

Essays on eco-innovation

Serdal Ozusaglam

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ESSAIS SUR I'ECO-INNOVATION (Essays on Eco-Innovation)

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Resumé

Le **Chapitre 1** propose un survol de la littérature consacrée à l'éco-innovation. A la lumière des travaux passés en revue, l'éco-innovation n'est pas encore un concept bien stabilisé : sa définition même continue d'évoluer, appelant des corrections et des précisions. Ce chapitre accorde une attention particulière au rôle de la règlementation, qui est souvent présentée dans la littérature comme un déterminant essentiel de l'éco-innovation, notamment quand l'innovation concerne un procédé de fabrication. Une controverse demeure à ce propos, en particulier quand la règlementation repose sur des normes environnementales peu contraignantes, voire facultatives (telle la norme ISO 14001). Le processus de diffusion de l'éco-innovation est un autre objet de controverse dans la littérature. Nous refermons ce chapitre en concluant qu'il convient de focaliser les recherches empiriques sur ces deux points de controverse, ce que nous essayons de faire dans les chapitres suivants.

Le Chapitre 2 va donc s'intéresser au premier point de controverse, à savoir le rôle des règlementations environnementales ayant une dimension facultative, comme les normes de type ISO 14001. A l'aide d'une analyse par appariement sur les scores de propension, nous examinons l'effet de l'adoption de normes de type ISO 140001 sur la performance des entreprises (mesurée à partir de la Valeur Ajoutée). La première étape de l'analyse consiste à prédire l'adoption de ces normes à l'aide d'un modèle Probit afin de calculer le score de propension. La seconde étape consiste à calculer, à l'aide d'un algorithme d'appariement non paramétrique, l'effet de l'adoption sur la VA des entreprises ayant adopté les normes (par rapport aux entreprises ayant un score de propension similaire mais n'ayant pas adopté ces normes). L'analyse est menée sur les données françaises de l'enquête COI 2006 combinées aux Enquêtes Annuelles d'Entreprises de 2003 à 2006. Ces données nous permettent de distinguer parmi les entreprises adoptant des normes de types ISO 14001, celles les ayant adopté précocement et celles les ayant adopté tardivement. Afin de tenir compte de ces deux comportements d'adoption, le score de propension est calculé à l'aide d'un Probit multivarié. Nos résultats indiquent qu'adopter des normes environnementales de type ISO 14001 sur une base volontaire permet d'accroître la VA des entreprises, que l'adoption soit précoce ou tardive. Une analyse de sensibilité conduite en estimant un modèle linéaire à interactions complètes (Fully Interacted Linear Model) vient conforter ces résultats.

Le Chapitre 3 s'intéresse à savoir le processus de diffusion de l'éco-innovation. En raison des contraintes posées par les données disponibles, cette investigation se fait du point de vue des barrières à, et des déterminants de, l'innovation environnementale. L'analyse repose sur l'estimation de fonctions de production d'innovation à l'aide de modèles à deux équations. La première est une équation de sélection qui permet de tenir compte du fait que toutes les entreprises recensées dans un échantillon ne sont pas nécessairement innovantes. La seconde équation permet d'expliquer l'intensité de l'éco-innovation, mesurée de différentes manières. Ces modèles sont estimés par Maximum de Vraisemblance sur les données européennes micro-anonymisées (ou "bruitées") de l'Enquête Communautaire sur l'Innovation de 2004 (CIS 2004). Les données fournissent deux mesures approchées de l'intensité de l'écoinnovation: (1) une mesure de la "réduction de l'utilisation des consommations intermédiaires et de la consommation d'énergie par unité produite" (sur une échelle de 0 à 4) et (2) une mesure de la "réduction de l'impact environnemental ou sanitaire et amélioration de la sécurité" (également sur une échelle de 0 à 4). Ces deux mesures peuvent être combinées pour obtenir une mesure continue de l'intensité d'éco-innovation. Les deux variables peuvent également être utilisées telle quelles, et séparément, ce qui conduit à estimer deux modèles économétriques dans laquelle l'équation de deuxième étape est un Probit ordonné. Les modèles estimés sur l'ensemble de l'échantillon suggèrent que la coopération avec des partenaires extérieurs et le degré d'ouverture de l'entreprise sont les principaux facteurs associés à une plus grande intensité de l'éco-innovation. Ce résultat fait sens dans la mesure où l'éco-innovation repose moins sur le développement de produits de haute technologie que sur l'adoption de procédés de production moins polluants – procédés qu'une entreprise peut acquérir en externe, auprès de partenaires plus expérimentés dans ce domaine. Les analyses conduites sur les quatre sous-échantillons évoqués plus haut conduisent toutefois à nuancer ce résultat, dans la mesure où la R&D apparaît bien comme un déterminant de l'éco-innovation dans les pays d'Europe de l'Ouest et du Nord d'une part, et dans les pays d'Europe méditerranéenne d'autre part.

Le **Chapitre 4** exploitant les données des Enquêtes Communautaires sur l'Innovation de 2008 (CIS 2008), qui contiennent un module spécifiquement dédié à l'éco-innovation. Ainsi, à la différence du chapitre 3, nous établissons – à l'aide du module spécifique de l'enquête – une typologie originale permettant de distinguer, parmi les entreprises éco-innovantes: (1) les entreprises adoptant une éco-innovation de manière passive, (2) les entreprises adoptant

tardivement une éco-innovation de manière stratégique, (3) les entreprises adoptant précocement une éco-innovation de manière stratégique, (4) les éco-innovateurs stratégiques tardifs et (5) les éco-innovateurs stratégiques précoces. Ces cinq types d'entreprises écoinnovantes s'opposent aux entreprises innovantes mais non éco-innovantes, et toutes se distinguent des entreprises non innovantes. Notre objectif est ici d'identifier des facteurs spécifiques à un type d'entreprise éco-innovante. Pour ce faire, nous estimons par Maximum de Vraisemblance de modèles comprenant une équation de sélection et une équation de type Probit ordonné. Comme dans le Chapitre 3, le degré d'ouverture des entreprises et la coopération avec des partenaires extérieurs apparaissent prédominants, et ce quel que soit le type d'éco-innovateur considéré. Toutefois, la R&D joue ici un rôle aussi important que ces deux facteurs. Il est donc a priori difficile d'identifier les spécificités de chaque type d'écoinnovateur, bien que les éco-innovateurs stratégiques précoces soient plus susceptibles de recourir à l'adoption de normes environnementales volontaires. Par contraste, les entreprises modérément éco-innovantes ont besoin d'encouragements financiers, qui peuvent prendre la forme d'un soutien de l'UE. L'estimation sur des sous-échantillons permet de nuancer ces résultats, en soulignant notamment que l'importance de la coopération avec des partenaires extérieurs varie d'un groupe de pays à l'autre.

Abstract

Environmental innovation or eco-innovation¹ is considered one of the most significant paradigm shifts in the innovation behaviour. The specificity of eco innovation lies in the fact that environmental improvement, in addition to technological advancement, is the main reason of its development. By combining these two objectives, eco-innovation has become one of primary tools in the search to solve the world's environmental problems and sustainability challenges. Despite its prevalence, there is still more to be discovered within the eco-innovation literature. The drivers of and barriers to eco-innovation discussion is one of the most significant amongst all. More notably, the role of a firm's social, technological, economic and organizational characteristics within the eco-innovation process has been little studied. With this Ph.D. thesis we aim to fill the existing gap with four distinct research articles.

In order to provide a solid background for the discussion, **Chapter 1** proposes a literature survey that collects and presents the most prominent theoretical and empirical contributions in the field of eco-innovation. In view of those studies, we show that the concept and definition of eco-innovation is still evolving and policy and regulation appear as specific drivers of eco-innovation especially, for environmental process innovation. Nevertheless, some issues remains controversial i.e., the impact of some of the less stringent environmental regulations (such as, ISO 14001 and EMAS) and its diffusion process. We concluded that if eco-innovation is to be encouraged among firms and consumers, more empirical research is needed on these queries.

Chapter 2 investigates, firstly, whether firms' internal characteristics have an impact on the adoption of voluntary environmental standards. Secondly, the causal effect of adoption of environmental standards on the firms performance. Based on the empirical evidence obtained, we show that medium-size, high-tech manufacturing firms operating at the EU level and using quality standards are more likely to adopt these standards earlier. By contrast, late adopters are more likely to be large, low/medium tech manufacturing firms that rely on

¹ According to Measuring Eco-innovation (MEI¹, 2007) project, eco-innovation is assimilation or exploitation of a product, production process, service or management or business method that it is novel to the firm or user and which results, throughout its life cycle, in a reduction of environmental risk, pollution and other negative impacts of resources use (including energy use) compared to relevant alternatives (p. 3)

other quality standards. We also observe that both of these adopters could reap financial benefits through Total Value Added (VA) but we do not detect any significant difference between the early and late adopters. We concluded this chapter by asserting that voluntary environmental standards may be a good complement for market-based instruments (such as; tradable permits, emission taxes, subsidies etc.) and, command and control mechanisms to increase firms' economic performance while becoming greener.

In **Chapter 3**, we stressed the significance of a firm's structural characteristics in another context and aimed at investigating the role these characteristics play in realization of product and/or process eco-innovation. We contributed to the discussion of drivers of and barriers to eco-innovation by investigating the role of a firm's internal characteristics, its openness and cooperation with external partners within the eco-innovation process. Based on the empirical evidence collected, we show that a firm's cooperative behaviour may be more beneficial than some of their structural characteristics and this behaviour may even substitute its certain internal efforts i.e., R&D intensity. With regard to country specific differences, the results point out that established R&D tradition in Western and Mediterranean countries (when contrasted to Central European and Baltic countries) is indeed one of the most significant components of a firm's product and/or process eco-innovation capabilities. While fierce competition in European markets and existing technological differences among country groups may force Eastern European and Baltic firms to speed up and shift their approach (i.e., through networking and formal cooperation) to eco-innovate.

In **Chapter 4** we moved forward the analyses conducted in Chapter 3 to another context and examined the influence of firm's economic, technological, organisational capabilities on the eco-innovation behaviour for different types of product, process, organisational or marketing eco-innovators. The results of our empirical analyses point out the marginal impacts of firm's characteristics changes with respect to type of eco-innovator. More notably, a firm's voluntary engagement in eco-innovation is as important as the environmental regulations for early strategic eco-innovators. While financial funds provided by the EU seem to spur firms' propensity to become a late strategic eco-innovator and strategic eco-adopters. Country-specific analyses indicate that within the Western European and Baltic countries firms' structural characteristics, voluntary engagement in eco-innovation and environmental regulation are the main factors that leads firms to be early strategic eco-innovators. The

marginal impact of the same factors is of higher importance for the late strategic ecoinnovators within the Eastern European and Mediterranean countries. We concluded this chapter by claiming that besides a firm's internal characteristics and innovation capabilities, the conceptual and regulatory framework being involved and, the interrelations with other external factors may also shape their innovation strategy.

LIST OF FIGURES

Chapter 1
2.2.2. Possible distribution of firms according to eco-activities
Chapter 2
3.3.1. Systematic visualisation of the EMS adopters

LIST OF TABLES

Chapter 2

3.1. Summary of key variables before and after merging the datasets	77
3.1.1. Frequency of adopters and non-adopters according observation periods	80
3.1.2. Summary statistics on the variables used in the econometric analysis	82
3.1.2. Aggregated Nace codes according to OECD's technology classification	83
4.1. Estimation Results	79
4.2. Average Treatment Effect on the Treated (ATT) obtained from PSM and FILM	90
4.2. Summary statistics on the choice of variables before and after matching	92

Chapter 3

3.3. Distribution of Eco-innovators and innovators according to industry classifications I	114
3.4. Summary statistics of the variables used in the econometric analysis	117
3.4. Descriptive statistics of the regressors for eco-innovators and conventional innovators l	118
4.1. Estimates of the Generalized Tobit Model for eco-innovation	123
4.2. Estimates of the Generalized Tobit model by group of countries (Western Europe and	
Mediterranean)1	125
4.2. Estimates of the Generalized Tobit model by group of countries (Eastern Europe and	
Baltic countries)	127

Chapter 4

3.3. Taxonomy of Firms (Terms used in this study)	.170
3.4. Summary statistics on the variables used in the empirical estimation	.177
3.4. Summary statistics on the variables used in the estimations by type of eco-innovator	.178
3.5. Descriptive statistics according to industry affiliations	. 180
3.5. Distribution of eco-innovators across industries for each type of eco-innovator	. 182
3.5. Summary statistics for different types of eco-innovators across all manufacturing	
industries and service sectors	. 184
4.1. Marginal effects for the whole sample	. 187
4.2. Distribution of innovators across countries	. 191
4.2.1. Marginal effects for the Western European group	. 193
4.2.1. Marginal effects for the Baltic group	. 195
4.2.1. Marginal effects for the Mediterranean group	197
4.2.1. Marginal effects for the Eastern European group	. 199

General introduction

The tendency to innovate plays a vital role in human development; it has enabled us to improve our health and living conditions, expand and share our knowledge and increase our productivity and income. Innovation has undoubtedly improved our lives in many ways; it has also, however, led to serious problems, challenges which we are now struggling to address. As the world population continues to grow, and with it, demands for improved standards of living, increased consumption is leading to worsening pollution, more severe climatic change, and the irreversible depletion of natural resources and biodiversity (Cawsey, 1996). It is no surprise to hear then, at both international and national levels, about the severity of the impacts of global climate change and possible solutions for reversing these impacts. It is clear that in order to address this problem we will have to invest extensive work, resources and time, but as we debate and discuss, we are running out of time and applicable solutions while environmental degradation continues to grow. In order to escape this inertia, we need to make fundamental changes in policy (such as efficient use, definition and governance of regulatory frameworks, standards and norms), institutions (from education to research) and practices (from the household level to industrial organization and to society as a whole). These goals are indeed well beyond the reach of single handed efforts of individual firms or nations and require conjoint efforts of individuals, organizations, businesses and governments. Ekins (2010) states that the scale of the changes that seem to be envisaged goes well beyond individual technologies and artefacts, and involves system innovation through 'a technological transition' where this transition aims at both reducing environmental impacts and the use of natural resources. The literature calls this type of paradigm shift environmental innovation (a.k.a. Eco-innovation), which is rather a new concept (see Fussler and James, 1996). Despite its recent appearance it succeeded to attract the attentions of many. Its success can be related with increasing global environmental problems (air, water soil and sound pollution along with decreasing stock of renewable resources), consumer awareness and demand, pressures from regulatory authorities and, the economic gains that it can provide. On a micro scale, firms consider eco-innovation as a crucial technological opportunity to reduce their carbon foot-prints and to obtain substantial financial returns through decreased input use and increasing demand for their product, process, organisational and marketing eco-innovations.

Eco-industries (e.g., air pollution control, waste water management, solid waste management, soil remediation, and renewable energies and recycling) are among the fastest

growing industries in the world and they are likely to be worth around 2\$ trillion by 2020 (European Commission², 2012). On a macro scale, countries or group of countries such as the EU approach this specific paradigm shift as pathway to a more sustainable, competitive and knowledge based economy i.e. the EU's Lisbon agenda (2000). Therefore, it may be the only way to sustain the current quality of life (or even increase it) in the long run. The most significant difference between conventional innovation and eco-innovation is the environmental concern embedded in the latter. By joining two objectives that for much of history appeared to be mutually exclusive, eco-innovation has become one of the most pivotal developments in the search to solve the world's environmental problems and sustainability challenges.

Previous studies and motivation of this research

Undoubtedly, introduction and development of eco-innovation is important for a better future but also for a more sustainable economy. If successfully realized, it could provide substantial financial return through i.e., decreased energy and/or material use, increased labour and capital productivity in the production process and through commercialization of the product, production, organizational and marketing innovations. Despite its increasing importance there are still many issues that deserve a thorough investigation. Difficulties inherent in measuring the environmental aspects of an innovation has led to certain problems in collecting accurate and reliable data, which, resulted in lack of empirical research and knowledge. These complications have led the researchers to deal with certain aspects of eco-innovation and thus the empirical eco-innovation literature calls for a thorough empirical research. For example, most researchers identified the drivers and barriers of eco-innovation mainly from the technology-push, market-pull and regulatory framework context. Even though these conventional factors should be considered as of the utmost importance for a successful eco-innovation, as an evolutionary mechanism, the traditional process of innovation requires steady change while the process becomes highly complex, interdependent and more difficult to coordinate. Evolutionary economists (i.e., Nelson and Winter, 1982; 2009) draw our attention to the significance of social interactions by stating that innovation is a path-dependent and a complex process where technology is developed through various applicable knowledge accrued by interacting with various actors and other

² The EU's, Number of Jobs Dependent on the Environment and Resource Efficiency Improvements project, Final Report, 2012.

factors. Combined with the double externality problem³ (Rennings, 2000), the process of ecoinnovation tends to be even more complex and is affected by even more factors. Kemp and Pearson (2007) assert that firms' propensity to eco-innovate is positively related to their ability to build on organisational capabilities. De Marchi (2012) emphasizes the role of engagement in continuous R&D activities and cooperation activities as a driver of ecoinnovation. Mazzanti and Zoboli (2005) show that networking activities may be a major driver for environmental innovation and they may be even more important than the structural characteristics of firms (such as size). Hemmelskamp (1997), in a study of German manufacturing firms, claims that eco-innovators tend to have lower R&D intensity and the use of external sources of information can be a good solution to countervail potential information gaps. In addition, Kesidou and Demirel (2012) and Horbach (2008) claim that the adoption of Environmental Management Systems (EMS), as an indicator of a firm's organisational capabilities, may indeed increase the firm's eco-innovation intensity. Porter and van der Linde (1995) raised considerable attention when they postulated that stricter environmental regulations would lead firms to eco-innovate and the economic gains that could be reaped from this investment would offset (partially or fully) the cost of the regulation itself, leading to a situation known as a win-win opportunity. The drivers of and barriers to eco-innovation discussion is one of the most significant subject for a thorough empirical investigation. More notably, the role of a firm's social, technological, economic and organizational characteristics within the eco-innovation process has been little studied. This Ph.D. thesis aims at contributing to both theoretical and empirical literature of ecoinnovation by questioning the role of a firm's economic, technological, organisational and social capabilities in eco-innovation process.

Aims and scopes of the research

This Ph.D. thesis aims at finding empirical evidence regarding the influence of firms' structural, economic, technical, managerial and social capabilities on various aspects of the eco-innovation process. This research question has been addressed using a firm-level perspective, embracing both the economic and the management literatures. This thesis thus

 $^{^{3}}$ Eco-innovations produce positive spill overs in both, the innovation and diffusion phase. Positive spill overs of R&D activities can be usually identified for all kinds of innovations. The peculiarity of eco-innovations is that positive spill overs appear also in the diffusion phase due to a smaller amount of external costs compared to competing goods and service on the market. This peculiarity of eco-innovations has been called the double externality problem (Rennings, 2000).

contributes to the discussion on the drivers of and barriers to eco-innovation discussion. It also provides further empirical evidence on the validity of the so-called "Porter hypothesis".

These questions are addressed from an empirical economic perspective, relying primarily on concepts drawn from the economics of innovation (such as the "innovation production function" and "innovativity") and on econometric analyses. While the questions at hand may concern environmental economics as well as innovation economics, the emphasis in this Ph.D. thesis is deliberately on the latter. One of the reasons for this (apart from a sheer interest in the economics of innovation) is that a recurrent problem in eco-innovation studies is the shortage of available quantitative data to conduct empirical research. This thesis tackles this challenge by utilizing to the utmost some of the most extensive databases available, which all pertain to some form of innovation (organizational innovation, ICT-related innovation, product and/or process innovation). Each chapter of the thesis focuses on a precise research question that fully exploits one of three datasets made available for this research. These three datasets are: (1) the French "Organisational Changes and ICT use" Survey of 2006 (Changement Organisationnel et Informatisation 2006, COI 2006) matched with the French Annual Firm Survey (Enquête Annuelle d'Entreprises, EAE) and (2) microanonymised data from the 2004 and 2008 waves of the Community Innovation Survey (CIS 2004 and 2008) available for several European Countries. The firm-level data provided by the COI 2006 survey matched with the relevant EAE was used in Chapter 2 to test a "weak or narrow" version of the Porter hypothesis. The third and the fourth chapters of the thesis exploit the micro-anonymised data from CIS 2004 and CIS 2008, respectively, to examine the barriers to and drivers of eco-innovation (in Chapter 3) and to identify the characteristics of different types of eco-innovators (in Chapter 4). While Chapter 2 was focused on France, using micro-anonymised CIS data available for several EU countries (and affiliates) allowed us to address the research questions in Chapters 3 and 4 from a truly international perspective. Using micro-anonymised CIS data, we were able to highlight country-specific differences with regard to firms' eco-innovation intensity and eco-innovative behaviour. Despite their usefulness, the aforementioned data still suffer from certain limitations, which constrained our empirical analyses. In particular, the French COI 2006 data does not allow researchers to distinguish between ISO 14001 environmental standards and other related nonmandatory standards, all of them being encompassed by a single variable. In another vein, while the micro-anonymised CIS data covers several countries, it often leaves out some of the core EU member States presenting a wealth of information on more recent entrants.

Given the reliance of this Ph.D. thesis on econometric analyses, the choice of appropriate econometric methodology and tools was of the utmost importance. Except in Chapter 1 (dedicated to a survey of the literature), we have tried, in each chapter, to address the research question with the most appropriate tools and state-of-the-art techniques. In Chapter 2, we thus use both Propensity Score Matching (PSM) Method and Fully Interacted Linear Models (FILM) to measure the impact of the adoption of ISO 14001-type standards on firm performance. Both methods aim at correcting selection biases with respect to observable characteristics, the latter allowing us, in addition, to test for the presence of heterogeneous effects (Goodman and Sianesi, 2005). In the third and fourth chapters, we adapted and estimated an innovation production function following the *innovativity* approach proposed by Mohnen and Mairesse (2002). This approach consists in estimating an innovation production function specified as a Generalized Tobit model (i.e., a variant of the Heckman (1979)'s selection model estimated by Maximum Likelihood), in order to address the selection bias inherent in the use of CIS data. One of our conceptual adaptation of this framework consists for instance in replacing Mohnen and Mairesse (2002)'s measure of innovation intensity with a measure of *eco-innovation* intensity. Other technical adaptations consist in replacing the linear intensity equation of the Generalized Tobit model with an Ordered Probit equation when our measure of eco-innovation is not a continuous, but an ordered categorical variable.

Another contribution of this thesis, in the second and the fourth chapters, is that we were able to distinguish different types of eco-innovators, further refining our analysis. In Chapter 2, we have thus separated the adopters of voluntary ISO 14001-type environmental standards into two categories: *early* adopters and *late* adopters. Since the data covers the period from 2003 to 2006, a firm that adopted this type of standards in 2003 is considered as an *early adopter* and firms that adopted the same standards in 2006 considered as a *late adopter*. Being an early adopter indicates that a firm is relatively more experienced in dealing with environmental problems and environmental management practices. In Chapter 4, we propose a new taxonomy of eco-innovators and eco-adopters with regard to a firm's strategic behaviour towards eco-innovation. Adapting and refining Kemp and Pearson (2007)'s categorization, we could classify eco-innovators into five mutually exclusive categories: *(i)* "passive" *eco-adopters*, *(ii)* "late" *strategic eco-adopters*, *(iii)* "early" *strategic eco-adopters*. To the best of our knowledge, this type of distinction between eco-innovators has not been attempted by any other empirical research based on the econometric analysis of firm-level data.

Structure of The Ph.D. Thesis

Based on our research question, this Ph.D. thesis comprises four mutually-exclusive research articles on eco-innovation⁴.

Chapter 1, entitled "Environmental Innovation: A Concise Review of the Literature⁵", aims at providing a solid background for the discussion, and proposes a literature survey that collects and presents the most prominent theoretical and empirical contributions in the field of eco-innovation. This paradigm is rather new and there is a need for an extensive review of the existing researches in the field of eco-innovation. With the first chapter of this Ph.D. thesis, we provide the reader with a detailed presentation of the concept of eco-innovation from a historical perspective. Its definition and its peculiarities are the issues addressed in the main section of this chapter. The other sections include an extensive body on the issue of drivers and barriers of eco-innovation and its diffusion process. Based on both theoretical and empirical literature surveys, we show that the concept and definition of eco-innovation is still evolving and, besides the conventional supply-side and demand-side determinants of innovation, policy and regulation appear as specific drivers of eco-innovation. Empirical evidence also indicates that environmental policy and an appropriate regulatory framework both have a strong impact on eco-innovation, and especially on environmental process innovation. Nevertheless, the impact of some of the less stringent environmental regulations (e.g., voluntary proactive approaches such as; ISO 14001 and EMAS⁶) on the introduction of eco-innovation remains controversial. We suggest that an in-depth empirical analysis is required to stress the importance of voluntary 'ISO-14001 type' environmental standards and norms for the eco-innovation process. Finally, we can also claim that more (empirical) research on the diffusion processes of eco-innovation is needed, so that it can be efficiently encouraged by policy makers among both firms and consumers.

Building upon the conclusions of the Chapter 1, **Chapter 2**, entitled "Voluntary environmental standards: A new strategy to promote greener business?"⁷, investigates, first, whether firms' internal characteristics have an impact on the adoption of 'ISO 14001-

⁴Eco-innovation is assimilation or exploitation of a product, production process, service or management or business method that is novel to the firm or user and which results, throughout its life cycle, in a reduction of environmental risk, pollution and other negative impacts of resource use (including energy use) compared to relevant alternatives" (MEI, Kemp and Pearson (2007, p. 3).

⁵ This chapter was published as an article in *VSE* (*Vie et Science de l'Entreprise*) in January 2013.

⁶ Environmental Management and Audit Schemes

⁷This chapter corresponds to a research paper co-authored with Stéphane Robin (Université Paris 1 Panthéon-Sorbonne) and about to be submitted for publication.

type'voluntary environmental standards and second, the effect of the adoption of these standards on firms performance. The OECD (2009) refers to organisational innovation as "incremental or radical change of a firm's processes and responsibilities, which reduces environmental impact and supports organisational learning". We can therefore consider adopters of EMAS and ISO 14001-type standards as organisational-eco-innovators. This type of environmental standards is often considered as one of the most efficient voluntary proactive approaches and a useful supplement to traditional mandatory command and control regulations (e.g., Khanna and Damon, 1999; Alberini and Segerson, 2002). They do not impose environmental performance requirements and help organisations to form their own EMS⁸. Many organisations consider these types of standards as a good and flexible approach to clean their operations and a good opportunity to identify themselves as eco-innovators. But less is known about the factors that lead firms to adopt ISO 14001-type standards and about the potential returns to the adoption of such standards (in terms for instance of increased sales or profits). The average cost of implementation of such environmental standards can reach to hundreds of thousand dollars. If the financial return is not guaranteed, a firm's motivation to adopt these standards deserves a thorough empirical investigation. The proposed research is a direct attempt to test the validity of the "narrow or weak" version (see Jaffe and Palmer, 1997) of the Porter hypothesis. The empirical analyses rely on the Propensity Score Matching (PSM) method and Fully Interacted Linear Model (FILM). These methods are applied using firm-level data acquired from a large French survey (COI 2006) matched with administrative EAE (2006) data. As a result of this merger, we worked with a panel of 11,168 observations (5,584 firms observed in 2003 and in 2006) with more than 11 employees. Within the context of this chapter, a firm that adopted an ISO 14001-type standard in 2003 is considered as an early adopter and a firm that adopted the same type of standard in 2006 is considered as a late adopter. Being an early adopter basically indicates a firm's relative experience in dealing with environmental problems and environmental management practices. Our empirical results suggest that a firm's structural characteristics organisational capabilities, business market and technological complexity are some of the factors that have a significant impact on a firm's propensity to adopt voluntary EMS. We also showed that respective marginal effect of these factors may differ between the early and late adopters of the EMS. Last but not least, we also observed that both types of adopters reap financial benefits (as measured by an increase in Value Added) from the adoption of environmental standards, but we do not detect any

⁸ Environmental Management Standards (EMS)

significant difference in this respect between early and late adopters. These results support the Porter hypothesis to a certain extent. We concluded this chapter by asserting that voluntary environmental standards may be a good complement for market-based instruments (such as; tradable permits, emission taxes, subsidies etc.) and command and control mechanisms to make firms greener while increasing their economic performance.

In Chapter 3, entitled "Which firm characteristics matter? Identifying the drivers for eco-innovation", we stress the significance of a firm's structural characteristics in another context and aim at investigating the role these characteristics play in the realization of product and/or process eco-innovation. We suggest that due to environmental concerns and the double externality problem⁹ (Rennings, 2000), the process of eco-innovation tends to be even more complex than the conventional innovation process and it is likely to be affected by even more factors. These complexities, therefore, imply that it is essential to identifying those firms' technical, organizational and/or social capabilities which are liable to affect their engagement in eco-innovation. We contribute to this discussion by investigating how a firm's openness and of formal cooperation with external partners affect their eco-innovation intensity. We test the significance of these factors by conducting empirical analyses which focus on the estimation of innovation production functions following the innovativity framework proposed by Mohnen and Mairesse (2002). Formally, this approach relies on the estimation (by Maximum Likelihood) of a Generalized Tobit model in which the first equation addresses selection into (general) innovation and the second equation addresses the intensity of eco-innovation. We performed our estimations on micro-anonymised data from the 4th wave of the Community Innovation Survey (CIS 4). This database provides us with comparable firm-level data on innovation activities in European countries. The empirical analysis is first conducted on the whole sample covering 104,943 observations across 15^{10} different countries, and then by country group¹¹. The empirical results show that ecoinnovative firms tend to be large, autonomous firms that are open to the use of internal and

⁹ Eco-innovations produce positive spill overs in both, the innovation and diffusion phase. Positive spill overs of R&D activities can be usually identified for all kinds of innovations. The peculiarity of eco-innovations is that positive spill overs appear also in the diffusion phase due to a smaller amount of external costs compared to competing goods and service on the market. This peculiarity of eco-innovations has been called the double externality problem (Rennings, 2000).

¹⁰ Belgium, Bulgaria, Czech Republic, Germany, Estonia, Spain, Greece, Hungary, Lithuania, Latvia, Norway, Portugal, Romania, Slovenia and Slovakia.

¹¹ We created four country groups: Western European, Eastern European, Mediterranean and Baltic countries

external sources of information ('openness'). We found evidence that eco-innovative firms tend to countervail basic R&D intensity and engagement in continuous R&D by cooperating with public research institutes, universities and other higher education institutions, as well as other non-scientific partners, which supporting Hemmelskamp's (1997) claims. We then have reasons to believe that a firm's cooperative behaviour may be more beneficial to the development of an eco-innovation than some of their structural characteristics, and that this behaviour may even, to a certain extent, act as a substitute for internal R&D efforts. With regard to country specific differences, we concluded that countries' economic, social and technological levels indeed have a significant impact on a firm's eco-innovation strategy and the way eco-innovation is developed in each country groups. The results point out that established R&D tradition in Western and Mediterranean countries (when contrasted to Central European and Baltic countries) is indeed one of the most significant components of a firm's product and/or process eco-innovation capabilities. Fierce competition in European markets and existing technological differences among country groups may nevertheless force Eastern European and Baltic firms to speed up and shift their approach (e.g., through networking and formal cooperation) to eco-innovation.

In Chapter 4, entitled "A Study on the Types of Eco-innovators Using Micro-anonymised Data: Case of eco-innovators and eco-adopters", we moved forward the analyses conducted in Chapter 3 to another context and examined the influence of firm's economic, technological, and organisational capabilities on the eco-innovation behaviour for different types of product, process, organisational or marketing eco-innovators. Until now there has been little or no attempt, in empirical studies, to distinguish true eco-innovators from mere eco-adopters, which is partly due to a lack of empirical data. In this chapter, we used microanonymised data from the 2008 wave of the Community Innovation Survey (CIS 2008), a wave which includes a special module on eco-innovation and provides us with detailed information about the environmental benefits of product, process, organisational or marketing innovations introduced during the observation period (from 2006 to 2008). In this final chapter, we propose a new typology of eco-innovators by adapting and refining Kemp and Pearson (2007)'s categorization. Using (i) the environmental impacts of an innovation and (ii) the way the eco-innovations are implemented, we were able to classify eco-innovators into five mutually exclusive categories First, the measured environmental impacts of realised product, process, organisational and marketing innovations are used to qualify a firm as an eco-innovator. Second, the information about the way eco-innovations are implemented is

used to separate "true" strategic eco-innovators from strategic eco-adopters. Combining these two dimensions of eco-innovation yield the five aforementioned categories of eco-innovators: passive eco-adopters, early strategic eco-adopters, late strategic eco-adopters, early strategic eco-innovators and late strategic eco-innovators. Our empirical analyses aim at identifying specific differences between these different types of eco-innovators, taking into account the fact that they have to be distinguished from non-innovators and non-eco-innovators. Again, we adapted the innovativity framework first proposed by Mohnen and Mairesse (2002) to our purposes. Our extension consists in estimating a model accounting first for selection into innovation, and second for the probability to be one of the five aforementioned types of ecoinnovators rather than a "conventional" innovator that does not eco-innovate. To do so, we estimated an Ordered Probit model with selection. This model has a Probit selection equation (to model selection into innovation), and an Ordered Probit equation (replacing the innovation intensity equation of Mohnen and Mairesse, 2002), the dependent variable of which has six categories: the five aforementioned categories of eco-innovators, and the base category (conventional innovators that do not eco-innovate). Both equations are estimated simultaneously by Full Information Maximum Likelihood. The model is first estimated on the whole sample, which comprises 79,972 observations across 11¹² EU countries. As in Chapter 3, we then create four country groups and replicate our estimations within these country groups. The results of our empirical analyses allow us to shed light on the factors and characteristics that lead a firm towards a more active, strategic eco-innovation behaviour. In particular, we find that voluntary engagement in eco-innovation is as important as environmental regulations for early strategic eco-innovators. By contrast, financial funds provided by the EU seem to spur firms' propensity to be a "late strategic eco-innovator" or a "strategic eco-adopter". This latest result suggests that environmental regulation might be an efficient tool for attracting new firms to eco-innovation, or at least for making their innovative behaviour more sustainable. If efficiently used, these tools may have a significant impact on firm's eco-innovation efforts and increase the number of eco-innovators. Countryspecific analyses show that a firm's cooperative behaviour, internal R&D efforts, international market conditions and financial funds provided by the EU are the main factors that lead a firm towards a more active eco-innovation behaviour. Forming formal cooperation with scientific and other partners and being open to external sources of information appears to be especially important for eco-innovators. Eco-adopters, as well as eco-innovators, tend to

¹² Bulgaria, Cyprus, Czech Republic, Germany, Estonia, Hungary, Ireland, Lithuania, Portugal, Romania and Slovakia

look for external financial support to mitigate financial risks involved in the eco-innovation process. We conclude this chapter by claiming that besides a firm's internal characteristics and innovation capabilities, the regulatory framework the interrelations with other external actors also shape their innovation strategy.

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Chapter 1

Environmental Innovation: A Concise Review of the Literature

Table of Contents

1. Introduction
2. Eco-innovation: A Specific form of Innovation?
2.1. From Innovation to Eco-innovation
2.2. What is Eco-innovation?
2.2.1. Toward a Definition of Eco-innovation
2.2.2. Classification of Eco-innovators and Eco-innovation
3. Drivers of and Barriers to Eco-innovation
3.1. Drivers
3.2. Barriers
4. Environmental Policy
4.1. Recognition of Environmental Policy as a Driver of Eco-innovation
4.2. Effectiveness of Environmental Policy as a Driver of Eco-innovation
4.3. Proactive Approaches to Environmental Protection
5. Diffusion of Eco-innovation
5.1. From Normal Innovation to Eco-innovation
5.2. Hindrances to Diffusion of Eco-innovation
6. Conclusion

Environmental Innovation: A Concise Review of the Literature

Abstract

The importance of environmental innovation or "eco-innovation" has become well recognized in the innovation literature. The present survey proposes a review of the most prominent theoretical and empirical contributions in the field, without aiming at being exhaustive. The review of the theoretical works suggests that the definition of eco-innovation is still evolving, although recent definitions of eco-innovation all tend to be based on environmental performance rather than on environmental aim. The review of the empirical research indicates that the regulatory framework and environmental policy both have a strong impact on eco-innovation, besides conventional technology-pushed and demand-pulled factors. This is mostly a result of eco-innovation sharing some of the characteristics of a public good, thus giving rise to a double-externality problem. We also consider some issues that remain to be explored, such as the diffusion of eco-innovation. If it is as beneficial as expected (i.e., likely to overcome environmental problems without hampering economic performance), a better grasp of its adoption mechanisms is still required, so that it can be efficiently encouraged by policy makers among both firms and consumers.

Keywords: Eco-innovation, Barriers and Drivers, Environmental policy, Diffusion

1. Introduction

Innovation has been long been considered as a subject of the utmost importance in the managerial and economic literature. At the level of an economy, innovation is one of the most important factors leading to development, growth and competitiveness. At the firm level, innovation, because its focus is on change and on the creation and/or commercialization of novelty, requires specific, flexible forms of management. These considerations have led to an abundant literature on innovation. Environmental innovation (a.k.a. "eco-innovation"), a specific form of innovation aiming at reducing the impact of products and production processes on the natural environment, has only recently appeared in the innovation literature. It has since then attracted the attention of scholars, who have attempted to define the concept of eco-innovation and identify its drivers and barriers at various levels of analysis (from consumer and firm levels to industry and national levels).

Among these contributions, the seminal work of Porter and Van der Linde (1995) has led to the much-debated "Porter hypothesis", according to which environmentally benign innovations can lead to an increase in firms performance, for instance through a reduction of energy or materials use. Since firms are not always aware of the opportunities offered by ecoinnovation, a strict and effective environmental regulation is required, so that they can be brought, through compliance, to this awareness. In this conception, environmental policy appears as an important driver of eco-innovation and deserves a specific attention.

Although it kick-started an increased interest in eco-innovation, the Porter hypothesis remains controversial, due to contrasting empirical evidence. More generally, despite its increasing importance, research on eco-innovation is still in an early stage as far as the gathering of quantitative and qualitative evidence is concerned. Furthermore, provided that environmental innovation is as beneficial as expected, issues such as its diffusion remain largely unexplored. The present survey proposes a summary of the most relevant and prominent researches in the field of eco-innovation. This literature review does not pretend to be exhaustive, but aims at providing relevant insights, which may give readers a better grasps of various aspects of eco-innovation.

The paper is organized as follows. In the first and second sections, we examine the specifics of environmental innovation (or "eco-innovation") compared to "normal" innovation and touch upon the definition of eco-innovation issue. The third section is dedicated to the identification of the drivers and barriers to eco-innovation, while the fourth section focuses

on the specific role played by environmental policy and regulation. The fifth section considers the issue of the diffusion of eco-innovations. We conclude in a final section.

2. Eco-innovation: A Specific Form of Innovation?

2.1. From Innovation to Eco-Innovation

Technological innovation has long been strictly defined as "the introduction of new products, processes or service into the market" (UNCTAD¹³, 2006). With the development of innovation studies as an interdisciplinary field (involving business research, economics, epistemology, management and sociology, among others), this definition soon appeared as too narrow and somewhat restrictive. A somewhat broader definition has been given by the "OSLO Manual" (OECD, 1997:2009), according to which innovation is "the implementation of a new or significantly improved product (good or service), or process, a new marketing method, or a new organizational method in business practice". This definition stresses that an innovation needs not be novel to the market to be qualified as such; it is sufficient that it be novel to the firm which implements it. Hence, any new process or business practice which has been developed somewhere else and then adapted to the said firm can also be counted as an innovation. One may then wonder whether this broader definition is not actually too broad, being so all-encompassing that it might lose the specificity of what makes an innovation.

This broader definition can find a justification, though, in the fact that technological change is generally consider to take place in three stages: invention, innovation and diffusion. This classification, originally proposed by Schumpeter, is still widely used even though it is often regarded as over-simplistic (Foxon *et al.* 2007). In spite of its supposed crudeness, this classification is of a practical interest here, as it allows one to point out that, while innovation should be distinguished from invention, diffusion can be regarded as an extension of innovation.

The distinction between invention and innovation is well-known: The concept of invention refers to discovery, whereas most innovations are not based on discovery. They rather are the outcome of applied research and development informed by theoretical knowledge, engineering experience and knowledge about user needs. By contrast, diffusion can be considered as another, important part of the innovation process. Kemp and Pearson (2007), among others, have underlined that innovation continues in the diffusion stage of the innovation process. New uses and users may be found during diffusion. Advances in

¹³ United Nations Conference on Trade and Development

technology, increasing R&D efforts and feedback from users and suppliers help sellers improve their products, while competition may bring the price of an innovative product down.

When it comes to innovation, the choice of a definition is important, if only because the assessed impact of an innovation varies considerably depending on the concept of the innovation used. This makes the effects of innovation sometimes hard to measure (Kline and Rosenberg, 1986). The availability, since the early 1990s, of innovation surveys constructed according to the guidelines provided by the Oslo manual has allowed for significant advances in that domain (for an overview, see Kleinknecht and Mohnen (2002) and OECD (2009), for instance). However, these econometric studies mostly focus on the impact of innovation on productivity. Even then, distinguish the respective impacts of different forms of innovation (e.g. product innovation, process innovation and organisational innovations) remains difficult due to imprecision and measurement errors (Mairesse and Robin, 2012).

Innovation can be classified into different forms alongside different dimensions, one of them being the distinction between incremental and radical innovation. Another dimension, which is a core concern of the present survey, is the degree of "environmental friendliness" of an innovation. This dimension deserves a specific examination, since it has given rise to the notion of environmental innovation or "eco-innovation", which is often presented as a specific form of innovation supposedly presenting a large number of positive impacts at different levels: the economy, the society, and the planet. In the next section we attempt to clarify and define the notion of eco-innovation.

2.2. What is eco-innovation?

2.2.1. Toward a Definition of Eco-innovation

The claim that the economy is embedded in the human society, which is itself embedded in the natural, physical environment of planet Earth is not recent, and has already been made by environmentally-conscious economists such as René Passet (see Passet, 1979 in particular). However, the concept of "eco-innovation" per se is rather recent, since it first appeared in the innovation literature in a book by Fussler and James (1996). These authors defined eco-innovations as "new products and processes which provide customer and business value, but significantly decrease environmental impacts". However, since this term has first been coined, the concept of eco-innovation has attracted the attention of many scholars, and – as was already the case for innovation – various definitions of eco-innovation have been proposed in the literature. Some of them are strictly based on the environmental aim or environmental performance of innovations. Early studies of eco-innovation have focused on environmentally motivated innovation, overlooking the potential environmental gains that may be derived from "normal" innovations¹⁴. However, most current definitions of eco-innovation are based on environmental performance rather than on environmental aim, since it is not the aim that is of interest, but whether there are positive environmental effects related to the innovation (Kemp and Pearson, 2007). Today, the most widely accepted definition of eco-innovation may be that proposed by Kemp and Pearson (2007) as part of the MEI¹⁵ project:

"assimilation or exploitation of a product, production process, service or management or business method that it is novel to the firm or user and which results, throughout its life cycle, in a reduction of environmental risk, pollution and other negative impacts of resources use (including energy use) compared to relevant alternatives (p. 3)".

The above definition includes not only innovation aimed at reducing environmental impacts, but also cases where innovation leads to reduced impacts without this being an explicit aim. In that sense, normal innovations which have positive environmental effect are also counted as eco-innovations. As mentioned by OECD (2009), eco-innovation may be environmentally motivated, but may also occur as a side-effect of other goals, such as complying with regulations and norms, increasing productivity, reducing input costs (and hence production costs). Thus, according to this definition, a normal innovation should also be considered as a potential eco-innovation. Moreover, innovations thus defined are not Schumpeterian in the sense that they need not create a new market (or be novel to the existing market) to be considered as innovations: being novel to the firm is enough.

The MEI definition embraces the idea of adopting environmental performance rather than aim as the fundamental defining criterion. According to this definition, eco-innovations need not have an environmental aim in either the development phase or the use of the product/process. They just have to be better than their alternatives on a life cycle basis (Speirs *et al.* 2008). According to the contributors to the MEI project, the concept of eco-innovation should not be limited to new or better environmental technologies, which results in every

¹⁴ "Normal" innovations are those developed for usual market-oriented reasons such as saving costs or providing better service to users.

¹⁵ "Measuring Eco-Innovation" Project co-funded by the European Commission within the Sixth Framework Programme (2002-2006).

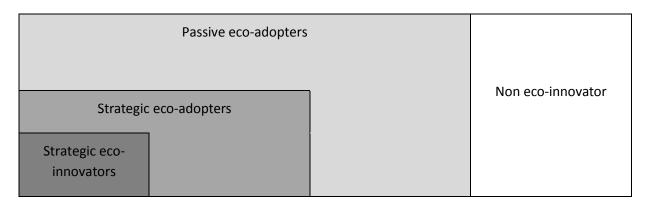
environmentally-improved product or service counting as an eco-innovation. The justification for this is that only taking into account the aim of an innovation would narrow too much the boundary of the term.

2.2.2. Classifications of Eco-innovators and Eco-innovation

Taking a critical look at the definition given in Section 2.1, one may fear that such an allencompassing definition may be too imprecise and too optimistic as regards the effects of innovation on the environment. This is not necessarily the case. Indeed, the contributors to the MEI project acknowledge that eco-innovation is likely to occur in different manners in the whole economy (Kemp and Pearson, 2007). Given this, it is necessary to distinguish between innovations that will have a major positive impact on the environment and those that will have only a limited impact.

Thus, although the MEI project states that "any company adopting a good, service, production process management or business method with environmental benefit is an ecoinnovator (Kemp and Foxon, 2007b, p. 17)", it also acknowledge the necessity to distinguish between four types of eco-innovators (i.e., firms with different behaviours regarding ecoinnovation): (1) **Strategic eco-innovators** are active in eco-equipment and service sectors, and/or develop eco-innovations for sale to other firms; (2) **Strategic eco-adopters** intentionally implement eco-innovators have no specific strategy to eco-innovate, although they may accidentally implement innovations that result in environmental benefits; (4) **Non** eco-innovators do not develop either intentional or unintended innovations with environmental benefits. This typology of eco-innovators, suggested by (Kemp and Pearson, 2007), is summarized in Figure 1.

Figure 1: Possible distribution of firms according to eco-activities



Source: MEI, Kemp and Pearson (2007)

Due to the all-encompassing nature of the MEI definition of eco-innovation, distinguishing between four types of eco-innovators is not enough to capture the diversity of the concept. It is important to be able to classify eco-innovations themselves, according to the nature of the innovations involved. One such classification is proposed by Kemp and Foxon (2007a), who distinguish between: (i) environmental technologies, (ii) organisational innovation for the environment, (iii) product and process innovations offering environmental benefits and (iv) green system innovations. Each category includes a variety of items, which we briefly summarized below.

Environmental technologies gather together all pollution control technologies and cleaning technologies that treat pollution released in the natural environment. This includes: cleaner process technologies (i.e., new manufacturing processes that are less polluting and/or more resource efficient than relevant alternatives), waste management equipment (e.g., waste water treatment technologies), environmental monitoring and instrumentation, green energy technologies, water supply and noise and vibration control.

Organizational innovation for the environment refers to the introduction of organizational methods and management systems dealing with environmental issues encountered in the production process. Organizational methods include all pollution prevention schemes aimed at the prevention of pollution through input substitution, more efficient operation of processes and small changes to production plants (e.g., avoiding or stopping leakages). Environmental management and auditing systems include all formal systems of environmental management involving measurement, reporting and responsibilities concerning issues of material use, energy, water and waste (EMAS and ISO 14001 are examples of environmental management systems). Such organizational innovations may be

extended to the whole value chain. This chain management requires the commitment of a larger number of actors, since it involves cooperation between companies in order, for instance, to close material loops and to avoid environmental damage across the value chain (following products and treating hazardous materials "from the cradle to the grave").

Product and service innovation offering environmental benefits include all new or environmentally-improved products as well as environmentally beneficial service. These encompass: new or environmentally improved goods including eco-houses and buildings; green financial products (such as eco-leases or climate mortgages); environmental service (e.g., waste and water management, management, environmental consulting, testing and engineering); service that are less pollution and resource intensive (car sharing, for example).

Green system innovations are alternative systems of production and consumption, which are more environmentally benign than the existing ones. Biological agriculture and renewable-based energy systems are examples of green system innovation.

Although it is fairly widespread, this categorization is not the only existing one in the literature. For instance, Andersen (2005) has proposed a classification based on five categories of eco-innovation, which partially overlap those of Kemp and Foxon (2007a). The categories suggested in Andersen (2005) are: (i) add-on innovations (pollution- and resource-handling technologies or service), (ii) integrated innovations (cleaner technological processes or products), (iii) eco-efficient technological system innovations (new technological paths), (iv) eco-efficient organizational system innovations (new organizational structures) and (v) general purpose eco-efficient innovations.

Add-on innovations are products (artefacts or service) that improve the environmental performance of the customer. The product in itself need not be environmentally friendly. Integrated innovations are innovations which contribute to the solution of environmental problems within the company or other organizations (public institutions, families...). It is in this sense that they are integrated. Eco-efficient technological system innovations are innovations that represent a technological discontinuity. These radical innovations have wide systemic effects. They built on new theories, competencies and practices and may demand a change of both production and consumption patterns. Eco-efficient organizational system innovations entail new concepts for an eco-efficient way of organizing society. This means new ways of organizing production and consumption at the societal level, with new functional interplays between organizations, e.g. between companies ("industrial symbiosis"), between families and workplaces ("urban ecology"), etc.

Finally, **general purpose eco-efficient innovations** are technologies that affect the economy profoundly, because they have an underlying importance and feed into a range of other technological innovations. Information and Communication Technologies (ICT) may be the archetypical example of general purpose innovation in the context of "normal" innovations. While ICT as they are now are unlikely to have direct green effects (some even claim it may quite the contrary), it is important to look for eco-oriented changes in general-purpose technologies. These technologies are so fundamental that any "green change" they undergo will have a major effect on all other eco-innovations and on the environment itself. It is interesting to note that, in spite of some degree of overlap in their categorization, Kemp and Foxon (2007a) tend to disagree with this last point. They state that incremental innovations are important sources of productivity and that therefore, environmental improvements coming from incremental innovations. This may be because they are too optimistic as regards the likelihood of any radical eco-innovation appearing with general-purpose innovations in the near future.

Now that we have highlighted what eco-innovations may consist in, the next step is to consider how they may come into being. As for regular innovations, the generation of eco-innovations largely depends on the benefits received by the innovator. Successful innovations must provide higher value or reduce costs and, ultimately, either increase revenues from existing customers or attract new customers (Carrillo-Hermosilla *et al.* 2010). The eco-innovator may benefit from its innovative activity both directly and indirectly. According to Kemp and Andersen (2004), direct benefits for an eco-innovator consist in operational advantages such as cost savings from greater resource productivity and better logistics, and sales from commercialization. Indirect benefits include a better image, better relations with suppliers, customers and authorities, an enhanced innovation capability overall thanks to contacts with knowledge holders, health and safety benefits and greater worker satisfaction. Hence, the indirect benefits mostly have value in the long run, and while they may be overlooked by firms looking for short-run profits, they might well be the most important drivers for proactive green behaviour. These considerations lead us to the issue of drivers of and barriers to eco-innovation.

3. Drivers of and Barriers to Eco-innovation

For eco-innovation to unfold its full potential, firms and decision makers need to be able to identify the drivers of (and barriers to) eco-innovation. Policy makers need to be aware of the many drivers of eco-innovation (Kemp and Foxon, 2007a) in order to spur eco-innovativeness, while many eco-innovations fail or are abandoned because firms are unable to overcome the manifold barriers they encounter (Bleischwitz *et al.* 2009). Drivers are thoroughly explored in Section 2.1., while Section 2.2. deals with barriers.

3.1. Drivers

The innovation literature has long been dominated by two contrasting explanations of what drives technological change: the push of the technology itself ("technology-push") on the one hand, and the pull of the market ("demand-pull") on the other. Empirical evidence has shown that both forces are very likely to be at play in the actual world (Pavitt, 1984). It is therefore necessary to examine drivers and barriers from both the supply side (where technology push is supposed to occur) and the demand side (which exerts the market pull). As far as eco-innovation is concerned, technology push factors include all new eco-efficient technologies, whereas market pull (or demand pull) factors include consumers' preference for environmentally friendly products and the need for companies to maintain their environmentally-responsible reputation (Rennings, 1998). Environmental innovation and "normal" innovation notably differ in that institutional and political factors are likely to play an even more important role in the former than in the latter. The main features of ecoinnovations are correlated to the determining role of regulations. Table 1 lists these three types of determinants: (1) technological and supply-side related, (2) market and demand-side related, and (3) policy related. Since environmental policy may interact with both supply and demand sides, it will be examined in the next section, and the present section focuses on the first two types of determinants.

3.2. Barriers

Broadly defined, barriers to eco-innovation include every element that may hinder the development and/or diffusion of environmental innovations. Rather than thinking in terms of supply and demand, as was done for drivers, the literature usually classifies barriers and drivers into categories such as political, informational, financial etc. (Bleischwitz *et al.* 2009).

Among these categories, informational and financial barriers are often deemed to be the most important ones. Without being exhaustive, we can consider here some established typologies.

For instance, Bleischwitz *et. al.* (2009) consider the following barriers to eco-innovation: (*i*) **informational barriers** arise from an asymmetric distribution of knowledge about material and resource efficiency among various actors, such as users and producers; (*ii*) **financial barriers** are generally the result of a splitting of financial incentives between actors (e.g., between user and investor) with contrasting interests as regards the introduction of eco-innovation; (*iii*) **a gap between R&D and market launch** often occurs when the risks associated with R&D expenditures are high, in which case a firm will only accept to act as a "first mover" (i.e. to introduce an eco-innovation) if it can benefit from a sufficient patent protection.

Reid and Miedzinski (2008), relying on an analysis of CIS3¹⁶ indicators, found that the most significant barriers for firms that are classified as eco-innovators, were (1) the high costs of innovation activity (for almost 30% of these firms), (2) the lack of an appropriate source of funding (for %23) and (3) the excessive economic risks (perceived by around 20% of these firms). This analysis thus points to financial barriers as the most important ones. However, one should be aware that any analysis of barriers conducted using CIS data is likely to be biased by the fact that the data refer to barriers encountered by innovators, and not to barriers that may have caused a firm to remain a non-innovator.

Reid and Miedzinski (2008) suggest that, whatever the data limitation, any classification of drivers and barriers to eco-innovation is difficult because it depends to a large extent on the cultural, institutional and historical context of the country. According to his conclusion, socio-cultural factors that can be considered as barriers to eco-innovation are (1) low or low-quality education at all levels, (2) low environmental awareness and lack of clear information, (3) low openness of society (e.g., "fear of change", closed networks, risk aversion, etc.), (4) limited access to human resources and expert knowledge, (5) lack of organisational capacities, (6) persistent power structures within societies ("institutional inertia" and historical path dependency), (7) short-term decision-making, (8) weak social corporate responsibility. The risk with such an approach, however, is to be of little practical usefulness if what is really needed for the study of eco-innovation is a general model or framework. In this case, saying that the object of the study is mostly country-specific for cultural and historical reasons is hardly a step towards such a framework.

¹⁶ Third Community Innovation Survey, conducted in 2001 across all EU Member States.

Institutional and social barriers certainly have to be taken into account, nevertheless; and they precisely seem to be absent of classifications such as the following one, proposed in European Commission (ETAP, 2004). This classification distinguishes: (*i*) economic barriers, ranging from market prices which do not reflect the external costs of products or service (such as health care costs due to urban air pollution) to the higher cost of investments in environmental technologies (e.g., because of the complexity of switching from traditional to "green" technologies); (*ii*) regulations and standards can also act as barriers to innovation when they are unclear or too detailed, while a good legislation can stimulate environmental technologies; (*iii*) insufficient research efforts, coupled with an inappropriate working of the research system in European countries and weaknesses in information and training; (*iv*) lack of market demand from the public sector, as well as from consumers. The weakness of this typology is that almost every barrier it lists is of an economic nature, and it fails to encompass purely technological barriers as well as social factors.

In the end, one has to go back to a work such as that of Ashford (1993) to find an extensive and complete list of barriers. Those are: (*i*) **Technological barriers** (e.g., unavailability of certain technologies for specific applications, or lack of alternative substances to substitute for the hazardous components; (*ii*) **Financial barriers** (e.g., costs of R&D, or non-comprehensive cost evaluations and cost-benefit analyses leading to a lack of funding for eco-innovation); (*iii*) **Labour force-related barriers** (e.g., lack of persons in charge of the management, control, and/or implementation of waste reduction technology); (*iv*) **Regulatory barriers** (e.g., lack of incentives to invest in reuse and recovery technologies); (*v*) **Consumer-related barriers** (e.g., risk of customer loss if output properties change slightly or if product cannot be delivered for a certain period); (*vi*) **Supplier-related barriers** (lack of supplier support in terms of product advertising, good maintenance service, expertise of process adjustments, etc.); (*vii*) **Managerial barriers** (e.g., lack of top management commitment, or reluctance on principle to initiate change in the company).

Ashford (1993) emphasizes that most of the barriers listed above can be disaggregated to a more detailed level. For instance, the "lack of top management commitment" may be caused by various factors. It could come from a lack of information regarding the profitability of waste reduction technologies in general. It could also stem from a lack of confidence in the performance of new technologies, or, more crucially, from short-sightedness: if the top management is not aware of the long-term benefits that may be derived, for instance, from waste reduction, then waste reduction will remain a low-priority issue.

Once a general framework such as the one above has been proposed, further analyses may help identify which type of barriers to eco-innovation should be addressed first, and which type could be considered as secondary. This is where the institutional and cultural context matters, since the most important type of barriers may differ across economic sectors and even across countries (Hitchens *et al.*, 2003). For instance, in a study of environmental performance in EU countries, Kemp and Foxon (2007a) found that financial barriers were more important in Germany than in the UK, whereas lack of time was a greater barrier in the UK than in Germany.

4. Environmental Policy

4.1. Recognition of Environmental Policy as a Driver of Eco-innovation

Empirical evidence shows that the regulatory framework and environmental policy have a strong impact on eco-innovation (Green *et al.* 1994; Porter and van der Linde 1995; Kemp, 1997; Hemmelskamp, 1997; Cleff and Rennings, 1998; Berman and Bui, 2001). However, according to Rennings (1998), environmental *product* innovation tends to be more driven by the strategic market behaviour of firms (market-pulled effect). By contrast, regulation tends to drive more environmental *process* innovation, because the public-good character of clean technology leads to under-investment in environmental R&D (Rennings and Rammer, 2009). Therefore, environmental regulation is especially necessary to foster environmental process innovation. When the reduction of the environmental impact of a firm's activity offers little operational or commercialization benefits, then regulation may become the primary driver of eco-innovation (Kemp and Foxon, 2007a). For example, regulations to protect local air quality have stimulated an innovation such as catalytic converters, which have led to dramatic reductions in the emission of pollutants from vehicles (Kemp and Foxon, 2007a).

Therefore, environmental policy is a potentially strong driving force for eco-innovation, which deserves to be studied separately. Environmental policies may fall under the "command-and-control" or "market-based" types. Market-based instruments such as pollution charges, subsidies, tradable permits, and some types of information programs can encourage firms or individuals to undertake pollution control efforts that are in their own interests and that collectively meet policy goals (Jaffe *et al.*, 2002). By contrast, command and control regulations tend to force firms to take on similar shares of the pollution burden,

regardless of the cost. They often do this by setting uniform standards for firms, the most prevalent of which are performance- and technology-based standards (*ibid*).

In the wake of the Lisbon agenda, the European Council adopted in 2001 the Sustainable Development Strategy and introduced the "Environmental Technologies Action Plan" (ETAP) in 2004. In the ETAP, the European Union acknowledged the strategic importance of eco-innovation. Although the actual impact of ETAP remains to be precisely assessed, it seems to have led to an increased recognition of environmental problems (and of the need for eco-innovation as an answer) in the public and political consciousness.

The results from a survey across ten OECD countries show that an increasing number of countries now perceive environmental challenges not as a barrier to economic growth but as a new opportunity (OECD, 2009). This new understanding has made environmental policy appear as an important driver of eco-innovation, thus reconciling real-world policy with the theoretical considerations of Porter and Van der Linde (1995). These authors argued in the mid-nineties that environmental progress requires companies to improve their resource productivity through dedicated innovation, so that regulation becomes not an obstacle but a driver for innovation. In Porter and Van der Linde (1995), implicitly, the more prescriptive the regulation is, the more confined will the innovation be.

More recently, Rennings (2000) also emphasized that environmental policy is becoming the main driver of eco-innovations. According to him, eco-innovations differ from normal innovations because they produce a double externality, consisting in (1) the usual knowledge externalities in the research and innovation phases and (2) externalities in the adoption and diffusion phases due to the positive impact upon the environment (Oltra, 2008). In other words, the beneficial environmental impact of environmental innovations makes their diffusion always socially desirable. However, these positive external effects lead to market failures which may hinder eco-innovation. The private return on R&D in environmental technology is less than its social return due to its public good nature, which in turn causes a lack of private incentives leading firms to under-invest in environmental R&D and innovation (Oltra, 2008). Therefore, environmental policy and/or an appropriate regulatory framework appear as a requirement for eco-innovation.

4.2. Effectiveness of Environmental Policy as a Driver of Eco-innovation

According to Porter and van der Linde (1995), environmental standards can foster innovation under three conditions. First, they must create maximum opportunities for ecoinnovation, letting the industry (and not a standard-setting agency) choose its own approach to innovation. Second, regulation should foster continuous improvement, rather than locking in any particular technology. Third, the regulatory process should leave as little room as possible for uncertainty at every stage.

Therefore, the type of regulation/policy and the way it is implemented is significantly important. It should lead firms to effectively address environmental problems rather than restrict firms in a specific technology and leave the environmental problem unsolved. The stringency of the policy and the terms in which it is defined are equally important, since uncertainty depends on these factors. In spite of on-going controversies on whether environmental regulation actually has an impact on innovation and on the most efficient policy instruments (see for instance Greenstone, 2002; Jaffe *et al.*, 2002), many empirical studies (European Commission, 2001; Rennings *et al.* 2006, Belin *et al.* 2009) find a positive correlation between innovation and regulation.

Porter and van der Linde, (1995), Kemp *et al.* (1998) and, Jänicke and Jacob (2002) all predict that strict environmental regulations stimulate innovation in a number of ways (e.g. first mover advantages created by the development of "green" technologies). These predictions are in line with the so-called "Porter hypothesis" postulates that "there are win-win opportunities through environmental regulation, where simultaneously pollution is reduced when having an increase in productivity". As mentioned above, this hypothesis has fuelled controversies¹⁷, but its argument remains at the core of current research on eco-innovation.

The rationale behind this argument is that firms do not detect the potential of environmental innovations because they are "still inexperienced in dealing creatively with environmental issues" (Porter and van der Linde, 1995, p. 99). Environmentally and economically benign innovations are not realized because of incomplete information, and of organizational and/or coordination problems (*ibid*). Firms are not able to recognize the cost saving potentials (e.g. energy or materials savings) of environmental innovation (Frondel et al. 2007). This leads many of them to believe that an environmentally-virtuous behaviour is a burden rather than an asset (Kemp and Andersen, 2004). Therefore, regulations and policies can be a catalyst and help them to understand the potential benefits of environmental innovations.

¹⁷ For instance, Jaffe *et al.* (2002) state that while it is theoretically valid, its relevance with regard to the magnitude of win–win potentials is debatable.

However, as Porter and Van der Linde (1995) hinted at, the types of environmental policies used matter. Popp (2009) argues that in general, market-based policies are thought to provide greater incentives for innovation, as they provide rewards for continuous improvement in environmental quality. In contrast, command-and-control policies penalize polluters who do not meet the standard, but do not reward those who do better than mandated as the command-and-control regulations direct a specific level of performance, such as pounds of sulphur dioxide (SO2) emissions per million BTUs of fuel burned, or the percentage of electricity that must be generated using renewable sources.

Popp (2009) also argues that what matters is not just the type of policy instruments used: differences within one policy type (e.g. market-based policies) may also matter. His ideas find some support in the results of Johnstone *et al.* (2008), who analyse the effect of different policy instruments on renewable energy innovation in 25 OECD countries. In their study, they compare price-based policies such as tax credits and feed-in tariffs¹⁸ to quantity-based policies such as renewable energy mandates. They find that quantity-based policies favour the development of wind energy while, by contrast, direct investment incentives are effective in supporting innovation in solar and waste-to-energy technologies. However, designing such instruments efficiently for different pollutants is sophisticated and costly.

Moreover, the introduction of such policies has often met with opposition from at least some industries (e.g., Arimura *et al.* 2008). As a result, in recent years, voluntary proactive approaches to environmental protection are considered useful supplements to traditional mandatory command and control regulations and economic incentives (e.g., Khanna and Damon, 1999; Alberini and Segerson, 2002).

4.3. Proactive approaches to environmental protection

The most important voluntary approaches to environmental protection are subsumed under ISO 14001, EMAS¹⁹ and 33/50, the first two being international and the latter being specific to the US. ISO 14001 comprises standards that must be adopted. EMAS requires facilities – besides third-party audits with independent environmental verifiers and registration bodies – to publish an environmental statement. Finally, 33/50, which was launched by the U.S. Environmental Protection Agency (EPA) in 1991, aimed at reducing aggregate emissions of 17 chemicals reported to the Toxic Release Inventory (TRI⁾ (Ziegler

¹⁸ Feed-in tariffs, used in various European countries, guarantee renewable energy producers a minimum price for the electricity they produce

¹⁹ Eco-Management and Audit Scheme

and Nogareda, 2009). The main reasons for introducing these voluntary approaches are (1) that they are more flexible and cheaper than government based policies, and (2) they are likely to improve corporate environmental performance. For example, EMAS-certified facilities in Germany benefit from regulatory relief and from more (and higher) subsidies based on the EMAS privilege regulation (Ziegler and Nogareda, 2009). In line with these implications, Rehfeld *et al.* (2007) emphasizes that a careful design of EMAS is important for both the environmental and economic performance of a facility. It is important to note that public voluntary programs such as 33/50 do not imply any penalties for withdrawal at any time. In other words, these measures do not guarantee an improvement in environmental performance (Anton et al. 2004). For this reason, critics have arisen that consider Voluntary Environmental Programmes (VEP) as "green washing". According to these critics, VEP fail to lead participants to clean their operations due to the absence of significant obligations or enforcements (e.g., Potoski and Prakash, 2005). However, some studies, such as Fischer et al. (2003), do not find a clear ranking when comparing VEP to other instruments. Fischer et al. (2003) claim that no instrument is generally preferable, and the welfare gain of environmental policy instruments (VEP, command-and-control policies, and market-based regulation) depends on different sets of circumstances, such as the number of polluting firms, the costs of an innovation, and the costs of imitating an innovation. As a result, the debate on appropriate policy instruments for developing eco-innovation is still open and the definition of an adequate policy may change depending on the situation. This asserts that more empirical investigation needed to stress the importance of voluntary proactive approaches to stimulate eco-innovation.

The OECD (2009)'s policy brief discussion pays considerable attention to notable policies that have actually been introduced in various countries. This discussion gives examples and provides detailed examples of certain policies such as the United States Department of Energy (DOE)'s Technology Commercialization Fund (TCF). This **funding for technology deployment** complements "angel investment", or early-stage corporate product development. The TCF brings the DOE's national laboratories and industry together to identify technologies that are promising, but face the "commercialization valley of death" (OECD, 2009). Another programme is Japan's **Top Runner Programme**, launched in 1998, which adopts a process of setting and revising standards by taking the current highest energy-efficiency rate of products in some 21 product groups as a benchmark, instead of setting fixed targets. This flexible standard-setting creates incentives and competition among manufacturers to improve product performance without providing financial support. The

European Automobile Manufacturers Association (ACEA) voluntary agreements is another programme discussed within this discussion, these agreement are an important element of the EU's strategy to reduce CO2 emissions from passenger vehicles and more generally to improve fuel efficiency. With these agreements, the automobile industry committed to reduce total new passenger fleet average CO2 emissions according to specific targets and timetables (Leitner et al. 2010). One interesting distinction between the Japanese Top Runner programme and the corresponding European Automobile Manufacturers' Association approach (based on voluntary agreements) is that the ACEA sets standards at the industry level, while the Top Runner Programme sets standards at the company level. This latter approach has the advantage that companies are more directly involved. It is remarkable that only about half of the European car manufacturers mentioned the ACEA standard in their annual reports (Tressel et al., 2007). As a last example, let us mention the "Bonus-Malus" (reward-penalty) scheme introduced in France in 2007 to encourage manufacturers of personal cars to develop low-emission vehicles by guiding consumer choice. This carrot and stick scheme for diffusing eco-innovative product provides a subsidy to those who purchase a new car which emits less than 130 grams of CO2 per kilometre, and imposes a penalty on those who buy a new car that emits over 160 g CO2/km (Popp, 2009). As mentioned above, these policies based on goodwill are supplements to traditional environmental policies and therefore do not "punish" the failing firms or manufacturers. The "bonus-malus" programme implemented in France could be considered as an exception, but it must be noted that the "stick" in this programme targets consumers, and not manufacturers.

5. Diffusion of Eco-Innovation

5.1. From Normal Innovation to Eco-innovation

The last policy example examined in the previous section leads us to the issue of the diffusion of eco-innovations. In the innovation studies literature, Rogers (1962:2010) was among the first to study the diffusion of innovations, using an original approach based on well-established theories in sociology, psychology, and communications. According to Rogers (1962), diffusion is "the process by which an innovation is communicated through certain channels over time among the members of a social system". Therefore, diffusion can be considered as a special type of communication concerned with the spreading of messages that are perceived as new ideas for the potential users.

Moreover, diffusion is a complex process. The widespread diffusion of new technologies can take anywhere from between five to fifty years, depending on the innovation (Mansfield, 1968). This is mainly because of the perceived uncertainty and the risk about the new technology. Countless studies have confirmed that the diffusion of new technologies follows a predictable temporal pattern; technologies are adopted rather slowly at first, then more rapidly, and then slowly again as a technology specific "adoption ceiling" is reached. Thus, a plot of the number or the proportion of adopters over time assumes a sigmoid shape (Blackman, 1999) as Rogers has indicated in his innovation diffusion model (Rogers, 1983). Therefore, Moore (1998) suggested that the early stage of new product development hinges on innovators and early adopters, as their reaction to the innovation will determine the product success or failure. Early adopters play a crucial role in the diffusion process, because they are taken as examples for others to follow.

In the case of eco-innovation, the diffusion of clean technologies is also likely to follow an S-shaped pattern, although far less research has been done on clean than on "normal" technologies. Kemp and Volpi (2007) explain that diffusion of a clean technology is initially slow because it is initially not well known, and surrounded with even greater uncertainty than technologies involved in "normal" innovation. In spite of this uncertainty, there are good reasons to support the development and diffusion of eco-innovations. First and foremost is an environmental reason: the world population continues to increase and there are demands for higher standards of living, with more consumption leading to increased pollution, climatic change, and the depletion of natural resources and biodiversity (Cawsey, 1996). In this context, eco-innovation may be the only way to sustain the current quality of life (or even increase it) in the long run.

Among secondary reasons for supporting the diffusion of eco-innovations is a purely economic reason. Eco-industries (e.g., air pollution control, waste water management, solid waste management, soil remediation, and renewable energies and recycling) are among the fastest growing industries in the world and they are likely to be worth around 2\$ trillion by 2020 (European Commission²⁰, 2012). As an example, European eco-industries are a small but quickly growing portion of the EU economy representing 2.1% of the EU's GDP and sustain 3.5 million full-time jobs in the EU. Around 75% of these jobs are in labour intensive sectors such as water and solid waste management (European Commission, 2004). We can

²⁰ The EU's, Number of Jobs Dependent on the Environment and Resource Efficiency Improvements project, Final Report, 2012.

thus expect that, as the awareness of environmental issues increases, eco-industries will create greater opportunities for new employment.

5.2. Hindrances to Diffusion of Eco-innovation

There are specific problems inherent to eco-innovative activities which make the diffusion of eco-innovations difficult. One of the main problems is the double externality issue mentioned in Section 3.1: since this double externality leads firms to under-invest in eco-innovation, it also strangles the diffusion of eco-innovations at its very root.

Another problem associated with eco-innovation is that in many cases the adoption and the diffusion of environmental technologies can be viewed as a typical case of technological competition between an established technology and an alternative environmental technology, or a set of alternative environmental technologies (Oltra, 2008). In order to be successful in the diffusion process, the alternative technology should become a viable substitute to the existing technology. The more efficient the new environmental technology is on the mainstream characteristics, the more likely its diffusion becomes (Christensen, 1997). But, the superiority of the eco-innovation compared to the established technology is always a debatable issue and needs further investigation. Moreover, most industries shy away from technological environmental innovations in the first instance because of their complexity and of the financial risks they entail.

The heterogeneity of potential adopters only adds to the second problem: a technology that is generally superior will not be equally superior for all potential users, and may remain inferior to the existing technology for some users for an extended period of time after its introduction (Jaffe *et al.* 2002). Therefore, it is important to understand how potential users, be they consumers or firms, value the environmental characteristics of the new technology. Very often, clean technologies or products are adopted at the expense of other technologies or other products. As a result, potential users want to gain some advantages from their investments (e.g., an increased competitiveness in the case of firms that have adopted a clean technology) and generally want to know the expected returns before the implementation process.

In general, when users are consumers (and not firms) they do not have adequate information about the new technology or product. This lack of adequate information constitutes another great obstacle to the diffusion of the new technology. Consumers have to cope with the typical problem of experience goods since they cannot know *ex ante* the environmental characteristics of goods and/or technologies (Oltra, 2008). This issue gives

more importance to communication channels, since better explanations and increasing awareness about environmental issues may enable a more widespread adoption of ecoinnovations. Increasing awareness about the shortcomings of the existing technologies and identifying the required technological improvements may also increase the speed of adoption. This is the reason why policy instruments such as information provision and eco-labels are necessary to inform consumers efficiently and to stimulate the diffusion of environmental products (Oltra, 2008). For example, ISO 14000 standards can be a good tool for increasing environmental awareness. All the above-mentioned obstacles raise the question of how to give a kick not only to the creation of eco-innovations, but also to their diffusion process. So far, however, the empirical literature seems to have focused on the determinants and/or effects of eco-innovation, or on testing the Porter Hypothesis. There is still scope for empirical work on the diffusion processes of environmental innovations.

6. Conclusion

Conventional or "normal" innovation has been the subject of countless research until now. However, eco-innovation, as a specific form of innovation, still deserves further exploration. The present survey summarized the most relevant and prominent researches in the field, in order to highlight the most important insights.

Our review of theoretical researches shows that the concept and definition of ecoinnovation is still evolving, although all current definitions of eco-innovation tend to be based on environmental performance rather than environmental aim. Moreover, the degree of novelty required for a product or process to be defined as an innovation is fairly minimal: it is sufficient for the product or process to be new at the firm level rather than, for instance, at the level of an industry or an economy. Direct and indirect benefits perceived by the innovator have significant impact on eco-innovativeness. The direct benefits are relevant in the short term while, the indirect benefits are more relevant in the long term. The literature also asserts that, besides the conventional supply-side and demand-side determinants of innovation, policy and regulation appear as specific drivers of eco-innovation.

Empirical evidence shows that environmental policy and an appropriate regulatory framework both have a strong impact on eco-innovation. Environmental product innovation tends to be more driven by the strategic market behaviour of firms (market pull), whereas policy and regulation tend to drive environmental process innovation, because the public-good character of clean technologies leads to under-investment in environmental R&D. Nevertheless, the impact of some of the less stringent environmental regulations (e.g.,

voluntary proactive approaches such as; ISO 14001 and EMAS) on the introduction of ecoinnovation remains controversial. This suggest that more in-depth empirical investigation to stress the importance of voluntary environmental standards and norms for the eco-innovation process is beneficial to support (or contradict) the so-called Porter hypothesis. Last but not least, more (empirical) research on the diffusion processes of eco-innovation is also needed. If it is as beneficial as the Porter hypothesis suggests (i.e., likely to overcome environmental problems without hampering economic performance), then a better grasp of its adoption mechanisms is still required, so that it can be efficiently encouraged by policy makers among both firms and consumers.

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Chapter 2

Voluntary environmental standards: A new strategy to promote greener business?

Table of Contents

1. Introduction
2. Environmental Management Systems and ISO 14001 standards73
2.1. A brief history73
2.2. Requirements to adopt an EMS73
2.3. EMS, Firm performance and the Porter hypothesis
3. Data and Research Methodology
3.1. Data
3.1.1. Dependent variable and Treatment group
3.1.2.Explanatory variables and Descriptive statistics
3.2. Methodology
3.2.1. Propensity Score Matching and Fully Interacted Linear Matching84
4. Econometric Results
4.1. Estimations obtained with PSM85
4.2. PSM Estimates of the ATT90
4.3. Robustness of the Matching Method91
5. Conclusion

Voluntary environmental standards: A new strategy to promote greener business?

Abstract

The direct impact of Environmental Management System (EMS) on firms' business performance is still a debated issue. The aim of this research is to identify the organisational capabilities of firms that facilitate the adoption of "ISO 14001 type" standards and the impact of their adoption on (early and late) adopters' business performance. The business performance of a firm is measured through Total Value Added (hereafter VA). This empirical analysis is based on firm-level data acquired from a large French survey (COI 2006) matched with administrative data, comprising 11,168 manufacturing and service firms. Empirical results obtained by the Propensity Score Matching (PSM) and Fully Interacted Linear Matching (FILM) methods showed that a firm's structural characteristics, organisational capabilities, business market and technological complexity are some of the factors that have a significant impact on a firm's propensity to adopt voluntary EMSs. We also showed that the respective marginal effect of these factors may differ between the early and late adopters of the EMSs. Last but not least, we observed that the adoption of voluntary EMSs may have a significant impact on adopters' VA but we did not observe any significant difference between the early and late adopters of the standards with respect to return on VA.

1. Introduction

The Porter Hypothesis (PH) (Porter and van der Linde, 1991; 1995) is one of the most investigated issues within the environmental innovation (eco-innovation) literature. Many economists and policy makers give considerable attention to the claims of the Porter Hypothesis as it has challenged the foundations of the established opinion in regards to the relationship between environmental regulations, innovativeness and firm performance. According to one of the first assumptions of this hypothesis, firms are not profit maximizing organisations, hence they are not able to capture available market opportunities without proper and well-designed environmental policies and regulations. These environmental policies and regulations assume to help organisations in reducing their carbon foot-print through environmental innovations. More precisely the PH postulates that;

"[r]educing carbon foot-print (or pollution) is often coincident with improving productivity with which resources are used and strict environmental standards can encourage innovations that increase competitiveness. Thus, the right environmental policies can greatly reduce the costs of environmental policies and can even make companies more profitable by creating "win-win" opportunities through environmental regulation, where simultaneously pollution is reduced when having an increase in productivity (Porter and van der Linde, 1991, 1995).

Even though the predictions of the PH sound appealing, the hypothesis has been heavily criticized by many scholars (i.e. Palmer, Oates and Portney, 1995). The main criticisms relate to the fact that the hypothesis is based on case studies, hence far from generalization, and incompatibility with the assumption that firms are profit maximizing firms. Therefore, researchers have generally disaggregated the PH into its component parts ("weak²¹", "narrow" and "strong²²" versions) in order to test the theory and collect empirical evidence. This research distinguishes itself from the existing empirical investigations by testing the validity of the "narrow" version of the PH, which argues that flexible regulatory policies give firms greater incentives to innovate and thus are better than prescriptive forms of regulation (Jaffe and Palmer, 1997).

²¹ The "weak" version of the hypothesis claims that properly designed environmental regulation may spur innovation.

²² The "strong" version of the hypothesis claims that properly designed environmental regulation may spur innovation.

Accordingly, in the first step of our investigation we try to identify the structural characteristics and organisational capabilities of firms that facilitate the adoption of voluntary Environmental Management Standards (EMS). In the second step, we examine the impact of voluntary EMS adoption (i.e., ISO 14001 and Eco-Management and Audit Scheme (EMAS) - type standards) on (early and late) adopters' business performance. A firm's business performance is measured by its Total Value Added with factor costs (hereafter VA). These types of voluntary standards are considered as the most efficient tools in order to support the market-based instruments (i.e. tradable permits, emission taxes, subsidies etc.) as they do not impose environmental performance requirements but helping organisations to form their own EMS. In addition, they do not aim at improving adopters' business performance. However, they may help firms to introduce product improvements, to reduce their carbon foot-print and input cost, and increase employee, consumer, investor, shareholder and insurer trust by reflecting environmentally friendly image (ISO²³). This 'greener' image may also provide substantial marketing advantages to capture additional market share, provide access to green markets, reduce insurance charges and input costs, which in turn may have a significant impact on a firm's business performance. Some recent investigations (i.e., Hart and Ahuja, 1996; Newell et al., 1999; Berman and Bui, 2001; Konar and Cohen, 2001) show that there may be a significant relationship between environmental regulations and a firm's business performance but the impact of the regulation depends on different factors such as regulation stringency of environmental regulation, timing of regulation, measure of business performance etc. In this research, we explicitly introduced a distinction between the 'early' and 'late' adopters of voluntary EMS depending on the adoption date. This type of distinction is non-existent in the eco-innovation literature and thus, it is one of the most significant added-values of this research. The data constraint for the empirical analyses is acknowledged as a common problem for eco-innovation studies. To tackle this problem the empirical analysis is based on firm-level data acquired from a large French survey (COI 2006, Organisational Changes and Computerization) matched with administrative data (EAE, The Annual Enterprise Survey). This new dataset comprises 11,168 manufacturing and services firms, and it allows us to investigate all significant differences between relatively "early" and "late" adopters of voluntary EMSs. This paper is organized as follows. Section 2, reviews the related literature of EMS and provides insights about the relationship between the EMS and the PH. We present our data and discuss our research methodology in section 3. We will

²³ International Organization for Standardization

present the econometric results in section 4 and finally, in section 5, we draw some conclusions and offer direction for further research.

2. Environmental Management Systems and ISO 14001 Standards

2.1. A brief history

EMS is a management structure that provides firms with a framework to minimise their environmental impact, ensure compliance with environmental laws and regulation, and reduce wasteful uses of natural resources. EMS could be integrated in a firm's daily operation via different environmental standards (e.g. ISO 14001, EMAS - Environmental Management and Audit Scheme, Responsible Care, etc.). These central Voluntary Environment Programmes (VEPs) are unilateral agreements by firms regarding EMS and the most important program in this respect is the ISO 14001 (Ziegler and Nogareda, 2009). This standard is part of ISO's environmental management standards family (ISO 1400) and it helps organisations to form their own EMS. They are considered as one of the most efficient tools to support the market-based instruments (i.e. tradable permits, emission taxes, subsidies etc.) and they do not impose environmental performance requirements (Khanna and Damon, 1999; Alberini and Segerson, 2002). This feature of the standards ensure that organisations are responsible for setting their own environmental targets and make the standards applicable to wide variety of organisations, in size and in occupation, (such as; manufacturing companies, branch offices, plants, construction sites, refineries, administration centres, etc.).

2.2. Requirements to adopt an EMS

Implementation of an EMS requires several steps to be taken and successful implementation of the standard depends on how well the 'Plan, Do, Check, Act (PDCA)' model is applied. The main components of the standards include 1) environmental policy, 2) performance goals: objectives and targets, 3) environmental programs: action plan to meet objectives, 4) roles and responsibilities, 5) training and awareness, 6) communication—internal and external, 7) documentation of the system, including procedures or operational controls, 8) monitoring, measurement and record keeping, 9) procedures for corrective and preventive action, 10) Environmental Management System audits and 11) management reviews (adapted from ISO²⁴). Capturing the potential advantages of the ISO 14001 standards

²⁴ Environmental Management Systems, Requirements with Guidance for Use, ISO 14001

depends on a firm's capabilities (structural, organisational, technical, networking etc.) to detect and reduce its eco foot-print. Therefore, these steps make sure that the organisations identify and minimise the negative effect of their operations, comply with existing laws and continually improve in the above. Darnall *et al.* (2001) assert that basically the documentation requirements of the standard itself may increase firms' understanding of how their operations affect the natural environment. Hence, following these requirements may cause firms to systematically analyse some of their environmental impacts for the first time and lead them to consider their impacts related to various inputs and their production process.

2.3. EMS, firm performance and the Porter hypothesis

EMS is not designed to improve a firm's business performance. However, the relationship between the EMS and firm performance is a significant research question as its potential advantages can benefit industries and society at large. If successfully implemented, the standard may provide substantial financial return via different channels. For example; firms may reduce input cost (raw materials, energy, water, labour), mitigate environmental liability and create a sustainable brand image that can attract new customers and/or reduce trade barriers to reach certain markets (i.e. European markets). Some studies claim that welfare gain of environmental policy instruments depends on different sets of circumstances (Fischer et al. 2003) (i.e. on the number of polluting firms, the costs of an innovation and the costs of imitating an innovation) and none of the policy instruments is generally preferable when the endogeneity of technological progress is taken into account, as is done in evolutionary economics as well as in the new institutional and growth theory (Rennings and Rammer, 2009). According to Hart and Ahuja (1996) there is a positive relationship between emissions reductions and financial performance. However, there is a two-year lag until financial performance benefits are reaped. Newell et al. (1999) showed that environmental regulations may indeed induce energy efficient eco-innovations. By estimating the impact of local air pollution regulation on abatement investment, Berman and Bui (2001) showed that heavily regulated firms face Total Factor Productivity increases compared to non-regulated ones. However, Konar and Cohen (2001) claimed that legally emitted toxic chemicals have a significant effect on the intangible asset value of publicly traded companies and, therefore, on their future performance. Furthermore, some empirical investigations (Delmas and Pekovic, 2013, Frondel et al., 2007) show that even implementing an EMS may have a positive effect on a firm's performance such as improving internal efficiency, reducing costs or enhancing the firm's reputation and providing access to green markets. Hence, the adoption of an EMS

may allow a firm to decrease their environmental liabilities while having an increase in their performance, which is called "win-win" opportunity for firms. The "win-win" opportunity through environmental regulation was first proposed by the Porter Hypothesis. Based on case studies, the PH asserts that strict environmental regulations may help firms to exploit existing market opportunities via development and/or adoption of eco-innovations, and the cost of the regulation might be mitigated with the productivity gains that can be accrued through realized/adopted eco-innovation (Porter, 1991; Porter and van der Linde, 1995). Nevertheless, even after two decades, the literature does not provide clear evidence to support or contradict the PH. Ambec et al. (2013) states that empirical studies that test the validity of the PH generally fall into three categories: those testing the "weak" version of the hypothesis, those testing a "strong" version focused on firm-level performance, and those testing a "strong" version focused on country-level competitiveness. We contribute and extend the discussion about the PH by investigating another particular case of the hypothesis, namely the "narrow" version of the PH, which argues that flexible regulatory policies give firms greater incentives to innovate and thus are better than prescriptive forms of regulation (Jaffe and Palmer, 1997). As mentioned above, EMS is not designed to improve a firm's business performance therefore, mitigating the EMS implementation cost is not the primary aim of this type of management standards. As the financial return is not guaranteed, identifying the organisational capabilities of a firm that facilitates the adoption of an EMS and investigating the impact of this adoption on a firm's business performance becomes a highly significant research question. Many of the previous studies that investigate the determinants of productivity sensu lato have suffered from several technical difficulties such as lack of adequate data, missing environmental regulation stringency, timing and firm performance measures that does not allow for complete empirical investigation. According to Ambec et al. (2013) some of the researchers have often regressed productivity at time 0 on proxies of environmental regulation stringency at time 0 as well. Our extensive database, however let us to overcome this challenge and allow us to consider this time lag between regulation and its actual effect on firm performance. We therefore, introduced a clear distinction between relatively "early" and "late" adopters of voluntary EMSs by considering the adoption year. By doing so, we could identify the relationship between the adoption of an EMS and its impact on early and late adopters business performance and, provide robust empirical evidence to support/contradict the "narrow" version of the PH.

3. Data and Research Methodology

3.1. Data

This research is based on two data sources. The first one is the French "Organisational Changes and, Information and Communication Technologies use 2006" (COI 2006) survey. The COI survey provides a matched employer / employee dataset on organisational change and ICT use, compiled through the joint efforts of the INSEE (National Institute for Statistics and Economic Studies), DARES (Ministry of Labour) and CEE (Centre for Labour Studies). This comprehensive survey contains various questions related to changes and is addressed to companies and to their employees. The latest survey was conducted in 2006 on a representative sample (7700 firms) of the population of French firms from all industries except agriculture, forestry and fishing. Each of these firms filled in a questionnaire concerning the use of ICT and organisational practises in 2003, in 2006 and during the period in between. The COI 2006 survey is a retrospective survey in which firms answer in 2006 a questionnaire that includes questions pertaining to 2003 and to the period in between. Furthermore, the survey design allows for an easy matching with administrative databases. In order to obtain information about value added, material expenditures, number of employees and some other accounting variables, we merged the COI survey with another French administrative dataset; the Annual Enterprise Survey (EAE). As a result of this merger, we worked with a panel of 11,168 observations (5584 firms observed in 2003 and in 2006) of firms with more than 11 employees. We tried to eliminate the outliers by excluding firms that employ less than 11 employees and have more than 100% annual growth rate (in terms of totals sales). Table 1 presents "before merger" and "after merger" means and the standard deviation of all the key variables used in our empirical analyses. The test for differences in means indicates that merging datasets did not lead to severe selection biases and there aren't any significant changes in mean values of the key variables before and after merger. Therefore, we can use the merged database for our empirical investigations.

		EAE			
Variables:	Mean	Std. Dev.	Mean	Std. Dev.	P-value
Firm Size	691,85	3739,84	653,7	3735,85	0,29
Turnover	168558,5	1132874	152650,1	1034387	0,21
Firm part of a Group in 2006	0,34	0,48	0,35	0,48	0,38
Firm part of a Group in 2003	0,36	0,48	0,36	0,48	0,37
Firm active in Local or Regional markets in					
2006	0,13	0,34	0,13	0,34	0,48
Firm active in National market in 2006	0,3	0,46	0,3	0,46	0,24
Firm active in European market in 2006	0,54	0,5	0,53	0,5	0,25
Firm active in International market in 2006	0,64	0,48	0,64	0,48	0,41
Firm active in Local or Regional market in 2003	0,14	0,34	0,14	0,34	0,46
Firm active in National market in 2003	0,31	0,46	0,3	0,46	0,2
Firm active in European market in 2003	0,55	0,5	0,55	0,5	0,26
Firm active in International market in 2003	0,65	0,48	0,65	0,48	0,39
Having ISO 9001 certificate in 2006	0,49	0,5	0,47	0,5	0,1
Having ISO 14001certificate in 2006	0,8	0,4	0,79	0,41	0,2
Having ISO 9001 certificate in 2003	0,53	0,5	0,52	0,5	0,08*
Having ISO 14001 certificate in 2003	0,85	0,36	0,84	0,37	0,19
Firm belongs to High-tech. manufacturing	0,03	0,18	0,04	0,19	0,27
Firm belongs to High-medium tech.					
manufacturing	0,09	0,29	0,1	0,29	0,16
Firm belongs to Low-medium tech.	0.1	0.2	0.1	0.21	0.14
manufacturing	0,1	0,3	0,1	0,31	0,14
Firm belongs to Low-tech manufacturing	0,14	0,35	0,15	0,36	0,08*
Firm belongs to Knowledge intensive services	0,18	0,39	0,18	0,38	0,32
Firm belongs to Other services	0,46	0,5	0,44	0,5	0,02*
Disaggregated industry affiliations:	0.00	0.4.4	0.00	0.4.5	0.00
Textile, clothing, accessories	0,02	0,14	0,02	0,15	0,32
Wood	0,01	0,09	0,01	0,09	0,37
Paper and printing	0,03	0,17	0,03	0,17	0,26
Coal, petroleum and nuclear products	0	0,05	0	0,05	0,43
Chemical products	0,03	0,17	0,03	0,17	0,28
Rubber and plastic	0,02	0,15	0,02	0,15	0,31
Non-metallic mineral products	0,02	0,13	0,02	0,13	0,32
Metal and metal products	0,06	0,23	0,06	0,24	0,23
Machinery and equipment	0,03	0,18	0,03	0,18	0,3
Electrical and electronic equipment	0,04	0,19	0,04	0,2	0,28
Transportation equipment	0,02	0,14	0,02	0,15	0,3
Other manufactured products; recycling	0,02	0,13	0,02	0,13	0,34
Electricity, gas and water	0	0,07	0,01	0,07	0,4
Construction	0,07	0,26	0,07	0,26	0,21
Car dealing and repair	0,03	0,16	0,03	0,17	0,27
Wholesale trade	0,08	0,27	0,08	0,27	0,15
Retail trade	0,09	0,29	0,1	0,3	0,13
Hotels and restaurants (horeca)	0,03	0,18	0,04	0,18	0,29
Transportation & communication services	0,09	0,29	0,1	0,29	0,16
Housing and real estate	0,02	0,14	0,02	0,14	0,34
Rental	0,02	0,13	0,02	0,13	0,37
ICT services, R&D	0,04	0,21	0,03	0,18	0,*
Services to firms	0,12	0,32	0,12	0,33	0,09*
Culture, entertainment, sports	0,01	0,1	0,01	0,1	0,45
Financial and insurance services	0,04	0,2	0,04	0,2	0,2

3.1.1. Dependent variable and Treatment group

In this research, we used another, rather rare, approach to distinguish 'early' and 'late' adopters of voluntary EMS (see figure A. for systematic representation). Table 2 below represents frequency of the adopters and non-adopters with respect to observation periods. According to our approach, firms that already adopted voluntary environmental standards and obtained the certificates in 2003 are considered "early adopters" and similarly, firms that did not have the certification in 2003 but obtained the certificate either in 2003 and 2006 are considered "late adopters". Firms that did not have an EMS certificate either in 2003 or 2006 are considered non-adopter. In addition, firms that had an EMS certification in 2003 but not in 2006 are excluded from our sample.

We may assume that a certain amount of time is required before the standards can have significant effect on firms' performance. Empirical evidence (Hart and Ahuja, 1996) assert that there may be a positive relationship between emissions reductions and financial performance but one may have to wait as long as two-years to have any significant financial return. Ambec *et al.* (2013) claims that in many of the previous studies that investigate the determinants of productivity, researchers have often regressed productivity at time 0 on proxies of environmental regulation stringency at time 0 as well, which does not allow time for the innovation process to occur. Introducing a clear distinction between the early and late adopters of voluntary EMS therefore, allows us to test this claim. The database does not distinguish between the standards (i.e. ISO 14001, organic labelling, AB, fair trade, etc.) since they were put together under the same category in the survey. Nevertheless, the literature acknowledges that the ISO 14001 environmental standards and EMAS as the most widely used EMS in the world.

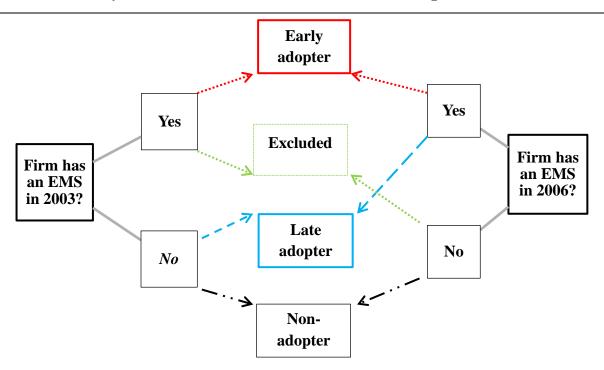


Figure A: Systematic visualisation of the EMS adopters

	Yes	No
# of early adopter	927 (%15.5)	1248 (%21)
# of late adopter	321 (%5.4)	5661 (%94.6)
Total in both year	1248 (%21)	4734 (%79)

Table 2. Frequency of adopters and non-adopters according observation periods

3.1.2. Explanatory Variables and Descriptive Statistics

We are first interested in observing the impact of a firm's size on voluntary EMS adoption decision. We measure the firm size with its actual number of workers. We also included "Size²" variable, which is measured by taking the square of a firm's size. These two variables together could point out the direction of the relationship between a firm's size and their propensity to adopt the standards. In practice, large firms have higher chances of securing additional financial resources for innovative activities and they are more likely to be faced with growing international pressures to clean up their operations and minimize environmental hazards. Most empirical studies found that the probability of implementing environmental standards increases with firm size (e.g. Delmas and Montiel, 2009; Grolleau *et al.* 2007a, b). However, difficulties in reorganization and stabilization of certain routines, red-tape, lack of managerial capabilities, complex, costly and the time-consuming nature of the adoption process should also be accounted for. Firms that are too large may find the adoption of voluntary EMS unnecessary and its adoption process highly demanding, which in return may have a significant impact on a firm's adoption decision.

Being part of a "Group" is another control variable and it would help us to understand whether being part of a group gives an advantage when adopting voluntary EMS. According to Pekovic (2010) and Zyglidopoulos (2002), a group firm may have relatively easier access to a larger amount of financial resources to invest in new practices. However, having a larger amount of financial resources is only one side of the coin. We should also consider the organisational challenges pertinent to large organisations that comprise several sister companies. Accordingly, the relationship among the parent and sister companies and their respective organisational, structural, managerial and other characteristics may play significant role in the adoption decision. Accordingly, if the sister companies are bounded to the parent company's strategic decision-making and general business strategy, being part of a larger group of firms may also constitute certain organisational issues. Hence, firms may or may not have to adopt the standards depending on parent company's business strategy.

We also would like to control for the impact of having other types of quality standards (ISO 9001) and customer-related services (Labelling and Delivery) on a firm's propensity to adopt voluntary EMS. The ISO 9001 quality standards are one of the most widely used management tools in the world today and the isomorphic pressures from different countries and industries makes the adoption of this standard virtually mandatory. By integrating ISO 14001-like standards into the existing ISO 9001 management system, organisations' environmental responsibility may become a component of their product quality. Customer services along with quality standards are the other efficient tools to maintain customer satisfaction, increase efficiency, reduce time and cost etc. In this research, we assume that a firm contractually undertaking to deliver or supply goods or services in a fixed deadline requires a close relation with customers. Such a relation with customers then, may help firms to obtain feedback and precisely identify customers' needs. Corporate reputation can also be affected by information about an organisation's environmental performance (Arora and Gangopadhyay, 1995; Konar and Cohen, 1997; Marshall and Mayer, 1991). Hence, firms that have close links with their customers may also have strong incentives to demonstrate goodwill by adopting and implementing successful environmental management systems (Nishitani, 2009). We, therefore included three dummy variables to investigate whether a firm's existing experience in implementing i) ISO 9001 quality standards, i)) Labelling services and iii) Delivery services has any significant impact on a firm's decision to adopt voluntary EMS.

We have also included a categorical variable indicating whether a firm is active in "National", "European" and "International" markets, where the local and regional markets category is the base category for comparison. By including this variable we aim to observe the respective significance of different business markets on a firm's propensity to adopt voluntary EMS. Increasing local, national and international pressures to reduce carbon foot-prints and increasing consumer awareness, especially within the EU markets, created perceivable pressures on firms to take preventive measures. As a consequence, we may observe an upward shift in firms' behaviour towards the adoption standards, as they may now consider the adoption a key strategy and a significant advantage to capture additional market share or to enter new markets.

The industry classification dummies are also included in our econometric model. These dummies therefore indicates whether a firm is i) High-tech manufacturing, i)) High-medium tech manufacturing, iii) Low-medium tech manufacturing, iv) Low-tech manufacturing and v) Knowledge-intensive firm. These categories are identified following the OECD's

technological levels classification, which is based on the ratio of the total R&D expenditures in total turnover. These dummy variables then, help us to investigate whether a firm's technological complexity has a significant impact on the adoption decision. We assume that firms with high technological complexity are more likely to have higher carbon footprint. Operating at the high technology level generally involves using precise, complex and energyintensive technologies and, processing rare materials with the use of toxic and harmful substances. Therefore, we expect to see a positive relationship between the propensities to adopt the standards and a firm's respective technological complexity.

Table 3. shows the summary statistics on the variables used in the econometric analysis, while table 4. presents the classification of the technological levels by the industries and their respective Nace codes.

Variables used:	Mean	Std. Dev.	Min	Max
Firm size (in numbers)	640.4	3611.2	11	116.989
Firm size ² (in numbers)	1345	32704.7	0.0121	1368643
Firm part of a Group	0.34	0.47	0	1
Having ISO9001certificate	0.51	0.5	0	1
Having Delivery services	0.69	0.46	0	1
Having Labelling services	0.34	0.47	0	1
Firm active in Local market	0.87	0.34	0	1
Firm active in National market	0.71	0.45	0	1
Firm active in European market	0.47	0.5	0	1
Firm active in International market	0.36	0.48	0	1
Firm belongs to High-tech manufacturing	0.025	0.15	0	1
Firm belongs to High-medium tech. manufacturing	0.095	0.29	0	1
Firm belongs to Low-medium tech manufacturing	0.10	0.31	0	1
Firm belongs to Low-tech manufacturing	0.15	0.36	0	1
Firm belongs to Knowledge intensive sectors	0.15	0.36	0	1

Table 3. Summary statistics on the variables used in the econometric analysis

Technology	Nace codes				
classification					
High-technology	24.4 Manufacture of pharmaceuticals, medicinal chemicals and				
	botanical products				
	30 Manufacture of office machinery and computers				
	32 Manufacture of radio, television and, communication equipment				
	and apparatus				
	33 Manufacture of medical, precision and optical instruments, watches				
	and clocks				
	35.3 Manufacture of aircraft and spacecraft				
High-medium-	24 Manufacture of chemicals and chemical product, excluding 24.4				
technology	Manufacture of pharmaceuticals, medicinal chemicals and botanical				
	products				
	29 Manufacture of machinery and equipment n.e.c				
	31 Manufacture of electrical machinery and apparatus n.e.c				
	34 Manufacture of motor vehicles, trailers and semi-trailer				
	35 Manufacture of other transport equipment, excluding 35.1				
	Building and repairing of ships and boats and excluding 35.3				
	Manufacture of aircraft and spacecraft.				
Low-medium-	23 Manufacture of coke, refined petroleum products and nuclear fuel				
technology	25 to 28 Manufacture of rubber and plastic products; basic metals and				
	fabricated metal products; other non-metallic mineral products				
	35.1 Building and repairing of ships and boats.				
Low-technology	15 to 22 Manufacture of food products, beverages and tobacco;				
	textiles and textile products; leather and leather products; wood and				
	wood products; pulp, paper and paper products, publishing and				
	printing				
	36 to 37 Manufacturing n.e.c.				
Knowledge-	61 Water transport				
intensive services	62 Air transport				
(KIS)	64 Post and telecommunications				
	65 to 67 Financial intermediation				
	70 to 74 Real estate, renting and business activities				
	80 Education				
	85 Health and social work				
	92 Recreational, cultural and sporting activities				

 Table 4. Aggregated Nace codes according to OECD's technology classification

 Technology

3.2. Methodology

3.2.1. Propensity Score Matching and Fully Interacted Linear Matching

The aim of this research is twofold. First, we are interested in discovering the intrinsic characteristics of firms that may have an impact on the adoption of environmental standards. Secondly, discovering the causal effect of adoption of environmental standards on the firms performance (the average effect of treatment on the treated – ATT) through Total Value Added (VA). Even though, there are several other performance measures such as labour, material, energy productivity, specific issues and problems call for an appropriate measure of performance. In our approach, we use a performance indicator to measure efficiency in the use of resources approximated with total price of sales minus the total production cost including factor costs (i.e., material cost along with subventions, taxes and subsidies) as the Porter hypothesis postulates. Therefore, we consider it appropriate to use Total Value Added (VA) as the relevant measure of performance proxy. In order to check the robustness of our empirical estimation, we also specified other performance measures i.e. VA / Total turnover. Nevertheless, the estimation results obtained with alternative dependent variables remains robust to our main results. Therefore, we will only discuss the results pertinent to our benchmark model.

In order to estimate the impact of adoption on adopters' VA, we need to compare the average VA of these firms to the average VA that these same firms would have achieved had they not adopted the standards. However, since a firm either adopts the standards or does not, the average VA that firm would have achieved had they not adopt the standards remains an unobserved counterfactual since only one outcome is observed. More precisely, what would have resulted had the firm not been adopted the standards (treated) cannot be observed, which gives a rise to the 'Evaluation Problem' (Vandenberghe and Robin, 2004). The evaluation problem consists in providing unbiased estimates of this average counter-factual through the use of appropriate methods and usually untestable assumptions (Goodman and Sianesi, 2005). Hence, we used the Propensity Score Matching (PSM) method proposed by Rosenbaum and Rubin (1983) and further developed by Heckman *et al.* (1997, 1998). The matching method is a non-parametric alternative to Instrumental Variable (IV) and Heckman type models for estimating a causal effect net of endogeneity bias.

In our research, relying on Ordinary Least Square method (OLS) may produce biased results when considering the fact that there would be some firms adopting the standards, which are not comparable to the firms in the non-adopting sample. In this sense, performing OLS might hide the fact that we are actually comparing incomparable firms by using the linear estimation (Goodman and Sianesi, 2005). The propensity score matching method (PSM) then offers a unique advantage by excluding the firms from the sample that are not comparable to any other firm in the non-adopting sample, leaving only firms from both group that have same features and hence, comparable. Even though, the PSM offers certain advantages over the OLS, we can bring the OLS estimations closer to the PSM estimates by imposing common support before running the OLS and by allowing the impact of treatment (in our case adoption) to vary for each observable variable. This technique is generally referred as Fully Interacted Linear Model (FILM) and it allows the impact of adoption to vary for each observable variable. This method does not provide evidence heterogeneous return (Goodman and Sianesi, 2005). If the method does not provide evidence for heterogeneous return, the estimation results will simply coincide with simple OLS. Hence, both the PSM and FILM methods have been carried out in this research in order to provide comparability between the estimation results.

4. Econometric Results

4.1. Estimations obtained with PSM

The first step of our empirical analysis is to estimate the propensity score i.e. the probability of adopting environmental standards on a voluntary basis conditional on observable control variables that may affect this probability as well as the response variable. Table 5, below, displays the Probit estimation results obtained by using the PSM as well as the conditional marginal effects obtained after the PSM. The Probit estimation results are generally hard to interpret alone, therefore we displayed the marginal affects after the PSM so that we can interpret the estimation results better.

The estimation results obtained by the PSM imply that a firm's size is one of the most significant firm characteristics and it facilitates the adoption of voluntary EMS. The existing literature assert that the probability of implementing environmental standards increases with the firm size (Delmas and Montiel, 2009; Grolleau *et al.*, 2007a, 2007b) and larger organizations are more likely to possess the knowledge-based skills that may be critical factors in their capacity to commit to environmental initiatives (Hart and Ahuja, 1996). Our results, on the one hand, support these claims and show that medium and large size firms are more likely to adopt voluntary EMS. On the other hand, they point out that propensity to adopt the standards may decrease as the firm size reaches to a threshold. We can note that

size-square (size²) variable has a negative sign, which implies that the relationship between adoption of these standards and firm size can be represented by an inverted U-shape. The existence of such a relationship may indicate that some big firms may suffer from certain implementation problems (i.e., red tape and stabilization of certain routines) hence they may shy away from the adoption. Moreover, the marginal effects obtained after the PSM imply the same and show that the marginal effect of a firm's size on the adoption decision of voluntary EMS is significantly higher for the early adopters than the late adopters.

The estimation results also reveal an important insight about the relationship between voluntary EMS adoption and group ownership. Like medium and large size firms, group firms or sister companies tend to have significant advantages over individual firms to access financial resources, human capital and technical capabilities for technology transfer and knowledge outsourcing. The estimation results show that these potential advantages alone might not provide sufficient incentives to adopt an EMS. The results obtained by the PSM show that being a sister company tends to lower a firm's propensity to adopt the standards. We can also note that the marginal impact of being a sister company on the adoption decision is significant and negative, both for the early and late adopters. We can relate this negative relationship to the implementation cost inherent in the adoption of voluntary EMS and to the role of a parent company's general business strategy. We presume that individual firms are more likely to be (relatively) smaller than group firms in terms of firm size and, as we noted earlier, some big firms may suffer from certain implementation problems that can hinder the adoption of voluntary EMS. When this is the case, it is reasonable to believe that individual firms rather than group firms are more likely to adopt and implement the standards. Another explanation can also be made by assuming that that a parent company's organisational and technical capabilities, general business strategy, environmental consciousness and long-term strategic thinking may have a direct and significant impact on sister company's' decision making. If the sister company has enough managerial space to plan and execute strategic actions then they may easily choose to adopt the standards. But if the relationship between the parent company and its sister companies is rather prescriptive, sister companies may leave little or no space for individual decision making and will have to bound the strategic decisions (adopting or not adopting the standards) made by the parent company. Our estimation results show that the second scenario is more likely to occur and left us to conclude that individual firms, rather than firms that are part of a larger group, are more likely to adopt the standards.

Furthermore, Probit estimation results obtained by the PSM indicate that having other type of quality standards (ISO 9001) and customer related services (Labelling and Delivery) may significantly increase a firm's propensity to adopt voluntary EMS. On the one hand, we note that the early adopters tend to have both quality standards and customer related services and, they are more likely to use their existing experience in implementing these standards.

On the other hand, we observe that the late adopters of voluntary EMS tend to have only the ISO 9001 quality standard. This may imply that having already one type of quality standards from a standards family (i.e. ISO 9001 standards) may facilitate the adoption of another type of standards from the same family (i.e. ISO 14001). Many industries and countries often adopt ISO 9001-type standards as requirements for doing business, which makes their adoption virtually mandatory. The isomorphic pressures from industries and countries might be a significant factor for the adoption of such standards but we should also consider the role of a firm's strategic thinking and general marketing strategy to adopt them. If efficiently used, a green marketing strategy may help an adopting firm to capture additional market share and it may even grant access to new markets. Therefore, by looking at the estimation results, it is reasonable to believe that firms that have existing experience in implementing such standards and services tend to consolidate their reputation (or even to improve it) by implementing other types of quality and management standards i.e. ISO 14001 and EMAS - type environmental standards. Moreover, we can also note that the marginal impact of having these types of quality standards and customer related services on the adoption of voluntary EMS is significantly higher for the early adopters than the late adopters. Especially, the marginal impact of having ISO 9001 quality standards is significantly higher for the early adopters. This result is not a surprise and it attests that especially the early adopters of voluntary EMS tend to rely on their existing experience in implementing other type of quality standards.

Besides a firm's structural characteristics and organisational capabilities, a firm's business market might be another important factor affecting its decision to adopt and/or when to adopt voluntary EMS. Unlike ISO 9001 type quality standards, the diffusion of voluntary EMSs are still in the catching-up phase as their adoption is solely voluntary (ISO 9000 family is first introduced in 1987 while ISO 14001 EMS was first introduced in 1996). The ISO 14001-type standards are especially flexible (more flexible than EMAS) and applicable to wide variety of organisations, in size and in occupation, such as companies, branch offices, construction sites, administration centres etc. However, we expect that firms that are active in a relatively greater context i.e. International or European markets, are more likely to adopt

the standards as the adoption decision might be influenced by the isomorphic pressures from industries and countries. The Probit estimation results obtained by the PSM imply that being active in national markets significantly decreases a firm's propensity to adopt the standards. On the contrary, we observe that being active in the European markets tend to be a significant factor that increases a firm's propensity to adopt them. Furthermore, the marginal effects obtained after the PSM show that the effect of being active in the European markets is significant and positive only for the early adopters. This has led us to conclude that the European markets may perceive the environmental standards adoption as a compulsory action as a result of the isomorphic pressures within industry hence, imposing the adoption of the standards for certain firms, i.e. early adopters.

Finally, the Probit estimation results obtained by the PSM suggest that a firm's technological complexity may have significant impact on EMSs adoption. We observe that 'high-medium technology manufacturing' firms are more likely to adopt the standards earlier than the firms within other technological complexity levels. On the contrary, 'low-tech manufacturing' firms are more likely to adopt the standards relatively later. In addition, the marginal effects obtained after estimating the PSM indicate that being a high-medium technology manufacturing firm has a significant and positive impact on the early adoption decision. This result, at the best, may imply the existence of mimicking behaviour performed by less hazardous and technologically less complex low-technology manufacturing companies. We expect that the marginal benefit that the low technology manufacturing firms that would reap from adopting voluntary EMS tends to be lower when compared to technologically more complex and more hazardous industries i.e. high-medium technology manufacturing. Hence, firms within low tech. manufacturing industries might be reluctant to adopt the standards and might choose the 'wait and see' strategy to observe whether the standards adoption provide any return, both environment and business performance wise, before taking the adoption decision. When we consider the fact that the adoption of EMS is solely voluntary and depend on a firm's general business strategy, our results make more sense. The results also provide important insights about the adoption behaviour of 'knowledge-intensive services' sectors. Accordingly, the results suggest that firms within 'knowledge-intensive services' sectors are less likely to adopt the standards. The marginal effect of being active in knowledge-intensive services sectors on the adoption decision is also significant and negative. Given that the knowledge-intensive sectors tend to have less carbon foot-print compared to the manufacturing industries in general, we have reasons to believe

that the adoption of the voluntary EMS may depend on a firm's carbon foot-print level and the technical complexity of their production process.

Dep. Var.: Adoption of Voluntary	Matching Met		Marginal effects after matching		
EMS	Func		E males a domésia	Tertereter	
Constant	Early adopter	Late adopter	<i>Early adopter</i> -0.29***	Late adopter	
Constant		-2.06***		-0.16***	
	(0.11)	(0.152)	(0.018)	(0.013)	
Control Variables:					
- Firm size	0.087***	0.071**	0.00015***	0.000054**	
	(0.013)	(0.023)	(0.00024)	(0.000017)	
- Size ²	-0.00075***	-0.0009*	-0.00013***	-0.0008**	
	(0.00014)	(0.0005)	(0.000026)	(0.00004)	
- Firm part of a group	-0.241***	-0.145*	-0.043***	-0.013**	
	(0.056)	(0.078)	(0.009)	(0.006)	
Having other standards:					
- ISO 19001	1.02***	0.72***	0.178***	0.058***	
	(0.059)	(0.088)	(0.009)	(0.006)	
- Labelling	0.19***	0.085	0.035***	0.005	
-	(0.049)	(0.066)	(0.008)	(0.005)	
- Delivery	0.12***	0.097	0.021**	0.007	
	(0.059)	(0.083)	(0.01)	(0.006)	
<u>Active Business Market:</u>					
- National	-0.137*	-0.187*	-0.023**	-0.014**	
	(0.066)	(0.092)	(0.011)	(0.007)	
- International	0.094	0.081	0.015	0.009	
	(0.066)	(0.089)	(0.011)	(0.006)	
- European	0.142*	0.09	0.024**	0.009	
-	(0.071)	(0.1)	(0.012)	(0.007)	
Industry affiliations:					
- High-tech. manufacturing	0.21*	0.091	0.035	0.003	
c c	(0.122)	(0.167)	(0.021)	(0.013)	
-High-medium manufacturing	0.4***	-0.0008	0.071***	-0.0022	
	(0.078)	(0.111)	(0.013)	(0.008)	
- Low-medium manufacturing	0.058	0.11	0.01	0.0051	
	(0.078)	(0.104)	(0.013)	(0.008)	
- Low-tech manufacturing	0.098	0.16*	0.017	0.01	
	(0.072)	(0.095)	(0.012)	(0.007)	
- Knowledge services	-0.33***	-0.235**	-0.06***	-0.022**	
	(0.079)	(0.12)	(0.013)	(0.009)	
N	5455	4867	5582	5584	
Pseudo R ²	0.2073	0.094	0.2094	0.0975	
LR chi ² (13)	990.42	194.2	896.63	247.25	
Likelihood ratio test:	-1893.54	-931.83	-1913.75	-1055.03	

Table 5. Estimation Results

-* significant at 10%, ** significant at 5%, *** significant at 1%

-Standards errors are below each coefficient and in parentheses

4.2. Average Treatment Effect on the Treated (ATT)

In order to provide robust empirical evidence to support/contradict the PH, the second step of our empirical analysis involves in estimating the average treatment effects on the treated (i.e. the impact of environmental standards adoption on Total Value Added). We calculated the ATT given the propensity scores calculated as predictions of the Probit models estimated. The dependent variable for the ATT is log of total value added indicated by VA (total price of sales minus total production costs) as the PH postulates. Table 6 below reports the results of the ATT estimations. In order to provide a robustness check of our primary estimation method, namely the PSM, table 6 also presents the ATT results obtained by using the FILM method.

Table 6. Average Treatment Effect on the Treated (ATT) obtained from PSM and FILM

Dep. Var.: Total Value Added	Propensity Sc (PS	0	Fully Interacted Linear Matching (FILM)		
	Early adopter Late adopter		Early adopter	Late adopter	
ATT	0.49***	0.51***	0.38***	0.28***	
	(0.07)	(0.1)	(0.076)	(0.071)	
On support (untreated)	4588	4602	9995	9995	
On support (treated)	867	264	1173	1172	
Off support (treated)	0	1	0	1	
Ν	5455	4867	11168	11168	

-* significant at 10%, ** significant at 5%, *** significant at 1%

-Standards errors are below each coefficient and in parentheses

The ATT results calculated by the FILM estimation indicate that adoption of voluntary EMS has a significant impact on a firm's Total Value Added (VA). The ATT's coefficient is significant and has positive sign while the impact of the adoption on firms' total value added is slightly higher for the early adopters.

The ATT calculated by the PSM estimation is also in line with the results obtained by the FILM estimation but with a slight difference in the coefficients. According to our benchmark approach, the adoption of voluntary EMS may reveal slightly higher VA for the late adopters. However, the difference takes place in the last decimal of the calculated coefficient and it might be attained to computational differences between these two approaches. Therefore, the estimation results may point out that adoption voluntary EMS may indeed increase a firm's business performance through VA, regardless of the adoption year. These results are

especially important for establishing the validity of the narrow version of the Porter hypothesis and they provide us a preliminary proof for "win-win" situation. Even though, we could not find significant difference between the early and late adopters of voluntary EMS in terms of financial performance that they can reap. The voluntary EMS can be efficient tools to decrease a firm's environmental liabilities while enabling them to reap competitive advantage. Even though increasing firms' business performance and competitiveness is not the primary aim of ISO 14001 type standards, lesser environmental liability, reduced input costs, environmentally friendly image and, increased employee, consumer, investor, shareholder and insurer trust might enable firms to increase their VA. We can conclude this section by asserting that the "win-win" opportunity indeed exists even without relying on the strict environmental regulations. However, in order to reach a certain level of global environmental efficiency the number of adopting firms should be augmented rapidly. In order to do so, the policy makers should efficiently use their communication channels to explain the environmental and economic impact of voluntary EMS and, assist them through their transition to become less environmentally hazardous and more profitable organisations.

4.3. Robustness Test of the Matching Method

The next step of the matching method is to check whether or not the "balancing property" is achieved. Accordingly, we will control whether the propensity score effectively balances characteristics between the treatment and comparison group. If the balancing property is achieved, we are confident that the bias associated with observable characteristics is reduced. Table 7 presents the balancing tests that are performed for the two groups, "adopter" versus "non-adopter". In addition to reporting the mean values of the groups and the t-statistics, we also report the standardised difference, that is, the size of the difference in means of a conditioning variable (between the treatment and comparison units), scaled by (or as a percentage of) the square root of the average of their sample variances.

Adopter vs. Non- adopter	Before matching			After matching			
Variables	Treated (adopter)	Control (non- adopter)	t-test	Treated	Control	t-test	% reduction in bias
ISO 9001	0.9	0.42	27.8***	0.9	0.89	0.68	97.9
Group	1.16	1.39	-12.41***	1.16	1.17	0.49	96
Delivery	0.84	0.65	11.05***	0.84	0.84	0.01	9.99
Size	1590.1	464.1	8.5***	1590.1	1254.1	1.14	70
Size ²	0.005	0.0067	3.45***	0.0049	0.0031	0.7	58.2
Local	0.78	0.88	-7.74***	0.78	0.76	0.7	85.8
National	0.82	0.69	7.66***	0.82	0.81	0.52	92.5
European	0.67	0.43	13.66***	0.67	0.67	0.12	99
International	0.56	0.32	13.98***	0.56	0.57	0.34	96.7
High-tech man & High-medium man.	0.29	0.096	16.38***	0.29	0.32	1.03	88.5
Low-medium man. & Low-tech man.	0.32	0.26	3.41***	0.32	0.31	0.55	78.2
Knowledge services	0.08	0.18	-7.07***	0.08	0.086	-0.21	97

Table 7. Summary statistics on the choice of variables before and after matching

- * Significant at 10%, ** significant at 5%, *** significant at 1%

Before-matching results show that there is clearly an imbalance between the treated and the control groups. All t-statistics are highly significant indicating that the null hypothesis of joint equality of means in the matched sample is rejected. By contrast, in the after-matching results, we clearly see that the differences are no longer statistically significant therefore, we are confident that our matching significantly reduced bias.

5. Conclusion

Since it was first introduced the Porter Hypothesis (Porter and van der Linde, 1991, 1995) has been subject to various empirical investigations to prove or disprove its validity. Many researchers have generally disaggregated the PH into its component parts ("weak", "narrow" and "strong" versions) in order to test the theory and collect empirical evidence (see Jaffe and Palmer, 1997). According to Ambec *et al.* (2013) empirical investigation of the PH generally fall into three categories: those testing the "weak" version of the hypothesis, those testing a "strong" version focused on firm-level performance, and those testing a "strong" version focused on country-level competitiveness. This research distinguishes itself from the existing empirical investigation in several ways. First of all, it tries to identify the drivers of

voluntary EMS adoption from firms' organisational capabilities perspective. Accordingly, we try to find out whether a firm's structural characteristics and organisational capabilities facilitate "early and/or late" adoption of voluntary EMSs. In the second step, we investigate the validity of the 'narrow' version of the PH, which argues that flexible regulatory policies give firms greater incentives to innovate and thus are better than prescriptive forms of regulation (Jaffe and Palmer, 1997). Following this claim the second step of our analyses consists in examining the impact of voluntary EMS adoption (i.e., ISO 14001 and Eco-Management and Audit Scheme (EMAS) - type standards) on early and late adopters' business performance. In our approach, we use a performance indicator to measure efficiency in the use of resources - total price of sales minus the total production cost that is Total Value Added (VA) - as the PH postulates. The last contribution of this paper has to do with classification of the adopters of voluntary EMS which is one of the most significant addedvalues of this research. We explicitly introduced a distinction between the 'early' and 'late' adopters of the standards depending on the adoption year. By doing so, we can investigate potential differences (in terms of organisational capabilities and structural characteristics and, return on VA) between the early and late adopters of voluntary EMS. To our knowledge, this type of distinction is non-existent in the eco-innovation literature and therefore, it is a significant contribution to existing the eco-innovation literature.

The results obtained in the first step of our empirical analyses show that a firm's structural characteristics organisational capabilities, business market and technological complexity are some of the factors that facilitate the adoption of voluntary EMS. Moreover, we also showed that the respective marginal effect of these factors may differ between the early and late adopters of the EMS. More precisely, we observe that medium and big size firms are more likely to adopt the standards but the results also suggest that some big firms may suffer from certain implementation problems due to i.e. red tape and stabilization of certain routines. Having established ISO 9001 quality standards and customer related services (i.e. labelling and delivery services) are the other significant factors that facilitate the adoption of voluntary EMS. Moreover, the results indicate that having existing experience in implementing such standards and services can greatly affect a firm's decision to adopt the standards in a relatively earlier stage. In the same vein, we assert that being active in European markets can be a significant factor for voluntary EMS adoption. We associate this result with the isomorphic pressures from industries and countries that makes the adoption of EMS virtually a mandatory action to survive within highly competitive and environmentally conscious European markets. The marginal impact of being active in European markets is

statistically significant and positive for the early adopters, which supports our deduction to a great extent. Finally, the results stress the importance of a firm's technological complexity as a driver to adopt voluntary EMS. Accordingly, high-medium technology manufacturing firms (R&D / Total turnover ratio is between 5% and 3%) are more likely to adopt the standards earlier, while Low technology firms (R&D / Total turnover ratio is between 0.09% and 0%) are more likely to adopt the standards in a later stage.

In the second step, we showed that the adoption of voluntary EMS may have a significant impact on adopters' VA. Furthermore, we did not observe any significant difference between the early and late adopters of the standards with regards to return on VA. We tried to ensure the reliability of the results by performing robustness check with the use of alternative estimation method (namely, Fully Interacted Linear Model) and highlighted that the empirical results obtained with this approach are robust to the results obtained by Propensity Score Matching method. In short, we concluded this research paper by asserting that even though government intervention through environmental regulations are still the most important factor leading firms to lower their hazardous impact on environment, the voluntary environmental standards may be a good complement for market-based instruments (such as tradable permits, emission taxes, subsidies etc.) and, command and control mechanisms to increase firms' economic performance while becoming greener. Hence, the policy setting institutions should use this opportunity to communicate and explain positive economic and social gains that could be reaped from the adoption of environmental standards. By doing so, a domino effect could be launched and the number of adopters could be multiplied while leading to less environmentally hazardous and more profitable organisations, enabling them to enjoy the "win-win" opportunity as the narrow version of the Porter Hypothesis postulates.

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Chapter 3

Which firm characteristics matter? Identifying the drivers for eco-innovation

Table of Contents

1. Introduction	105
2. Review of the Literature	106
2.1. Defining eco-innovation	106
2.2. Drivers of eco-innovation	107
3. Data and Methodology	108
3.1. Data	108
3.2. Specification and estimation of the model	110
3.3. Innovators and environmental innovators	112
3.4. Explanatory variables	115
3.5. Choice of an exclusion variable	119
4. Empirical Results	120
4.1. Estimations conducted on the whole sample	120
4.2. Estimations conducted by group of countries	124
5. Conclusion	129

Which firm characteristics matter? Identifying the drivers for eco-innovation

Abstract

Up until now, researchers have emphasized the importance of conventional technology-push, market-pull drivers and regulatory framework for successful eco-innovation. The importance of firms' structural characteristics, of their openness and of formal cooperation with external partners have been comparatively disregarded. The aim of this research is to examine the importance of firms' structural characteristics and to understand the role these characteristics play in the realization of eco-innovation. We estimate a Generalized Tobit model on micro-anonymised data from the 4th wave of the Community Innovation Survey (CIS 4). We find that eco-innovative firms tend to be large, autonomous firms that are open to use of internal and external sources of information. Moreover, we find evidence of a possible substitution effect between formal cooperation with external partners and basic R&D efforts. Our results are robust to country-specific analyses.

Key words: Eco-innovation, Generalized Tobit, CIS 4, Firm characteristics, Environmental consequences.

JEL codes: **D2**- Production and Organizations, **L2**- Firm Objectives, Organization, and Behaviour, **Q5**-Environmental Economics, **O3**- Technological Change

1. Introduction

Over the last decade, there has been a rise in the introduction and development of environmentally benign innovation (a.k.a. Eco-innovation). Global environmental problems and fierce competition among firms may have played a significant role in the emergence of this type of innovation. A consequence of increasing global environmental consciousness and of concomitant firms' willingness to reduce environmental hazards related to their activities, eco-innovation is presented as one of the most promising tools to solve the world's sustainability issues. Eco-innovation may help firms overcome environmental challenges by modifying, re-designing, creating and using alternative products, processes and organizational methods. Despite its recent emergence (see Fussler and James, 1996); the concept of eco-innovation has gained content and scope in a relatively short period of time. The complexity of the eco-innovation process suggests that, besides conventional innovation drivers, there should be certain firm specific capabilities and competencies that could be complements or even substitutes to conventional innovation drivers. These competencies and capabilities might be related to experience, knowledge base, managerial and technological capabilities, degree of openness, cooperation and collaboration with internal and external agents, etc.

These capabilities may be of even greater importance for eco-innovative firms than they are for conventional innovators since eco-innovation incorporates environmental concerns in the usual innovation-related challenges. However, the eco-innovation literature is still in need of extensive empirical research to identify these capabilities. Our research aims at filling this gap by identifying those firm characteristics that are inherent in the realization of innovations with positive environmental consequences.

The policy implications of the proposed research are twofold. First, identifying the aforementioned characteristics may give policy makers a better apprehension of the drivers and other facilitating factors of eco-innovation. This improved comprehension may in turn help policy makers design eco-innovation policies that are more effective and easier to implement. Second, a better grasp of the above-mentioned capabilities and competencies might give firms themselves useful guidelines for the successful implementation of eco-innovation.

Our empirical analysis exploits data from the 4th Community innovation Survey (CIS 4) that comprises 104,943 firms observed across 15 EU member States (or associated countries) over the years 2002 to 2004. The target population is made of firms of at least 10 employees.

By conducting firm-specific analyses with such a large dataset, we aimed at finding evidence at the micro level in a macro context. Moreover, besides conducting an empirical analysis on the whole dataset, we also conducted country specific analyses in order to highlight the potential differences among the European countries with regard to firm characteristics that are inherent in the realization of innovations with positive environmental consequences. The remainder of this paper is organized as follows. Section 2 reviews the literature related to eco-innovation and state our research objective. Section 3 presents the data, estimation strategy and choice of variables. Section 4 reports the estimation results. Concluding remarks are given in a final section.

2. Review of the Literature

2.1. Defining eco-innovation

The innovation literature defines eco-innovation as "Assimilation or exploitation of a product, production process, service or management or business methods that it is novel to the firm or user and which results, throughout its lifecycle, in a reduction of environmental risk, pollution and other negative impact of resource use (including energy use) compared to the relevant alternatives" (Kemp and Pearson, 2007). As one can clearly see, the definition is quite general and rather minimalist. It embraces every environmentally improved product, production process, services, management or business method as potential eco-innovation, even if the environmental concerns are not the corner-stone of the realized innovation. Accordingly, unintentionally reduced environmental hazard could be sufficient enough to characterize an innovation as an eco-innovation. However, this aspect of eco-innovation raises some identification and measurement problems, all the more since every different definition tends to focus on different aspects of eco-innovation. For instance, the Japanese government's Industrial Science Technology Policy Committee defines eco-innovation as "a new field of techno-social innovations [that] focuses less on products' functions and more on [the] environment and people" (adapted from OECD, 2009). This definition considers ecoinnovation as a central component in order to overcome societal challenges on the way to sustainable development. However, other, more technical aspects can be emphasized. For instance, a product innovation in one firm (e.g. product innovation in machinery, packaging etc.) could be considered as a process eco-innovation in another firm (e.g. Schmookler, 1966; Scherer 1982a, b). Such complications explain why eco-innovation has to be defined in such a general way.

Whatever its definition, its promise in terms of competitiveness and sustainability have attracted the general attentions of both firms and countries towards eco-innovation. This has led to an increase in the number of eco-innovating firms and to significant market development for eco-innovative technologies. For instance, Stern (2006) claims that markets for low-carbon energy products are likely to be worth at least 500 billion dollars per year by 2050. Other prominent example of this development is the EU's Lisbon agenda, which aims at significant targets for sustainable growth such as: (1) reducing greenhouse gas emissions by 20% (compared to 1990 levels) by 2020, (2) increasing the share of renewable energies in final energy consumption to 20%, and (3) moving towards a 20% increase in energy efficiency (European Commission). The targets set by the EU do not solely aim at being environmentally benign, but also at increasing the competitiveness of the Union by reaching certain levels of efficiency in input use and in productivity. As Porter (1991) and, Porter and van der Linde (1995) asserted pollution is generally associated with unnecessary, inefficient or incomplete use of resources, or with loss of energy potential. Hence, reducing pollution would not only make the world a better place, but would also improve resource productivity, which could trigger drastic competitive advantage for successful innovators over noninnovators.

2.2. The drivers of eco-innovation

The innovation literature highlights the significance of technology-push and market-pull factors as the main drivers of successful conventional innovation. According to Pavitt (1984), the technology push factors are essential in the first stages of a product's lifecycle, whereas the market related factors are significant in the diffusion phase. For successful eco-innovation however, regulatory push/pull effect can also be considered of the utmost importance. As stated in van Leeuwen and Mohnen (2013), environmental regulation and public funding of R&D are the first impetus to have more green technologies developed by individual firms. Despite the on-going controversies on this issue, the acceptance of environmental regulation as one of the main drivers of eco-innovation took some time and considerable efforts. Porter and van der Linde were the pioneers of this sometimes heated discussion and naturally have been heavily criticized by many scholars (e.g. Palmer, Portney and Oates, 1995).

The Porter hypothesis (Porter, 1991; Porter and van der Linde, 1995) assert that stricter environmental regulation would lead firms to eco-innovate and the economic gains that could be reaped from this investment would offset (partially or fully) the cost of the regulation itself, leading to a "win-win" situation. Later on, Rennings (2000) consolidated the importance of environmental regulation when he introduced the notion of "double externality" problem. According to him the eco-innovation comprises two distinct externalities, unlike other innovations. These externalities are i) the usual knowledge externalities in the research and innovation phases and ii) the environmental externalities in the adoption and diffusion phases (Oltra, 2008). More precisely, the beneficial environmental impacts of eco-innovations make their diffusion always socially desirable (Belin *et al.* 2011). However, these positive external effects bring some market failures that reduce the incentives for firms to invest in eco-innovation (Rennings, 2000), which makes the environmental regulations one of the most crucial factors affecting the introduction and development of eco-innovation.

Nevertheless, the conventional drivers of innovation and environmental regulation are not the only factors likely to affect a firm's involvement in the eco-innovative process. As the eco-innovative process becomes more complex, the role of (and need for) firm-specific capabilities and competencies is getting increasingly important. The acquisition of these capabilities and competencies often comes to rely on cooperation. The time where innovation (and eco-innovation) was seen only a "one-man work" is long gone. Cooperation with internal and external agents has become a strategic move to attain success in innovative activities. This seems to be even more the case when eco-innovation is concerned. Thus, Hemmelskamp (1999) found that eco-innovative companies tend to have lower R&D intensity and tend to countervail this relatively low R&D effort with use of external sources information. Additionally, eco-innovation is a highly complex activity which involves increasing interdependencies and relies on external knowledge and information (Andersen, 1999, 2002; Foxon and Andersen, 2009). The complexity of the eco-innovation process suggests that, besides the conventional innovation drivers, there should be certain firmspecific capabilities and competencies that could act as complements or even substitutes to the impulsion of technology, the pull of the market and the environmental regulation. The aim of this research is to identify those capabilities and competencies that are liable to assist firms in the eco-innovative process.

3. Data and Methodology

3.1. Data

This research exploits micro-ananoymised data drawn from the 4th wave of the Community Innovation Survey (CIS 4). The CIS is a survey that aims at providing comparable firm-level data on innovation activities in the European countries. It is coordinated throughout Europe by Eurostat in cooperation with the statistical offices of the Member States. It relies on a harmonized questionnaire based on the Oslo Manual (1997), which provides invaluable information on innovation activities at the firm level (see Eurostat, 2005). The CIS data is available as micro (firm-level) data for a given country (from this country's national statistical office), but also as micro-anonymised data for several EU Member States (from Eurostat).

The present research uses the micro-anonymised CIS 4 data, available for 15 EU Member States (or associated countries). Micro-anonymised CIS is firm-level data which has been made anonymous by adding random "error terms" to the raw firm-level data. This method makes it almost impossible to break the anonymity of individual firms, while allowing researchers to conduct econometric analyses (which will take out the added noise). This process is explained in details in Mohnen, Mairesse and Dagenais (2006). Experiments on specific countries have shown that econometric analyses conducted on the micro-anonymised data yield results that are not significantly different from those of analyses conducted on the corresponding firm-level data (Mohnen, Mairesse, and Dagenais, 2006).

The CIS 4 survey contains information about the main economic activity of each sampled firm, location, number of employees, total turnover, expenditures in intramural and extramural R&D, innovation output, innovation collaboration and knowledge sources for innovation. Compared to the traditional R&D and patent data, the innovation output indicators used in the CIS surveys have the advantage of measuring innovation directly, which means they capture the innovations that were introduced to the market and their relative weight for the innovators total sales (Kleinknecht *et al.* 2002).

The 4th wave of the CIS survey does not contain information which directly pertains to eco-innovation. However, there are some variables that can help researchers identify the environmental impact of product and/or process innovations introduced during the observation period (2002 to 2004). The environmental outcome of the product and/or process innovations can thus be measured by (*i*) "reduced materials and energy per unit output" and (*ii*) "reduced environmental impact or improved health and safety". In what follows, we will use these two measures as proxies for the environmental outcomes of innovation activities. Hence, using this dataset may allow us to compare and contrast the specific characteristics of eco-innovation compared to non-environmental (or "conventional") innovations. The dataset

includes 104,943 firms observed across 15 different countries²⁵ over the years 2002 to 2004. This sample is representative of the population of firms with 10 employees or more based in these 15 countries. As such, it is one of the most extensive datasets on innovation available, which will let us conduct our empirical investigation beyond the confines of a specific industry or geographical area.

3.2. Specification and Estimation of the model

The aim of this research is to identify firms' characteristics and capabilities that have significant impact on firms' eco-innovation intensity. To do so, we deemed it relevant to estimate an innovation production function following the "innovativity" approach proposed by Mairesse and Mohnen (2002). The main departure from this approach is that our measure of innovation output will actually be a measure of eco-innovation. We chose to estimate a Generalized Tobit model (i.e., a variant of the Heckman (1979)'s selection model estimated by Maximum Likelihood) in order to reduce the selection bias inherent in the use of CIS data. This bias arises due to the presence of "filtering" questions in the questionnaire. These questions are designed to characterize firms as "innovating" or "non-innovating", and only the former have to fill in the full questionnaire (see Mohnen, Mairesse and Dagenais (2006) for further details). This attribute of CIS surveys can endanger the reliability of the empirical analysis if one does not take into account the selection problem.

Our dependent variables of interest (the two proxies mentioned in the previous section) are categorical variables coded from 0 to 3 (a higher number denoting a larger positive effect of the innovation on the environment). However, continuous variables would be better suited to the intensity equation of the Generalized Tobit model. To address this issue, we applied a factor analysis on our two categorical eco-innovation proxies. Both loaded to a single continuous factor with an eigenvalue close to 1, asserting that there is a single underlying variable that accounts for most (91%) of the total variance with a 0.67 factor loading. We use the scores of this underlying variable (first factor) to build a continuous measure of eco-innovation intensity, which we then used as the intensity variable in our Generalized Tobit model. In order to make interpretation easier, we have normalized this variable so that it has a standard normal distribution with zero mean and a standard deviation of 1.

²⁵ Belgium, Bulgaria, Czech Republic, Germany, Estonia, Spain, Greece, Hungary, Lithuania, Latvia, Norway, Portugal, Romania, Slovenia and Slovakia.

The Generalized Tobit model is a two-equation model, both equations being estimated simultaneously by Maximum Likelihood. In our application, the first equation (or selection equation) models the probability of being a product and/or process innovator. This equation is specified as a Probit model, as is usually done in the literature. The second equation (or intensity equation) models the intensity of eco-innovation activities for firms that innovate. The dependent variable used in this equation is the normalized continuous index built from the factor analysis, as explained above. The intensity equation is central to our empirical analysis, as it is the equation which allows us to identify the determinants of eco-innovation.

$$Z_i^* = W_i + u_i \tag{1}$$

$$Z_{i} = \begin{cases} 1 & if \ Z_{i}^{*} > \mathbf{0} \\ 0 & if \ Z_{i}^{*} \leq \mathbf{0} \end{cases}$$
(2)
$$Y_{i} = \begin{cases} X_{i}\beta + \epsilon_{i} & if \ Z_{i}^{*} > \mathbf{0} \\ - & if \ Z_{i}^{*} \leq \mathbf{0} \end{cases}$$
(3)
$$u_{i} \sim \mathrm{N}(0,1)$$
$$\epsilon_{i} \sim \mathrm{N}(0,\sigma^{\dagger}\mathbf{2})$$
$$\mathrm{Corr}(u_{i}i, \epsilon_{i}i) = \rho$$

In Equation (1), Z_i^* represents the unobserved innovation decision variable which measures the propensity to innovate, W_i represents the vector of explanatory variables determining the innovation decision and the γ 's are the coefficients to be estimated for each explanatory variable. If a firm's propensity to innovate is bigger than zero, $(Z_i^* > 0)$, then the firm is considered as an innovator (Equation 2). In Equation (3), Y_i is the normalized variable measuring the intensity of eco-innovation. X_i is the vector of explanatory variables for this equation, and β denotes the vector of associated parameters to be estimated. In the model, u_i and ϵ_i are standard normal random error terms with correlation coefficient ρ . Taken together, Equation (2) and Equation (3) defined the Generalized Tobit model that is estimated by Maximum Likelihood. The model is estimated first on all firms (i.e., the 104,943 observations taken from all of the 15 countries covered in the dataset). In this second step, we distinguish Western European countries (Belgium, Germany, and Norway²⁶), Eastern

²⁶ Although Norway is not part of the EU, it is a Northern European country and part of the larger European Economic Area. We thus find it relevant to include Norway alongside Belgium and Germany for the purpose of our analysis.

European countries (Bulgaria, Czech Republic, Hungary, Romania, Slovakia and Slovenia), Mediterranean countries (Spain, Greece, Portugal), and Baltic countries (Estonia, Latvia, Lithuania), and estimate the model for these country groups.

3.3. Innovators and Environmental innovators

Our empirical analysis hinges on the distinction between non innovators, innovators, and environmental innovators (or eco-innovators). The selection equation of our Generalized Tobit model relies on a binary variable indicating whether a firm introduced an innovation or not during the period of observation (i.e. from 2002 to 2004). The dependent variable used in the intensity equation of the model is based on two categorical variables indicating whether the innovation had "low", "medium" or "high" environmental effects or "no" environmental effect at all.

To take a first glance at the data, we can recode these two variables as binary (with 1 corresponding to an environmental effect and 0 corresponding to none) and define an ecoinnovator as a firm with a value of 1 in either of these two variables (or both), and a non-ecoinnovator as a firm with a value of 0 in both variables. Table 1 gives the distribution of innovators and eco-innovators across industries for the 15 countries covered in the survey. Table 1 displays a slight difference in the distribution of eco-innovators across the services and manufacturing industries. The number of eco-innovative firms is higher in the manufacturing industries than in the services sectors. As regard services, we can note that that the highest concentration of eco-innovators is within "Computer and Related Activities" (27.8%), while the lowest concentration lies within "Real Estate Activities" (9.3%). We can also detect that "R&D and Other Business Activities" and "Wholesale Trade" have the highest number of eco-innovators.

The picture is a bit different within manufacturing industries. The average number of eco-innovating firms is higher among manufacturing firms, which highlights the manufacturing companies' greater need to reduce their ecological foot-print. By nature, manufacturing industries tend to pollute more than the services sectors and hence become the main target of regulatory authorities. Moreover, the voluntary proactive approaches to lower environmental hazards can be considered as another significant factor increasing the number of eco-innovators within these industries.

The "Coke and Petroleum products" and "Chemicals and Chemical products" industries have the highest (53.1%) share of eco-innovators among manufacturing industries. The

highest number of eco-innovators is within the "Electrical and Optical Equipment" industries (1984 firms) with 41% of all firms within this industry having reported a positive environmental impact of their realized innovations. "Textile and Textile Products" has the lowest share (13.7%) of eco-innovators among all manufacturing industries. Overall, our summary statistics suggest that firms involved in eco-innovation activities are concentrated in industries where the production process presents a higher technological complexity and involves more negative environmental externalities.

Summary statistics for the eco-innovation proxies (in their categorical and binary forms) together with those for the continuous measure of eco-innovation built from these variables are given in Table 2.

classifications Classification	Total no. of	% of eco-	% of
Classification	firms	innovators	innovators
Services and Other Sectors:	<i>Juni</i> 3		
50: Sale and Repair of Motor	1179	172 (14.6%)	249 (21.1%)
Vehicles		(,.)	(, .)
51: Wholesale Trade	13669	1687 (12.3%)	2668 (19.5%)
52: Retail Trade	2498	248 (9.9%)	410 (16.4%)
60_61_62: Transport	4930	558 (11.3%)	747 (15.2%)
63: Auxiliary Transport Activities	2206	355 (16.1%)	540 (24.5%)
64: Post and telecommunications	1155	289 (25%)	447 (38.7%)
70: Real Estate Activities	728	68 (9.3%)	111 (15.3%)
71: Renting Machinery and	294	38 (12.9%)	69 (23.5%)
Equipment		``´´	
72: Computer and Related Activities	2589	720 (27.8%)	1485 (57.4%)
73_74: R&D and Other Business	7897	1760 (22.3%)	2491 (31.5%)
Activities			
E: Electric, Gas and Water Supply	1921	437 (22.7%)	517 (26.9%)
F: Construction	4353	628 (14.4%)	855 (19.6%)
H: Hotels and restaurants	1424	153 (10.7%)	211 (14.8%)
J: Financial Intermediation	2354	544 (23.1%)	1053 (44.7%)
Manufacturing:			
20_21: Manufacture of wood, pulp	4122	919 (22.3%)	1148 (28%)
and paper			
22: Publishing and printing	2474	562 (22.7%)	811 (32.8%)
27: Basic Metal	1360	479 (35.2%)	534 (39.3%)
28: Fabricated Metal Products	4096	1457 (35.6%)	1710 (35%)
C: Mining and Quarrying	1444	270 (18.7%)	323 (22.4%)
DA: Food, Beverages and Tobacco	7907	2070 (26.2%)	2565 (32.5%)
DB: Textiles and Textile Products	8905	1222 (13.7%)	1652 (18.5%)
DC: Leather and Leather Products	1746	269 (15.4%)	334 (19.1%)
DF_DG: Coke, Petroleum Products &			
Chemicals, Chemical products	2657	1410 (53.1%)	1578 (59.4%)
DH: Rubber and Plastic Products	2531	913 (36.1%)	1059 (41.8%)
DI: Non-metallic Mineral Products	2855	864 (30.3%)	1034 (36.2%)
DK: Machinery and Equipment	4315	1736 (40.2%)	2036 (47.2%)
DL: Electrical and Optical Equipment	4838	1984 (41%)	2418 (50%)
DM: Transport Equipment	2643	1033 (41.2%)	1192 (45%)
DN: Other Manufacturing	3810	1034 (27.1%)	1282 (33.7%)
Total	103710	23879 (23.02%)	31529 (30.64%)

Table 1: Distribution of Eco-innovators and innovators according to industry classifications

3.4. Explanatory variables

Size is one of the most important characteristics of a firm and it is necessary to include it as a control variable in our regressions. The literature stresses the importance of firm size as a determinant of innovation (del Rio Gonzalez, 2005, 2009; Kammerer, 2009), suggesting either a linear relationship (as bigger firms can more easily make large investments in R&D) or an inverted-U relationship (as the intensity of innovation may decrease after firms reach a certain size, due to red tape and stabilization of certain routines). The measure of size available in our dataset is an ordered variable comprising three categories ("small", "medium" or "large") depending on the number of employees, small being the category of reference in our estimations.

Besides firm size, being part of a larger group may also affect firms' propensity to innovate (Cainelli et al., 2011). On the one hand, we can assert that being part of a group may give easier access to the financial resources required for investment in R&D. On the other hand, this may negatively affect their involvement in eco-innovation since as subsidiaries they may have to follow the parent company's innovation strategy. Accordingly, we include a binary variable indicating whether a firm is part of a group.

Market characteristics such as maturity of the market, existing technological opportunities, established and/or future regulations, standards and norms, and consumer awareness can have an impact on the introduction of both innovation and eco-innovation. These characteristics may encourage (or discourage) firms to actively engage in eco-innovation. We include two dummy variables, indicating whether a firm operates mainly on the national or international market, in order to control for the impact of different markets on the introduction of eco-innovation.

Firms' engagement in continuous R&D and R&D intensity are two conventional innovation inputs measuring firms' determination in innovation. The knowledge created and the experience obtained during the research phase is of a great importance for the on-going and future innovative projects. A continuous engagement in R&D not only enhances a firm's scientific and technological capabilities, but also maintains the absorptive capacity needed to undertake successful innovation (Hemmelskamp, 1997; Horbach, 2008). Interestingly, Hemmelskamp (1999) claims that eco-innovative companies are likely to have a lower R&D intensity as they tend to rely more on cooperation and external sources information. In our estimations, we include two distinct variables to control for the respective impacts of R&D intensity and of engagement in continuous R&D. Since in CIS surveys, some firms do not

engage in R&D on a continuous basis (and some firms do not engage in R&D at all), it is usual to use both variables as regressors in econometric models of the innovation production function (see for instance Mairesse and Mohnen, 2002). In our application, R&D intensity is classically measured by the share of (intramural and extramural) R&D expenditures in total sales, while the continuous engagement in R&D is captured by a binary variable equal to 1 in case of continuous R&D and to 0 otherwise.

Hemmelskamp (1999)'s claim on the use of external sources of information took our attention to another crucial point: firms may need to be more open in order to acquire the resources and competences they need for eco-innovation. Following Laursen and Salter's (2004, 2006) approach, we include an ordered variable that controls for the importance of the openness of the firm for the development of eco-innovation. According to Laursen and Salter (2004, 2006), we can measure the degree of openness of a firm by looking at the use of general internal and external sources of information within the innovative process. Following their methodology, we recode the "sources of information" variables provided by CIS4 as binary variables equal to 1 if a source is used and to 0 otherwise. These variables are then summed up to provide the ordered variable, a greater value of which indicates a larger degree of openness. In our application, we use the following information sources²⁷: (1) sources from within the enterprise or its group, (2) suppliers of equipment or materials, (3) clients or customers, (4) competitors and other enterprises in same industry, (5) consultants, commercial labs or private R&D institutes, (6) professional conferences, trade fairs, meetings, (7) scientific journals, trade/scientific publications, (8) professional and industry associations. According to this method, the greatest possible openness score (which is attained if a firm uses all information sources) is equal to 8 while the lowest possible score is 0.

Openness in the above sense does not entail a formal cooperation with the sources of information used (e.g., a firm might obtain information from a competitor without formally cooperating in R&D with this competitor). Formal cooperation may thus be distinct from openness, especially when academic knowledge is involved (see Footnote 3). To take this into account in our model, we include two more variables, as in Robin and Schubert (2013). The first variable is an indicator of scientific cooperation (i.e., cooperation with either

²⁷ As in Laursen and Salter (2004) and Robin and Schubert (2013), we do not include academic sources of information because most firms report no (or a very low) use of these sources. This is a recurrent feature of CIS data, which may be due to the fact that relying on academic knowledge to innovate generally implies some measure of formal cooperation. Thus, the use of academic knowledge is better captured by other variables, such as indicators of cooperation, as will be done here.

universities and higher education institutions or government/public research institutes). The second variable is an indicator of cooperation with other partners, which captures interactions between firms and non-scientific cooperation partners. Cooperating with "other" partners entails a formal cooperation between a firm and at least one of the following: (i) other enterprises within the group, (ii) suppliers of equipment, materials, components, or software, (iii) clients or customers, (iv) competitors or other enterprises in the firm's industry and (v) consultants, commercial labs, or private R&D institutes.

Variables	Mean	Std. dev.
Dependent :		
Firm introduced a product innovation (yes/no)	%23	0.42
Firm introduced a process innovation (yes/no)	%23	0.42
Innovator (Product and/or Process)	%30.4	0.46
Proxy #1 for innovation with environmental impact: EMAT ^a	%83.2	1.02
Proxy #2 for innovation with environmental impact: $EENV^{b}$	%93.5	1.11
Explanatory:		
Size in 2002 ^c	%47.6	0.69
Size in 2004	%51.1	0.69
Part of a group (yes/no)	%23.3	0.42
R&D intensity	%0.33	16.08
Engagement in continuous R&D (yes/no)	%9.7	0.3
Indicator of Openness of the company	7.2%	1.8
Cooperation with public research institutions (yes/no)	%4.71	0.21
Cooperation with others (yes/no)	%9.9	0.3
Firm active in international markets (yes/no)	%35.7	0.48
Firm active in national market (yes/no)	%59.6	0.49
Financial support for innovation activities by the E.U (yes/no)	%2.1	0.14
Exclusion:		
Soviet	%65.4	0.47

 Table 2: Summary statistics of the variables used in the econometric analysis

^a - Reduced materials and energy per unit output (0: Not relevant, 1: Low, 2: Medium and 3: High)

^b - Reduced environmental impacts or improved health and safety (0: Not relevant, 1: Low, 2: Medium and 3: High)

^c - 0: Small, 1: Medium and 2: Large, small being the base category

	Eco-innovators		Inn	ovators
Variables	Mean	Std. Dev.	Mean	Std. Dev.
Size 2002 ^a	76.4%	0.79	70.3%	0.771
Size 2004 ^b	80.3%	0.783	74.3%	0.767
Group	39.2%	0.484	37.6%	0.484
National market	73.9%	0.439	73.6%	0.44
International market	58.4%	0.492	54.8%	0.5
Fund	6.8%	0.253	6%	0.24
Cont. R&D	33.1%	0.47	29.2%	0.46
R&D intensity	0.72	22.71	0.754	24.74
Openness	6.23	1.98	5.8	2.26
Scientific cooperation	16.5%	0.371	14.5%	0.352
Other cooperation	33%	0.47	31%	0.462
Soviet countries	47.9%	0.499	50.1%	0.5

 Table 3: Descriptive statistics of the regressors for eco-innovators and conventional innovators

^a and ^b- 0: Small, 1: Medium and 2: Large, small being the base category

According to van Leeuwen and Mohnen (2013), environmental regulation and public funding of R&D are the first impetus to have more green technologies developed by individual firms. Moreover, Porter and van der Linde (1995) assert that stricter environmental regulation would lead firms to eco-innovate and the economic gains that could be reaped from this investment would offset (partially or fully) the cost of the regulation itself, leading to a "win-win" situation (Porter, 1991; Porter and van der Linde, 1995). Even though these authors consider the regulatory framework as being of utmost importance for the development of eco-innovation, CIS 4 lacks indicators that would allow us to test this claim. The best we have at our disposal is a binary variable indicating whether a firm has received support from the EU's 5th or 6th Framework Program, part of which are designed to foster environmental innovation and sustainable growth. We thus include in our estimations a dummy variable equal to 1 if a firm received any public financial support for innovation activities from the EU's FP during 2002-2004, and to 0 otherwise.

Finally, in order to capture the technological conditions as regard to eco-innovation among different industry segments and among various European countries, we include twodigit Nace codes along with country dummies in our regressions.

3.5. Choice of an exclusion variable

Although the Generalized Tobit model can be identified through its non-linearity alone, it is preferable to rely, for identification, on an exclusion variable which should be correlated with the error term in selection equation, but not with the error term of the intensity equation. This is generally a hard task, which often implies relying on variables sourced in external databases compatible with the one at hand. In our application, such external sources were not available, so we created a dummy variable indicating whether the country to which a firm belongs was an ex-Soviet nation or not. This variable is equal to one if a country is an ex-Soviet Union country *lato sensu* (Estonia, Latvia and Lithuania, Bulgaria, Hungary, Czech Republic, Romania, Slovakia and Slovenia), which was strictly tied to the Soviet Union as a result of having a communist regime. We refer to all of these countries as "Soviet" for simplification. This variable is included among the selection equation regressors, but not among those of the intensity equation, because we expect it to have an impact on being a conventional innovator, but not an eco-innovator.

The rationale for choosing this variable as the exclusion variable is as follows: Communist regimes were well known for stressing the importance of productivity at the national level, and for dedicating a significant amount of resources to science and technology. At the same time, industrial expansion in the Soviet countries often occurred at the expense of environment (of which the draught of the Aral Sea constitutes one of the most striking and extreme examples). We claim that a potential eco-innovator would have technical capabilities to develop knowledge and all the more, it would have environmental consciousness to turn this created knowledge to an applicable solution. The former is definitely one of the fundamental characteristics of the ex-Soviet Union countries. However, the latter is suspected to be one of the most significant barriers to eco-innovate for the same countries. We assume that having a communist regime for long years may have affected the ecologic awareness of former Soviet countries as compared to Western European countries. Especially, with growing global competition firms within these countries have taken advantage of unregulated economic system and political regimes, absent environmental protection framework, problems in implementing existing regulations and standards, auditing problems etc. Therefore, as a result of not having the same environmental consciousness levels with their Western European counterparts, firms within the ex-Soviet countries are presumed to be less eco-innovative. We acknowledge the fact many of these countries are now EU member States (some of them became members in 2004, others in 2007), and that there are certain criteria to

be fulfilled before entering the EU - including ecological criteria. These criteria are negotiated before the acceptance of the candidate country and put into action before their entry into the EU. However, it would naïve to expect total compliance to the legislations right from the start. More precisely, there will be a certain time span between the implementation of the European legislations and the observation of significant changes (in particular as regards an increased social awareness of environmental issues and the widespread adoption of eco-innovation). As a result, it is admissible to assume that establishing widespread environmental consciousness within those countries would take some time even after their acceptance into the EU. This assumption is all the more valid since the observation period of our micro-anonymised CIS 4 dataset 2002-2004, and most of former Soviet countries have become member States in 2004. Two countries, Bulgaria and Romania, have even become member States in 2007, i.e. after the period of observation considered in CIS4. Accordingly, choosing the indicator of a former Soviet regime as an exclusion variable (i.e., assuming it is can be correlated with the error term of the "selection into innovation" equation, but not with the error term of the "intensity of eco-innovation" equation) seems reasonable. This can only be done, however, when we estimate our model on all firms (i.e. the whole sample of 104,943 observations spread across 15 countries). In the second series of estimations, performed by groups of countries, we have to rely on the nonlinearity of the Generalized Tobit model for identification (as do Mairesse and Mohnen, 2002, for instance).

4. Empirical Results

4.1. Estimations conducted on the whole sample

This section discusses the estimation results obtained from the Heckman selection model. Table 4 displays the estimations results from two distinct approaches. We first estimated the Generalized Tobit model with a country fixed effect and without an exclusion variable (since our exclusion variable is an indicator of former Soviet countries, it cannot be combined with a country fixed effect). In this first approach, the model is identified through its non-linearity alone. Second, we have estimated the model with the exclusion variable (which might improve identification) but without the country fixed effect. We present the estimates of the Generalized Tobit with a country fixed-effect (but no exclusion variable) in Columns I and II of Table 4, and those of the Generalized Tobit with an exclusion variable (but without a country fixed effect) in Columns III and IV of Table 4. Table 4 first reveals that five variables (firm size, belonging to a group, type of market, openness and scientific cooperation) are significant within the intensity equations of both models. Therefore, these factors appear as those that are primarily relevant to eco-innovation. We can attribute the significance of firm size firms to financial risks inherent in the eco-innovative process. Small and medium size firms may not have easy access to financial sources and/or technological opportunities required for a successful eco-innovation which may result lack of interest or, no or low success in eco-innovative projects. In general, the innovation literature assumes that the larger firms have easier access to the financial sources thus less effort to innovate (del Rio Gonzalez, 2005, 2009; Kammerer, 2009). These assumptions are consolidated with our estimation results.

In both models, being part of a group is negatively associated with the intensity of ecoinnovation. Firms that have to closely follow the innovation strategy of their parent company may have less opportunity to get actively involved in eco-innovation. Being part of a group may indeed require high level of interdependence among the firms affecting firms' innovation strategy and their decision-making mechanism. Firms' organic network within the same group could diffuse important knowledge related to other firms' innovation activities. However, the diffused knowledge may entail firms to be more similar and homogenous in terms of innovative activities and technological levels. For instance, a general innovation strategy might be binding for all the subsidiaries within the group, which would certainly affect their behaviour towards environmentally benign innovations. Thus, in our case being a conventional innovator rather than an eco-innovator could be the results of such a strategy.

Table 4 also reveals that firms operating mainly on national markets have a lower intensity of eco-innovation, whereas those operating mainly on international markets tend to have (in the second specification of the model, at least), a higher intensity of eco-innovation. This may be because firms exposed to an international market face a more heterogeneous demand, including a higher demand for "greener" products from countries which are more environmentally aware.

According to Table 4, a greater openness of the firms is associated with a higher intensity of eco-innovation, and the same goes for cooperation with science (i.e. universities or public research institutes). Reid and Miedzinski (2008) state that the main features of innovation are determined early in the innovation process. This deduction is especially true for eco-innovations since without the correct apprehension of the environment and environmental aspects of the intended innovation it would be challenging, if not impossible, to develop eco-innovations. A greater openness may allow firms to get a better grasp of environmental issues

and thus allow them to perform better as eco-innovators. In the same vein, firms' ability to establish scientific cooperation may increase their eco-innovation intensity because the complexity of eco-innovation may indeed lead firms to look for other means to complement their capabilities during the eco-innovative process. As stated in Robin and Schubert (2013), the Framework Program (FP) of the EU is the prime example of the cooperation between public research institutions and industry and this claim is consolidated with our results.

Additionally, cooperation with other non-scientific partners is highly significant in the intensity equation of the second model. This indicates that besides the scientific knowledge, firms' formal relationships with other agents is of high importance for actively pursuing new ideas and develop eco-innovations. The results in Table 4 also give us significant clues about various aspects of eco-innovation and its relationship with the conventional innovation inputs. We note that the R&D intensity has no significant impact on the intensity of ecoinnovation as in De Marchi (2012) and Horbach (2008). On the contrary, engagement in continuous R&D is significantly positive in our benchmark model. Hemmelskamp (1997) and Horbach (2008) found evidence that stresses the significance of continuous R&D as one of the conventional innovation inputs that can enhance a firm's scientific and technological capabilities as well as the absorptive capacity needed to undertake a successful innovation. Moreover, De Marchi (2012) suggests that the internal R&D effort is a key component of innovative process, increasing the effectiveness of incoming information and knowledge to the development of innovation. However, we should note here that the intentions of the R&D activities are considerably important for successful eco-innovations. More precisely, simply increasing R&D related expenditures (and, hence, the R&D intensity) may not be enough to develop innovations with positive environmental impacts. In order to do so, one should incorporate the environmental aspects to the R&D process since the beginning of innovative process. On the one hand, incorporating environmental aspects in basic R&D efforts may require several specific competences, knowledge-base, experience, considerable effort, financial sources and of course, time. On the other, some of these prerequisites may not be available within the company, which may yield low or no success in the innovative process. Following the same intuition, our result supports Hemmelskamp's (1999) claim that ecoinnovative companies are likely to have lower R&D intensity, and they tend to countervail the low R&D effort with external use of information sources. This may affirm that conventional innovation and eco-innovation are indeed different in nature and therefore, they need to be treated as two distinct entities. The estimated coefficients of our openness and cooperation variables seem to support Hemmelskamp's claims to a great extent.

	With a country	fixed-effect	With an exclusio	n variable
	(I) Selection equation	(II) Intensity equation	(III) Selection equation	(IV) Intensity equation
Constant term	0.23***	-0.93***	0.94***	-0.71***
	(0.043)	(0.044)	(0.037)	(0.027)
<u>Firm size</u>				
Medium	0.17***	-0.001	0.11***	-0.03*
	(0.015)	(0.013)	(0.014)	(0.013)
Large	0.2***	0.1***	0.11***	0.06***
	(0.021)	(0.017)	(0.021)	(0.016)
<u>Characteristics</u>				
Being part of a Group	0.1^{***}	-0.03*	-0.01	-0.06***
	(0.016)	(0.012)	(0.016)	(0.012)
Active in National market	0.23***	-0.03*	0.25***	-0.04**
	(0.014)	(0.014)	(0.013)	(0.013)
Active in Inter. Market	0.28***	0.02	0.27***	0.03*
	(0.014)	(0.013)	(0.014)	(0.013)
Funded by EU	0.53***	0.03	0.62***	0.03
	(0.072)	(0.022)	(0.069)	(0.023)
Engagement in Continuous R&D	1.16***	0.02	1.01***	0.07***
D&D intensity	(0.029) 2.03***	(0.021) 0.05	(0.028) 1.73***	(0.018) -0.08
R&D intensity	(0.31)	(0.066)	(0.282)	-0.08 (0.065)
Openness	-0.29***	0.16***	-0.3***	0.14***
Openness	(0.004)	(0.004)	(0.004)	(0.004)
Scientific cooperation	0.02	0.09***	-0.03	0.08***
~ · · · · · · · · · · · · · · · · · · ·	(0.054)	(0.018)	(0.051)	(0.019)
Cooperation with others	1.53***	-0.03	1.63***	0.09***
	(0.037)	(0.022)	(0.034)	(0.022)
Exclusion variable				
Soviet	-	-	0.238***	
			(0.013)	
Industry dummies (chi ² test)	2677.04***		2720.58***	
Country Dummies (chi ² test)	4736.90***		-	
No of observations	64919		64919	
Censored observations	37910		37910	
Uncensored observations	27009		27009	
Chi ² (df)	9328.71(53)***		6637.76(39)***	
	///////////////////////////////////////		3037.70(37)	

Table 4: Estimates of the Generalized Tobit Model for eco-innovation

-Associated standard errors are in parenthesis below each coefficient

- *** significant at the 1% level, *significant at the 1% level, significant at the 1% level
- For the 1st model Wald test of indep. Eqs. is: 7.06 with Prob>Chi²:0.008** and rho is: -.11
- For the 2nd model Wald test of indep. Eqs. is: 0.29 with Prob>Chi²:0.589 and rho is: 0.16

Due to the double externality involved in the development of eco-innovation and its pubic good nature, firms' propensity to engage in eco-innovation can be easily affected by the regulatory framework. The central authorities tend to influence firms' behaviour towards eco-innovation by using various tools such as command and control mechanisms, market based instruments, incentives and subsidies. Accordingly, our funding proxy indicates that the EU's Framework Programmes are efficient in assisting firms throughout their conventional innovative process. However, we fail to observe any significant impact of the same Programmes on the intensity of eco-innovation. This could be due to lack of demand for available funds, lack of successful eco-innovative projects, problems inherent in the distribution of funds, excessive applications from the conventional innovators or a relative inefficiency of the EU funding to generate eco-innovation in the period covered by CIS 4.

Lastly, our exclusion variable asserts that former soviet countries are indeed active innovators and they tend to engage in conventional innovation slightly more than their other European counterparts. This result could be the reflection of on-going process of technological, economic and social change within the ex-soviet countries.

4.2. Estimations conducted by groups of countries

The analyses presented in the Section 4.1 aimed at drawing a general picture of ecoinnovators within the 15 EU countries considered in the survey. However, 104,943 observations make for a huge sample, which may results in having some explanatory variables appear more significant in the estimations than they really are (due to the sheer number of degrees of freedom available). Therefore, it makes sense to perform additional estimations on sub-samples, which is what we do in our analysis by groups of countries. Indeed, another important aim of this research is to compare and contrast the profiles of ecoinnovators between different countries (or groups of countries). In order to do so, we have separated the countries according to their geographic proximity. As a result, four different country groups "Western Europe", "Mediterranean", "Eastern Europe" and "Baltic" have been created. Accordingly, the Western European group comprises Germany, Belgium and Norway where Germany is the country of reference for comparison. We deemed it more relevant to include Norway in this group not because of its geographic proximity to Germany or to Belgium but as a result of its economic, technologic and social proximity to these countries. The Mediterranean group includes Spain, Greece and Portugal where Spain is the country of reference within this group. Lithuania, Latvia and Estonia form the third group,

the Baltic countries. Lastly, the Eastern Europe group comprises countries such as Bulgaria (the country of reference), Romania, Slovenia, Slovakia, Czech Republic and Hungary. Table 5.a displays the results of the estimations conducted for the Western Europe and Mediterranean countries, while the Table 5.b presents the results for the Eastern European and Baltic countries.

	Western	Europe	Mediter	anean
	(I) Selection	(II) Outcome	(III) Selection	(IV) Outcome
	equation	(intensity)	equation	(intensity)
		equation		equation
Constant term	-1.45***	-0.55***	1.74***	-0.64***
	(0.078)	(0.111)	(0.084)	(0.047)
<u>Firm size</u>				
Medium	0.22***	-0.01	0.06*	-0.01
	(0.034)	(0.031)	(0.027)	(0.022)
Large	-0.28***	0.199***	0.03	0.074*
	(0.052)	(0.039)	(0.042)	(0.031)
Characteristics				
Being part of a Group	0.04	0.001	0.05	-0.04
	(0.031)	(0.028)	(0.03)	(0.022)
Active in National market	0.123***	-0.1**	0.21***	-0.02
	(0.036)	(0.035)	(0.029)	(0.031)
Active in Inter. Market	0.27***	-0.03	0.23***	-0.02
	(0.033)	(0.031)	(0.027)	(0.024)
Funded by EU	-0.24*	0.16**	0.47***	-0.006
	(0.104)	(0.053)	(0.115)	(0.038)
Engagement in Continuous R&D	0.84***	0.08*	1.22***	0.05
	(0.043)	(0.036)	(0.049)	(0.031)
R&D intensity	1.78***	-0.09	0.6	0.24*
	(0.454)	(0.145)	(0.393)	(0.112)
Openness	0.04***	0.1***	-0.4***	0.13***
	(0.007)	(0.007)	(0.01)	(0.008)
Scientific cooperation	-0.002	0.03	0.11	0.004
	(0.075)	(0.04)	(0.108)	(0.034)
Cooperation with others	1.11^{***}	-0.06	1.59***	0.05
	(0.061)	(0.041)	(0.073)	(0.034)
Industry dummies (chi ² test)	1057.65***		844.52***	
Country Dummies (chi ² test)	156.91***		585.09***	
No of observations	10147		20975	
Censored observations	5672		11737	
Uncensored observations	4475		9238	
Chi ² (df)	1538(38)***		2273.36(41)***	

-Associated standard errors are in parenthesis below each coefficient

***significant at the 1% level, **significant at the 1% level, *significant at the 1% level
For the 1st model Wald test of indep. Eqs. is: 4.47 with Prob>Chi²:0.03* and rho: -0.13
For the 2nd model Wald test of indep. Eqs. is: 0.24 with Prob>Chi²:0.624 and rho: 0.026

At first glance, we observe that only the firm size and openness indicator has established its significance for the eco-innovators within each country group. This indicates that large firms are more likely to be eco-innovators while their openness is another factor that assists them throughout the eco-innovation process. Additionally, the country groups diverge from each other with regard to the significance of conventional innovation inputs and their cooperative behaviour. The former has a more significant effect among Western European and Mediterranean countries, whereas the latter is more significant in Eastern European and Baltic countries.

Furthermore, besides being large in size, and being open to use of external sources of information, the Western European eco-innovators tend to engage in R&D on a continuous basis. This may point out the significance of the technical and scientific knowledge and experience that could be accrued from doing R&D. Financial funding provided by the EU is another significant factor for the Western European eco-innovators. Indeed, considering the double externality problem (Rennings, 2000), financial funds are becoming increasingly important to change firms' innovative behaviour to a more sustainable one. Due to the presence of two distinct externalities inherent in eco-innovation (the first being the usual knowledge spill-overs that occur during the production of an innovation and the second being positive environmental externalities), financial funds and subsidies can mitigate the financial risks involved in eco-innovation.

Mediterranean eco-innovators also tend to be large in size and open to use of external sources of information. Moreover, unlike their Western European counterparts, they tend to eco-innovate by increasing internal and external R&D expenditures, hence R&D intensity. This indicates that both in the Western and the Mediterranean countries conventional innovation inputs are regarded as of the utmost importance for eco-innovation. This result may also be an indication of a comparatively longer R&D history within these countries (when contrasted to Central European and Baltic countries).

	Eastern	Europe	Ba	ltic
	(I) Selection equation	(II) Outcome (intensity) equation	(I) Selection equation	(II) Outcome (intensity) equation
Constant term	5.41***	-0.78***	2.87***	-0.86***
	(0.261)	(0.037)	(0.247)	(0.093)
<u>Firm size</u>				
Medium	0.25***	0.01	0.11	-0.04
	(0.026)	(0.021)	(0.084)	(0.047)
Large	0.488***	0.09***	0.46***	0.11
C	(0.033)	(0.026)	(0.115)	(0.081)
<u>Characteristics</u>				
Being part of a Group	0.14***	-0.03	0.25**	-0.1*
	(0.031)	(0.02)	(0.097)	(0.047)
Active in National market	0.32***	0.005	0.41**	0.01
	(0.024)	(0.019)	(0.151)	(0.062)
Active in Inter. Market	0.35***	0.07***	0.41***	0.16**
	(0.028)	(0.021)	(0.091)	(0.057)
Funded by EU	1.08***	0.02	0.63	0.01
	(0.211)	(0.034)	(0.427)	(0.062)
Engagement in Continuous R&D	1.6***	0.02	0.95***	0.01
	(0.115)	(0.027)	(0.228)	(0.068)
R&D intensity	4.72***	-0.002	-2.2	-0.08
	(1.384)	(0.097)	(1.3)	(0.315)
Openness	-0.71***	0.19***	-0.32***	0.14***
-	(0.03)	(0.007)	(0.025)	(0.019)
Scientific cooperation	0.16	0.17***	-0.09	0.17*
*	(0.171)	(0.031)	(0.262)	(0.065)
Cooperation with others	1.78***	-0.04	1.69***	0.06
	(0.108)	(0.027)	(0.177)	(0.087)
Industry dummies (chi ² test)	1059.7***		610.83***	
Country Dummies (chi ² test) No of observations	2028.1*** 30942		343.06***	
Censored observations	30942 19461		2855 1040	
Uncensored observations	19461		1815	
Chi ² (df)	5044.38(44)***		1015	

Table 5.b. Estimates of the Generalized Tobit model by group of countries

-Associated standard errors are in parenthesis below each coefficient - *** significant at the 1% level, ** significant at the 1% level, * significant at the 1% level -For the 1st model Wald test of indep. Eqs. is: 10.85 with Prob>Chi²:0.001** and rho: -0.13 -For the 2nd model Wald test of indep. Eqs. is: 0.1 with Prob>Chi²:0.749 and rho: 0.09

Eastern European eco-innovators distinguish themselves from Western and Mediterranean eco-innovators by the importance of establishing formal cooperation with scientific partners. Provided that openness is another significant characteristic of the ecoinnovators within this country group, we can claim that information is a key factor that can increase firms' eco-innovation intensity. Moreover, eco-innovators within this country group tend be active in international markets which create an adequate environment for firms to learn about new environmental standards, technological advancements and establish networks that can provide invaluable knowledge and information that can be utilized throughout the eco-innovation process. Mazzanti and Zoboli (2005) show that networking activities may constitute great importance for environmental innovation, even more important than structural characteristics of firms such as size.

Eco-innovators in the Baltic countries tend to have similar characteristics as their Eastern European counterparts. Accordingly, besides being large in size, and being open to use of internal and external source of information, they are also active in international markets and more likely to establish formal cooperation with scientific partners. Additionally, they also tend to be individual firms that are responsible from their own innovative strategy. Therefore, both the Eastern European and the Baltic eco-innovators, in general sense, tend to be more social in their attitude towards eco-innovation incorporating various other partners within the eco-innovation process.

In general, we can conclude that the results that obtained from the global model are robust to country-specific analyses. Both level of analyses show that certain firm characteristics associated with a higher intensity of eco-innovation (namely, size and openness) are commonly shared among firms within all country groups. Other favourable characteristics (being active on the international market and cooperating with scientific partners) appear in two country groups, namely Central European counties and Baltic countries.

Moreover, the importance of conventional R&D inputs and networking activities change respective to the country group, pointing out the differences inherent in each country groups with regard to their eco-innovation behaviour. Both Western European and Mediterranean countries are long-term EU members and enjoyed the welfare gains and standardization enabled by the EU. The EU aims at establishing standardization among countries, especially with regard to environmental issues, and attains a higher level of innovative efforts towards eco-innovation by providing incentives for innovation and passing new laws and regulations to solve sustainability problems. The EU's Gothenburg Summit (2001) and Lisbon strategy (2010) are the primary examples of this strategic behaviour. Therefore, expectedly Western European and Mediterranean firms are relatively more mature than their Eastern European and Baltic counterparts in dealing with environmental issues and moreover, in ecoinnovating. The fact that most of Eastern Europeans and Baltic countries become EU's member states by the end of our observation period (2004) may indicate that firms within these country groups may be in need of additional time to comply with existing rules, standards and regulation and moreover, to eco-innovate at the same level as their Western European and Mediterranean counterparts. While fierce competition in European markets and existing technological differences among country groups may force Eastern European and Baltic firms to speed up their innovation production process, the realization of a successful eco-innovation, as mentioned earlier, is more difficult. It may require additional knowledge, information, technical and organizational capabilities that potential eco-innovative firms in the latter two country groups do not possess. In order to reduce these differences, ecoinnovative firms within the Eastern European and the Baltic countries may use their networking activities for their advantage. As Hemmelskamp (1999) stated, potential ecoinnovative companies are likely to have a lower R&D intensity and tend to countervail their low R&D effort with the use of external sources of information. This is especially true in the case of lack of technological capabilities and increasing need for technical and organizational knowledge for a successful eco-innovation. Indeed, the networking, be it by being open to the use of external sources of information, or by cooperating with scientific or other partners, may convey invaluable knowledge required to undertake an eco-innovation. De Marchi (2012) supports our claim by asserting the role of engagement in continuous R&D activities and cooperation activities as a driver of eco-innovation. We observed that both of these factors are indeed crucial to undertake a successful eco-innovation.

5. Conclusion

Despite its relatively new appearance in the literature, eco-innovation succeeded to attract the attention of many as one of the most important factors to overcome world's sustainability challenges. On the one hand, the complexity of measuring the environmental aspect of an innovation makes it troublesome to gather data and conduct empirical analysis, which has hitherto resulted in a relative lack of empirical evidence about the driving forces of eco-innovation. On the other hand, the eco-innovation literature acknowledges the importance of both conventional (technology-push and market-pull) innovation drivers and environmental regulatory framework as the primary drivers of eco-innovation. However as the eco-innovative process becomes more complex, the role of (and need for) firm-specific capabilities and competencies is getting increasingly important. The acquisition of these capabilities and competencies often comes to rely on cooperation.

us to investigate, besides conventional innovation drivers and regulatory framework, the role of other firm-specific characteristics and capabilities in the eco-innovation process. Accordingly, this paper aimed at identifying, among firm capabilities that are thought to be essential to the innovation process, those characteristics which more specifically favoured the introduction of innovations with positive environmental consequences.

To do so, we estimated an (eco-)innovation production function on micro-anonymised data constructed from firm-level CIS4 (2004) data and covering 104,943 observations across 15 countries. Adapting the "innovativity" approach proposed by Mairesse and Mohnen (2002), we estimated a Generalized Tobit model in which the first equation deals with selection into innovation and the second equation measures the intensity of eco-innovation. This model allowed us to address the selectivity issue that the CIS 4 data may cause. Moreover, we estimated two different specifications of the model in order to check the robustness of the results. The models were first estimated on the whole sample of 104,943 micro-anonymised observations and then within four mutually exclusive groups of countries (Western European countries, Mediterranean countries, Eastern/Central European countries, and Baltic countries).

Our global estimation results indicate that, among the structural characteristics of a firm, being a large and individual firm (i.e., not a subsidiary) is positively associated with a higher intensity of eco-innovation. Being open to the use of external sources of knowledge is another significant attribute of successful eco-innovators. Moreover, we find evidence supporting Hemmelskamp's (1999) claims that eco-innovative firms tend to countervail a lower R&D intensity by cooperating with public research institutes, universities and other higher education institutions. Indeed, increasing R&D intensity may not be sufficient to successfully develop eco-innovation. This suggests one should treat eco-innovation apart from conventional innovation, since the former may require specific intentions, capabilities, considerations, knowledge stock, know-how, strategic planning and more in order to be successfully achieved. Without embedding the environmental issues into the basic R&D efforts one cannot come up with an eco-innovation (ruling out, here, the innovations with unintended positive environmental consequences). Furthermore, while it is generally assumed that more competitive market conditions would lead firms to be more innovative, here only being active on the international market seems to matter for eco-innovation. Being active on the international market does not necessarily imply a fiercer competition, but could expose firms to a more heterogeneous demand, with more widespread requests for "greener" products.

The by-group analyses show that the main results obtained with the global model are robust to country-specific analyses, while highlighting some divergences between different groups of countries as regards their eco-innovation behaviour. More precisely, firm size and openness appear as the main determinants of eco-innovation within all country groups. By contrast, the importance of conventional R&D inputs and networking activities change with the country groups. The former are more significant in Western European and Mediterranean countries while the latter is more significant in Eastern European and Baltic countries. We attribute this divergence to fierce competition in the European markets and the existing technological differences among these country groups. These factors may give Eastern European and Baltic firms incentives to accelerate their innovation production process in order to ameliorate their technological base and catch up with the overall European standards. In order to do so firms may require additional knowledge, information, technical and organizational capabilities that the potential eco-innovative firm does not possess. Therefore, closing this technological gap among countries might be dependent on how well they can complement the eco-innovation process with the external information and knowledge.

To sum it up, this research does not deny the importance of conventional innovation drivers (such as firm size) as drivers of eco-innovation. It does, however, point out to some firm characteristics (such as openness and cooperation) that may act as specific drivers of eco-innovation. This suggests that eco-innovation differs from conventional innovations to a certain extent and that both should be treated as two separate instances. In particular, firms R&D effort is clearly not enough when eco-innovation is the concern. Cooperation, openness to external sources of knowledge and networking are key factors to a successful eco-innovation. It seems, thus, that successful eco-innovations depend not only on firms' individual efforts but, also (and perhaps primarily) on joint efforts realized by governments, firms and (private and public) research institutions.

APPENDICES

To test the robustness of the previous results, we considered also different specifications of the dependent variable. Accordingly, we specified the dependent variables in two distinct ways. In our first approach, we separately re-coded our two eco-innovation proxies i) reduced environmental impacts or improved health and safety (ECO1) and ii) reduced materials and energy per unit output (ECO2) as binary dependent variables and conduct two separate estimations. The first dependent variable takes the value of 1, if the realized innovation has a relevant reduced environmental impact or improved health and safety and 0 otherwise. While, the second proxy is coded as 1, if the realized innovation has led to a reduced materials and energy per unit output and 0 otherwise. The results of this approach are presented in the Appendix A.

In our second approach, our dependent variables of interest are categorical variables coded from 0 to 3 in order to indicate the relative importance of the environmental impacts, i) reduced environmental impacts or improved health (EENV) and, safety and ii) reduced materials and energy per unit output (EMAT), of their product and/or process innovation introduced during the three years 2002 to 2004. Accordingly, the responses take value from 0 to 3 (0 being not relevant to 3 being highly important). The results of this approach are presented in the Appendix B.

In both of these approaches we explicitly take into account sample selection. In the first approach (Appendix A), this leads to the estimation of a "Heckman Probit" in which the selection equation and the intensity equation (both specified as Probit equations) are jointly estimated by ML. In the second approach (Appendix B), this amounts to the joint estimation of a Probit selection equation and an Ordered Probit intensity equation. The joint estimation is performed by ML using the cmp module for Stata (Roodman, 2009).

In what follows, Table A.1 and A.3 in Appendix A (and B.1 and B.3 in Appendix B) present the estimates of the models with a country fixed-effect (and no exclusion variable) and with the "Soviet" exclusion variable (and no country fixed effect) respectively. Tables A.2.a and A.4.a in Appendix A (and B.2.a and B.4.a in Appendix B) display the results of the estimations conducted for the Western Europe and Mediterranean countries, while Tables A.2.b and A.4.b (B.2.b and B.4.b in Appendix B) present the results for Eastern European and Baltic countries.

Appendix A

A.1. The results of global model on ECO1

At first glance, we observe that the significance of the estimated coefficients in our new approach deviates from our previous results to a certain extent. This time the country fixedeffect model indicates that the funds provided by the EU, continuity of R&D efforts and moreover, R&D intensity are among the factors that significantly increases firms' ecoinnovation intensity. The significance of firm size, being an individual firm, international market conditions, firms' openness and formal cooperation between scientific organisations are also consolidated. Furthermore, the benchmark model points out the same direction. Accordingly, eco-innovative firms tend to be individual medium and large size firms that are active in international markets. They distinguish themselves from the other firms by giving significant importance to the funds provided by the EU, engagement in continuous R&D, being open to the internal and external sources of knowledge and, formal cooperation with the scientific and non-scientific organisations. We also detect a substantial difference in significance of two variables. Accordingly, in our new estimation results we also observe that the impact of being medium size firm and impact of R&D intensity is also significantly positive, unlike in our global model.

	With a country	fixed-effect	With an exclusion	variable
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	0.23***	-1.31***	0.94***	-1.09***
	(0.043)	(0.062)	(0.036)	(0.041)
<u>Firm size</u>				
Medium	0.18***	0.045*	0.11***	0.05*
	(0.015)	(0.02)	(0.014)	(0.02)
Large	0.2***	0.17***	0.11***	0.18***
	(0.021)	(0.027)	(0.021)	(0.026)
<u>Characteristics</u>				
Being part of a Group	0.01***	-0.06**	-0.01	-0.01***
	(0.016)	(0.02)	(0.016)	(0.019)
Active in National market	0.24***	0.01	0.26***	0.0
	(0.015)	(0.022)	(0.013)	(0.020)
Active in Inter. Market	0.28***	0.05*	0.26***	0.04*
	(0.015)	(0.02)	(0.014)	(0.02)
Funded by EU	0.53***	0.1^{**}	0.61***	0.13***
	(0.072)	(0.038)	(0.068)	(0.038)
Engagement in continuous R&D	1.17***	0.08**	1.1***	0.08**
	(0.029)	(0.029)	(0.028)	(0.029)
R&D intensity	1.99***	0.29**	1.73***	0.29**
	(0.31)	(0.107)	(0.282)	(0.105)
Openness	-0.29***	0.2***	-0.3***	0.19***
	(0.005)	(0.006)	(0.004)	(0.008)
Scientific cooperation	0.01	0.12***	-0.03	0.11***
	(0.054)	(0.03)	(0.051)	(0.029)
Cooperation with others	1.53***	0.01	1.63***	0.05***
	(0.037)	(0.03)	(0.034)	(0.037)
Exclusion variable				
Soviet	-	-	0.24***	
			(0.013)	
Industry dummies (chi ² test)	2466.61***		2561.77***	
Country Dummies (chi ² test)	3201.38***		_	
No of observations	65058		65058	
Censored observations	37910		37910	
Uncensored observations	27148		27148	
Chi ² (df)	4653.28(53)***		4312.88(39)***	

Table A.1. Estimates of the Heckman Probit Model for e	eco-innovation (ECO1)
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-Associated standard errors are in parenthesis below each coefficient

***significant at the 1% level, **significant at the 1% level, *significant at the 1% level
For the 1st model Wald test of indep. Eqs. is: 1.39 with Prob>Chi²:0.2377 and rho: 0.048
For the 2nd model Wald test of indep. Eqs. is: 6.25 with Prob>Chi²:0.0124* and rho: 0.12

A.2. The results of the country specific analyses on ECO1

The country specific analyses display another picture of the relationship between the selection into innovation and intensity of eco-innovation, and firm characteristics.

Accordingly, the Western European eco-innovators tend to be large firms that can efficiently use the funds provided by the EU and more importantly, they assist their innovation process with the use of internal and external sources of information hence by being open. Mediterranean eco-innovators, however, tend to be large individual firms that give significant importance to the R&D intensity in order to produce successful innovations while using the internal and external sources of information. Therefore, the main difference between the Western European and Mediterranean eco-innovators is the relative importance given to the R&D intensity.

	Western Europe		Medite	rranean
	(I) Selection equation	(II) Outcome	(III) Selection equation	(IV) Outcome (intensity)
		(intensity) equation		equation
Constant term	-1.43***	-0.53	1.74***	-1.01***
	(0.078)	(0.294)	(0.084)	(0.075)
<u>Firm size</u>				
Medium	0.22***	0.01	0.06*	0.06
	(0.033)	(0.052)	(0.027)	(0.035)
Large	-0.26***	0.27***	0.03	0.17***
-	(0.052)	(0.067)	(0.042)	(0.049)
Characteristics				
Being part of a Group	0.035	-0.05	0.05	-0.11**
	(0.031)	(0.045)	(0.03)	(0.035)
Active in National market	0.134***	-0.05	0.21***	-0.01
	(0.036)	(0.058)	(0.029)	(0.049)
Active in Inter. Market	0.26***	-0.08	0.23***	-0.04
	(0.036)	(0.054)	(0.027)	(0.037)
Funded by EU	-0.25*	0.19*	0.47***	0.04
ý	(0.104)	(0.091)	(0.115)	(0.062)
Engagement in Continuous R&D	0.84***	0.07	1.22***	0.08
6.6	(0.043)	(0.083)	(0.049)	(0.051)
R&D intensity	1.8***	-0.2	0.5	0.38*
5	(0.452)	(0.249)	(0.393)	(0.185)
Openness	0.04***	0.18***	-0.4***	0.2***
I	(0.007)	(0.014)	(0.01)	(0.015)
Scientific cooperation	-0.02	0.04	0.11	0.02
I	(0.076)	(0.066)	(0.108)	(0.055)
Cooperation with others	1.12***	-0.13	1.59***	0.06
r i i i i i i i i i i i i i i i i i i i	(0.061)	(0.096)	(0.074)	(0.056)
Industry dummies (chi ² test)	699.48***		845.88***	
Country Dummies (chi ² test)	157.97***		509.85***	
No of observations	10194		20977	
Censored observations	5672		11737	
Uncensored observations	4522		9240	
Chi ² (df)	713.66(38)***		1526.77(41)***	

Table A.2.a. Estimates of the Global model by group of countries

-Associated standard errors are in parenthesis below each coefficient - ***significant at the 1% level, **significant at the 1% level, *significant at the 1% level -For the 1st model Wald test of indep. Eqs. is: 1.02 with Prob>Chi²:0.3133 and rho: -0.16 -For the 2nd model Wald test of indep. Eqs. is: 0.05 with Prob>Chi²:0.8218 and rho: 0.02

$\begin{array}{ c c c c c c } \hline equation (intensity) equation (intensity) equation equation (intensity) equation (intensity) equation (intensity) equation (intensity) equation (intensity) equation (intensity) $		Eastern 2	Europe	Ba	ltic
Immediate (0.26) (0.058) (0.246) (0.157) Firm size (0.26) (0.058) (0.246) (0.157) Medium 0.25^{***} 0.01 0.11 0.07 (0.026) (0.032) (0.034) (0.075) Large 0.48^{***} 0.13^{**} 0.46^{***} 0.13 (0.031) (0.041) (0.114) (0.123) CharacteristicsBeing part of a Group 0.15^{***} -0.04 0.26^{**} -0.08 (0.031) (0.032) (0.097) (0.077) Active in National market 0.32^{***} 0.04 0.42^{**} 0.23^{**} (0.024) (0.03) (0.15) (0.093) Active in Inter. Market 0.35^{***} 0.13^{***} 0.4^{***} 0.35^{***} (0.028) (0.031) (0.09) (0.086) Funded by EU 1.07^{***} 0.13^{*} 0.63 0.21 (0.213) (0.063) (0.419) (0.172) Engagement in Continuous R&D 1.6^{***} 0.001 0.95^{***} 0.15 (0.17) (0.03) (0.179) (1.269) (0.561) Openness -0.71^{***} 0.21^{***} -0.32^{***} 0.21^{***} (0.03) (0.009) (0.025) (0.022) (0.021) (0.25) (0.17) (0.03) (0.09) (0.25) (0.022) (0.17) (0.03) (0.07) (0.048) (0.25) (0.022) (0.17) </th <th></th> <th></th> <th>(intensity)</th> <th></th> <th>•</th>			(intensity)		•
Firm size Medium 0.25^{***} (0.026) (0.032) (0.032) (0.084) (0.075) (0.075) (0.033) (0.041) (0.114) (0.114) (0.113) (0.113) (0.114) (0.113) (0.114) (0.113) Characteristics Being part of a Group 0.15^{***} (0.031) (0.031) (0.032) (0.032) (0.097) (0.077) Active in National market (0.024) (0.024) (0.031) (0.031) (0.031) (0.031) (0.031) (0.031) (0.097) (0.077) Active in Inter. Market (0.028) (0.028) (0.031) (0.091) (0.091) (0.093) Active in Inter. Market (0.028) (0.0213) (0.063) (0.042) (0.031) (0.09) (0.095) (0.042) (0.031) (0.09) (0.095) (0.0213) (0.063) (0.419) (0.172) Engagement in Continuous R&D $(1.6^{***}$ 0.051 (0.115) (0.042) (0.042) (0.042) (0.256) (0.094) R&D intensity (0.15) (0.173) (0.173) (0.179) (1.269) (0.251) (0.251) (0.221) Scientific cooperation (0.15) (0.171) (0.033) (0.077) (0.078) (0.177) (0.048) (0.251) (0.176) 1003^{***} (0.177) (0.048) (0.251) (0.176) 10012^{***} (0.071) (0.038) (0.176) (0.075) (0.096) (0.107) (0.038) (0.176) 10012^{***} (0.075) (0.078) (0.075) (0.078) (0.075) (0.096) (0.176) 10012^{***} (0.071) (0.038) (0.176) 10012^{***} (0.076) (0.075) (0.076) Industry dummies (chi² test) (0.107) 1012.05^{***} 1003^{***} 1003 10071.13^{***} 360.6^{***} 360.6^{***} 360.6^{***} 3	Constant term	5.4***	-1.16***	2.87***	-1.46***
Medium 0.25^{***} 0.01 0.11 0.07 Large 0.48^{***} 0.13^{**} 0.46^{***} 0.13 Large 0.48^{***} 0.13^{**} 0.46^{***} 0.13 Characteristics 0.033 (0.041) (0.114) (0.123) Being part of a Group 0.15^{***} -0.04 0.26^{**} -0.08 (0.031) (0.032) (0.097) (0.077) Active in National market 0.32^{***} 0.04 0.42^{**} 0.23^{*} (0.024) (0.03) (0.15) (0.093) Active in Inter. Market 0.35^{***} 0.13^{***} 0.4^{***} 0.35^{***} (0.028) (0.031) (0.09) (0.086) (0.031) (0.09) (0.086) Funded by EU 1.07^{***} 0.13^{*} 0.63 0.21 (0.213) (0.063) (0.419) (0.172) Engagement in Continuous R&D 1.6^{***} 0.001 0.95^{***} 0.15 (0.17) (0.042) (0.226) (0.094) R&D intensity 4.67^{***} 0.58^{**} -2.26 -0.08 (0.33) (0.009) (0.225) (0.225) (0.225) (0.225) Openness -0.71^{***} 0.23^{***} -0.32^{***} 0.21^{***} (0.03) (0.077) (0.048) (0.25) (0.228) Cooperation 0.15 0.23^{***} -0.09 0.42^{***} Cooperation with others 1.77^{***} -0.05 1.69^{***		(0.26)	(0.058)	(0.246)	(0.157)
Large (0.026) $0.48***$ (0.033) (0.032) (0.041) (0.084) $0.46***$ (0.114) (0.075) 0.13 (0.123) Characteristics (0.033) (0.011) (0.041) (0.011) (0.114) (0.123) (0.123) (0.077) Chiracteristics (0.031) (0.031) (0.032) (0.097) (0.077) (0.077) Active in National market 0.32^{***} (0.024) 0.04 (0.03) 0.15^{***} (0.093) Active in Inter. Market 0.35^{***} (0.028) 0.13^{***} (0.031) 0.4^{***} (0.031) 0.35^{***} (0.031) Funded by EU 1.07^{***} (0.213) 0.063 (0.419) (0.172) (0.260) (0.074) (0.071) Engagement in Continuous R&D (0.115) 1.6^{***} (0.0115) 0.001 (0.025) 0.95^{***} (0.126) 0.15^{***} (0.094) Qpenness (0.03) (0.175) (0.172) (0.226) (0.072) (0.022) (0.022) (0.022) Scientific cooperation 0.15 (0.177) 0.21^{***} (0.038) (0.25) (0.176) (0.176) (0.072) Industry dummies (chi^2 test) 1003^{***} 1012.05^{***} 1071.13^{***} 360.6^{***} 31032 2855 2855 2855 1051 1040 Industry dummies (chi^2 test) 1003^{***} 1072.05^{***} 1024 1040 1040 Uncensored observations 11571 1815 1071.13^{**}	<u>Firm size</u>				
Large 0.48^{***} 0.13^{**} 0.46^{***} 0.13 (0.033)(0.041)(0.114)(0.123)CharacteristicsBeing part of a Group 0.15^{***} -0.04 0.26^{**} -0.08 Active in National market 0.32^{***} 0.04 0.42^{**} 0.23^{*} Active in National market 0.32^{***} 0.04 0.42^{**} 0.23^{*} Active in Inter. Market 0.35^{***} 0.13^{***} 0.4^{***} 0.35^{***} (0.024) (0.03) (0.15) (0.093) Active in Inter. Market 0.35^{***} 0.13^{***} 0.4^{***} 0.35^{***} (0.028) (0.031) (0.09) (0.086) Funded by EU 1.07^{***} 0.13^{*} 0.63 0.21 (0.213) (0.063) (0.419) (0.172) Engagement in Continuous R&D 1.6^{***} 0.58^{**} -2.26 -0.08 (1.373) (0.179) (1.269) (0.561) Openness -0.71^{***} 0.21^{***} 0.22^{***} 0.21^{***} (0.03) (0.009) (0.025) (0.022) (0.096) Scientific cooperation 0.15 0.23^{***} -0.09 0.42^{***} (0.177) (0.048) (0.25) (0.108) Cooperation with others 1.77^{***} -0.05 1.69^{***} 0.06 (0.176) 1002.5^{***} 360.6^{***} 360.6^{***} No of observations 31032 2855 2855 </td <td>Medium</td> <td>0.25***</td> <td>0.01</td> <td>0.11</td> <td>0.07</td>	Medium	0.25***	0.01	0.11	0.07
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Industry dummies (chi² test) 1003*** 1071.13*** Country Dummies (chi² test) 1012.05*** 360.6*** No of observations 31032 2855 Censored observations 19461 1040 Uncensored observations 11571 1815	Cooperation with others		-0.05	1.69***	· · · · · · · · · · · · · · · · · · ·
Country Dummies (chi ² test) 1012.05*** 360.6*** No of observations 31032 2855 Censored observations 19461 1040 Uncensored observations 11571 1815	•	(0.107)	(0.038)	(0.176)	(0.096)
No of observations310322855Censored observations194611040Uncensored observations115711815	Industry dummies (chi ² test)	1003***		1071.13***	
No of observations310322855Censored observations194611040Uncensored observations115711815	Country Dummies (chi ² test)	1012.05***		360.6***	
Censored observations194611040Uncensored observations115711815	•	31032		2855	
Uncensored observations 11571 1815					
$(h_1/(d_1))$ $(X/(A_1) \times X)$	Chi ² (df)	1834.76(44)***		1015	

Table A.2.b. Estimates of the Global model by group of countries

-Associated standard errors are in parenthesis below each coefficient

***significant at the 1% level, **significant at the 1% level, *significant at the 1% level
For the 1st model Wald test of indep. Eqs. is: 0.996 with Prob>Chi²: 0.00*** and rho: -0.0002
For the 2nd model Wald test of indep. Eqs. is: 1.1 with Prob>Chi²: 0.2935 and rho: 0.21

We observe a similar difference between the Eastern European and Baltic ecoinnovators. We note that each country group value different innovation inputs. For example, the estimation results indicate the fact that the Eastern European eco-innovators put more emphasis on the importance of R&D intensity even though they do not necessarily engage in intramural R&D efforts. They also complement their shortcomings with internal and external information and, cooperative efforts with the scientific institutions. The eco-innovators in the Baltic countries however, choose to countervail technical aspects of innovation process with the use of internal and external sources of information and, establishing formal bond with the scientific institutions. Besides their attitude towards the innovative process, their structural characteristics also deviate. We observe that eco-innovators in the Eastern European countries tend to be large in size and in search of external financial aid to engage in ecoinnovation activities. Nevertheless, international market condition seems to be a significant factor affecting the eco-innovation intensity of firms within both of the country groups. Additionally, national market condition perceived as a significant factor only in the Baltic countries.

A.3. The results of global model on ECO2

Table A.3 below presents the estimation results obtained from the Heckman Probit with a country fixed-effect and the Heckman Probit with an exclusion variable. As mentioned in the beginning of this section, the dependent variable used in this approach is a binary response variable indicating whether the realized innovation has led to "reduced materials and energy per unit output" (ECO2).

By inspecting the estimation results we can notice that the country fixed-effect model and the benchmark model contradict each other only on the impact of size, where this variation could be attributed to the inclusion of an exclusion variable. Accordingly, the country fixed-effect estimation result points out that eco-innovator tend to be large individual firm that pursues continuous R&D while trying to assist the innovative process with the use of internal and external sources of knowledge. Moreover, we can also note that national market conditions tend to have a significant negative impact on the intensity of ecoinnovation. Furthermore, the result of the benchmark model shows that being medium and large size firm, national market conditions, the funds provided by the EU and the R&D intensity tend to negatively effects firms' eco-innovation intensity. While, international market conditions, engagement in continuous R&D efforts, the use of internal and external sources of knowledge and, formal cooperation with other non-scientific organizations tend to increase eco-innovation intensity. Even though, these two models contradicts each other and show a slight deviation from each other the significance of most of the variables is established.

	With a country f	ixed-effect	With an exclusion	variable
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	0.23***	-1.17***	0.94***	-1.03***
	(0.043)	(0.066)	(0.037)	(0.041)
Firm size				
Medium	0.18***	0.015	0.114***	-0.1***
	(0.015)	(0.021)	(0.014)	(0.02)
Large	0.2***	0.14***	0.111***	-0.077**
	(0.021)	(0.028)	(0.021)	(0.025)
Characteristics				
Being part of a Group	0.01***	-0.06**	-0.009	-0.077
	(0.017)	(0.021)	(0.016)	(0.019)
Active in National market	0.24***	-0.055*	0.256***	-0.11***
	(0.015)	(0.023)	(0.013)	(0.019)
Active in Inter. Market	0.28***	0.023	0.265***	0.052**
	(0.015)	(0.023)	(0.014)	(0.02)
Funded by EU	0.52***	-0.022	0.613***	-0.097**
	(0.073)	(0.038)	(0.069)	(0.036)
Engagement in Continuous R&D	1.17***	0.01***	1.088***	0.31***
	(0.029)	(0.029)	(0.028)	(0.029)
R&D intensity	1.99***	0.11	1.745***	-0.28**
	(0.311)	(0.108)	(0.284)	(0.098)
Openness	-0.29***	0.205***	-0.297***	0.16***
	(0.005)	(0.006)	(0.004)	(0.008)
Scientific cooperation	0.01	0.016	-0.033	0.04
~	(0.054)	(0.03)	(0.051)	(0.029)
Cooperation with others	1.53***	-0.034	1.627***	0.19***
	(0.038)	(0.03)	(0.034)	(0.038)
Exclusion variable				
Soviet	-	-	0.243***	
			(0.013)	
Industry dummies (chi ² test)	1742.91***		1849.82***	
Country Dummies (chi ² test)	4218.52***		-	
No of observations	65058		65058	
Censored observations	37910		37910	
Uncensored observations	27148		27148	
Chi ² (df)	4099.12(53)***		3542.65(39)***	

 Table A.3. Estimates of the Heckman Probit Model for eco-innovation (ECO2)

-Associated standard errors are in parenthesis below each coefficient - ***significant at the 1% level, **significant at the 1% level, *significant at the 1% level -For the 1st model Wald test of indep. Eqs. is: 1.16 with Prob>Chi²: 0.2819 and rho: -0.045

-For the 2nd model Wald test of indep. Eqs. is: 1.01 with Prob>Chi²: 0.3152 and rho: 0.05

A.4. The country specific analyses on ECO2

The country specific analyses give us significant clues in order to identify the ecoinnovators within different country groups. The results assert that the eco-innovators in the Western European countries are tend to be large firms that engage in continuous R&D and open to the use of internal and external sources of information. Moreover, we can also note that the national market tend to hinder the eco-innovation intensity indicating that the national market conditions are still away from being a significant driver for the introduction and development of innovations with the reduced materials and energy per unit output (ECO2). The Mediterranean countries however, tend to differ from their Western European counterparts in regard to size. Accordingly, the size is not a significant factor affecting firms' eco-innovation intensity in Mediterranean countries. Nevertheless, we can observe that R&D intensity, continuous engagement in R&D efforts and firms' openness to the use of internal and external knowledge sources have significant positive impact on the firms' eco-innovation intensity.

Table A.4.a. Estimates of the C	Western Europe		Mediterranean	
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	-1.42***	-1.22***	1.74***	-0.79***
	(0.078)	(0.232)	(0.084)	(0.072)
Firm size				
Medium	0.22***	0.04	0.06*	-0.02
	(0.033)	(0.051)	(0.027)	(0.034)
Large	-0.26***	0.26***	0.03	0.08
-	(0.052)	(0.066)	(0.042)	(0.048)
Characteristics				
Being part of a Group	0.035	-0.035	0.06	-0.06
	(0.031)	(0.046)	(0.03)	(0.035)
Active in National market	0.13***	-0.14*	0.21***	-0.08
	(0.036)	(0.058)	(0.029)	(0.048)
Active in Inter. Market	0.26***	0.07	0.23***	-0.02
	(0.033)	(0.052)	(0.027)	(0.036)
Funded by EU	-0.24*	0.07	0.47***	-0.06
	(0.104)	(0.087)	(0.115)	(0.059)
Engagement in Continuous R&D	0.84***	0.164*	1.22***	0.12*
D&D intensity	(0.043) 1.8^{***}	(0.07) 0.03	(0.049) 0.59	(0.048) 0.43*
R&D intensity	(0.455)	(0.237)	(0.393)	(0.182)
Openness	0.04***	0.18***	-0.4***	0.17***
Openness	(0.007)	(0.013)	(0.01)	(0.014)
Scientific cooperation	-0.015	-0.09	0.11	-0.07
i i i i i i i i i i i i i i i i i i i	(0.076)	(0.065)	(0.109)	(0.052)
Cooperation with others	1.2***	0.025	1.6***	0.03
-	(0.061)	(0.083)	(0.073)	(0.053)
Industry dummies (chi ² test)	668.76***		494.26***	
Country Dummies (chi ² test)	157.77***		496.09***	
No of observations	10194		20977	
Censored observations				
	5672		11737	
Uncensored observations	4522		9240	
Chi ² (df)	728.14(38)***		1005.51(41)***	

-Associated standard errors are in parenthesis below each coefficient

*** significant at the 1% level, ** significant at the 1% level, *significant at the 1% level
For the 1st model Wald test of indep. Eqs. is: 0.06 with Prob>Chi²: 0.8089 and rho:0.03
For the 2nd model Wald test of indep. Eqs. is: 1.15 with Prob>Chi²:0.2836 and 0.083

	Eastern Europe		Baltic	
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	5.41***	-1.27***	2.87***	-1.06***
	(0.261)	(0.067)	(0.249)	(0.154)
Firm size				
Medium	0.25***	0.055	0.11	-0.05
	(0.026)	(0.035)	(0.085)	(0.074)
Large	0.48***	0.13**	0.45***	0.11
	(0.033)	(0.044)	(0.113)	(0.124)
<u>Characteristics</u>				
Being part of a Group	0.14^{***}	-0.075*	0.26**	-0.04
	(0.031)	(0.035)	(0.097)	(0.075)
Active in National market	0.32***	-0.02	0.41**	0.14*
	(0.024)	(0.034)	(0.151)	(0.091)
Active in Inter. Market	0.34***	0.09**	0.4***	0.16***
	(0.028)	(0.035)	(0.092)	(0.086)
Funded by EU	1.07***	-0.05	0.63	0.25
	(0.212)	(0.065)	(0.426)	(0.175)
Engagement in Continuous R&D	1.6***	0.09*	0.96***	0.16
R&D intensity	(0.114) 4.65^{***}	(0.044) -0.11	(0.231) -2.23	(0.091) 0.58
R&D Intensity	(1.373)	(0.182)	-2.23 (1.297)	(0.579)
Openness	-0.71***	0.24***	-0.32***	0.2***
openness	(0.03)	(0.009)	(0.025)	(0.022)
Scientific cooperation	0.15	0.09	-0.08	0.12***
L.	(0.17)	(0.05)	(0.252)	(0.105)
Cooperation with others	1.76***	-0.09*	1.69***	-0.001
	(0.107)	(0.041)	(0.177)	(0.098)
Industry dummies (chi ² test)	726.7***		928.77***	
Country Dummies (chi ² test)	1736.22***		344.23***	
No of observations	31032		2855	
Censored observations	19461		1040	
Uncensored observations	11571		1815	
Chi ² (df)	5176.33(88)***		1015	
	51/0.55(00)		•	

Table A.4.b. Estimates of the Global model by group of countries

-Associated standard errors are in parenthesis below each coefficient - *** significant at the 1% level, ** significant at the 1% level, * significant at the 1% level -For the 1st model Wald test of indep. Eqs. is: 0.55 with Prob>Chi²: 0.4597** and rho: -0.18 -For the 2nd model Wald test of indep. Eqs. is: 0.32 with Prob>Chi²: 0.5744 and rho: 0.12

Appendix B

B.1. The results of global model on EENV

In the main analyses, our dependent variable was the normalized continuous index built from the factor analysis on two distinct eco-innovation proxies. In appendix A, we changed our approach and we used a binary response model with sample selection hence a binary response variable as the dependent variable. As mentioned before, in this section we change our approach towards the main dependent variables and we use two distinct ordered response variables (EENV and EMAT).

In the present section we discuss the results with those pertaining to EENV. By inspecting the result of the country fixed-effect model, we note that the results slightly deviate from the main results. Accordingly, we affirm that, in general, the eco-innovators tend to be large individual firms that are mainly active in international markets and can efficiently use the funds provided by the EU. More importantly, we can also claim that the formal cooperation with scientific organizations and the use of internal and external sources of knowledge are other significant characteristics commonly shared among the eco-innovators.

	With a country fi	ixed-effect	With an exclusion	on variable
	(I) Selection	(II)	(III) Selection	(IV)
	equation	Outcome	equation	Outcome
		(intensity)		(intensity)
		equation		equation
Constant term	0.23***		0.94***	
	(0.043)	-	(0.037)	-
<u>Firm size</u>				
Medium	0.18***	-0.001	0.11***	-0.003*
	(0.015)	(0.017)	(0.014)	(0.017)
Large	0.2***	0.1^{***}	0.11***	0.12***
C	(0.021)	(0.022)	(0.021)	(0.021)
Characteristics				
Being part of a Group	0.1^{***}	-0.03*	-0.01	-0.1***
	(0.016)	(0.016)	(0.016)	(0.016)
Active in National market	0.24***	0.01	0.25***	0.008
	(0.014)	(0.019)	(0.013)	(0.017)
Active in Inter. Market	0.28***	0.03*	0.27***	0.016
	(0.015)	(0.018)	(0.014)	(0.017)
Funded by EU	0.52***	0.13***	0.61***	0.15***
	(0.072)	(0.03)	(0.069)	(0.03)
Engagement in Continuous R&D	1.16***	0.04	1.09***	-0.017
8-8-	(0.029)	(0.03)	(0.028)	(0.029)
R&D intensity	1.99***	0.01	1.73***	0.12
	(0.31)	(0.089)	(0.281)	(0.086)
Openness	-0.29***	0.16***	-0.29***	0.16***
- F	(0.004)	(0.007)	(0.004)	(0.009)
Scientific cooperation	0.012	0.14***	-0.03	0.11***
	(0.054)	(0.023)	(0.051)	(0.023)
Cooperation with others	1.53***	0.02	1.63***	0.001
	(0.037)	(0.033)	(0.034)	(0.04)
Exclusion variable	(0.007)	(01000)	(01001)	(0101)
Soviet	-	_	0.24***	
			(0.013)	
Industry dummies (chi ² test)	2335.83***		2460.98***	
Country Dummies (chi ² test)	3289.64***			
No of observations	65058		- 65058	
	15297.01(106)***		14538.38(79)***	
Chi ² (df)	13297.01(100)		14330.30(79)****	
Cut-off points		1 11444		074***
Cut-off - 1		1.11^{***}		0.74***
		(0.063)		(0.04)
Cut-off - 2		1.65***		1.28***
		(0.064)		(0.04)
Cut-off - 3		2.52***		2.13***
		(0.064)		(0.041)

Table B.1. Ordered Res	ponse Models with Samp	ple Selection (using	g EENV Proxy)
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-Associated standard errors are in parenthesis below each coefficient - ****significant at the 1% level, **significant at the 1% level, *significant at the 1% level

The benchmark model again consolidates the significance of our key variables. The only significant difference between the country fixed-effect model and the benchmark model is the role of the international market conditions. The significance of the estimated coefficients of the rest of the variables is consolidated in our benchmark model.

B.2. Country specific analyses on EENV

The country specific analyses again give another picture of the eco-innovators within each country groups. We observe that the Western European eco-innovators tend to be large in size, receive financial support from the EU and are open to the use of internal and external sources of knowledge. However, we also note that the national and international market conditions seem to be far away from offering significant opportunities to engage in ecoinnovative activities. Instead, they seem to hinder firms' eco-innovation intensity.

	Western E	Curope	Mediterra	anean
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	-1.43***	-	1.74***	
	(0.078)	-	(0.085)	-
<u>Firm size</u>				
Medium	0.22***	-0.03	0.06*	-0.01
	(0.033)	(0.043)	(0.027)	(0.028)
Large	-0.26***	0.15**	0.028	0.08*
	(0.052)	(0.054)	(0.042)	(0.039)
<u>Characteristics</u>				
Being part of a Group	0.035	-0.017	0.06	-0.05
	(0.031)	(0.039)	(0.03)	(0.028)
Active in National market	0.13***	-0.094*	0.2***	0.01
	(0.036)	(0.048)	(0.029)	(0.042)
Active in Inter. Market	0.26***	-0.11*	0.23***	-0.03
	(0.033)	(0.043)	(0.027)	(0.033)
Funded by EU	-0.245*	0.23**	0.47***	0.05
	(0.104)	(0.069)	(0.115)	(0.049)
Engagement in Continuous R&D	0.84***	0.05	1.22***	0.046
D&D interneiter	(0.043) 1.79***	(0.054)	(0.049)	(0.068)
R&D intensity	(0.455)	-0.127 (0.22)	0.59 (0.393)	0.23 (0.15)
Openness	0.04***	(0.22) 0.14***	-0.4***	0.15***
Openness	(0.007)	(0.013)	(0.01)	(0.021)
Scientific cooperation	-0.015	0.083	0.11	0.053
	(0.076)	(0.054)	(0.109)	(0.042)
Cooperation with others	1.12***	-0.092	1.59***	0.04
1 I	(0.061)	(0.062)	(0.073)	(0.075)
Industry dummies (chi ² test)	692.33***		795.21***	
Country Dummies (chi ² test)	145.23***		461.23***	
No of observations	10194		20977	
Chi ² (df)	2818.34(76)***		4489.24(82)***	
Cut-off points	(, _)			
Cut-off - 1		0.34		0.65***
Cut-off – 2		(0.18) 1.08^{***} (0.18)		(0.07) 1.2*** (0.07)
Cut-off - 3		(0.18) 1.92*** (0.19)		(0.07) 2.05*** (0.072)

Table B.2.a. Ordered Response Models with Sample Selection (EENV Proxy) by group of countries

-Associated standard errors are in parenthesis below each coefficient

- *** significant at the 1% level, ** significant at the 1% level, * significant at the 1% level

Furthermore, the results claim that the Mediterranean eco-innovators tend to be large firms that are open to the use of internal and external sources of knowledge. We fail to observe a significance relationship between the eco-innovation intensity and the rest of the firm characteristics.

	Eastern I	Europe	Baltic		
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation	
Constant term	5.4***		2.9***		
	(0.26)	-	(0.247)	-	
<u>Firm size</u>					
Medium	0.25***	-0.005	0.11	-0.03	
	(0.026)	(0.028)	(0.085)	(0.064)	
Large	0.48***	0.08*	0.46***	0.14	
	(0.033)	(0.034)	(0.113)	(0.111)	
<u>Characteristics</u>					
Being part of a Group	0.14***	-0.008	0.25**	-0.15*	
	(0.031)	(0.026)	(0.097)	(0.064)	
Active in National market	0.32***	0.05	0.41**	0.16*	
	(0.024)	(0.026)	(0.151)	(0.081)	
Active in Inter. Market	0.35***	0.1***	0.41***	0.3***	
	(0.028)	(0.027)	(0.092)	(0.079)	
Funded by EU	1.07***	0.16***	0.63	0.32*	
Engagement in Continuous P&D	(0.213) 1.6***	(0.047) -0.01	(0.419) 0.95***	(0.144) 0.12	
Engagement in Continuous R&D	(0.115)	(0.036)	(0.226)	(0.084)	
R&D intensity	4.68***	0.1	-2.22	-0.19	
Red intensity	(1.375)	(0.126)	(1.279)	(0.505)	
Openness	-0.71***	0.18***	-0.32***	0.16***	
1	(0.03)	(0.009)	(0.025)	(0.022)	
Scientific cooperation	0.22***	0.22***	-0.107	0.29***	
	(0.036)	(0.036)	(0.262)	(0.085)	
Cooperation with others	-0.014	-0.014	1.69***	0.1	
	(0.035)	(0.035)	(0.177)	(0.101)	
Industry dummies (chi ² test)	955.8***		291.98***		
Country Dummies (chi ² test)	975.36***		359.12***		
•					
No of observations	31032		2855		
Chi ² (df)	4829.98(88)***		1614.74(80)***		
Cut-off points					
Cut-off - 1		0.84***		1.18***	
Critical ff 2		(0.054)		(0.137)	
Cut-off – 2		1.32^{***}		1.7^{***}	
Cut-off - 3		(0.055) 2.2***		(0.14) 2.65***	
Cut-011 - 5		(0.057)		(0.15)	

Table B.2.b. Ordered Response Models with Sample Selection (EENV Proxy) by group of countries

-Associated standard errors are in parenthesis below each coefficient - ***significant at the 1% level, **significant at the 1% level, *significant at the 1% level

B.3. The results of global model on EMAT

This section presents the results pertaining to the eco-innovation with "reduced materials and energy per unit output" (EMAT). The estimated coefficients obtained from the global models take our attention to the significant differences between the two models.

On the one hand, our country fixed-effect model shows that eco-innovative firms tend to be large and individual, while they also tend to be open to the use of internal and external sources of information. Moreover, the national market conditions seem to have a significant negative impact on firms' eco-innovation intensity.

On the other hand, our benchmark model claims that the international market conditions, engagement in continuous R&D, openness of the company and formal cooperation with other non-scientific partners tend to have significantly positive impact on the intensity of eco-innovation. While, being medium size firm, national market conditions, funds provided by the EU and intensity of R&D efforts seem to hinder firms' eco-innovation intensity. As a result, we obtain two distinct estimation results from two distinct models. Our benchmark model consolidates our previous estimation results while our country fixed-effect model tends to yield significantly different results. We suspect that specification of the models could lead such results.

	With a country	fixed-effect	With an exclus	ion variable
	(I) Selection equation	(II) Outcome (intensity) equation	(III) Selection equation	(IV) Outcome (intensity) equation
Constant term	0.23***		0.94***	
	(0.043)	-	(0.037)	-
<u>Firm size</u>				
Medium	0.17***	0.001	0.11***	-0.078***
	(0.015)	(0.017)	(0.014)	(0.017)
Large	0.2***	0.13***	0.11***	-0.017
	(0.021)	(0.022)	(0.021)	(0.021)
<u>Characteristics</u>	0.1***	0.042**	0.000	0.016
Being part of a Group	(0.016)	-0.043** (0.016)	-0.009 (0.016)	-0.016 (0.015)
Active in National market	0.24***	-0.04*	0.26***	-0.084***
Active in National market	(0.014)	(0.019)	(0.013)	(0.017)
Active in Inter. Market	0.28***	0.02	0.265***	0.047**
	(0.015)	(0.018)	(0.014)	(0.017)
Funded by EU	0.53***	-0.05	0.612***	-0.081**
5	(0.072)	(0.03)	(0.069)	(0.03)
Engagement in Continuous	1.17***	0.2	1.087***	0.16***
R&D	(0.029)	(0.03)	(0.028)	(0.029)
R&D intensity	1.99***	0.044	1.75***	-0.241**
	(0.31)	(0.089)	(0.284)	(0.083)
Openness	-0.29***	0.18***	-0.29***	0.137***
	(0.004)	(0.007)	(0.004)	(0.006)
Scientific cooperation	0.014	0.02	-0.033	0.025
	(0.054)	(0.023)	(0.051)	(0.022)
Cooperation with others	1.53***	-0.033	1.63***	0.186***
Evolution voriable	(0.037)	(0.033)	(0.034)	(0.026)
Exclusion variable Soviet			0.242***	
500100	-	-	(0.013)	
Industry dummies (chi ²	1660.79***		1706.23***	
test)	1000.77		1700.25	
Country Dummies (chi ²	67252.82***		_	
test)	07232.02			
No of observations	65058		65058	
Chi ² (df)	122495.55		13927.95 (79)***	
()	(106)***			
Cut-off points	()			
Cut-off - 1		1.02***		0.81***
		(0.066)		(0.036)
Cut-off – 2		1.77***		1.5***
		(0.068)		(0.037)
Cut-off - 3		2.71***		2.37***
		(0.07)		(0.038)

Table B.3. Ordered Response Models with Sample Selection (EMAT Proxy)

-Associated standard errors are in parenthesis below each coefficient - ***significant at the 1% level, **significant at the 1% level

B.4. Country specific analyses on EMAT

The country specific analyses aim at displaying the differences among country groups. The results show that we indeed achieved to show profiles of successful eco-innovators within different country groups.

When we inspect the estimation results, we can note that there are significant differences inherent in the relative importance of firm characteristics for each country groups. The comparison of Western European and Mediterranean countries indicate that the Western European eco-innovators tend to be bigger in size and moreover, they also engage in continuous R&D efforts. Mediterranean countries, however, give greater importance to the R&D intensity rather than the continuity of R&D efforts. The openness measure, on the other hand, is the only factor commonly shared between these country groups. Therefore, the significance of the openness measure for the eco-innovators is again consolidated with these new estimation results.

	Western	Europe	Mediterr	anean
	(I) Selection equation	(II) Outcome (intensity) equation	(I) Selection equation	(II) Outcome (intensity) equation
Constant term	-1.42***		1.74***	
	(0.078)	-	(0.085)	-
Firm size				
Medium	0.22***	0.011	0.057*	-0.01
	(0.033)	(0.042)	(0.027)	(0.028)
Large	-0.26***	0.268***	0.03	0.054
C .	(0.052)	(0.052)	(0.042)	(0.038)
Characteristics				
Being part of a Group	0.034	0.006	0.056	-0.026
	(0.031)	(0.038)	(0.03)	(0.028)
Active in National market	0.13***	-0.13**	0.205***	-0.051
	(0.036)	(0.047)	(0.029)	(0.042)
Active in Inter. Market	0.26***	0.06	0.234***	-0.003
	(0.033)	(0.042)	(0.027)	(0.032)
Funded by EU	-0.24*	0.108	0.472***	-0.054
•	(0.104)	(0.066)	(0.115)	(0.049)
Engagement in Continuous R&D	0.845***	0.124*	1.219***	0.086
	(0.043)	(0.051)	(0.049)	(0.068)
R&D intensity	1.795***	-0.011	0.596	0.328*
	(0.453)	(0.21)	(0.393)	(0.138)
Openness	0.039***	0.138***	-0.404***	0.124***
a	(0.007)	(0.013)	(0.01)	(0.018)
Scientific cooperation	-0.017	-0.036	0.105	-0.054
	(0.076)	(0.052)	(0.109)	(0.042)
Cooperation with others	1.118***	-0.043	1.588***	0.086
	(0.061)	(0.06)	(0.073)	(0.063)
Industry dummies (chi ² test)	644.5***		463.52***	
Country Dummies (chi ² test)	146.76***		657.81***	
No of observations	10194		20977	
Chi ² (df)	277.36(76)***		4187.3(82)***	
Cut-off points	277.50(70)		(10).5(02)	
Cut-off - 1		0.87***		0.65***
		(0.171)		(0.065)
Cut-off – 2		1.81***		1.4***
		(0.174)		(0.066)
Cut-off - 3		2.72***		2.36***
		(0.178)		(0.067)

Table B.4.a. Ordered Response Models with Sample Selection (EMAT Proxy) by group of countries

-Associated standard errors are in parenthesis below each coefficient - ***significant at the 1% level, **significant at the 1% level, *significant at the 1% level

	Eastern I	Europe	Baltie	c
	(I) Selection equation	(II) Outcome (intensity) equation	(I) Selection equation	(II) Outcome (intensity) equation
Constant term	5.4***		2.871***	
	(0.261)	-	(0.247)	-
Firm size				
Medium	0.252***	0.032	0.112	-0.066
	(0.026)	(0.028)	(0.085)	(0.063)
Large	0.483***	0.152***	0.458***	0.071
-	(0.033)	(0.035)	(0.113)	(0.107)
Characteristics				
Being part of a Group	0.145***	-0.068*	0.255**	-0.062
	(0.031)	(0.027)	(0.097)	(0.064)
Active in National market	0.325***	0.023	0.41**	0.083
	(0.024)	(0.028)	(0.151)	(0.083)
Active in Inter. Market	0.349***	0.103***	0.404***	0.129
	(0.028)	(0.029)	(0.092)	(0.079)
Funded by EU	1.073***	-0.099*	0.63	0.02
F (1	(0.212)	(0.05)	(0.419)	(0.122)
Engagement in	1.602*** (0.115)	0.041 (0.034)	0.95*** (0.226)	0.117
Continuous R&D	. ,	× ,	. ,	(0.088)
R&D intensity	4.668***	-0.152	-2.204	0.32
0	(1.377)	(0.145)	(1.297)	(0.444)
Openness	-0.709***	0.22***	-0.322***	0.161***
Scientific cooperation	(0.03) 0.15	(0.009) 0.078*	(0.025) -0.086	(0.029) 0.054
Scientific cooperation	(0.17)	(0.036)	(0.256)	(0.082)
Cooperation with others	1.767***	-0.019	1.692***	0.073
cooperation with others	(0.107)	(0.039)	(0.177)	(0.127)
Industry dummies (chi ²	750.13***		675.12***	
test)				
Country Dummies (chi ²	26019.42***		340.42***	
test)				
No of observations	31032		2855	
Chi ² (df)	42958.02(88)***		1622.62(80)***	
Cut-off points				
Cut-off - 1		1.04***		0.86***
		(0.06)		(0.135)
Cut-off – 2		1.74***		1.51***
		(0.062)		(0.139)
Cut-off - 3		2.68***		2.51***
		(0.065)		(0.146)

Table B.4.b. Ordered Response Models with Sample Selection with EMAT Proxy by group of countries

-Associated standard errors are in parenthesis below each coefficient

- *** significant at the 1% level, ** significant at the 1% level, * significant at the 1% level

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A Study on the Types of Eco-innovators Using Micro-Anonymised Data:

The case of eco-innovators and eco-adopters

Table of Contents

1. Introduction
2. Literature Review
2.1. Definition of Eco-innovation164
2.2. What Drives Eco-innovation
3. Econometric Analysis
3.1. The Data
3.2. The Estimation Strategy and the Econometric Model
3.3. A Typology of Eco-innovators
3.4. Choice of Explanatory Variables
3.5. Descriptive Statistics
4. The Empirical Results
4.1. The Results of Estimations Conducted on the Whole Sample
4.2. The Country Specific Analyses
4.2.1. The Results of Estimations Conducted by Group of Countries
5. Conclusion

A Study on the Types of Eco-innovators Using Micro-Anonymised Data

Abstract

Despite its accelerating importance, the eco-innovation literature lacks consistent empirical evidence reflecting the role of firm characteristics and capabilities within the eco-innovation process. Furthermore, there has been little or no attempt, in empirical studies, to distinguish true eco-innovators from mere eco-adopters. This research aims at filling this gap by examining the influence of a firm's economic, technological, organisational and social capabilities on the eco-innovation behaviour for different types of eco-innovators. It exploits micro-aggregated data drawn from the 2008 wave of the Community Innovation Survey, which includes a special module on eco-innovation. Our results stress the importance of conventional R&D inputs, firms' formal cooperative behaviour and firms' openness. Moreover, we find evidence that firms' voluntary engagement in eco-innovation is as important as the environmental regulations for early strategic eco-innovators, while financial funding from the EU seem to spurs eco-innovativeness among late strategic eco-innovators and strategic eco-adopters. Country-specific analyses indicate that the role of cooperative behaviour may be more important in some groups of countries than in others, whereas the significance of environmental regulations and voluntary engagement in eco-innovation is established in each country group.

Keywords: Eco-innovation, micro-aggregated data, CIS 2008, Firm capabilities, cooperative behaviour, Environmental regulations

JEL codes: **D2**- Production and Organizations, **L2**- Firm Objectives, Organization, and Behaviour, **Q5**-Environmental Economics, **O3**- Technological Change

1. Introduction

Environmentally friendly innovation, green innovation, sustainable innovation, and environmentally benign innovation are terms commonly used for a specific type of innovation, labelled in the innovation literature as "eco-innovation". The specificity of ecoinnovation lies in the fact that it makes environmental concerns, in addition to technological advancement, its prime objective. By combining these two objectives (that may appear mutually exclusive), eco-innovation has become one of primary tools in the search to solve the world's environmental problems and sustainability challenges. The EU's Lisbon strategy (2000) and the Gothenburg Summit (2001) are the most visible examples of the macro-scale acknowledgement of its significance. These strategies give considerable importance to the introduction and development of eco-innovation: according to the Lisbon Strategy it is supposed to play a key role in making the EU's economy one of "the most competitive and dynamic knowledge-based economy in the world capable of sustainable economic growth with more and better jobs and greater social cohesion" (the Lisbon Agenda, 2000).

As an evolutionary mechanism, the conventional innovation process requires a steady change even as it becomes more complex, interdependent and difficult to coordinate. Because of the double externality problem highlighted by Rennings (2000), the eco-innovation process tends to be even more complex and is affected by even more factors, asserting that sustainability cannot be achieved through policies only but requires profound changes in the strategic thinking of firms. Hence, to become successful eco-innovators and in order to survive within a competitive market, firms need to be more systematic in their strategic behaviour towards innovation, well-informed on the environmental regulatory framework and able to capture existing and future opportunities. If elaborated within this context, a firm's economic, technological, organisational and social capabilities appear to be significant components of the eco-innovative process. However, due to its recent emergence in innovation research and to the scarcity of relevant quantitative data, there is still much debate and unexplored issues with regard to the concept of eco-innovation.

The exploration of the drivers of and barriers to eco-innovation is one of the most debated issues in the eco-innovation literature and it must be approached with caution. Most empirical studies address the drivers and barriers from the conventional market-pull, technology-push, and institutional (regulatory) factors perspective (Porter and Van der Linde, 1995; Rennings, 2000; Brunnermeier and Cohen, 2003; Jaffe et al., 2002). A few studies highlight the importance of other dynamic factors, such as firms' economic, technical,

managerial and social capabilities (Horbach, 2008; Borghesi et al., 2012; De Marchi, 2012). A significant part of these empirical studies are focused on specific geographical areas, industry segments or types of eco-innovation (Horbach, 2008). Additionally, Kemp and Pontoglio (2011, p. 34) claim that a significant amount of econometric research has focused on eco-innovation that is new to the world, disregarding the potential differences in characteristics and capabilities among eco-innovators and eco-adopters.

The present research aims at overcoming (some of) these limitations. It contributes to the discussion about the drivers of eco-innovation by introducing a distinction between eco-innovators and eco-adopters on the basis of firms' strategic behaviour and the ways in which they innovate. In order to do so, we adapted Kemp and Pearson's (2007, p. 18) approach in defining different types of eco-innovators. Accordingly, within this research, eco-innovators are classified as "early" and "late" *strategic eco-innovators*, "early" and "late" strategic *eco-adopters* and "passive" *eco-adopters*. We thus contribute to the literature by proposing an empirically relevant typology of eco-innovators and by identifying the technical and organisational characteristics and capabilities that are specific to each type of eco-innovators. Our research is one of the first quantitative empirical studies to classify eco-innovators according to their strategic behaviour and to identify type-specific drivers of eco-innovation.

Our econometric analysis relies on micro-aggregated data drawn from the 2008 wave of the Community Innovation Survey (CIS 2008), which includes a special module on ecoinnovation. The data covers 79972 firms observed between 2006 and 2008 across 11 EU member states, and allows us to control for national and sectoral differences in the factors that are relevant for innovation and performance. It also allows us to separate and categorise eco-innovators and to distinguish the firm-specific characteristics of each type of eco-innovator.

The remainder of this paper is organised as follows. Section 2 provides a definition of eco-innovation and reviews its possible drivers. Section 3 presents our data, our typology of eco-innovators, our estimation strategy and choice of variables. Section 4 presents the results of our global and country-specific analyses. We draw a general conclusion in Section 5.

2. Literature Review

2.1. Definition of Eco-innovation

Due to its increasing relevance, the concept of eco-innovation has attracted the attention of many. Because the concept of eco-innovation is rather novel, the innovation literature is abundant with various definitions. In this research we adopt one of the most widely used definitions, proposed by the EU's Measuring Eco-innovation project (MEI). According to this definition, eco-innovation is:

"assimilation or exploitation of a product, production process, service or management or business method that is novel to the firm or user and which results, throughout its life cycle, in a reduction of environmental risk, pollution and other negative impacts of resource use (including energy use) compared to relevant alternatives" (MEI, Kemp and Pearson (2007, p. 3).

Upon completion of the MEI project in 2008, the concept of eco-innovation has acquired a solid theoretical base that helps and guides researchers from various fields of study. According to this background, "any company adopting a good, service, production process, management or business method with environmental benefit is considered an eco-innovator" (Kemp and Pearson, 2007, p. 17). This definition is somehow general and tends to address various types of eco-innovators. Therefore, there are also other definitions addressing specific type of eco-innovators. For example, the OECD (2009) refers to organisational ecoinnovation as "incremental or radical change of a firm's processes and responsibilities, which reduces environmental impact and supports organisational learning". Even though, this definition is rather direct, it does not specifically address the adopters of environmental management standards (such as; EMAS and ISO 14001-type standards). Therefore, there is no consensus on whether the adopter's environmental management standards should be considered as organisational eco-innovator. De Marchi (2012) stated the general definition of eco-innovation is based on the effect of the innovation activities independently of their initial intent, and includes both incremental and radical improvements. Hence, the overall environmental impact of an innovation, rather than its original intention, is the point of interest within this definition. This broad definition intrinsically brings in certain complications while categorizing different types of eco-innovators. It is efficient enough to separate eco-innovators from non-eco-innovators but it is too fuzzy to distinguish real ecoinnovators from mere eco-adopters. As a result, most empirical studies have used the general categorization of eco-innovation, neglecting the potential differences between real ecoinnovators and eco-adopters. We have concluded that at the best case scenario, these standards may help organisations to become aware of their environmental responsibilities and lower their carbon foot-print through changes in their organisational structure. Therefore, for

our empirical analyses the adopters of these standards are considered as organisational-ecoinnovator.

2.2. What Drives Eco-innovation

The eco-innovation literature points out to three external settings as the primary drivers of eco-innovation. Amongst them, conventional market-pull and technology push factors are relevant to the technical aspects of eco-innovation while regulation (both in the form of command and control and market-based instruments) is relevant to the environmental aspects of eco-innovation. The double externality problem highlighted by Rennings (2000) is characteristic of eco-innovation that makes environmental regulation a necessity if ecoinnovation is to occur at all. More precisely, the expected positive environmental impacts of eco-innovation make it desirable, and policy maker encourage firms to take actions and be a part of the solution to environmental issues. However, the negative externalities inherent in the innovation production phase, (through the usual knowledge spill overs) make ecoinnovation rather costly, with a financial return that is even more uncertain than that conventional innovation. When firms faced with such options, they may show a tendency to shy away from eco-innovation as the competitive market conditions may require them to do so. In their seminal work, Porter and Van Der Linde (1995) have stressed the significance of strict environmental regulation as a mean to channel firms' innovative efforts towards more sustainability. They also claim that productivity gains and competitive advantages accrued from eco-innovation (through reduced material and energy use and patenting and learning effects) could offset the costs of innovation, leading to a "win-win" situation. This claim is also known as the Porter Hypothesis. Palmer et al. (1995) have criticized Porter and Van Der Linde's statements on the grounds that they are supported only by case studies and hence, provide insufficient evidence for generalization. Since then, several studies have been conducted on this issue and the literature still lacks solid empirical evidence to support or reject the Porter Hypothesis (see Wagner, 2003 for a detailed review). On the one hand, in their empirical investigation van Leeuwen and Mohnen (2013) have found a supporting evidence for the weak version of the Porter hypothesis hence, they concluded that environmental considerations (i.e., government regulations, market pressures) seem to be an important factor in the decision making of firms to invest in R&D and eco-investments and they also play a crucial role in the introduction of different types of eco-innovations. However, they did not observe any direct impact of environmental regulations on Total Factor Productivity (TFP) and on firms' decision to invest in eco R&D. Therefore, they

couldn't approve or disapprove the strong version of the hypothesis. On the other, Ozusaglam and Robin (2013) have concluded that government intervention through environmental regulations is indeed one of the most important factors leading firms to lower their hazardous impact on environment. However, they also show that voluntary environmental standards i.e. ISO 14001 and EMAS may be a good complement for market-based instruments (such as; tradable permits, emission taxes, subsidies etc.) and, command and control mechanisms to increase firms' economic performance through Total Value Added. Given that these voluntary environmental standards help firms to reduce their carbon foot-print, their statement supports the weak version of the Porter hypothesis.

As can be seen, there is no consensus on the overall validity of the Porter hypothesis which can be largely attributed to problems inherent in measuring eco-innovation, the lack of available quantitative data and the stringency measures used in the analyses. Mulatu et al. (2001) stated that environmental regulation-competitiveness linkage should be investigated with data on the industrial level and with better stringency measures. These problems have led researchers to analyse the drivers of and barriers to eco-innovation issue from the conventional innovation perspective. This resulted in underestimating the importance of firmspecific characteristics and capabilities within the eco-innovation process. However, there are rather few investigating the role of more dynamic factors such as firms specific characteristics, technical, organisational and social capabilities within the eco-innovation process. Amongst them, Kemp and Pearson (2007) asserted that firms' propensity to ecoinnovate is positively related to their ability to build on organisational capabilities. De Marchi (2012) emphasized the role of engagement in continuous R&D activities and cooperation activities as a driver of eco-innovation. This author found, though, (as did Horbach, 2008) that firms' R&D intensity was not a significant driver of the eco-innovation process. In addition, Kesidou and Demirel (2012) and Horbach (2008) claim that the adoption of Environmental Management Systems (EMS), as an indicator of a firm's organisational capabilities, may indeed increase the firm's eco-innovation intensity. Ozusaglam and Robin (2013) postulate that firms that fall in "high-technology and high-medium manufacturing" level are more prone to adopt voluntary environmental standards and thus they might be an efficient choice for making firms more competitive and greener. Nevertheless, a through empirical research on the role of a firm's internal characteristics and technical, organisational and social capabilities in the eco-innovation process is rather scarce. This research article aims to investigate the respective role of these characteristics and capabilities by conducting in-depth empirical analyses.

3. Econometric Analysis

3.1. The Data

The relative lack of quantitative data appears to be one of the most significant obstacles to empirical studies in the field of eco-innovation. The problem can be associated with the recent appearance of the concept of eco-innovation and the difficulties involved in measuring it. Nevertheless, certain improvements have taken place within the last few years and the 2008 wave of the Community Innovation Survey (CIS 2008) includes a special module on eco-innovation. The CIS is coordinated throughout Europe by Eurostat in cooperation with the statistical offices of the Member States. It relies on a harmonised questionnaire based on the Oslo Manual (OECD, 1997), which provides invaluable information on the characteristics of innovation activities at the firm level (see Eurostat, 2005). The CIS data is available either as micro (firm-level) data for a given country (from each country's national statistical office), or as micro-aggregated data for several EU Member States (from Eurostat).

Micro-aggregated data is firm-level data which has been made anonymous by adding random "error terms" to the raw firm-level data. This method makes it almost impossible to break the anonymity of individual firms, while allowing researchers to conduct econometric analyses (which will, by their very nature, take out the added noise). This process is explained in detail by Mohnen, Mairesse and Dagenais (2006). Experiments on specific countries have shown that econometric analyses conducted on the micro-aggregated data yield results that are not significantly different from those of analyses conducted on the corresponding firm-level data (Mohnen, Mairesse, and Dagenais, 2006). Moreover, all enterprises included in the data set conform with the minimum coverage (10 employees or more). The data provides information about (among others) main economic activity, location, number of employees, turnover, expenditure in intramural and extramural research and development (R&D), innovation output, innovation collaboration, knowledge sources, primary objective of environmental innovations and environmental benefits from innovation. In its original design, the survey does not aim at extracting information directly related to eco-innovation. However, CIS 2008 includes a special module on eco-innovation which lists a series of environmental benefits for product, process, organisational or marketing innovations introduced during the observation period (from 2006 to 2008). This module makes it an adequate dataset for our purposes.

3.2. The Estimation Strategy and the Econometric Model

Our empirical analysis aims at identifying a firm's characteristics and capabilities that lead a firm towards a more active and strategic eco-innovation behaviour. The analysis relies on the innovativity approach first proposed by Mairesse and Mohnen (2002). Within the context of this research, this approach consists in estimating an innovation production function specified as an Ordered Probit model with selection. Using some sort of selection model is necessary to address the selection issues that arise with CIS 2008 (as with all waves of the CIS). A potential selection bias arises because of the presence of "filtering" questions in the CIS questionnaire. These questions are designed to characterize firms as "innovating" or "non-innovating", and only the former have to fill in the full questionnaire (see for instance Mohnen and Mairesse (2002) or Mairesse, Mohnen and Dagenais (2006) for more detailed explanations). This attribute of CIS surveys can endanger the reliability of the empirical analysis if one does not take into account the selection problem.

The econometric model used in this research contains a selection equation (selection into conventional product, process, organizational or marketing innovation) which is specified as a Probit model, as is usually done in the literature. The outcome equation (intensity equation) of the model is specified as an Ordered Probit model rather than as a linear equation. We adapted this model since there is a natural ranking in our typology; the different types of eco-innovators are ranked from the less eco-innovative (passive eco-adopters) to the more eco-innovative (early strategic eco-innovators). Therefore, the Ordered Probit will give the effect of each explanatory variable on eco-innovation "as a whole", with a different "cut-off point" for each type of eco-innovator while non-eco-innovators being taken as the category of reference. This model is estimated using the Conditional Mixed Process estimator (CMP) programme that runs with Stata econometric software (Roodman, 2009).

3.3. A Typology of Eco-innovators

The eco-innovation literature lacks consistent empirical research on the eco-innovation behaviour of different types of eco-innovators and the role of firm specific characteristics as drivers to eco-innovate for each type of eco-innovator. We aim at filling this gap with this research. To do so, we adapted and refined the categorization of eco-innovators proposed by Kemp and Pearson (2007). Their categorization is as follows:

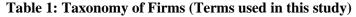
Strategic eco-innovators: active in eco-equipment and service sectors, develop eco-innovations for sale to other firms.

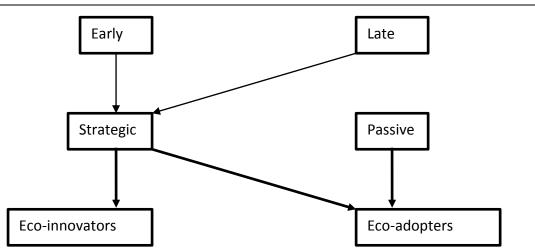
Strategic eco-adopters: intentionally implement eco-innovations, either developed inhouse, acquired from other firms, or both.

Passive eco-innovators: process, organisational, product innovations etc. that result in environmental benefits, but where there is no specific strategy to eco-innovate.

Non eco-innovators: no activities for either intentional or unintended innovations with environmental benefits.

In our approach, eco-innovators are classified as "early" and "late" *strategic eco-innovators*, "early" and "late" *strategic eco-adopters*, "passive" *eco-adopters* and *non-eco-innovators*. Table 1 displays the taxonomy used for different type of eco-innovators.





Our proposed definitions allow us to distinguish the different types of eco-innovators alongside two dimensions: (1) the **environmental impacts** of the realised product, process, organisational or marketing innovation are used to determine qualification as an eco-innovator; (2) the **way the eco-innovations are implemented** is used to separate "true" strategic eco-innovators from strategic eco-adopters. Regarding the first dimension, CIS 2008 lists the direct environmental impacts of the realised product, process, organisational or marketing innovation within the enterprise as follows: (i) Reduced material use per unit of output, (ii) Reduced energy use per unit of output, (iii) Reduced CO₂ 'footprint' (total CO₂ production) by the enterprise, (iv) Replaced materials with less polluting or hazardous substitutes, (v) Reduced soil, water, noise, or air pollution and (vi) Recycled waste, water, or materials. In addition to the direct environmental benefits within the enterprise, the direct environmental benefits after sale, by the end-user such as (i) Reduced energy use, (ii)

Reduced air, water, soil or noise pollution and (iii) Improved recycling of product after use, are also included in the survey. These environmental impact indicators are utilised in order to separate eco-innovators from non-eco-innovators.

After introducing this separation, we disentangle eco-innovators according to their strategic behaviour towards eco-innovation (that is to say, how they innovate). We rely on the adoption date of voluntary environmental procedures at the firm level (e.g., environmental audits, the setting environmental performance goals, ISO 14001 certification) to differentiate firms that are at the leading edge of eco-innovation and laggards. Indeed, adopting such procedures amounts to implementing an Environmental Management Standard (EMS). Many researchers (Rennings *et al.*, 2006; Rehfeld *et al.* 2007; Wagner, 2008; Khanna *et al.*, 2009) have stated that the implementation of an EMS is a relevant form of eco-innovation, because EMS are voluntary, proactive approaches that reflect the environmental awareness of firms. Horbach *et al.* (2012) have asserted that the EMS help overcome problems related to incomplete information as well as organizational and coordination problems within a firm, and that they are also important for the introduction of cost-saving cleaner technologies.

In practice, we use a question on the adoption of the above-mentioned procedures that is featured in the module dedicated to environmental innovation of the CIS 2008 survey. For firms that have adopted such procedures (or, in other words, that have implemented an EMS), the question let us know whether the EMS was implemented before January 2006 (i.e., before the start of CIS 2008) or after January 2006 (i.e., during the time period covered by the CIS 2008). Utilizing the year 2006 as the separation measure can raise some question as this year might be perceived as rather random. However, we assert that by doing so we are able to determine whether an eco-innovator as defined alongside the first dimension is *relatively* early or late in implementing an EMS, indicating their experience in managing the EMS and their relatively higher environmental consciousness. By combining the two above-mentioned dimensions (the introduction of an innovation with an environmental impact and the date of implementation of an EMS), we can assign all innovative firms to one of five mutually exclusive categories.

A **passive eco-adopter** is defined as an innovator that introduces a product, process, organisational or marketing innovation that results in environmental benefits (eco-innovation), even though the firm has no specific strategy to eco-innovate (no established environmental management standards, norms, procedures or regulations within the enterprise). Therefore, this group represents a set of eco-innovators in which the firms show

no sign of commitment to solve environmental problems and moreover, have no overt intention or strategic plan to eco-innovate. The introduction of eco-innovation could be a mere coincidence and/or the positive environmental impacts could be side effects of their conventional innovations. As can be expected this category comprises the majority of the population and includes a heterogeneous sub-populations from various technological levels (from low-tech manufacturing to high-tech manufacturing). Accordingly, a passive ecoadopter might be a high-tech manufacturing firm engaging in conventional i.e., product or process innovation, where the firm has no intention or strategic plan to eco-innovate.

A late strategic eco-adopter represents an innovator that established environmental management standards (EMS), norms, procedures or regulations in place to regularly identify and reduce the enterprise's environmental impacts "after 2006" but did not introduce any eco-innovation. This type of eco-adopter is labelled as "late" due to fact that they adopt the EMS's at a relatively later date. Nevertheless, even if the adoption took place after the base year (2006), it can still signal an on-going change in their commitment to reduce their ecological foot-print.

An **early strategic eco-adopter** is defined as an innovator that has the same environmental management standards, norms, procedures or regulations in place "before and after 2006" but did not introduce any eco-innovation. Firms within this group indeed show a relatively higher degree of environmental commitment than the aforementioned groups. Adopting the EMS's before the base year may point out the existence of an established longterm environmental awareness and experience that may offer significant advantages in the effort to tackle environmental problems. However, their proactive approach is not sufficient to develop and/or introduce a successful eco-innovation.

A late strategic eco-innovator is an eco-innovator that adopted environmental management standards, norms, procedures or regulations "after 2006" and succeeded in introducing an eco-innovation during the observation period. Therefore, firms within this group show signs of significant efforts to reduce their ecological foot-print. Moreover, they tend to embed their environmental concerns in their day to day interactions by adopting EMS's even if the adoption took place relatively later than that of other firms.

Last but not least, an **early strategic eco-innovator** is an eco-innovator that established the environmental management standards, norms, procedures or regulations "before 2006" and succeeded in introducing an eco-innovation during the observation period. This group of firms is significantly more strategic than the rest. They tend to have a longer history of environmental commitment and are more experienced in the application of environmental management standards. Therefore, identifying the characteristics and capabilities of this type of eco-innovator is crucial for this research.

In order to prevent potential confusion, we must indicate here that throughout this research, we will use the word "eco-innovator" to represent all of these five sub-groups, while their respective titles will be used when differentiation is necessary.

3.4. Choice of Explanatory Variables

The eco-innovation literature highlights the significance of firm size as one of the most important firm characteristics and it is included as our first control variable. It is generally assumed that bigger firms tend to have easier access to financial resources to invest in innovation inputs (del Rio Gonzalez, 2005, 2009; Kammerer, 2009) such as R&D related activities, acquisition of knowledge and equipment etc. In the micro-aggregated CIS 2008 survey data, the indicator of firm size is an ordered variable comprising three categories ("small" "medium" or "large") depending on the number of employees (where small is the category of comparison).

Besides a firm's size, being part of a larger group (being a subsidiary) can also be considered another significant firm characteristic that can affect firm's propensity to innovate. Various studies (i.e., Jeppesen and Hansen, 2004; Cainelli *et al.*, 2011) have shown that being part of a group can be a significant opportunity for firms to learn about existing environmental standards and other ways to reduce their carbon foot-print. In the best-case scenario, a subsidiary may need to spend less effort to access the financial resources, knowledge and technical support needed for successful innovation than individual firms. However, dependence on the parent company's decisions may also have an adverse effect on a firm's innovative efforts. Being strictly tied to the parent company's innovation strategy may restrict the creative process of an otherwise innovative firm. We will attempt to test these assumptions by including a binary indicator of group ownership as the second control variable.

As was previously mentioned, one of the significant advantages of the CIS survey is that it covers multiple EU member states and allows us to control for national and sectoral differences in the factors that are relevant for innovation and performance. Therefore, we aim to use this aspect of the survey to our advantage by including both manufacturing and service sectors in the regressions. Accordingly, 2-digit industry affiliation dummies (Nace codes) included in our regressions. In the same vein, in order to capture the technological conditions as regard to eco-innovation among different we also included country dummies in the regressions.

Besides firm specific characteristics, a firm's contextual surroundings can also shape its innovative behaviour; we can consider the market conditions as the prime example. Maturity of the markets, existing technological opportunities, sufficient market demand, consumer awareness and existence of regulatory framework are the other examples of the market characteristics. Accordingly, we included two dummy variables indicating whether a firm is active in "national" or "international" market. By including these dummy variables we aimed at controlling for the impact of different markets on the introduction of eco-innovation.

In addition to abovementioned control variables, we also acknowledge R&D efforts (engagement in continuous R&D and R&D intensity) as "classic" inputs in the innovation production function. For example, a continuous engagement in R&D is generally considered as one of the conventional innovation inputs that not only can enhance a firm's scientific and technological capabilities but increases a firm's absorptive capacity needed to undertake successful innovation (Hemmelskamp, 1997; Horbach, 2008). Through a firm-level analysis Rennings et al. (2006) claimed that the internal R&D efforts could be of greater importance for eco-innovation than for conventional innovation. Moreover, a firm's "R&D intensity" can help identify the importance the firm gives to the development of future products and processes. Successful outcomes of these efforts can be considered innovation, and this success can augment the intensity of R&D in return. Interestingly, Hemmelskamp (1999) claims that eco-innovative companies are likely to have a lower R&D intensity as they tend to rely more on cooperation and external sources information. Since in CIS surveys, some firms do not engage in R&D on a continuous basis (and some firms do not engage in R&D at all), it is usual to use both variables as regressors in econometric models of the innovation production function (see for instance Mairesse and Mohnen, 2002). The continuous engagement in R&D is captured by a binary variable equal to 1 in case of continuous R&D and to 0 otherwise while R&D intensity is classically measured by the share of (intramural and extramural) R&D expenditures in total sales.

The innovation production is a long and a challenging practice characterized as a highly complex process that requires increased interdependencies and information/knowledge exchange among incorporating partners. This complexity makes the absorption of external knowledge an invaluable factor within the innovation process (Cohen and Levinthal, 1989). However, the significance of the absorptive capacity through external knowledge is stressed mainly for the conventional product and process innovation (Zahra and George, 2002) while little has been made with respect to eco-innovation. In order to observe the significance of interdependencies between involved partners and the importance of the information that they convey, we introduced three mutually exclusive variables namely: (i) Openness, (ii) Scientific cooperation and (iii) Other cooperation.

According to Laursen and Salter (2004, 2006), we can measure the degree of "openness" of a firm by looking at the use of general internal and external sources of information within the innovative process. Following their methodology, we recode the "sources of information" variables provided by CIS 2008 as binary variables equal to 1 if a source is used and to 0 if it is not. These variables are then summed up to provide the ordered variable, in which a greater value indicates a larger degree of openness. In our application, we use the following information sources²⁸: (1) sources from within the enterprise or its group, (2) suppliers of equipment or materials, (3) clients or customers, (4) competitors and other enterprises in the same industry, (5) consultants, commercial labs or private R&D institutes, (6) professional conferences, trade fairs, meetings, (7) scientific journals, trade/scientific publications, and (8) professional and industry associations. According to this method, the greatest possible openness score (which is attained if a firm uses all eight sources of information) is equal to 8, while the lowest possible score is 0.

Openness in the above sense does not entail formal cooperation with the sources of information used (for example, a firm might obtain information from a competitor without formally cooperating in R&D with this competitor). Therefore, the second variable, labelled "scientific cooperation", controls for formal cooperation with scientific partners as is discussed by Robin and Schubert (2013) (i.e., cooperation with universities and higher education institutions or with government/public research institutes). This may be distinct from openness, especially when academic knowledge is involved (see Footnote 2).

The third variable, labelled "other cooperation", is an indicator of cooperation with other partners, which captures interactions between firms and non-scientific cooperation partners. Cooperating with other partners entails a formal cooperation between a firm and at least one of the following: (i) other enterprises within the group, (ii) suppliers of equipment, materials,

²⁸ As in Laursen and Salter (2004) and Robin and Schubert (2013), we do not include academic sources of information because most firms report no (or a very low) use of these sources. This is a recurrent feature of CIS data, which may be due to the fact that relying on academic knowledge to innovate generally implies some measure of formal cooperation. Thus, the use of academic knowledge is better captured by other variables, such as indicators of cooperation, as will be done here.

components, or software, (iii) clients or customers, (iv) competitors or other enterprises in the firm's industry and (v) consultants, commercial labs, or private R&D institutes.

In addition, (i) available financial funds, (ii) regulatory framework and (iii) voluntary efforts are also considered exogenous factors that can shape a firm's eco-innovative behaviour. Therefore, three binary variables included in the analyses. The variable "fund" indicates whether a firm received any public financial support for innovation activities from the EU's 6th and 7th Framework Programme. When considering the second component of the 'double externality problem', namely externality through the usual knowledge spill overs, firms may need incentives (such as financial help, credits, tax reduction etc.) to engage in eco-innovation. Absence of these incentives may cause some firms to shy away from engaging in such an activity. Therefore, by including this dummy variable we aim to test the significance of the EU's Framework Programmes.

In the same vein, the "regulation" variable indicates whether a firm introduced an environmental innovation in response to 'existing environmental regulations or taxes on pollution' while the "voluntary" variable identifies whether 'the voluntary codes or agreements for environmental good practice' are the main motivation to introduce an eco-innovation. Table 2, below, presents the summary statistics on the main variables used in the econometric analysis while Table 3 displays the summary statistics on the variables used in the empirical estimation for each type of eco-innovator.

Variables	Mean	Std. dev.	Min	Max
Dependent variables:				
Firm introduced a product innovation (yes/no)	0.22	0.41	0	1
Firm introduced a process innovation (yes/no)	0.25	0.43	0	1
Firm introduced an organisational innovation	0.26	0.44	0	1
(yes/no)				
Firm introduced an marketing innovation (yes/no)	0.21	0.41	0	1
Firm is an innovator (yes/no)	0.43	0.49	0	1
Firm is a passive eco-adopter (yes/no)	0.12	0.33	0	1
Firm is a late strat. eco-adopter (yes/no)	0.042	0.2	0	1
Firm is an early strat. eco-adopter (yes/no)	0.092	0.29	0	1
Firm is a late strat. eco-innovator (yes/no)	0.063	0.24	0	1
Firm is an early strat. eco-innovator (yes/no)	0.069	0.25	0	1
Firm is an eco-innovator (yes/no)	0.26	0.43	0	1
Explanatory variables:				
Size 2006 ^a	0.46	0.66	0	2
Size 2008 ^a	0.48	0.66	0	2
Group (yes/no)	0.21	0.41	0	1
R&D intensity (Total R&D exp. / Total turnover	584.8	97251.6	0	1.73e+07
Continuous R&D (yes/no)	0.07	0.26	0	1
Openness (0-8) ^b	6.47	2.49	0	8
Scientific Cooperation (yes/no)	0.048	0.21	0	1
Other Cooperation (yes/no)	0.11	0.31	0	1
International market (yes/no)	0.39	0.49	0	1
National market (yes/no)	0.68	0.46	0	1
Local market (yes/no)	0.67	0.46	0	1
Funds (yes/no)	0.023	0.15	0	1
Regulation (yes/no)	0.154	0.36	0	1
Voluntary (yes/no)	0.16	0.36	0	1

Table 2: Summary statistics on the variables used in the empirical estimation

^{a-} Small, medium and large, small being the base category
 ^{b-} Openness is an ordered variable ranging from 0 to 8 where the highest score is 8 and the lowest score is 0

Variables	Passive eco- adopter	Late strat. eco-adopter	Early strat. eco-adopter	Late strat. eco- innovator	Early strat. eco-innovator
Dependent variables:	uuopiei	eco-uuopiei	eco-uuopiei	innovator	eco-mnovalor
Introduction of a product	0.51	0.11	0.11	0.52	0.51
innovation (yes/no)	(0.5)	(0.31)	(0.31)	(0.52)	(0.5)
Introduction of a process	0.55	0.15	0.14	0.63	0.62
•	(0.49)	(0.36)	(0.35)	(0.48)	(0.48)
innovation (yes/no)					. ,
Introduction of an	0.57	0.18	0.15	0.67	0.6
organisational innovation	(0.49)	(0.38)	(0.36)	(0.47)	(0.49)
(yes/no)					
Introduction of a marketing		0.13	0.13	0.54	0.49
innovation (yes/no)	(0.5)	(0.33)	(0.33)	(0.5)	(0.5)
Firm is an innovator	0.87	0.3	0.3	0.88	0.87
(yes/no)	(0.33)	(0.46)	(0.45)	(0.32)	(0.32)
Explanatory variables:					
Size 2006 ^a	0.54	0.67	0.63	0.86	1.03
	(0.68)	(0.7)	(0.68)	(0.76)	(0.76)
Size 2008 ^a	0.58	0.7	0.59	0.9	0.98
	(0.68)	(0.7)	(0.66)	(0.74)	(0.75)
Group (yes/no)	0.28	0.2	0.21	0.4	0.45
	(0.45)	(0.4)	(0.41)	(0.49)	(0.5)
R&D intensity	2100.4	0.02	0.02	0.02	0.086
Continuous D&D (was/na)	<i>(190357.1)</i> 0.17	(0.11)	(0.15)	(0.12)	(2.81)
Continuous R&D (yes/no)	(0.37)	0.03	0.02	0.17	0.23
Openness (0-8) ^b	4.8	(0.18) 7.07	(0.16) 7.25	(0.37) 5.63	(0.42) 5.63
Openness (0-8)	(2.62)	(2.05)	(1.86)	(2.4)	(2.4)
Scientific Cooperation	0.11	0.03	0.02	0.15	0.19
(yes/no)	(0.32)	(0.16)	(0.13)	(0.36)	(0.39)
Other Cooperation (yes/no)		0.06	0.05	0.32	0.32
1 0 /	(0.43)	(0.23)	(0.21)	(0.46)	(0.46)
International market	0.55	0.4	0.37	0.58	0.61
(yes/no)	(0.5)	(0.5)	(0.48)	(0.49)	(0.48)
National market (yes/no)	0.78	0.57	0.54	0.74	0.73
	(0.41)	(0.49)	(0.5)	(0.43)	(0.44)
Local market (yes/no)	0.69	0.57	0.61	0.63	0.62
	(0.46)	(0.49)	(0.48)	(0.48)	(0.48)
Funds (yes/no)	0.058	0.012	0.008	0.06	0.07
	(0.23)	(0.11)	(0.09)	(0.24)	(0.26)
Regulation (yes/no)	0.41	0.08	0.035	0.7	0.62
Voluntomy (weaker)	(0.49)	(0.27)	(0.18)	(0.45)	(0.48)
Voluntary (yes/no)	0.29	0.05	0.02	0.5	0.44
	(0.45)	(0.21)	(0.14)	(0.5)	(0.49)

Table 3: Summary statistics on the variables used in the estimations by type of ecoinnovator

- Standard deviations are displayed in parenthesis

a - Small, medium and large, small being the base category

b - Openness is an ordered variable ranging from 0 to 8 where the highest score is 8 and the lowest score is 0

3.5. Descriptive Statistics

In this section we present the general characteristics of the firms included in the sample by providing simple summary of the collected data. Table 4 displays the distribution of conventional innovators and eco-innovators across the manufacturing industries and service sectors. The main objective of this research being to stress the differences across the types of eco-innovators we have defined earlier, we present in Tables 5.A and 5.B summary statistics by type of eco-innovators across all manufacturing industries and service sectors.

We can first note significant differences between different industry segments with regard to the distribution of innovators and eco-innovators. Table 4 shows that there are more conventional innovators (product, process, organizational and marketing innovator) than ecoinnovators in all sectors and both of these innovators are concentrated within two industries. More precisely, among the manufacturing industries "Manufacture of Computer, Electronic and Electrical equipment" and "Manufacture of Coke, Chemicals, Chemical products" sectors have the highest concentration of conventional innovators and eco-innovators. Moreover, Table 5.A and 5.B show that these two industries comprise the most "early" and "late" strategic eco-innovators. In addition, the low-tech industries such as "Manufacture of Textiles, Wearing apparel and Leather products", "Manufacture of Food, Beverages and Tobacco products" and "Manufacture of Wood, Paper and Printing" have high rates of "passive eco-adopters" and, "early" and "late" strategic eco-adopters. This may point out that the technical sophistication of an industry may be one of the main determining factors that lead firms to pursue different innovative strategies and therefore, the reduction of negative impact might be the most relevant for the high-tech and high medium-tech industries than the low-tech industries.

Table 4: Descriptive statistics according to industry affiliati Nace codes	No# obs.	Mean (Innovator)	Mean (Eco- innovator)	
(B): Mining and Quarrying	1038	36.42%	31.21%	
(C10-C12): Manufacture of Food, Beverages and Tobacco products	6080	49.42%	26.74%	
(C13-C15): Manufacture of Textiles, Wearing apparel and Leather products	7855	31.1%	14.4%	
(C16-C18): Manufacture of Wood, Paper and Printing	4400	41.95%	29.43%	
(C19-C23): Manufacture of Coke, Chemicals and Chemical products	6331	53.23%	38.16%	
(C24-C25): Manufacture of Basic metals and Fabricated metal products	5344	46.44%	32.62%	
(C26-C30): Manufacture of Computer, Electronic and Machinery	6711	60.02%	43.5%	
(C31-C33): Manufacture of Furniture and Repair and installation of machinery	4598	45.65%	27.45%	
(D): Electricity, Gas, Steam and Air conditioning supply	1224	43.06%	37.17%	
(E): Water supply (F): Construction	2259 2202	40.9% 28.93%	41.08% 22.71%	
(G): Wholesale and Retail trade	13815	31.45%	15.88%	
(H49-H51): Land, Water transport and Air transport	4748	26.31%	17.17%	
(H52-H53): Warehousing and support activities for transportation and, Postal activities	1856	41.43%	22.84%	
(I): Accommodation and Food service activities	164	46.34%	44.51%	
(J58-J60): Publishing, Broadcasting and Video production activities	1119	50.04%	21.36%	
(J61-J63): Telecommunications, Computer programming, Consultancy and Information service activities	3160	55.44%	20.03%	
(K): Financial and Insurance activities	2534	57.1%	20.6%	
(L): Real estate activities	75	46.67%	30.67%	
(M69-M70): Legal and accounting activities and, Activities of head offices	450	53.56%	24.22%	
(M71-M73): Architectural, Engineering, Advertising, Market research and Scientific R&D activities	3070	50.81%	28.14%	
(M74-M75): Other professional, Scientific, Technical and Veterinary activities	112	57.14%	38.39%	
(N): Administrative and Support service activities	827	46.07%	32.89%	
Total	79972	42.78%	26.02%	

Table 4: Descriptive statistics according to industry affiliations

To be more precise, we can report that within the "Manufacture of Computer, Electronic and Electrical equipment" sector 60% and 43% of all the firms are conventional innovator and eco-innovators, respectively. Within the "Manufacture of Coke, Chemicals and Chemical products" sector these rates are 53% and 38%. The least innovative, both in terms of conventional innovation and eco-innovation, manufacturing sectors is the "Manufacture of Textiles, Wearing apparel and Leather products" by 31% and 14.4%, respectively. Among the service sectors, "Accommodation and Food service activities" appear to be the most eco-innovative. 45% of all the firms within this sector reported as eco-innovator.

Inspecting Table 5.A and 5.B gives a clearer picture of the distribution of different type of eco-innovators. In Table 5.A, we observe that the "passive eco-adopters" represent the biggest share of all types of eco-innovators. Accordingly, 12% of the eco-innovators have no specific strategy to eco-innovate and moreover, 31% of all firms do not eco-innovate at all. This may indicate that the most of the firms' lack environmental commitment and hence, firms' deficiency in embedding environmental concern into their general innovation strategy. Among all the manufacturing industries the "Manufacture of Computer, Electronic and Electrical equipment" sector has the highest concentration of passive eco-adopters (19%) while the "Accommodation and Food service activities" sector is the frontrunner within the service sectors (32%).

As in the case of passive eco-adopters, the "Manufacture of Computer, Electronic and Electrical" sector makes up the biggest proportion of late strategic eco-adopters (4.7%), while "Construction" has the biggest share of late strategic eco-adopters (8.9%) within the service sectors.

Nace codes	No# obs.	Non eco- innovator	Passive eco- adopter	Late strategic eco-adopter
			-	-
(B): Mining and Quarrying	1038	28.42%	12.04%	4.43%
(C10-C12): Manufacture of Food, Beverages and	6080	28.78%	12.38%	3.22%
Tobacco products	7055	27 40/	7.400/	4.020/
(C13-C15): Manufacture of Textiles, Wearing apparel and Leather products	7855	27.4%	7.49%	4.03%
(C16-C18): Manufacture of Wood, Paper and	4400	29.84%	13.75%	3.7%
Printing	4400	29.0470	13.7570	5.770
(C19-C23): Manufacture of Coke, Chemicals and	6331	25.44%	15.28%	3.22%
Chemical products	0551	23.4470	15.20%	5.2270
(C24-C25): Manufacture of Basic metals and	5344	29.67%	15.02%	3.79%
Fabricated metal products	5511	27.0770	15.0270	5.1770
(C26-C30): Manufacture of Computer, Electronic	6711	27.67%	18.89%	4.7%
and Machinery				,.
(C31-C33): Manufacture of Furniture and, Repair	4598	32.92%	14.28%	4.52%
and installation of machinery				
(D): Electricity, Gas, Steam and Air conditioning	1224	33.33%	15.68%	5.39%
supply				
(E): Water supply	2259	23.37%	13.28%	5.62%
(F): Construction	2202	60.94%	8.81%	8.9%
(G): Wholesale and Retail trade	13815	25.8%	7.49%	4.86%
(H49-H51): Land, Water transport and Air	4748	24.22%	9.49%	2.86%
transport				
(H52-H53): Warehousing and support activities	1856	37.44%	9.21%	5.6%
for transportation and, Postal activities				
(I): Accommodation and Food service activities	164	52.43%	31.7%	1.21%
(J58-J60): Publishing, Broadcasting and Video	1119	41.28%	14.29%	4.02%
production activities				
(J61-J63): Telecommunications, Computer	3160	42.12%	13.13%	4.33%
programming and Consultancy				
(K): Financial and Insurance activities	2534	43.21%	12.74%	3.23%
(L): Real estate activities	75	60%	24%	6.66%
(M69-M70): Legal and accounting activities and,	450	65.11%	20.88%	2%
Activities of head offices				
(M71-M73): Scientific R&D, Architectural and	3070	40.84%	15.47%	4.23%
Engineering activities				
(M74-M75): Other professional, Scientific,	112	55.35%	30.35%	0.89%
Technical and Veterinary activities				
(N): Administrative and Support service activities	827	53.08%	21.88%	2.29%
Total	79972	31.05%	12.33%	4.23%

Table 5.A.: Distribution of eco-innovators across industries for each type of eco-innovator

In Table 5.B, we can observe that the proportion of early strategic eco-adopters does not vary much between sectors. For example, across the manufacturing industries it varies only slightly from 7.41% to 13% indicating the existence of an established but rather low environmental consciousness within the manufacturing industries. Among the service sectors, however, the proportion of early strategic eco-adopters is much more variable (ranging from 1.11% to 16%). Among all the manufacturing industries, the "Mining and Quarrying" sector have the highest percentage of early strategic eco-adopters (13%), and "Financial and Insurance activities" have an even higher rate, as the leader within the service sector (16%). The divergence between the manufacturing industries and service sectors could be associated to significantly higher social and environmental costs pertinent to the manufacturing industries than the service sectors.

We can also note that the three industries have the highest concentration of early strategic eco-innovators and late strategic eco-innovators. Overall, the "Water Supply" sector has the highest rate of both types of strategic eco-innovators, with 11.8% of early strategic eco-innovators and 15.5% of late strategic eco-innovators. Within the manufacturing industries "Manufacture of Coke, Chemicals and Chemical products" and "Manufacture of Computer, Electronic and Electrical equipment" have the highest concentrations of both types of strategic eco-innovators make up 10.6% of the former sector, while this rate is 10% for the latter. Moreover, the early strategic eco-innovators constitute 11.6% of all the firms within "Manufacture of Coke, Chemicals and Chemical products" and Chemical products" and 14% of all the firms within "Manufacture of Coke, Chemicals and Electrical equipment". This may indeed point out the importance of reducing carbon footprint for the high-tech manufacturing industries.

Nace codes	No#	Early		Early strategic	
	obs.	strategic eco- adopter	eco-innovator	eco-innovator	
(B): Mining and Quarrying	1038	13%	9.63%	8.57%	
(C10-C12): Manufacture of Food, Beverages	6080	7.46%	6.61%	7.4%	
and Tobacco products	0000	7.4070	0.0170	7.770	
(C13-C15): Manufacture of Textiles, Wearing	7855	9.12%	2.71%	3.95%	
apparel and Leather products	1000	2.1270	2.7170	5.9570	
(C16-C18): Manufacture of Wood, Paper and	4400	9.06%	7.27%	7.86%	
Printing	1100	2.0070	7.2770	1.0070	
(C19-C23): Manufacture of Coke, Chemicals	6331	8.33%	10.61%	11.64%	
and Chemical products					
(C24-C25): Manufacture of Basic metals and	5344	7.41%	8.85%	8.15%	
Fabricated metal products					
(C26-C30): Manufacture of Computer,	6711	9.17%	10.04%	13.96%	
Electronic and Machinery		2			
(C31-C33): Manufacture of Furniture and	4598	8.61%	6.17%	6.65%	
Repair and installation of machinery					
(D): Electricity, Gas, Steam and Air	1224	11.02%	9.55%	11.51%	
conditioning supply					
(E): Water supply	2259	12.12%	11.81%	15.49%	
(F): Construction	2202	7.44%	7.31%	6.58%	
(G): Wholesale and Retail trade	13815	10.93%	4.14%	4.02%	
(H49-H51): Land, Water transport and Air	4748	7.79%	3.83%	3.39%	
transport					
(H52-H53): Warehousing and support activities	1856	12.23%	6.25%	6.73%	
for transportation and, Postal activities					
(I): Accommodation and Food service activities	164	1.82%	5.48%	7.31%	
(J58-J60): Publishing, Broadcasting and Video	1119	9.2%	3.03%	3.48%	
production activities					
(J61-J63): Telecommunications, Computer	3160	9.43%	3.51%	3.1%	
programming, Consultancy and Information					
service activities					
(K): Financial and Insurance activities	2534	15.78%	3.78%	3.7%	
(L): Real estate activities	75	2.66%	2.66%	4%	
(M69-M70): Legal and accounting activities	450	1.11%	1.77%	0.88%	
and, Activities of head offices					
(M71-M73): Architectural, Engineering,	3070	8.69%	5.96%	6.02%	
Advertising, Market research and Scientific					
R&D activities					
(M74-M75): Other professional, Scientific,	112	2.67%	5.35%	2.67%	
Technical and Veterinary activities					
(N): Administrative and Support service	827	3.38%	4.83%	5.19%	
activities		2.2070			
Total	79972	9.29%	6.3%	6.96%	

Table 5.B.: Summary statistics for different types of eco-innovators across all manufacturing industries and service sectors

4. Empirical Results

4.1. The Results of Estimations Conducted on the Whole Sample

As mentioned in section 3.3., we have estimated an innovation production function specified as an Ordered Probit model with selection. The dependent variable of the selection equation is specified as a Probit model and the dependent variable of the intensity equation (or outcome equation) is an ordered variable ranking firms from the less (non eco-innovator) to the more eco-innovative (early strategic eco-innovator). The results obtained from the ordered probit model are rather hard to interpret; one should also estimate the marginal effects of the independent variables in order to get robust and meaningful deductions. Accordingly, Table 6 reports marginal effects calculated after estimating the selection equation and the outcome equation. The first column of Table 6 displays the marginal effects for the selection equation. In the same vein, the other columns of Table 6 displays the marginal effects for each type of eco-innovator computed after estimating the ordered probit with selection.

The estimation results indicate that the firm-specific characteristics, organizational and technical capabilities and moreover, firms' cooperative behaviour are significantly important for eco-innovators. Voluntary engagement and environmental regulation are the other crucial factors that have a significant impact on firms' eco-innovation behaviour. Nevertheless, we also note that the marginal impact of these factors tend to change according to different types of eco-innovators. More precisely, we observe that the voluntary engagement is significantly more important for the early strategic eco-innovators than the rest, while the marginal impact of the environmental regulation is more significant for the late strategic eco-innovators than the other types of eco-innovators. We note a similar relationship between the voluntary engagement and environmental regulations and their respective marginal impact on early and late strategic eco-adopters. This indicates that self-determination is a key point for early strategic eco-innovators and early strategic adopters while environmental regulation might be an efficient tool for attracting new firms and channelling their innovative behaviour to a more sustainable approach.

Table 6 shows that being part of a larger group has a significantly positive marginal impact on each type of eco-innovator, except passive eco-adopters. This indicates that firms in groups, rather than individual firms, are more likely engage in eco-innovation. Firms that are part of a larger group are 1.5% and 2.4% more likely to be early strategic eco-innovator and late strategic eco-innovator, respectively. This can be explained through relatively easier

access to the financial resources and to technological and organisational assistance required for the realisation of eco-innovation. By considering the proximity to these essential factors, group members may perceive the eco-innovation process as less risky which may, in return, have a positive impact on firm's eco-innovative behaviour. Another reason for this effect could be the mutual innovative strategy of the group firms, in general. When the group members share the same concerns, in this case environmental concerns, their innovative approaches may become alike, which may explain the positive impact of group membership on eco-innovation behaviour.

The other control variable, firm size also indicates a positive relationship between firm size and firms' strategic behaviour towards eco-innovation. The estimation results show that both medium and large firms are more likely to be eco-innovators, especially the strategic eco-innovators. Large firms are 6% more likely to be early strategic eco-innovators and 7.7% more likely to be late strategic eco-innovators, while the marginal effect for early strategic eco-adopters and late strategic eco-adopters is 2.2% and 0.9%, respectively. As in the case of group membership large firms presumably have easier access to the financial resources required for their R&D investments. This assumption is supported by these results to a certain extent. Indeed, the uncertain market demand and financial risks inherent in the innovative process may hinder a firm's eco-innovative behaviour leaving only the medium and large size firms to engage in eco-innovation as they may be the only ones that have the required financial resources to actively engage in eco-innovation.

	Selection	Type of eco-innovator						
	equation Selection into Innovation	Early strategic eco-	Late strategic eco-	Early strategic eco- adopter	Late strategic eco-adopter	Passive eco- adopter	Non eco- innovator	
Group	0.065**	<i>innovator</i> 0. 015***	<i>innovator</i> 0.024***	0.008***	0.003***	0.001	-0.051***	
Gloup	(0.027)	(0.002)	(0.002)	(0.001)	(0.0004)	(0.001)	(0.006)	
Medium size	0.12***	0.02***	0.031***	0.01***	0.004***	0.001	-0.066***	
Wiedrum Size	(0.025)	(0.002)	(0.002)	(0.001)	(0.0004)	(0.001)	(0.006)	
Large size	0.25***	0.06***	0.077***	0.022***	0.009***	-0.023***	-0.145***	
	(0.045)	(0.004)	(0.004)	(0.001)	(0.0005)	(0.003)	(0.007)	
R&D intensity	-0.0002***	0.004***	0.006***	0.003***	-0.007***	-0.005**	-0.001***	
	(0.0)	(0.0002)	(0.0003)	(0.0001)	(0.0009)	(0.0001)	(0.0001)	
Continuous	0.7***	0.012***	0.017***	0.005***	0.002***	0.0001	-0.036***	
R&D	(0.055)	(0.0002)	(0.003)	(0.001)	(0.0005)	(0.0006)	(0.007)	
Openness	-0.076***	0.003***	0.005***	0.0022***	0.0008***	0.001**	-0.012***	
-	(0.004)	(0.0003)	(0.0005)	(0.0002)	(0.0001)	(0.0002)	(0.001)	
Scientific	-0.04	0.013***	0.021***	0.0063***	0.003***	-0.0003	-0.043***	
cooperation	(0.077)	(0.003)	(0.004)	(0.001)	(0.0005)	(0.0009)	(0.008)	
Other	0.93***	0.015***	0.023***	0.0076***	0.003***	0.0004	-0.049***	
cooperation	(0.05)	(0.002)	(0.003)	(0.001)	(0.0005)	(0.0008)	(0.007)	
International	0.21***	0.007***	0.013***	0.0042***	0.0022***	0.0016*	-0.028***	
market	(0.024)	(0.001)	(0.002)	(0.001)	(0.0004)	(0.0006)	(0.005)	
National market	0.21***	-0.005**	-0.01**	-0.003**	-0.0014**	-0.0006	0.02**	
	(0.026)	(0.002)	(0.003)	(0.001)	(0.0004)	(0.0003)	(0.006)	
Public funds	0.28***	0.012**	0.019**	0.006**	0.0022**	-0.0062	-0.033**	
	(0.087)	(0.003)	(0.004)	(0.001)	(0.0006)	(0.0007)	(0.009)	
Voluntary	0.52***	0.082***	0.11***	0.03****	0.01***	-0.034***	-0.198***	
-	(0.034)	(0.003)	(0.003)	(0.001)	(0.0006)	(0.003)	(0.006)	
Regulation	0.48***	0.016***	0.15***	0.045***	0.019***	-0.03***	-0.2***	
	(0.031)	(0.003)	(0.003)	(0.001)	(0.007)	(0.004)	(0.0061)	

Table 6: Marginal effects for the whole sample

- This table reports marginal effects

- Standard deviation is reported in parentheses below each coefficient -**** and **** denote significance at the 10%, 5% and 1% level, respectively

-Wald chi2 (88): 18562.74***

-Global significance of Industry dummies: 1403.46*** and Global significance of Country dummies: 8953.09***

- Number of observation: 30637 and Rho: -0.069

Amongst the conventional innovation inputs, the marginal effect of R&D intensity is significantly positive for early strategic eco-innovator (0.03%), late strategic eco-innovator (0.05%) and early strategic eco-adopter (0.02%). The impact of R&D intensity is negative for the rest of the eco-innovators and non-eco-innovators. We suggest that the R&D intensity

may point out the importance given to the introduction and/or development of new products and processes developed in-house therefore observing a relatively higher marginal impact for the strategic eco-innovators could support this claim. Nevertheless, the marginal impact of the R&D intensity is rather low. This could be due to the nature of eco-innovation itself, which often requires knowledge and skills that may fall outside of the firm's expertise. Hemmelskamp (1999) claims that potential eco-innovative companies are likely to have lower R&D intensity as they tend to countervail the low R&D effort with use of external sources of information. The other reason could be the original intentions of the innovative project. More precisely, if environmental concerns are at the heart of the innovative behaviour, a potential eco-innovator should and would incorporate environmental concerns and objectives into the innovative process from the beginning. Failing to do so may not be sufficient to yield eco-innovation per se. When the former is the case, the case of increasing need for additional technical and scientific knowledge, the eco-innovative firms may draw on the existing knowledge base accrued from engaging in continuous R&D and incorporate this knowledge with other competencies such as cooperation with external partners. Indeed, the estimation results point out this direction. Engagement in continuous R&D activities is the second conventional innovation input that reveals a positive marginal impact on a firm's propensity to become an eco-innovator especially, the "early" and "late" strategic ecoinnovator. Accordingly, firms that engage in continuous R&D activities are 1.7% more likely to be late strategic eco-innovator while the rate is 1% for being an early strategic ecoinnovator. The marginal effect for early strategic eco-adopters (0.5%) is higher than late strategic eco-adopters (0.2%) indicating the significance of scientific knowledge, technological capabilities and absorptive capacity that can be acquired by engaging in R&D on a continuous basis. These competencies may be of greater importance for strategic ecoinnovators than strategic eco-adopters since technical capabilities are more essential for developing an eco-innovation than for adopting them.

In addition, the results take our attention to the significance of the use of external sources of information (openness) for each type of eco-innovators. These firms are 0.5% and 0.3% more likely to be late strategic eco-innovators and early strategic eco-innovators, respectively. The marginal effect of openness for different types of eco-adopters is relatively less but still significant, indicating that the heterogeneous information is of great importance for each type of eco-innovator. We assume that firms would need to establish formal and informal networks to associate their existing knowledge with the other external knowledge. As Robin and Schubert (2013) claimed most lines of research agree on the fact that

interactions between industry and science are among the most prominent institutional interfaces for knowledge diffusion which is a crucial component for value creation and successful innovation. Networks may provide an adequate environment for potential ecoinnovators to fill the gap between their competencies and the competencies required to undertake successful innovation. Accordingly, indicators of formal cooperation, both with scientific and non-scientific partners, imply the strategic importance of establishing formal cooperation. Both of these formal cooperation indicators reveal a significantly positive marginal impact on the eco-innovation behaviour of different types of eco-innovators. Firms that establish cooperation with non-scientific partners are 1.5% and 2.3% more likely to be early strategic eco-innovators and late strategic eco-innovators, respectively. These percentages are 1.3% and 2.1% for those that established formal cooperation with scientific partners. The estimation results on the use of internal and external sources of information and the use of formal cooperation partners indeed indicate the complexity of the eco-innovation process and the heterogeneity of the knowledge and the information required to undertake successful eco-innovation. Besides the technical complexity of the innovation process, the environmental concerns inherent in the eco-innovation process can turn a promising innovative attempt into a complex mission. In order to avoid this type of inertia, firms can involve other partners and mitigate the risks inherent in the innovative process and complement their competencies with the information obtained from their formal and informal partners.

International market conditions appear to be another factor that has a positive impact on a firm's strategic eco-innovation behaviour. Firms that are active in international markets are 0.7% more likely to be early strategic eco-innovators while this impact is almost twice as high (1.3%) for late strategic eco-innovators. The results, however, also indicate that national market conditions may not be as efficient as international market conditions with regard to incentives that it may provide and hence, firms that are active in national markets are less likely to introduce eco-innovation and/or adopt eco-innovation. These results, in general, imply that only international competition can be regarded as a relevant driver for successful eco-innovation and/or adoption of eco-innovation, which may rule out the national and local market conditions as significant drivers for eco-innovating firms.

The EU's 6th and 7th Framework Programmes aim at providing funding opportunities to consolidate and develop the EU's leading edge in becoming one of the world's most knowledge-based and sustainable economies. The estimation results show that the financial support provided by the EU is, indeed, one of the significant components of an eco-

innovative firm's strategy. Accordingly, firms that have received the EU funds are 0.1% more likely to be early strategic eco-innovators while this rate is 1.6% for being a late strategic eco-innovator. We can also observe that the marginal effect of the EU funds is significantly positive for eco-adopters.

4.2. The Country Specific Analyses

The main objective of the above section is to draw a general picture of the firm-specific characteristics and capabilities of different types of eco-innovators found within the eleven EU countries and the respective roles of these characteristics within the eco-innovation process. Running a regression on the whole sample for the eleven European countries may, sometimes, be misleading since the sheer sample size may reflect some regressors as more significant than they actually are. In order to prevent any dispute on this issue we move the analysis one step ahead and conduct country specific analyses on sub-samples.

Accordingly, we constructed four different country groups out of the eleven European countries included in the CIS 2008 survey. The separation of country groups is based on the geographic proximity of each country to other countries in the group. Thus, the four different country groups are as follows: Western European (Germany and Ireland), Mediterranean (Cyprus and Portugal), Eastern European (Bulgaria, Romania, Slovakia, Czech Republic and Hungary) and Baltic (Lithuania and Estonia). To provide greater detail Table 7 (below) presents the distribution of eco-innovators and conventional innovators for each of the eleven countries. We can see that, there is significant heterogeneity among the countries in terms of the number of observations, the distribution of eco-innovators and non-eco-innovators.

Amongst all the countries included in the regressions, Germany and Portugal appear to be the most innovative. 67% of all German firms and 65% of all Portuguese firms included in the survey introduced at least one conventional innovation. When we consider the ecoinnovative performance, we observe that Portuguese firms appear to be slightly more ecoinnovative than German firms, which according to us a bit peculiar. Thus, 59% of all Portuguese firms and 53% of all German firms included in the survey appear to be ecoinnovators. Furthermore, with respect to innovative and eco-innovative performance, Bulgaria is the least innovative country in our dataset. Among all the firms included in the survey, 30% of all firms within this country introduced a conventional innovation while only 7.3% of them were able to introduce an eco-innovation during the observation period. Considering the heterogeneity inherent in the distribution of innovators in each country, it makes sense to conduct country specific analysis in order to shed light on the role of firm specific characteristics and capabilities that are assumed to have determining role on firm's strategic eco-innovation behaviour. The following section will discuss the regression results that are obtained from the country specific analyses.

Country	No# obs.	Innovator (Mean)	Eco-innovator (Mean)
Bulgaria	31718	29.52%	7.31%
Cyprus	1024	58.11%	15.82%
Czech Republic	6804	57.55%	48.5%
Germany	6026	66.69%	52.7%
Estonia	3986	62.07%	30.76%
Hungary	5390	33.04%	21.04%
Ireland	2178	58.72%	42.1%
Lithuania	2111	42.02%	22.64%
Portugal	6512	64.71%	58.71%
Romania	9631	39.57%	32.79%
Slovakia	4592	40.94%	24.35%
Total	79972	42.78%	26.02%

Table 7: Distribution of innovators across countries

- An innovator is product, process, organisational and marketing innovator without any environmental impact

4.2.1. The Results of Estimations Conducted by Group of Countries

As was the case for our global regression model, we start the discussion of the results with those pertaining to the marginal effects obtained after conducting the Ordered Probit model with selection. Table 8.A, 8.B, 8.C and 8.D displays the estimation results pertinent to the country specific sub-samples.

When we inspect the estimation results, we can see that only the control variables (group membership and firm size), voluntary engagement and regulation seem to be important for all the eco-innovators within all country groups. Moreover, the marginal impacts of these factors appear to be more significant for the strategic eco-innovators than the strategic eco-adopters in all country groups but their respective importance varies accordingly. More precisely, within the Western European and Baltic countries firms' structural characteristics, voluntary engagement in eco-innovation and environmental regulation are the main factors that leads firms to be early strategic eco-innovators. The marginal impact of the same factors is of higher importance for the late strategic eco-innovators within the Eastern European and Mediterranean countries highlighting the differences in eco-innovation strategies of each country group. Nonetheless, our analysis reveals new insights and highlights the existence of significant differences between country groups with regard to their cooperative behaviour, importance of conventional innovation inputs, market conditions and financial funds provided by the EU.

To begin with, we can observe that within the Western European countries engagement in R&D on a continuous basis and being active in the international markets are distinguishing characters shared by all types of eco-innovators. As expected, the respective importance of these factors is significantly more important for the early strategic eco-innovators than the rest. Moreover, the marginal impact of voluntary engagement is also higher than the impact of the regulatory framework for the early strategic eco-innovators asserting the significance of voluntary proactive approaches to eco-innovation for this type of eco-innovators. Lastly, we can note that the marginal impact of the EU funds is significantly positive only for the late strategic eco-innovators and late strategic eco-adopters indicating that the EU's 6th and 7th Framework Programmes are indeed efficient tools in providing incentives and attracting new firms to engage in eco-innovation.

	Selection	T	ype of eco-inn	iovator			
Variables	equation Selection into Innovation	Early strategic eco- innovator	Late strategic eco- innovator	Early strategic eco- adopter	Late strategic eco- adopter	Passive eco- adopter	Non eco- innovator
Group	0.031	0.028**	0.013**	0.003**	0.002**	0.004	-0.05**
	(0.06)	(0.009)	(0.004)	(0.001)	(0.0004)	(0.003)	(0.014)
Medium size	0.11	0.041***	0.019***	0.004***	0.002**	0.004	-0.07***
	(0.057)	(0.009)	(0.004)	(0.001)	(0.0005)	(0.004)	(0.015)
Large size	0.17	0.148***	0.052***	0.012***	0.005***	-0.047**	-0.17***
	(0.09)	(0.019)	(0.006)	(0.002)	(0.001)	(0.016)	(0.015)
R&D	-0.66**	0.009	0.004	0.001	0.0004	0.002	-0.017
intensity	(0.24)	(0.033)	(0.015)	(0.003)	(0.001)	(0.007)	(0.061)
Continuous	0.52***	0.025**	0.012**	0.003*	0.001*	0.003	-0.044**
R&D	(0.097)	(0.009)	(0.004)	(0.001)	(0.0005)	(0.003)	(0.016)
Openness	0.28***	0.002	0.0008	0.0002	0.00	0.0004	-0.0034
	(0.018)	(0.002)	(0.0009)	(0.0002)	(0.01)	(0.0006)	(0.003)
Scientific	-0.073	0.02	0.0082	0.0023	0.0008	0.0017	-0.033
cooperation	(0.129)	(0.013)	(0.006)	(0.001)	(0.0006)	(0.002)	(0.02)
Other	0.31**	0.02	0.008	0.002	0.001	0.002	-0.033
cooperation	(0.113)	(0.011)	(0.005)	(0.001)	(0.0005)	(0.002)	(0.018)
International	0.138*	0.0172*	0.007*	0.002*	0.0008*	0.004	-0.031*
market	(0.06)	(0.008)	(0.004)	(0.001)	(0.0004)	(0.003)	(0.015)
National	0.091	-0.0042	-0.0012	-0.0006	-0.0001	-0.0008	0.007
market	(0.062)	(0.01)	(0.005)	(0.001)	(0.0004)	(0.002)	(0.018)
Public funds	0.6	0.037	0.016*	0.003	0.001*	-0.002	-0.055*
	(0.34)	(0.021)	(0.008)	(0.001)	(0.0008)	(0.007)	(0.026)
Voluntary	0.34***	0.147***	0.05***	0.011***	0.004***	-0.046***	-0.166***
	(0.083)	(0.014)	(0.005)	(0.001)	(0.001)	(0.013)	(0.012)
Regulation	0.36***	0.104***	0.04***	0.009***	0.004***	-0.014	-0.143***
	(0.073)	(0.01)	(0.004)	(0.001)	(0.0009)	(0.009)	(0.013)

Table 8.A: Marginal effects for the Western European group

- This table reports marginal effects

- Standard deviation is reported in parentheses below each coefficient -^{*}, ** and **** denote significance at the %10, %5 and %1 level, respectively

-Wald chi² (64): 1702.67***

-Global significance of Industry dummies: 305.31***

- Number of observation: 4016 and Rho: -0.26**

Within the Baltic countries, formal cooperation with non-scientific partners is of significant importance for all type of eco-innovators especially, for the early strategic eco-innovators. As stated in Von Hippel (1988) and Belderbos *et al.* (2004) cooperation with non-scientific partners (such as customers, suppliers, etc.) may indeed reduce the risks inherent in the market introduction phase and can increase the technological base of the firm. Therefore, we found supporting evidence for this claim. Furthermore, unlike the Western European case, we now see that the marginal impact of environmental regulation is slightly higher than the marginal impact of voluntary engagement for the early strategic eco-innovators. This result indicates that firms within the Baltic countries. However, the estimated coefficients of the voluntary engagement and regulatory framework are quite similar to each other asserting that they are the main factors shaping the firms' eco-innovation strategy within the Baltic countries.

	Selection		Type of eco	-innovator			
	equation						
Variables	Innovator	Early Strategic Eco- Innovator	Late Strategic Eco- Innovator	Early Strategic Eco- Adopter	Late Strategic Eco- Adopter	Passive Eco- Adopter	Non Eco- Innovator
Group	-0.07	0.021*	0.019**	0.005*	0.002*	0.003	-0.051**
	(0.08)	(0.008)	(0.007)	(0.002)	(0.0008)	(0.002)	(0.019)
Medium size	0.32***	0.044***	0.041***	0.012***	0.004***	0.008	-0.111***
	(0.085)	(0.008)	(0.007)	(0.002)	(0.001)	(0.005)	(0.02)
Large size	0.54***	0.104***	0.07***	0.016***	0.005***	-0.017	-0.178***
	(0.13)	(0.028)	(0.014)	(0.003)	(0.001)	(0.012)	(0.036)
R&D intensity	0.073	-0.09	-0.083	-0.025	-0.008	-0.017	0.226
	(0.052)	(0.067)	(0.062)	(0.019)	(0.006)	(0.017)	(0.168)
Continuous	0.45**	0.009	0.008	0.002	0.0009	0.001	-0.023
R&D	(0.169)	(0.01)	(0.008)	(0.002)	(0.0009)	(0.001)	(0.023)
Openness	-0.33***	0.004	0.003	0.001	0.0003	0.0007*	-0.01
	(0.02)	(0.003)	(0.002)	(0.0007)	(0.0002)	(0.0003)	(0.006)
Scientific	0.0	0.003	0.003	0.001	0.0003	0.0006	-0.009
cooperation	(0.0)	(0.013)	(0.011)	(0.003)	(0.001)	(0.002)	(0.032)
Other	0.5***	0.033***	0.03***	0.008**	0.003**	0.004	-0.08***
cooperation	(0.112)	(0.008)	(0.007)	(0.002)	(0.001)	(0.004)	(0.021)
International	0.48***	0.017	0.016	0.005	0.001	0.004	-0.044
market	(0.088)	(0.009)	(0.009)	(0.003)	(0.001)	(0.004)	(0.026)
National market	0.175*	0.003	0.003	0.0009	0.0003	0.0006	-0.008
	(0.086)	(0.009)	(0.009)	(0.002)	(0.001)	(0.002)	(0.025)
Public funds	-0.083	0.016	0.014	0.004	0.001	0.001	-0.038
	(0.247)	(0.014)	(0.012)	(0.003)	(0.001)	(0.001)	(0.031)
Voluntary	0.45**	0.174***	0.11***	0.026***	0.007***	-0.028*	-0.29***
	(0.146)	(0.017)	(0.009)	(0.003)	(0.001)	(0.014)	(0.019)
Regulation	0.98***	0.196***	0.126***	0.03***	0.009***	-0.024	-0.338***
	(0.118)	(0.014)	(0.01)	(0.004)	(0.002)	(0.014)	(0.024)
Lithuania	-0.96***	-0.077***	-0.074***	-0.023***	-0.008***	-0.024	0.21***
	(0.091)	(0.009)	(0.011)	(0.004)	(0.002)	(0.013)	(0.034)

Table 8.B: Marginal effects for the Baltic group	Table 8.B:	Marginal	effects f	for the	Baltic	group
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- Standard deviation is reported in parentheses below each coefficient -^{*,**} and ^{***} denote significance at the %10, %5 and %1 level, respectively -Wald chi² (67): 2332.2***

-Global significance of Industry dummies: 322.8*** - Number of observation: 3320 and Rho: -0.001

Within the Mediterranean countries, engagement in R&D on a continuous basis is a significant factor for each type of eco-innovators however, it is relatively more important for the late strategic eco-innovator than the rest of the eco-innovators. Besides, international market conditions and their interactions with external partners seem to be other significant exogenous factors for the eco-innovators within this country group. We observe that the eco-innovators tend to be open to the use of internal and external sources of information and, establish a formal bond with other cooperation partners. Additionally, both of the strategic eco-adopters tend to take this cooperative behaviour one step forward and incorporate the scientific partners within the eco-innovative process. Therefore, we can claim that firms within the Mediterranean countries are more likely to have a cooperative behaviour and the late strategic eco-innovators are driven by mostly with their voluntary motivation while the environmental regulation is of second degree importance.

	Selection		Type of eco-in	novator			
	equation						
Variables	Innovator	Early Strategic Eco- Innovator	Late Strategic Eco- Innovator	Early Strategic Eco- Adopter	Late Strategic Eco- Adopter	Passive Eco- Adopter	Non Eco- Innovator
Group	0.19**	0.02***	0.043***	0.004***	0.002***	-0.016**	-0.053**
	(0.062)	(0.005)	(0.009)	(0.001)	(0.0005)	(0.005)	(0.011)
Medium size	0.06	0.022***	0.046***	0.004***	0.002***	-0.02***	-0.058***
	(0.058)	(0.004)	(0.008)	(0.0009)	(0.0005)	(0.005)	(0.01)
Large size	0.23	0.088***	0.128***	0.009***	0.004***	-0.097***	-0.132***
	(0.167)	(0.015)	(0.014)	(0.002)	(0.001)	(0.018)	(0.013)
R&D	3.21	-0.009	-0.022	-0.002	-0.001	0.007	0.029
intensity	(4.08)	(0.012)	(0.028)	(0.003)	(0.001)	(0.008)	(0.036)
Continuous	1.6***	0.022***	0.044***	0.004***	0.002***	-0.02**	-0.053***
R&D	(0.31)	(0.006)	(0.011)	(0.001)	(0.0005)	(0.006)	(0.012)
Openness	-0.56***	0.003**	0.007**	0.0007**	0.0004**	-0.002*	-0.009***
	(0.033)	(0.001)	(0.002)	(0.0002)	(0.0001)	(0.001)	(0.002)
Scientific	-0.44	0.011	0.024	0.002*	0.001*	-0.009	-0.03*
cooperation	(0.283)	(0.006)	(0.013)	(0.001)	(0.0006)	(0.006)	(0.015)
Other	1.73***	0.009*	0.021*	0.002*	0.001*	-0.007*	-0.026*
cooperation	(0.175)	(0.004)	(0.009)	(0.001)	(0.0005)	(0.003)	(0.012)
International	0.125*	0.009*	0.02*	0.002*	0.001*	-0.006*	-0.027*
market	(0.052)	(0.003)	(0.008)	(0.0009)	(0.0005)	(0.002)	(0.01)
National	0.143*	-0.003	-0.007	-0.0007	-0.0004	0.002	0.009
market	(0.058)	(0.005)	(0.011)	(0.001)	(0.0006)	(0.003)	(0.013)
Public funds	0.46	-0.0004	-0.001	-0.0001	0.0	0.0003	0.001
	(0.36)	(0.007)	(0.016)	(0.001)	(0.0009)	(0.004)	(0.021)
Voluntary	0.53***	0.091***	0.157***	0.014***	0.007***	-0.083***	-0.186***
	(0.061)	(0.006)	(0.009)	(0.001)	(0.001)	(0.009)	(0.011)
Regulation	0.43***	0.062***	0.113***	0.01***	0.005***	-0.06***	-0.133***
	(0.07)	(0.005)	(0.008)	(0.001)	(0.0007)	(0.007)	(0.01)

Table 8.C: Marginal effects for the Mediterranean group

- This table reports marginal effects

- Standard deviation is reported in parentheses below each coefficient -*,** and **** denote significance at the %10, %5 and %1 level, respectively

-Wald chi² (64): 2540.48***

-Global significance of Industry dummies: 452.37***

- Number of observation: 6599 and Rho: -0.06

In Table 8.D we observe that the Eastern European eco-innovators tend to differentiate themselves from the other eco-innovators within other country groups. The main difference has to do with use of internal and external sources of information and, the formal bonds with

external partners. Accordingly, openness is one of the distinguishing characteristic of the Eastern European eco-innovators along with the cooperative behaviour with the scientific and other non-scientific partners. Moreover, their relative importance is higher for the late strategic eco-innovators than the rest. As in the Mediterranean countries, continuous R&D effort is another significant factor leading firms to be more eco-innovative especially, for the late strategic eco-innovators. International market condition is also another significant factor. Therefore, firms that are active in the international markets tend to be eco-innovative, while national market conditions tend to reveal exact opposite impact. Finally, we also observe that financial funds provided by the EU's 6th and 7th Framework Programmes are perceived as another significant factor by the Eastern European eco-innovators. These funds are of higher importance for the late strategic eco-innovators than the rest hence, cannot be disregarded.

	Selection equation						
Variables	Innovator	Early Strategic Eco- Innovator	Late Strategic Eco- Innovator	Early Strategic Eco- Adopter	Late Strategic Eco- Adopter	Passive Eco- Adopter	Non Eco- Innovator
Group	0.035	0.011***	0.022***	0.01***	0.004***	0.001	-0.049***
	(0.059)	(0.002)	(0.004)	(0.001)	(0.0008)	(0.0006)	(0.009)
Medium Size	0.022	0.015***	0.03***	0.014***	0.006***	0.001	-0.068***
	(0.051)	(0.002)	(0.003)	(0.001)	(0.0008)	(0.0008)	(0.008)
Large Size	0.003	0.044***	0.073***	0.03***	0.012***	-0.013***	-0.146***
	(0.078)	(0.004)	(0.005)	(0.002)	(0.0009)	(0.003)	(0.009)
R&D	0.005	0.0007	0.001	0.0006	0.0003	0.0001	-0.003
intensity	(0.007)	(0.0004)	(0.0008)	(0.0004)	(0.0001)	(0.0001)	(0.002)
Continuous	-0.133	0.005*	0.01*	0.004*	0.002*	0.0006*	-0.023*
R&D	(0.084)	(0.002)	(0.005)	(0.002)	(0.001)	(0.0002)	(0.012)
Openness	0.143***	0.002***	0.005***	0.002***	0.001***	0.0006**	-0.011***
openness	(0.011)	(0.0003)	(0.0007)	(0.0003)	(0.0001)	(0.0001)	(0.001)
Scientific	-0.31**	0.012***	0.023***	0.01***	0.004***	-0.0002	-0.05***
cooperation	(0.097)	(0.003)	(0.006)	(0.002)	(0.001)	(0.001)	(0.012)
Other	0.55***	0.011***	0.022***	0.01***	0.004***	0.0008	-0.049***
cooperation	(0.086)	(0.002)	(0.004)	(0.002)	(0.0009)	(0.0006)	(0.009)
International	0.136**	0.007***	0.015***	0.007***	0.003***	0.001**	-0.035***
market	(0.05)	(0.001)	(0.003)	(0.001)	(0.0007)	(0.0006)	(0.008)
National	0.131**	-0.008***	-0.016***	-0.007***	-0.003***	-0.0008	0.036***
market	(0.05)	(0.002)	(0.004)	(0.001)	(0.0008)	(0.0004)	(0.008)
Public funds	-0.007	0.007*	0.013*	0.006*	0.002*	0.0004	0.03*
	(0.092)	(0.003)	(0.006)	(0.002)	(0.001)	(0.0005)	(0.013)
Voluntary	0.31***	0.056***	0.089***	0.035***	0.013***	-0.022***	-0.17***
-	(0.087)	(0.004)	(0.005)	(0.002)	(0.0009)	(0.003)	(0.008)
Regulation	0.23***	0.09***	0.14***	0.054***	0.021***	-0.03***	-0.277***
C	(0.062)	(0.004)	(0.005)	(0.002)	(0.001)	(0.003)	(0.007)
Romania	0.34***	0.446***	0.217***	0.026***	-0.003**	-0.27***	-0.41***
	(0.081)	(0.011)	(0.006)	(0.002)	(0.001)	(0.006)	(0.005)
Slovakia	0.4***	0.054***	0.083***	0.031***	0.011***	-0.027***	-0.15***
	(0.118)	(0.006)	(0.007)	(0.002)	(0.0009)	(0.005)	(0.011)
Czech	0.11	0.063***	0.098***	0.038***	0.014***	-0.027***	-0.18***
republic	(0.074)	(0.005)	(0.005)	(0.002)	(0.001)	(0.004)	(0.009)
Hungary	-0.52***	0.101***	0.13***	0.043***	0.014***	-0.067***	-0.22***
6	(0.072)	(0.008)	(0.007)	(0.002)	(0.0009)	(0.007)	(0.009)

Table 8.D: Marginal effects for the Eastern European group

- This table reports marginal effects

- Standard deviation is reported in parentheses below each coefficient -**** and **** denote significance at the %10, %5 and %1 level, respectively

-Wald chi² (75): 9778.03***

-Global significance of Industry dummies: 507*** - Number of observation: 13660 and Rho: -0.212***

5. Conclusion

Eco-innovation is one of the most pivotal developments in the search to solve the world's environmental problems and sustainability challenges. Most of the empirical research addresses the drivers and barriers of eco-innovation from the conventional market-pull, technology-push, and institutional (regulatory) factors perspective (Porter and Van der Linde, 1995; Rennings, 2000; Brunnermeier and Cohen, 2003; Jaffe et al., 2002) while few studies highlight the importance of other dynamic factors, such as firms' economic, technical, managerial and social capabilities (Horbach, 2008; Borghesi et al., 2012; De Marchi, 2012). Highlighting the significance of these factors then, is our primary object. In addition to that significant amount of econometric research has focused on eco-innovation that is new to the world, disregarding the potential differences in characteristics and capabilities among ecoinnovators and eco-adopters (Kemp and Pontoglio, 2011, p. 34) which constitutes our second research question that is addressed within the context of this research. Therefore, we aim at overcoming these limitations and try to identify the role of firms' internal characteristics and organizational capabilities for different type of eco-innovators by employing a unique database drawn from the last wave of the Community Innovation Survey (CIS) 2008. The most significant added value of this research is its ability to distinguish eco-innovators into five mutually exclusive categories namely "Passive eco-adopters", "Late strategic ecoadopter", "Early strategic eco-adopter", "Late strategic eco-innovator" and "Early strategic eco-innovator". Besides, we also conducted country specific analyses in order to compare and contrast the country specific differences with regard to strategic behaviour of eco-innovators and the role of firm specific characteristics within the eco-innovation process.

The estimation results indicate that a firm's structural characteristics and organizational and technical capabilities are crucial factors affecting its strategic behaviour toward ecoinnovation. Those factors also have significant impact on firms' strategic behaviour towards eco-innovation. We note that large and medium size firms and moreover, firms that are part of a larger organization (group firms) are more likely to be eco-innovators. Firms' cooperative behaviour (such as, cooperation with scientific partners and/or other partners) and their ability to use external sources of information are the other factors that increase firms' probability to engage in eco-innovation. We have evidence to believe that potential eco-innovators are more likely to be active in larger markets (e.g. international market). All the more, we observe that the marginal impact of certain factors tend to change according to type of eco-innovator. More precisely, early strategic eco-innovators distinguish themselves from the rest of the eco-innovators with a highest marginal effect of the adoption of voluntary environmental agreements. Strategic adopters, by contrast, display a higher marginal effect of environmental regulation and available public funding, such as the financial funds provided by the EU's 6th and 7th FP. This, on the one hand, demonstrates the efficiency of regulatory framework in attracting firms' attention to eco-innovation. On the other, it suggests that we still in need of individually determined risk takers to achieve higher level of eco-innovative efforts. If efficiently used these tools may have a significant impact on firm's eco-innovator firms' innovative behaviour to a more sustainable approach.

We also found some intriguing results in regard to the role of conventional R&D inputs as determinants of a firm's eco-innovation behaviour. Our estimation results show that the marginal impact of a firm's R&D intensity is rather low for eco-innovators. This result indeed point out the role of some other external factors (such as; scientific cooperation, cooperation with others and informal cooperation 'openness') within a firm's eco-innovation process. Given that eco-innovation process is very complex, presumably even more than conventional innovation, realization of eco-innovation relies on multiple interactions with various actors (such as, research institutes, independent experts of government agencies etc.). Therefore, a firm's single-handed efforts, such as increasing R&D expenditures, may not be enough to allow for the introduction of an eco-innovation. This has let us conclude that potentially eco-innovative companies are likely to have lower R&D intensity, since they tend to countervail the low R&D effort with the use of external sources information as Hemmelskamp's (1999) suggested.

By acknowledging the technical problems that running a regression on the whole sample, we also conducted country specific analyses on sub-samples. Our overall estimation results consolidated with the country-specific estimation results. In short, the analyses showed that different country groups diverge from each other with respect to the role of their cooperative behaviour and the role of conventional R&D inputs. Within Western European and Baltic countries, firms' structural characteristics, voluntary engagement in eco-innovation and environmental regulation are the main factors associated with a higher probability to be early strategic eco-innovators. Within Eastern European and Mediterranean countries, the marginal impact of those same factors is higher for late strategic eco-innovators, which may highlight differences with respect to eco-innovation strategies across country groups. This is to say, besides the firms' technical capabilities, their global network can also be crucially important for successful eco-innovation. Forming formal cooperation with scientific and other partners

and, being open to the external sources of information appear to be especially important for eco-innovators. Eco-adopters, as well as eco-innovators, tend to look for external financial support to mitigate financial risks involved in the process. According to our understanding, this shows the complex nature of eco-innovation process which may require external information, knowledge and skills that may not be at the firm's disposal. When this is the case, entering into partnership with scientific and non-scientific partners could be of greater importance especially for the less R&D-intense sectors. Within these sectors, the cost of external knowledge can be costly and time consuming to obtain. Firms that possess well established global network, both with formal and informal partners, could use their acquaintance to lower the associated cost and the time required.

Hence, we can conclude this chapter by claiming that a firm's structural characteristics and organizational and technical capabilities are crucial factors affecting its strategic behaviour toward eco-innovation. All the more, the conceptual framework being involved (e.g., available public funds, environmental regulation, international market conditions) and, the interrelations with other external factors (through the use of external source of information and through formal cooperation with scientific and non-scientific partners) may also shape a firm's innovation strategy. These factors tend to have even greater significance as the environmental and technical challenges inherent in eco-innovation process makes its development and the introduction more complex and a dynamic process (maybe even more so than conventional innovation). In that respect, eco-innovation should be treated apart from conventional innovation, since the former may require specific organisation, knowledge, know-how, and technical capabilities in order to be realized

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General Conclusion

This Ph.D. thesis aims at contributing to both theoretical and empirical literature of ecoinnovation by questioning the role of firms' structural, economic, technical, managerial and social capabilities in eco-innovation process. Eco-innovation differentiates itself from the conventional innovation by embedding environmental concerns as its primary objective. This contextual shift in innovation behaviour makes eco-innovation one of the most significant developments in the search to solve the world's environmental problems and sustainability challenges. Besides its desired environmental impacts such as reducing firms' carbon footprints, an eco-innovation may also provide firms a substantial financial return through decreased inputs use and increasing demand for their product, process, organisational and marketing eco-innovations. Moreover, if realized on a macro-scale, it can be also a pathway to a more sustainable, competitive and knowledge based economy as the Lisbon Agenda (2000) postulates. However, in order to achieve a perceivable environmental efficiency and observe environmental amelioration, we need to make fundamental changes in policy, institutions and practices. Therefore, development and introduction of eco-innovation is pertinent to people, firms, industries, countries and in short to the Earth. Eco-industries such as air pollution control, waste water management, solid waste management, soil remediation, and renewable energies and recycling are leading the way to solve world's environmental problems and they are also able to capture sizeable business opportunities. According to Bleischweitz et al. (2009) these industries expected to worth around 1\$ trillion worldwide by 2015. Danish Wind Industry Association (DWIA²⁹, 2013) claims that 28% of Denmark's electric supply is produced by the wind power and by 2050 Denmark is to be free of fossil fuels. The DWIA (2013) also states that Danish wind energy industry holds a quarter of the global turnover and the industry exported 7 Billion EUR worth of electric in 2012. These statements indeed demonstrate the economic and environmental significance of ecoinnovation not only for a firm or a geographic area but for a country as a whole. Up until now, researchers have emphasized the importance of conventional technology-push, marketpull drivers and regulatory framework for successful eco-innovation. However, the significance of a firm's structural, economic, technical, managerial and social capabilities was rarely explicitly analysed. This thesis, therefore, aimed at filling this gap by identifying

²⁹ Danish Wind Industry Association (DWIA) (2013): visited on 20-11-2013 available at: http://www.windpower.org/en/knowledge/statistics.html

the determinants of eco-innovation from the firms' economic, technological, organisational and social capabilities perspective. This research question is addressed with four mutually exclusive chapters.

In each chapter, we have addressed the determinants of eco-innovation issue in different contexts. The advantage of carrying out multi-aspect analysis of the determinants of eco-innovation is that it allows us to gain a deeper knowledge about the firm specific characteristics and capabilities that are associated with development and introduction of eco-innovation. By conducting these separate empirical analyses, we can draw specific conclusions from each study and make links between all four chapters. The empirical findings of this Ph.D. thesis are derived through the application of various statistical and econometric tools to a number of firm-level databases. In each chapter, we used the most adequate econometric tools that can control for selection and endogeneity biases. By using original databases, we are able to examine the impact of various firm specific characteristics and, technical, managerial, social and organisational capabilities on development and introduction.

The Chapter 1 of this thesis forms a literature review of the most prominent theoretical and empirical contributions in the field. Within the first chapter, we provide a detailed analysis of eco-innovation concept from a historical perspective. To the best of our knowledge, a review of the eco-innovation literature and the empirical investigation of the role of firm specific characteristics and capabilities are still lacking. Therefore, we aim at filling this existing gap and provide a solid theoretical background for our research. In Chapter 2 aims at contributing to the drivers of and barriers to eco-innovation discussion. By conducting empirical analyses, we try to shed light on the role of a firm's internal characteristics on the adoption of voluntary environmental standards (such as, ISO 14001 and EMAS) and the causal effect of adoption of environmental standards on the firms' business performance i.e., through VA. We developed further the extent of our research question by diversifying the adopters of these standards as *early* and *late* adopters. This chapter makes a clear contribution to discussions about the so-called Porter hypothesis. In Chapter 3, we stressed the significance of a firm's structural characteristics in another context. With this chapter, we aimed at providing a clearer understanding of the relationship between firms' structural characteristics, capabilities, cooperative behaviour and development of product and/or process eco-innovations. Besides conducting an empirical analysis on the whole sample (disregarding the country specific differences), we also conducted country specific analyses by clustering various European countries into several groups. Country specific analyses conducted in Chapter 3 are significantly important as the eco-innovation literature lacks empirical evidence that compares and contrasts the importance of certain firm characteristics for the eco-innovation in different European countries. With **Chapter 4** we moved our empirical investigation to another level. Consequently, we aim at identifying the role of a firm's internal characteristics and, technical and organizational capabilities on the eco-innovation (product, process, organisational and marketing eco-innovation) behaviour for different type of eco-innovators. We suggest a new typology of eco-innovators and ecoadopters for empirical purposes and aim at highlighting the importance of a firm's structural characteristics amongst these eco-innovators and eco-adopters. Moreover, within this chapter, in order to compare and contrast the results obtained from the global estimation, we again conducted country specific analyses. The discussion below summarizes the main findings and their policy implications.

Main results

Our review of theoretical researches in **Chapter 1** show that eco-innovation is a rather new paradigm in innovation studies, and thus the very definition of eco-innovation is still evolving. Some trends can however be identified. For instance, most current definitions of eco-innovation tend to be based on environmental performance rather than environmental aim. Current definitions also tend to be broad, in the sense that they simply require the environmentally benign (product, process, organizational or marketing) innovation to be novel at the firm level rather than at the level of a market, an industry or an economy. Therefore, the issue of drivers and barriers to eco-innovation requires to be addressed with certain precision in terms of level of analysis. In this thesis, we aimed at identifying the drivers and barriers to eco-innovation at the firm-level rather than at the industry or macro level.

Our survey of the literature also suggests that, besides conventional supply-side and demand-side determinants of innovation, policy and regulation appear as specific drivers of eco-innovation. The existing empirical evidence tend to support this assertion and shows that environmental product innovation tends to be more driven by the strategic market behaviour of firms (market pull), whereas policy and regulation tend to drive environmental process innovation. The latter aspect could be associated with the public-good character of clean technologies which leads to under-investment in environmental R&D. Therefore, firms tend

to perceive environmental regulations and standards as an opportunity (rather than an obstacle) to become eco-innovative. We could not find, however, an extensive body of work investigating the impact of some of the less stringent environmental regulations (e.g., voluntary proactive approaches such as; ISO 14001 and EMAS) on the introduction of eco-innovation. Their role remains controversial, and we suggest that more empirical investigation is required precisely assess the importance of voluntary environmental standards and norms in the eco-innovation process. This question is investigated in the second chapter of this thesis.

Last but not least, based on our survey of the literature we claim that if diffusion of ecoinnovation is as beneficial as the Porter hypothesis suggests (i.e., likely to overcome environmental problems without hampering economic performance), then more (empirical) research on the diffusion processes of eco-innovation is needed. Conducting a thorough empirical analysis on this issue could assist policy makers in promoting eco-innovation among both firms and consumers. Within the context of this research, we could not touch upon this important issue as the available quantitative data does not allow such an empirical investigation. Therefore, it could be an important research question to be investigated in the future.

The empirical findings in Chapter 2 suggest that firms' internal characteristics, such as size, indeed play a significant role in the adoption process of environmental standards. Moreover, firms that already have adopted ISO 9001 quality standards and provide customeroriented services are also more likely to adopt voluntary 'ISO 14001-type' environmental standards. This, in its basic sense, could point out that an existing experience in managing quality standards and in some other practices may indeed provide firms with a certain ease in implementing ISO 14001 and EMAS-type environmental standards. We also note that the adoption of these types of standards seems more relevant for firms that are active in wider market (e.g., the European or international market rather than the national market). This result is consistent with the fact that the EU is one of the first political entities to come up with a road-map (the Lisbon Agenda of 2000, predecessor of Europe 2020) towards becoming one of the most competitive and knowledge-based economy in the world. Indeed, increasing environmental consciousness amongst European customers and regulatory pressures from the EU are giving firms strong incentives to clean up their operation by replacing, re-organizing and/or re-designing their production process, products. organizational behaviour or marketing methods.

We also observe that the firms' strategic behaviour towards the adoption of these standards varies with the technological complexity and the environmental impact of the industry in which a firm operates. Accordingly, the *early* adoption of voluntary standards is more relevant for high-technology and high/medium technology manufacturing firms. Firms operating in high tech industries tend to have higher carbon footprints, and voluntary environmental standards might therefore be an efficient choice to become both more competitive and greener. These environmental standards might not only help firms to clean up their operations, they might also enable firms to market themselves as "environmentally friendly". Firms could use the adoption of such standards as a signal in order to acquire additional market shares, enter new markets or simply reduce their inputs costs by replacing them with less environmentally hazardous ones.

In a nutshell, the results in Chapter 2 tend to support the predictions of the so-called Porter Hypothesis to a certain extent. We can claim that voluntary environmental standards may be good complements to market-based instruments (such as tradable permits, emission taxes or subsidies) and "command and control" mechanisms, as they are liable to make firms greener without harming their economic performance (quite to the contrary, they might even improve it). Policy makers could use this opportunity to communicate and explain the positive economic and social gains that could be reaped from the adoption of environmental standards. Doing so would increase the number of adopters, up to a certain threshold. Indeed, one can imagine that as the number of adopters increase pas a certain threshold, the competitive advantages and gains begin to decrease, as environmentally-conscious consumers have a larger number of "green" suppliers to turn to for fulfilling their everyday needs. Our empirical analysis suggest that we are still far from this theoretical situation, however – and firms desiring to keep a competitive advantage can always increase their efforts to become greener.

In **Chapter 3** we identify the firm characteristics and economic factors that lead firms to become more eco-innovative. The empirical analyses conducted in this chapter show that large and individual firms tend to have a higher intensity of eco-innovation. Besides, firms that are open to the use of external sources of knowledge and active in larger markets (e.g. international market) are also likely to have a higher intensity of eco-innovation. We also found evidence that eco-innovative firms tend to have a lower engagement in "continuous" R&D and comparatively low R&D intensity. They however tend to countervail their lower R&D activity by cooperating with external partners (in particular, public research institutes

and universities) which supports Hemmelskamp (1999)'s claims. Although this conclusion may be initially met with some suspicion, at least two closely-related elements can be brought forward to support it. The first is that the eco-innovation process is very complex – even more so than conventional innovation – and by its very nature relies on multiple interactions with various actors. The second element is that eco-innovation cannot generally be realized single-handedly: while increasing a firm's R&D intensity is often necessary to introduce important innovations, it may not be enough to allow for the introduction of an ecoinnovation. In that respect, eco-innovation should be treated apart from conventional innovation, since the former may require specific intentions, capabilities, considerations, knowledge stock, know-how, and strategic planning in order to be realized. Without embedding environmental issues into their basic R&D efforts, firms can rarely come up with an eco-innovation (in saying so, we rule out conventional innovations with unintended positive environmental consequences). And this embedment generally relies on cooperation with external partners, be they research institutes, independent experts of government agencies.

The country-specific analyses showed the robustness of the results obtained with the global estimation. Certain firm-specific characteristics ('firm size' and 'openness') are important within all country groups, while the importance of conventional R&D inputs and networking activities varies with country groups. The reliance on conventional R&D inputs tend to be more significant in Western European and Mediterranean countries, while the reliance on networks is more significant for Eastern European and Baltic countries. We can attribute this divergence to the existing technological gap among these country groups, which may indeed force Eastern European and Baltic firms to accelerate their innovation production processes in order to ameliorate their technological gap among countries might be dependent on how well they can complement the eco-innovation process with the external information and knowledge.

To conclude on this chapter, we do not deny the importance of the conventional innovation drivers for eco-innovation. However, our results assert that the nature of eco-innovation differs from the nature of conventional innovation to a certain extent and that both should be treated as separately. Conventional innovation drivers, such as basic R&D efforts, are clearly not sufficient to explain the intensity of eco-innovation, which largely depends on

cooperation and joint efforts realized with external actors such as government agencies and public research institutions.

Chapter 4 consolidates and builds upon the deductions obtained in Chapter 3. In Chapter 4, we show that a firm's structural characteristics and organizational and technical capabilities are crucial factors affecting its strategic behaviour toward eco-innovation. Besides, we also show that firms' eco-innovation strategies could be affected by other external factors (such as environmental regulation and available public funding) and by their reactions to these external factors (such as the adoption of voluntary environmental agreements). Firms' cooperative behaviour and their ability to use external sources of information are other factors that increase firms' probability to engage in eco-innovation. While all these factors matter for all types of eco-innovators, their marginal effect tends to change across types of eco-innovator. Our empirical evidence shows that early strategic ecoinnovators distinguish themselves from the rest of the eco-innovators with a highest marginal effect of the adoption of voluntary environmental agreements. Strategic adopters, by contrast, display a higher marginal effect of environmental regulation and available public funding, such as the financial funds provided by the EU's 6th and 7th FP. Moreover, eco-innovators in general tend to have a comparatively lower R&D intensity, which they countervail by cooperating with external partners (in particular, public research institutes and universities) and by using of external sources information, as Hemmelskamp's (1999) suggested.

Following the reasoning proposed in Chapter 3, we also conducted country-specific analyses. These analyses show that country groups differ from each other with respect to the role of their cooperative behaviour and the role of classic R&D inputs. Within Western European and Baltic countries, firms' structural characteristics, voluntary engagement in ecoinnovation and environmental regulation are the main factors associated with a higher probability to be early strategic eco-innovators. Within Eastern European and Mediterranean countries, the marginal impact of those same factors is higher for late strategic ecoinnovators, which may highlight differences with respect to eco-innovation strategies across country groups. Overall, these results points out towards certain factors that qualifies as specific determinants of eco-innovation. In order to introduce eco-innovation, firms tend to give considerable importance to their existing network and to the contextual framework in which they operate. Therefore, creating a formal partnership with scientific and non-scientific partners and/or informal interaction with external sources of information could be crucially important, especially, for less R&D-intense sectors. As far as we understand, this gives important clues about the nature of eco-innovation process. Besides internal characteristics and their ability to use external information, firm's knowledge capital and skills are of an even greater importance for eco-innovation than for innovation. The required knowledge could pertain to a specific technology, process, product, input or to a new marketing method that may not necessarily be obtained or developed internally. We can conclude that the development and the introduction of eco-innovations is a complex and a dynamic process (maybe even more so than conventional innovation), it's the realization of which goes well beyond the reach of the single-handed efforts of individual firms. In that respect, ecoinnovation should be treated apart from conventional innovation, since the former may require specific organisation, knowledge, know-how, and technical capabilities in order to be realized.



Serdal OZUSAGLAM ESSAIS SUR L'ECO-INNOVATION



Résumé étendu en français

L'innovation environnementale ou "éco-innovation" est un concept apparu assez récemment dans la littérature consacrée à l'innovation. La spécificité de ce concept est qu'il vise à combiner deux objectifs: réaliser une avancée technologique ou un changement dans les procédés ou l'organisation de l'entreprise d'une part, et préserver ou restaurer l'environnement naturel d'autre part. Bien que le concept d'éco-innovation ait déjà fait l'objet d'une abondante littérature, de nombreuses interrogations demeurent, en particulier quand il s'agit de quantifier ses effets sur la performance économique des entreprises, ou encore d'identifier les facteurs conduisant ces dernières à adopter un comportement éco-innovant. Cette thèse vise à apporter des éléments de réponse empiriques à ces questions, au travers de quatre chapitres correspondant à quatre articles de recherche distincts.

En premier lieu, afin de fournir un cadre de référence solide à l'analyse empirique, le Chapitre 1 propose un survol de la littérature consacrée à l'éco-innovation. A partir de ce socle, les chapitres suivants proposent trois analyses originales sur des thèmes identifiés comme pertinents dans le premier chapitre. Le Chapitre 2 réexamine, dans une perspective empirique, l'hypothèse dite de Porter, selon laquelle une réglementation environnementale plus stricte conduirait non seulement à préserver l'environnement mais aussi à une performance économique accrue des entreprises. Cette hypothèse est examinée ici dans une version « faible », le Chapitre 2 s'intéressant au rôle des règlementations environnementales facultatives, comme les normes de type ISO 14001, dont l'adoption est volontaire. Nous y examinerons l'effet de l'adoption de normes de ce type sur la performance des entreprises, à l'aide des données françaises de l'enquête COI 2006 combinées aux Enquêtes Annuelles d'Entreprises de 2003 à 2006. Le Chapitre 3 porte sur le processus de diffusion de l'éco-innovation, abordée (en raison de contraintes liées aux données) du point de vue des déterminants de l'innovation environnementale, et des barrières à cette dernière. L'analyse, qui vise à identifier ces déterminants et barrières, utilise des modèles de « fonction de production d'innovation » estimés sur les données européennes micro-anonymisées ("bruitées") de l'Enquête Communautaire sur l'Innovation de 2004 (CIS 2004). Le Chapitre 4 étend cette analyse en exploitant l'Enquête Communautaire sur l'Innovation de 2008 (CIS 2008), qui contient un module spécifiquement dédié à l'innovation environnementale. L'analyse repose toujours sur des modèles économétriques de « fonction de production d'innovation », mais l'exploitation des données ("bruitées") CIS 2008 permet de changer l'angle d'approche, en proposant une classification originale des différents types d'entreprises éco-innovantes. Le contenu de ces quatre chapitres est détaillé ci-après.

CHAPITRE 1 : REVUE DE LA LITTERATURE

A la lumière des travaux passés en revue dans le Chapitre 1, il apparaît que l'éco-innovation n'est pas encore un concept bien stabilisé : sa définition même continue d'évoluer, appelant à des corrections et à des précisions. Fussler et James (1996) furent sans doute les premiers auteurs à distinguer l'écoinnovation de l'innovation au sens usuel du terme, en la définissant comme l'introduction de nouveaux produits et/ou procédés qui créent de la valeur tout en réduisant significativement leurs impacts sur l'environnement. A mesure que le concept d'éco-innovation attirait l'attention des chercheurs, sa définition même évoluait, au risque d'en faire un concept polysémique. Les premières définitions du concept, assez strictes, se basaient uniquement sur les innovations ayant un objectif environnemental explicite, négligeant ainsi les bénéfices environnementaux que peuvent potentiellement procurer les innovations au sens usuel du terme. Les définitions les plus récentes, toutefois, insistent sur le fait que c'est sa performance en matière environnementale, plutôt que son objectif, qui caractérise une éco-innovation. L'objectif en lui-même ne présente guère d'intérêt, dans la mesure où une innovation ayant les meilleures intentions en matière d'amélioration de l'environnement peut très bien échouer. Il est donc plus pertinent d'observer les retombées environnementales d'une innovation, et de la caractériser a posteriori (ex post, disent les économistes) comme une éco-innovation ou comme une innovation « conventionnelle ». S'il fallait aujourd'hui trouver une définition faisant la quasi-unanimité au sein de la communauté des chercheurs, elle pourrait prendre la forme de celle proposée par Kemp et Pearson (2007). Selon ces auteurs, l'écoinnovation est « l'assimilation ou l'exploitation d'un produit, procédé de production, service ou méthode de gestion nouvelle pour l'entreprise ou l'utilisateur et qui résulte, tout au long de son cycle de vie, en une réduction des risques environnementaux, de la pollution et des autres impacts négatifs de l'utilisation de ressources (y compris énergétiques) comparées à des alternatives pertinentes » (Kemp et Pearson, 2007, p. 3).

Cette définition est extrêmement générale, et englobe différents comportements éco-innovants (cf. Figure 1). Elle inclut ainsi non seulement les innovations qui visent explicitement à réduire les impacts sur l'environnement, mais aussi celles qui conduisent à une réduction des impacts environnementaux sans que ce cela soit leur objectif initial. Ainsi, toute innovation « conventionnelle » qui aurait des effets bénéfiques pour l'environnement comptera comme une éco-innovation. Comme le souligne l'OCDE (qui conçoit l'éco-innovation comme un outil permettant de relever les défis sociétaux sur la voie du développement durable), si l'éco-innovation peut découler d'un objectif environnemental à la base, elle peut également survenir comme effet secondaire d'autres

objectifs, tels que l'accroissement de la productivité, la réduction du coût des facteurs de production, ou encore le respect de la réglementation et de normes environnementales (OCDE, 2009).

Eco-adoptions passives	
Eco-adoptions stratégiques	Comportement non éco- innovant
Eco-innovations stratégiques	

Figure 1 : catégorisation des comportements des entreprises en matière d'éco-innovation

Source : adapté de Kemp et Pearson (2007)

Ce premier chapitre accorde précisément une attention toute particulière au rôle de la règlementation, qui est souvent présentée dans la littérature comme un déterminant essentiel de l'écoinnovation, notamment quand l'innovation concerne un procédé de fabrication (moins polluant que les procédés existants). Une controverse demeure à ce propos, en particulier quand la règlementation repose sur des normes environnementales peu contraignantes, voire facultatives (telles que la norme ISO 14001). Cette controverse trouve sa source dans l'hypothèse dite de Porter (Porter et Van der Linde, 1995), selon laquelle une réglementation environnementale plus stricte conduirait d'une part à des innovations ayant des effets bénéfiques pour l'environnement, et d'autre part à une performance économique accrue des entreprises. Bien qu'elle soit à l'origine d'un intérêt accrue pour l'éco-innovation, l'hypothèse de Porter alimente la controverse, car les études empiriques qui tentent de la tester ou de la mettre à l'épreuve ne permettent pas de l'étayer systématiquement (sur un plan plus théorique, Jaffe et al. (2002) soulignent que même en admettant sa validité, l'ampleur des gains réalisés, tant en matière de performance économique qu'en matière environnementale, demeure sujette à caution).

En fait, comme le suggéraient déjà Porter et Van der Linde (1995), tout dépend du type de politiques environnementales utilisées. Popp (2009) affirme qu'en général, les politiques s'appuyant sur le levier du marché (étendent à fournir davantage d'incitations à innover, dans la mesure où elles récompensent une amélioration continue de la qualité de l'environnement. Par opposition, les mesures contraignantes de type « command and control » (imposition d'une norme obligatoire, par exemple) pénalisent les pollueurs qui ne respectent pas la norme en vigueur mais ne récompensent pas les entreprises qui, en matière de respect de l'environnement, parviennent à faire mieux que la norme imposée. Par exemple, si une norme imposait que 10% de l'électricité utilisée par les entreprises d'un

pays doit être issue de sources renouvelables, une entreprise qui utiliserait 15% d'électricité « renouvelable » ne serait pas récompensée pour cela par rapport à une autre entreprise qui s'en tiendrait juste à la norme. De même, si une norme imposait de ne pas rejeter plus de x tonnes de CO2 par an, une entreprise qui rejetterait une quantité strictement inférieure à x n'en retirerait aucun avantage comparativement à une entreprise qui rejetterait une quantité juste égale à x. Popp (2009) considère par ailleurs qu'au-delà du type de mesure utilisée (recours aux normes versus recours au marché), les différences existant au sein d'un type de mesure (différences entre plusieurs instruments basés sur le recours au marché par exemple) peuvent également se traduire par des bénéfices environnementaux plus ou moins importants.

La question des politiques environnementales aboutit à un autre point, sinon controversé, du moins encore relativement inexploré dans la littérature : la diffusion des éco-innovations. Depuis les travaux de Rogers (1962) – situés aux confins de la sociologie, de la psychologie et de la théorie de la communication), la diffusion d'une innovation (au sens usuel du terme) est généralement définie comme le processus par lequel les acteurs d'un système social se communiquent une innovation au fil du temps, à travers certains canaux. Cette diffusion est un processus complexe, dont l'aboutissement peut prendre aussi bien 5 ans que 50 ans, selon la formule de Mansfield (1968). La lenteur du processus de diffusion d'une innovation est liée à l'incertitude et au risque lié à la commercialisation et/ou à l'utilisation d'un produit, d'une technologie, ou d'un procédé nouveau. Il existe dans la littérature un consensus selon lequel la diffusion d'une innovation suit un schéma temporel assez prévisible : les nouvelles technologies, nouvelles méthodes ou nouveaux procédés sont tout d'abord adoptés assez lentement, puis plus rapidement et à nouveau plus lentement à mesure qu'un seuil d'adoption (spécifique à chaque nouveauté) est atteint. L'évolution du nombre ou de la proportion d'entreprises (ou plus généralement d'acteurs socio-économiques) adoptant une innovation peut ainsi être représentée à l'aide d'une courbe en S (Rogers, 1983 ; Blackman, 1999). Il s'ensuit que le succès d'une innovation repose d'une manière cruciale sur les réactions des premières entreprises (ou des premiers consommateurs) à l'adopter, dans la mesure où celles-ci (ceux-ci) sont considérés comme des exemples à suivre par les acteurs économiques plus « timorés » (Moore, 1998).

Dans le cas de l'éco-innovation, il est probable que la diffusion de technologies « propres » (ou « vertes ») suive également une courbe en S. Kemp et Volpi (2007) expliquent que la diffusion d'une technologie dite propre (au sens environnemental du terme) est très lent au début, car la technologie est initialement méconnue et entourée d'une incertitude encore plus grande que les technologies impliquées dans le cadre d'innovations « conventionnelles ». En dépit de cette incertitude, il ne fait guère de doute que la diffusion des éco-innovations est souhaitable : d'une part, l'éco-innovation est une des pistes privilégiés pour maintenir (voire accroître) la qualité de vie à long terme ; d'autre part, les industries éco-innovantes sont à l'origine de profits non négligeables et pourraient le devenir encore davantage dans les années à venir (Commission Européenne, 2012).

Toutefois, en raison de l'incertitude mentionnée plus haut, laisser le marché être l'unique vecteur de diffusion des éco-innovations est risqué dans la mesure où celles-ci pourraient être introduites trop tardivement pour inverser les tendances actuelles de dégradation de l'environnement. Il reste donc à trouver des politiques environnementales permettant d'encourager la diffusion des éco-innovations.

De telles politiques sont toutefois difficiles à mettre en œuvre dans la mesure où les freins à l'innovation environnementale sont encore mal connus, de même que les facteurs qui la favorisent. Elles semblent néanmoins indispensables, dans la mesure où il existe une double externalité inhérente à l'innovation environnementale, comme le souligne Rennings (2000). Les éco-innovations présentent en effet, en premier lieu, les externalités usuelles liées à la production de connaissances pendant la phase de R&D. Elles présentent également, en second lieu, des externalités positives sur l'environnement pendant la phase d'adoption et de diffusion, externalités qui profitent à tous et qui rendent la diffusion des éco-innovations toujours souhaitable. Toutefois, ces externalités peuvent conduire à une situation où les rendements privés de la R&D sur des technologies « vertes » sont inférieurs aux rendements sociaux – en raison de la nature de quasi « biens publics » des éco-innovations, ce qui peut amener les entreprises (en l'absence de politiques environnementales appropriées) à sous-investir dans la R&D environnementale.

Nous refermons ce chapitre en concluant qu'il convient de focaliser les recherches empiriques sur les deux points évoqués ci-dessus : d'une part, la controverse sur l'hypothèse de Porter, en particulier quand la règlementation repose sur des normes environnementales peu contraignantes, et d'autre part, l'identification des facteurs qui influencent le processus de diffusion de l'innovation environnementale (que ce soit en la freinant ou en l'encourageant). C'est ce que nous essayons de faire dans les chapitres suivants.

CHAPITRE 2 : LES NORMES ENVIRONNEMENTALES A ADOPTION VOLONTAIRE : DE NOUVELLES STRATEGIES POUR PROMOUVOIR UN « BUSINESS VERT » ?

Le **Chapitre 2** s'intéresse au premier point de controverse évoqué dans le Chapitre 1, à savoir le rôle des règlementations environnementales ayant une dimension facultative, comme les normes de type ISO 14001, dont l'adoption est volontaire. A l'aide d'une analyse par appariement sur les scores de propension, nous examinons l'effet de l'adoption de normes de type ISO 140001 (ou un système de management environnemental, ci-après SME) sur la performance des entreprises (mesurée à partir de la Valeur Ajoutée, ci-après VA). La première étape de l'analyse consiste à prédire l'adoption de ces normes à l'aide d'un modèle Probit afin de calculer le score de propension (cette première étape permet également d'identifier un certain nombre de déterminants de l'adoption). La seconde étape consiste à calculer, à l'aide d'un algorithme d'appariement non paramétrique (ici, le kernel), l'effet de l'adoption sur la VA des entreprises ayant adopté les normes (par rapport aux entreprises ayant un

score de propension similaire mais n'ayant pas adopté ces normes). L'analyse est menée sur les données françaises de l'enquête COI 2006 (dont le champ est constitué des entreprises françaises de plus de 10 salariés), combinées aux Enquêtes Annuelles d'Entreprises (EAE) de 2003 à 2006. Ces données nous permettent de distinguer parmi les entreprises adoptant des normes de types ISO 14001 (ou un SME), celles (*early adopters*) les ayant adopté précocement, c'est-à-dire en 2003 ou avant, et celles (*late adopters*) les ayant adopté tardivement, c'est-à-dire entre 2003 et 2006. Ces deux types d'adoption s'opposent au comportement de non-adoption, comme indiqué dans la Figure 2 (il n'existe pas, dans nos données, d'entreprises ayant adopté un SME en 2003 puis l'ayant abandonné en 2006). Afin de tenir compte de l'existence de ces deux comportements d'adoption, le score de propension est calculé à l'aide d'un Probit multivarié et non pas d'un Probit simple.

En fusionnant l'enquête COI 2006 et les données administratives des EAE, nous obtenons, après un nettoyage standard des données fusionnées (consistant à exclure les entreprises de moins de 10 salariés et celles ayant un taux de croissance annuel du chiffre d'affaires supérieur à 100%, indice d'une fusion-acquisition) un échantillon de 5584 entreprises observées en 2003 et 2006 (autrement dit, un panel de deux ans). Le Tableau 1 présente – avant et après la fusion des bases de données – la moyenne et l'écart-type des principales variables utilisées dans l'analyse. Un test de différences des moyennes indique que la fusion et le nettoyage des bases de données n'ont pas conduit à des biais de sélection significatifs.

Figure 2 : modalité d'adoption d'un Système de Management Environnemental (SME) par les entreprises de la base COI 2006

	L'entreprise dispose d'un SME en 2006	L'entreprise ne dispose pas d'un SME en 2006
L'entreprise dispose d'un SME en 2003	Adoption précoce	catégorie non renseignée
L'entreprise ne dispose pas d'un SME en 2003	Adoption tardive	Non-adoption

	COI COI et EAE après		E après nettoyage	Test	
Variables:	Moyenne	Ecart Type	Moyenne	Ecart Type	P-valeur
Taille (nombre de salariés)	691,85	3739,84	653,70	3735,85	0,29
C.A. en euros	168558,50	1132874,00	152650,10	1034387,00	0,21
Filiale d'un groupe en 2006	1,34	0,48	1,35	0,48	0,38
Filiale d'un groupe en 2003	1,36	0,48	1,36	0,48	0,37
Ppal marché en 2006 :local	1,13	0,34	1,13	0,34	0,48
national	1,30	0,46	1,30	0,46	0,24
européen	1,54	0,50	1,53	0,50	0,25
international	1,64	0,48	1,64	0,48	0,41
Ppal marché en 2003 : local	1,14	0,34	1,14	0,34	0,46
national	1,31	0,46	1,30	0,46	0,20
européen	1,55	0,50	1,55	0,50	0,26
international	1,65	0,48	1,65	0,48	0,39
Certifiée ISO 9001 en 2006	1,49	0,50	1,47	0,50	0,10
" ISO 14001 en 2006	1,80	0,40	1,79	0,41	0,20
Certifiée ISO 9001 en 2003	1,53	0,50	1,52	0,50	0,08*
" ISO 14001 en 2003	1,85	0,36	1,84	0,37	0,19
Secteur d'activité:					
Industrie haute tech.	0,03	0,18	0,04	0,19	0,27
Industrie moy./haute tech.	0,09	0,29	0,10	0,29	0,16
Industrie moy./basse tech.	0,10	0,30	0,10	0,31	0,14
Industrie basse tech.	0,14	0,35	0,15	0,36	0,08*
Services intensifs en					
connaissance	0,18	0,39	0,18	0,38	0,32
Autres services	0,46	0,50	0,44	0,50	0,02*

Tableau 1 – Distribution des principales variables avant et après fusion/nettoyage des données

Les résultats (effets marginaux) de l'estimation du Probit multivarié sont fournis dans le Tableau 2. Ils convient de rappeler que ces résultats concernent des entreprises françaises (ou implantées en France) dans la première décennie du XXIème siècle. Ces résultats suggèrent que, contrastés au comportement de non-adoption (des standards de type ISO 14001), l'adoption précoce et l'adoption tardive présentent plus de similitudes que de différences – les principaux déterminants étant la taille de l'entreprise, l'appartenance à un groupe, et l'adoption préalable de standards de type ISO 9001. En ce qui concerne la taille des entreprises, nos résultats suggèrent une relation en U inversé entre la taille (mesurée par le nombre de salariés) et la probabilité d'adopter des standards de type ISO 14001 (précocement ou tardivement). Autrement dit, la probabilité d'adoption s'accroît jusqu'à un certain seuil (l'adoption pouvant se révéler encore trop coûteuse pour des entreprises de petite taille) puis décroît au-delà de ce seuil (l'adoption pouvant se révéler trop complexe à gérer pour des entreprises de très grande taille, surtout si elles sont réparties sur plusieurs pays).

	Adoption précoce		Adoption tardi	ve
	Coefficients	(Err. Std)	Coefficients	(Err. Std)
Taille	0.00015***	(0.000024)	0.000054***	(0.000017)
Taille au carré	-0.00013***	(0.000026)	-0.0008**	(0.00004)
Filiale d'un groupe	-0.043***	(0.009)	-0.013*	(0.006)
Certifiée ISO9001	0.178***	(0.009)	0.058***	(0.006)
Label de qualité	0.035***	(0.008)	0.005	(0.008)
Livraison type « juste à temps »	0.021**	(0.010)	0.007	(0.006)
Active sur marché national	-0.023**	(0.011)	-0.014**	(0.007)
Active sur marché européen	0.024**	(0.012)	0.009	(0.007)
Active sur marché international	0.015	(0.011)	0.009	(0.006)
Industrie (réf. : Services)				
Haute tech.	0.035	(0.021)	0.003	(0.013)
Haute/moyenne tech.	0.071***	(0.013)	-0.0022	(0.008)
Moyenne/basse tech.	0.010***	(0.013)	0.0051	(0.008)
Basse tech.	0.017	(0.012)	0.01	(0.007)
Services intensifs				
en connaissance	-0.060***	(0.013)	-0.022**	(0.009)
N	5455		4867	
Pseudo R ²	0.21		0.1	
Khi 2 du R.V. (12 dl)	975.72		192.82	
Statistique du R.V.		-1900.8971		

Tableau 2 – Effets marginaux du Probit multivarié sur l'adoption d'un standard de type ISO 14001

* significatif au seuil de 10%, ** significatif au seuil de 5%, *** significatif au seuil de 1% Erreurs Standards entre parenthèses

L'appartenance à un groupe est associée à un probabilité plus faible d'adopter des standards de type ISO 14001 (les filiales devant généralement suivre les pratique managériales de la maison mère, il est possible qu'une relative frilosité de celle-ci se répercute sur celles-là). Enfin, le fait d'avoir adopté des standards de qualité de type ISO 9001 est associé à une probabilité accrue d'adopter des standards environnementaux facultatifs de type ISO 14001. Sans doute faut-il voir là un effet cumulatif : les entreprises ayant déjà adoptés des standards de qualité peuvent en percevoir – au delà de l'aspect contraignant – les retombées bénéfiques pour leur image de marque et leur activité en général. Cela les rend sans doute plus réceptives aux potentialités des standards environnementaux facultatifs de type ISO 14001, qui se situent dans le prolongement des normes de types ISO 9001.

Comme indiqué plus haut, la distinction entre adoption précoce et adoption tardive des SME ne repose que sur un petit nombre de facteurs. En premier lieu, le recours à des labels de qualité (autres que la certification ISO 9001) et/ou à des pratiques garantissant le respect de certaines conditions de livraison (de type juste-à-temps pour des fournisseurs, par exemple) est associé à une probabilité d'adoption précoce plus élevée (alors que ce recours n'a aucun effet sur le probabilité d'une adoption tardive). De même, opérer au niveau européen (plutôt que sur le seul marché national) est associé à une probabilité d'adoption précoce plus élevée, ce qui s'explique sans doute par le fait que les entreprises opérant à ce niveau géographiques sont plus sensibilisées à l'importance des

normes, standards et certifications (nombre d'entre elles étant requises pour opérer à travers les pays de l'UE).

	Adoption pré	doption précoce		tion tardive
	Coefficients	(Err. Std)	Coefficients	(Err. Std)
PSM	0.49***	(0.070)	0.51***	(0.100)
FILM	0.38***	(0.076)	0.28***	(0.071)

Tableau 3 – estimation de l'effet moyen du traitement sur les traités (ATT)

* significatif au seuil de 10%, ** significatif au seuil de 5%, *** significatif au seuil de 1% Erreurs Standards entre parenthèses

L'estimation du modèle Probit dont nous venons de commenter les résultats ne constitue toutefois qu'une étape préalable à l'analyse qui nous intéresse ici. En effet, nous voulons surtout déterminer si l'adoption de standards facultatifs de type ISO 14001 à un effet significatif sur la performance économique des entreprises, et si cet effet est positif ou négatif. Au-delà des résultats factuels qu'il fournit, le modèle Probit nous permet avant tout de calculer le score de propension, que l'on peut définir ici comme la probabilité prédite d'adopter des standards environnementaux conditionnellement aux caractéristiques observées (le variables explicatives du modèle Probit) des entreprises les ayant effectivement adopté (dites entreprises « traitées » selon la terminologie habituelle de cette méthode) mais aussi des entreprises ne les ayant pas adopte (le « groupe de contrôle » selon la terminologie de cette méthode). L'appariement sur score de propension consiste à comparer la performance économique des entreprises du groupe traité à celle des entreprises du groupe de contrôle en choisissant parmi ces dernières un ensemble d'entreprises (le « support commun ») présentant des caractéristiques similaires à celles du groupe traité (c'est-à-dire un score de propension ayant des valeurs proches de celles observées dans le groupe traité). L'appariement se fait à l'aide d'un algorithme non paramétrique (ici, le kernel). Cette méthode permet d'éliminer un certain nombre de biais de sélection et d'endogénéité, pour se rapprocher de l'estimation d'un effet causal sur des données non expérimentales.

L'effet estimé est appelé « effet du traitement sur les traités » (ici, effet de l'adoption de normes environnementales de type ISO 14001) ou en abrégé ATT (pour *Average effect of the Treatment on the Treated*). L'estimation de l'ATT obtenu dans le cas de notre étude est présentée dans la première ligne du Tableau 3, pour les comportements d'adoption précoce (colonne de gauche) et d'adoption tardive (colonne de droite). Nos résultats indiquent qu'adopter des normes environnementales de type ISO 14001 sur une base volontaire permet d'accroître significativement la

VA des entreprises (toutes choses égales par ailleurs), que l'adoption soit précoce ou tardive (la valeur de l'ATT étant de 0,49 dans le premier cas et de 0,51 dans le second).

Une analyse de sensibilité conduite en estimant un modèle linéaire à interactions complètes (*Fully Interacted Linear Model* ou FILM) vient conforter ces résultats, en donnant toutefois un léger avantage à l'adoption précoce (la valeur de l'ATT, présentée dans la seconde ligne du Tableau 3, étant alors de 0,38 pour une adoption précoce contre 0,28 seulement pour une adoption tardive). Le principe du modèle FILM est d'imposer le « support commun » évoqué plus haut avant de procéder à une estimation par les moindres carrés ordinaires (MCO) d'un modèle linéaire dont la variable dépendante est (dans notre étude) la mesure de performance des entreprises et dans lequel une variable indicatrice d'adoption (des standards de type ISO 14001) intervient comme variable explicative. Le modèle FILM autorise l'effet de cette variable à varier avec chaque autre variable explicative du modèle linéaire (en introduisant autant de termes en interaction que nécessaire).

Enfin, l'utilisation d'une variable alternative permettant de tenir compte des effets de taille (en l'occurrence, productivité du travail définie comme la VA sur le nombre de salariés) conduit à des résultats qualitativement similaires à ceux que nous venons de présenter, quelle que soit la méthode utilisée (appariement sur les scores de propension ou modèle FILM). En conclusion, l'analyse menée dans ce chapitre fournit des indices qui soutiennent l'hypothèse de Porter dans sa version faible, en France pour la période 2003-2006.

CHAPITRE 3 : IDENTIFICATION DES DETERMINANTS DE L'ECO-INNOVATION.

Le **Chapitre 3** s'intéresse au second point d'intérêt évoqué dans le Chapitre 1, à savoir le processus de diffusion de l'éco-innovation. En raison des contraintes posées par les données disponibles, cette investigation se fait du point de vue des déterminants de, et des barrières à, l'innovation environnementale. L'analyse repose sur l'estimation de fonctions de production d'innovation à l'aide de modèles à deux équations, du type :

(1.a)
$$z_i^* = \gamma . W_i + u_i$$

(1.b) $y_i^* = \beta . X_i + \varepsilon_i$

L'Equation (1.a) est une équation de sélection qui permet de tenir compte du fait que toutes les entreprises recensées dans un échantillon ne sont pas nécessairement innovantes. Cette équation est généralement spécifiée empiriquement à l'aide d'un modèle Probit:

(2)
$$\Pr(z_i = 1) = \Phi(\gamma, W_i)$$
 avec $z_i = 1 z_i^*$ is $z_i^* > 0$ et $z_i = 0$ si $z_i^* \le 0$

La seconde équation, (1.b), est dite équation d'*intensité* car elle permet d'expliquer l'intensité de l'éco-innovation, mesurée de différentes manières (voir ci-après). Ces modèles sont estimés de manière simultanée par Maximum de Vraisemblance sur les données européennes micro-anonymisées (ou "bruitées") de l'Enquête Communautaire sur l'Innovation de 2004 (CIS 2004). La base de données

rassemble 104943 entreprises observées à travers 15 pays, qui peuvent être répartis en quatre groupes: pays d'Europe du Nord et de l'Ouest (Allemagne, Belgique et Norvège), pays d'Europe Centrale et de l'Est (Bulgarie, Hongrie, République Tchèque, Roumanie, Slovaquie et Slovénie), pays méditerranéens ou d'Europe du Sud (Espagne, Grèce et Portugal) et pays baltes (Estonie, Lettonie, Lituanie).

Bien que l'enquête CIS 2004 ne soit pas spécifiquement centrée sur l'innovation environnementale, les données fournissent deux mesures approchées de l'intensité de l'éco-innovation: (1) une mesure de la "réduction de l'utilisation des consommations intermédiaires et de la consommation d'énergie par unité produite" (sur une échelle de 0 à 3) et (2) une mesure de la "réduction de l'impact environnemental ou sanitaire et amélioration de la sécurité" (également sur une échelle de 0 à 3).

Avec l'échelle ordonnée utilisée, une entreprise présentant une valeur non nulle de la première variable a réduit (plus ou moins intensément) ses consommations intermédiaires et/ou sa consommation d'énergie. De même, une entreprise présentant une valeur non nulle de la seconde variable a réduit (plus ou moins intensément) l'impact environnemental ou sanitaire de son activité. Le Tableau 4 présente, dans chaque industrie prise en compte par l'enquête CIS 2004, le pourcentage d'entreprises "éco-innovantes", définies comme présentant une valeur non nulle de l'une ou l'autre de nos deux mesures approchées.

	Nombre	% d'entreprises
	d'entreprises	éco-innovantes
<u>Services (secteur tertiaire) et autres secteurs</u>		
50: Vente et réparation de véhicules à moteur	1179	249(21,1%)
51: Commerce de gros	13669	2668 (19,5%)
52: Commerce de détail	2498	410 (16,4%)
60_61_62: Transport	4930	747 (15,2%)
63: Activités auxiliaires des transports	2206	540 (24,5%)
64: Postes et télécommunications	1155	447 (38,7%)
70: Immobilier et activités associées	728	111 (15,3%)
71: Location de machines et d'équipement	294	69 (23,5%)
72: Services informatiques	2589	1485 (57,4%)
73_74: R&D et autres activités	7897	2491 (31,5%)
E: Fourniture d'électricité, de gaz et d'eau	1921	517 (26,9%)
F: Construction	4353	855 (19,6%)
H: Hôtellerie et restauration	1424	211 (14,8%)
J: Intermédiation financière	2354	1053 (44,7%)
<u>Industrie (secteur secondaire)</u>		
20_21: Bois, pulpe et papier	4122	1148 (28%)
22: Imprimerie et édition	2474	811 (32,8%)
27: Métallurgie	1360	534 (39,3%)
28: Produits métalliques	4096	1710 (35%)
C: Exploitation des mines et carrières	1444	323 (22,4%)
DA: Agro-alimentaire, boissons et tabac	7907	2565 (32,5%)
DB: Textile	8905	1652 (18,5%)
DC: Cuir et peaux	1746	334 (19,1%)
DF_DG: Coke et produits du pétrole,		
Chimie et produits chimiques	2657	1578 (59,4%)
DH: Plastique et caoutchouc	2531	1059 (41,8%)
DI: Produits minéraux non ferreux	2855	1034 (36,2%)
DK: Machinerie et équipement	4315	2036 (47,2%)
DL: Equipement électrique et optique	4838	2418 (50%)
DM: Equipement de transport	2643	1192 (45%)
DN: Autres	3810	1282 (33,7%)
Total	103710	31529 (30,64%)

Tableau 4 – pourcentage d'entreprises éco-innovantes par secteur d'activité (enquête CIS 2004)

Pour l'analyse économétrique proprement dite, ces deux mesures peuvent être combinées pour obtenir (à l'aide d'une analyse factorielle) une mesure continue de l'intensité d'éco-innovation. Dans ce cas, le modèle économétrique estimé prend la forme d'un modèle Tobit généralisé. Les deux variables peuvent également être utilisées telle quelles, et séparément, ce qui conduit à estimer deux modèles économétriques dans laquelle l'équation de deuxième étape est un Probit ordonné.

	Effet fixe pay	/S	Variable d'e	xclusion
	Equation de sélection	Equation d'intensité	Equation de sélection	Equation d'intensité
Constante	0.232***	-0.933***	0.939***	-0.707***
	(0.043)	(0.044)	(0.037)	(0.027)
Taille (réf.: petite)				
Moyenne	0.174***	-0.001	0.112***	-0.034*
	(0.015)	(0.013)	(0.014)	(0.013)
Grande	0.198***	0.104***	0.11***	0.060***
	(0.021)	(0.017)	(0.021)	(0.016)
Filiale d'un groupe	0.097***	-0.032*	-0.012	-0.055***
	(0.016)	(0.012)	(0.016)	(0.012)
Marché principal: national	0.234***	-0.029*	0.253***	-0.042**
	(0.014)	(0.014)	(0.013)	(0.013)
Marché principal: international	0.284***	0.016	0.267***	0.028*
	(0.014)	(0.013)	(0.014)	(0.013)
Soutien financier de l'UE	0.53***	0.03	0.615***	0.031
	(0.072)	(0.022)	(0.069)	(0.023)
Activité de R&D continue	1.163***	0.015	1.087***	0.069***
	(0.029)	(0.021)	(0.028)	(0.018)
Intensité de R&D	2.003***	0.043	1.732***	-0.076
	(0.31)	(0.066)	(0.282)	(0.065)
Ouverture	-0.292***	0.158***	-0.297***	0.138***
	(0.004)	(0.004)	(0.004)	(0.004)
Coopération scientifique	0.018	0.087***	-0.030	0.078***
1	(0.054)	(0.018)	(0.051)	(0.019)
Coopération (autre)	1.53***	-0.031	1.631***	0.085***
	(0.037)	(0.022)	(0.034)	(0.022)
Ancien pays Soviétique	-	-	0.238***	× ,
			(0.013)	
Indicatrices sectorielles (stat de test)	2677.04***		2720.58***	
Indicatrices pays (stat de test)	4736.90***		-	
Nombre d'observations			64919	
Test du R.V. (d.l.)	9328.71(53)*** 6637.76(39)***		k	
Test de Wald (statistique)	7.6		0.29	
(Indépendance des deux équation du modèle)				
P-valeur du test d'endogénéité	0.008**		0.589	

Erreurs standard entre parenthèses sous chaque coefficient

****significatif au seuil de 1%, **significatif au seuil de 5%, *significatif au seuil de 10%

Les résultats obtenus avec la première méthode sont présentés dans le Tableau 5. Nos mesures approchées de l'éco-innovation ont toutes deux conduites à un unique facteur, avec une valeur-propre proche de 1. Ce résultat indique qu'une seule variable latente ("structurelle" selon la terminologie de l'analyse factorielle) explique la quasi-totalité (en l'occurrence, 91%) de la variance observée, avec un poids de 0,67. Nous utilisons alors les scores associés à la variable latente (autrement dit, au premier facteur) pour construire une mesure continue de l'intensité de l'éco-innovation. Cette mesure continue est utilisée comme variable dépendante du modèle Tobit généralisé évoqué dans le paragraphe précédent. Pour faciliter l'interprétation, nous avons normalisé notre mesure continue, avec une

moyenne de 0 et un écart-type de 1, de telle sorte qu'elle a une distribution normale centrée réduite. Le modèle Tobit généralisé est estimé avec et sans variable d'exclusion. L'utilisation d'une variable d'exclusion (corrélée avec la variable dépendante de l'équation de sélection, mais pas avec celle de l'équation d'intensité) peut permettre d'améliorer l'identification, mais il est en général difficile de trouver une variable propre à jouer ce rôle. Ici, la meilleure variable que nous ayons trouvée est une indicatrice de l'appartenance à un ancien pays soviétique – l'idée étant que ces pays accordaient une importance aussi grande à l'innovation que les pays de l'Ouest, alors que la sensibilisation à l'écologie y fut beaucoup plus tardive. Cette variable d'exclusion étant très imparfaite (en particulier parce qu'elle n'est pas mesurée au niveau de l'entreprise et parce qu'elle est parfaitement corrélée avec un éventuel effet fixe pays), nous avons également estimé le modèle sans y recourir (mais en ajoutant alors un effet fixe pays).

Estimés sur l'ensemble de l'échantillon, les deux variantes du modèle Tobit généralisé présentées dans le Tableau 5 suggèrent que les principaux facteurs associés à une plus grande intensité de l'éco-innovation sont la coopération avec des partenaires extérieurs et le degré d'ouverture de l'entreprise (au sens de Laursen et Salter, 2004; 2006). Par contraste, l'engagement continu dans une activité de R&D et l'intensité de R&D, s'ils sont bien associés à une probabilité d'innover (au sens usuel du terme) plus élevée, ne semblent pas influencer significativement l'intensité de l'écoinnovation. Ces résultats font sens dans la mesure où l'éco-innovation repose moins sur le développement de produits de haute technologie que sur l'adoption de procédés de production moins polluants - procédés qu'une entreprise peut acquérir en externe, auprès de partenaires plus expérimentés dans ce domaine, et qu'elle n'a pas forcément intérêt à développer en interne. Les analyses conduites sur les quatre sous-échantillons (groupes de pays) évoqués plus haut conduisent toutefois à nuancer ce résultat, dans la mesure où la R&D apparaît bien comme un déterminant de l'éco-innovation dans les pays d'Europe de l'Ouest et du Nord d'une part, et dans les pays d'Europe méditerranéenne d'autre part. Il se peut que les entreprises de ces deux groupes de pays jouent un rôle moteur dans le processus d'éco-innovation, alors que celles des pays d'Europe Centrale et de l'Est d'une part, et des pays baltes d'autre part, auraient une attitude plus attentiste.

Les résultats obtenus avec la seconde méthode (estimation par Maximum de Vraisemblance, pour chaque mesure approchée de l'intensité de l'éco-innovation, d'un modèle à équations simultanées combinant un Probit simple pour l'équation de sélection et un Probit ordonnée pour l'équation d'intensité) sont présentés dans les Tableaux 6.a (quand la mesure ordonnée de l'intensité d'innovation est la réduction des consommations intermédiaires et/ou d'énergie par unité produite) et 6.b. (quand la mesure ordonnée de l'intensité d'innovation est la réduction de l'intensité d'innovation est la réduction de l'intensité d'innovation est la réduction de l'impact environnemental ou sanitaire lié à l'activité de l'entreprise). Les résultats obtenus avec cette méthode alternative corroborent globalement ceux présentés plus haut, tout particulièrement quand l'intensité de l'éco-innovation est mesurée par la réduction des consommations intermédiaires et/ou d'énergie par

unité produite. Ils n'appellent donc pas à davantage de commentaires, en-dehors du fait que le rôle de la R&D (déjà commenté plus haut apparaît plus déterminant quand l'intensité de l'éco-innovation est mesurée par la réduction de l'impact environnemental ou sanitaire lié à l'activité de l'entreprise).

	Effet fixe pays		Variable d'exclusi	on
	Equation de	Equation	Equation de	Equation
	sélection	d'intensité	sélection	d'intensité
Constante	0.230***		0.941***	
	(0.043)		(0.037)	
Taille (réf.: petite)				
Moyenne	0.176***	-0.001	0.112***	-0.003*
	(0.015)	(0.017)	(0.014)	(0.017)
Grande	0.202***	0.097***	0.109***	0.119***
	(0.021)	(0.022)	(0.021)	(0.021)
Filiale d'un groupe	0.099***	-0.032*	-0.01	-0.098***
	(0.016)	(0.016)	(0.016)	(0.016)
Marché principal: national	0.238***	0.012	0.253***	0.008
* *	(0.014)	(0.019)	(0.013)	(0.017)
Marché principal: international	0.284***	0.035*	0.267***	0.016
* *	(0.015)	(0.018)	(0.014)	(0.017)
Soutien financier de l'UE	0.522***	0.126***	0.612***	0.147***
	(0.072)	(0.030)	(0.069)	(0.03)
Activité de R&D continue	1.163***	0.04	1.087***	-0.017
	(0.029)	(0.03)	(0.028)	(0.029)
Intensité de R&D	1.998***	0.011	1.734***	0.118
	(0.31)	(0.089)	(0.281)	(0.086)
Ouverture	-0.292***	0.165***	-0.297***	0.164***
	(0.004)	(0.007)	(0.004)	(0.009)
Coopération scientifique	0.012	0.137***	-0.032	0.109***
1 1	(0.054)	(0.023)	(0.051)	(0.023)
Coopération (autre)	1.53***	0.024	1.628***	0.001
	(0.037)	(0.033)	(0.034)	(0.04)
Ancien pays Soviétique	_	-	0.240***	
r - J			(0.013)	
Indicatrices sectorielles	2335.83***		2460.98***	
Indicatrices pays	3289.64***			
Nombre d'observations	65058		65058	
Test du R.V. (d.l.)	15297.01(106)**	*	14538.38(79)*	
			**	

Tableau 6.a - Déterminants de l'éco-innovation (modèle Probit ordonné avec équation de sélection, variable ordonnée = réduction des consommations intermédiaires et/ou d'énergie par unité produite).

Erreurs standard entre parenthèses sous chaque coefficient

****significatif au seuil de 1%, **significatif au seuil de 5%, *significatif au seuil de 10%

	Effet fixe pays		Variable d'excl	usion
	Equation de sélection	Equation d'intensité	Equation de sélection	Equation d'intensité
Constante	0.235***		0.939***	
	(0.043)		(0.037)	
Taille (réf.: petite)				
Moyenne	0.175***	0.001	0.114***	-0.078***
	(0.015)	(0.017)	(0.014)	(0.017)
Grande	0.200***	0.132***	0.111***	-0.017
	(0.021)	(0.022)	(0.021)	(0.021)
Filiale d'un groupe	0.098***	-0.043**	-0.009	-0.016
	(0.016)	(0.016)	(0.016)	(0.015)
Marché principal: national	0.237***	-0.042*	0.256***	-0.084***
	(0.014)	(0.019)	(0.013)	(0.017)
Marché principal: international	0.283***	0.020	0.265***	0.047**
	(0.015)	(0.018)	(0.014)	(0.017)
Soutien financier de l'UE	0.526***	-0.051	0.612***	-0.081**
	(0.072)	(0.030)	(0.069)	(0.03)
Activité de R&D continue	1.167***	0.2	1.087***	0.160***
	(0.029)	(0.03)	(0.028)	(0.029)
Intensité de R&D	1.989***	0.044	1.749***	-0.241**
	(0.31)	(0.089)	(0.284)	(0.083)
Ouverture	-0.294***	0.180***	-0.297***	0.137***
	(0.004)	(0.007)	(0.004)	(0.006)
Coopération scientifique	0.014	0.021	-0.033	0.025
	(0.054)	(0.023)	(0.051)	(0.022)
Coopération (autre)	1.53***	-0.033	1.628***	0.186***
-	(0.037)	(0.033)	(0.034)	(0.026)
Ancien pays Soviétique	-	-	0.242***	
			(0.013)	
Indicatrices sectorielles	1660.79***		1706.23***	
Indicatrices pays	67252.82***		-	
Nombre d'observations	65058		65058	
Test du R.V. (d.l.)	122495.55(106)***		13927.95(79)***	

Tableau 6.b - déterminants de l'éco-innovation (modèle Probit ordonné avec équation de sélection, variable ordonnée = réduction de l'impact environnemental ou sanitaire lié à l'activité de l'entreprise.)

Erreurs standard entre parenthèses sous chaque coefficient

****significatif au seuil de 1%, **significatif au seuil de 5%, *significatif au seuil de 10%

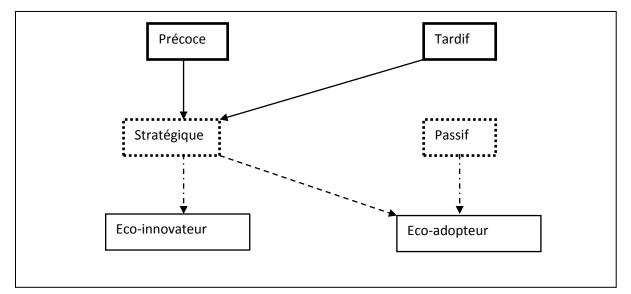
CHAPITRE 4 : ETUDE DES TYPES DE COMPORTEMENTS D'ECO-INNOVATION A L'AIDE DE DONNEES MICRO-ANONYMISEES

Le **Chapitre 4** étend l'analyse conduite dans le Chapitre 3, en exploitant les données "bruitées" des Enquêtes Communautaires sur l'Innovation de 2008 (CIS 2008), qui contiennent un module spécifiquement dédié à l'innovation environnementale. Les modèles économétriques estimés dans ce chapitre restent similaires à ceux utilisés dans le Chapitre 3, mais l'exploitation des données CIS 2008 permet de changer l'angle d'approche. Ainsi, plutôt qu'essayer de mesurer approximativement l'intensité de l'éco-innovation, nous établissons à l'aide du module spécifique de l'enquête une

typologie originale permettant (comme indiqué dans la Figure 3) de distinguer, parmi les entreprises éco-innovantes: (1) les entreprises adoptant une éco-innovation de manière passive (*passive eco-adopters*), (2) les entreprises adoptant tardivement une éco-innovation de manière stratégique (*late strategic eco-adopters*), (3) les entreprises adoptant précocement une éco-innovation de manière stratégique (*late strategique (early strategic eco-adopters*), (4) les éco-innovateurs stratégiques tardifs (*late strategic eco-innovators*) et (5) les éco-innovateurs stratégiques précoces (*early strategic eco-innovators*).

Ces cinq types d'entreprises éco-innovantes suivent une gradation, de la moins éco-innovante à la plus éco-innovante. Ainsi, les entreprises adoptant une éco-innovation de manière passive (type (1) dans notre classification) n'ont pas développé elles-mêmes une éco-innovation, mais ont adopté – sans avoir de stratégie d'éco-innovation spécifique – un produit, procédé de production ou méthode d'organisation ayant des effets bénéfiques sur l'environnement. Les entreprises adoptant tardivement une éco-innovation de manière stratégique (type (2) de notre classification) n'ont pas développé ellesmêmes une éco-innovation, mais ont adopté entre 2006 et 2008 - et après avoir mis en place des procédures spécifique pour évaluer et réduire leur empreinte environnementale - une innovation développée par d'autres et ayant des effets bénéfiques sur l'environnement. Les entreprises adoptant précocement une éco-innovation de manière stratégique (type (3) de notre classification) n'ont pas développé elles-mêmes une éco-innovation, mais ont adopté avant 2006 - et après avoir mis en place des procédures spécifique pour évaluer et réduire leur empreinte environnementale – une innovation développée par d'autres et ayant des effets bénéfiques sur l'environnement. Les éco-innovateurs stratégiques tardifs (type (4) de notre classification) ont introduit une éco-innovation entre 2006 et 2008, après avoir mis en place des procédures spécifique pour évaluer et réduire leur empreinte environnementale. Les éco-innovateurs stratégiques précoces (type (5) de notre classification) ont introduit une éco-innovation avant 2006, après avoir mis en place des procédures spécifique pour évaluer et réduire leur empreinte environnementale. Ces cinq catégories d'entreprises éco-innovantes s'opposent aux entreprises innovantes au sens conventionnel du terme mais non éco-innovantes, toutes se distinguent des entreprises non innovantes.

Figure 3 : construction de notre typologie d'entreprises éco-innovantes



Les données micro-anonymisées de l'enquête CIS 2008 rassemblent 79972 entreprises observées à travers 11 pays, qui peuvent être répartis en quatre groupes: pays d'Europe du Nord et de l'Ouest (Allemagne et Irlande), pays d'Europe Centrale et de l'Est (Bulgarie, Hongrie, République Tchèque, Roumanie et Slovaquie), pays méditerranéens ou d'Europe du Sud (Chypre et Portugal) et pays baltes (Estonie et Lituanie). Le Tableau 7 fournit la proportion d'entreprises éco-innovantes (tous types confondus) par pays, et le Tableau 8 donne la proportion de chaque type d'entreprises éco-innovantes par secteur d'activité pour l'ensemble de la base de données.

Pays	Nombre d'observations	% d'entreprises éco-innovantes
Allemagne	6026	52.7%
Bulgarie	31718	7.31%
Chypre	1024	15.82%
Estonie	3986	30.76%
Hongrie	5390	21.04%
Irlande	2178	42.1%
Lituanie	2111	22.64%
Portugal	6512	58.71%
République Tchèque	6804	48.5%
Roumanie	9631	32.79%
Slovaquie	4592	24.35%
Total	79972	26.02%

Tableau 7 – proportion d'entreprises éco-innovantes par pays

Notre objectif dans ce chapitre est d'identifier des facteurs spécifiques à un type d'entreprise éco-innovante. Pour ce faire, nous estimons par Maximum de Vraisemblance des modèles comprenant une équation de sélection (pour tenir compte des questions-filtres distinguant, dans les enquêtes CIS, les entreprises innovantes des non-innovantes) et une équation de type Probit ordonné (dans laquelle la variable dépendante est le type d'entreprise). Cette méthode correspond à la seconde approche utilisée dans le chapitre précédent. Les résultats des estimations sont fournis dans le Tableau 9.

Comme dans le Chapitre 3, le degré d'ouverture des entreprises et la coopération avec des partenaires extérieurs apparaissent prédominants, et ce quel que soit le type d'éco-innovateur considéré. Toutefois, la R&D joue ici un rôle aussi important que ces deux facteurs (là encore quel que soit le type d'entreprise considéré). Il est donc *a priori* difficile d'identifier les spécificités de chaque type d'éco-innovateur, bien que les éco-innovateurs stratégiques précoces (les entreprises les plus éco-innovantes de notre typologie) soient plus susceptibles de recourir à l'adoption de normes environnementales volontaires (de type ISO14001). Par contraste, les entreprises modérément éco-innovantes (*late strategic eco-adopters, early strategic eco-adopters* et *late strategic eco-innovators*) ont besoin d'encouragements financiers, qui peuvent prendre la forme d'un soutien de l'UE. Comme dans le Chapitre 3, l'estimation sur des sous-échantillons permet de nuancer ces résultats, en soulignant notamment que l'importance de la coopération avec des partenaires extérieurs varie d'un groupe de pays à l'autre.

	Nombre	Non éco-	Passive eco-	Late strategic
Codes NACE	d'entreprises	innovantes	adopter	eco-adopter
(B): Mines et Carrières	1038	28.42%	12.04%	4.43%
(C10-C12): Agro-alimentaire, boissons et tabac	6080	28.78%	12.38%	3.22%
(C13-C15): Textiles et cuir	7855	27.4%	7.49%	4.03%
(C16-C18): Bois, papier et imprimerie	4400	29.84%	13.75%	3.7%
(C19-C23): Coke, chimie et produits chimiques	6331	25.44%	15.28%	3.22%
(C24-C25): Métallurgie et produits métalliques	5344	29.67%	15.02%	3.79%
(C26-C30): TIC, électronique et machinerie	6711	27.67%	18.89%	4.7%
(C31-C33): Mobilier, réparation et installation de machinerie	4598	32.92%	14.28%	4.52%
(D): Fourniture d'électricité, de gaz, de vapeur et d'air conditionné	1224	33.33%	15.68%	5.39%
(E): Fourniture d'eau	2259	23.37%	13.28%	5.62%
(F): Construction	2202	60.94%	8.81%	8.9%
(G): Commerce de gros et détail	13815	25.8%	7.49%	4.86%
(H49-H51): Transport	4748	24.22%	9.49%	2.86%
(H52-H53): Stockage, soutien au transport, postage	1856	37.44%	9.21%	5.6%
(I): Services de logement et de restauration	164	52.43%	31.7%	1.21%
(J58-J60): Publication, radio- et élédiffusion, vidéo	1119	41.28%	14.29%	4.02%
(J61-J63): Télécommunications, informatique et consultance	3160	42.12%	13.13%	4.33%
(K): Finance et assurance	2534	43.21%	12.74%	3.23%
(L): Immobilier	75	60%	24%	6.66%
(M69-M70): Direction, droit, comptabilité	450	65.11%	20.88%	2%
(M71-M73): R&D, Architecture et ingénierie	3070	40.84%	15.47%	4.23%
(M74-M75): Autres activités scientifiques et techniques	112	55.35%	30.35%	0.89%
(N): Activités administratives	827	53.08%	21.88%	2.29%
Total	79972	31.05%	12.33%	4.23%

Tableau 8 – pourcentage de chaque type d'entreprises éco-innovantes par secteur d'activité

Tableau 8, suite (trois dernières colonnes)

Codes NACE	Early strategic eco-	Late strategic eco-	Early strategic eco-	
Could NACE	adopter	innovator	innovator	
(B): Mines et Carrières	13%	9.63%	8.57%	
(C10-C12): Agro-alimentaire, boissons et tabac	7.46%	6.61%	7.4%	
(C13-C15): Textiles et cuir	9.12%	2.71%	3.95%	
(C16-C18): Bois, papier et imprimerie	9.06%	7.27%	7.86%	
(C19-C23): Coke, chimie et produits chimiques	8.33%	10.61%	11.64%	
(C24-C25): Métallurgie et produits métalliques	7.41%	8.85%	8.15%	
(C26-C30): TIC, électronique et machinerie	9.17%	10.04%	13.96%	
(C31-C33): Mobilier, réparation et installation de machinerie	8.61%	6.17%	6.65%	
(D): Fourniture d'électricité, de gaz, de vapeur et d'air conditionné	11.02%	9.55%	11.51%	
(E): Fourniture d'eau	12.12%	11.81%	15.49%	
(F): Construction	7.44%	7.31%	6.58%	
(G): Commerce de gros et détail	10.93%	4.14%	4.02%	
(H49-H51): Transport	7.79%	3.83%	3.39%	
(H52-H53): Stockage, soutien au transport, postage	12.23%	6.25%	6.73%	
(I): Services de logement et de restauration	1.82%	5.48%	7.31%	
(J58-J60): Publication, radio- et télédiffusion, vidéo	9.2%	3.03%	3.48%	
(J61-J63): Télécommunications, informatique et consultance	9.43%	3.51%	3.1%	
(K): Finance et assurance	15.78%	3.78%	3.7%	
(L): Immobilier	2.66%	2.66%	4%	
(M69-M70): Direction, droit, comptabilité	1.11%	1.77%	0.88%	
(M71-M73): R&D, Architecture et ingénierie	8.69%	5.96%	6.02%	
(M74-M75): Autres activités scientifiques et techniques	2.67%	5.35%	2.67%	
(N): Activités administratives	3.38%	4.83%	5.19%	
Total	9.29%	6.3%	6.96%	

Variables	Equation de	Type d'entreprise						
	sélection	Early strategic	Late strategic	Early strategic	Late strategic	Passive eco-	Non éco-	
		eco-Innovator	eco-Innovator	eco-adopter	eco-adopter	adopter	innovante	
Filiale d'un	0.065**	0.015***	0.024***	0.008***	0.003***	0.001	-0.051***	
groupe	(0.027)	(0.002)	(0.002)	(0.001)	(0.0004)	(0.001)	(0.006)	
Taille : Moyenne	0.12***	0.02***	0.031***	0.01***	0.004***	0.001	-0.067***	
	(0.025)	(0.002)	(0.002)	(0.001)	(0.0004)	(0.001)	(0.006)	
Grande	0.25***	0.06***	0.077***	0.022***	0.009***	-0.022***	-0.145***	
	(0.045)	(0.004)	(0.004)	(0.001)	(0.0005)	(0.003)	(0.007)	
Intensité de R&D	-0.0002***	0.0003***	0.0005***	0.0002***	-0.0008***	-0.0006**	-0.0012***	
	(0.0000)	(0.0002)	(0.0003)	(0.0001)	(0.0009)	(0.0001)	(0.0001)	
R&D Continue	0.700***	0.01***	0.017***	0.005***	0.002***	0.0002	-0.036***	
	(0.055)	(0.0002)	(0.003)	(0.001)	(0.0005)	(0.0006)	(0.007)	
Ouverture	-0.076***	0.003***	0.005***	0.002***	0.001***	0.0006**	-0.012***	
	(0.004)	(0.0003)	(0.0005)	(0.0002)	(0.0001)	(0.0002)	(0.001)	
Coopération	-0.040	0.013***	0.021***	0.006***	0.003***	-0.0003	-0.043***	
scientifique	(0.077)	(0.003)	(0.004)	(0.001)	(0.0005)	(0.0009)	(0.008)	
Autre coopération	0.930***	0.015***	0.023***	0.007***	0.003***	0.0004	-0.049***	
	(0.050)	(0.002)	(0.003)	(0.001)	(0.0005)	(0.0008)	(0.007)	
Marché	0.210***	0.007***	0.013***	0.004***	0.002***	0.001*	-0.028***	
international	(0.024)	(0.001)	(0.002)	(0.001)	(0.0004)	(0.0006)	(0.005)	
Marché national	0.210***	-0.005**	-0.009**	-0.003**	-0.001**	-0.0006	0.02**	
	(0.026)	(0.002)	(0.003)	(0.001)	(0.0004)	(0.0003)	(0.006)	
Soutien public	0.280***	0.001**	0.016**	0.005**	0.002**	-0.0002	-0.033**	
(UE)	(0.087)	(0.003)	(0.004)	(0.001)	(0.0006)	(0.0007)	(0.009)	
Adoption de	0.520***	0.084***	0.011***	0.03****	0.012***	-0.034***	-0.198***	
normes	(0.034)	(0.003)	(0.003)	(0.001)	(0.0006)	(0.003)	(0.006)	
facultatives								
Réglementation	0.48***	0.001***	0.123***	0.034***	0.014***	-0.033***	-0.237***	
	(0.031)	(0.003)	(0.003)	(0.001)	(0.007)	(0.004)	(0.006)	

Tableau 9 – déterminants du type d'entreprises éco-innovantes (effets marginaux, modèle Probit multivarié avec équation de sélection)

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