
High-growth firms and innovation in European countries

Este artículo analiza el efecto de la innovación en una empresa de rápido crecimiento (HGF). Los microdatos pertenecen a la *Community Innovation Survey 2008*, facilitada por Eurostat, y cubre el período 2006-2008 para 15 países europeos. Clasificamos estos países en dos grupos de acuerdo al gasto empresarial en I+D sobre el PIB: los líderes (Alemania, Eslovenia, República Checa, Noruega, Portugal, España e Italia) y los rezagados (Estonia, Hungría, Eslovaquia, Lituania, Rumanía, Bulgaria, Letonia y Chipre). Las empresas en los países líderes invierten más en I+D, pero la presencia de HGF es más moderada que en los países rezagados. Nuestros principales resultados muestran que los determinantes de la innovación y de ser una empresa HGF son diferentes entre los países europeos. En los países líderes, la presencia de empresas HGF está relacionada con las inversiones en I+D y la innovación, mientras que, en los países rezagados, depende directamente del tamaño de la empresa y la tasa de turbulencia del conjunto de la economía. En una visión de conjunto, nuestros resultados arrojan luz sobre los diferentes ecosistemas empresariales en los países de la Unión Europea.

Artikulu honek aztertzen du hazkunde azkarreko enpresan (HGF ingelesez) berrikuntzak duen eragina. Mikro-datuak Eurostatek emandako Berrikuntzari buruzko 2008ko Erkidego Inkestakoak dira. Inkesta horrek 2006-2008 aldia jaso-tzen du Europako 15 herrialdeetarako. Europako herrialdeak bi taldetan sailkatu ditugu BPGaren gainean enpresek egiten duten G+B gastuaren arabera: buruan dauden herrialdeak (Alemania, Eslovenia, Txekiar Errepublikak, Norvegia, Portugal, Espainia eta Italia) eta atzean dauden herrialdeak (Estonia, Hungaria, Eslovakia, Lituania, Errumania, Bulgaria, Letonia eta Zipre). Buru diren herrialdeetan, HGBko enpresen agerpena B+G eta berrikuntza inbertsioekin erlazionatuta dago; aitzitik, herri atzeratuetan, agerpen hori enpresak duen tamaninaren eta ekonomia osoaren nahaste-tasaren araberakoa da. Oro har ikusita, gure emaitzak lagungarri dira Europar Batasuneko herrialdeetako enpresa-ekosistemak ezagutzeko.

This paper analyses the effect of innovation on a high-growth firm (HGF). The micro-data belongs to the *Community Innovation Survey 2008* provided by Eurostat covering the period 2006-2008 for 15 European countries. We classify the EU countries in two groups according to the share of business R&D on GDP: leader countries (Germany, Slovenia, Czech Republic, Norway, Portugal, Spain and Italy) and laggard countries (Estonia, Hungary, Slovakia, Lithuania, Romania, Bulgaria, Latvia and Cyprus). Firms in leader countries are more prone to invest in R&D but the presence of HGFs is more moderated than firms in laggard countries. Our main results show that the drivers to innovate and become a HGF differ across European countries. In leader countries, the HGF firms presence is related to R+D inversions and innovation, whereas in laggard countries depend directly on the size of the firm, as well as the turbulence rate in the economy as a whole. All in all, our results show light on the different entrepreneurial ecosystems of the European Union member countries.

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Palabras clave: Empresas de rápido crecimiento, crecimiento empresarial, actividad innovadora.

Keywords: High-growth firms, firm growth, innovation activity.

JEL codes: L11, L25, O30.

1. INTRODUCTION

During last decades, Europe has performed deficiently in generating innovative high-growth firms (henceforth HGFs) that quickly become global leaders in comparison with other economies located in the technological frontier. Recently, this gap has generated an increasing concern between European institutions. Accordingly, policymakers have shown increasing interest in fostering fast growing innovative firms as they are seen as a key driver of economic growth and employment. Hence, HGFs have attracted increasing interest since it is suggested that they contribute significantly to create new jobs, to foster the industrial productivity and to ensure a sustainable aggregate economic growth.

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In line with this, the Horizon 2020 framework proposes a new enterprise policy which adopts a systematic approach in order to foster SMEs' capacity to innovate and to generate new jobs. The new European enterprise policy aims to generate environmental factors that promote firm's competitiveness which drive productivity growth, internationalization, innovation and investment in order to create jobs with higher levels of education¹.

Concerning the effects of HGFs on the employment growth and the economic growth, scholars have paid attention to the tent-shaped distribution of firm growth where a small group of firms located in the heavy tails grows faster than their counterparts (Bottazzi and Secchi, 2006; Bottazzi *et al.*, 2011). This group of firms has attracted the attention of researchers due to their economic contributions. First, they create most new jobs (Birch and Medoff, 1994; Davidsson and Henrekson, 2002; Delmar *et al.* 2003; Acs and Mueller, 2008; Acs, 2011; Coad *et al.*, 2014a; Daunfeldt and Halvarsson, 2015). Second, they exert spillover effects which are beneficial to the growth of other firms (Mason *et al.*, 2009). Third, HGFs contribute to the creation of knowledge (Colombelli *et al.*, 2014)². Fourth, from a social point of view, they employ the young, less educated, immigrants and long-term unemployed individuals (Coad *et al.*, 2014b).

Consequently, policymakers have created different initiatives to support HGFs in order to capture their potential capacity to be a driver of job creation, innovation and economic growth (Acs *et al.*, 2008; OECD, 2002). However, the capacity that an economy has to reinforce HGFs is limited. In fact, at European level the different initiatives to promote the presence of HGFs in Europe have failed to catch up the share of HGFs in US. This difference may be in part explained by the lack of young innovative companies (YICs) which largely become HGFs (Veugelers and Cincera, 2015). In fact, the relationship between HGFs and YICs has been recently pointed out by Decker *et al.* (2016)³. However, the existence of imperfect information may cause government failures by focusing their policies in a selected group of «winners» to the detriment of all SMEs.

¹ According to Joint Declaration on a horizontal EU SME policy, 5 April 2016, there are more than 21 million SMEs in the EU, that represent 99.8% of all European firms, 67% of employment and 58% of gross added value.

² Using data for 335 firms from UK, Germany, France, Sweden, Italy, and the Netherlands between 1988 and 2005, Colombelli *et al.* (2014) analyse the impact of HGFs in the knowledge creation process. These authors find that «HGFs are key actors in the creation of new technological knowledge, and showing also that firms that achieve higher than average growth focus on exploration based on familiar technology.»

³ According to Decker *et al.* (2016), «if rapid firm-level growth reflects efficient movement of labour toward high-productivity producers, then reductions in the number and impact of such firms may be a cause for concern.»

Given the current interest of Europe to promote HGFs and innovation⁴, there seems necessary to show light on the relationship between innovation and firm growth. In fact, it is crucial to analyse the pattern of HGFs across countries (Bravo-Biosca, 2010, 2011). A scarce number of studies have tackled with the behaviour of HGFs at country level. Some exceptions are Schreyer (2000); Bravo-Biosca (2010, 2011); Hölzl (2009), Navaretti *et al.* (2014) and Teruel and de Witt (forthcoming). However, the majority of these works have aggregated level. Hence, an analysis at firm level may be more informative of the firm performance and its linkages at macroeconomic level.

In line with Hölzl (2009), we analyse the data from Community Innovation Survey (henceforth CIS). However, here we consider the simultaneous behaviour of research, innovation and HGFs at firm level. The entrance in the EU of new countries with a technological gap may accelerate their process of technological catch-up, but also their growth activity. Hence, we aim to analyse HGFs in countries with a large investment in R&D in comparison with those with a lower level. Our assumption is that the heterogeneous market structure and R&D effort among European countries have generated HGFs with different R&D and innovation patterns.

With this purpose in mind, we focus on the behaviour of two groups of countries according to R&D investment effort. Here, we aim to analyse the different behaviour of HGFs in reference with their innovation activity. Given the recent findings from Decker *et al.* (2016) and Audretsch *et al.* (2014), we may expect that there are unobserved characteristics which affect simultaneously the innovation performance and the probability of becoming a HGF.

Our database is drawn from the CIS between 2006 and 2008 for 15 European countries. After the dataset treatment, our sample contains 67,279 firms. According to the features of our data, we apply a biprobit model to take into account the simultaneity between the innovation output and the probability that a firm becomes a HGF. With this methodology we control for the unobserved characteristics that may potentially affect simultaneously that a firm becomes a HGF, but it also innovates. Our results show that the drivers of HGFs in countries with a low business R&D effort differ from HGFs located in countries with a high business R&D effort.

The article makes several contributions. First, we use a database that covers 15 European countries that allow us to observe the differences between high and low R&D intensive countries. Second, we consider the simultaneous relationship between the innovation inputs and the innovation outputs on the probability to become a HGF.

The structure of the article is the following. The second section reviews the empirical literature of HGFs. The third section presents our database and the

⁴ See European Commission (2011).

main statistical descriptive. The fourth section shows the econometric methodology. The fifth section reports our main results and the final section presents our concluding remarks.

2. LITERATURE REVIEW

2.1. HGFs: concept and stylized facts

Birch's (1979) work was the starting point to observe the contribution of a group of firms which were contributing more than their counterparts. According with his findings, small firms were contributing more to the job creation. Despite the criticisms to his work (see Haltiwanger *et al.*, 2013), his research constitutes a point of reference in the literature of HGFs. Furthermore, the tent-shaped distribution of firm growth has risen the attention to a small group of firms located in the heavy tails that grow faster than their counterparts (Bottazzi and Secchi, 2006; Bottazzi *et al.*, 2011).

The wide interest in the phenomenon of HGFs has generated that the delimitation of the concept is far from easy. In that sense, Parker *et al.* (2010) point out the lack of a commonly accepted denomination used for 'high-growth' firms. In this regard, the literature has referred to fast-growth firms (Deutschmann, 1991; Storey, 1994; Almus, 2002); high-growth impact firms (Acs *et al.*, 2008), high-growth firms (Schreyer, 2000), «superstar» fast-growth firms (Coad and Rao, 2008), rapidly expanding firms (Schreyer, 2000), and gazelles (Birch, 1981, among others).

At empirical level, there are also differences. First, firm growth is a multidimensional phenomenon (Delmar *et al.*, 2003) which may be measured in terms of sales, employment, profit, productivity and added value. Second, HGFs are identified according with different measures. They may be identified as a certain share of the fastest growing firms (often 5% or 10%) during a period, using the Eurostat-OECD measure which considers HGFs as firms with at least ten employees in the starting year, and an annualized employment growth larger than 20% during a 3-year period (Eurostat-OECD, 2007), or the Birch index which is a mixture between absolute and relative growth rates (Birch, 1981). Finally, the evidence shows that firms classified as HGFs with one measure may differ according with another measure (Daunfeldt *et al.*, 2014). Hence, each variable and measure has advantages and disadvantages depending on the policy focus and they will be more appropriate according with the purpose of analysis.

Synthetically, the main stylized facts of the HGFs phenomena are the following⁵: 1) a small share of firms become HGFs; 2) they are present at all sectors (Schreyer,

⁵ For a review, see Henrekson and Johansson (2010), Coad *et al.* (2014a) and Moreno and Coad (2015), among others.

2000); 3) they are more present among young firms⁶; 4) there is a low persistence of HGFs (Delmar *et al.* 2013)⁷ or in other words HGFs are «one hit wonders» (Daunfeldt and Halvarsson, 2015); 5) small HGFs tend to have more organic growth, while large HGFs grow more with mergers and acquisitions; 6) they are more R&D intensive (Segarra and Teruel, 2014; Coad *et al.*, 2016)⁸; 7) HGFs usually export more than they counterparts (Parsley and Halabisky, 2008; Mason and Brown, 2010); 8) they show a larger internationalization and integration in global value chains (Mason and Brown, 2010; Du and Temouri, 2015); 9) they have more human capital (Daunfeldt *et al.*, 2015).

2.2. The empirical evidence at country level

According to the Schumpeterian theory of creative destruction, HGFs may be a revulsive for the innovation and growth of countries. Their capacities to generate new jobs and to exploit their competitive advantages represent a shake-out in the market distribution. Consequently, policymakers have focused their attention in HGFs. However, according with a recent survey from Mason and Brown (2013) and Brown and Mawson (2015), the theoretical basis that have generated current public governmental policies are supported on incorrect theories⁹. The authors suggest that policies should base on the «dynamic capabilities» instead of the traditional resource-based views. They suggest that «growth accelerators should become much less resource based and more ‘competency-based’. Therefore, assistance to help with the external orientation of the firm will be important». In part, this mistake is due to the high potential growth of high-tech sectors (see Daunfeldt *et al.*, 2015).

While promoting HGFs may be difficult at country level, the challenge is still more prominent for wider regions such as the EU. For instance, the European Commission has applied policies to SMEs HGFs¹⁰. In an economic context where countries differ in terms of their technological gap, their economic growth and their institutional and market structures, it is necessary to analyse the differences of HGFs at country level. According to Daunfeldt *et al.* (2015) the fact that «conditions may differ across countries and over time» may cause the disparity of non-homogenous results of the impact of R&D on firm growth. The issue is relevant given that the allocation of the

⁶ In fact, authors such as Schreyer (2000), Delmar *et al.* (2003), Haltiwanger *et al.* (2013) and Daunfeldt *et al.* (2014) have pointed that it is age and not size the variable which most affect firm growth.

⁷ However, Hölzl (2014) points out that it depends on the growth measure.

⁸ According with Mazzucato and Parris (2015), «HGF have the most to gain from increasing their R&D intensity. However, the benefits of investing in R&D are conditional on the competitive environment, even for firms in the top growth quantiles».

⁹ The authors point out that there are some misconceptions based on the fact that «HGFs will predominantly emerge from the stock of high-tech firms within an economy» and the fact that «high growth potential NTBFs confront certain ‘market failures’».

¹⁰ See European Commission (2010, 2011).

scarce public budget must be addressed to the most convenient firms. However, Bravo-Biosca (2010, 2011) points out to the necessity that policies must address structural reforms at country level that remove barriers to entry and growth (product, labour, land and financial barriers)¹¹ to overcome differences across countries.

From a territorial perspective, SMEs become HGFs more frequently in innovative ecosystems such as clusters and other business networks promoting innovation and value chains. Hence, new innovation policies can stimulate the appearance of HGFs locally by supporting firm's initiatives and sectorial clusters activities to drive greater growth through collaborative actions.

Consequently, it seems necessary to adopt a country level approach to evaluate HGFs. However, the majority of the empirical evidence has focused in a particular country, while scarce contributions have analysed the behaviour of HGFs across countries. The most outstanding articles are those from Schreyer (2000), Hölzl (2009), Bravo-Biosca (2010, 2011), Navaretti *et al.* (2014) and Teruel and de Wit (*forthcoming*).

Using data from five OECD countries and Quebec, Schreyer (2000) analyses the pattern of HGFs at industry level between 1980s and 1990s. His results show that HGFs are more technology intensive than the average firm. Furthermore, this author observes that HGFs are found in all industries and regions. Concerning the R&D effort, HGFs are more R&D intensive. Finally, he confirms that HGFs account for a disproportionately large share of job creation.

Later Bravo-Biosca (2010, 2011) analyses the industrial behaviour of 12 OECD countries between 2002 and 2005. He focuses on the relationship between TFP growth and the dynamics of the growth distribution. He finds two interesting findings. First, countries with larger share of firms which remain static show a lower productivity growth in a country. Second, countries with a higher share of shrinking and growing firms show a faster productivity growth. Both authors, Schreyer (2000) and Bravo-Biosca (2010, 2011), observe a stylized fact in EU: firms are more static in EU than in US. According with these authors, this is the reason why Europe shows a lower productivity growth at the aggregate level.

Teruel and de Witt (*forthcoming*) explore data from 17 OECD countries between 1999 and 2005. They focus on the incidence of macroeconomic determinants of three driving forces of high growth, such as the entrepreneurship, institutional settings, and opportunities for growth. Results highlight the importance of the entrepreneurship to increase the presence of HGFs in a country and the existence of institutional obstacles such as the labour market protection and the administrative barriers.

¹¹ Bravo-Biosca (2010) considers that the causes of differences across countries are related to the role of institutions and the barriers of growth.

Due to the inherent difficulties in accessing business-level data simultaneously in several countries, only Hölzl (2009) explores the behaviour of HGFs at firm level for different countries. Using data from the CIS for 16 countries for the period 1998-2000, this author analyses the determinants across countries that a firm becomes a HGF. After applying a matching procedure, he estimated quantile regressions to analyse the determinants of firm growth. The main result is that HGFs show a larger R&D intensity than non-HGFs in countries closer to the technological frontier.

More recently, for a sample of French, Italian and Spanish manufacturing firms with more than ten employees in the period from 2001 to 2008, Navaretti *et al.* (2014) apply a quantile methodology to analyse the determinants of firm growth. These authors find that the number of employees in R&D activities and graduates is positively correlated with the firm growth in the largest quantiles, while product and process innovations only have a significant positive incidence for the lowest quantiles.

Our database is similar to Hölzl (2009) but with more restricted information and for the period 2006-2008. Nevertheless, we consider the unobserved characteristics that may potentially affect simultaneously that a firm becomes a HGF, but it also innovates. As we have seen previously, there is empirical evidence that HGFs show a larger R&D and innovation intensity. However, there is no evidence on the underlying relationship. In that sense, we consider that HGFs depend on the innovation activity, and where their capacity to innovate depends on the R&D activity.

3. DATABASE AND STATISTICAL DESCRIPTIVE

3.1. CIS data and country clusters criteria

The empirical application was carried out using the Community Innovation Survey (CIS), in particular, we use the CIS 2008 wave which covers the period 2006-2008. The CIS is a harmonized survey at firm level that provides information on firm's innovation behaviour, type of innovators, sectors and size classes. CIS surveys are carried out every two years by EU member states as well as several other non-EU countries (e.g. Norway, Iceland). Although most of European countries participate in each CIS survey, data are only available for a limited set of EU members' states. Hence, scholars must focus their work on a restricted sample of countries. Despite these limitations of data availability, this paper analyses the determinants of HGFs using an extensive sample of firms belonging to 15 countries: Bulgaria, Cyprus, Czech Republic, Estonia, Germany, Hungary, Italy, Latvia, Lithuania, Norway, Portugal, Romania, Slovakia, Slovenia and Spain.

The main advantage of the CIS data is that it contains detailed information on the innovation behaviour at the firm level in much greater detail than in other datasets. Thus, CIS data makes it possible to study the innovation behaviour of HGFs

and, in general, of SMEs. Additionally, the CIS data are internationally comparable based on a common survey questionnaire and methodology, which makes the corresponding data set suitable for cross-country comparison.

Table 1. COUNTRY DISTRIBUTION OF FIRM SAMPLES

	Country	Business R&D (% GDP)	Number of firms	Number of firms (%)	HGFs (%)	Sales growth 06-08 (%)
Country group 1 -leaders						
	Germany	1.84	4,028	5,99	10,80	15,15
	Slovenia	1.07	1,594	2,37	19,45	22,56
	Czech Republic	0.91	3,812	5,67	23,08	26,05
	Norway	0.87	2,449	3,64	19,19	22,41
	Portugal	0.76	4,578	6,80	13,30	12,11
	Spain	0.74	19,316	28,71	12,76	8,77
	Italy	0.60	8,219	12,22	8,80	5,77
Total group 1	7 countries		43,996	65,40%	13,36	11,90
Country group 2 -laggards						
	Estonia	0.56	2,710	4,03	16,09	12,81
	Hungary	0.53	3,207	4,77	20,11	20,72
	Slovakia	0.20	1,011	1,50	41,54	49,13
	Lithuania	0.19	825	1,23	22,42	22,00
	Romania	0.18	5,680	8,44	24,30	24,49
	Bulgaria	0.15	8,821	13,11	32,32	38,66
	Latvia	0.15	444	0,66	20,05	15,06
	Cyprus	0.11	585	0,87	22,74	28,92
Total group 2	8 countries		23,283	34,60%	26,36	28,89
	Total firms		67,279	100%		

Source: CIS 2008, Eurostat, own calculation.

HGF High-growth firms.

Business R&D: Expenditures on R&D performed by the business sector. Unit: % of GDP. Source: Eurostat.

We should also point out that CIS 2008 database has some drawbacks for the analysis of firm growth. First, CIS data is a cross-sectional dataset. In fact, analysing HGFs time-series data would allow us to investigate further questions, for instance,

which share of them continues growing fast, or which role the firm life cycles plays in the high-growth phenomenon. Second, CIS data has little financial information, which is a crucial variable for firm growth. Thus, we cannot use it to answer the question whether HGFs grow fast because they are already more profitable than the average firm, or whether they grow fast in order to achieve above average profitability. Third, some questions are «subjective». In this regard, the assessment of the innovation has a qualitative dimension and its proxies depend partially on the personal appreciation of the respondents. And fourth, although the CIS provides comparable innovation data for European countries, the country coverage varies substantially depending on the indicators considered.

Finally, our database was subject to a filtering process. First, we selected firms from the manufacturing and service sectors (including high-tech and low-tech sectors). Second, we restricted observations to those with a growth or decline of sales and employees smaller than 250% per year in order to control the presence of outliers. Although the filtering process reduced the initial database from 125,496 to 67,279 firms, the sample improved in the consistency of the data.

After the filtering process and according to the share of expenditures on R&D performed by the business sector on GDP during the period 2008, we group the 15 European countries into two categories. The first group (leader cluster) comprises a set of countries: six EU members (Germany, Slovenia, Czech Republic, Portugal, Spain and Italy) and a country that is member of the European Economic Area (Norway). All of them have been integrated on the EU project for many years, but Slovenia and Czech Republic that were integrated in 2004. These countries are close to the technological frontier where the share of business R&D on GDP is higher than 0.6 percent. In contrast, the second group comprises eight laggards EU members (Estonia, Hungary, Slovakia, Lithuania, Romania, Bulgaria, Latvia and Cyprus). All of them were integrated in 2004 in the EU, except two (Romania and Bulgaria) that were integrated in 2007. All of these countries were characterised by showing lower shares of business R&D on GDP and being far enough of the technological frontier of EU. The first group includes 43,996 firms and the second one 23,283 firms.

3.2. HGF indicators

Departing from this final selection of firms, we identify HGFs. We adopt the criteria adopted by OECD and Eurostat in the Manual on Business Demography Statistics (Eurostat-OECD 2007) which defines HGFs as: «*All enterprises with average annualised growth in employees (turnover) greater than 20% a year, over a 3-year period, and with 10 employees at the beginning of the observation period*».

Both of the most frequently used measures -sales and employment growth- have advantages and disadvantages. One drawback of the sales variable is inflation (Delmar *et al.*, 2003). Given that policy makers are concerned with reducing the

unemployment rate, employment is generally considered to be an interesting measure of firm growth (Storey, 1994). However, employment growth is highly affected by increases in labour productivity (Delmar *et al.*, 2003) and by the distance from the sectorial minimum efficient scale that enables them to survive (Sutton, 1997). On the other hand, it would be difficult to apply a consistent HGF threshold across all countries participating in the data collection. In fact, the countries with a larger share of SMEs and young firms will have more probabilities to have a larger percentage of HGFs than those countries with a larger share of large and mature firms.

Following this definition, we considered a HGF as a firm with a turnover growth equal or superior to the 20% between the years 2006 and 2008. In general, the empirical literature applies different HGF measures such as sales and employees. However, the CIS information provided by Eurostat does not offer the number of employees (this variable is recoded between three size classes: firms with 10–49, 50–249 and 250 and more employees) but CIS data offers the turnover. Consequently, our definition of HGFs will base on growth in terms of sales. Hence, we deflated this variable, as well as the rest of monetary variables, by an industrial price index.

Our final data has 67,279 firms, of which 12,020 (17.87 %) were HGFs. In the cluster of countries close to the technological frontier the percentage of HGFs are 13.36 %. In contrast, in the laggard group this percentage increase up to 26.36 %. The difference among the percentage of HGFs in both groups of firms may be explained in part by the fact that firms in laggard countries are taking benefit by the fact that these countries are still in a process of economic convergence and also with market structures which are weaker.

3.3. Statistical descriptive

Here we offer the main features that distinguished HGFs between the two country clusters that we have taken into account in this study. In particular, Table 2 presents descriptive statistics for the two country groups, leaders versus laggards' countries. The main characteristics that distinguish HGFs in both groups of countries are the following:

- a) The presence of HGFs is higher in countries classified as laggards than in leaders. In particular, the percentage of HGFs in laggards doubles to leaders.
- b) Despite the above result, the probability of being an innovative firm is higher among the most dynamic countries (leader group), in terms of innovation, than in the other countries (laggard group). In addition, firms that belong to the first group are more prone to invest in R&D, both internal and external, cooperate in R&D projects and perform acquisitions of machinery and other hardware more frequently.

Table 2. DESCRIPTIVE STATISTICS BY COUNTRY GROUPS
(mean values)

	Whole sample	Country group 1	Country group 2	Mean difference
HGF	0.1786	0.1336	0.2636	0.1299***(0.0030)
Innovative	0.5551	0.6175	0.4376	-0.1799***(0.0039)
Innovation input				
intRD	0.2602	0.3350	0.1187	-0.2163***(0.0034)
extRD	0.1277	0.1616	0.0638	-0.0977***(0.0026)
Machinery	0.2879	0.2949	0.2749	-0.0199***(0.0036)
Sources of information				
Internal	0.2259	0.2689	0.1449	-0.1239***(0.0033)
Market	0.2205	0.2503	0.1644	-0.0855***(0.0033)
Institutional	0.0402	0.0496	0.0224	-0.0272***(0.0015)
Other	0.0664	0.0745	0.0509	-0.0236***(0.0020)
Innovation output				
TechInnov	0.4488	0.5060	0.3410	-0.1649***(0.0039)
Non-TechInnov	0.4083	0.4632	0.3048	-0.1584***(0.0039)
Individual characteristics				
Size				
Size <50	0.5730	0.5862	0.5484	-0.0378***(0.0040)
Size 50-249	0.3210	0.3010	0.3591	0.0511***(0.0037)
Size >249	0.1056	0.1126	0.0923	-0.0202***(0.0024)
Group	0.2885	0.3359	0.1988	-0.1371***(0.0036)
Cooperation	0.1607	0.1825	0.1196	-0.0629***(0.0029)
Public funds				
Regional	0.0670	0.0997	0.0051	-0.0946***(0.0019)
National	0.0823	0.1056	0.0383	-0.0672***(0.0022)
EU	0.0342	0.0356	0.0316	-0.0040***(0.0014)
Export	0.5290	0.5628	0.4652	-0.0976***(0.0040)
Aggregate determinants				
Birth rate	21.02	17.29	28.07	10.78***(0.1251)
Observations	67,297	43,996	23,283	

Source: CIS 2008, Eurostat, own calculation.

HGF: high-growth firms.

Country group 1: Czech Republic, Germany, Italy, Norway, Portugal, Spain and Slovenia. Country group 2: Bulgaria, Cyprus, Estonia, Hungary, Latvia, Lithuania, Romania and Slovakia.

Note: Comparison of the two samples by the statistical t-test.*** Significant at 1%

Table 3. DESCRIPTIVE STATISTICS OF HFGS AND NON-HFGS BY COUNTRY GROUPS (mean values)

	Whole sample		Country group 1		Country group 2		Mean difference
	HGFs	Non-HGFs	HGFs	Non-HGFs	HGFs	Non-HGFs	
Innovative	0.5505	0.5562	0.6716	0.6091	0.4345	0.4386	-0.2370*** (0.0088)
<i>Innovation input</i>							
intRD	0.2396	0.2647	0.3776	0.3285	0.1073	0.1228	-0.2703*** (0.0073)
extRD	0.1275	0.1278	0.1911	0.1570	0.0666	0.0628	-0.1245*** (0.0059)
Machinery	0.2916	0.2871	0.3188	0.2912	0.2656	0.2782	-0.0531*** (0.0082)
<i>Sources of information</i>							
Internal	0.2259	0.2260	0.3021	0.2638	0.1529	0.1421	-0.1492*** (0.0075)
Market	0.2208	0.2205	0.2742	0.2466	0.1697	0.1626	-0.1045*** (0.0075)
Institutional	0.0430	0.0396	0.0622	0.0477	0.0245	0.0216	-0.0376*** (0.0036)
Other	0.0740	0.0647	0.0865	0.0727	0.0620	0.0470	-0.0244*** (0.0047)
<i>Innovation output</i>							
TechnInnov	0.4437	0.4500	0.5505	0.4991	0.3414	0.3409	-0.2091*** (0.0088)
Non-TechInnov	0.4199	0.4059	0.5352	0.4521	0.3094	0.3031	-0.2257*** (0.0087)

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Individual characteristics										
Size										
Size <50	0.5931	0.5688	0.5857	0.5863	0.6002	0.5298	0.0144(0.0089)			
Size 50-249	0.3136	0.3228	0.3104	0.2996	0.3166	0.3744	0.0061(0.0084)			
Size >249	0.0931	0.1083	0.1037	0.1140	0.0830	0.0957	-0.0206***(0.0053)			
Group	0.2797	0.2904	0.3696	0.3307	0.1935	0.2007	-0.1761***(0.080)			
Cooperation	0.1761	0.1574	0.2331	0.1747	0.1215	0.1189	-0.1116***(0.0068)			
Public funds										
Regional	0.0596	0.0686	0.1159	0.0972	0.0057	0.0049	-0.1102***(0.0042)			
National	0.0836	0.0821	0.1309	0.1017	0.0384	0.0383	-0.0924***(0.0049)			
EU	0.0391	0.0332	0.0460	0.0340	0.0324	0.0313	-0.0136***(0.0035)			
Export	0.4563	0.5449	0.5271	0.5684	0.3884	0.4927				
Aggregate determinants										
Birth rate	25.65	20.01	20.63	16.77	30.45	27.22	-0.1386***(0.0090)			
Observations	12,020	55,259	5,881	38,115	6,139	17,144				

Source: CIS 2008, Eurostat, own calculation.

HGF high-growth firms.

Country group 1: Czech Republic, Germany, Italy, Norway, Portugal, Spain and Slovenia. Country group 2: Bulgaria, Cyprus, Estonia, Hungary, Latvia, Lithuania, Romania and Slovakia.

Note: Comparison of the two samples by the statistical t-test. *** Significant at 1%.

- c) Firms that belong to country group 1 introduce more innovations, both technological and non-technological, and receive more public funds than their counterparts.
- d) Finally, laggard countries have a higher birth rate compared to the other countries, which might be explained due to increasing presence of HGF in these countries. The existence of more HGFs may be due to the fact that they are in a process of economic convergence but also to the characteristics of market structure where there is a larger predominance of small and young firms.

Table 3 shows the statistical descriptive of HGFs and Non-HGFs among both groups of countries. The main characteristics can be summarised in the following aspects:

- a) HGFs in laggard countries are less prone to invest in R&D and to cooperate in R&D projects than HGFs in the countries close to the technological frontier.
- b) HGFs in the laggard countries are smaller, dominated by firms with less than 50 employees.
- c) HGFs in laggard countries are less oriented to international markets, since they present a lower percentage of exports in comparison with leader countries.
- d) While the picture for the HGFs in leader countries seems to be that they are more innovative firms that carry out both technological and non-technological innovations, firms that often cooperate in R&D projects firms that are, oriented to international markets and with better access to public subsidies in R&D.
- e) In general, regardless we compare HGFs and Non-HGFs there are not significant differences (both groups show similar innovative patterns, similar innovative effort, and propensity to innovate). However, HGFs in laggard countries are smaller and they export less than Non-HGFs.
- f) In comparison with Non-HGFs, HGFs in leader countries are more innovative, they are more prone to invest more in R&D and cooperate in R&D projects, they use more frequently the different sources of information, and they have more likelihood to receive public funds. However, they tend to export less than Non-HGFs.

4. ECONOMETRIC METHODOLOGY

In order to analyse the relationship between the innovation activity and the probability of becoming a HGF, we apply a bivariate probit procedure. Our approach is based on the model labelled as CDM model (Crépon, *et al.*, 1998) where a firm's innovation effort has an impact on the capacity to innovate. And finally the innovation effort will have an impact on the firm performance measured in productivity.

Since we have a cross-section our estimations will not be able to capture the impact of innovation on the probability of being a HGF but the interrelationships. Hence, we adapt the CDM model and we consider a simultaneous model where the innovation inputs, innovation outputs and the capacity of the firm to become a HGF are interrelated.

By applying the bivariate probit regression model to estimate jointly the propensity to innovate and the capacity to become a HGF we will be able to control for unobserved common determinants¹².

$$Innovative_{it} = X'_{i,t} \beta_{11} + \gamma_{11} intRD_{i,t} + \gamma_{12} extRD_{i,t} + \gamma_{13} Machinery_{i,t} + Z'_{i,t} \beta_{12} + \varepsilon_{1i,t} \quad \text{Eq. [1]}$$

$$HGF_{it} = X'_{i,t} \beta_{21} + \gamma_{21} TechInnov_{i,t} + \gamma_{22} Non - TechInnov_{i,t} + \varepsilon_{2i,t} \quad \text{Eq. [2]}$$

and where

$$\begin{pmatrix} \varepsilon_1 \\ \varepsilon_2 \end{pmatrix} \sim N \left\{ \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{bmatrix} 1 & \rho_{12} \\ \rho_{22} & 1 \end{bmatrix} \right\} \quad \text{Eq. (3)}$$

Equation (1) measures the probability that a firm innovates depending on a set of determinants common in the current literature¹³. *Innovative* is a dummy variable that takes the value 1 if the firm reports having introduced new or significantly improved technological or non-technological innovations between 2006 and 2008. As explanatory variables which are specific in this equation we include the innovation inputs such as whether the firm invests in internal R&D (*intRD*), external R&D activity (*extRD*) and in machinery (*Machinery*) and a set of explanatory variables (*Z*) which belong to the different sources of information for innovation activities such as whether the firm considers important the information from sources within the

¹² The probability of innovating and the probability of becoming a HGF must be estimated simultaneously, since there may be unobserved characteristics that explain the capacity of firms to innovate and their capacity to become a HGF (see for instance Segarra and Teruel, 2014; Coad *et al.*, 2016). Also, Decker *et al.*, (2016) has recently pointed out the relationship between the presence of YICs and HGFs. Furthermore, the uncertainty associated with the innovation output depends also on unobservable firm-specific risk factors, which may affect also the capacity of a firm to introduce its goods in the market and, consequently, it may affect the capacity to become a HGF. In other words, there may be variables such as the access to financial resources, human resources with capacity to innovate and introduce their capacity in the market, among others. The bivariate probit model takes the correlations between the likelihood of innovating and the probability a firm becomes a HGF. The bivariate probit estimation, where we assume normality of the error terms, provides a correlation parameter that yields information about the co-variation of the error terms of the two estimations.

¹³ Please, check the variables definitions in Table A.1 and Table A.2 for the correlation matrix.

enterprise or group (*Internal sources*), from suppliers, clients, competitors or private R&D institutions (*Market sources*), from universities, public research organizations or technology centres (*Institutional sources*), and from conferences, scientific reviews or professional associations (*Other sources*).

Equation (2) estimates the probability of being a HGF (*HGF* is a dummy variable which indicates if the firm is a HGF measured in terms of sales or not) depending on whether the firm introduces technological (*TechInnov*) or non-technological innovations (*Non-TechInnov*) during the period 2006-2008.

Furthermore, both equations introduce different common sets of control variables (X). First, we introduce firm characteristics such as firm size (dummies that identify firms between 50 and 249 employees and those with 250 or more employees), if the firm belongs to a group, if the firm exports or cooperates, and if it has received public funds (at regional, national or EU level). Finally, we include some macroeconomic variables such as the birth entry rate, country dummies and sectoral dummies.

We assume that ε_i are independently and identically distributed and hence they follow a normal density function. The correlation coefficient between the disturbances (ρ) accounts for the possible existence of omitted or unobservable factors that affect simultaneously the decision to innovate and the likelihood of becoming a HGF. If ρ is equal to 0, the probability of becoming a HGF will not be correlated with the error term in Equation (1) and the probability of innovating will not be affected by the error term in Equation (2). While if ρ is different from 0, a joint estimation is required to obtain consistent estimates. Our results show that the coefficient ρ is significantly different to 0 when we estimate simultaneously all the countries. However, when we split up according with our two categories of countries the parameter is not significant. We present the joint results but our results remain similar when we estimate the equations independently.

5. RESULTS

This section reports the estimated parameters of the determinants of the probability to innovate and to be a HGF in terms of sales. Table 4 offers the estimations for the whole database and for both country groups considered in this paper. Our empirical results suggest, as might be expected, that the likelihood of innovating depends closely on investments in internal R&D, purchases of machinery and participation in cooperative R&D projects. With respect to the relationship between innovation and the probability to become a HGF our results suggest that technological and non-technological innovations are not really decisive determinants. However, firms belonging to leader countries show a positive relationship between introducing non-technological innovations and the probability of becoming a HGF.

Given these relationships we can conclude that while success in innovation is closely related to previous investments in R&D, to be a HGF is not associated with the innovative activity at firm level while other variables will show an impact.

With respect to other factors affecting the probability of being an innovative firm, the main results are the following. First, firms between 50 and 249 are more prone to innovate. However, for the largest firms the relationship is negative in the leader countries, while the relationship is positive among the laggard countries. Second, belonging to a group shows a positive relationship with the innovative capacity of the firm, especially in the group of laggard countries. Third, firms that cooperate and export show a greater correlation of being an innovative firm in both groups of countries. Fourth, regional public funds seem to be important to introduce technological and non-technological innovations.

Furthermore, public funds from the EU seem to show a positive relationship with the probability to be an innovative firm in the group of laggard countries. Finally, internal, market and other sources of information are the most important sources of information for innovation activities across all technology country groups. All in all, our results seem to show that firms' innovation activity in both groups of countries is ruled by different models innovation stages.

Regarding the other factors that affect the probability of becoming a HGF the differences between both country groups are relevant. First, our results confirm previous empirical evidence on the negative relationship between firm size and the probability of being a HGF. Hence, small firms have a larger propensity to become a HGF. Second, firms that belong to a group or they cooperate show also a positive association with the likelihood of becoming a HGF. However, both variables are not significant for the group of leaders while only the variable *Cooperation* is significant for firms located in the laggard countries. Third, the export activity has a significant negative relationship with being a HGF. Our result is not able to confirm the previous evidence which shows that firms with international activity will have a larger likelihood of being a HGF. We must have in mind that our data does not have temporal lags so we are not capturing the causal relationship.

Concerning the access to public funds, in general there is a positive relationship between firms that have received public funds at regional and national level and the probability of being a HGF. However, these relationships are only significant among firms in leader countries. Finally, we observe that the probability of being a HGF is positively associated with the existence of high business dynamics. Our result may point out that those countries with a larger firm entry rate may have a larger percentage of firms which become HGF. This may be linked with the fact that they are countries with higher market opportunities or that there are more competitive pressure.

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Table 4. BIVARIATE PROBIT OF THE PROBABILITY OF INNOVATING AND THE PROBABILITY OF BECOMING A HGF

	Whole database	Country group 1	Country group 2
Probability of becoming a HGF			
Innovation output			
TechInnov	0.0247 (0.015)	0.0296 (0.020)	0.0154 (0.024)
Non-TechInnov	0.0987*** (0.017)	0.1274*** (0.021)	0.0356 (0.029)
Individual characteristics			
Size			
Size: 50-249	-0.0534*** (0.014)	-0.0313 (0.019)	-0.0788*** (0.022)
Size>249	-0.1524*** (0.024)	-0.1545*** (0.030)	-0.1420*** (0.038)
Group	0.0284* (0.016)	0.0271 (0.019)	0.0304 (0.027)
Cooperation	0.0405** (0.020)	0.0333 (0.024)	0.0615* (0.035)
Public funds			
Regional	0.1058*** (0.027)	0.0949*** (0.028)	0.0423 (0.130)
National	0.0570** (0.025)	0.0658** (0.028)	0.0047 (0.054)
EU	-0.0333 (0.034)	-0.0385 (0.043)	-0.0026 (0.057)
Exports	-0.0649*** (0.014)	-0.0693*** (0.018)	-0.0530** (0.022)
Aggregate determinants			
Birth rate	0.0024*** (0.001)	0.0028** (0.001)	0.0029*** (0.001)
Constant	-1.4161*** (0.035)	-1.5574*** (0.043)	-0.3975*** (0.031)
Probability of innovate			
Innovation input			
intRD	0.5856*** (0.025)	0.6124*** (0.026)	0.7200*** (0.093)
extRD	-0.0661* (0.037)	-0.0824** (0.038)	0.4417*** (0.152)
Machinery	1.4730*** (0.022)	1.2294*** (0.026)	1.9164*** (0.040)

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	Whole database	Country group 1	Country group 2
Individual characteristics			
Size			
Size: 50-249	0.1690*** (0.015)	0.1413*** (0.018)	0.2154*** (0.026)
Size>249	0.0466 (0.028)	-0.0978*** (0.034)	0.3695*** (0.049)
Group	0.0416** (0.017)	0.0104 (0.020)	0.1757*** (0.035)
Cooperation	0.6797*** (0.038)	0.6427*** (0.041)	0.6536*** (0.093)
Exports	0.2185*** (0.014)	0.2596*** (0.016)	0.1076*** (0.026)
Public funds			
Regional	0.2510*** (0.045)	0.3053*** (0.044)	5.6923*** (0.176)
National	-0.0445 (0.046)	-0.0321 (0.048)	-0.0082 (0.151)
EU	0.1541* (0.080)	-0.0726 (0.086)	0.5645*** (0.200)
Sources of information			
Internal	0.9539*** (0.029)	0.9731*** (0.030)	0.7016*** (0.080)
Market	0.9701*** (0.030)	0.9047*** (0.032)	1.1532*** (0.082)
Institutional	-0.3454*** (0.077)	-0.4136*** (0.082)	-0.1607 (0.212)
Others	0.4415*** (0.066)	0.38433*** (0.072)	0.7485*** (0.161)
Aggregate determinants			
Birth rate	-0.0005 (0.007)	-0.0012 (0.001)	0.0012 (0.001)
Constant	-0.8893*** (0.042)	-0.8911*** (0.042)	-0.6938*** (0.034)
ρ	-0.0230* (0.012)	-0.0136 (0.016)	-0.0235 (0.020)
Wald test of χ^2	14648.84 0.000	8502.38 0.000	5489.57 0.000
Observations	67,279	43,996	23,283

Source: CIS 2008, Eurostat, own calculation.

HGF: high-growth firms.

Country group 1: Czech Republic, Germany, Italy, Norway, Portugal, Spain and Slovenia. Country group 2: Bulgaria, Cyprus, Estonia, Hungary, Latvia, Lithuania, Romania and Slovakia.

Estimations control for country and sector dummies. *, **, *** indicate levels of significance equal to 10, 5 and 1 %. Robust standard errors in parentheses.

6. CONCLUSIONS

Since the current economic recession, the European governments and the European Commission have shown an increasing interest in creating a favourable environment for innovation and generation of firms with high growing potential. Hence, this paper examines in depth the role that innovation plays to firms become HGFs. The CIS 2008 data provided by Eurostat covers most of the EU countries. Our country selection is based on an extensive sample of 67,279 innovative firms from 15 European countries distributed in two samples. The first sample includes 43,996 innovative firms from leader countries, and the second one includes 23,283 innovative firms from laggard countries. The aim of this study is to find the main stylized facts that define the relationship between R&D and innovation activities of European firms and their capacity to become HGFs.

The descriptive analysis confirms that HGFs are present in all the countries, although the amount of firms and their features considerably differ. The relative macroeconomic position of the country just like the technology nature and the market structure of the industry greatly define the capacity of the firms to become HGFs. In our analysis we show that leader countries have fewer HGFs but their driving forces are related to the R&D activities -firms are especially high growth innovative firms-, while the laggard countries have more HGFs driven by important business opportunities. These opportunities are provided by the catching-up process that these countries are experiencing towards the leading European countries. Most of these countries were recently satellite economies under the Soviet Union hegemony, which were finally joined to the EU during 2004 and 2007 with a considerable gap regarding previous EU country members.

In summary, in laggard countries HGFs are less oriented to international markets with smaller and underinvestment in R&D projects and little access to public subsidies than their counterparts in leader countries. Also in laggard countries, when we compare HGFs and Non-HGFs the differences between both groups are little in terms of innovative patterns, however in terms of size and export propensity HGFs are smaller and less export oriented than Non-HGFs. On the other hand, in leader countries we observe that HGFs are more innovating than Non-HGFs. They are more prone to invest in R&D and cooperate with R&D projects, they use more frequently the different sources of information and are more likely to receive public funds. Nevertheless, HGFs tend to export less than Non-HGFs.

Related to the determinants of the likelihood of innovating and becoming a HGFs, we should say that the capacity to innovate depends closely on investments in internal R&D, the purchase of machinery and the participation in cooperative R&D projects of the innovative firms. However, related to the link between innovation and the probability to become a HGF, our results suggest that technological and non-technological innovations are not really decisive determinants.

Other factors that affect the probability to innovate are the following: middle size firms -between 50 and 249 employees- are more prone to innovate; belonging to a group shows a positive impact on the innovative capacity of the company, especially in the laggard countries group; firms which export show a greater capacity to innovate; and regional public funds seem to be important to introduce technological and non-technological innovations. In conclusion, our results seem to show that firms' innovation activity in leaders and laggards countries is ruled by different model innovation stages.

Regarding the factors that affect the probability of becoming a HGF, the differences between both country groups are relevant. The results confirm the previous empirical evidence on the negative relationship between firm size and the probability of being a HGF. The most relevant stylized facts obtained are the following: small firms have a higher propensity to become a HGF; firms that belong to a group take advantage of becoming a HGF; and the export activity has a significant negative relationship about being a HGF. This final result is not able to confirm the previous evidence which shows that firms with international activity will have more likelihood of being a HGF, it is probably because our data does not have temporal lags so we are not capturing the causal relationship.

This paper highlights the evidence that the amount and, above all, the profile of HGFs are different between countries. The macro, sectorial and technological conditions are determinant keys about HGFs' features. In our analysis, we have found that in seven leader European countries the rate of HGFs is lower and they are more innovative HGFs, however in eight laggard European countries the rate of HGFs is higher and they invest less in innovation activities.

During these years, the European Commission together with states and regional governments share the goal to foster the creation and the survival of HGFs. This requires create the appropriate atmosphere to facilitate the emergence of new business projects that generate qualified high risk job opportunities and ensure the economic growth of the European countries. To achieve these objectives European Commission and the European governments improve the conditions in order to support the creation of HGFs. Up to now, frameworks such as «*Promoting innovative and high growth firms*» have taken an excessive simple and homogeneous vision. Studies like this one can help to understand that the European reality is much more complex. Therefore, in front of a complex reality we need an appropriate public policy in order to achieve the suggested aims.

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APPENDIX

Table A.1. VARIABLE DEFINITIONS

Dependent variables	
HGF	Dummy variable which takes the value 1 if the firm becomes a HGF measured in sales; 0 if not
Innovative	Dummy variable which takes the value 1 if the firm has introduced technological innovations or non-technological innovations; 0 if not
Independent variables	
<i>Innovation sources</i>	
intRD	Dummy variable which takes the value 1 if the firm invests in internal R&D; 0 if not
extRD	Dummy variable which takes the value 1 if the firm invests in external R&D; 0 if not
Machinery	Dummy variable which takes the value 1 if the firm acquires machinery; 0 if not
<i>Sources of information</i>	
Internal sources	Dummy variable which takes the value 1 if information from sources within the enterprise or group has high importance; 0 if not
Market sources	Dummy variable which takes the value 1 if information from suppliers, clients, competitors or private R&D institutions has high importance; 0 if not
Institutional sources	Dummy variable which takes the value 1 if information from universities, public research organizations or technology centres has high importance; 0 if not
Other sources	Dummy variable which takes the value 1 if information from conferences, scientific reviews or professional associations has high importance; 0 if not
<i>Innovation output</i>	
TechInnov	Dummy variable which takes the value 1 if the firm has introduced product or process innovations; 0 if not
Non-TechInnov	Dummy variable which takes the value 1 if the firm has introduced marketing or organisational innovations; 0 if not

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<i>Individual characteristics</i>	
Size	Set of size dummy variables according to the firm's number of employees. Categories are: <49 employees, 50–249 employees and 250 or more employees. Note: In CIS 2008 questionnaire Estonian and Latvian firms are only classified between two groups of size less than 50 employees and 50 or more employees.
Group	Dummy variable that takes a value equal to 1 if the firm belongs to a group; 0 if not
Cooperation	Dummy variable that takes a value equal to 1 if the firm cooperates with other agents; 0 if not
Regional public funds	Dummy variable that takes a value equal to 1 if the firm receives public financial support for innovation activities from local or regional authorities; 0 if not
National public funds	Dummy variable that takes a value equal to 1 if the firm receives public financial support for innovation activities from central government; 0 if not
EU public funds	Dummy variable that takes a value equal to 1 if the firm receives public financial support for innovation activities from the EU; 0 if not
Export	Dummy variable that takes a value equal to 1 if the firm sells goods or services in other European Countries or all other countries; 0 if not
<i>Aggregate determinants</i>	
Birth rate	Number of firm births in the reference period (t) divided by the number of firms active in t (%)
Industry	Set of industry dummies according to the firm's main CIS business activities (NACE 2-digit level)
Country	Set of country dummies belonging to country group 1 (Czech Republic, Germany, Italy, Norway, Portugal, Spain and Slovenia) and country group 2 (Bulgaria, Cyprus, Estonia, Hungary, Latvia, Lithuania, Romania and Slovakia).

Table A.2. CORRELATION MATRIX

	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)	(11)	(12)	(13)	(14)	(15)	(16)	(17)	(18)	
1.HGF	1.000																		
2.Innovative	-0.004	1.000																	
3.intRD	-0.021*	0.455*	1.000																
4.extRD	-0.000	0.293*	0.485*	1.000															
5.Machinery	0.003	0.517*	0.353*	0.284*	1.000														
6.Internal sources	-0.001	0.451*	0.478*	0.305*	0.366*	1.000													
7.Market sources	0.003	0.446*	0.390*	0.285*	0.414*	0.455*	1.000												
8.Institutional sources	0.006	0.160*	0.241*	0.252*	0.093*	0.202*	0.224*	1.000											
9.Other sources	0.014*	0.220*	0.251*	0.213*	0.177*	0.247*	0.302*	0.515*	1.000										
10.TechInnov	-0.004	0.807*	0.538*	0.351*	0.619*	0.535*	0.528*	0.185*	0.258*	1.000									
11.Non-TechInnov	0.011*	0.743*	0.381*	0.263*	0.360*	0.347*	0.343*	0.130*	0.187*	0.485*	1.000								
																			...

