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A Counting Multidimensional Innovation Index for SMEs

We developed a Counting Multidimensional Innovation Index (MII) framework for measuring and benchmarking innovation of Small and Medium Enterprises (SMEs), groups of SMEs, industries, regions, and countries. The methodology behind the MII is similar to the methodology behind the United Nations Multidimensional Poverty Index and follows the innovation definitions stipulated by the OECD Oslo Manual, covering dimensions and partial indicators suggested by this Manual and/or adapted from the Innovation Union Scoreboard (IUS) and from the Global Innovation Index (GII). To illustrate the MII framework, a survey was conducted among SMEs of the metalworking industry in Portugal.

Keywords: Innovation, SME, Multidimensional Innovation Index, Portuguese Metalworking Industry

1. Introduction

This research was motivated by the importance attached to innovation in the literature, by the role of Small and Medium Enterprises (SMEs)¹ as major sources of innovation, by the need of characterizing the innovation of sectors dominated by SMEs such as the metalworking industry in Portugal, and by the difficulties of gathering information on SMEs. The main objective of this research is the development of a Multidimensional Innovation Index (MII) innovation measurement framework, capable of overcoming the difficulties in measuring and benchmarking innovation of SMEs, groups of SMEs, industries, regions, and countries. The MII framework is able to generate not only individual profiles of innovation but also group composite measures of innovation, by partial indicator and/or dimension and multidimensionally, allowing to compare SMEs, groups of SMEs, industries, regions, and countries.

Composite indicators have been increasingly used to support data based narratives for political advocacy. Saltelli (2007) questioned the simplified messages they provide. Nardo *et al.* (OECD 2008) recommended best practices for the construction of composite indicators. In this research we followed the innovation definitions stipulated by the Organisation for

¹ In the European Union, SME's are firms employing 10 to less than 250 employees, which have an annual turnover or an annual balance sheet not exceeding, respectively, € 50 million and € 43 million (European Commission 2015).

Economic Cooperation and Development Oslo Manual (OECD 2005) and built 20 partial indicators of innovation, mostly adapting suggestions of this Manual and/or partial indicators appearing in the Innovation Union Scoreboard (IUS - UNU-MERIT, 2015) and in the Global Innovation Index (GII - Cornell University, INSEAD, and WIPO, 2015).² Hence, the partial indicators being considered in the MII are not new in the innovation literature. The novelty of the MII is the counting dual cut-off method employed to establish innovative and non-innovative firms, by partial indicator and multidimensionally. This counting dual cut-off method was proposed in the poverty literature by Alkire and Foster (2011), being motivated by Atkinson's (2003) discussion of counting methods for measuring multidimensional deprivation, and incorporating Sen's (1993) view of poverty as capability deprivation. It has been applied to compute the United Nations (UN) Multidimensional Poverty Index (MPI) developed by Alkire and Santos (2010, 2014), and more recently to measure multidimensional poverty in Europe (Alkire and Apablaza, 2016). So, the MII framework is based on the method proposed by Alkire and Foster (2011) and has similarities with the UN MPI framework (Alkire and Santos 2010, 2014). Like the MPI, the MII is based on micro data, employs a counting dual cut-off method that demands individual simple yes or no answers to a set of questions, and it is easy to compute.

The paper unfolds as follows. Section 2 addresses the concept of innovation and the theories behind the OECD Oslo Manual, the IUS, the GII, and the partial indicators of the proposed MII. Section 3 describes the counting dual cut-off method employed to compute the MII, providing an illustrative example. Section 4 yields an application of the MII framework to the SMEs of the metalworking industry in Portugal, presenting and discussing the survey conducted and the results obtained. Section 5 yields the concluding remarks.

² The Innovation Union Scoreboard (IUS) is published annually by the European Commission. The Global Innovation Index (GII) is co-published annually by Cornell University, INSEAD, and the World Intellectual Property Organization (WIPO), an agency of the United Nations (UN). In addition to the Oslo Manual, the IUS, and the GII, a few Portuguese national systems were also taken into account to build the 20 partial indicators (see, for instance, COTEC 2014 and IAPMEI 2014).

2. Innovation and SMEs

Several theories form the basis for the OECD Oslo Manual, the IUS, the GII, and thus the proposed MII framework. These theories establish the concept of innovation and its relationship with productivity, identify the territories where innovation occurs, provide a view of innovation as a system, stress the importance not only of products and processes but also of organisational, management and marketing practices, discuss the driving forces behind innovation, characterize the existence of radical and incremental innovations and of open and closed innovations, describe the links between innovation and firm size and between innovation and firm growth, discuss the explanatory power of patents, and many other aspects. They justify not only the choice of the 20 partial indicators used in the MII framework but also the environmental variables that can be considered ex-post explaining differences in innovation across firms.

According to Schumpeter (1911), innovation is a process of creating something new and destroying what becomes outdated. It can be understood as the creation of new combinations of means of production, which may include the introduction of a new good, the introduction of new production methods, the opening of a new market, the conquest of a new source of supply of raw materials, or the implementation of a new organization in any industry (Schumpeter 1911). For Joyce, Nohria, and Roberson (2004), to be considered innovative an organization must change its industry somehow. Godin (2008) considers innovation as a process that leads to an outcome that is a novelty, by definition, since the object or the new way of doing something did not previously existed. Chauvel (2011) sees innovation as the ability of a firm to overcome perfect competition, originating a situation of temporary monopoly by creating a new market for its products. For the purpose of this research, we have adopted the OECD Oslo Manual definition of innovation. “An innovation

is the implementation of a new or significantly improved product (good or service), or process, a new marketing method, or a new organisational method in business practices, workplace organisation or external relations” (OECD 2005, pp 46).

Innovations and creativity (or inventions) are related but not the same thing. Inventions necessarily happen before innovations, which explains the attention given in the literature to creativity (Florida 2002; Landry 2008; Shearmur 2012). While creativity refers to the generation of ideas, innovation concerns their implementation.³ Innovation is a process that transforms inventions into value at the firm level (Lazonick 2005). Creativity is an important element in innovation, as it aims to develop organizations with the ability to process innovation, so that the result is new, unique and meaningful (Badawy 1986). National culture does influence economic creativity, and innovation implementation explains some of the variation in prosperity across countries (Williams and McGuire 2010). Networks contribute to innovation and performance of SMEs (Gronum, Verreynne, and Kastle 2012). Human capital can have a significant impact on reducing the barriers to innovation represented by knowledge shortages and market uncertainties (D’Este, Rentocchini, and Vega-Jurado 2014). Small firms provide the most conducive environment for entrepreneurship and innovation, given the commitment and close cooperation of their members (Sahut and Peris-Ortiz 2014).

Robert Solow (1956, 1957) noted more than 60 years ago that rising incomes should largely be attributed not to capital accumulation, but to technological progress – to learning how to do things better (Stiglitz 2014). Technological progress is embodied in Neoclassical and New Growth Theory models (for example Romer 1990) and is pointed as the principal way through which economic growth can be stimulated (Ray 1998). There are other perspectives on innovation, besides the ones provided by Neoclassical and New Growth

³ “A creative idea that doesn’t generate value isn’t technology. It’s art” Matthew Ganz (Boeing Company).

theories. According to Evolutionary and New Schumpeterian approaches, innovation and technological changes are path dependent (Nelson and Winter 1982; Dosi 1982; Freeman and Louçã 2001; Verspagen 2005).

Concerning the territories where innovation occurs, many authors show the existence of links between innovation and cities. Currid (2007) presented the case of New York. Bettencourt, Lobo, and Strumsky (2007), referring to the number of innovation patents, concluded that the latter occur predominantly in cities. For a few authors cities are the *loci* of innovation and creativity (Montgomery 2007; Florida 2009). Cohendet, Grandadam, and Simon (2010) studied the anatomy of the creative city by defining three different layers - the upper ground, the middle ground and the underground. Each one of these layers intervenes with specific characteristics in the creative process, and enables new knowledge to transit from an informal micro level to a formal macro-level. Shearmur (2012) investigated the correlation between innovation and cities. The author argues that cities convert innovation into value, though innovation may occur outside the cities. That is, cities may be dependent on activities occurring outside. He concluded that the only types of innovation specific to the cities might well be social and political innovations, designed to address issues specifically earmarked for housing developments. Lööf and Johansson (2014) studied the influence of metropolitan externalities on productivity for different types of long-run R&D engagement. They found that firms in Sweden with persistent R&D have a productivity premium that is about 14 per cent in the largest cities and 8 per cent in non-metro locations. Lee and Rodríguez-Pose (2014) studied the link between local creative industries concentration and SMEs' innovation in the UK. The results suggest that firms in local economies with high shares of creative industries' employment are significantly more likely to introduce entirely new products and processes than firms elsewhere.

Regarding the relationship between innovation and productivity, Hall (2011) found an economically significant impact of product innovation and a somehow more ambiguous impact of process innovation, the latter result being primarily due to difficulties in measuring the effect. Bloom *et al.* (2012) show that high management scores are strongly and positively related with countries' level of development. Bloom *et al.* (2014) provide evidence that an important explanation for the substantial differences in productivity among firms and countries are variations in management practices. Their preliminary estimates suggest that around a quarter to a third of cross-country and within-country TFP gaps appear to be management related. Still according to Bloom *et al.* (2014), higher management scores are positively and significantly associated with higher productivity, firm size, profitability, sales growth, market value and survival. Factors such as competition, governance, ownership, human capital, asymmetric information, financial constraints, etc., help to account for the variation in management (Bloom *et al.* 2014). Several authors analyse organizational structures and the processes of learning and adjustment to changes in technology and in the firm environment, including the market (e.g. Lam 2005). Thus, management seems to matter, both qualitatively and quantitatively. Many authors found that the effect of R&D on firm's productivity is positive (Lichtenberg and Siegel 1991; Klette, and Johansen 1998; Harhoff 1998; Lotti, and Santarelli 2001; Janz, Löf, and Peters 2004; Van Leeuwen and Klomp 2006; Parisi, Schiantarelli, and Sembenelli 2006) and a few suggested that the returns from R&D have been declining over the years (Klette and Kortum 2004).

Innovations can be radical or incremental. For Schumpeter (1911), radical innovations cause world changes, while incremental innovations fill in the process of continuously change. For Stiglitz (2014), 'While some of the productivity increase reflects the impact of dramatic discoveries, much of it has been due to small, incremental changes'. A few authors relate radical innovations with networking SMEs and incremental innovations with large and

hierarchical organizations operating in markets with few innovations (Freeman and Soete 1997). Other relate these innovation types with network relationships of firms, with radical innovations requiring strong collaborative ties with costumers while incremental innovations being commercialized through different types of downstream networks (Partanen, Chetty, and Rajala 2014). Maes and Sels (2014) investigated if and how internally and externally oriented knowledge capabilities can stimulate radical product innovation in SMEs. They concluded that externally oriented learning processes are dependent on firms' capabilities that increase knowledge diversity and sharing among the employees, which in turn has a strong and direct influence as potential wellsprings of radical innovation.

Regarding the sources of information, the literature distinguishes closed and open innovations. According to Chesbrough (2013) closed innovations happen when an organization uses internal sources of information to innovate, such as its own employees, while open innovations occur when an organization uses external sources of information to innovate. The author suggests that many innovative firms have moved to a model of open innovation, using a wide range of actors and external sources to help them achieve and sustain innovation. Laursen and Salter (2006) explored the relationship between the firm's external openness and its innovation performance. The authors found the most open firms to be more likely to get better innovation performance and the benefits of openness being subject to decreasing returns. Barge-Gil (2010) addressed the relationship between the openness of firms' innovation strategies and firm characteristics by distinguishing three firm strategies - open, semi-open and closed. Using a panel of Spanish firms (2004–2006), he concluded that open innovators are smaller and less R&D intensive than semi-open ones, although larger and more R&D intensive than closed innovators. Parida, Westerberg, and Frishammar (2012) analysed inbound open innovation activities in high-tech SMEs. They found technology sourcing being linked to radical innovation performance,

whereas technology scouting being linked to incremental innovation performance. Lasagni (2012) investigated the role of external relationships as key drivers of European SMEs innovation. The author found innovation performance being higher in SMEs that are proactive in strengthening their relationships with innovative suppliers, users, and customers. Brockman, Jones, and Becherer (2012) showed that the overall positive influence of customer orientation on SMEs performance is stronger as risk-taking, innovativeness, and opportunity focus increase. Spithoven, Vanhaverbeke, and Roijakkers (2013) studied open innovation (OI) practices in SMEs. They found that the effects of OI practices in SMEs often differ from those in large firms. SMEs are more effective in using different OI practices simultaneously when they introduce new products on the market, whereas this is less the case for large firms.

Against the Schumpeterian hypothesis of a positive relationship between firm size and innovative activity, a few authors (e.g. Acs and Audretsch 1988, 1990) argued that small companies are the engines of technological change and innovation activity. Although preliminary, Hall, Lotti, and Mairesse's (2009) results indicate that firm size is negatively associated with R&D intensity and positively associated with the likelihood of having process or product innovations. Revilla and Fernández (2013) studied the effects of firm size on innovative activity on a sample of Spanish manufacturing firms. They found that technological dynamism negatively moderates the effects of firm size on the economic productivity of R&D. Lejarraga and Martinez-Ros (2014) extended the size–innovation debate by proposing that the size of firms affects the scale and quality of product innovation, through the adoption of different decision making styles. Using longitudinal data of Spanish firms, they showed that size is negatively related with the scale and positively related with the quality of product innovation. Antonelli and Scellato (2015) analysed the effects of the size of the firm in the direction of technological innovations in a panel of 6600 Italian

firms (1996-2005). They found small firms being more likely to introduce biased technological changes, directed toward the most intensive use of locally abundant production factors, and large firms being more likely to introduce neutral technological changes, shifting the production frontier.

Other authors addressed the link between innovation and firm growth. According to Audretsch, Coad, and Segarra (2014), a large number of applied papers found a positive link between innovation and firm growth. For example, Deschryvere (2014) explored the relationships between growth and R&D investment. Using Finnish data, he found continuous product innovators and occasional process innovators to have the stronger associations between sales growth and subsequent R&D growth.

A few authors questioned the innovation explanatory power of patent citations (Breschi and Lissoni 2001; Thompson and Fox-Kean 2005). Indeed, as pointed out by Acs (2002), the use of patents as an indicator of innovation reveals some problems such as: i) the fact that patents only register major product innovations; ii) the existence of firms patenting new ideas and having no intention of launching them into the market; and iii) SMEs that would rather rely on the secrecy of the product, not disclosing their inventions.

Many other aspects are addressed in the literature, such as the role of entrepreneurial motivations (Verbees and Meulenbergh 2004; Carsrud and Brännback 2011), the context dependence of innovation performance (Rosenbusch, Brinckmann, and Bausch 2011), the impact and effectiveness of policy support to innovation (Parrilli and Elola 2012; Foreman-Peck 2013; Kobayashi 2014; Castillo *et al.* 2014; Brancati 2015), the joint dynamic of export and R&D activities (Esteve-Pérez and Rodríguez 2013), the impact of the business cycle (Madrid-Guijarro, García-Pérez-de-Lema, and Auken 2013), the family and non-family nature of the firm (for example Classen *et al.* 2014), the joint effect of patents

and reduced competition (Beneito, Rochina-Barrachina, and Sanchis 2014), the use of CEO, managers, and non-managerial employees ideas in small firms (Andries and Czarnitzki 2014), the Total factor Productivity (TFP) elasticity with respect to R&D capital (Cchini and Venturini 2014), the comparison of international innovation strategies of emerging and developed markets (Wang, Sutherland, and Ning 2014), marketing theories and normative approaches to market exchanges (Hunt, 1983), consumer markets (Burr 2014), and other aspects.

Table 1 describes the 20 partial indicators of innovation employed in the MII framework. Recognizing the existence of outputs and inputs of innovation (see OECD 2005; UNU-MERIT 2015; Cornell University, INSEAD, and WIPO 2015), the 20 partial indicators of innovation are classified as outputs (or results of) and inputs (or efforts towards) innovation. The 8 output partial indicators considered are subdivided in 2 classes: i) Knowledge and technology (3 indicators); and ii) Economic effects (5 indicators). The 12 input partial indicators considered are subdivided in 5 classes corresponding to business functions: i) Human resources (2 indicators); ii) Processes and infrastructure (4 indicators); iii) Strategy and organization (3 indicators); iv) Finance (2 indicators); and v) Marketing (1 indicator).

Table 1
MII partial indicators

Partial Indicators
Outputs, Knowledge and Technology
1. Product and/or processes innovations. Measures whether companies were able to introduce innovations in products and processes in their markets.

<p>2. Number of workers. Aims to assess whether the company has managed to increase its number of qualified employees</p> <p>3. PCT Patents. Aims to appreciate the company's ability to create patents, since the company's ability to develop new products will determine their competitive advantage, and is therefore a good indicator of the innovation rate in new products.</p>
Outputs, Economic effects
<p>4. Volume of exports. Company's ability to increase its volume of exports.</p> <p>5. Turnover. Company's ability to increase its turnover</p> <p>6. Patent revenues.</p> <p>7. New customers. Company's ability to get new customers.</p> <p>8. New markets. Company's ability to enter into new markets.</p>
Inputs, Human resources
<p>9. Workers by skill level. Aims to identify the degree of qualification of employees in the company, and the percentage of those with masters or doctoral qualification.</p> <p>10. Training investment. The purpose is to know whether companies invest in the training of employees.</p>
Inputs, processes and infrastructure
<p>11. Innovations made in products and/or processes. Intends to evaluate whether a company has achieved some innovation in their products and/or processes internally.</p> <p>12. Partnerships with external entities for innovation. Aims to measure the degree of involvement of companies with external entities for innovation.</p> <p>13. ICT use. It is intended to illustrate whether companies are using computer tools or specialized software during the process of manufacturing and services.</p> <p>14. Existence of R&D department in the company. Aims to find out if the company has a space exclusively dedicated to R & D within its facilities.</p>
Inputs, Strategy and organization

<p>15. Actions by top management in innovation activities. Indicates whether top management participates actively in the innovation process.</p> <p>16. Encouraging innovation by organizational culture. Aims to assess to what extent the organization's culture encourages entrepreneurship, and the risk-taking behaviour of their workers.</p> <p>17. Organizational innovations. Seeks to measure the ability of firms to innovate organizationally, according to the concept of organizational innovation defined.</p>
Inputs, Finance
<p>18. R&D investment. Evaluates the financial efforts in R&D made by the company.</p> <p>19. Credit access for R&D investment. Reveals the company's ability to obtain external financing to invest in R&D.</p>
Inputs, Marketing
<p>20. Marketing innovations. The purpose is to establish to what extent companies have achieved innovations in marketing in accordance with the definitions set out.</p>

3. A Counting Multidimensional Innovation Index (MII) for SMEs

The MII is a vector that contains the following 4 multidimensional innovation measures:

$$MII_o = H_o \times A_o \quad (1)$$

$$MII_i = H_i \times A_i \quad (2)$$

$$MII_r = MII_o / MII_i \quad (3)$$

$$MII_a = (MII_o + MII_i) / 2 \quad (4)$$

These 4 composite measures mimic the 4 composite indicators proposed in the GII (see Cornell University, INSEAD, and WIPO 2015):

- The MIIo measures the incidence and intensity of multidimensional innovation in terms of outputs. H_o is the proportion of multidimensional output innovative SMEs and A_o is the average intensity of multidimensional output innovation of these SMEs. The MIIo measure lies on the interval $[0,1]$.
- The MIIi measures the incidence and intensity of multidimensional innovation in terms of inputs. H_i is the proportion of multidimensional input innovative SMEs and A_i is the average intensity of multidimensional input innovation of these SMEs. The MIIi measure lies on the interval $[0,1]$.
- The MIIr is the ratio between the MIIo and the MIIi. It measures the Total Factor Productivity (TFP) of innovation of the group of SMEs under analysis. It is similar to the GII Innovation Efficiency Ratio (Cornell University, INSEAD, and WIPO 2015). The MIIr measure lies on the interval $[0, + \infty]$.
- Finally, the MIIa is the average between the MIIo and the MIIi. It is an overall measure of innovation similar to the IUS Summary Innovation Index (UNU-MERIT 2015) and to the GII Global Innovation Index (Cornell University, INSEAD, and WIPO 2015). The MIIa measure lies on the interval $[0,1]$.

Like the MPI (Alkire and Santos 2010, 2014; Alkire and Apablaza 2016), the mathematical structure of the MIIo and the MIIi measures corresponds to the one of the M_0 poverty measure, the latter being the Adjusted Head Count Ratio proposed by Alkire and Foster (2011). The M_0 measure has several important properties:

1. It is robust when using ordinal or cardinal variables, as it dichotomizes the variables;

2. By adjusting multidimensional incidence by intensity, it satisfies dimensional monotonicity;
3. It is decomposable by population sub-groups; and
4. It can be broken down by partial indicator and/or dimension.

See Alkire and Foster (2011) for a detailed presentation of the M_0 properties.

Constructing the MII measures implies the following steps:

1. Establishing the set of output and input partial indicators and dimensions to be considered in the composite measures. In section 2, we presented a list of 20 output and input partial indicators.
2. Setting the cut-offs for each partial indicator above which the SME is considered to be innovative in each indicator. These cut-offs are set to be 0.5, since all the variables are dichotomized (1, if the SME is innovative in the partial indicator; 0 if the SME is not innovative in the partial indicator) and we consider 0.5 as the multidimensional output (input) cut-off. Consequently, all partial indicators missing values are to be filled with the value 0.5.
3. Applying the cut-offs to verify in which partial indicators each SME is and is not innovative.
4. Selecting the relative weights of each indicator such that the sum of the weights of all output (input) indicators adds up to one. Like in the MPI, we have chosen equal weights for all output (input) dimensions, and equal weights for the indicators within each dimension.
5. Computing the weighted innovation output (input) composite score for each SME.
6. Determining the multidimensional output (input) innovation cut-off, that is, the proportion of weighted innovation achievements a SME needs to have to be

considered output (input) multidimensional innovative. We set the output (input) cut-off to be 0.5.

7. Computing the proportion of SMEs that have been identified as output (input) multidimensionally innovative in the sample. This is the headcount ratio H_o (H_i).
8. Computing the average score of multidimensional output (input) innovative SMEs. This is the average intensity of multidimensional output (input) innovation, A_o (A_i).
9. Computing MII_o and MII_i .
10. Computing MII_r and MII_a respectively as the ratio and the average of MII_o and MII_i .

The cut-offs in 2. and in 6. are normative. If the cut-offs in 6. are changed, the cut-offs in 2 must be changed accordingly, because of the treatment given to the missing values.

The weights of the partial indicators are also normative. They need not to be equal across output (input) dimensions and across indicators in each dimension. The only restriction is the sum of the weights of all output (input) indicators to add up to one. Nonetheless, equal weights make easier the interpretation of the composite indicators (on this see Atkinson 2003; Alkire and Foster 2011).

The list of 20 partial indicators considered can be altered. For instance, it is possible to have a set of alternative partial indicators more suited to address SMEs' innovation in developing countries.

To compute the MII, we only need the group of SMEs under analysis to answer twenty 'yes' or 'no' questions that feed the twenty partial indicators of innovation being considered. Table 2 shows the twenty partial indicators of innovation measured for each SME and their weights.

Table 2
MII partial indicators and corresponding Weights

	Dimension	Partial indicator (weight)
Outputs	Knowledge and Technology (50%)	Product and/or Processes Innovations (16.67%) Number of Workers (16.67%) PCT Patents (16.67%)
	Economic Effects (50%)	Volume of Exports (10%) Turnover (10%) Patent Revenues (10%) New Costumers (10%) New Markets (10%)
Inputs	Human Resources (20%)	Workers by Skill Level (10%) Training Investment (10%)
	Processes and Infrastructures (20%)	Innovations Made in Products or Processes (5%) Partnerships with External Entities for Innovation (5%) ICT Use (5%) Existence of R&D Department in the Company (5%)
	Strategy and Organization (20%)	Actions by Top Management in Innovation Activities (6.67%) Encouraging Innovation by organization Culture (6.67%) Organizational Innovations (6.67%)
	Accounting and Finance (20%)	R&D Investment (10%) Credit Access for R&D Investment (10%)
	Marketing (20%)	Marketing Innovations (20%)

Table 2 allows to build for each SME under analysis a profile of innovation with ‘1’s and ‘0’s, which is easy to interpret, and to compute individual multidimensional innovation output add input scores. Once each SME is classified in either multidimensional output (input) innovative or non-innovative, the next step is to compute the MII vector that characterizes the group of SMEs under analysis. We next show an illustrative example of the MII framework computation for a group of three hypothetical SMEs: X, Y, and Z. The first step is to fill each SME innovation profile and check if it is multidimensional output (input) innovative, by computing the corresponding multidimensional innovation scores. Table 3 shows the values of the 20 partial indicators for each of the three SMEs considered.

Table 3
Innovation Profiles of SMEs X, Y, and Z

	Dimension	Partial indicator (weight)	X	Y	Z
Outputs	Knowledge and Technology (50%)	Product and/or Processes Innovations (16.67%)	1	0	0
		Number of Workers (16.67%)	1	0	1
		PCT Patents (16.67%)	0	0	0
	Economic Effects (50%)	Volume of Exports (10%)	1	0	1
		Turnover (10%)	1	0	1
		Patent Revenues (10%)	0	0	0
		New Costumers (10%)	1	1	1
New Markets (10%)		1	0	0	
Inputs	Human Resources (20%)	Workers by Skill Level (10%)	0	0	0
		Training Investment (10%)	1	1	1
	Processes and Infrastructures (20%)	Innovations Made in Products or Processes (5%)	1	0	1
		Partnerships with External Entities for Innovation (5%)	1	0	1
		ICT Use (5%)	1	0	0
		Existence of R&D Department in the Company (5%)	1	0	0
	Strategy and Organization (20%)	Actions by Top Management in Innovation Activities (6.67%)	1	1	1
		Encouraging Innovation by organization Culture (6.67%)	1	1	1
		Organizational Innovations (6.67%)	0	0	1
	Accounting and Finance (20%)	R&D Investment (10%)	1	0	1
Credit Access for R&D Investment (10%)		1	0	0	
Marketing (20%)	Marketing Innovations (20%)	0	0	1	

From Table 3, the multidimensional innovation output and input scores of SME X are respectively:

- $1*0.1667+1*0.1667+0*0.1667+1*0.1+1*0.1+0*0.1+1*0.1+1*0.1 = 0.733 > 0.5$, for outputs, and
- $0*0.1 + 1*0.1 + 1*0.05 + 1*0.05 + 1*0.05 + 1*0.05 + 1*0.0667 + 1*0.0667 + 0*0.0667 + 1*0.1 + 1*0.1 = 0,633 > 0.5$, for inputs.⁴

We conclude that SME X is multidimensional output innovative ($0.733 > 0.5$) and multidimensional input innovative ($0,633 > 0.5$). Similar calculations can be performed with SMEs Y and Z. Table 4 shows the multidimensional output and input innovation scores of the three SMEs. Only SME X is multidimensional output innovative. SMEs X and Z are both

⁴ It is also possible to compute for each SME the TFP of innovation. The TFP score of SME X is: $0.733/0.633 = 1.158$.

multidimensional input innovative. SME Y is neither multidimensional output innovative nor multidimensional input innovative.

Table 4
SMEs X, Y and Z Multidimensional Output and Input Innovation Scores

	X	Y	Z
Multidimensional Output Innovation Score	0.733	0.100	0.467
Is the firm Output-Innovative?	Yes	No	No
Multidimensional Input Innovation Score	0.633	0.233	0.700
Is the firm Input-Innovative?	Yes	No	Yes

From Table 4, it is possible to compute the MII vector measures for the group constituted by the three SMEs:

$$MII_o = H_o \times A_o = 1/3 \times 0.733 = 0.244 \quad (5)$$

$$MII_i = H_i \times A_i = 2/3 \times [(0.633+0.700)/2] = 0.444 \quad (6)$$

$$MII_r = MII_o/MII_i = 0.244/0.444 = 0.550 \quad (7)$$

$$MII_a = (MII_o+MII_i)/2 = 0.244/0.444 = 0.344 \quad (8)$$

4. Applying the MII framework to the SMEs of the Metalworking Industry in Portugal

An online survey was conducted in late 2013 among a universe of 700 SMEs of the metalworking industry in Portugal. The questionnaire asked information referring to the 2012 civil year. A total of 45 SMEs responded, which represents about 6.4 percent of firms surveyed.

Sample

Figure 1 and Figure 2 display respectively the distribution of the universe and of the sample by NUTS III territories and by Class of Economic Activity (CAE). Table 5 indicates the average scale of operation – number of workers and turnover – of the universe and of the sample. Figures 1 and 2 and Table 5 show the sample to be fairly representative of the universe. Most of the firms are from ‘Grande Porto’ NUTS III and from CAE 25 – ‘metal products, except machinery and equipment’, in the universe and in the sample (see Appendix for mainland Portugal map of NUTS III territories and for the description of the Portuguese metalworking industry CAE). Average number of workers and turnover are similar, in the universe and in the sample.

Figure 1
Universe and Sample by NUTS III

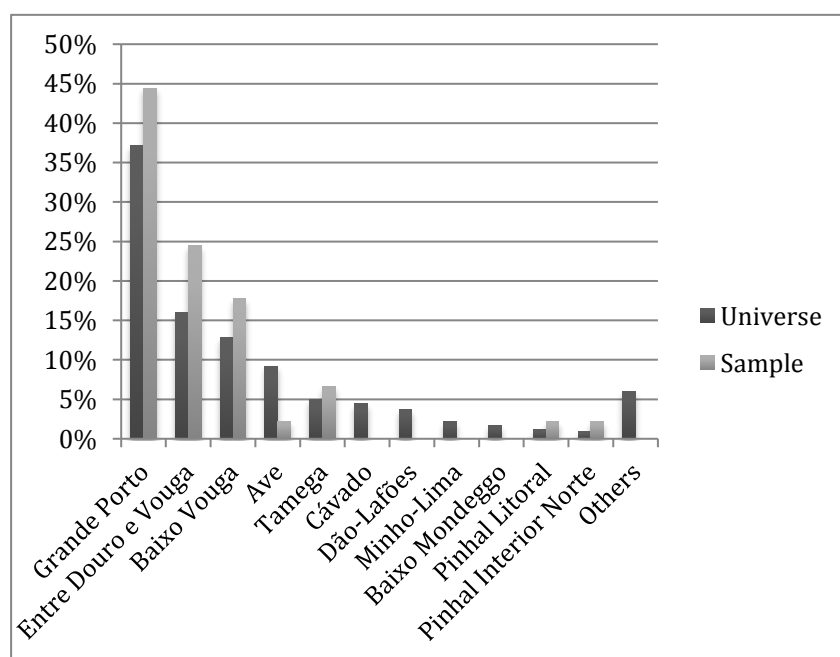


Figure 2
Universe and sample by Class of Economic Activity (CAE).

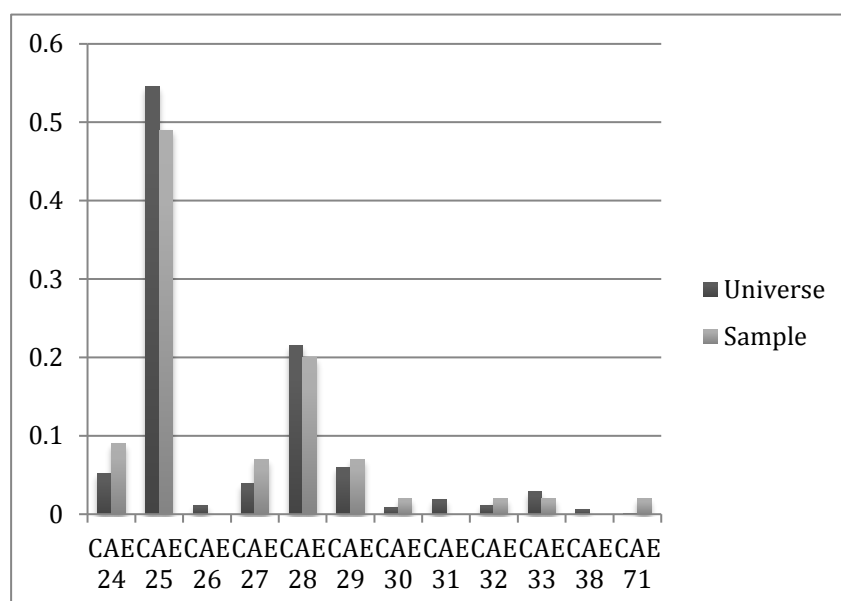


Table 5
Average Scale of Operation in 2012.

	Average number of workers	Average turnover per firm
Universe	55	€ 5352515
Sample	58	€ 5409632

Source: AIMMAP/Authors

Table 6 yields sample average characteristics of Portuguese metalworking industry SMEs in 2012. The average turnover of sample SMEs was slightly above € 5 million in 2012. ‘Exports’ represented 36.5 percent of the turnover, while ‘R&D and innovation expenditures’ only 1.4 percent. ‘Sales to new markets’ accounted for approximately 4.5 percent of ‘Turnover’, the weight of ‘Sales of innovative products’ being 6.5 percent. ‘New customers’ were about 9.1 percent of ‘Total customers’. Only two sample SMEs had ‘Patents’.

Table 6
Sample SMEs Average characteristics in 2012.

Variable	Average value in 2012
Number of workers	58
Turnover	€ 5409632.53
Exports	€ 1977192.55
R&D and innovation expenditures	€ 78094.61
Sales to new markets	€ 258642.90
Sales of innovative products	€ 404590.82
New customers	11.2
Total customers	123.3
Patents	0.7

Figures 3, 4, and 5 illustrate the distribution of workers by gender and by level of education in the sample. More than 75 percent of workers are males. More than 80 percent have a level of education III – ‘Upper Secondary education geared for further study at a higher level or less’. The weight of females in the workforce is superior in intermediate levels of education when compared to other levels, representing half of workers with level IV - ‘Upper Secondary Education obtained in basic education courses or dual certification or geared for further studies at a higher level plus internship - minimum of 6 months.’

Figure 3
Workers by Gender

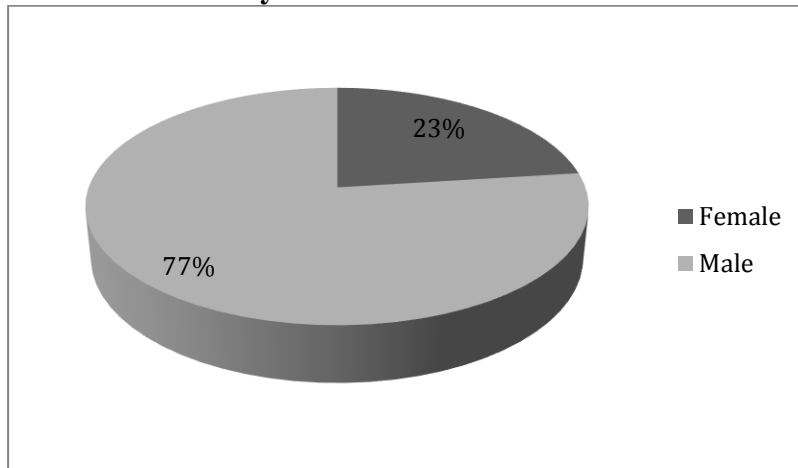
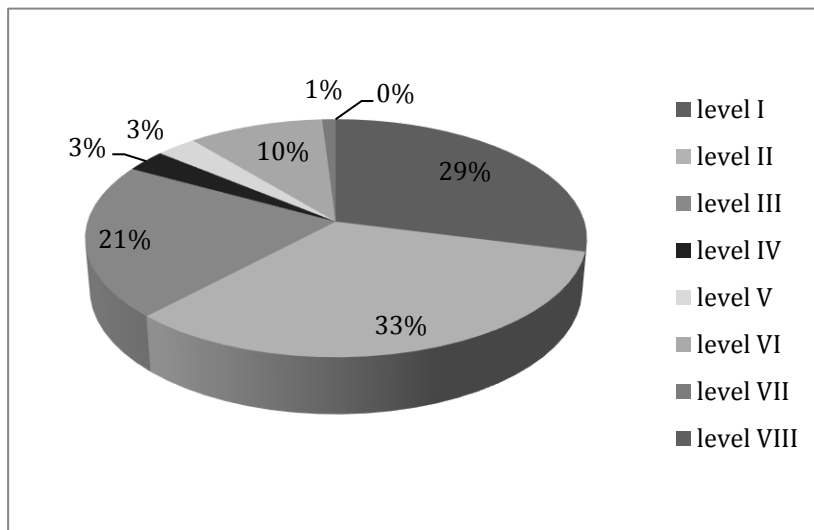


Figure 4
Workers by Level of Education



Level I – Lower Secondary Education

Level II – Upper Secondary Education obtained in basic education courses or dual certification

Level III – Upper Secondary education geared for further study at a higher level

Level IV – Upper Secondary Education obtained in basic education courses or dual certification or geared for further studies at a higher level plus internship - minimum of 6 months

Level V – Qualification not greater than post-secondary level with credits for further study at higher level

Level VI – Bachelor

Level VII – Master

Level VIII – Doctoral

Figure 5

Workers by Gender and Level of Education.

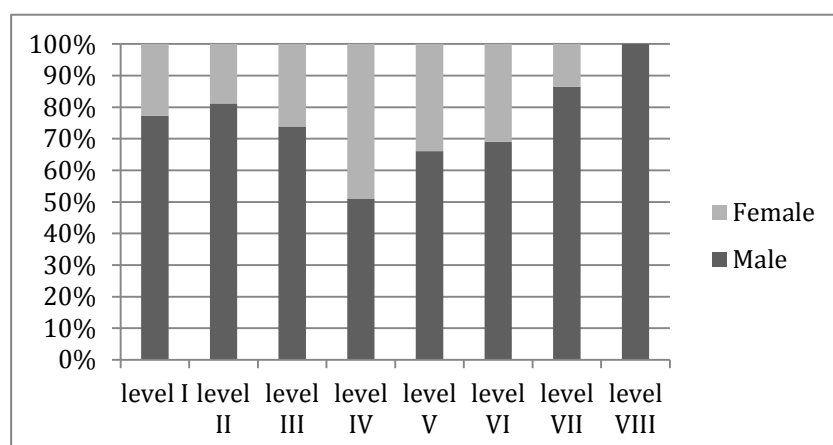


Figure 6 shows the distribution of sample firms by legal status. Sample SMEs are private limited firms, public limited firms and sole proprietorships. About 64 percent of the firms are private limited firms.

Figure 6

Distribution of Firms by Legal Status

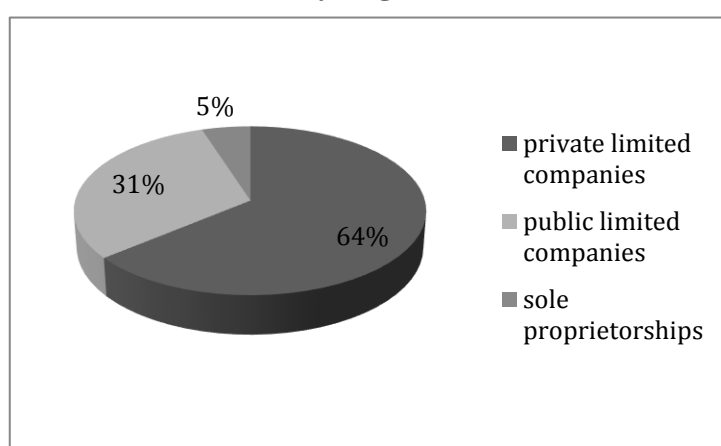
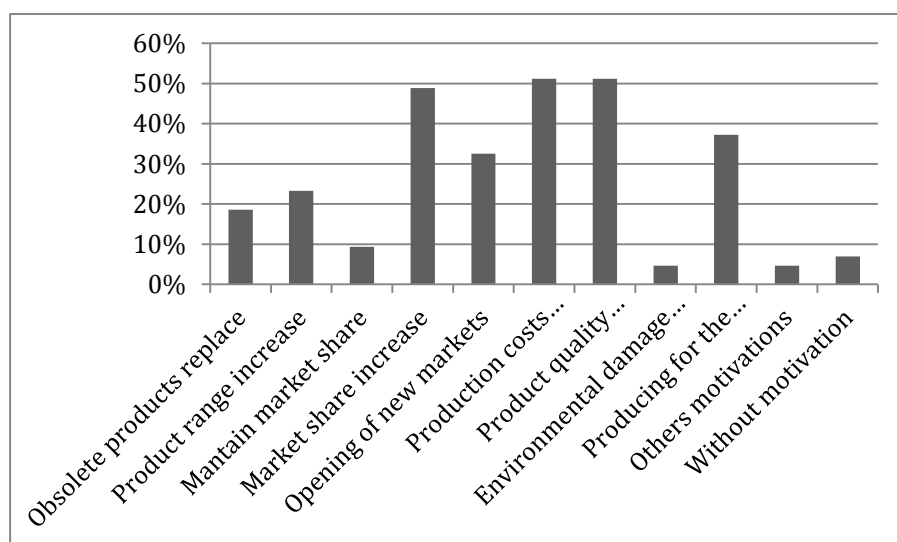


Figure 7 illustrates the motivations behind innovation of sample SMEs. About 50 percent of sample SMEs selected '*Market share increase*', '*Production costs reduction*', and '*Product quality improvement*' as motivations to invest in innovation. More than 30 percent

selected ‘*Opening new markets*’ and ‘*Producing for the customer*’ as their motivations. Around 20 percent selected ‘*Obsolete products replace*’ and ‘*Product range increase*’. ‘*Environmental damage reduction*’ and ‘*Maintain market share*’ do not seem to be major motivations to invest in innovation for nearly all the firms. Finally, 5 percent of sample SMEs state the existence of ‘*Other motivations*’ besides the ones listed to invest in innovation and 7 percent answered ‘*No motivation*’.

Figure 7
Portuguese Metalworking Industry SMEs’ Motivations for Innovation



MII results

Table 7 gives the proportion of sample SMEs that show to be innovative in each of the twenty partial indicators.

Table 7**Portuguese Metalworking Industry Percentage of Innovative SMEs by Partial Indicator**

Dimension	Partial Indicator	Percentage of SMEs
Knowledge and Technology	Product and/or processes innovations	44%
	Number of workers	44%
	PCT patents	3%
Economic effects	Volume of exports	59%
	Turnover	46%
	Patent revenues	6%
	New customers	66%
	New markets	43%
Human Resources	Workers by skill level	2%
	Training investment	82%
Processes and Infrastructure	Innovations made in products and/or processes	62%
	Partnerships with external entities for innovation	46%
	ICT use	51%
	Existence of R&D department in the company	27%
Strategy and Organization	Actions by top management in innovation activities	70%
	Encouraging innovation by organizational culture	61%
	Organizational innovations	53%
Accounting and Finance	R&D investments	46%
	Credit access for R&D investment	12%
Marketing	Marketing innovations	32 %

Table 8 shows the MII composite measures and elements for sample SMEs.

Table 8**Sample SMEs MII Measures and Elements**

Ho	Ao	MIIo	Hi	Ai	MIHi	MIIr	MIHa
0.289	0.605	0.175	0.422	0.633	0.267	0.654	0,221

Ho = 0.289, that is, 28.9 percent of the firms are multidimensional innovative in terms of outputs. Ao = 0.605 is the average intensity of multidimensional innovation in terms of outputs of innovative firms. Thus, MIIo = Ho x Ao = 0.175 measures the incidence of innovation on the interval [0, 1] in what concerns to outputs of sample SMEs. Hi = 0.422, that is, 42.2 percent of the firms are multidimensional innovative in terms of inputs. Ai = 0.633 is the average intensity of multidimensional innovation in terms of inputs of innovative firms. Thus, MIHi = Hi x Ai = 0.267 measures the incidence of innovation on the interval [0, 1] in what concerns to inputs of sample SMEs. MIIr = MIIo/MIHi = 0.654, lies on the interval [0, +inf], and corresponds to the TFP of multidimensional innovation of sample SMEs. MIHa = (MIIo+MIHi)/2 = 0.404 is the average incidence of innovation in terms of outputs and inputs.

Differences across SMEs

We analysed the links between individual output (input) multidimensional composite scores of the sample SMEs and the variables used to characterize the sample, which from now on we designate as Z variables. The exercise allowed us to identify groups of most innovative SMEs and to characterize them by calculating the corresponding MII specific vector (MIIo, MIHi, MIIr, MIHa).

Available Z variables refer to scale of operation (*‘Number of workers’* and *‘Turnover’*), the share of exports on the turnover (*‘Export ratio’*), classes of economic activity (*‘CAE #’*), the legal nature of the firms (*‘Public limited companies’* and *‘Sole proprietorship’*), and NUTS III territorial location (*‘Ave’*, *‘Baixo Vouga’*, *‘Entre Douro e Vouga’*, *‘Pinhal Interior Norte’*, *‘Pinhal Litoral’*, *‘Tâmega’*). In the case of qualitative

variables, such as CAE, legal nature of the firms, and NUTS III, dummy variables were considered distinguishing each class. NUTS III dummy variables capture observable and non-observable effects not captured by other Z variables taken into account and they were considered just for estimation consistency.

The regression results are presented in Table 9 (see Table A2 in the Appendix for a description of the data). Concerning outputs, with a level of significance of 1 percent, ‘CAE 33’ affects multidimensional output innovation negatively. The effects of ‘CAE 71’ and ‘Sole proprietorship’ are positive. In what refers to inputs, with a level of significance of 1 percent, ‘CAE 71’ affects SMEs’ multidimensional input innovation negatively. With a level of significance of 5 percent, the ‘Number of workers’ affects multidimensional input innovation positively, and ‘CAE 29’ has a negative effect. The coefficients of other variables such as ‘Export ratio’ are not significant, although having the expected signs.

Table 9
Explaining the Differences in Multidimensional Innovation across SMEs of the Portuguese Metalworking Industry

	Individual multidimensional output innovation scores			Individual multidimensional input innovation scores		
	Coef.	Std. Err.	P> t	Coef.	Std. Err.	P> t
Number of workers	-.0408019	.122759	0.742	.193586	.0960733	0.055***
Turnover	-.0125735	.1041714	0.905	-.148551	.0992244	0.147
Export ratio	.1401962	.1395888	0.325	.0717037	.1740312	0.684
CAE 24	-.0170588	.0646838	0.794	.1831002	.1860233	0.344
CAE 27	.1591182	.1228386	0.207	.0030383	.1947514	0.988
CAE 28	-.1161001	.1275327	0.371	-.0400064	.1026777	0.700
CAE 29	-.1644721	.1345675	0.233	-.1896955	.1064041	0.087***
CAE 30	-.0352127	.0992237	0.190	-.0761066	.1144514	0.635
CAE 32	-.3159792	.1104988	0.726	.1422944	.1177671	0.225

CAE 33	-.7529799	.1384438	0.008*	-.0421088	.10533	0.724
CAE 71	.2873827	.2131566	0.000*	-.7507715	.158575	0.000*
Public limited companies	.117701	.1006021	0.253	.0259934	.0990508	0.795
Sole proprietorship	.4992659	.1393381	0.001*	.5328428	.095483	0.000*
NUTS III Ave	.2402359	.115921	0.049**	.1481365	.2107581	0.489
NUTS III Baixo Vouga	.0368534	.1116215	0.744	.0688489	.1316794	0.606
NUTS III Entre Douro e Vouga	.0785803	.1220637	0.526	.0764345	.1163557	0.517
NUTS III Pinhal Interior Norte	-.2149487	.0975117	0.037**	.1206598	.100672	0.242
NUTS III Pinhal Litoral	.3185134	.1187465	0.013**	.4271834	.1096129	0.001*
NUTS III Tâmega	-.2268606	.1248031	0.081**	-.2134692	.1739322	0.231
Constant	.3205571	.0956796	0.003*	.327097	.1070552	0.005*

*p < 0.01; **p < 0.05, and ***p < 0.10; N=45; Robust estimation.

The results of the regressions must be interpreted with caution, given the small size of the sample and of each group of observations.⁵ Nonetheless, they confirmed some of the findings of the literature. In particular, the '*Number of workers*' (a measure of scale of operation) has a significant and positive impact on multidimensional input innovation, which supports the Schumpeterian hypothesis of a positive relationship between firm size and innovative activity, within the SMEs of the metalworking industry in Portugal. In Table 10, we distinguish the MII composite measures for the groups of small SMEs (with a number of workers from 10 to less than 50) and medium SMEs (with a number of workers from 50 to less than 250).

⁵ For instance, Sole proprietorship corresponds to only two firms of the sample (see Table A3 in the Appendix).

Table 10
Portuguese Metalworking Industry SMEs MII Composite Measures

	Sample	Small SMEs (10 to less than 50 workers)	Medium SMEs (50 to less than 250 workers)
MIIo	0.175	0.122	0.308
MIIf	0.267	0.192	0.449
MIIf	0.654	0.636	0.686
MIIf	0.221	0.157	0.404

From Table 10, it is noteworthy that the group of small (medium) SMEs has lower (higher) MII composite measures than the total sample.

5. Conclusions

We developed a Counting MII framework for measuring and benchmarking innovation of SMEs.

The methodology behind the MII is similar to the methodology behind the UN MPI and follows the innovation definitions stipulated by the OECD Oslo Manual, covering dimensions and partial indicators appearing in the IUS and in the GII. The MII framework starts by computing individual partial and multidimensional innovation output and input scores. The MII vector considers 4 multidimensional innovation composite measures that characterize the group of SMEs, the industry or the country being analysed: i) the first (MIIo) measures the incidence and intensity of innovation in terms of outputs, and lies on the interval [0, 1]; ii) the second (MIIf) measures the incidence and intensity of innovation in terms of inputs, and lies on the interval [0, 1]; and iii) the third (MIIf) measures the TFP of innovation

and lies on the interval $[0, +\infty]$; the fourth (MIIa) measures the average multidimensional innovation in terms of outputs and inputs, and lies on the interval $[0, 1]$.

To illustrate the MII, a survey was conducted among SMEs of the metalworking industry in Portugal. Results show that, in 2012, about 29 percent of SMEs in the Portuguese metalworking industry were multidimensional innovative in terms of outputs and 42 percent were multidimensional innovative in terms of inputs. The average intensity of multidimensional innovation of innovative SMEs in terms of outputs was 0.605, while the average intensity of multidimensional innovation of innovative SMEs in terms of inputs was 0.633. The MII vector was thus $[MII_o; MII_i; MII_r; MII_a] = [0.175, 0.267, 0.654, 0.221]$, showing that there is still much to be done in what refers to innovation of the industry SMEs.

Significant differences were found across sample SMEs in what concerns to individual multidimensional output and input innovation scores. These differences are explained by factors such as the number of workers, class of economic activity, and legal status of SMEs, a few of the Z variables considered. They allow to identify, characterize, and compare different groups of SMEs in the industry in what refers to multidimensional innovation, by computing the corresponding and specific MII vectors. This exercise was done for the groups of small and medium sample SMEs. When compared with the sample, the group of small (medium) SMEs has lower (higher) MII composite measures, which supports the Schumpeterian hypothesis of a positive relationship between firm size and innovative activity, within the SMEs of the industry.

There are issues to explore in future research. The set of partial indicators of innovation considered can be adjusted, particularly if analysing SMEs of developing countries. The weights and the cut-offs of the counting dual cut-off method employed are normative and they can also be adjusted by experts. These changes do not alter the nature and

the properties of the counting dual cut-off method employed to measure multidimensional innovations of SMEs.

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APPENDIX

TABLES

Table A1

Description of Portuguese Metalworking Industry CAE

CAE #	Description
CAE 24	Based metal manufacture.
CAE 25	Manufacture of fabricated metal products, except machinery and equipment.
CAE 27	Manufacture of electrical equipment.
CAE 28	Manufacture of machinery and equipment, n.p.
CAE 29	Manufacture of motor vehicles, trailers, semi-trailers and components for motor vehicles.
CAE 30	Manufacture of other transport equipment.
CAE 32	Other manufacturing industries.
CAE 33	Repair and installation of machinery and equipment.
CAE 71	Activities architectural, engineering and related techniques; activities of testing and analysis techniques.

Source: Statistics Portugal

Table A2

Description of regression variables.

Output Multidimensional Innovation Score	Input Multidimensional Innovation score	NUTS III #	CAE #	Number of Workers (10 ²)	Turnover (10 ⁶ euros)	Export Ratio	Legal Status
0.6334	0.7334	3	28	0.47	0.4577886	0.41	2
0.8001	0.6334	2	25	0.18	0.075926389	0.32	2
0.7001	0.4667	4	25	0.11	0.0622335	0.03	2
0.1667	0.3667	2	25	0.24	1.5	0.10	2
0.4834	0.4334	2	25	0.16	0.12	0.01	2
0.1	0.3	3	25	0.06	0.515191284	0.35	2
0.3	0.3	3	25	0.55	0.515191284	0.35	2

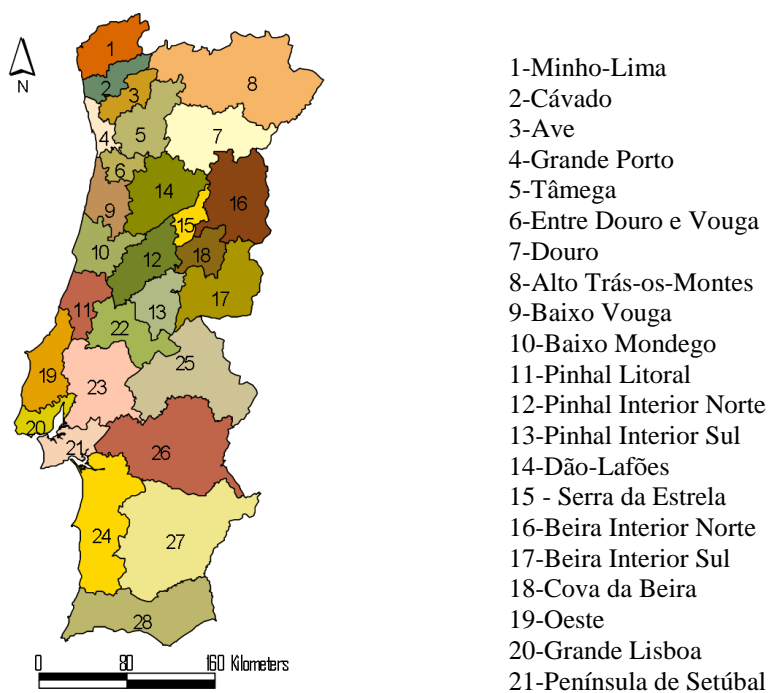
0.1667	0.2167	3	28	0.16	0.515191284	0.35	2
0.5501	0.4	4	25	0.3	0.36	0.70	2
0.2334	0.1	4	25	0.1	0.036	0.00	2
0.1834	0.1	4	29	0.23	0.19	0.00	2
0.8501	0.7334	4	28	0.12	0.075	0.25	3
0.5334	0	4	28	2.11	0.515191284	0.35	1
0	0.2	7	25	0.53725	0.515191284	0.35	2
0.6501	0.3667	4	24	0.06	0.515191284	0.35	2
0.15	0.3334	4	25	0.19	0.2	0.25	2
0.4334	0.1667	3	28	0.53725	0.515191284	0.35	2
0.2834	0.4667	4	29	1.02	0.5	0.96	1
0.4001	0.1667	2	25	0.36	0.17	0.53	2
0.76675	0.6334	1	24	0.78	0.7362103	0.93	2
0.4001	0.3667	7	27	0.14	0.515191284	0.35	2
0.5834	0.4667	2	24	1	0.50951	0.57	1
0	0	7	28	0.2	0.2470774	0.57	1
0.1	0.1	4	71	0.22	0.515191284	0.35	3
0.8001	0.7334	3	25	1.7	1.2	0.40	1
0.5084	0.4667	4	32	0.16	0.3783171	0.53	1
0	0	4	25	0.13	0.0230658	0.00	2
0.5501	0.7334	4	25	0.53725	0.515191284	0.35	2
0.5001	0.6834	2	27	1.3	0.7	0.11	1
0.3501	0.1	4	33	0.14	0.095893343	0.73	2
0.2834	0.4167	3	25	0.9	0.619872871	0.17	1
0.1	0	3	25	0.24	0.515191284	0.35	2
0.71675	0.3667	3	28	0.7	0.543006667	0.20	1
0.13335	0.38335	2	27	0.07	0.515191284	0.35	2
0.6834	0.5667	2	25	0.55	0.679	0.85	1

In the regression, we considered dummy variables for CAE, Legal status, and NUTS III classes. CAE, see Table A2 for description. Legal status: 1- Private limited firms, 2- Public limited firms, 3- Sole proprietorship. NUTS III, see Figure 1A for identification of number.

FIGURES

Figure 1A

NUTS III Mainland Portugal Map



Source: Wikimedia Commons