Identifying enablers of innovation in developed economies: A National Innovation Systems approach

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Abstract. It has been recognized that innovation drives the long-run economic growth of nations and increasingly governments are placing innovation at the center of their economic growth strategies. International variation in the investment on innovation presents an opportunity to examine key enablers of innovation-driving policy choices. Countries find themselves at different stages of economic development and innovation performance and so their relative levels of innovation inputs and outputs are likely to be different. In this study we employ a systems of innovation approach to examine what enables improvements in innovation potential among developed countries. Using data from the 2017 Global Innovation Index (GII) Report, we subjected 770 data measures to an analysis of 242 relationships involving changes in the GII’s innovation inputs/outputs scores and overall innovation potential of 35 OECD countries over a five year period (2012 to 2016). Our findings suggest that instituting policies that improve access to open and competitive markets is the most significant enabler for raising a developed country’s innovation potential.

Keywords. Innovation Enablers, Innovation Policy, OECD, Developed Economies, National Innovation Systems.


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1. Introduction

Countries increasingly recognize that innovation drives long-run economic growth and governments are putting innovation at the center of their growth strategies (INSEAD, 2017). Investment in innovation is an important contributor to productivity and the growth of an economy (Hassan and Tucci, 2009; Roberts, 1998; Chen, 2009; Chol, 1999). The extant literature on innovation’s role on economic growth is voluminous (Mansfield, 1980; Romer, 1986; Griliches and Mairesse, 1986; Stokey, 1995; Fagerberg, 1994; Kirchoff, 1994; Wennekers, 1999; Audretsch, 1995; Abiad et. al., 1989). Growing attention to the Schumpeterian theory of economic growth has drawn attention to the role that technological change and innovation play in achieving economic growth (Watkins et. al., 2015). This has attracted considerable research interest in the enablers of the innovation process itself (Crespo and Crespo, 2016; Furman, Porter and Stern, 2002; Lui and Buck, 2007; Grossman and Helpman, 1991). International variation in the investment on innovation presents an opportunity to examine key enablers of innovation-driving policy choices. Countries find themselves at different stages of economic development and innovation performance and so their relative levels of innovation inputs and outputs are likely to be different (Vivarelli, 2014; Watkins et. al., 2015; Crespo and Crespo, 2016). For innovation policy makers in developed countries the driver for improving their innovation potential is premised on the realization that while innovation (particularly that of a process orientation) can reduce employment, improving a country’s potential to innovate will result in increased investment in capital goods, lower prices to drive consumption, new products and higher average wages (Vivarelli, 2014). A comprehensive review of the general literature dealing with innovation enablement was undertaken to determine a suitable conceptual framework and provide empirical support for the relationship within inputs and outputs of innovation (predictor variables) and with the Global Innovation Index (GII), (criterion variable). An analysis of the rankings of indices helps policymakers and governments, particularly in industrialized countries, to identify paths for future development and design innovation policies (Balzat and Hanusch, 2004; Mahroum and Al-Saleh, 2013). The principal research question driving this study is: What enables improvements in innovation potential among developed countries?

In total, 770 data measures were subjected to an analysis of 242 relationships involving changes in the GII’s innovation inputs/outputs scores and overall innovation potential of 35 OECD countries over a five year period (2012 to 2016). This study adopted an indicators method of national innovation performance and considered the following characteristics (Grupp and Schubert, 2010): the indicators are of comparable importance as measures of the concept under study, the indicators are based on reliable statistics; the indicators hold their value over time; and the indicators are relevant to medium and long-term policy issues (Carayannis and Grigorioudis, 2014). To understand the key determinants that drive innovation, this study used Pearson correlation coefficients and stepwise linear regression modeling to observe the relationships among these determinants that might inform innovation policy-makers. This follows a “systems of innovation approach” (NIS) in terms of enablers of, or factors influencing, innovation. This approach includes economic, social, political organizational, institutional and other factors that influence the diffusion and use of innovation (Edquist, 1997b; Nelson, 1993, Lundvall, 1992).

We have organized our article by starting with a general discussion regarding innovation com-
petitiveness and globalization. This is followed by some background regarding the concept of national innovation systems and then a description of what is involved in the GII framework. The next section describes the methodology we employed in our study followed by our results and discussion. We conclude with a discussion of the implications of our findings and identify some of the limitations of our study.

2. Innovation Competitiveness

Technology entrepreneurs often compete in dynamic and fierce environments (D’Aeni, 1994) and need to develop and commercialize new products in order to exploit promising market opportunities, generate cash flows and make a profit (Haeussler et. al., 2012). A knowledge-based global economy emphasizes the need of a country to develop its innovative potential, given that the competitive performance of a national economy depends on the formation of intellectual capital and the society’s capacity to innovate (Carayannis and Grigoroudis 2014). Yet, developing, testing and commercializing innovations can be a costly and time consuming process and financial outcomes risky. In order to decrease the risks of innovation and achieve the benefits of innovativeness, it is important to understand the key enablers. One of the most important stimuli of innovativeness is public policy support (Wojnicka-Syez and Syez, 2016). The role of public policy in Schumpeterian economics generally reflects the importance of entrepreneurial start-ups in generating innovation, economic growth and competitiveness in globally linked markets (Ferreira et. al., 2017; Audretsch et. al., 2012). Researchers have sought to explain the determinants of innovation, from a micro perspective, by identifying a number of critical success factors of innovation such as the firm’s size (Fritsch and Meschede, 2001), strategy and social capital (Balachandra and Friar, 1997). Other research has shown innovation determinates from a macro perspective, mainly from the research and development (R&D) function, patents and governance (Fagerberg and Srholec, 2008; Becheikh at al., 2006a). According to an OECD survey on a sample of 20 OCED countries over the period 1982-2001, the main determinants of countries innovativeness appear to be the availability of scientist and engineers, research conducted in the public sector, business-academic links, the degree of product-market competition and the a high level of financial development and access to foreign inventions (OECD, 2005: 33).

The link between R&D and other macro functions is of great interest to governments (MacPher- son, 1997). Governments formulate policies and offer services destined to promote and support technological innovation, with the hope that it will translate into higher levels of innovation, growth and internationalization of firms (Raymond and St. Pierre, 2010; Ouellet and Raoub, 2006). Ultimately this strategy leads to innovation as a competitive advantage. Porter’s seminal work developed this linkage when he identified two existing models that make a country and organization competitive: the first model uses efficiency as the source of competitive advantage and is mainly operated by multinationals; the second model is based on innovation and growth to meet the individualized needs of the consumer (Porter, 1991). This environment emphasizes technological progress powered by entrepreneurship and innovation and leads to the creation of product and process innovations. Later work by the Global Entrepreneurship Monitor (2016) described this type of environment as “innovation driven” and found within advanced global economies. These economies have the ability to produce innovative products and services at the
global technological frontier using the most advanced methods and are the dominant source of competitive advantage. At this stage, the national business environment is characterized by the presence of deep clusters. Clusters become critical motors not only in generating productivity but encouraging innovation at the world frontier. Institutions and incentives supporting innovation are also well developed, increasing the efficiency of cluster interaction. Companies compete with unique strategies that are often global in scope, and invest strongly in advanced skills, the latest technology, and innovative capacity (Delgado et. al., 2012). Therefore, innovation-driven competitiveness is critical for a country’s long run economic performance (Carayannis and Grigoroudis, 2014).

3. A Systems Approach to Innovation

The concept of National innovation systems (NIS) have its origin in the influential work of Joseph Schumpeter. He recognized that innovation and ultimately economic growth is not autonomous, being dependent on factors outside of it. Since these factors are many, no one-factor can ever be satisfactory (Schumpeter, 1947). Schumpeter’s insight was further supported by research introduced in the late 1980s (Freeman, 1987; Dosi et. al., 1988) and expounded upon in subsequent years (Lundvall, 1992; Nelson, 1993; Edquist, 1997). Furman et al., (2002) defined NIS as the ability of a country – as both a political and economic entity- to produce and commercialize a flow of new-to-the world technologies over the long term and Oglobina et al. (2002) and Godin (2007) found that such ability relies on a larger system composed of national institutional sectors and their environments. Lunval (2007) identified two schools of thought in the literature about NIS. The first, tends to define innovation in a narrow sense by focusing on science and technology policy, and mostly analyzes the systemic relationships between R&D efforts in firms. The other school of thought looks at innovation in a broader sense and defines innovation as a continuous cumulative process involving not only radical and incremental innovation, but also the diffusion, absorption and use of innovation, beside science (Nasierowski, 2009). The framework also emphasized the relationships between the components or sectors as the “cause” explaining the performance of innovation systems (Godin, 2007). In short, a national innovation system can be perceived as a historically grown subsystem of the national economy in which various organizations and institutions interact with and influence one another in the carrying out of innovative activity (Balzat and Hanusch, 2004). The systemic approach to innovation is based on the notion of non-linear and multidisciplinary innovation processes. Interactions on the organizational level as well as the interplay between organizations and institutions are given central significance (Nelson, 1993). This in turn is consistent with the dynamics capabilities concept of strategy, that sees a firm’s ability to deploy and exploit resources as critical to its competitiveness (Amit and Schoemaker,1993; Teece et. al., 1997). For firms engaged in innovation commercialization within fast-changing environments, a Schumpeterian approach to innovation is likely to be more important (Lim et al., 2013). Ultimately, the NIS approach not only contributes to innovation, but is almost totally defined in terms of, and dedicated to, innovation as commercialization of technological invention (Godin, 2006a).

Connecting the level of NIS development with the level of economic advancement along structural and institutional development is critical in order to avoid defining pathways that are impossible
to achieve (Gu, 1999, Watkins et al. 2015). High income and low income countries demonstrate different paradigms of economic development, and as such, the dimensions that support their innovation performance should be distinct (Watkins et. al., 2015; Crespo and Crespo, 2016).

3.1 GII Innovation Framework

Since the late 1990s, the most collective approach to compare the performances of different innovation systems is the use of indices and rankings. The appeal of indices and rankings is based on political and operational importance for decision-making. An analysis of the rankings of these indices helps policymakers and governments throughout the world, particularly in industrialized countries, to identify paths for future development and design innovation policies (Sonorexza and Moodie, 2013). In 2007, the INSEAD Business School, Cornell University, and the World Intellectual Property Organization (WIPO) developed the GII to evaluate the level of innovative potential in national socioeconomic systems and to support the development of policies and practices that stimulate innovation. The use of this particular index as a measure of national innovation capability has been recognized in the recent innovation literature (Al-Sudairi et al., 2014; Crespo & Crespo, 2016; Jackson et al., 2016; Sohn et al., 2016; Prim et al., 2017). The GII depends on two sub-indices, each one developed on several enablers. Over time, this index has improved, and by 2015 included 79 indicators divided into 5 input enablers (institutions, human capital and research, infrastructure, market sophistication and business sophistication and two output enablers, knowledge and technology outputs and creative outputs.

4. Methodology

Each of the innovation inputs and innovation outputs contained within the GII framework are measured for each country by using a number of metrics that are combined to provide their respective scores. The input/output scores are then aggregated to provide an overall innovation potential score for each country in the study. The measures included in the GII framework are described in the Appendix. For the purpose of further analysis we have created a conceptual model as shown in Figure 1. The authors (Dutta et al. 2016, pg. 49) of the GII recognize that it is not a tool that can provide “the ultimate and definitive ranking of economies with respect to innovation” but they support their framework by stating that “The rich metrics can be used – on the level of the index, the sub-indices, or the actual raw data of individual indicators – to monitor performance over time and to benchmark developments against countries in the same region or of the same income category.”

With that observation in mind, and the empirical evidence within the literature that supports isolating high income from low income countries when using the GII for analysis (Crespo and Crespo, 2016), we have chosen to focus on the GII data associated with OECD member countries. Accordingly our principal research question that we seek to address is “what enables improvements in innovation potential among developed countries”?

To measure improvement or deterioration in a country’s innovation potential we have chosen to
compare the change in a country’s GII score with changes in both innovation inputs and outputs over a five year period (2012-2016) relying on data published annually by Cornell University, Insead and the World Intellectual Property Organization in their Global Innovation Index report. Our sample is comprised of 35 country members of the Organisation for Economic Co-operation and Development as listed in Table 1. Given the continuous nature of the scoring data, a Pearson correlation was run to determine if there were any significant relationships between increases or decreases in a country’s GII score over that five year period when compared to any increases or decreases in a country’s scores for each of the innovation inputs and outputs. Upon determining statistically significant relationships, and in order to explore for any predictive variables, stepwise linear regression modelling was undertaken. Our sample of developed countries are missing only three of the total number of developed economies as identified by the United Nations; Bulgaria, Croatia and Cyprus.¹

![Fig. 1. Drivers of Innovation Research Model after Dutta et al. 2016](image)

The change in GII score is the dependent variable and changes in those innovation inputs or outputs that were statistically significantly correlated to GII are the independent variables. In addition, we wish to further examine any innovation inputs or outputs that might influence the predictive variables by identifying any statistically significant correlations between them. As the GII score is determined using the respective country’s innovation inputs and output scores we are watchful for any collinearities that might exist in the data. Accordingly we will test for multi-collinearity between variables by calculating the variance inflation factor (VIF) in our regressions and employ the Durbin-Watson test on any regression models.

5. Results and Discussion

In total, 770 data measures were subjected to an analysis of 242 relationships. Table 2 provides the results of the Pearson correlations between the 5 year change in GII score and the corresponding changes in scores for innovation inputs. A statistically significant and strongly positive relationship \(R = 0.515, p = .002\) was found between a change in a country’s Trade, Competition and Market Scale (Trade et al) score and the change in its GII score. This result is not surprising given that improving the market conditions for trade, encouraging competition that might stimulate innovation among market participants, and increasing the scale of the domestic market and its innovation capacity could positively impact a country’s ability to advance its innovation potential. Our result also finds support in the literature. Baldwin and Gu (2004) in their empirical study of trade liberalization and increased export capacity in Canada found that improving access to export markets increased the innovative capacity of firms. In a literature review on the relationship of market scale and innovation, Dubois et al. (2015) found that for large industries like the pharmaceutical industry, the larger the market the greater the amount of innovation and while Stagni et al. (2017) found that increased competition can limit the type of innovation they also discovered that competition increases firm diversification and innovation exploration. A more moderate, yet statistically significant, positive relationship was found to exist between the change in a country’s GII score and its knowledge absorption score.

Table 1. OECD sample countries

<table>
<thead>
<tr>
<th>Country</th>
<th>Gross Domestic Product (GDP) (2016) in US$ billions(^1)</th>
<th>GII Income Group</th>
<th>GII Region</th>
</tr>
</thead>
<tbody>
<tr>
<td>Australia</td>
<td>1,256.6</td>
<td>High Income</td>
<td>SE Asia, East Asia, and Oceania</td>
</tr>
<tr>
<td>Austria</td>
<td>387.3</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Belgium</td>
<td>470.2</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Canada</td>
<td>1,532.3</td>
<td>High Income</td>
<td>Northern America</td>
</tr>
<tr>
<td>Chile</td>
<td>234.9</td>
<td>High Income</td>
<td>Latin America and the Caribbean</td>
</tr>
<tr>
<td>Czech Republic</td>
<td>193.5</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Denmark</td>
<td>302.6</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Estonia</td>
<td>23.5</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Finland</td>
<td>239.2</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>France</td>
<td>2,488.3</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Germany</td>
<td>3,494.9</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Greece</td>
<td>195.9</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Hungary</td>
<td>117.1</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Iceland</td>
<td>19.4</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Ireland</td>
<td>307.9</td>
<td>High Income</td>
<td>Europe</td>
</tr>
<tr>
<td>Israel</td>
<td>311.7</td>
<td>High Income</td>
<td>Northern Africa and Western Asia</td>
</tr>
<tr>
<td>Italy</td>
<td>1,852.5</td>
<td>High Income</td>
<td>Europe</td>
</tr>
</tbody>
</table>
---|---|---|---
Japan | 4,730.3 | High Income | SE Asia, East Asia, and Oceania
Korea, Republic of | 1,404.4 | High Income | SE Asia, East Asia, and Oceania
Latvia | 27.9 | High Income | Europe
Luxembourg | 61.0 | High Income | Europe
Mexico | 1,063.6 | Upper-Middle Income | Latin America and the Caribbean
Netherlands | 769.9 | High Income | Europe
New Zealand | 179.4 | High Income | SE Asia, East Asia, and Oceania
Norway | 376.3 | High Income | Europe
Poland | 467.4 | High Income | Europe
Portugal | 205.9 | High Income | Europe
Slovakia | 90.3 | High Income | Europe
Slovenia | 44.1 | High Income | Europe
Spain | 1,252.2 | High Income | Europe
Sweden | 517.4 | High Income | Europe
Switzerland | 662.5 | High Income | Europe
Turkey | 735.7 | Upper-Middle Income | Northern Africa and Western Asia
United Kingdom | 2,649.9 | High Income | Europe
United States | 18,561.9 | High Income | Northern America

1 As reported by GII

The GII measures knowledge absorption principally through the ability of a country’s high tech sector to acquire information and information technology related to innovation, to encourage inflows of foreign investment that will allow domestic firms to be exposed to global innovation, and the measure of local professionals engaged in acquiring intellectual property. To find in the data a positive relationship between greater knowledge absorption and a county’s improved GII score is also reasonable and is consistent with the work of Forés and Camisón (2016) who found that the absorptive capacity of firms has a significant positive effect on innovation performance.

Table 3 contains the results of the Pearson correlations between the 5 year change in GII score and the corresponding changes in scores for innovation outputs. Moderate and statistically significant relationships exist between the increase in a country’s overall innovation score and its ability to improve knowledge creation capabilities (R=.475, p = .004) and the ability to diffuse that knowledge (R=.386, p = .022). The latter result is interesting because the GII framework is designed so that knowledge diffusion output measures are a “mirror” of the knowledge absorption
input measures (Dutta et al. 2016, p.55). As opposed to the ability to acquire information as a benefit for enhancing innovation, the diffusion output measures the value received (royalties and fees, technology exports, foreign direct investment) for being able to disseminate the knowledge derived from innovation activities. Given the framework design, to find similar results for both as significant contributors to a country’s overall innovation potential is therefore not unsurprising. Furthermore, these results are consistent with the description by Zanello et al. (2016) that the innovation process is one where investments are made in knowledge building activities that support the creation, diffusion and absorption of innovation. Finally, a moderate relationship exists between that the intangible assets associated specifically with creative outputs and the overall growth in a nation’s innovation standing (R=.486, p = .004). The output measures include national trademark and design applications as well as the use of ICT within the businesses and organizations of the country. The literature continues to support these measures as contributors to improving regional or national innovation potential and support our findings here (Billon et al. 2017; Khedhaouria and Thurik, 2017).

Having identified some significant relationships between improvements in specific inputs and outputs and overall improvement of a country’s innovation standing we now focus on their predictive capabilities. Stepwise regression was employed to determine the relative predictive contribution of each

**Table 2. Correlations – Change in Innovation Inputs Score with Change in Global Innovation Index Score (n=35)**

<table>
<thead>
<tr>
<th>Human Capital and Research</th>
<th>Education</th>
<th>Tertiary Education</th>
<th>Research and Development</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pearson Corr.</td>
<td>.008</td>
<td>.063</td>
<td>.231</td>
</tr>
<tr>
<td>Sig.</td>
<td>.963</td>
<td>.720</td>
<td>.181</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Infrastructure</th>
<th>Information Communication Technology (ICT)</th>
<th>General Infrastructure</th>
<th>Ecological Sustainability</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pearson Corr.</td>
<td>.080</td>
<td>.214</td>
<td>-.297</td>
</tr>
<tr>
<td>Sig.</td>
<td>.648</td>
<td>.217</td>
<td>.083</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Market Sophistication</th>
<th>Credit</th>
<th>Investment</th>
<th>Trade, Competition &amp; Market Scale</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pearson Corr.</td>
<td>-.157</td>
<td>-.135</td>
<td>.515**</td>
</tr>
<tr>
<td>Sig.</td>
<td>.367</td>
<td>.438</td>
<td>.002</td>
</tr>
</tbody>
</table>
Table 3. Correlations – Change in Innovation Outputs Score with Change in Global Innovation Index Score (N=35)

<table>
<thead>
<tr>
<th>Knowledge and Technology</th>
<th>Creative Outputs</th>
</tr>
</thead>
<tbody>
<tr>
<td>Knowledge Creation</td>
<td>Intangible Assets</td>
</tr>
<tr>
<td>Knowledge Impact</td>
<td>.475**</td>
</tr>
<tr>
<td>Knowledge Diffusion</td>
<td>.486**</td>
</tr>
<tr>
<td>Pearson Corr.</td>
<td>.004</td>
</tr>
<tr>
<td>Sig.</td>
<td>.004</td>
</tr>
</tbody>
</table>

* Correlation is significant at the 0.05 level (2-tailed).
** Correlation is significant at the 0.01 level (2-tailed).

to multi-variate regression models. Table 4 provides the results of the regressions with Trade et al, Knowledge Diffusion and Intangible Assets being included in the predictive models and Knowledge Absorption and Knowledge Creation being excluded. Model 3 combined all three predictive variables improving the total correlation co-efficient or strength of association (R=.714) and resulting in a relatively high explanation of variance (R²=.549). The validity of Model 3 was examined by firstly adjusting the R² for the number of independent variables included in the model. The result indicating that the model continues to explain or predict over half of the variability of the increase in the GII score (Adjusted R² = .505). A Durbin-Watson test was conducted to identify if any autocorrelation or non-independence exists within the variables and the test outcome approaches 2 (1.984) confirming the independence of the data. Finally, multicollinearity within the data was analyzed to determine if the variance of the co-efficient estimate was being influenced by any collinearity between the predictor variables. Both the Tolerance and VIF tests approach 1 suggesting little multicollinearity. We rely on the work of Knofczynski and Mundfrom (2008) to confirm whether the level of predictive strength given the sample size of 35 is appropriate. With an R² of 0.549 and three predictor variables in our model, their sample size recommendation for a “good prediction level” (pg. 438, Table 1) lies between 33 and 39.

Table 4. Predictive model (stepwise regression) for GII improvement

<table>
<thead>
<tr>
<th>Model</th>
<th>R</th>
<th>R²</th>
<th>Adjusted R²</th>
<th>R Square Change</th>
<th>Durbin-Watson</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>.515*</td>
<td>.265</td>
<td>0.243</td>
<td>0.265</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>.695b</td>
<td>.484</td>
<td>0.451</td>
<td>0.218</td>
<td></td>
</tr>
</tbody>
</table>
The strongest singular predictive variable; Trade et al., is moderately and positively ($R = .489$, $p = .003$) associated with the Research and Development score. This association was also found by Silaghi et al. (2014) in their economic study of eastern and central European countries where they found that increases in research and development by businesses in those countries were linked to economies that encouraged more trade and competitive (open) markets. Furthermore, scores for both Trade et al. and Research and Development...
were also found to have moderate positive relationships ($R = .569 \ p = .000$ and $R = .343 \ p = .043$ respectively) with improvement in the Intangible Assets score, another of Model 3’s predictive variables. The association of trademark applications, industrial design applications, and ICT with (Intangible Assets measures) the measures of Research and Development (R&D expenditures) would seem logical as they are recognized outputs of the R&D process. More significant is the relatively strong relationship with greater trade, more competitive markets and increased market scale, further confirming the collective role they have in promoting growth in innovation.

There are a three negative associations that have arisen from our analysis and they are worth discussing here. The strongest and most statistically significant is the inverse relationship between the change in a country’s Trade et al. score and its Investment score ($R = -.659 \ p = .000$). This result suggests that those countries who increased trade and market competition experienced decreased levels of domestic investment in publicly-traded companies. With increased competition at home, domestic investors might wish to seek opportunities elsewhere to invest their capital especially if external markets are opened as a result of the introduction of open trade policies. We found another more moderate negative correlation ($R = -.347 \ p = .041$) between a country’s Credit score and its Knowledge Diffusion score. As countries improved the ability to borrow they saw a decline, in some combination of high-tech and ICT exports, related royalty and licence fees, and net outflows of FDI as a percentage of their gross domestic product (GDP). It is likely the effect is due to the latter, rather than the former, as improving domestic borrowing capacity should encourage domestic lenders to loan capital at home and perhaps limit their investment in external economies, thus reducing the FDI outflows. With the former, increased access to capital should support the ability to fund, and therefore increase, exports. Our final negative association was also a moderate one and was observed between the Trade et al. and Ecological

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Sustainability innovation inputs \((R = -0.337 \ p = 0.047)\). From our finding it could be argued that even in developed countries the improving of trade, competition and market scale results in a reduction of some combination of energy efficiency, and environmental performance and compliance. This observation is consistent with the literature in finding that increased open trade and market scale led to greater negative environmental footprints (Al-Mulali and Ozturk, 2015), even specifically within OECD countries (Shafiei and Salim, 2014).

6. Conclusion and Limitations

The principal research question driving this study was: What enables improvements in innovation potential among developed countries? Developing economies can no longer rely on labour-intensive industries for economic growth and job creation as those industries have moved to developing countries with lower wages and lower costs of operation. In developed countries, policy innovation should remain focused on encouraging inputs and outputs that improve a nation’s potential to innovate. This study explored innovation enablers that have a relationship with high innovation performance using an NIS approach. We relied on correlations and regression to identify the specific enablers that contribute to improving a country’s innovation potential. By using GII data, we have attempted to understand what NIS’ key elements may be more relevant to designing policy for enhancing innovation performance. In this study, we subjected 770 data measures to an analysis of 242 relationships between GII’s innovation inputs, outputs and overall innovation potential among 35 OECD countries resulting in the identification of statistically significant relationships between the improvement of certain input/output scores and a nation’s total GII score. We used those relationships to produce a predictive regression model that could inform policy on improving a country’s innovation capacity. Before stating our conclusions we need to be clear that we are not suggesting that policies to promote innovation within developed countries should exclude any of the many inputs and outputs identified in the GII. Rather, we have attempted to isolate those input/outputs that policy makers may wish to focus on when considering innovation policy.

In terms of our results, we were surprised to find that, with the large number of innovation inputs and outputs contained within the GII framework, there were such few statistically significant relationships found when comparing changes in input/output scores with changes in the total GII score over a recent five year period. However, we find the aggregate predictive model that did emerge is significant in its strength and fit. Therefore we conclude that for the developed nation, the initial focus might be on policies that promote open, competitive markets and the enabling of hi-tech and ICT exports. Furthermore, policy support for continued creative outputs and the development and use of ICT might also be considered. As discussed in our results, there are important relationships with other innovation inputs and outputs that must be considered when structuring policy as suggested. Certainly, the resulting positive association between open, competitive markets that encourage trade, competition and increase in market scale and increased research and development and creative output would be seen by policy makers as beneficial to increased innovation capacity.

However, caution must be taken as some moderate to strong negative associations may exist when
focusing policy on the results of our predictive model. The most significant may be the potential for decreased investment in domestic capital markets as policies for expanding open-trade and increased competition might encourage domestic investment to seek opportunities externally, especially if reciprocal open market agreements provides access to external markets. The other is the potential environmental impact arising from policies that stimulate economic growth such as easing trade and opening markets. Policy makers may need to consider the importance of balance between innovation and economic need with protection of the environment. Perhaps more importantly, policy that continues to support environmental innovation can be integrated into open market policies in order to mitigate some of the potential harm that may arise.

There are some limitations with our study. We have relied on the integrity of the data presented within the GII report and would recommend that readers review the methodology contained within that report in order to assist in understanding why we have chosen to use this data. Furthermore, we have chosen the scores for each of the inputs and outputs identified by the GII for our statistical measures but we recognize that each input/output is an aggregation of a number of measures that make up the score. Further research might look into specific correlations and regressions within this larger data set. We have also limited ourselves to the most recent five year period and the changes in score of each input/output for the 35 OECD countries to generalize on innovation policy for developed nations. The statistical significance of the correlations and the prediction level of our model may be reasonable but we must rely on the fact that our sample represents almost all of the countries currently defined as developed countries by the United Nations in order for us to make our generalizations.

7. References


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Biographies

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