

COMPLEMENTARITY AMONG INNOVATION STRATEGIES: EVIDENCE FROM COLOMBIA*

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ABSTRACT

Using Innovation Survey Data from Colombia, we apply a new methodology to assess complementarity relationships among different innovative inputs. This approach allows scholars to prevent serious data drawbacks, such as the lack of proper performance measures and is also fairly general, allowing cross equations restrictions. Empirically, we found evidence for the absorptive capacity hypothesis and the key role played by ICT.

Key words: *Innovation, Innovation Survey, Latin America, Complementarity.*

RESUMEN

A partir de los datos de la Encuesta de Innovación de Colombia, se aplicó una nueva metodología para evaluar las relaciones de complementariedad entre los distintos “inputs” innovativos. Este enfoque permite a los académicos evitar graves inconvenientes en los datos, tales como la falta de medidas para un desempeño adecuado y además es bastante general, permitiendo cruzar las restricciones de las ecuaciones. Empíricamente, encontramos evidencias para la hipótesis de la capacidad absorptiva y del papel fundamental que desempeñan las TIC.

Palabras clave: *Innovación, Encuesta de Innovación, Latinoamérica, Complementariedad.*

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RESUMO

Usando os dados da pesquisa de Inovação provenientes da Colômbia, nos aplicamos uma nova metodologia para avaliar as relações de complementaridade entre os diferentes “inputs” inovadores. Esta abordagem permite que os acadêmicos evitem graves inconvenientes nos dados, tais como a falta de medidas pelo desempenho adequado e também bastante geral, permitindo as restrições das equações sejam cruzadas. Empiricamente, encontramos evidências para a hipótese da capacidade absorptiva e do papel fundamental que desempenham as TIC.

Palavras-chave: *Inovação, Levantamento de Inovação, América Latina, Complementaridade.*

JEL: O30, O33, O54.

INTRODUCTION

Modern growth theory identifies R&D as innovation *tout cours*. In fact the Neo Schumpeterian tradition à la Aghion-Howitt (Aghion and Howitt, 1998) endogenizes the technical progress through a relation between new products discovery and R&D activity.

Nevertheless, the empirical literature has provided large evidence that R&D is only part of the overall innovation expenditure (Mohnen and Roller, 2005; Mairesse and Mohnen, 2004 for evidence using Community Innovation Survey data). On the one hand, the global markets and the recent advances in science and technology make very difficult for a firm to rely only on internal sources (Arora and Gambardella, 1990; Cassiman and Veugelers, 2006). On the other hand, there is a large group of industries that perform innovative efforts through acquisition of know-how, adoption of capital embodied technology and incremental and non-formalized process innovation. This latter statement goes back to the very origin of the economics of innovation and the possibility to classify industries according to their technological trajectories (Bogliacino and Pianta, 2013; Dosi, 1988; Pavitt, 1984).

Allowing for complex interactions among innovation activities raises a set of related questions. If we can speak of *knowledge production function* in the same way in which standard economics deal with activity analysis, can we speak of complementarity/substitutability relationships?

As an example, some contributions put forth the hypothesis of absorptive capacity (Leiponen, 2005; Griffith et al., 2003; Cohen and Levinthal, 1990), stressing the key role of R&D labs in the absorption of knowledge produced elsewhere. In other words, engaging in some research activity generates internal capabilities that can be exploited to learn or imitate the codified knowledge available for sale (or through spillover) on the market.

We talk about complementarity whenever the return on one input is increased by the contemporaneous adoption of other activities. Generally speaking, these properties are local, in the sense that they may not work everywhere in the domain of the activities space. As an example, there can be threshold effects: complementarity holds only once a certain level of use of some of the inputs is implemented. Topkins (1998) and Milgrom and Roberts (1995) provide a full characterization of these properties using the theory of *supermodular functions on lattice* structures.¹

Recently the issue of complementarity has been raised for environmental friendly innovation. Given the urgency of reducing Carbon Dioxide emissions due to the threat of Global Warming, the issue of

¹ A lattice is a partially ordered set in which any two elements have a *supremum* (also called a least upper bound or join) and an *infimum* (also called a greatest lower bound or meet). A function is defined supermodular if for any two elements of its domain, the sum of the counterimages of the two elements are lower than or equal to the sum of the counterimages of their supremum and infimum.

complementarity has a strong policy relevance to achieve the required targets (Mazzanti and Zoboli, 2009 and 2008).

Econometrically, testing for complementarity in this general approach poses a series of empirical problems, starting from the need of an explicit measure of performance, which may not be available in innovation surveys. This is precisely the kind of problem we have to face with the Colombian data from the *Segunda Encuesta de Desarrollo e Innovación Tecnológica* (Second Innovation Survey). In the literature, most of the scholars have relied on indirect measures. Although we follow this line of research, one of the main novelty of this work is the use of a more general approach, based on a Tobit I system.

Complementarities are important from a policy perspective. If firms are not able to internalize them, inefficiencies arise: as in the case of any positive externality, economic agents tend to underinvest. Managerial slackness, lack of proper organizational practices scarce information circulation in compartmentalized large firms may be reason behind difficulties to capture these externalities.

This paper proceeds as follows: Section II discusses Data and Methodology, Section III presents the results, Section IV concludes.

I. DATA AND METHODOLOGY

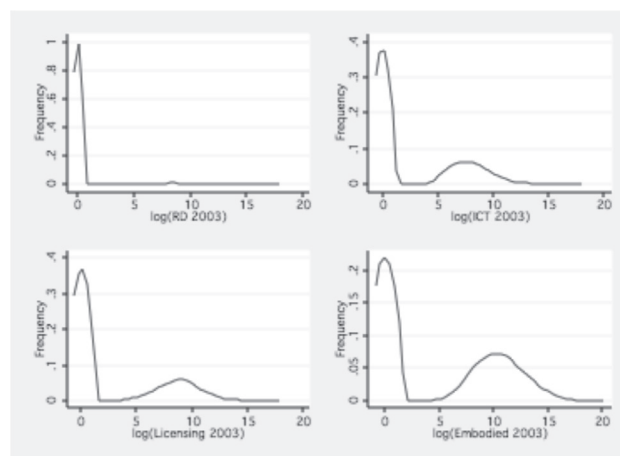
We use data from the *Segunda Encuesta de Desarrollo y Innovación Tecnológica*, an innovation survey over 6670 Colombian manufacturing firms. For each company, the survey collects information over innovative expenditure in 2003 and 2004. The Survey design is based on the *Manual de Bogotá (2001)*, an adaptation of the Oslo Manual (OECD 2005). The expenditure is deflated using the GDP Deflator (taken from the Colombian Statistical Office, the DANE) using as a base year the 2003.

We consider four innovation inputs: R&D, licensing, Information and Communication Technology (ICT) expenditure, and capital embodied technology (CET) acquisition. In Figure 1 and 2 we plot the distributions of the variables in the two years. The pattern is quite striking: as can be seen there is a density mass in zero, given by a large number of firms without any expenditure; secondly, there is a relatively large support of the distribution. The pattern is qualitatively similar across types of expenditure and within years.

In order to use log demand without introducing selection bias, we perform the following data transformation: we let the demand equal to zero whenever the expenditure is null and equal to the log of the expenditure whenever the latter is positive (there are no “1” value, so there isn’t induced measurement error).

There are two different approaches to empirically test for complementarity (for a general discussion see Catozzella and Vivarelli, 2007): a direct approach and an indirect approach.

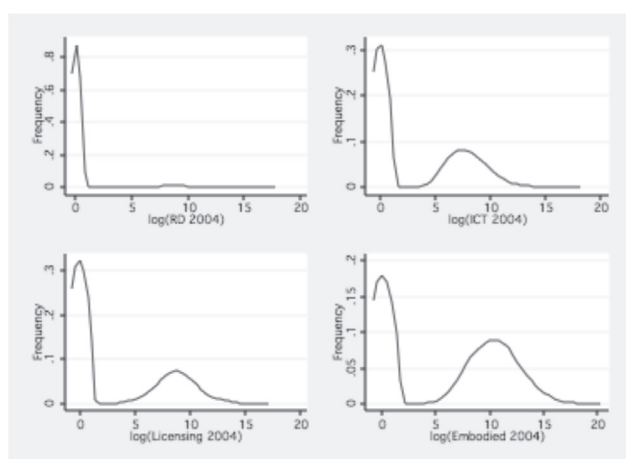
Figure 1. The Distribution of Innovative Expenditure in 2003.



The direct approach is based on the idea that the simultaneous adoption of different inputs should prove to be more valuable than adopting each

one individually. Empirically, it can be assessed through the estimation of the effect of alternative combinations of inputs over some measure of economic performance. This is equivalent to identifying the effect of the (cross) second derivative. We can write a system of inequalities and test it. It is computationally demanding but feasible with the proper data (Lokshin et al., 2004; Catozzella and Vivarelli, 2004).

Figure 2. The Distribution of Innovative Expenditure in 2004.



Given the difficulties related to the direct approach, most authors use an indirect one. They test the implications of complementarity instead of complementarity itself. In fact if the latter holds firms will at least partly internalize it and *ceteris paribus* demands will move together. The advantage of this method is that we do not need a measure of payoff. Econometrically, some authors use cluster techniques (Athey and Stern, 1998), other ones propose a correlation analysis (Arora and Gambardella, 1990).

In this article we follow the indirect approach. The rationale behind this choice is the lack of economic performance measure in Colombia's Innovation Survey.

To assess co-movements among innovative input demands, we calculate the unconditional

correlations, as a first measure, and the robust conditional correlations, i.e. controlling for all the observable characteristics of the firm and of the environment. To compute the residuals, we directly estimate the system of input demands (for four innovative inputs), keeping into account the potential corner solutions and accommodating cross-equations restrictions.

There are two basic econometric issues to face. First of all, the investment is highly persistent in the period covered by the analysis. Regressing the 2004 value over the one for the previous year, we get very high coefficients, as shown in Table 1. This is consistent with the evidence of firm level investment and the existence of financial frictions (Holt, 2003). Its time profile is thus not continuous. For our analysis there are two possibilities: either using the lag of the expenditure variable as a regressor or considering the total amount spent over the two years.

Table 1. Persistence in Expenditure Variables

Variable	Lag's coefficient
ICT	0.79 [92.00]***
R&D	0.87 [102.21]***
Licensing	0.65 [62.61]***
CET	0.58 [56.02]***

T statistics in brackets, * significant at 10%, ** at 5%, *** at 1%

Source: *Segunda Encuesta de Desarrollo e Innovación Tecnológica*

The first option ends up not being viable: since we have just one point in time for all variables (except for the expenditure ones), we cannot control for unobserved heterogeneity, so the

lag is potentially endogenous. Moreover, since we are in presence of corner solutions models, we need to use nonlinear models. For the latter, identification require sequential exogeneity of the regressors. We decided to run test of exogeneity, using a simple two steps procedure illustrated in Woolridge (2002), which is strongly confirmed;² so we proceed using the total amount spent. We transfer all the money to 2003, using a nominal interest rate at 1 year.

The second issue is that the demands for the various inputs have to be estimated jointly: we have to write a system of non-linear equations, since the corner outcome at zero require a Tobit I estimation. As one may think, writing a full simulated maximum likelihood program (SML) is going to pose major problem in terms of computational effort, and convergence may be very difficult. However, we know from Cameron and Trivedi (2005) that by maximizing the sum of the individual equation (quasi maximum likelihood, QML) we get consistent estimates of the *Betas* (allowing the computation of the residuals). Moreover, Barslund (2007) recently compares SML and QML for a demand system (using Monte Carlo simulations): his finding point out that the former never provides a statistically significant improvement. For these reasons, we proceed by estimating the full system with QML, then computing the residuals and finally calculating the correlation matrix.

After a careful revision of the literature we use the following battery of controls:

- The share of graduated workers, the share of workers with some college, the share of workers with secondary education or less as a measure of skills and capabilities (Piva and Vivarelli, 2009);
- The foreign share of capital (which controls for multinational), a dummy for large firms (more

than 250 employees) as a measure of scale effects (Cohen and Levin, 1989);

- A set of dummies for hampering factors declared by the firm (Crépon et al. 1998);
- The rate of change of value added at three digits CIIU³ (for both 2002-2003 and 2003-2004)⁴, to control for demand pull effects (Piva and Vivarelli, 2007; Kleinknecht and Verspagen, 1990; Scherer, 1982);
- A dummy for firms that innovate under cost pressures, as a measure of technology push driver (Mowery and Rosenberg, 1979);
- Finally, a dummy for the source of innovation (equal to one if the idea comes from an internal department).

II. RESULTS

In Table 2 and 3 we report the unconditional and conditional correlation matrices, respectively.

The results suggest two main conclusions: on the one hand, there is the strong correlation between licensing and R&D, which we interpret as empirical evidence for absorptive capacity (e.g. without a lab it is difficult to manage a patent); on the other side, we underline that the ICT is correlated with all inputs, and show the largest coefficients, which suggests that investing on new technologies can help manage other innovative inputs.

At the same time, the poor but positive correlation between R&D and CET is certainly driven by the contrasting effect between the make or buy decision (from one point of view CET is an alternative to internal research) and the complementarity between the lab machinery and R&D.

2 Results are available from the authors upon request.

3 Clasificación Industrial Internacional Uniforme de todas las Actividades Económicas (harmonized with the International Standard, ISIC).

4 Data are taken from the Annual Manufacturing Survey (*Encuesta Anual Manufacturera*).

Table 2. Unconditional Correlation Analysis

	R&D	Licensing	ICT	CET
R&D	1			
Licensing	0.2746 [0.0000]	1		
ICT	0.2059 [0.0000]	0.4613 [0.0000]	1	
CET	0.1898 [0.0000]	0.4109 [0.0000]	0.4375 [0.0000]	1

p-values in brackets

Source: *Segunda Encuesta de Desarrollo e Innovación Tecnológica***Table 3.** Conditional Correlation Analysis

	R&D	Licensing	ICT	CET
R&D	1			
Licensing	0.3247 [0.0000]	1		
ICT	0.1801 [0.0000]	0.3148 [0.0000]	1	
CET	0.0943 [0.0000]	0.1491 [0.0000]	0.1573 [0.0000]	1

p-values in brackets

Source: *Segunda Encuesta de Desarrollo e Innovación Tecnológica***Table 4.** The system for Innovative Input Demand

	(1) Log (Licensing)	(2) Log (ICT)	(3) Log (R&D)	(2) Log (CET)
Total Copyright	0.898 [11.13]***	0.434 [6.47]***	0.850 [4.47]***	0.162 [4.34]***
Graduated Employees	0.003 [0.77]	0.010 [4.20]***	0.045 [6.42]***	0.006 [3.16]***
Employees with Some College	0.001 [0.79]	-0.001 [-0.52]	-0.015 [-2.91]***	-0.000 [-0.14]
Employees with Secondary School	0.001 [1.28]	0.001 [2.15]**	0.003 [1.73]*	0.002 [4.76]***
Foreign Share of Equity	0.008 [1.23]	0.004 [0.75]	-0.003 [-0.17]	0.021 [6.52]***
Large Firms Dummy	1.021 [1.80]*	0.545 [1.08]	0.115 [0.07]	2.646 [8.39]***
Growth of Value Added at CIIU 3 dig, 2002-2003	-0.997 [-1.12]	0.302 [0.43]	0.783 [0.27]	0.981 [1.94]*
Growth of Value Added at CIIU 3 dig, 2003-2004	-0.661 [-0.61]	1.037 [1.20]	-10.061 [-2.73]***	0.904 [1.41]
Factor Saving Innovators	1.679 [5.81]***	1.031 [4.34]***	3.089 [3.24]***	2.152 [12.77]***
Internal Source of Idea	2.624 [3.27]***	3.414 [4.80]***	11.000 [2.80]***	5.350 [8.67]***
Constant	-4.833 [-5.95]***	-2.727 [-3.82]***	-40.169 [-10.16]***	0.561 [0.90]
Sigma	8.252 [79.68]***	6.950 [78.96]***	16.713 [58.58]***	5.438 [65.42]***
Hampering Factors	Yes	Yes	Yes	Yes
Observations	4813			
Log Pseudo Likelihood	-27930.59			

Tobit I system, estimation with QML.

Robust z statistics in brackets

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

Source: *Segunda Encuesta de Desarrollo e Innovación Tecnológica*

The above results suggest that innovation policies should be multifaceted and especially focused on supporting research activities and adoption of ICT.

Table 4 reports the full system. Regarding this last one, it should be stressed how firms with formalized innovation (captured through the copyright measure) invest more in all inputs. Human capital impacts significantly over three of the four inputs. Size and multinational dimension seem to impact only on CET expenditure, probably reflecting the general feature of the Colombian Industrial structure, where R&D and other more “make” choice of innovation are likely to be less important.

CONCLUSIONS

In this paper we show how to assess complementarity in the knowledge production function when data do not provide performance measure. This approach can be very important in dealing with Latin American Innovation Surveys that still suffer some limitations.

Our identification strategy follows a new approach; we assess complementarity through an indirect approach based on the estimation of the demand system, using a QML estimator, which is both robust and computationally cheap.

Our results stress a significant role played by ICT in driving innovative effort and the role of R&D as absorptive capacity.

Through the identification of the complementarity relationships, this framework allows a deeper exploration of the effects of public interventions and subsidies. In fact most of the literature considers additionality versus crowding-out in a pure unidimensional input perspective without considering neither the possible interrelationships between different inputs nor the relationship between input and output (Catozzella and Vivarelli, 2011).

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