

Comments on “Research, Innovation and Productivity in Four European Countries”

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Introduction

- The combination of the substantive questions, econometric modelling, and the CIS data makes this an exciting line of research.
- I was left with the desire to know more about the CIS questionnaires and data.
- The contribution of the paper is mainly empirical: Exploring variants of the Crépon-Duguet-Mairesse model with a new wave of the CIS data for four countries.
- Main results seem to be in line with previous findings.
- More work is needed to understand the sources of cross-country heterogeneity.
- I will provide a quick summary of the model, together with some comments focusing on the empirical strategy.

A Model for innovators

- Determinants of R&D

- *R&D participation*: Probit for being an R&D performer

$$\Pr(s = 1 | x) \equiv p(x) = \Phi(xb_0)$$

$$s^* = xb_0 + u_0$$

- *R&D intensity*: Letting r be log R&D expenditure per employee for performing firms

$$E(r | x, s = 1) = xb_1 + \gamma\lambda(xb_0) = xb_1 + g[p(x)]$$

$$r^* = xb_1 + u_1$$

- Production of knowledge:

- *Process innovation* probit ($pc = 1$ if firm is a process innovator). Letting $x = (x_A, x_B)$:

$$pc1(a_2s^* + x_Ab_2 + u_2 > 0)$$

$x_A = \{\text{demand pull, cost push}\}$. $x_B = \text{external instruments}$.

I abstract from a subset of x that appear in all equations.

- *Product innovation*. Let $spd = 1$ if product innovator; $zinno = \text{logit-share of innovative sales for } spd = 1 \text{ firms}$:

$$spd = 1(a_3s^* + x_Ab_3 + u_3 > 0)$$

$$zinno = a_4s^* + x_Ab_4 + u_4$$

Other versions use r^* in place of s^* .

- Productivity equation. Let $prod$ be log output per worker:

$$prod = a_5zinno + u_5$$

Other versions use s^* or r^* instead of $zinno$.

Comments about the model

- No discussion of the selection problem arising from the fact that only innovating firms are used. There are two aspects:
 - (a) Effect of R&D on innovation: some non-innovating firms may be R&D performers.
 - (b) Effect of innovation on productivity: there may be different effects at the non-innovation margin.
- Not clear why using s^* or r^* in the innovation equations as opposed to s or r .
- The variable s^* is a propensity to perform R&D, but one would expect production of innovations to depend on actual performance rather than propensities.
- The construction of the potential outcome r^* is particularly unsatisfactory:
 - (a) It lacks interpretation,
 - (b) it relies on distributional assumptions, and
 - (c) no determinant of selection is excluded from the potential outcome equation.

Using predicted R&D versus Instrumenting R&D

A) Linear Models

- Let $f(z) = E(x|z)$ and consider

$$\text{Model I: } y = bx + u \quad E(u|z) = 0$$

$$\text{Model II: } y = af(z) + v \quad E(v|z) = 0.$$

- Model I is the IV model: b measures the effect of x on y holding u constant in a situation where x and u are potentially correlated. Model II is a regression model that relates y to expected x .
- They have different economic interpretations: in Model I agents respond to x whereas in Model II they respond to $f(z)$. But replacing $x = f(z) + e$ in the first equation
$$a = b$$
$$v = b * e + u.$$
- The implication is that whether we use $f(z)$ as a regressor or as an instrument we are estimating the same coefficient.
- The fact that $a = b$ is a peculiarity of the linear case, which does not hold more generally, i.e. it would not hold for:

$$\text{Model I}' : y = g(x, b) + u \quad E(u|z) = 0$$

$$\text{Model II}' : y = g[f(z), a] + v \quad E(v|z) = 0$$

B) Binary Choice

- Suppose a probit case such that u and e are normal

$$y = 1 (bx + u > 0)$$
$$y = 1 (bf(z) + v > 0)$$

- Here derivative effects on the probability are different:

$$\frac{\partial}{\partial f(z)} \Pr(y = 1) = \frac{b}{(1 + b^2\sigma_e^2)^{1/2}} \phi \left[\frac{bf(z)}{(1 + b^2\sigma_e^2)^{1/2}} \right]$$
$$\frac{\partial}{\partial x} \Pr(y = 1) = b\phi(bx).$$

C) Predicted Latent R&D

- So far we dealt with predicted R&D. If we have a discrete choice or selection model, we could consider using
 - “predicted propensity to perform R&D”
 - or “predicted potential R&D intensity”as explanatory variables, where x^* denotes de propensity or potential outcome, and $f^*(z)$ its conditional mean.
- In those cases, the model that substitutes $f^*(z)$ could be equivalent to a hypothetical model that instruments x^* .
- But they are different economic models to those using x or $f(z)$ because they are modelling the response to different explanatory variables.

Comments about the results

- We would expect the effect of R&D on productivity to be roughly the product of the effect of R&D on innovation and the effect of innovation on productivity,
- but ONLY if demand pull and cost push variables have been included in the productivity equation;
- i.e. combining the equations for *prod* and *zinno* we obtain the semi-reduced form equation:

$$prod = (a_5 a_4) s^* + x_A (b_4 a_5) + (a_5 u_4 + u_5)$$

- If there is measurement error in sales, the product innovation share coefficient in the labor productivity equation may be downward biased.
- More generally, if *zinno* is a very noisy measure of product innovation, it may be easier to identify $(a_5 a_4)$ than a_5 .
- The importance of this is that for policy evaluation we might just be interested in the productivity of R&D.
- Understanding the production of innovations seems an interesting but separate concern.

Other comments about the econometrics

- From the policy evaluation perspective of modern labor econometrics some concerns are:

(a) *Functional form restrictions and common support*: In a matching exercise, assuming that s is exogenous given x , one would calculate an average return of R&D performance on productivity as:

$$\int [E(\text{prod} \mid p(x), s = 1) - E(\text{prod} \mid p(x), s = 0)] dG(p)$$

where $E(\text{prod} \mid p(x), s = j)$ is a non-parametric simple regression and $p(x)$ is a propensity score.

(b) *Instrument validity*: Why are quality improvement, cooperation, international market, and sources of information valid instruments for the innovation and productivity equations?

(c) *Heterogeneity of returns*: If returns of R&D performance are homogeneous we would expect the regression of productivity on the propensity score to be linear. Testing for nonlinearity can be regarded as a test of heterogeneity.