

Innovation and Employment: Evidence from Italian Microdata

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Technology and employment: theory

- At the macroeconomic level, concern about **negative employment impact of technology** is not new (Hobsbawm, 1968).
- The fear comes especially from the **labour-saving** effect of **process innovation** (Ricardo, 1951).
- Some “**compensation mechanisms**” can counterbalance the employment negative impact of technology (Marx, 1961; Vivarelli, 1995).

Technology and employment: empirical evidence (1)

- Empirical literature is developed at **three levels** depending on the **disaggregation of data** (macroeconomic, sectoral and firm level analysis) and using **different proxies** for **technology**.
- Focusing on **firm level analysis**, empirical evidence **cannot capture all** the sectoral and macroeconomic **effects of innovation** (*business stealing*), but fully captures the **direct** labour-saving effect of innovation.

Technology and employment: empirical evidence (2)

- **Previous empirical evidence** at the **firm level** shows a generalised **positive impact** of technology on employment both in *cross-section and panel data* analysis in Germany (Entorf-Pohlmeier, 1990; Smolny, 1998), UK (Blanchflower *et al.* 1991; Van Reenen, 1997), France (Greenan-Guellec, 1996), US (Doms *et al.*, 1997), Australia (Blanchflower-Burgess, 1998).
- **Negative employment impact** of technology just in the Netherlands (Brouwer *et al.*, 1993) and Norway (Klette-Førre, 1998).

Microeconomic analysis on Italian firm level data

- The aim of this analysis is to assess the **microeconomic employment impact of innovation** in **Italy** mainly characterised by **capital-embodied intermediate technologies**.
- Firm level data come from **Mediocredito Centrale**. A **balanced panel** dataset of **575 manufacturing firms** (with no less than 11 employees) covering the **1992-1997** period has been used.

The model (1)

- Starting from a **perfectly competitive firm** maximising profits under a **CES function**:

$$Y = A[(\alpha L)^\rho + (\beta K)^\rho]^{1/\rho}$$

- The **stochastic version of labour demand augmented by innovation** (*inn*) can be derived for a panel of firms (*i*) over time (*t*):

$$l_{i,t} = \beta_1 y_{i,t} + \beta_2 w_{i,t} + \beta_3 inn_{i,t} + (\varepsilon_i + v_{i,t})$$

where $i = 1, \dots, n$ and $t = 1, \dots, T$.

The model (2)

- In order to **introduce dynamic regressors (employment and innovation)** and to **avoid biased and inconsistent estimators**, the **first difference** specification is adopted:

$$\Delta l_{i,t} = \alpha \Delta l_{i,t-1} + \beta_1 \Delta y_{i,t} + \beta_2 \Delta w_{i,t} + \beta_{3,1} \Delta inn_{i,t} + \beta_{3,2} \Delta inn_{i,t-1} + \Delta v_{i,t}$$

- l = number of employees
- y = sales
- w = average wage per employee
- inn = value of innovative investments (peculiarities of Italian manufacturing)
- v = usual error term

Estimation method

- In order to overcome common problems concerning the **endogeneity of the lagged depend variable** (correlation $\Delta l_{i,t-1}$ and $\Delta v_{i,t}$) and **other potentially endogenous** variables, it is necessary to rely on **instrumental variables techniques**: ***GMM-DIF*** and ***GMM-SYS*** (Arellano-Bond, 1991; Blundell-Bond, 1998).
 - **GMM-SYS estimate** turns out to be the most **efficient** due to:
 - 1) **persistence** of the dependent variable
 - 2) $(\sigma_\varepsilon)^2 / (\sigma_v)^2$ **large** in short panels
 - 3) **Differenced Sargan** test

Table 1: Descriptive statistics: 575 Italian manufacturing firms (1992-1997)

| | All firms | | Innovators | | Occasional innovators | | Non innovators | |
|--|-----------|--------|------------|--------|-----------------------|--------|----------------|-------|
| | Mean | S.D. | Mean | S.D. | Mean | S.D. | Mean | S.D. |
| Output | 96196 | 339341 | 86381 | 144344 | 124147 | 528650 | 34010 | 60875 |
| Average output growth (1992-1997) | 5.12% | 10.21 | 4.54% | 8.23 | 6.42% | 12.87 | 3.12% | 8.35 |
| Employment | 271 | 550 | 291 | 567 | 270 | 567 | 132 | 254 |
| Average employment growth (1992-1997) | 2.34% | 8.62 | 2.25% | 8.54 | 2.67% | 9.44 | 1.41% | 3.89 |
| Real wage | 54.13 | 18.74 | 55.22 | 19.32 | 53.42 | 18.65 | 49.47 | 13.47 |
| Innovative investments | 2351 | 7755 | 2989 | 6628 | 1892 | 9762 | 0 | 0 |
| Number of firms | 575 | | 318 | | 212 | | 45 | |
| Observations | 3450 | | 1908 | | 1272 | | 270 | |

Table 2: Dependent variable: **employment**

| | (1) OLS | (2) WITHIN | (3) GMM-SYS |
|------------------------------------|---------------------|---------------------|---------------------|
| Employment (-1) | 0.93*** (0.005) | 0.52*** (0.016) | 0.86*** (0.040) |
| Sales | 0.06*** (0.004) | 0.14*** (0.009) | 0.13*** (0.031) |
| Wages | -0.12*** (0.009) | -0.35*** (0.016) | -0.20*** (0.034) |
| Innovative investments | 0.007*** (0.001) | 0.004*** (0.001) | 0.005** (0.002) |
| Innovative investments (-1) | -0.003** (0.001) | -0.002* (0.001) | -0.003 (0.002) |
| Constant | 0.11*** (0.035) | | 0.13 (0.147) |
| Time dummies | Yes | Yes | Yes |
| AR(1) | | | -5.76*** |
| AR(2) | | | 0.28 |
| Sargan test | | | 53.28 |
| Observations | 2875 | 2875 | 2875 |

Notes:

In brackets: White robust standard errors;*=10% significant, **=5% significant, ***= 1% significant. In column (3) lagged employment and sales are considered as endogenous, innovative investments as predetermined, and wages as exogenous.

AR(1) and AR(2) are tests - with distribution $N(0,1)$ - on the serial correlation of residuals.

The Sargan-test has a $\chi^2(43)$ distribution under the null of validity of the instruments.

Overall long-run employment-innovation elasticity turns out to be 0.0143.

Table 3: robustness checks; dependent variable: **employment**

| | (1) GMM-SYS | (2) GMM-SYS | (3) GMM-SYS | (4) GMM-SYS |
|--|---------------------|---------------------|---------------------|---------------------|
| Employment (-1) | 0.87*** (0.037) | 0.86*** (0.037) | 0.86*** (0.039) | 0.85*** (0.043) |
| Sales | 0.11** (0.027) | 0.12** (0.029) | 0.12*** (0.030) | 0.13*** (0.032) |
| Wages | -0.20*** (0.034) | -0.20*** (0.034) | -0.20*** (0.033) | -0.21*** (0.042) |
| Innovative investments | 0.005* (0.002) | 0.005* (0.002) | 0.005** (0.002) | 0.005** (0.002) |
| Innovative investments (-1) | -0.003 (0.002) | -0.003 (0.002) | -0.003 (0.002) | -0.003 (0.002) |
| Constant | 0.16 (0.159) | 0.14 (0.168) | 0.19 (0.15) | 0.16 (0.166) |
| Sectoral dummies (13 ATECO sectors) | Yes | | | |
| Sectoral dummies (21 ATECO sectors) | | Yes | | |
| Area dummies (4 macro-regions) | | | Yes | |
| Size dummies (5 classes) | | | | Yes |
| Time dummies | Yes | Yes | Yes | Yes |
| AR(1) | -5.67*** | -5.70*** | -5.75*** | -5.91*** |
| AR(2) | 0.32 | 0.32 | 0.29 | 0.25 |
| Sargan test | 54.59 | 55.23 | 55.89* | 54.55 |
| Observations | 2875 | 2875 | 2875 | 2875 |

Conclusions

- Using a panel dataset of 575 Italian manufacturing firms, the microeconomic analysis shows a **significant, although small in size, positive** relationship between innovation - measured through **innovative investments - and employment**.
- Innovative investments are not just a proxy of process innovation, but rather a **mark of innovativeness** (complementarity between process and product innovations).
- The job-creating impact of innovation proves **robust** after checking for time, industry, size of firm and geographical fixed effects.
- Results are **consistent with previous studies**, but **cannot be easily generalised**.