

Exploring the nexus between appropriability and productivity in highly innovative and globalised companies

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- ① Introduction and motivation
- ② Literature
- ③ Empirical model
- ④ Conclusion

- What** The paper investigate the **the delicate balance** between MNEs R&D activities, IPR protection and their impact on firm level **labour productivity**
- Where** We explore this relationship on data of **most R&D intensive MNEs in the world**, between 2007-2016 (COR&DIP database–JRC/OECD & ORBIS)
- How** We characterise the determinants of labour productivity as a function of tangibles, intangibles and **the institutional context**, whereby we explore potential non-linearities.
- Why** Lack of consensus in the literature on the role of appropriability in spurring innovation (Hall Sena 2017, Breschi Malerba Orsenigo 2000, Nelson Wolff 1997, Berge-Gill Lopez 2014, Gelabert Fosfuri Tribo 2009, Falk 2006, Cohen Walsh 2000, Cohen Nelson Walsh 2000, Ceccagnoli 2009)

The paper in a nutshell: contributions

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Role of Appropriability (IPR) at the firm level

Identification of the role of appropriability at the **level of the firm** on productivity

Bright side The results show that patent protection **could be beneficial** in terms of productivity if associated to R&D expenditure, supporting the dual role (direct and indirect) of R&D at the firm level (Cohen Levinthal 1989)

Dark side The results also show the existence of non-linearities in the impact of R&D and IPR on productivity: too much patent protection **ceases to be beneficial** in terms of productivity even if associated with R&D expenditures

Methodological contribution The paper applies a methodology proposed by Amoroso Bruno Magazzini 2022 to identify time-invariant variables (TIV) in panel FE

- Productivity and R&D (Griliches 1979)
- R&D → productivity (Hall and Mairesse 1995; Crépon Duguet Mairesse 1998)
- Formal and Informal IPR, investment in R&D (Hall Sena 2017; Hall Helmers Rogers Sena 2013; Hall Helmers Rogers 2014)
- Schumpeter Hypotheses (Levin Cohen Mowery 1985; Nelson Wolff 1997; Breschi Malerba Orsenigo 2000; Berge-Gil Lopez 2014; Lee 2005):
 - ① Market structure
 - ② Role of demand
 - ③ Technology opportunities
 - ④ R&D appropriability
- Degree of Appropriability and effect on public R&D support (Gelabert Fosfuri Tribo 2009)
- Appropriability and performance (Griliches Cockburn 1988; Ceccagnoli 2009; Hall Sena 2017)

- Dependent variable: Labour productivity (VA/emp.)
- Main independent variables:
 - ▷ R&D/emp (intangible)
 - ▷ Tangible capital (/emp)
 - ▷ Labour (emp)
- Moderation: Intellectual property rights as international patent systems strength
- Controls:
 - ▷ Unobserved heterogeneity: firms fixed effects
 - ▷ Time fixed effects
 - ▷ Sector trends (sector#time) Region trends (region#time) vs. use of country-industry deflators

Empirical analysis: determinants of productivity

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- Data: COR&DIP database (JRC/OECD reports) merged with ORBIS & ORBIS-IP
- Starting from a sample of the top 2000 worldwide R&D spenders over the period 2007-2016
 - ▷ Missing values → 1000-1200 firms in the analysis
- Firm productivity as a function of tangibles, intangibles & appropriability (in logs, following the specification of Hall Mairesse, 1995)

$$\ln\left(\frac{VA}{L}\right) = \beta_1 \ln\left(\frac{K}{L}\right) + \beta_2 \ln\left(\frac{RD}{L}\right) + \beta_3 \ln(L) + \beta_4 App + \beta_5 \ln\left(\frac{RD}{L}\right) \times App + \varepsilon$$

How to measure appropriability?

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- Measure of the IP-environment **faced by the firms**
- We rely on two measures of “appropriability” based on measurement of the international patent systems strength (country-level information)
 - EFW*: Economic Freedom Index, by World Economic Forum
Survey of business executives asked to rate the extent to which a country’s legal framework allows individuals to acquire, hold, and utilize private property and the extent to which these rights are secured by clear laws that the government enforces effectively
(1-7; 151 countries, 2007-2017)
 - PEI*: Patent Enforcement Index by Papageorgiadis Sofka
(2020)
Patent system strength as determined by laws and enforcement – degree to which patents confer rights to their holders, but also how effectively and efficiently they can be enforced
(0-10; 51 countries, 1998-2017)
- In our sample, the correlation between the two measures is 0.95

How to measure appropriability?

The IP-environment faced by the firm

- To build a measure of appropriability at the firm level, PEI and EFW are weighted on the basis of firm's patenting across countries (Bruno et. al. 2022)

$$App_{it} = \sum_c \frac{\#Pt_{ict}}{\#Pt_{it}} App_{ct} = \sum_c s_{ict} App_{ct}$$

- ▷ App_{ct} is, respectively, EFW and PEI , in country c at time t
- ▷ $\#Pt_{ict}$: number of patent of firm i at time t granted in country c
- ▷ $\#Pt_{it}$: total number of patent of firm i at time t
- App_{it} : weighted average of the IPR strength in the countries in which firm i chooses to protect its rights at time t
 - ▷ Endogeneity concerns – to be discussed later
- At the firm level, correlation equals 0.78

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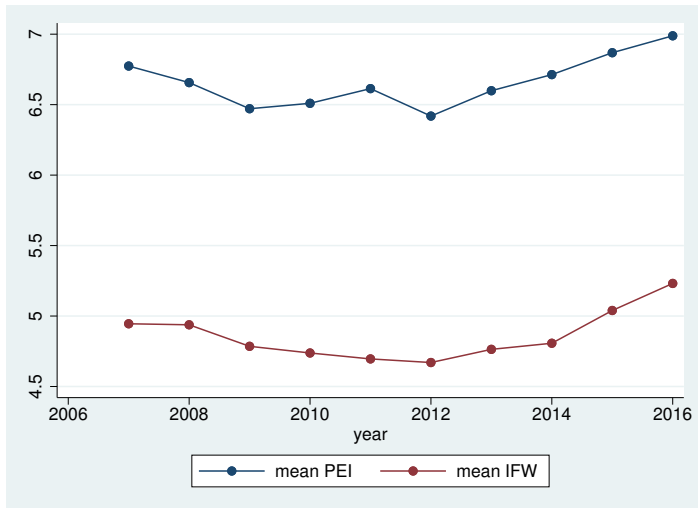
Evolution of patent protection at the firm level

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Variable	orig.	imput.	defl.	defl.
$\ln(K/L)$.533*** (.030)	.529*** (.030)	.572*** (.032)	.572*** (.032)
$\ln(RD/L)$.527*** (.160)	.488*** (.157)	.315* (.170)	.382* (.161)
$\ln(L)$	-.058 (.103)	-.088 (.099)	-.122 (.098)	
App_{it}	.362 (.247)	.335 (.243)	.233 (.171)	.220 (.170)
$\ln(RD/L)$ $\times App$	-.029 (.022)	-.027 (.021)	-.031 (.024)	-.030 (.024)
N	7058	7446	5846	5846

Note: time FE included in all specifications.

Orig. & imputed also includes sector \times year, region \times year.

Variable	orig.	imput.	defl.	defl.
$\ln(K/L)$.530*** (.030)	.527*** (.030)	.572*** (.032)	.572*** (.032)
$\ln(RD/L)$.567*** (.183)	.560*** (.176)	.507*** (.191)	.577*** (.184)
$\ln(L)$	-.051 (.103)	-.084 (.099)	-.140 (.097)	
App_{it}	.627* (.378)	.660* (.365)	.645** (.259)	.616** (.260)
$\ln(RD/L)$ $\times App$	-.044 (.033)	-.048 (.032)	.083** (.037)	-.080** (.037)
N	7044	7431	5845	5845

Note: time FE included in all specifications.

Orig. & imputed also includes sector \times year, region \times year.

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$$\begin{aligned}\ln\left(\frac{VA}{L}\right) &= \beta_1 \ln\left(\frac{K}{L}\right) + \beta_2 \ln\left(\frac{RD}{L}\right) + \beta_3 \ln(L) \\ &+ \beta_4 App + \beta_5 App \times \ln\left(\frac{RD}{L}\right) \\ &+ \beta_6 App^2 + \beta_7 App^2 \times \ln\left(\frac{RD}{L}\right) + \varepsilon\end{aligned}$$

Exploring non-linearities

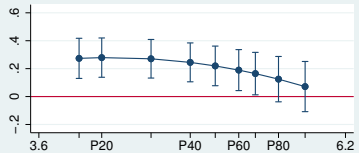
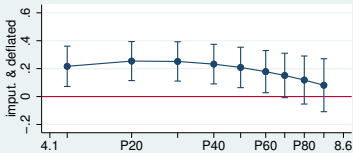
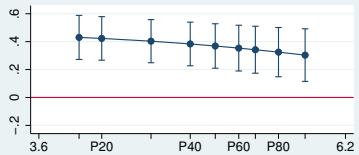
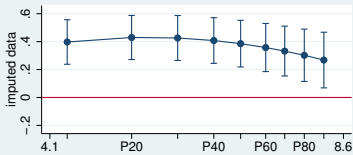
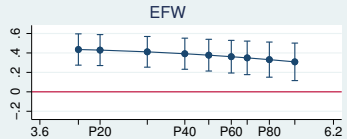
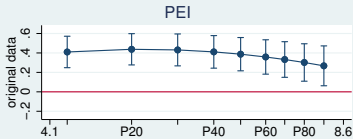
Polynomial modeling: APE of $\ln(RD)$ as a function of App

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- On the basis of previous results, we selected three “levels”: (i) below the median; (ii) between 2nd and 3rd quartile (median to P75); (iii) above 3rd quartile

$$\begin{aligned}\ln\left(\frac{VA}{L}\right) &= \beta_1 \ln\left(\frac{K}{L}\right) + \beta_2 \ln\left(\frac{RD}{L}\right) + \beta_3 \ln(L) \\ &+ \beta_4 A_{(Q2.Q3)} + \beta_5 A_{(Q2.Q3)} \times \ln\left(\frac{RD}{L}\right) \\ &+ \beta_6 A_{(Q3+)} + \beta_7 A_{(Q3+)} \times \ln\left(\frac{RD}{L}\right) + \varepsilon\end{aligned}$$

Exploring non-linearities

Dummy for selected thresholds – PEI

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Variable	orig.	imputed	imp. & defl	imp. & defl
$\ln(K/L)$.532*** (.030)	.528*** (.030)	.571*** (.032)	.571*** (.032)
$\ln(RD/L)$.390*** (.111)	.364*** (.108)	.164* (.088)	.236*** (.069)
$\ln(L)$	-.045 (.103)	-.075 (.099)	-.114 (.098)	
$A_{(Q2.Q3)}$.773 (.584)	.682 (.563)	.685* (.363)	.668* (.364)
$A_{(Q3+)}$	1.61** (.689)	.579** (.683)	.820* (.463)	.821* (.460)
$A_{(Q2.Q3)} \times \ln(RD/L)$	-.063 (.052)	-.054 (.050)	-.091* (.053)	-.089* (.054)
$A_{(Q3+)} \times \ln(RD/L)$	-.136** (.061)	-.133** (.060)	-.112* (.066)	-.113* (.066)
N	7058	7446	5846	5846

Exploring non-linearities

Dummy for selected thresholds – EFW

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Variable	orig.	imputed	imp. & defl	imp. & defl
$\ln(K/L)$.529*** (.030)	.527*** (.030)	.5703*** (.032)	.571*** (.032)
$\ln(RD/L)$.400*** (.111)	.377*** (.109)	.176** (.089)	.259*** (.070)
$\ln(L)$	-.056 (.101)	-.088 (.098)	-.131 (.097)	
$A_{(Q2.Q3)}$.758 (.524)	.721 (.507)	.635** (.302)	.626** (.301)
$A_{(Q3+)}$	1.88*** (.659)	1.88*** (.641)	1.31*** (.414)	1.30*** (.413)
$A_{(Q2.Q3)} \times \ln(RD/L)$	-.057 (.046)	-.054 (.045)	-.090** (.044)	-.088** (.044)
$A_{(Q3+)} \times \ln(RD/L)$	-.146** (.058)	-.148*** (.056)	-.168*** (.059)	-.167*** (.059)
N	7044	7431	5845	5845

- Firms *choose* the countries in which to protect their innovation
 - Endogeneity concerns related to the measure of appropriability
 - Two strategies
1. Build an exogenous measure of appropriability

- ▷ Build a model for firm's patenting at the country level:

$$Pt_{ict} = f(\ln(RD)_{it}, \ln(\bar{RD})_i, \ln(GDP)_{ct}, \ln(GDP)_{it}, dist_{ic}, \alpha_c, D(cty_i), \tau_t)$$

- ▷ Use fitted number of patents to build the weights:

$$\hat{A}_{it} = \sum_c \frac{\# \hat{P}t_{ict}}{\# \hat{P}t_{it}} App_{ct}$$

2. Exploit pre-sample information available for PEI

Robustness checks

Exogenous measure of appropriability – PEI

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Variable	orig.	imputed	imp. & defl	imp. & defl
$\ln(K/L)$.514*** (.031)	.513*** (.031)	.573*** (.032)	.573*** (.032)
$\ln(RD/L)$.386*** (.116)	.356*** (.112)	.148* (.089)	.230*** (.072)
$\ln(L)$	-.059 (.105)	-.091 (.100)	-.129 (.098)	
$\hat{A}_{Q2.Q3}$.566 (.384)	.533 (.383)	.740*** (.237)	.727*** (.238)
\hat{A}_{Q3+}	1.07 (1.03)	.873 (.860)	.145 (.539)	.116 (.544)
$\hat{A}_{Q2.Q3} \times \ln(RD/L)$	-.069** (.034)	-.066* (.034)	-.109*** (.036)	.108*** (.036)
$\hat{A}_{Q3+} \times \ln(RD/L)$	-.103 (.093)	-.084 (.077)	-.011 (.078)	-.009 (.079)
N	6788	7171	5839	5839

Robustness checks

Exogenous measure of appropriability – EFW

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Variable	orig.	imputed	imp. & defl	imp. & defl
$\ln(K/L)$.515*** (.031)	.513*** (.031)	.569*** (.032)	.569*** (.032)
$\ln(RD/L)$.389*** (.116)	.360*** (.112)	.126 (.087)	.205*** (.070)
$\ln(L)$	-.055 (.105)	-.087 (.101)	-.123 (.099)	
$\hat{A}_{Q2.Q3}$.299 (.401)	.288 (.384)	-.067 (.224)	-.078 (.225)
\hat{A}_{Q3+}	.456 (.540)	.425 (.520)	.855*** (.285)	.847*** (.286)
$\hat{A}_{Q2.Q3} \times \ln(RD/L)$	-.052 (.034)	-.051 (.032)	.005 (.032)	.007 (.032)
$\hat{A}_{Q3+} \times \ln(RD/L)$	-.070 (.043)	-.067 (.041)	-.108*** (.039)	-.108*** (.039)
N	6788	7171	5839	5839

Robustness checks: pre-sample value (PEI only)

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- To mitigate endogeneity concerns, we exploit the availability of pre-sample information on PEI (average 2000-2006)
- Interaction term is identified in a FE framework, however the effect of time-invariant variables cannot be identified
 - ▷ Unless additional assumptions are imposed (e.g., Hausman Taylor 1981, Pesaran Zhou 2018)
- We exploit the framework in Breusch Mizon Schmidt 1989 to also identify the effect of appropriability
 - ▷ Need to find variable(s) with *homogenous* correlation with individual component

Identification of time-invariant variables

(Amoroso Bruno Magazzini 2022)

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$$y_{it} = \mathbf{x}'_{it}\beta + \mathbf{z}'_i\gamma + \varepsilon_{it} = \mathbf{x}'_{it}\beta + \mathbf{z}'_i\gamma + \tau_t + \alpha_i + e_{it}$$

- FE as IV with $\mathbf{x}_{it} - \bar{\mathbf{x}}_i$ as instruments (γ not identified)
- Seminal work by Hausman Taylor 1981

$$\mathbf{x}_{it} = (\mathbf{x}_{1it}, \mathbf{x}_{2it}) \quad \& \quad \mathbf{z}_i = (\mathbf{z}_{1i}, \mathbf{z}_{2i})$$

- Need to make assumption: \mathbf{x}_{1it} and \mathbf{z}_{1i} uncorrelated with α_i
- IV estimation with instruments $\mathbf{x}_{1it} - \bar{\mathbf{x}}_{1i}$, $\mathbf{x}_{2it} - \bar{\mathbf{x}}_{2i}$, \mathbf{x}_{1it} and \mathbf{z}_{1i}
- Set of instruments extended by Amemiya MaCurdy 1986 and Breusch Mizon Schmidt 1989

Identification of time-invariant variables

(Amoroso Bruno Magazzini 2022)

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- Breusch Mizon Schmidt 1989: assuming, for all i , $E(\mathbf{x}_{2it}\alpha_i)$ is the same for all t , the set of instrumental variables can be extended to include

$$\mathbf{x}_{2i1} - \bar{\mathbf{x}}_{2i}, \mathbf{x}_{2i2} - \bar{\mathbf{x}}_{2i}, \dots, \mathbf{x}_{2iT} - \bar{\mathbf{x}}_{2i}$$

- When only “2” vars. are available – Estimate β by:

$$E[(x_{it} - \bar{x}_i)(y_{it} - \beta x_{it} - \gamma z_i)] = 0$$

(γ not identified)

- Exploit the additional “BMS” conditions to identify γ

$$E[(x_{i2} - \bar{x}_i)(y_{it} - \beta x_{it} - \gamma z_i)] = 0$$

...

$$E[(x_{iT} - \bar{x}_i)(y_{it} - \beta x_{it} - \gamma z_i)] = 0$$

Robustness checks: pre-sample value (PEI only)

Results – imp. & defl. data (Amoroso Bruno Magazzini 2022)

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Variable	2SLS	LIML	2SLS	LIML
$\ln(K/L)$.576*** (.032)	.576*** (.032)	.577*** (.032)	.577*** (.032)
$\ln(RD/L)$.569** (.229)	.597** (.240)	-.014 (.090)	-.030 (.104)
$\ln(L)$	-.088 (.280)	-.097 (.310)	-.117 (.135)	-.139 (.161)
$\bar{A}_{00.06}$.602*** (.183)	.630*** (.191)		
$\bar{A}_{00.06} \times \ln(RD/L)$	-.090*** (0.034)	-.096*** (.036)		
$\bar{A}_{Q2.Q3}^{00.06}$.376 (.885)	.340 (.985)
$\bar{A}_{Q3+}^{00.06}$			1.08 (1.02)	1.29 (1.21)
$\bar{A}_{Q2.Q3}^{00.06} \times \ln(RD/L)$			-.039 (.153)	-.036 (.173)
$\bar{A}_{Q3+}^{00.06} \times \ln(RD/L)$			-.170 (.167)	-.210 (.201)
N	5848	5848	5848	5848
Hansen- j (p)	.173	.185	.355	.380
KP F -stat.	>1,000	>1,000	47.3	47.3

- We have identified a complex empirical relationship between R&D and productivity, especially when looking into the role of IPR
 - ▷ Patent protection **could be beneficial** in terms of productivity if associated with R&D expenditure, supporting the dual role (direct and indirect) of R&D at the firm level (Cohen Levinthal 1989)
 - ▷ However, non-linearities exist in the relationship between R&D, IPR and productivity: too much patent protection **ceases to be beneficial** in terms of productivity even if associated with R&D expenditures
- Further work
 - ▷ Better identification of the channels through which IPR “operates”
 - ▷ Better identification of the channels through which IPR interacts with R&D

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Thank you

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