Patents that Match your Standards: Firm-level Evidence on Competition and Innovation

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[The views expressed herein are ours and should not be attributed to BdF]

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- The ability to adapt to the new technology frontier is heterogeneous across firms and depends on past innovation choices
- This has implications for market structure, competition and (future) innovation

Example: ITU Standard 2015, IMT-2020 5G protocol

- Date of release: 9 June 2015
- Key minimum specifications:
 - ▶ bandwidth to be at least 100 MHz
 - ▶ bandwidths up to 1 GHz are required for higher frequencies (above 6 GHz)
 - ▶ connection density is 1 million devices per square kilometer
 - ▶ downlink peak data rate of 20 Gb/s
 - ▶ uplink peak data rate of 10 Gb/s
 - ▶ target downlink user experienced data rate of 100 Mb/s
 - ▶ target uplink user experienced data rate of 50 Mb/s
- Which firms were ready to adapt and comply to the new standard?

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 - 3. how to measure the proximity of the firm technology to the new frontier
- We use:
 - 1. data on **published standard documents** \leftarrow the new tech. frontier
 - 2. data on **firm-level patents** \leftarrow firms past innovation outcomes
 - 3. textual analysis to measure the semantic proximity of the firm' stock of patents to the newly published standards [!NOVELTY!] ← proximity to the frontier
- We cross this new measure with firm-level data (Compustat, Crisp, IBES)

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- We exploit the exogeneity of the timing and content of the standard to show that frontier firms
 - 1. temporarily gain in terms of sales and market-shares
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3. Industry-level results

- Growth is higher in industries with initial larger gap between leaders and laggards
- Growth is driven by frontier firms in the **short-run**, but by laggards in the **long-run**

Literature

1. The pros and cons of standardization (IO literature):

Katz and Shapiro 1985; Farrell and Saloner 1985; Leland 1979; Rysman and Simcoe 2008; Lerner and Tirole 2015; Schmalensee 2009; Llanes and Poblete 2014; Spulber 2019; Bekkers et al. 2017; Baron and Pohlmann 2018

- 2. Innovation literature:
 - Real Effects: Arrow 1962; Romer 1990; Aghion and Howitt, 1992; Grossman and Helpman 1991; Aghion et al. 2005; Aghion and Griffith 2005
 - Financial Effects: Kogan et al. 2017, Pakes 1985, Nicholas 2008; Daniel et al. 1998; Mitchell and Stafford 2000
- Text mining applied to semantic analysis of patents and standards Arts et al. 2018; Kuhn et al. 2020; Kelly et al. 2021; Argente et al. 2020; Bergeaud et al. 2017; Webb et al. 2018; Bloom et al. 2021; Brachtendorf et al. 2020

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- Score(p, s) measures the proximity of patent p to standard s

Mean SDр1 *p*5 *p*25 *p*50 *p*75 *p*95 p99 Ν (A) Keyword matching sample Score 715.7 1.766.6 141.3 151.8 211.8 315.2 638.7 2.345.7 6.289.0 100M

1.2 Firm-level Data

- CRISP for stock market abnormal returns
- IBES for market expectation on future EPS
- COMPUSTAT for firms fundamentals: sales, market shares (defined at NAICS 3-digit level), R&D, CapX, Tobin's Q, market capitalization, leverage, age, industry markup
- Aggregation of **patent-to-standard scores** at the firm level:

$$Shock_{i,t} = \sum_{s \in S} \sum_{p \in P} Score(p, s)_{i,t}$$

	Mean	SD	p1	<i>p</i> 5	p25	<i>p</i> 50	<i>p</i> 75	<i>p</i> 95	p99	N
(A) Standard	ization	Shock								
Shock	0.34	2.02	0.00	0.00	0.00	0.00	0.10	1.27	6.24	24,162
$\mathbb{I}[Shock > 0]$	0.48	0.49	0.00	0.00	0.00	0.00	1.00	1.00	1.00	24,162

2. Results

2.1 Patent-level Results

Is the score economically meaningful?

- For scores based on patents filed 1 year before the publication of each standard, consider this:

 $\log (value_i) = c + \alpha \log (1 + score_i) + \beta \log (1 + cit_i) + \gamma Z_i + \varepsilon_i$

where *value*; is the economic value of the patent from Kogan et al. 2017

	(1)	(2)	(3)	(4)	(5)	(6)
Score	0.0088***	0.0062***	0.0064***	0.0051***	0.0062***	0.0050***
	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]
Observations	1,165,487	1,165,487	1,165,462	1,165,462	1,163,913	1,163,913
R^2	0.05	0.06	0.10	0.11	0.13	0.14
Time	Yes	Yes	Yes	Yes	No	No
IPC	No	No	Yes	Yes	No	No
$Time \times IPC$	No	No	No	No	Yes	Yes

- Since the mean patent is valued at 16.3 mio USD, patents with a positive score raise their value by 310,000-592,000 USD.

2.2 Firm-level Results

Is the release of new standard exogenous to the firm? What are its implications for competition and innovation?

- For firms in the COMPUSTAT universe, consider the following:

$$Y_{i,t} = \alpha_i + \phi_{s,t} + \sum_{n=-12}^{N=16} \beta_n Shock_{i,t+n} + X'_{i,t-1}\eta + \varepsilon_{i,t}$$

where

- α_i is the firm FE
- $\phi_{s,t}$ is the Naics 3-digit FE interacted with a time FE
- ▶ $X_{i,t-1}$ controls for mkt cap, leverage, Q, dummy for tech. industries, age
- The distributed lead-lag model takes into account that different standards can be released in subsequent periods such that firms can receive multiple shocks throughout time

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- **Market Efficiency Hypothesis:** if the timing and content of the standard are unexpected, financial markets should react at the moment of the information release

- Consider the following dependent variables:
 - ▶ the abnormal return over a NAICS3-industry portfolio, i.e. $ar_{i,t}^{NAICS3}$
 - ► the change in the 1-year EPS forecast from professional agencies, i.e. $\Delta \mathbb{E}[EPS_{i,t+4}] = \mathbb{E}[EPS_{i,t+4}|\mathbb{I}_t] \mathbb{E}[EPS_{i,t+4}|\mathbb{I}_{t-1}]$



Figure: STANDARDIZATION SHOCK AND FINANCIAL MARKETS' REACTION

Sales & Market Shares

- Consider sales and market-shares (defined at the Naics 3-digit level) as dependent variables

Figure: STANDARDIZATION SHOCK, SALES AND MARKET SHARES



Is it a competition shock?

Figure: Aghion et al. 2005, Innovation vs. Competition



Competition

Is it a competition shock?

- Split Naics 3-digit industries in competitive and non-competitive according to their historical markup (De Loecker et al. 2020) and look at results for R&D

Figure: STANDARDIZATION SHOCK AND R&D



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Robustness Checks

- These results are robust
 - ▶ to different clustering methods:
 - firm-level clustering
 - industry-time double clustering
 - ▶ to sample selection:
 - results are not driven by frontier firms
 - results are not driven by listed firms only
 - when considering the intensive/extensive margin of the variable $Shock_{i,t}$
 - ▶ to different definitions of the variable *Shock*_{*i*,*t*}, i.e. they do not depend on our text analysis methodology

2.3 Industry-level Results

What are the sectoral implications of technology adoption in terms of growth? Is growth driven by leaders or followers in the long-run?

- Define as leaders (followers) those with $Shock_{i,t} > 0$ ($Shock_{i,t} = 0$)
- Aggregate variables at industry level and use same econometric model

	Industry	Leaders	Followers
Mean Sectoral Growth Rate (%)	1.64	1.04	0.60
(1yr-Cumulative) Change in Growth (pp)	-0.03	0.08	-0.11
(4yr-Cumulative) Change in Growth (pp)	0.11	0.02	0.09

- Consistently with the step-by-step model of Aghion et al. (1997,2001)
 - ► Short-run effect: leaders try to escape competition by increasing R&D
 - ► Long-run/Catching-up effect: followers adapt, increase research effort and surpass leaders

Followers explain sectoral growth in the long-run

- Consider the followers' share of industry aggregate sales and R&D expenditure

Figure: SECTORAL SALES AND R&D OF FOLLOWERS



Conclusion

- This paper studies the implications of large-scale technology adoption at the firm level
- To do so, we introduce a new measure of proximity of firms to the technological frontier
- This measure is based on text analysis of patents and standard documents and expresses how much the content of the patents of a firm overlaps with the content of a new standard
- We find that when a new standard is released, firms closer to the new frontier
 - ▶ gain in sales and market shares
 - ▶ increase R&D expenditure, if they operate in a competitive industry
- We can interpret the release of a new standard as a competition shock in favor of frontier firms
- Effects are only temporary. Laggards catch up through higher innovation efforts. This explains sectoral growth in the long-run

Thank you!

Appendix

Linking patents to standards

- Basic idea: look at keywords in the abstract of the patent and in the standard document
- Establish a list of all patent keywords with respective list of patent ids

keyword	patent id		
air condit	41		
aluminium	41,106		
beef	2		
electric	7,84,41,91,102		
screw	3,9,17,38		

Stablish a list of all standard keywords with respective list of standard ids

keyword	standard id		
air condit	B,M		
aluminium	В		
diameter	С		
rubber	A,E,G,H,R,Z		

Linking patents to standards

Find common keywords

keyword	patent id	standard id	idf
aluminium	41,106	В	0.2
air condit	41,65	B,M	0.5

• For each keyword: Cartesian product of respective patent and standard ids

keyword	match	idf
aluminium	41-B	0.2
aluminium	106-B	0.2
air condit	41-B	0.5
air condit	41-M	0.5

• Calculate score by adding up combinations of ids:

${\tt match}$	idf
41-B	0.7
41-M	0.5
106-B	0.2

Building up the score Back

- Evaluation of importance of each keyword k via calculation of inverse document frequency (IDF) across all patents (N):

$$\mathit{IDF}\left(k
ight) = 1 + \log\left(rac{1+\mathit{N}}{1+\mathit{N}(k)}
ight)$$

- For every patent $i \in \mathcal{P}$ and standard $j \in \mathcal{J}$, we extract
 - the set of k-grams $\mathcal{B}(i)$ for patent i
 - the set of k-grams $\mathcal{A}(j)$ for standard j
 - ▶ n(k, i), i.e. the number of times k-gram k appears in $\mathcal{B}(i)$
 - s(k), the length of k-gram k

- The final score from the matching of patent i to standard j is defined as

$$Score(i,j) = \sum_{k \in \mathcal{A}(j)} \sqrt{\left(\frac{n(k,i)}{|\mathcal{B}(i)|}\right)^{s(k)}} IDF(k) \left(|\mathcal{A}(j) \cap \mathcal{B}(i)|\right)$$