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Patent Commercialization

George Abi Younes
PhD Candidate
Chair of Innovation and IP Policy

Ecole Polytechnique Federale de Lausanne

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¹Shlomo Kalish and Gary L Lilien. "A market entry timing model for new technologies". In: *Management Science* 32.2 (1986), pp. 194–205.

²Glen L Urban et al. "Market share rewards to pioneering brands: An empirical analysis and strategic implications". In: *Management Science* 32.6 (1986), pp. 645–659.

 Companies need to strike a good compromise between commercializing too early and too late.¹

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- An early commercialization may risk pushing an underdeveloped product into the market.

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- Companies need to strike a good compromise between commercializing too early and too late.¹
- An early commercialization may risk pushing an underdeveloped product into the market.
- A late commercialization risks losing market share to other competitors who commercialize earlier.²

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 Many factors come into play when it comes to deciding market entry timing of a new product.³

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- Many factors come into play when it comes to deciding market entry timing of a new product.³
- This decision is even more complicated for patented inventions and technologies because of the extra factors and constraints that come into play.

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- This decision is even more complicated for patented inventions and technologies because of the extra factors and constraints that come into play.
- One factor that we believe potentially affects market entry timing of patented inventions is the **timing of patent grants**.

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- Many factors come into play when it comes to deciding market entry timing of a new product.³
- This decision is even more complicated for patented inventions and technologies because of the extra factors and constraints that come into play.
- One factor that we believe potentially affects market entry timing of patented inventions is the **timing of patent grants**.
- ⇒ Research Question: Are grant delays by patent offices slowing down product commercialization, and if so, to what extent?

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Several important milestones exist during patent commercialization

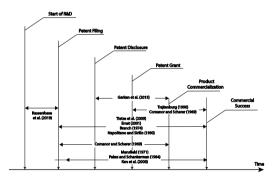


Figure: Important milestones in the process of patent commercialization

In our study, we investigate the relationship between the two following time variables:

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- Patent grant lag: duration between patent filing and patent grant.
- Product commercialization lag: duration between patent filing product market release.

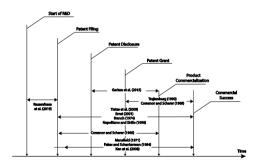


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• Not true. It is not always the case that companies commercialize after patent grants.

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- Even though it's rare, U.S. companies are allowed to commercialize their products up to 1 year prior to filing for the patent, which is also known as a grace period.⁴

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- Not true. It is not always the case that companies commercialize after patent grants.
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- In our data sample, we find that around 17% of the products have been commercialized prior to the grant date of its patent.

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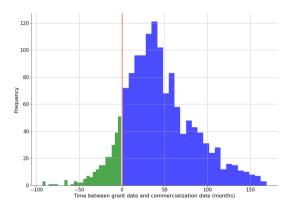


Figure: Duration between grant date and commercialization date

In building the dataset, we collect and supplement our data from various sources

• The building block of our data set is a product-patent pair, which we sample from IPRoduct⁵, which links patents to their associated products using VPM⁶ pages.

⁵https://iproduct.io/app/

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- We collect product-based measures (e.g. release date, ratings ...) by webscraping Amazon marketplace.

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- The database contains around 30000 patents from more than 700 companies and is continuously growing.

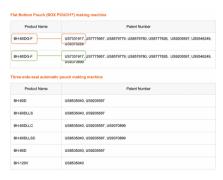


Figure: An example of a VPM page

Table: Summary Statistics

Variable	Definition	Mean	Std. dev.	Min	Max
Dependent Variable					
Commercialization Lag	Duration between commercialization date and patent application date (in months)	69.22	43.88	2	215
Patent Application Date	Date of patent application	-	-	_	-
Patent Grant Date	Date of patent grant by the USPTO		-	-	-
Product Commercialization Date	Extracted release date of product	-	-	-	-
Independent Variables					
Patent grant lag	Duration between grant date and patent application date (in months)	39.63	21.65	3	143
Control Variables					
Backward Citations	No. of backward citations referenced in the patent	45.62	39.08	1	179
Forward citations	Number of forward citations	5.86	6.9	0.00	50
Family size	Docdb family size of patent	102.44	378.47	1.00	5876.00
Geographic family size	Geographic family size of patent	5.81	5.14	1.00	24
$No.\ applicants$	Number of applicants for the patent	1.1	0.13	1.00	5
No. inventors	Number of inventors of the patent	2.90	2.4	1.00	29
No. claims	No. of claims made in the patent document	20.89	14.85	1.00	219
No. ind. claims	Number of independent claims	2.93	1.92	1.00	19
No. NPL citations	Number of non-patent literature citations	13.01	22.47	0	29
Patent originality	Patent originality index	0.59	0.23	0	0.95
No. Rejections	Number of rejections before patent grant	1.96	1.68	0	14
Process	Dummy variable indicating process patent	0.34	0.47	0	1
Company size	Categorical variable of company size	-	-	-	-
Company sector	Dummy variable indicating the sector of the company	-	-	-	-
$Product\ Age$	Number of days since the product release date	2103.91	1050.23	351.00	5667.00
$Product\ score$	Product ratings score on Amazon	3.38	1.59	0	5
No. Product ratings	Number of product ratings on Amazon	1041.66	3384.5	0	35394
Examiner Speed	Calculated speed of examiner	114.69	39.79	18.56	308.76

Empirical Setup

The two main variables of interest in our study are:

- GrantLag (GL) = PatentGrant PatentFiling
- $\bullet \ \textit{CommercializationLag} \ (\textit{CL}) = \textit{ProductCommercialization} \textit{PatentFiling} \\$

and our baseline model is:

$$CLi = GL_i + x_i \cdot \beta + s_i \cdot \theta$$

where x_i is the set of patent-based control variables and s_i the set of product-based control variables.

Unfortunately, the baseline model suffers from a potential endogeneity issue

- A patent applicant can purposefully delay the examination process by taking more time to respond to the patent office.
- By delaying the examination process, the applicant gets more time to add more features to the patent, thus making it stronger upon granting.
- **Solution**: We instrument our model using an IV⁷.

⁷Instrumental variable

The instrumental variable we resort to is examiner speed, which we calculate as follows:

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• For each patent examiner in our data set, we calculate the average grant lag divided by number of claims for patents he or she examined within a four-year window from the granting of the focal patent.

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- For each patent examiner in our data set, we calculate the average grant lag divided by number of claims for patents he or she examined within a four-year window from the granting of the focal patent.
- This gives us a notion of how fast, on average, the examiner goes through the examination process while taking into account the size of the patent document.

More explicitly...

examiner_speed =
$$\sum_{i=1}^{N} \frac{GL_{p_i}}{nb_claims_{p_i}}$$

where $p_i...p_N$ are the set of patents that have been examined by the focal examiner in a four-year window, taking the grant date of the focal patent as reference.

Empirical Results: Baseline Model

Table: Baseline model results

	(1)		(2)		(3)		(4)		(5)	
log_grant_lag_m	0.506***	(19.61)	0.490***	(15.60)	0.490***	(15.60)	0.495***	(12.29)	0.341***	(6.62)
nb_bcites			-0.000687	(-1.28)	-0.000687	(-1.28)	-0.00193**	(-2.85)	-0.00301***	(-3.79)
nb_fcites			0.0147***	(6.68)	0.0147***	(6.68)	0.0172***	(6.77)	0.0115***	(3.40)
nb_npl			-0.00468***	(-5.22)	-0.00468***	(-5.22)	-0.00296*	(-2.00)	0.00233	(1.34)
nb_ipc			0.00428	(0.89)	0.00428	(0.89)	-0.00277	(-0.48)	0.00233	(0.35)
nb_assignee			0.0142	(0.13)	0.0142	(0.13)	0.00275	(0.02)	0.0377	(0.22)
nb_inventor			-0.0108	(-1.61)	-0.0108	(-1.61)	-0.0114	(-1.54)	-0.0273**	(-2.91
nb_claims			-0.000924	(-0.69)	-0.000924	(-0.69)	-0.00108	(-0.58)	-0.00444*	(-2.00
nb_independent_claims			0.0165	(1.94)	0.0165	(1.94)	0.0378***	(3.32)	0.0511***	(3.40)
family size			0.000167**	(2.79)	0.000167**	(2.79)	0.000200**	(3.10)	0.000192*	(2.34)
geog family size			0.0186***	(3.89)	0.0186***	(3.89)	0.0284***	(4.59)	0.0211***	(3.61)
nb_rejections			-0.00935	(-0.87)	-0.00935	(-0.87)	-0.00485	(-0.38)	0.00996	(0.59)
process			-0.0651	(-1.78)	-0.0651	(-1.78)	0.0187	(0.40)	0.0151	(0.25)
originality_flex			0.0387	(0.48)	0.0387	(0.48)	0.0187	(0.20)	-0.101	(-0.97
young					0	(.)	0	(.)	0.0147	(0.12)
large					0	(.)	0	(.)	0.129	(1.84)
ratings							0	(.)	0.0000137*	(2.46)
score_1							0	(.)	0.0616	(1.43
nb_tms							0	(.)	7.41e-09	(1.48
Constant	2.405***	(26.24)	2.326***	(14.50)	2.326***	(14.50)	2.153***	(11.68)	2.479***	(7.55)
Observations	1374		1370		1370		992	. ,	1076	

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

Empirical Results: 2SLS

Table: Baseline model

	(1)		(2)		(3)		(4)		(5)	
log_grant_lag_m	0.806***	(17.55)	0.999***	(12.90)	0.999***	(12.90)	1.129***	(10.88)	0.935***	(7.70)
nb_bcites			-0.0000479	(-0.08)	-0.0000479	(-0.08)	-0.000816	(-1.04)	-0.00222**	(-2.60)
nb_fcites			0.00851***	(3.30)	0.00851***	(3.30)	0.0115***	(3.80)	0.00696	(1.90)
nb_npl			-0.00375***	(-3.76)	-0.00375***	(-3.76)	-0.00440**	(-2.59)	0.00108	(0.58)
nb_ipc			0.00333	(0.63)	0.00333	(0.63)	-0.00179	(-0.27)	0.00286	(0.40)
nb_assignee			-0.0449	(-0.36)	-0.0449	(-0.36)	-0.0477	(-0.37)	-0.00384	(-0.02)
nb_inventor			-0.0140	(-1.88)	-0.0140	(-1.88)	-0.0138	(-1.64)	-0.0270**	(-2.70)
nb_claims			-0.00171	(-1.15)	-0.00171	(-1.15)	-0.00228	(-1.07)	-0.00451	(-1.90)
nb_independent_claims			0.0171	(1.81)	0.0171	(1.81)	0.0382**	(2.94)	0.0539***	(3.37)
family_size			0.0000939	(1.41)	0.0000939	(1.41)	0.000154*	(2.10)	0.000190*	(2.19)
geog family size			0.0242***	(4.55)	0.0242***	(4.55)	0.0300***	(4.27)	0.0205***	(3.31)
nb_rejections			-0.111***	(-6.08)	-0.111***	(-6.08)	-0.122***	(-5.38)	-0.106***	(-3.79)
process			-0.0685	(-1.70)	-0.0685	(-1.70)	0.00344	(0.07)	0.0238	(0.37)
originality_flex			-0.0202	(-0.23)	-0.0202	(-0.23)	-0.0177	(-0.17)	-0.0716	(-0.65)
young					0	(.)	0	(.)	0.00566	(0.04)
large					0	(.)	0	(.)	0.0932	(1.25)
ratings							0	(.)	0.0000171**	(2.89)
score 1							0	(.)	0.132**	(2.77)
nb_tms							0	(.)	1.73e-08**	(3.10)
Observations	1370		1367		1367		989		1073	

t statistics in parentheses

^{*} p < 0.05, ** p < 0.01, *** p < 0.001

Comments to discuss with Gaétan

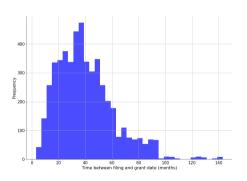
 two options for displaying results: either including fixed effects for products and firms, and thus excluding all the company and product specific variables or removing fixed effects like in column and only keeping sector fixed effects. Thee second option will allow us to observe the coefficients for the omitted variables.

Conclusion

References

Appendix

Table: Histogram: grant delay (months)



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