Perspectives on Computational Social Science

Problem Set #1

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1. Data Introduction and Access

The dataset I use is the NBER U.S. Patent Citations Data, which was collected and managed by the National Bureau of Economic Research. To be more specific, the exact data I would use is its patent data including constructed variables. There are 2,923,922 observations in this dataset, coming from all the U.S. utility patents granted between January 1963 and December 1999. Apart from basic information of the patents ("original variables", which consists of time, location of inventors, claims, and class), the dataset also contains information of their citations ("constructed variables", which consists of evaluations of each patent from different dimension). Furthermore, the collected patent covers six main technological categories: Computers and Communications, Drugs and Medical, Electrical and Electronics, Chemical, Mechanical, and Others. And for each technological category, there are several sub-categories.

NBER provides easy access to this dataset. This data, together with related documentations, now could be freely downloaded from its official website, under the website of the National Bureau of Economic Research.

2. Other key papers using this data

Most of the key papers having used this data are published on Research Policy, economic articles, and articles about innovation and technology. And the main concern of these papers is effects of patent policies.

Researchers have reviewed major changes in patent policy and their corresponding effects (Jaffe, 2000; Trajtenberg, 2001). More specifically, researchers have also analyzed the effect of patent policy on specific phenomenon, such as knowledge diffusion (Stolpe, 2002). But it is also indicated that problems might occur when highly interdependent technologies are attempted to be combined (Fleming & Sorenson, 2001).

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3. Data Collection

The data was produced with two means. The "original variables" came from records of the U.S. Patent Office, while the "constructed variables" were generated from the original data.

Table 1: Classification of Variables				
Original Variables	Constructed Variables			
Patent Number	Technological Category			
Grant Year	Technological Sub-Category			
Grant Date	Numer of Citations Made			
Application Year	Numer of Citations Received			
Country of First Inventor	Percent: Citations Made / Patents Granted)			
State of First Inventor (if US)	Measure of Generality			
Assignee Identifier (missing 1963-1967)	Measure of Originality			
Assignee Type	Mean Forward Citation Lag			
Number of Claims	Mean Backward Citation Lag			
Main Patent Class (3 digit)	Share of Self-Citations Made			
	Share of Self-Citations Received			

4. Descriptive statistics

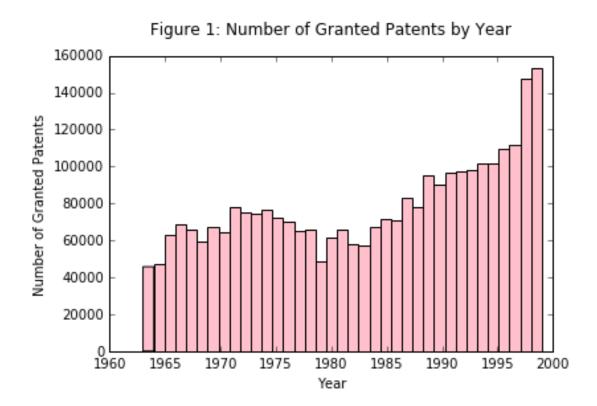
Since time, location of inventors, influence, and evaluations of generality and originality are key features of any patents, 10 corresponding variables are chosen for descriptive statistics. To make the structure more clear, I separate them into to two parts: continuous variables and categorical variables. Although variables of year is often treated as continuous variables in data analysis, it would be meaningless to treat them as continuous variables and calculate their mean and standard deviation. So in this part, I put them into categorical variables. And for categories variables with more than 10 categories, I aggregate them into fewer classes or cohorts.

Table 2: Data Description								
Continuous Variables								
Variables	Type	Min	Max	Mean	Std	Observations		
Num. of Citations Made	float	0	770	7. 720	9.000	2, 139, 314		
Num. of Citations Received	float	0	779	4.779	7. 346	2, 923, 922		
Generality	float	0	0.940	0.321	0. 285	2, 240, 348		
Originality	float	0	0. 951	0.349	0. 281	2, 042, 151		
Categorical Variables								
		Num. of Categories			Num. of			
Variables	Туре				Sub-	Observations		
					Categories			
Grant Year	int	37 years => 4 cohorts			37	2, 923, 922		
		1963-1970			8	481, 060		
		1971-1980 1981-1990 1991-1999			10	687, 818		
					10	737, 017		
					9	1, 018, 027		
Application Year	int	37 years => 5 cohorts <1963 1963-1970			37	2, 699, 606		
					41	6, 241		
					8	414, 390		
		1971–1980			10	658, 086		
		1981–1990 1991–1999			10	771, 006		
					9	849, 883		
Counry of First Inventor	str	162 => 2 cohorts US			162	2, 923, 922		
					0	1, 784, 989		
		Non-US			161	1, 138, 933		
State of First Inventor	str				57	1, 784, 989		
Assignee Type	int				7	2, 923, 922		
		unassigned				537, 988		
		U. S. NGO				1, 380, 310		
		Non-U.S. NGO				913, 470		
		U.S. Individual				24, 097		
		Non-U.S. Individual				9, 146		
		U.S. Government Non-U.S. Government				48, 323		
						10, 588		
Technological Category	int				6	1 ' '		
		Chemical			6	606, 934		
		Computers & Communications			4	290, 337		
		Drugs & Medical			4	204, 199		
		Electrical & Electronics			7	499, 741		
		Mechanical			6	681, 378		
	Others				9	641, 333		

5. Key Visualization

Patent production is a very significant indicator of technological innovation. Figure 1 shows number of patents produced each year. According to the histogram,

from 1963 to 1983, number of granted patents fluctuated within a limited range, but the trend is not positive. In 1979, the production of patents even reached its lowest point. Since 1980, number of granted patents has been increasing. The number even significantly increased after 1997, exceeding 150,000 for 1998 and 1999.



Generality and originality are also important features of patent production. In the more recent years, patterns of originality and generality have changed along the years of observation. According to Figure 2, the level of generality gradually increased from 1963 to 1975. Then it slightly fluctuated from 1975 to 1982, and has greatly decreased since 1983. In contrast, the level of originality has steadily increased since 1975. Although the data does not provide measure of originality from 1963 to 1974, the trend is still clearly shown by the plot. The comparison may mean that originality has grown to be more important than generality in recent years.

0.40 Mean Measure of Originality/Generality 0.35 0.30 0.25 0.20 0.15 0.10 Measure of Originality 0.05 Measure of Generality 0.00 1965 1970 1975 1980 1985 1990 1995 2000 Year

Figure 2: Mean Measure of Originality/Generality by Year

6. Conditional (slice) description

Patent production could have different patterns among disciplines. Figure 3 shows the number of patents of each year under each technological category. For Computers & Communications, and Drugs & Medical categories, the number of granted patents has kept growing. Particularly, it grew fast after 1990. But for the other four disciplines, the number of granted patents was fluctuating from 1963 to about 1997, then greatly increased in 1998 and 1999.

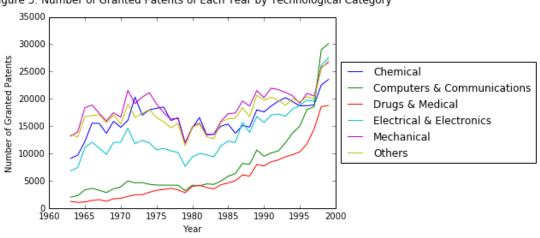


Figure 3: Number of Granted Patents of Each Year by Technological Category

It is also interesting to look at where the first inventors belong. According to Figure 4, at the beginning, the number of U.S. first inventors was almost 4 times as the number of non-U.S. first inventors. Then the number of non-U.S. first inventors kept growing, while the number of U.S. first inventors was fluctuating. Although the production of U.S. inventors started to increase after 1985, the difference between U.S. and non-U.S. has shrunk. In 1999, the number of U.S. first inventors is only 1.2 times of the number of non-U.S. first inventors.

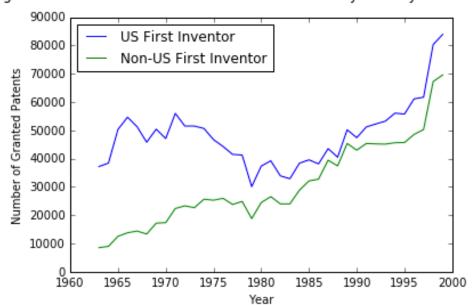


Figure 4: Number of Granted Patents of Each Year by Country of First Inventor

7. References

- [1] Hall B H, Jaffe A B, Trajtenberg M. The NBER patent citation data file: Lessons, insights and methodological tools[R]. National Bureau of Economic Research, 2001.
- [2] Fleming L, Sorenson O. Technology as a complex adaptive system: evidence from patent data[J]. Research policy, 2001, 30(7): 1019-1039.
- [3] Jaffe A B. The US patent system in transition: policy innovation and the innovation process[J]. Research policy, 2000, 29(4): 531-557.
- [4] Trajtenberg M. Innovation in Israel 1968–1997: a comparative analysis using patent data[J]. Research Policy, 2001, 30(3): 363-389.
- [5] Stolpe M. Determinants of knowledge diffusion as evidenced in patent data: the case of liquid crystal display technology[J]. Research Policy, 2002, 31(7): 1181-1198.