

Diversity of Firm Sizes, Complexity, and Industry Structure in the Chinese Economy

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Abstract

Among the phenomena in economics that are not yet well-understood is the fat-tailed (power-law) distribution of firm sizes in the world's economies. In the present paper we discuss different mechanisms suggested in the literature to explain this distribution of firm sizes. The paper uses the China Industrial Enterprises Database to study the distribution (firm size in terms of the number of employees, capital, and gross profit) for the provinces of China for the years 1998-2008. We estimate the power-law distribution and confirm its plausibility using the KS test and the log-likelihood ratio vs. log-normal and exponential distributions. The analysis on regional levels allows an assessment of regional effects on differences in the distribution; we discuss possible explanations for the observed patterns in the light of the recent regional economic development and the economic reforms in the PRC.

Keywords: Firm size distribution, Evolutionary industry dynamics, Power-law distribution, China

1. Introduction

The distribution of firm sizes in the world's economies seems to be remarkably homogeneous and stable across different countries as well as across time. It has been characterized as fat tailed, specifically as following a power law in the tail by different studies for different regions (e.g. the US (Axtell (2001)), the G7 countries (Gaffeo et al. (2003)), or China (Zhang et al.

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(2009))). Other studies Marsili (2005); Bottazzi et al. (2007); Dinlersoz and MacDonald (2009) found that the distribution only holds on an aggregate level across either sectors or firms of different ages - which may imply an evolutionary process or a self-organized criticality system depending on typical life-cycle developments of firms.¹ A consensus has not yet emerged and no generally accepted theory of how the distribution emerges has been found. There are, however, a number of candidate explanations as will be detailed in section 2 below - an ongoing debate to which the present study will contribute as well. While differences in the shape and parameters of the distribution have been investigated for sectors (Marsili (2005); Bottazzi and Secchi (2006); Bottazzi et al. (2007)) and international comparisons exist as well (Okuyama et al. (1999); Gaffeo et al. (2003)) there are to our knowledge no studies of regional differences within a country. While international differences also reflect differences in political systems and culture, regional variation may highlight what actually governs the emergence of the power law distribution and its particular shape. Of course this is also true for sectoral disaggregation, which has, however, already been studied extensively (Marsili (2005); Bottazzi et al. (2007); Bottazzi and Secchi (2006); Dosi (2007); Dosi et al. (2015)). This is related to the question of the very existence and persistence of diversity in firms and firm sizes (moreover in exactly this pattern), a question that also seems to require explanations related to evolutionary economics and self-organization (Nelson and Winter (1982); Dosi et al. (1995); Kwaśnicki (1998); Bottazzi and Secchi (2006)).

We use the People's Republic of China's² *China Industrial Enterprises Database* to analyze this distribution (the firm size in terms of 1) the number of employees, 2) capital, and 3) gross profit) for China for the years 1998 through 2008 both for the country as a whole and for individual provinces in section 4. It is attempted to model the distribution as a power law - a hypothesis that is supported by the Kolmogorov-Smirnov test for measure 1 (number of employees), as well as for some regions for measures 2 and 3 - following the well-established methods laid out in, e.g. Clauset et al. (2009) or Maschberger and Kroupa (2009). The analysis on regional (i.e. province level) and sectoral levels allows us to assess differences in the distribution

¹This idea was earlier proposed by Dosi et al. (1995).

²In the following, we occasionally refer to the PRC as China; we do not have data on either Taiwan or the Special Administrative Regions of the PRC (Macao and Hong Kong).

and their relation to the respective industry dynamics and regional specifics.

As firm size distributions are likely affected by - besides other factors - economic policy and development strategy, we devote section 3 to this aspect. The section covers the historical development of China, its economic policy since the 1980s, and how this may have impacted the distribution of the firm size differently across different provinces and in comparison to other countries. Specifically, reforms and economic transition towards a market economy in the 1980s and 1990s led to a very different development of different regions regarding their distinctive industrial profile, geographic advantages, and cultural characteristics. It may also have provided incentive for some regions to obtain and utilize first mover advantages and to create growth opportunities for other regions. A number of coastal cities and provinces were selected as the first special economic zones to be opened to the global market in 1980s, whereas the Western and Central areas did not benefit directly from the reforms until the late 1990s. This makes the regional firm size distribution in China a historically unique example the study of which may help in understanding the processes behind the astounding properties and similarities of firm size distributions in economies around the world.

Section 2 discusses the literature on firm size distribution with particular emphasis on the possible origins of the power law shape. This is followed by considerations on recent economic history in China in section 3. Section 4 proceeds with the empirical analysis and results for which potential explanations are introduced in section 5. Section 6 concludes.

2. Literature Review

It had first been hypothesized by Zipf and shortly afterwards empirically substantiated by Mandelbrot and Simon³ that firm sizes follow power law distributions, that is sizes s occur with frequency

$$p(s) = Cs^{-\alpha} \quad P(s > s^*) = \frac{C}{\alpha - 1} s^{-\alpha+1}.$$

In fact, it was conjectured with reasonable empirical evidence, that they specifically follow the Zipf distribution with exponent $\alpha = 2$.

³For a short historical overview, see Buendía (2013); for a comprehensive early theoretical account, see Ijiri and Simon (1977).

A number of theories have been put forward, why this might be the case (both, that the measure is power law distributed, if the second conjecture is accepted, that the exponent is $\alpha = 2$). Notable theories include:

1. A Gibrat process, a scale-invariant growth process which converges (fast enough) to a steady state distribution, will yield a power law with exponent $\alpha = 2$. Further, processes with almost any growth rate distribution will lead to a distribution converging to a power law as long as there is a lower bound to the distribution (Kesten process) (Gabaix (1999); Axtell (2001); Delli Gatti et al. (2005); Luttmer (2007)); see also item 4.
2. Exponentially mixed Gaussians, that is, lognormal distributions mixed with exponentials lead to power laws under certain conditions. The lognormal distribution can for instance be given by a normally distributed growth rate, the exponential distribution can be assumed for other measures, such as the firm age (Coad (2010)) or the firm's product diversity (Buldyrev et al. (2007)).
3. Aggregation over not power law distributed sectoral firm sizes as suggested by Dosi et al. (1995); Dosi (2007). The Bose-Einstein statistic, first suggested by Ijiri and Simon (1977) has been proposed as the distribution of growth opportunities that are subject to increasing returns following a generalized Eggenberger-Pólya-urn process and that result in Subbotin distributed (ideally, for the infinite limit Laplace distributed) firm growth rates (Bottazzi and Secchi (2006)).⁴
4. Multiplicative stochastic processes can give rise to power laws, the simplest such case being an AK model with A being drawn from a uniform distribution with nonzero mean (e.g. Zhang et al. (2009)).
5. The power law property is introduced through another variable and retained in the firm size. The variable often suspected to potentially act in this way is the productivity distribution across firms. A revenue

⁴It should be noted that exponential (in this theory Laplacian) growth rates should sum up to sizes that are Gamma distributed, not power law. There is limited evidence for sectoral size distributions that do not match the power law shape (Marsili (2005); Bottazzi et al. (2007)). In spite of overwhelming evidence for power law figures in the aggregated distributions, it has further been noted that scale free power law distributions should be persistent under disaggregation (Dosi et al. (2015)) which may cast doubt on whether the true distribution indeed follows a power law.

function which is exponential in the productivity with a positive exponent with an absolute value smaller than the productivity distribution's power law exponent will then yield another power law for the revenue distribution (Helpman et al. (2004)).

6. Preferential attachment in networks yields degree distributions that follow power laws. If the economy is modeled as a network where the firm size depends on the degree of the node (firm) and has a positive influence on further increase of that degree, the power law is retained in the firm size distribution (Stephen and Toubia (2009); Dahui et al. (2006)).
7. Other processes involving self-organized criticality (in a similar way to Gibrat and Kesten processes) could as well be used to model the emergence of power laws (Delli Gatti et al. (2005); Battiston et al. (2007)).

More recently, some scholars doubt the nature of the firm size distribution as a power law. Since its emergence is not well-understood, it is possible that the firm size actually follows an unidentified distribution similar to but different from a power law. As long as good fits can be obtained for a power law over several orders of magnitude, however, the interpretation of the parameters would remain largely unchanged; the assessment of some properties of the theoretical distribution (likelihood of 'black swans', extreme tail observations, existence of moments, convergence with increasing sample size) may change.

Empirical power law fits of the firm size distribution generally yield power law exponents around $\alpha = 2$; the specific values reported in the literature are summarized in table 1. There is evidence for some variation across sectors, countries, and time. However, very few scholars interpret these variations as actually meaningful. Notably, Gaffeo et al. (2003) have proposed and provided evidence for falling exponents during recessions. The literature tradition following Helpman et al. (2004) has attempted to establish that exponents firms differ between samples of exporting and non-exporting firms (di Giovanni et al. (2011); Sun and Zhang (2012)).

Further, there are indications that different measures of firm sizes yield slightly different exponents (see table 1). The exponent for sales is reported to be slightly lower than that for capital by Gaffeo et al. (2003) (but for some cases slightly higher in Fujimoto et al. (2011)). The exponent for the distribution in the number of employees is for many countries (including

China) reported to be higher than that for either sales or capital in Fujimoto et al. (2011).⁵

While the firm size distribution is generally stable across time and geographic region, some differences have been reported; of interest for the present study are those that directly concern China. Both Duschl and Peng (2015) and Yu et al. (2015) recently found that the characteristics of Chinese firms vary greatly with the ownership structure. Duschl and Peng (2015) report lower average growth for state-owned enterprises but a much higher probability to become high growth firms while foreign owned enterprises are also more likely to show high growth rates. Yu et al. (2015) indicate larger shifts in the productivity distribution in the 1990s and 2000s resulting from exit and changes in ownership and organizational structure. Other aspects seem unique for China as well: the variance scaling relationship (inverse linear relationship of the variance of growth rates and the logarithm of the size) differs from Western economies (the parameter being smaller) according to Duschl and Peng (2015) while Yu et al. (2015) report a very stable productivity growth distribution on the aggregated level compared to other emerging economies.

A related problem is the very existence and persistence of diversity in firms (moreover in exactly this pattern). Among the approaches put forward in relation to this question those from evolutionary economics (Nelson and Winter (1982); Dosi et al. (1995); Kwaśnicki (1998); Bottazzi and Secchi (2006); Dinlersoz and MacDonald (2009)), for a comprehensive overview, see e.g. Kwaśnicki (1998), more recent advances are discussed in, e.g. (Schwardt and Schwesinger (2013); Dosi et al. (2015)) seem to be the most promising. Several explanations have been put forward; most obviously, firms compete with different degree of success especially with respect to processes that are subject to uncertainty such as innovation and research. In fact, this is the core of the earliest Schumpeterian agent-based models of economic change (Nelson and Winter (1982)). More recent approaches emphasize that there may be a general mechanism that breaks the symmetry of the underlying growth and contraction process. Increasing returns (network externalities, for instance) lead to advantages for established competitors with large user bases for their products - which is arguably particularly important in mod-

⁵For disaggregated (sectoral) data (Bottazzi et al. (2007)) even report some cases of different modality for different firm size measures.

Author(s)	Com. CDF Exponent	PDF Expo- nent	Measure	Data
Okuyama et al. (1999)	0.72-1.4	1.72-2.4	income	USA, Japan 1990s
Axtell (2001)	0.994-1.0039	1.994-2.0039	employees	USA 1988-1997
Gaffeo et al. (2003)	0.81-0.97	1.81-1.97	sales	CAN, USA, JAP, ITA, FRA, GER, UK
	1.04-1.328	2.04-2.328	capital	
	0.73-0.84	1.73-1.84	debt	
Silverberg and Verspagen (2007)	1.0	2.0	innovation (patent)	USA, Europe
			revenue	
Helpman et al. (2004)	0.4-1.8	1.4-2.8	sales	Europe by sectors 1997
Zhang et al. (2009)	0.937-1.013	1.937-2.013	revenues	Top 500 firms, China 2002-2007
di Giovanni et al. (2011)	0.949-1.111	1.949-2.111	employees	France
	0.362-1.663	1.362-2.663	sales	
Fujimoto et al. (2011)	0.6315-1.2056	1.6315-2.2056	capital	worldwide
	0.8234-1.6516	1.8234-2.6516	employees	
	0.7241-1.3103	1.7241-2.3103	sales	
	0.867	1.867	capital	China
	1.2881	2.2881	employees	
	0.9972	1.9972	sales	
Sun and Zhang (2012)	0.411-0.81	1.411-1.81	sales	China by sectors 1998-2007
Buendía (2013)	1.096-1.328	2.096-2.328	market share	Telecom. and Aerospace sector, USA

Table 1: Power law exponent estimates for firm size from the literature.

ern ICTs (Arthur (1996); Bottazzi and Secchi (2006); Heinrich (2013, 2014); Buendía (2013)). An alternative approach is that in the presence of different (substitutable) input factors, firm types with different factor utilization (productivity) profiles will obtain different sustainable market shares (i.e., in the steady state) (Schwardt and Schwesinger (2013)).

3. Recent Economic Policy and Development in the PRC

Before we proceed with the empirical analysis and the evaluation of potential explanations in sections 4 and 5, we discuss three aspects of the recent economic history of China: the gradual transition to market economy (which happened at different times for different regions), the reform of state-owned enterprises, and the rise and fall of township and village enterprises. All of these and other aspects that we will merely touch upon likely have a strong impact on the current distribution of firm sizes in China. China is unique in that its transition to the market economy occurred stepwise, thus creating considerable regional variety, particularly between coastal, Central, and Western provinces, as well as resource-based provincial economies as is the case in Inner Mongolia. The differences can most clearly be seen in figure 15 which shows the median firm age in 2008⁶: low in Inner Mongolia and in the Central provinces (that had recently completed their market economy transition reforms), higher in coastal regions (that completed the transition less recently) and especially in the Western part of the country (where the transition was just under way). As the pattern is clearly visible in the age distribution of firms, an effect on the size distribution should clearly be expected.

3.1. *China's Gradual Transition to the Market Economy*

After the political and economic chaos in 1960s and 1970s, the Chinese government decided to make economic development a central strategic objective in 1978. Effecting a radical break with the economic policies of other socialist economies, China gradually started to transition from a planned economy to a market economy system.

The most striking reform strategy was to open the country to the global market by designating Special Economic Zones, in which the central government reduced some restrictions on international trade and offered special

⁶Computed from the Chinese Industrial Enterprises Database also used in section 4.

priorities and incentives in terms of tax benefits, financial support, and more flexible administration. In 1980, four cities, Shenzhen, Zhuhai, Shantou in Guangdong, as well as Xiamen in Fujian, were established as special economic zones, aiming at attracting international investment and building capabilities in international trade. Inspired by the success of the special economic zones, in 1984 the government added a further 14 coastal cities as Coastal Open Cities.⁷ The Yangtze River Delta, the Pearl River Delta, and the Minnan Delta (roughly geographically equivalent to Fujian province) were designated as Economic Open Zones shortly thereafter. By 1988, the Special Economic Zones, which jointly comprise 20% of the population of China provided tremendous increases in the volume of China's international trade. After 1992, the central government announced that China as a whole would gradually establish a market economy system, so many more special economic zones would be established also in other provinces while China would become more open to and increasingly interconnected with the global market.

A second critically important issue was China's "regional experimentation" strategy. That is, a trial and error approach with considerable regional autonomy was taken with the aim of pushing the reform forward before the central government is capable of evaluating its overall effects. This hence became a crucial factor for local economic development; the Special Economic Zones were just the boldest of the economic experiments, bearing considerable risk for the regional government but also entailing both funding from the central government and the potential fruits of successful development. In effect, the central government intended to create competition among local governments, which is perhaps unique to the Chinese case and may shape the country's future in many aspects Xu (2011). Further, the coastal provinces as the most developed areas in China had held more influence on the central government's policies, whereas the Central and Western provinces could not achieve their objectives as easily. It was not until the late 1990s, that the central government started to allocate more resources to the less developed provinces. This was finally done in the Great Western Development Strategy in 1999, the Northeast Area Revitalization Plan in 2003, and the Rise of Central China Plan in 2004.

⁷The 14 COCs were Tianjin, Shanghai, Dalian (Liaoning), Qinhuangdao (Hebei), Yantai (Shandong), Qingdao (Shandong), Lianyungang (Jiangsu), Nantong (Jiangsu), Ningbo (Zhejiang), Wenzhou (Zhejiang), Fuzhou (Fujian), Guangzhou (Guangdong), Zhanjiang (Guangdong), Beihai (Guangxi).

Migration to the coastal cities may have added to existing imbalances in developmental potential. The later development of the Central and Western provinces, while improving their economic position, did certainly not give them the same comparative advantages and the same capacity to attract investments and skilled labor.

Undoubtedly, China's gradual transition strategy in fact results in immense regional development disparities and accordingly leads to visible disparities in distribution of firms' sizes and average ages. In fact, our results in section 4 will show that the economic structure in terms of firm sizes is markedly different for the coastal areas and also between the Central and Western provinces that also enjoyed their respective development programs at different times.

3.2. The State-Owned Enterprise Reform

Since 1979, the Chinese government has also implemented a series of reforms in state-owned enterprises (SOEs). Within the time frame considered in the empirical study in this paper, more than 90% of SOEs were either privatized or discontinued (many went bankrupt). This might have had significant effects on the distribution of sizes, numbers, and ages of firms. For instance, in the database we used, SOEs account for 34% of the firms (56151 out of 165118) in 1998. This share drops to 2.2% (9264 out of 412212) in 2008.

When the profitability of SOEs did not improve in spite of the reforms, the central government took a more radical stance in 1995. Smaller SOEs were closed, privatized or bankrupted, while the larger ones continued to be maintained and even expanded through capitalization and mergers and acquisitions. Moreover, the SOEs are often comparatively large but are listed as several relatively smaller SOEs in the database we used. For example, the Fortune 500 in 2014 listed 100 Chinese firms, but 92 of 100 those firms were SOEs. Therefore, calculating firm size distributions including these large state-owned enterprises (with size measured in terms of capital) would make the distribution even more tail-heavy.

It is likely that this led to an unusual pattern of entries and exits (bankruptcies and new establishments) of firms. The regional distribution of discontinued SOEs and emerging private enterprises may have added to this and likely had an effect on the firm size distribution for the subsequent years including the time period studied here (between 1998 and 2008). For example, in the

early years after the reform of SOEs only the Eastern (coastal) provinces developed the private sector of the economy intensively. Currently, most SOEs are active in monopoly industries considered as vital and strategic sectors in terms of national security, for example, electricity, oil, natural resource, defense, etc.

3.3. Township and Village Enterprises

Another interesting case in China is the rise and fall of township and village enterprises. Generally, township and village enterprises are firms ran in the town or village by the local government; they contributed significantly to China's agricultural reform at the early stage. Emerging at the end of the 1970s, they grew very fast. The industrial output of township and village enterprises grew at 38.2% annually in the period from 1982 to 1988; the number of employees in those enterprises grew from 28 Million in 1978 to a peak of 135 Million in 1996. But due to the following reforms, the private economy was growing much faster after 1992, most township and village enterprises were bankrupted and only a minority was privatized. For example, alone in 1995 nearly 3 Million township and village enterprises were bankrupted. Therefore, this sort of historical breaking point might have led to significant changes in the distribution of size, number, and age of firms. Most crucially, if the power-law distribution in firm sizes is assumed to be the result of self-organization processes, the time at which this vast restructuring took place may have affected the speed with which the distribution converges to the power law, as well as, more directly, relocalization decisions of certain firms, bankruptcy and startup-establishment rates, and thus perhaps also the dispersion of firm sizes within the distribution and the parameters of the power law as suggested in section 5.

The reforms of state-owned enterprises and township and village enterprises may also provide an explanation why the average firms in Eastern provinces are younger than their Western and Central counterparts. In the Eastern provinces, township and village enterprises have even been developed very well and private economy is very active, hence, we can assume the birth/death rate of firms in those provinces to be higher than that of other regions.

Based on the recent economic development of China, we can expect to find considerable regional variation in the firm size distribution, although the nature of this variation is not clear from the aspects discussed here. On the one hand, either the fact that China started out as a planned economy or

the radical reforms may result in an economic structure and specifically a firm size distribution unlike that of any other country. We might not even find a heavy tailed distribution, power law or otherwise. On the other hand, the prevalence of large firms, particularly large SOEs might lead to a more tail-heavy firm size distribution, especially in the financial and industrial centers in the Eastern coastal areas. When fitted to a power law, this would yield lower exponents (provided the fit is plausible⁸). Further, the pattern of economic growth, bankruptcies, and entry of new firms may influence the firm size distribution via various mechanisms and may yield another, different result. We will return to this in the discussion of results in section 5.

4. Empirical Analysis

4.1. Data and Method

We use the Chinese Industrial Enterprises Database for the years between 1998 and 2008. The database contains data on between 165118 (1998) and 412212 (2008) firms in the People’s Republic of China.⁹ The firm size distribution is studied using three different measures, 1) the number of employees, 2) capital (specifically paid-in capital¹⁰), and 3) gross profit as they are included in the database.¹¹ The distributions are evaluated by year and by province.

The power law distribution is fitted using the well-established method laid out in detail in Clauset et al. (2009) or Maschberger and Kroupa (2009). That is, the minimum value and the exponent of the power law are fitted

⁸As explained below, the plausibility of power laws is assessed using a KS-Test. If the fit is plausible, this also holds for obtained characteristics and the resulting interpretation.

⁹Only firms with annual sales > 5 million RMB are included, which does, however, not affect the estimation the power law exponents as power laws only hold for the part of the distribution beyond some minimum value which is estimated simultaneously with the exponent (i.e., not for small firms). It under-represents small and medium sized firms, likely including many younger firms, in other variables from the database that are used later on, specifically in the computation of the median firm age.

¹⁰This includes share capital as well as additional revenue from shares sold in excess of par.

¹¹The number of employees is given as ”从业人数” (number of employees), ”全部职工” (all employees) etc. for different years; capital as ”实收资本” (paid in capital), and gross profits as ”利润总额” (total profit). The firm age is computed from founding year and month (”开工时间 (年)”, ”开工时间 (月)”) as given in the database.

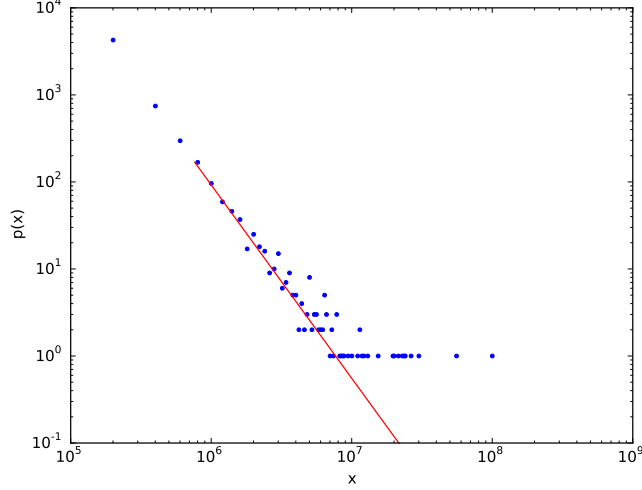


Figure 1: Firm size (measured in capital) distribution, China, 2008

simultaneously, assigning each candidate minimum value a maximum likelihood exponent and choosing the exponent that minimizes the Kolmogorov-Smirnov statistic (distance of empirical and fitted CDF). The standard error of the exponents is bootstrapped from samples of 500 artificial distributions with the fitted parameters. The goodness of fit of the estimates is evaluated using the KS test for plausibility and Vuong's test (log-likelihood ratio with Gaussian error < 0.1) for comparisons with fits to alternative distribution types (lognormal and exponential).

4.2. Results

The firm size distribution in terms of capital for the entire country is shown for 2008 (as an example) in figures 1 (PDF) and 2 (complementary CDF); that in terms of numbers of employees is shown in figures 3 (PDF) and 4.¹² The power law form is visually plausible. Further we show the firm size distribution for all three measures (number of employees, capital, gross profits) for Shanghai as an example in figures 5, 6, and 7 (all figures complementary CDFs for 2008).

¹²We do not show the figures for the firm size distribution in terms of gross profits; it is almost identical to that in terms of capital.

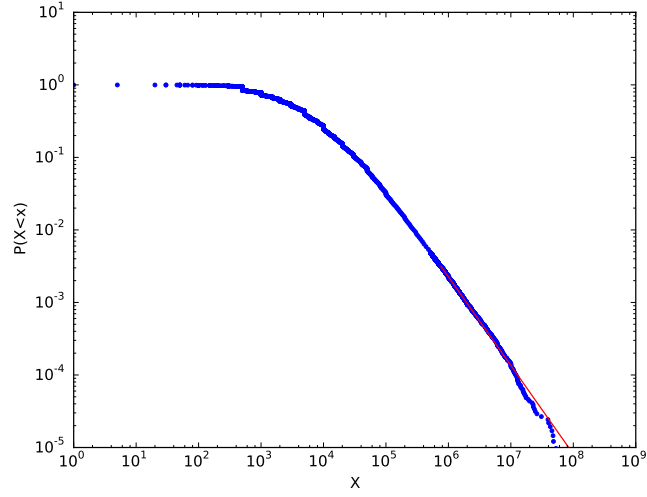


Figure 2: Survival function (complementary cumulative distribution function) for the firm size (measured in capital) distribution, China, 2008

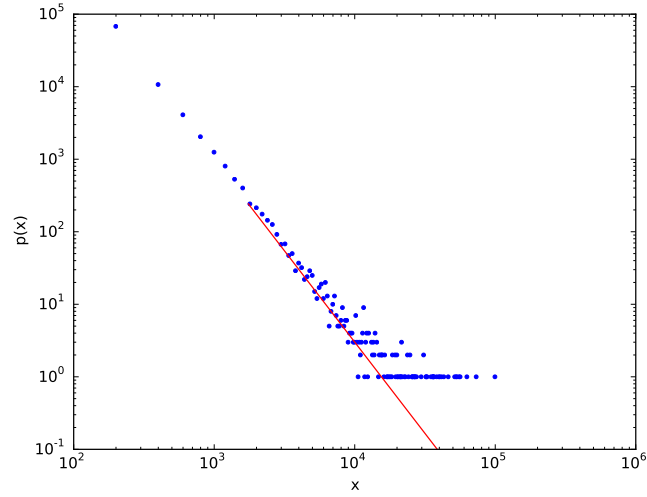


Figure 3: Firm size (measured in numbers of employees) distribution, China, 2008

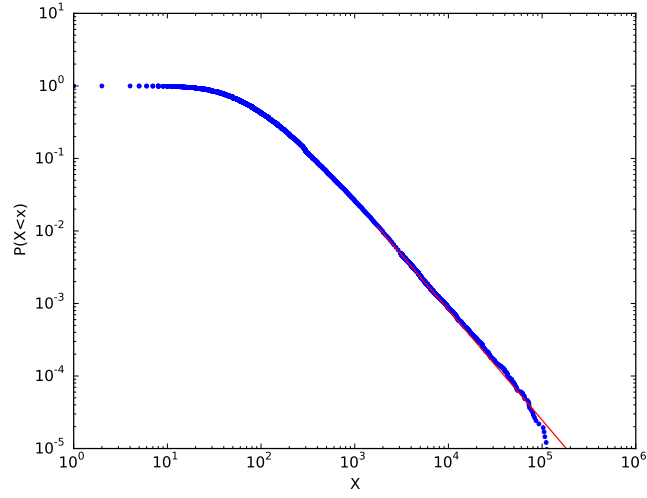


Figure 4: Survival function (complementary cumulative distribution function) for the firm size (measured in numbers of employees) distribution, China, 2008

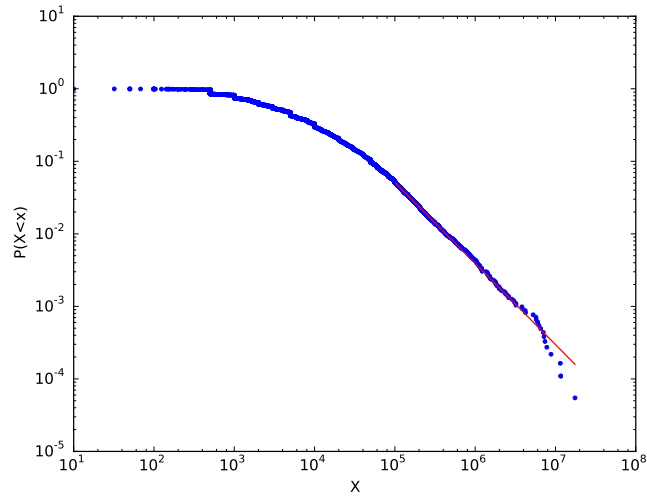


Figure 5: Survival function (complementary cumulative distribution function) for the firm size (measured in capital) distribution, Shanghai, 2008

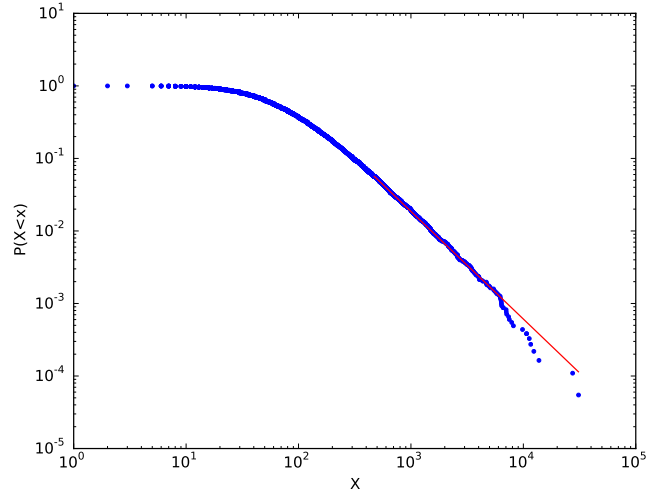


Figure 6: Survival function (complementary cumulative distribution function) for the firm size (measured in numbers of employees) distribution, Shanghai, 2008

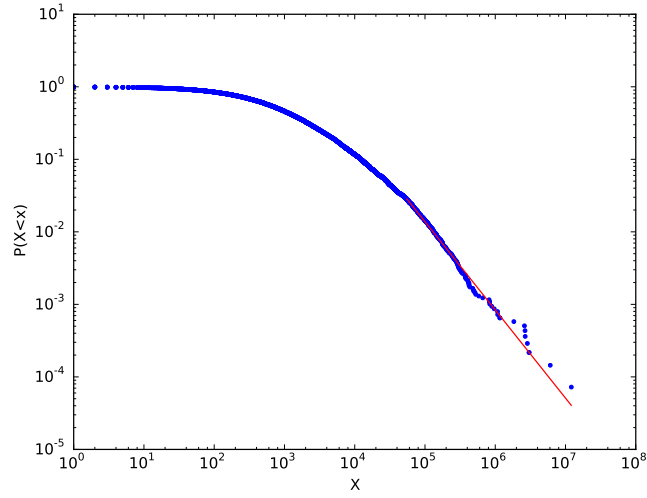


Figure 7: Survival function (complementary cumulative distribution function) for the firm size (measured in gross profits) distribution, Shanghai, 2008

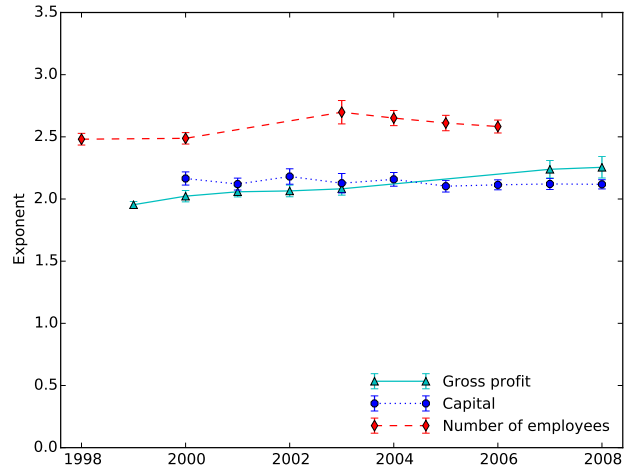


Figure 8: Development of power law exponents for firm size in terms of capital, gross profits, and number of employees for Shanghai, 1998-2008, significant values only (KS test with 10% threshold and log-likelihood ratio vs. lognormal and exponential fit).

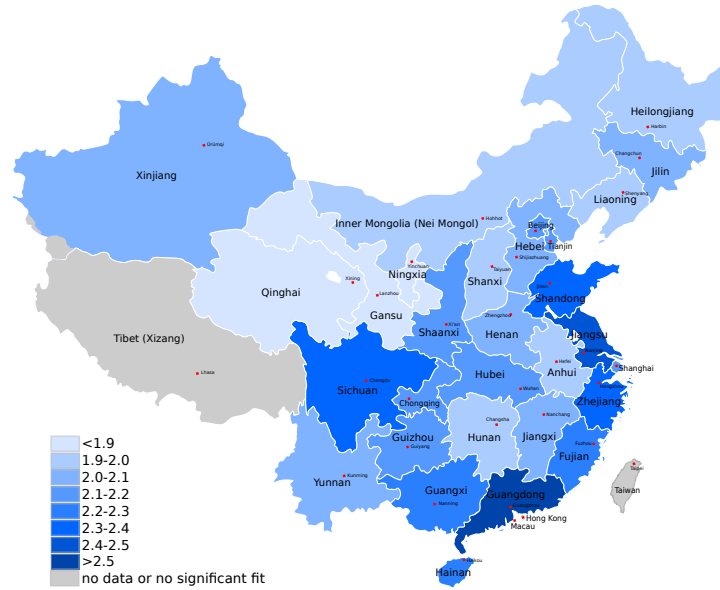


Figure 9: Power law exponents for firm size measured in capital, by province, 1998, values as in table 2.

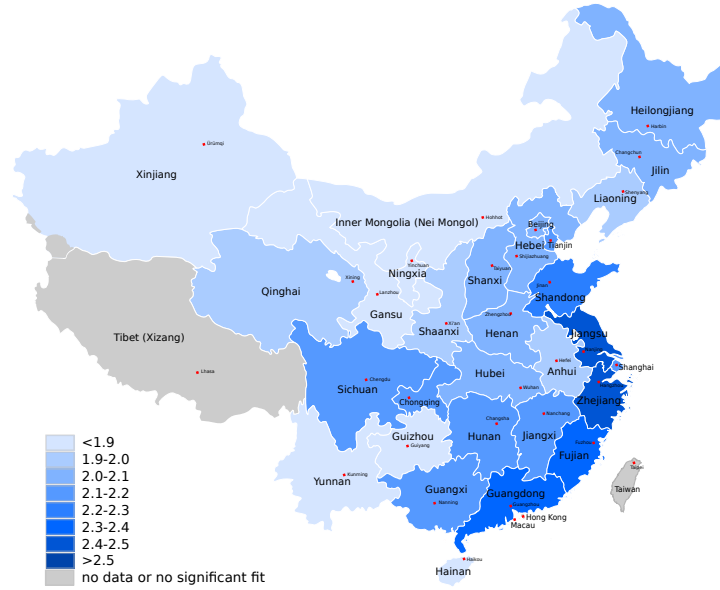


Figure 10: Power law exponents for firm size measured in capital, by province, 2008, values as in table 2.

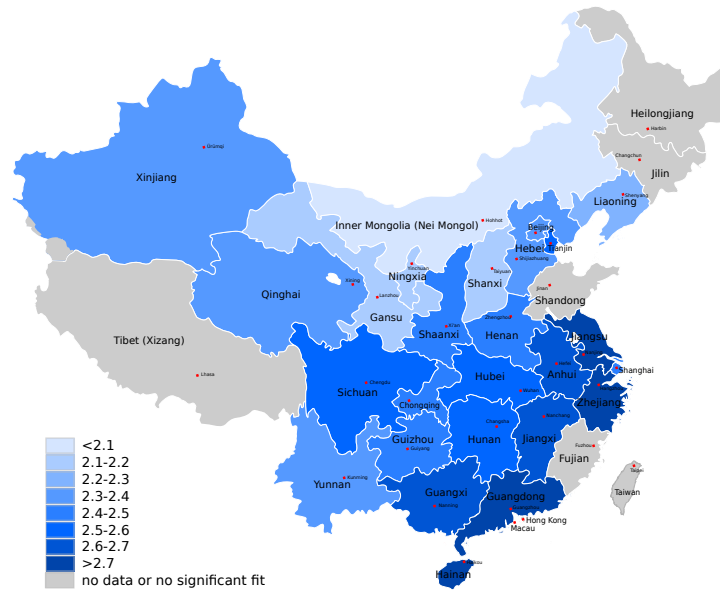


Figure 11: Power law exponents for firm size measured in number of employees, by province, 1998, values as in table 2.

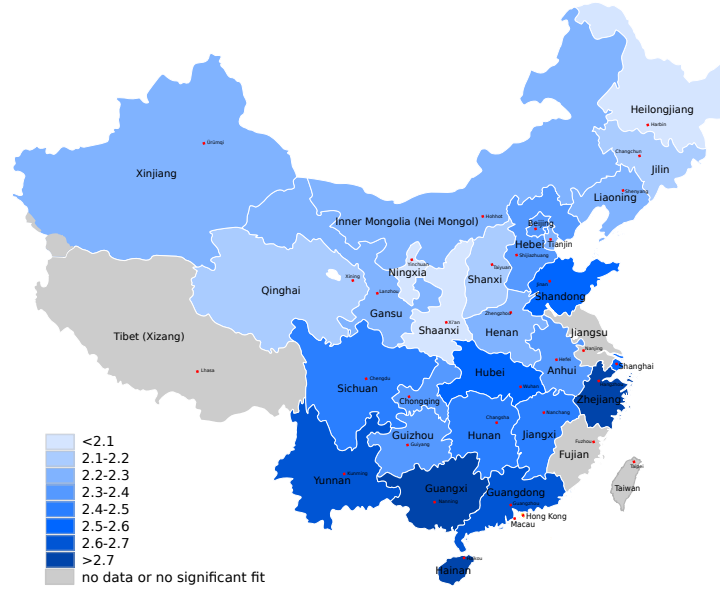


Figure 12: Power law exponents for firm size measured in number of employees, by province, 2008, values as in table 2.

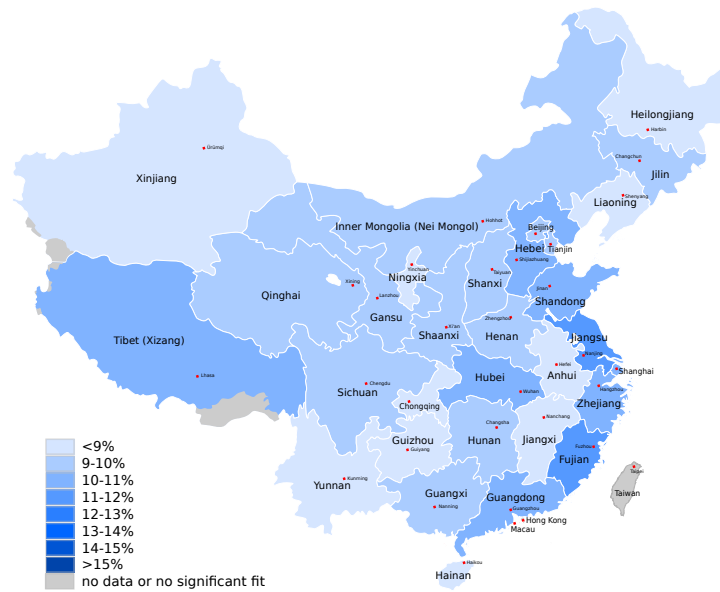


Figure 13: Annual GDP growth 1998 by province.

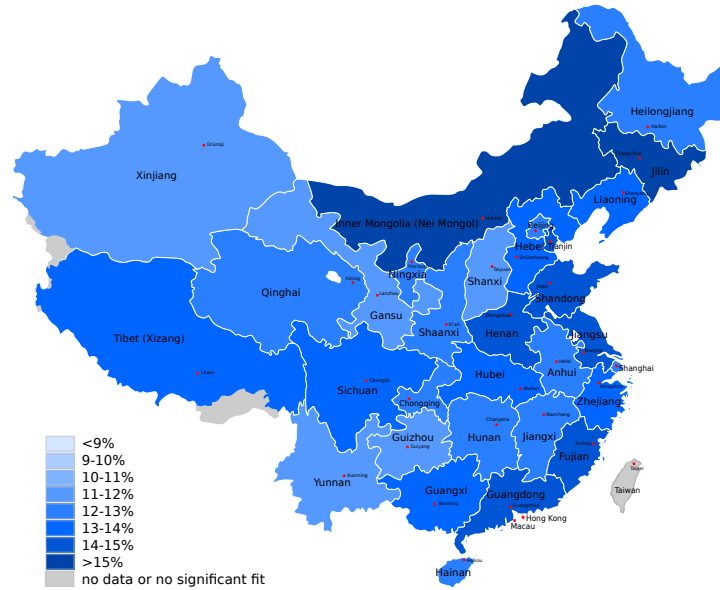


Figure 14: Annual GDP growth 2008 by province.

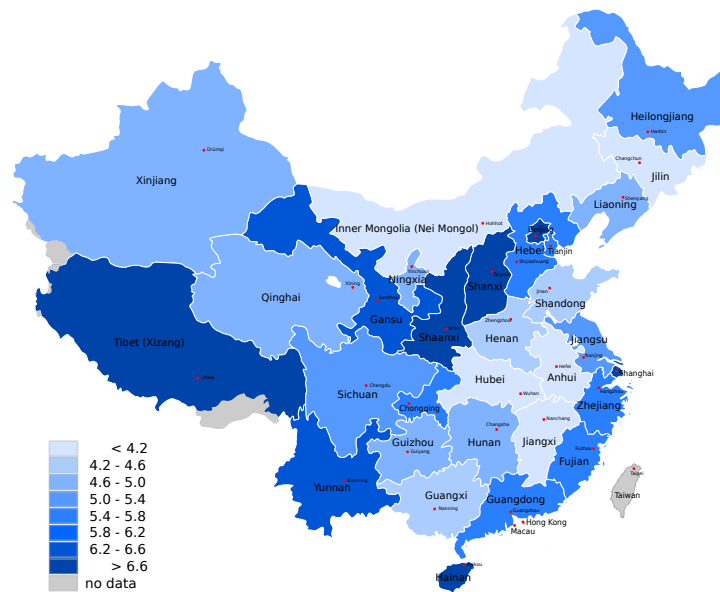


Figure 15: Median age of firms 2008 by province.

Region	α (#Emp. 1998)	KS p	α (#Emp. 2008)	KS p	α (Cap. 1998)	KS p	α (Cap. 2008)	KS p
Inner Mongolia	2.087 \pm 0.062	0.326 ³	2.248 \pm 0.041	0.176 ³	1.933 \pm 0.075 ¹²	0.304 ³	1.853 \pm 0.041 ¹	0.03 ³⁴
Shanxi	2.149 \pm 0.038	0.25 ³	2.152 \pm 0.033	0.128 ³	1.903 \pm 0.027	0.256 ³	2.0 \pm 0.058 ¹	0.164 ³
Hebei	2.317 \pm 0.043	0.356 ³	2.323 \pm 0.05 ¹	0.468 ³	2.014 \pm 0.043	0.624 ³	2.009 \pm 0.042 ¹	0.192 ³
Beijing	2.337 \pm 0.061 ¹	0.2 ³	2.301 \pm 0.047	0.88 ³	2.119 \pm 0.074	0.45 ³	2.068 \pm 0.078 ¹	0.122 ³
Liaoning	2.227 \pm 0.048 ¹	0.418 ³	2.226 \pm 0.043 ¹	0.462 ³	1.941 \pm 0.037 ¹	0.448 ³	1.95 \pm 0.048 ¹	0.474 ³
Jilin			2.14 \pm 0.068	0.65 ³	2.056 \pm 0.078 ¹	0.148 ³	2.046 \pm 0.084 ¹	0.42 ³
Heilongjiang			2.074 \pm 0.035 ¹	0.492 ³	1.927 \pm 0.045	0.14 ³	2.043 \pm 0.098	0.362 ³
Shanghai	2.481 \pm 0.047	0.578 ³	2.583 \pm 0.052 ¹	0.234 ³	2.165 \pm 0.053 ¹	0.268 ³	2.118 \pm 0.037	0.642 ³
Jiangsu	2.803 \pm 0.084 ¹	0.244 ³			2.416 \pm 0.087 ¹	0.04 ³⁴	2.411 \pm 0.051	0.266 ³
Anhui	2.606 \pm 0.083	0.11 ³	2.345 \pm 0.046	0.282 ³	1.931 \pm 0.034	0.36 ³	1.981 \pm 0.065	0.22 ³
Shandong			2.581 \pm 0.074	0.304 ³	2.308 \pm 0.063	0.164 ³	2.236 \pm 0.05	0.148 ³
Tianjin	2.587 \pm 0.091	0.256 ³	2.281 \pm 0.045 ¹	0.108 ³	2.144 \pm 0.071 ¹	0.14 ³	2.104 \pm 0.046	0.032 ³
Zhejiang	3.003 \pm 0.093	0.366 ³	2.857 \pm 0.055	0.422 ³	2.39 \pm 0.116 ¹	0.388 ³	2.441 \pm 0.053 ¹	0.186 ³
Jiangxi	2.656 \pm 0.111	0.306 ³	2.458 \pm 0.083 ¹	0.73 ³	2.066 \pm 0.089	0.634 ³	2.136 \pm 0.084 ¹	0.14 ³
Fujian					2.222 \pm 0.054	0.152 ³	2.353 \pm 0.104	0.426 ³
Chongqing	2.495 \pm 0.099	0.808 ³	2.391 \pm 0.071 ¹	0.696 ³	2.11 \pm 0.074	0.18 ³	2.153 \pm 0.079	0.2 ³
Hunan	2.584 \pm 0.112	0.516 ³	2.408 \pm 0.056 ¹	0.51 ³	1.947 \pm 0.032	0.074 ³⁴	2.145 \pm 0.08	0.458 ³
Hubei	2.55 \pm 0.09 ¹	0.284 ³	2.516 \pm 0.073	0.18 ³	2.169 \pm 0.081	0.506 ³	2.042 \pm 0.062 ¹	0.144 ³
Henan	2.465 \pm 0.054	0.17 ³	2.286 \pm 0.035 ¹	0.318 ³	2.033 \pm 0.04	0.376 ³	2.072 \pm 0.044	0.206 ³
Guangdong	2.819 \pm 0.068	0.068 ³⁴	2.636 \pm 0.048	0.044 ³⁴	2.671 \pm 0.1 ¹	0.52 ³	2.39 \pm 0.05 ¹	0.114 ³
Guangxi	2.63 \pm 0.101	0.546 ³	2.725 \pm 0.143 ¹	0.432 ³	2.206 \pm 0.06	0.12 ³	2.187 \pm 0.09 ¹	0.144 ³
Guizhou	2.482 \pm 0.164	0.566 ³	2.379 \pm 0.057	0.206 ³	2.134 \pm 0.104 ¹	0.576 ³	1.837 \pm 0.032 ¹	0.186 ³
Hainan	3.271 \pm 0.323 ¹	0.598	2.731 \pm 0.178 ¹	0.326 ³	2.285 \pm 0.148	0.246 ³	1.864 \pm 0.079 ¹	0.206 ³
Sichuan	2.586 \pm 0.108	0.1 ³	2.466 \pm 0.051	0.304 ³	2.36 \pm 0.089 ¹	0.146 ³	2.17 \pm 0.05 ¹	0.304 ³
Yunnan	2.399 \pm 0.078	0.238 ³	2.6 \pm 0.078	0.416 ³	2.017 \pm 0.053	0.554 ³	1.891 \pm 0.035 ¹	0.038 ³⁴
Shaanxi	2.496 \pm 0.125	0.236 ³	2.079 \pm 0.032	0.708 ³	2.182 \pm 0.087 ¹	0.146 ³	1.939 \pm 0.044 ¹	0.244 ³
Gansu	2.104 \pm 0.054 ¹	0.536 ³	2.259 \pm 0.059	0.254 ³	1.811 \pm 0.042	0.214 ³	1.846 \pm 0.037	0.126 ³
Ningxia	2.137 \pm 0.084	0.442 ³	2.06 \pm 0.063	0.292 ³	1.748 \pm 0.056 ¹	0.458 ³	1.805 \pm 0.055 ¹²	0.588 ³
Qinghai	2.337 \pm 0.17	0.514	2.127 \pm 0.134	0.502 ³	1.869 \pm 0.071	0.412 ³	1.931 \pm 0.121	0.502 ³
Xinjiang	2.391 \pm 0.117 ¹	0.226 ³	2.213 \pm 0.101	0.322 ³	2.025 \pm 0.074	0.25 ³	1.868 \pm 0.069	0.482 ³

Table 2: Fitted power law exponents and KS p-value for firm size measured by capital and number of employees, by province, 1998 and 2008. Significant fits only (KS-test and log-likelihood ratio vs. lognormal and exponential fits); for some values not significant fits were replaced with significant fits of successive or previous years (1999 or 2000 instead of 1998; 2006 or 2007 instead of 2008, indicated by superscript ¹); for some values not significant fits for capital were replaced for significant fits for gross profits (indicated by superscript ²); fits significantly superior to exponential fits (log-likelihood ratio) are indicated by superscript ³; fits that fail to achieve the 10% level of the KS p-value but still have a KS p-value > 2.5%, thus significant with a higher error probability, are indicated by superscript ⁴.

Not all the fits for all provinces and all years were plausible - some were rejected by either the KS-test or Vuong's test (vs. lognormal or exponential).¹³ Plausible fits were mostly stable in time (figure 8 shows the time development for Shanghai). The exponents for the firm size distribution in terms of the number of employees were found to be somewhat higher than that for gross profits or capital in the great majority of cases. The exponents for the firm size distributions in terms of capital and gross profits were found to be generally very close to one another (with sometimes the exponent for the gross profits being slightly lower than that for capital). These empirical findings as well as the numeric estimates (some of which are given in table 2) are in good agreement with the literature as discussed above (Zhang et al. (2009); Fujimoto et al. (2011); Gaffeo et al. (2003)).

Exponents for the early (1998) and the late part (2008) of the sample are given for all provinces in table 2 for the firm size distribution measured in number of employees and in capital. Rejected fits have, where possible, been exchanged for either better fits in the immediately succeeding or preceding years (1999 or 2000 for 1998 and 2006 or 2007 for 2008); where this was not feasible, we have attempted to replace rejected fits for capital by fits for gross profits (since the exponent for these measures were almost always very close). There are a few instances where even so, we could not give any plausible fits or approximations from succeeding/preceding years (missing values in the table).¹⁴

We study the correlations of the so obtained samples of exponents (for provinces) with GDP growth in 1998 and in 2008¹⁵ as well as with the average firm age in 2008. Moderately good correlations are obtained between growth 1998 and the capital exponents 1998 (0.488, p -value < 0.01), growth 1998 and the capital exponents 2008 (0.635, p < 0.001), growth 1998 and the numbers of employees exponents 2008 (0.321, p < 0.1), as well as between growth 2008 and the capital exponents 2008 (0.31, p < 0.1), and between the median firm age 2008 and the capital exponents 1998 (-0.533, p < 0.01) and 2008 (-0.531, p < 0.01) as well as median firm age 2008 and the number of employees exponents 1998 (-0.345, p < 0.1) and 2008 (-0.326, p < 0.1). The

¹³Most fits were significantly better than the exponential fit; comparisons with lognormal were mostly inconclusive; see table 2.

¹⁴For Tibet, the sample size was too small to obtain reasonably good estimates.

¹⁵Data from China Statistical Yearbooks from the National Bureau of Statistics in China, <http://www.stats.gov.cn>.

exponents are all positively correlated among themselves, $p < 0.01$.

The geographic distribution is illustrated in figures 9, 10, 11, 12, 13, 14, and 15. That the exponents are closely related can most clearly be seen from the Eastern and Southern coastal areas between Guangdong, Shanghai, and Tianjin. From figures 12, 13, and 14, it can also be seen that there is a connection between exponents on the one hand and both GDP growth and firm age on the other hand (as indicated in the correlation coefficients above).

5. Discussion of Results

5.1. *The Power Law and the Regional Pattern*

In the light of the recent economic development in China as detailed in section 3, it is likely that it also had a strong effect on the firm size distribution. As mentioned, the first regions where the transition to a more open market economy was implemented were the coastal regions, the very regions that also visibly stand out in the pattern of power law exponents estimated in section 4 (see figures 9 through 12). Power law exponents were higher in these regions, indicating 1) a steeper tail distribution with less weight on extreme tail events (super-large firms in this case), 2) a smaller dispersion, smaller skewness, etc., and, if the distribution indeed obeys a power law, 3) the finiteness of the mean of the distribution.¹⁶ This result is not intuitive as it is precisely the coastal region where some of the most dynamic industrial clusters, the financial centers, and the headquarters of some of the largest firms are located which would suggest a particularly large dispersion and high share of tail observations (i.e. small exponents). However, the rest of the distribution seems to adjust in a way that does not only counterbalance this but tilts the exponent into the other (higher) direction.

Potential consequences of different exponents include a relatively higher confidence in handling the tail of the distribution. In particular, the distribution financial risk associated with individual bankruptcies (i.e. not

¹⁶The i th moment of power law distributions with exponent α exists (is finite) if and only if $i \leq \alpha - 1$. Consequently, the mean exists for most of the cases observed in our study (different from some findings reported in the literature, see table 1), those with $\alpha \geq 2$, while the variance and higher moments are almost never finite. If the distribution was not a power law (we found no clear dominance between power law and lognormal fits for instance), this would be different.

taking into account ownership and lending interconnections, resulting potential bankruptcy cascades and systemic risk¹⁷⁾ will follow the same distribution if there is no additional connection between firm size and likelihood of bankruptcy. In this case, higher exponents will increase the predictability and, by extension, controllability of disastrous bankruptcies and probably (depending on the structure of the ownership and lending network) also that of systemic risk.¹⁸

The nature of the distribution as a power law, that we were able to establish plausibly, indicates as such that the shape of the distribution is likely invariant to scaling including aggregation and disaggregation as long as these operations remain unbiased.¹⁹

Although we were able to shed light on systematic regional differences in the characteristics of the distribution, the question of the origin of the power law shape remains unanswered as does the related question for the mechanism behind regional differences. Drawing on the literature discussed in section 2, we will now address potential explanations.

5.2. Possible Explanations of the Observed Pattern

As seen in figures 13 through 15, the Eastern coastal regions (that tend to have higher exponents) also differ markedly from Central and Western China in terms of growth rates and firm age, which suggests a possible connection, e.g. that the effect of economic development on the firm size distribution might be mediated by either growth or firm age or both.

Both of these effects are theoretically possible (as are many others) and can be modeled.

First, higher exponents may result from comparatively higher growth in the small and medium sized firms (low and middle part of the distribution) while the distribution maintains the characteristic overall shape as a result of one of the mechanisms discussed in section 2. The reason for such a

¹⁷The distribution resulting from this would be more skewed with even higher likelihood of tail events.

¹⁸This argument is particularly valid for the firm size measured in pay-in capital and perhaps gross profit (since it may indicate a higher capacity to react to liquidity problems), less so for the number of employees.

¹⁹As mentioned above, the failure to maintain the power law under sectoral disaggregation in some studies has cast doubt on the nature of the distribution; it is, however, possible that sectoral disaggregation is not always unbiased while regional disaggregation apparently is (at least for the scales we study).

process may lie in direct stimulation of growth in smaller firms, in better R&D performance of smaller firms, in the removal of market entry barriers, or in greater ease of doing business for smaller enterprises and startups. Any or all of these may be associated with the transformation to the market economy. Consider as a simple computational example a growth process satisfying this setting

$$y = g(x) = k \sqrt[h]{x}$$

(with $h > 1$ and k sufficiently large that $E(y) > E(x)$) acting on power law distributed firm sizes

$$p(x) = Cx^{-\alpha}.$$

The resulting distribution is

$$p(y) = \frac{dg^{-1}}{dy} p(g^{-1}) = Chk^{h(1-\alpha)} y^{-\alpha h - 1 + h} = \tilde{C} y^{-\alpha h - 1 + h}$$

where $\tilde{C} = Chk^{h(1-\alpha)}$ is the new constant and the new exponent is larger than the old one $\alpha h + 1 - h > \alpha$, see Appendix A.

Second, the reforms led to breakups (even bankruptcies) of some of the large state-owned enterprises which may have eliminated a part of the tail of the distribution²⁰ with smaller firms reacting to this thereby restoring the distribution's power law shape.

Third, if the origin of the power law distribution is indeed the Gibrat process (or the Kesten process), the distribution should yield an exponent $\alpha = 2$. However, the distribution may be subject to disturbances, perhaps caused by growth processes as suggested above, which temporarily create different slopes of the power law (i.e., different exponents) before reverting to the Gibrat exponent. Regions currently or recently in transition may experience stronger disturbances and perhaps also disturbances that systematically work into the same direction. (This would also offer an explanation why all deviations from the Gibrat exponent observed for China were upward while the literature provides numerous examples of deviations in both directions for other countries.)

Fourth, regions that experience higher growth may result in systematically different exponents as a result of more rapid technological progress,

²⁰This is in line with the assessment of the changes in the productivity distribution as given by Yu et al. (2015).

or of different distributions of productivity, firm age, or product diversity (following the models by Zhang et al. (2009), Helpman et al. (2004), Coad (2010), and Buldyrev et al. (2007) respectively).

Fifth, following preferential attachment models of firm size development Dahui et al. (2006); Stephen and Toubia (2009), the reforms may have led to the establishment of new connections, especially by and between smaller firms.

Sixth, the large shifts in the ownership structure associated with market reforms may have had an impact on the size distribution, especially since it is reported (though for more recent years) that dynamic characteristics of Chinese firms are strongly dependent on the ownership structure (Duschl and Peng (2015)).

Finally, firm ages, which were shown to be inversely related to higher power law exponents (steeper firm size distributions). They may have a direct effect (as in Coad (2010)) or may themselves be a result of a quantity that influences the firm ages, such as bankruptcy (which would also be in line with large numbers of township and village enterprises but also of state-owned enterprises that were disestablished in connection with the market reforms). A model of a firm-age mediated effect on the firm size distribution resulting in both a power law and a specific variation of the exponents between the regions as observed empirically could employ the Yule process. An instance of preferential attachment, the Yule process is known to yield power law distributions with exponent $\alpha = 2$ (Newman (2005)). The Yule process starts with a single node and proceeds to add 'children nodes' to randomly (with equal probabilities) selected nodes. The distribution of the number of children follows a power law with $\alpha = 2$. It can be modeled as a Markov chain with steady state conditions (following Newman (2005), with $p(b)$ being the frequency of nodes with b children, n the total number of nodes)

$$(n + 1)p(1) = np(1) - p(1) + 1$$

$$(n + 1)p(b) = np(b) - bp(b) + (b - 1)p(b - 1)$$

It follows $p(1) = 1/2$ and (where \sim gives the approximation for the tail, i.e. for large b)

$$\begin{aligned} p(b) &= \frac{b-1}{b+1}p(b-1) = \frac{2(b-1)!}{(b+1)!}p(1) = \frac{2(b-1)!}{(b-1)!b(b+1)}p(1) \\ &= \frac{2}{b^2+b}p(1) = \frac{2}{b^2+b} \sim b^{-2}. \end{aligned}$$

Consider a (very simple, stylized) model of increasing returns: Every firm establishes its own technology. Other, smaller, firms can adopt this technology which causes the firm to grow. The smaller firms, in turn, will interconnect their technology and the 'parent technology', which means all adopters of their technology will also adopt the 'parent technology' causing the 'parent firm' to grow further. This is an instance of the Yule process and would therefore create a power law with $\alpha = 2$.

To introduce variations in the exponent, consider random bankruptcy of firms. Assume firms that will bankrupt are chosen at random with an average value of z for each firm added to the system by the above Yule process. Further assume, these firms are immediately replaced by a firm of size 1.²¹ Consequently, the above Markov chain becomes

$$(n + 1)p(1) = np(1) - p(1) + 1 + (1 - z)p(1)$$

$$(n + 1)p(b) = np(b) - bp(b) + (b - 1)p(b - 1) - zp(1)$$

which yields $p(1) = 1/(1 + z)$ and

$$p(b) = \frac{\Gamma(b)\Gamma(3 + z)}{\Gamma(b + 2 + z)} \frac{1}{1 + z}$$

which can, as shown in Appendix B, be developed into

$$p(b) = Cb^{-2-z}.$$

That is, a bankruptcy rate z (at least for $z < 1$) will increase the exponent of the resulting power law by order z which can be generalized as the relation of death rate and birth rate. Note, however, that for real firm populations it is likely that birth and death rate are correlated, both increasing as the industry becomes more dynamic while rather static firm populations will have low birth and low death rates. The relation of the two influences the (median) firm age which could thus be used as a proxy. Figure 15 and the correlation of median firm ages with the exponents as cited above underlines this possible explanation.

²¹This ensures that the dynamic is not changed at the level of the total population and that $p(b)$ does not depend on $p(b + 1)$.

6. Conclusion

In the present study, we estimated the power law distribution for firm sized measured in terms of 1) the number of employees, 2) capital, 3) gross profit for the provinces of China for the years 1998 through 2008. The case of mainland China is unique for its stepwise transition to market economy in the 1980s and 1990s; this makes it an example that is worth studying not just in order to understand the firm size distribution in China or to prove yet again that firm sizes are indeed power law distributed, but also in order to investigate the effects behind the emergence of this distribution in virtually every case that was studied. Many scholars (Bak et al. (1988); Mandelbrot and Hudson (2004)) have hypothesized that scale free distribution in itself represents some kind of self-organization; that it indicates a process driven to converge against a distribution by a critical transition or otherwise; a process that likely results from evolutionary mechanisms in the underlying system.

In general agreement with the literature, we found exponents (validated with the KS test and Vuong's test) concentrated between 2.0 and 2.5 (with some few outliers) for the firm size distribution measured in terms of capital and gross profits and slightly higher for the firms size measured in the number of employees. As exponents reported in the literature range from 1.4 to 3.7, it would appear that the less steep part, the lower range of exponents (which ever of the expoanations put forward in section 5 holds) is missing in China.

Comparing the estimated exponents for different regions, a very clear geographical picture emerges. For the Eastern coastal regions, the estimated exponents tend to be higher with moderate levels in the Central part and rather low levels in the Western regions of the country. The same pattern can be recovered from the pattern of GDP growth across the provinces (with the exception of very high growth Inner Mongolia which is, however, mainly resource based). The pattern of median firm ages in 2008 on the other hand allows a very clear illustration of the regional stages of China's economic reforms: Eastern coastal regions (moderate to high firm ages) first, Central provinces (very low firm ages) second (then recent), Western regions afterwards. Higher power law exponents indicate smaller dispersion and a lower distribution mass on extreme tail observations; finding the highest exponents in precisely the economic centers of the country where the largest firms concentrate is therefore not an immediately intuitive but still very significant result. It hints at other effects that balance this and at the role of a mechanism that appears to stabilize the distribution in spite of changing

determinants.

Potential explanations put forward in the section 5 in turn connect the observed patterns to more rapid growth of smaller and medium sized firms as a result of the reforms on the one hand and to bankruptcy patterns on the other hand. The two approaches are not necessarily mutually exclusive; other explanations, involving changes in the ownership structure (away from absolute dominance of state-owned firms), product diversity, or firm age distribution, are conceivable as well.

Of course, any or all of these possible explanations may be connected to the patterns observed in section 4. Independent from the explanation, however, some stylized facts could be established: the firm size distribution in China likely follows a power law with slightly higher exponents than reported for other countries. This continues to hold in the regional firm size distributions where the parameters differ systematically and appear to be connected to growth and firm age.

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Appendix A. Derivation of the transformation $y = k\sqrt[h]{x}$ on the power law distribution

PDFs resulting from continuous invertible transformations $y = g(x)$ on continuous random variables x are given as

$$p(y) = \frac{dg^{-1}}{dy}p(g^{-1})$$

Hence, the transformation $y = g(x) = k\sqrt[h]{x} = kx^{1/h}$ (with $h > 1$ and $k >> 0$) on $p(x) = Cx^{-\alpha}$ yields

$$g^{-1}(y) = \left(\frac{y}{k}\right)^h$$

$$\frac{dg^{-1}}{dy} = hk^{-h}y^{h-1}$$

$$p(y) = \frac{dg^{-1}}{dy}p(g^{-1}) = hk^{-h}y^{h-1}Ck^{\alpha h}y^{-\alpha h} = Chk^{h(1-\alpha)}y^{-\alpha h-1+h} = \tilde{C}y^{-\alpha h-1+h}$$

where $\tilde{C} = Chk^{h(1-\alpha)}$ is the net constant and the new exponent is larger than the old one (if $\alpha > 1$) since

$$\begin{aligned}\alpha h + 1 - h &> \alpha \\ \alpha h - \alpha = \alpha(h - 1) &> h - 1 \\ \alpha &> 1.\end{aligned}$$

Appendix B. Approximation of the distribution $p(b, z)$ of the modified Yule process

The Markov chain

$$(n + 1)p(1) = np(1) - p(1) + 1 + (1 - z)p(1)$$

$$(n + 1)p(b) = np(b) - bp(b) + (b - 1)p(b - 1) - zp(1)$$

yields $p(1) = 1/(1 + z)$ and

$$\begin{aligned}p(b) &= \frac{b-1}{b+1+z}p(b-1) = \frac{\Gamma(b-1+1)\Gamma(2+z+1)}{\Gamma(b+1+z+1)}p(1) = \frac{\Gamma(b)\Gamma(3+z)}{\Gamma(b+2+z)}p(1) \\ &= \frac{\Gamma(b)\Gamma(3+z)}{\Gamma(b+2+z)} \frac{1}{1+z}\end{aligned}$$

Note that we have to use the Gamma function (which is the continuous extension of the factorial with $n! = \Gamma(n + 1)$) since the factorial is not defined over non-integer values. The Markov chain is unfortunately not as conveniently solved as the one above. However, using Stirling's approximation (Dutkay et al. (2012))

$$m! = \Gamma(m + 1) \approx \sqrt{2\pi m} m^{m+0.5} e^{-m} = \sqrt{2\pi m} \left(\frac{m}{e}\right)^m,$$

we obtain (\sim again giving the approximation of the tail, i.e. for large b)

$$\begin{aligned}p(b) &\approx \frac{\sqrt{2\pi(b-1)} \left(\frac{b-1}{e}\right)^{b-1} \sqrt{2\pi(2+z)} \left(\frac{2+z}{e}\right)^{2+z}}{\sqrt{2\pi(b+1+z)} \left(\frac{b+1+z}{e}\right)^{b+1+z}} \frac{1}{1+z} \\ p(b) &= \sqrt{2\pi \frac{(b+1)(2+z)}{b+1+z}} \left(\frac{2+z}{e}\right)^{2+z} e^{-2-z} \frac{1}{1+z} \frac{(b-1)^{b-1}}{(b+1+z)^{b+1+z}} \\ p(b) &= C \frac{(b-1)^{b-1}}{(b+1+z)^{b+1+z}} \sim C b^{(b-1)-(b+1+z)} = C b^{-2-z}.\end{aligned}$$

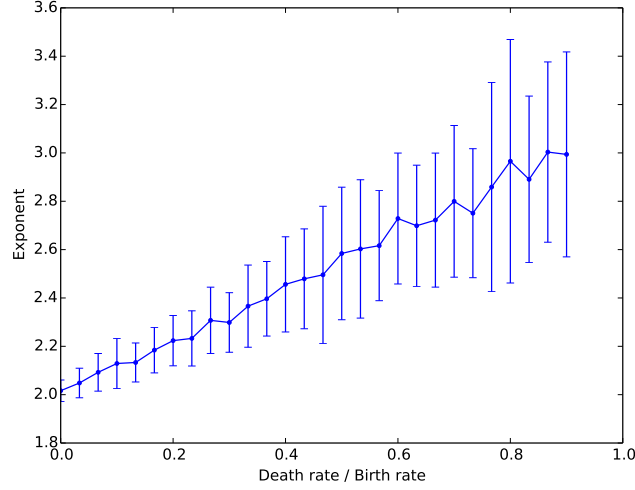


Figure B.16: Exponent estimated for Yule processes with random exit (death rate); averages over 100 simulations for each death/birth rate.

That is, a bankruptcy rate z (at least for $z < 1$) will increase the exponent of the resulting power law by order z . To illustrate this, a simulation of this system for different z is added in figure B.16. Note that z may be seen as a death rate in the resulting process that is contrasted to a birth rate 1 (the velocity of the Yule process itself, the speed at which new nodes are added, 1 per iteration). The resulting property can thus be generalized as $death\ rate / birth\ rate = z / 1 = z$.