UNIVERSITÀ DI PISA

SCUOLA NORMALE SUPERIORE PISA

Size and growth of firms: a statistical approach

Statistical Methods for Data Science A.Y. 2017/18
PROJECT PRESENTATION

Measuring size and growth

- First of all firm size can be measured in many different ways: number of employees, revenue/turnover, sales, etc.
- On long time series it is important to normalize monetary size measures (e.g. revenue, sales) controlling for inflation.
- The relevant variable for growth is either
  - $[S(t)-S(t-1)]/S(t-1)$
  - $\log [S(t)/S(t-1)]$

- The first stylized fact is an extreme heterogeneity of firm size, well described by a **power law (Pareto)** distribution (see figure below from Axtell, **Zipf's distribution** for U.S firm size, Science 2001)
- The qualitative character of such distributions seems independent of how size is defined.

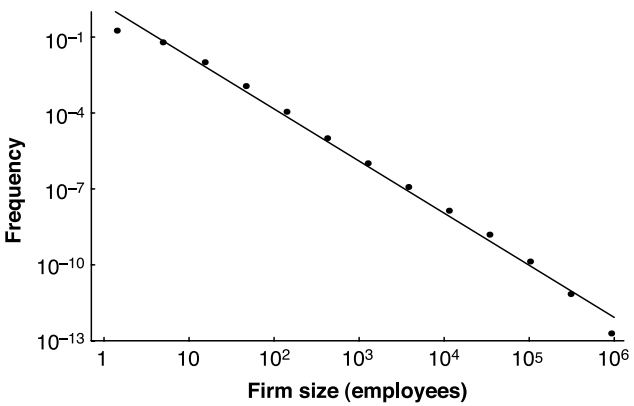

Fig. 1. Histogram of U.S. firm sizes, by employees. Data are for 1997 from the U.S. Census Bureau, tabulated in bins having width increasing in powers of three (*30*). The solid line is the OLS regression line through the data, and it has a slope of 2.059 (SE = 0.054; adjusted $R^2 = 0.992$), meaning that $\alpha = 1.059$; maximum likelihood and nonparametric methods yield similar results. The data are slightly concave to the origin in log-log coordinates, reflecting finite size cutoffs at the limits of very small and very large firms.

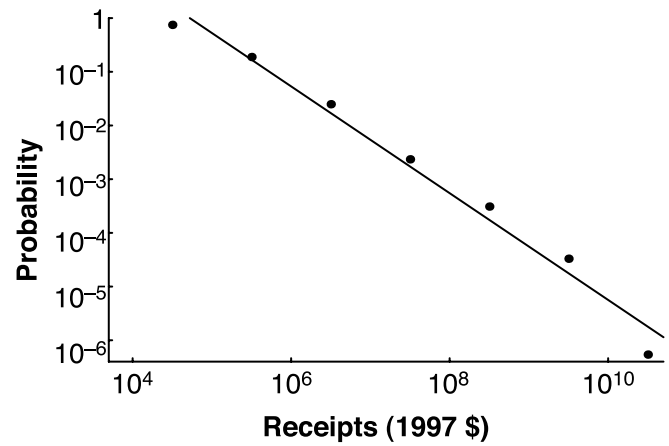

Fig. 2. Tail cumulative distribution function of U.S. firm sizes, by receipts in dollars. Data are for 1997 from the U.S. Census Bureau, tabulated in bins whose width increases in powers of 10. The solid line is the OLS regression line through the data and has slope of 0.994 (SE = 0.064; adjusted $R^2 = 0.976$).

$$\Pr[S \geq s_i] = \left(\frac{s_0}{s_i}\right)^{\alpha}, \; s_i \geq s_0, \; \alpha > 0$$

Robustness for different measures of size

Robustness in time

- Virtually all U.S. firms experienced significant changes: revenues, work force, merger and acquisition activity;

- During the years individual firms migrated up and down the Zipf distribution, but economic forces seem to have rendered any systematic deviations from it short-lived.

Table 3. Theoretical power law exponents for U.S. firms over a 10-year period. Note that even though the number of firms and total employees each increased over this period, as did the average firm size, the value of α was approximately unchanged.

Year	Firms	Employees	Mean firm size	α, from (*4*)
1997	5,541,918	105,299,123	19.00	0.9966
1996	5,478,047	102,187,297	18.65	0.9986
1995	5,369,068	100,314,946	18.68	0.9983
1994	5,276,964	96,721,594	18.33	1.0004
1993	5,193,642	94,773,913	18.25	1.0008
1992	5,095,356	92,825,797	18.22	1.0009
1991	5,051,025	92,307,559	18.28	1.0004
1990	5,073,795	93,469,275	18.42	0.9995
1989	5,021,315	91,626,094	18.25	1.0006
1988	4,954,645	87,844,303	17.73	1.0039

Gibrat's law

$$s_i(t+1) = \gamma(t) s_i(t)$$

- Under the assumptions
  - proportionate growth of a firm in a given period is a random variable independent of the initial firm size
  - statistical independence of successive growths

  Gibrat (1931) concluded that after a long period the logarithmic growth rates are Gaussian distributed and independent of the initial firm's size
- Gibrat's law has been tested empirically, but significant deviations from the normal distribution have been observed

Measuring the firm growth distribution

- Despite the availability of large firm size datasets, single firm models are difficult to test because the number of firms, N, is large, but the number of data points per firm, T, is very small (5-50 points)

- Two approaches used in the literature:
  - assume that the growth time series of each individual firm is a specific realization of the same stochastic process (Model Firm hypothesis)
  - assume that all firms in a balanced panel have the same specific functional form of the growth rate distribution, although the parameters that characterize the distribution may be different from firm to firm (Common distribution hypothesis)

- Gibrat law has been tested empirically (figures taken from Stanley et al., Nature **379**, 804 (1996)) and it has been found that the distribution of firm growth r depends on the size s_0

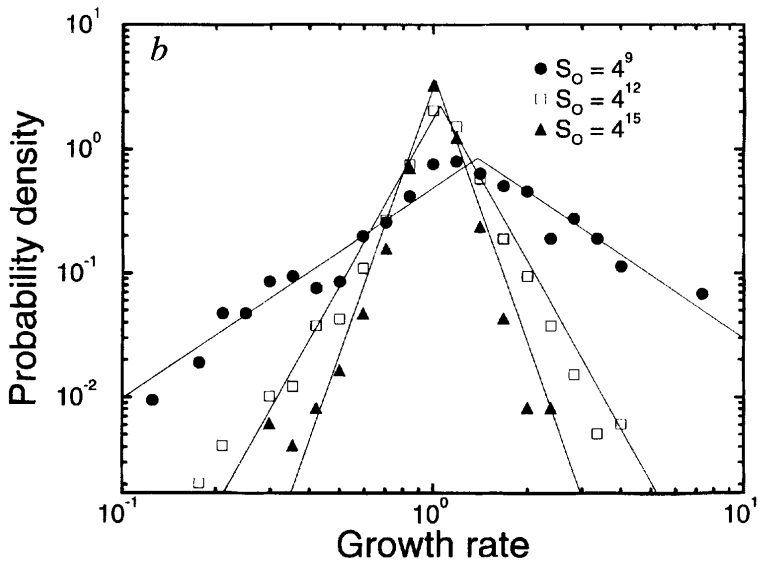

FIG. 1 *a*, Probability density $p(r \mid s_0)$ of the growth rate $r \equiv \ln(S_1/S_0)$ from year 1990 to 1991 for all publicly traded US manufacturing firms in the 1994 Compustat database with standard industrial classification index of 2000–3999. We examine 1991 because between 1992 and 1994 there are several companies with zero sales that either have gone out of business or are 'new technology' companies (developing new products). We show the data for two different bins of initial sales (with sizes increasing by powers of 4): $4^{11.5} < S_0 < 4^{12.5}$ (squares) and $4^{14.5} < S_0 < 4^{15.5}$ (triangles). Within each sales bin, each firm has a different value of *R*, so the abscissa value is obtained by binning these *R* values. The solid lines are fits to equation (1) (in the text) using the mean $\bar{r}(s_0)$ and standard deviation $\sigma(s_0)$ calculated from the data. *b*, Probability density $p(r \mid s_0)$ of the annual growth rate, for three different bins of initial sales: $4^{8.5} < S_0 < 4^{9.5}$ (circles), $4^{11.5} < S_0 < 4^{12.5}$ (squares) and $4^{14.5} < S_0 < 4^{15.5}$ (triangles). The data were averaged over all 16 one-year periods between 1975 and 1991. The solid lines are fits to equation (1) using the mean $\bar{r}(s_0)$ and standard deviation $\sigma(s_0)$ calculated from all data.

Tent shape

- Firm growth rates follow a Laplace distribution

$$p(r|s_0) = \frac{1}{\sqrt{2}\sigma(s_0)} \exp\left(-\frac{\sqrt{2}|r-\bar{r}(s_0)|}{\sigma(s_0)}\right)$$

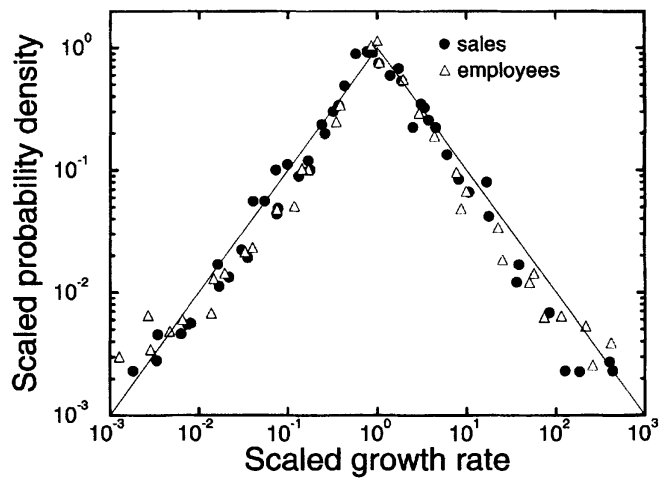

FIG. 3 Scaled probability density $p_{scal} \equiv 2^{1/2}\sigma(s_0)p(r \mid s_0)$ as a function of the scaled growth rate $r_{scal} \equiv 2^{1/2}[r - \bar{r}(s_0)]/\sigma(s_0)$ of sales (circles). The values were rescaled using the measured values of $\bar{r}(s_0)$ and $\sigma(s_0)$. Also we show (triangles) the analogous scaled quantities for the number of employees. All the data collapse upon the universal curve $p_{scal} = \exp(-\mid r_{scal} \mid)$ (solid line) as predicted by equations (1) and (2).

Subbotin family of distributions

- As specific distribution to test: consider the Subbotin family

$$p(r) = \frac{1}{2\gamma\beta^{1/\beta}\Gamma(1+1/\beta)} \exp\left(-\frac{1}{\beta}\left|\frac{r-\mu}{\gamma}\right|^{\beta}\right)$$

where μ is the mean, β characterizes the shape (kurtosis decreases with beta) and the standard deviation is

$$\sigma = \gamma\beta^{1/\beta}\sqrt{\frac{\Gamma(3/\beta)}{\Gamma(1/\beta)}}$$

- It includes the Laplace (β=1) and the Gaussian (β=2)

- The standard deviation $\sigma(S_0)$ depends on the initial size

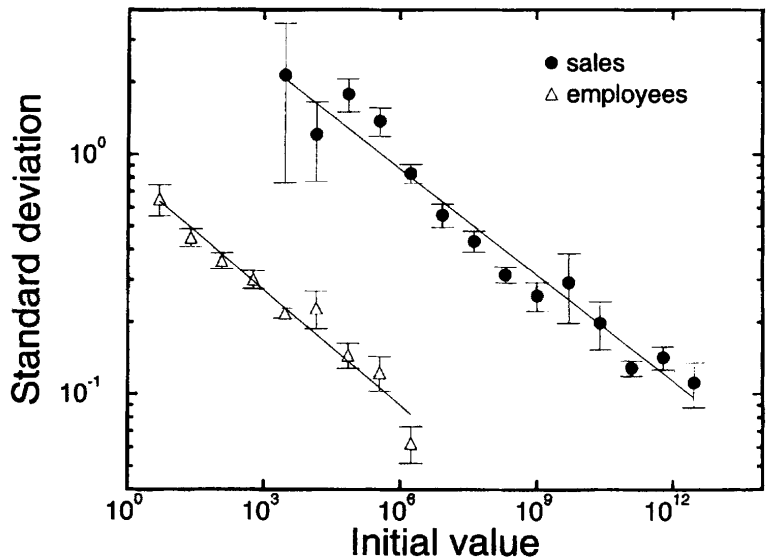

FIG. 2 Standard deviation of the one-year growth rates of the sales (circles) and of the one-year growth rates of the number of employees (triangles) as a function of the initial values. The solid lines are least-square fits to the data with slopes $\beta = 0.15 \pm 0.03$ for the sales and $\beta = 0.16 \pm 0.03$ for the number of employees. We also show error bars of one standard deviation about each data point. These error bars appear asymmetric as the ordinate is a log scale.

More recent empirical works have shown that

- the distribution is slightly asymmetric,
- the extreme tails are fatter than exponential,
- successive growth rates are slightly correlated,
- different sectors and sub-sectors of the economy can have different growth rates and therefore some of the above results might be driven by heterogeneity.

Subsectors

- We considered panels of firms which are homogeneous at the sub-sector level.
  - For the European Union (Amadeus)
    - Chemical Manufacturing (code 325)
    - Computer and Electronic Product Manufacturing (code 334)
    - Food Manufacturing (code 311).
  - For the US (Compustat)
    - Chemical Manufacturing (code 325)
    - Computer and Electronic Product Manufacturing (code 334)
    - Machinery Manufacturing (code 333).
- Since January 2008 Istat has adopted the new Ateco 2007 classification of economic activities for Italian firms

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Some questions about firms size (1)

- Quantifying the relation (correlation) between different measures of the firm size;
- Finding the family that best describes the size distribution of Italian firms. Use non-parametric (histograms and kernels) and parametric estimates (with maximum likelihood);
- Hypothesis test for alternative shapes.
- Is the size distribution of Italian firms at a given year with power law tails? Has the exponent changed in time?
- Do answers to the previous questions depend on geographical and/or sectorial heterogeneity ?

Qualche domanda sulla size (1)

- Quantificare la relazione (correlazione) tra le diverse possibili misure di size
- Quale famiglia descrive meglio la distribuzione della size delle aziende italiane? Usare stime non parametriche (istogrammi e kernel) e parametriche (con maximum likelihood)
- Test di ipotesi per forme alternative
- La distribuzione della size in un certo anno delle aziende italiane è con coda a legge di potenza? L'esponente è cambiato col tempo?
- Le risposte alle domande sopra dipendono dall'eterogeneità settoriale e/o geografica?

Some questions about firms growth (2)

- Does the Gibrat law hold for Italian firms?
- Finding the family that best describes the growth distribution of Italian firms. Use non-parametric (histograms and kernels) and parametric estimates (with maximum likelihood);
- Hypothesis test for alternative shapes.
- Is the mean growth statistically different from zero in each year?
- Is the distribution symmetric or asymmetric?
- Is there any relation between growth variance and size?
- Do answers to the previous questions depend on geographical and/or sectorial heterogeneity ?

Qualche domanda sulla crescita (2)

- La legge di Gibrat vale per le aziende italiane?
- Quale famiglia descrive meglio la distribuzione della crescita delle aziende italiane? Usare stime non parametriche (istogrammi e kernel) e parametriche (con maximum likelihood)
- La crescita media è statisticamente diversa da zero in ciascun anno?
- La distribuzione è asimmetrica?
- Esiste una relazione tra varianza della crescita e size?
- Test di ipotesi per forme alternative
- Le risposte alle domande sopra dipendono dall'eterogeneità settoriale e/o geografica?

Some questions about firms growth in time (2)

- Has the mean growth changed in time? (Hypothesis test)
- Has the growth distribution changed in time? (Hypothesis test)
- How yearly growths can be compared with growths on longer periods of time (bi-yearly, quinquennial)?
- Is it possible to measure a dependence between growth in subsequent years? (growth predictability)
- Do answers to the previous questions depend on geographical and/or sectorial heterogeneity?

Qualche domanda sulla crescita nel tempo (3)

- La crescita media è cambiata col tempo? (test di ipotesi)
- La distribuzione della crescita è cambiata nel tempo? (test di ipotesi)
- Come si confrontano crescite annuali con crescite su periodi più lunghi (ad esempio biennali o quinquennali)?
- Le risposte alle domande sopra dipendono dall'eterogeneità settoriale e/o geografica?
- Si può misurare una dipendenza tra la crescita in anni successivi? (predicibilità della crescita)