ECON641 – Problem Set 2

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1 The Firm Size Distribution

1.1 Power law in firm size

A random variable S follows a power law if

$$Pr[S > s] = Cs^{-\zeta}, \text{ with } C, s > 0$$

$$\implies \log \Pr[S > s] = \log C - \zeta s.$$
(1)

Note that Zipf's law refers to a power law distribution with exponent $\zeta \approx 1$. Recent research has shown that the distribution of firm size is approximately described by Zipf's law. Below, we're going to explore the application of these ideas to Compustat data.

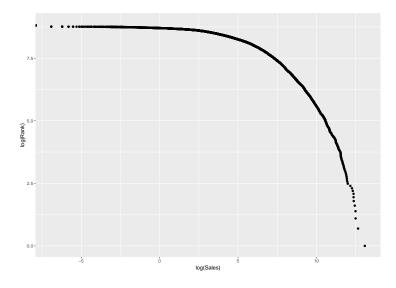
Before I get into the data work, let's briefly try to understand why we're doing all of this logrank/log-size stuff. Suppose S follows a power law as in (1). Then, draw N realizations and rank them in descending order $S_{(1)} > S_{(2)} > ... > S_N$. Because of the ranking, we get $i/N = 1 - F(S_{(i)})$ (since $F(S_{(i)})$ is distributed standard uniform). Thus, $i/N = CS_{(i)}^{-\zeta}$, or

$$\begin{aligned} \operatorname{Rank} &= NCS_{(i)}^{-\zeta} \\ \Longrightarrow & \log \operatorname{Rank} = \operatorname{constant} - \zeta \log S_{(i)}. \end{aligned}$$

1.1.1 Firm sales

First, I plot the log-log plot of rank vs. sales for all firms in the Compustat database for 2015. Clearly, the relationship is not linear, suggesting that the firm size distribution does not follow a power law. This result is consistent with Stanley, et. al. (1995), who also use Compustat data. Note that Axtell (2001) uses US Census data (which obviously contains a far more comprehensive sample of firms compared to Compustat) and finds that the firm size distribution does indeed follow Zipf's law.

Figure 1: Sales Distribution of All Firms in Compustat in 2015



Next, I plot the log-log sales plots for the top 500 and 100 firms in the Compustat database. In both cases, there seems to be a linear relationship, suggesting that the tail of the firm sales distribution follows a power law.

(a) Top 500 firms

(b) Top 100 firms

Figure 2: Sales Distribution for Top 500 and Top 100 Firms in 2015

Next, I estimate the power law coefficient ζ using the estimator proposed by Gabaix and Ibragimov (2011) for the samples of firms above. That is, I estimate the model

$$\log(\operatorname{Rank}_i - 1/2) = \operatorname{constant} - \zeta \log S_i + \epsilon_i$$

using OLS. I compute standard errors as in Gabaix and Ibragimov (2011). Table 1 shows the estimation results – suggesting that the sales distribution of the top 500 firms is approximately follows Zipf's law.

Table 1: Estimated Power Law Coefficients for Firm Sales in 2015

For	For different lower size cutoffs				
	All firms	Top 500	Top 100		
$\hat{\zeta}$	-0.27	-1.28	-2.06		
Std. err.	0.00	0.08	0.29		

- 1.1.2 Firm sales by industry
- 1.1.3 Employment
- 1.1.4 Employment by industry
- 1.1.5 Robustness check: 1985 data