

# Power Laws

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НАЦИОНАЛЬНЫЙ ИССЛЕДОВАТЕЛЬСКИЙ  
УНИВЕРСИТЕТ

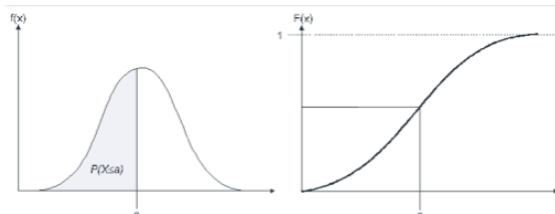
# Continues distribution

- Continues random variable  $X$
- Probability density function (PDF)  $f(x)$ :

$$Pr(a \leq X \leq b) = \int_a^b f(x)dx$$

- $f(x) \geq 0$ ,  $\int_{-\infty}^{\infty} f(x)dx = 1$
- Cumulative distribution function (CDF)

$$F(x) = Pr(X \leq x) = \int_{-\infty}^x f(y)dy$$



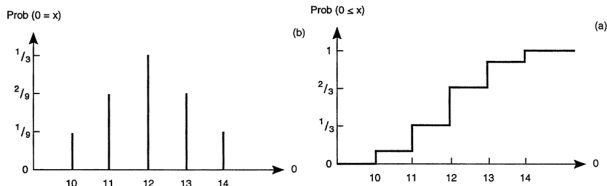
# Discrete distribution

- Discrete random variable  $X_i$
- Probability mass function (PMF)  $p(x)$ :

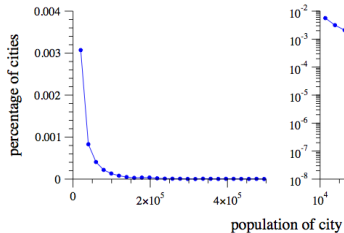
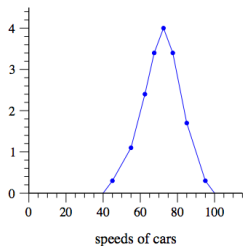
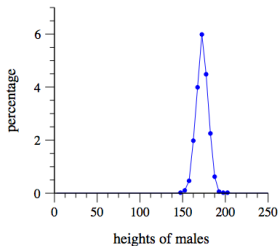
$$p(x) = Pr(X_i = x)$$

- $p(x) \geq 0, \sum_x p(x) = 1$
- Cumulative distribution function (CDF)

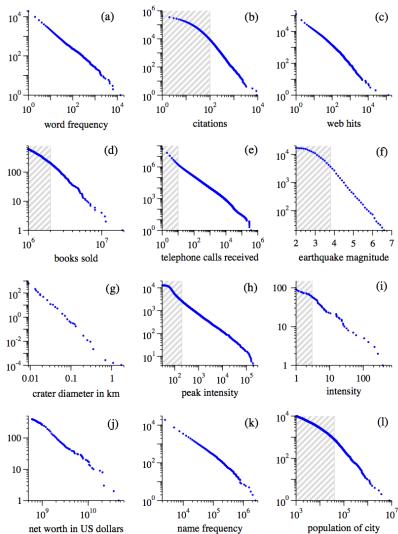
$$F(x) = Pr(X_i \leq x) = \sum_{y \leq x} p(y)$$



# Empirical distributions



# Empirical distributions



## Continues approximation

- Power law

$$p(x) = Cx^{-\alpha} = \frac{C}{x^\alpha}, \quad \text{for } x \geq x_{min}$$

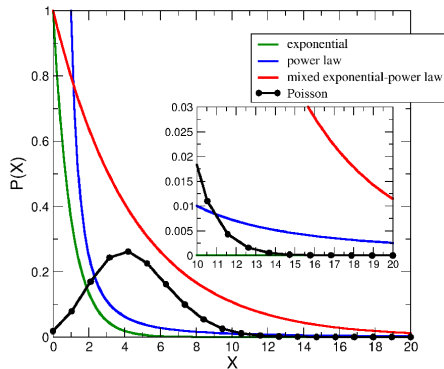
- Normalization ( $\alpha > 1$ )

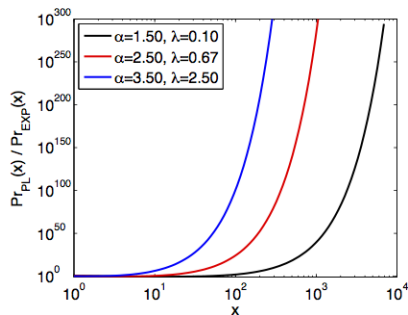
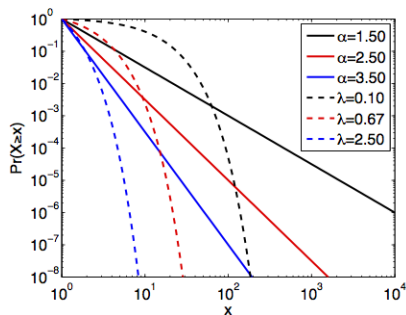
$$1 = \int_{x_{min}}^{\infty} p(x) dx = C \int_{x_{min}}^{\infty} \frac{dx}{x^\alpha} = \frac{C}{\alpha - 1} x_{min}^{-\alpha+1}$$

$$C = (\alpha - 1)x_{min}^{\alpha-1}$$

- Power law PDF

$$p(x) = \frac{\alpha - 1}{x_{min}} \left( \frac{x}{x_{min}} \right)^{-\alpha}$$





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

$$\log p(x) = \log C - \alpha \log x$$



- PDF

$$p(x) = \frac{C}{x^\alpha}$$

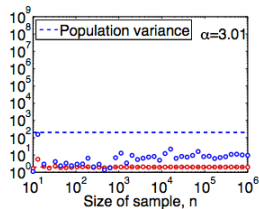
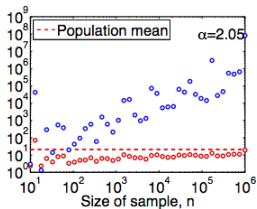
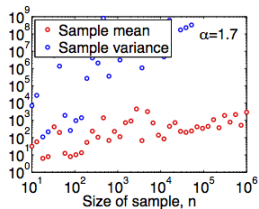
- mean ( $\alpha > 2$ )

$$\langle x \rangle = \int_{x_{\min}}^{\infty} xp(x)dx = C \int_{x_{\min}}^{\infty} \frac{dx}{x^{\alpha-1}} = \frac{\alpha-1}{\alpha-2} x_{\min}$$

- standard deviation ( $\alpha > 3$ )

$$\langle x^2 \rangle = \int_{x_{\min}}^{\infty} x^2 p(x)dx = C \int_{x_{\min}}^{\infty} \frac{dx}{x^{\alpha-2}} = \frac{\alpha-1}{\alpha-3} x_{\min}^2$$

# Moments



Clauset et.al, 2009

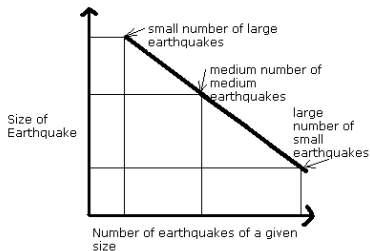
# Scale invariance

- Scaling of the density

$$p(bx) = C(bx)^{-\alpha} = Cb^{-\alpha}x^{-\alpha} \sim p(x)$$

- Scale invariance

$$\frac{p(10x_2)}{p(10x_1)} = \frac{p(x_2)}{p(x_1)}$$



- Cumulative distribution function CDF

$$F(x) = Pr(X \leq x) = \int_{-\infty}^x p(y)dy = \int_{x_{\min}}^x p(y)dy$$

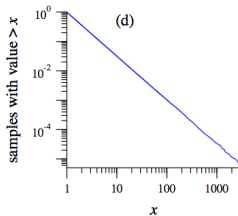
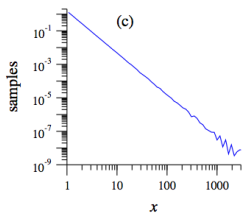
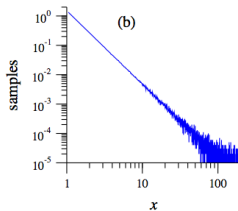
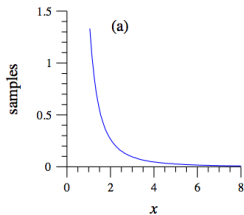
- Complimentary cumulative distribution function cCDF

$$\bar{F}(x) = 1 - F(x) = Pr(X > x) = \int_x^{\infty} p(y)dy$$

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

$$\bar{F}(x) = \frac{C}{\alpha - 1} x^{-(\alpha-1)} = \left( \frac{x}{x_{\min}} \right)^{-(\alpha-1)}$$

# Power law histograms



Newman et.al, 2005

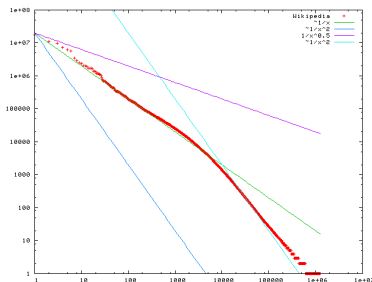
# Zips'f law

Zipf's law - the frequency of any word is inversely proportional to its rank in the frequency table: words are sorted by frequency in decreasing order.

- word frequency

$$f(k) \sim 1/k^s$$

- cCDF  $\bar{F}(x) = Pr(X > x)$ : for n-th word, there are n words with higher frequencies sorted before it
- Rank-frequency plot: rank(frequency) , x- frequency, y-rank



# Node degree distribution

- Node degrees  $k_i = 1, 2, \dots, k_{\max}$
- Degree distribution  $P(k) \equiv P(k_i = k)$

$$P(k) = \frac{n_k}{n} = \frac{n_k}{\sum_k n_k}$$

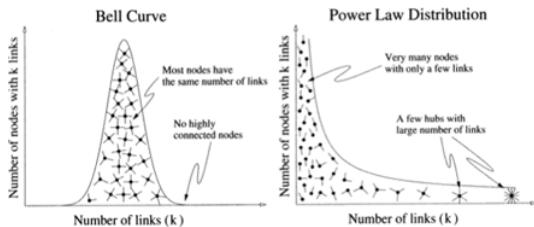
- Power law

$$P(k) = ck^{-\gamma} = \frac{c}{k^\gamma}$$

- Logarithmic coordinates

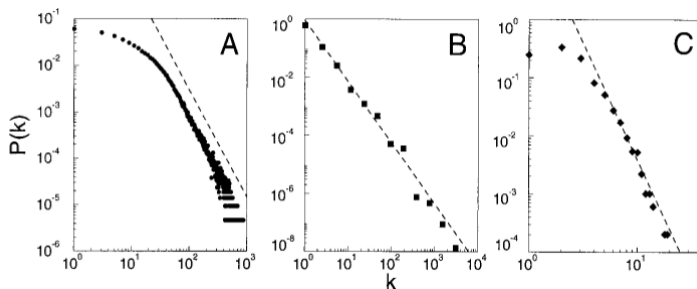
$$\log(P(k)) = -\gamma \log k + \log c$$

# Power law networks





# Power law networks



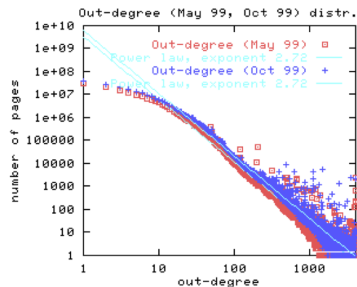
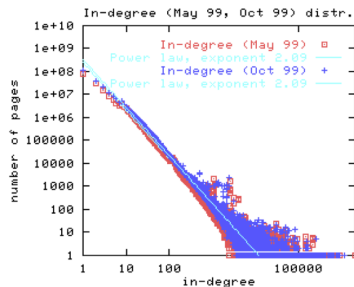
Actor collaboration graph,  $N=212,250$  nodes,  $\langle k \rangle = 28.8$ ,  $\gamma = 2.3$

WWW,  $N = 325,729$  nodes,  $\langle k \rangle = 5.6$ ,  $\gamma = 2.1$

Power grid data,  $N = 4941$  nodes,  $\langle k \rangle = 5.5$ ,  $\gamma = 4$

Barabasi et.al, 1999

# Power law networks



In- and out- degrees of WWW crawl 1999

Broder et.al, 1999

# Parameter estimation: $\alpha$

Maximum likelihood estimation of parameter  $\alpha$

- Let  $\{x_i\}$  be a set of  $n$  observations (points) independently sampled from the distribution

$$P(x_i) = \frac{\alpha - 1}{x_{\min}} \left( \frac{x_i}{x_{\min}} \right)^{-\alpha}$$

- Probability of the sample

$$P(\{x_i\}|\alpha) = \prod_i^n \frac{\alpha - 1}{x_{\min}} \left( \frac{x_i}{x_{\min}} \right)^{-\alpha}$$

- Bayes' theorem

$$P(\alpha|\{x_i\}) = P(\{x_i\}|\alpha) \frac{P(\alpha)}{P(\{x_i\})}$$

- log-likelihood

$$\mathcal{L} = \ln P(\alpha | \{x_i\}) = n \ln(\alpha - 1) - n \ln x_{\min} - \alpha \sum_{i=1}^n \ln \frac{x_i}{x_{\min}}$$

- maximization  $\frac{\partial \mathcal{L}}{\partial \alpha} = 0$

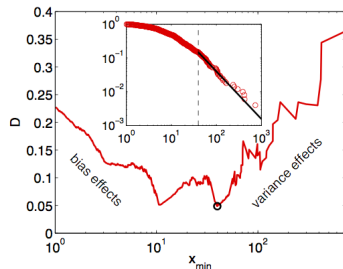
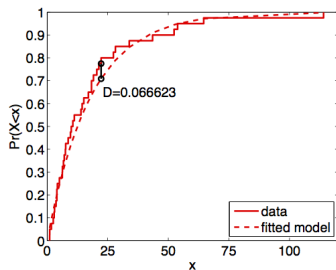
$$\alpha = 1 + n \left[ \sum_{i=1}^n \ln \frac{x_i}{x_{\min}} \right]^{-1}$$

- error estimate

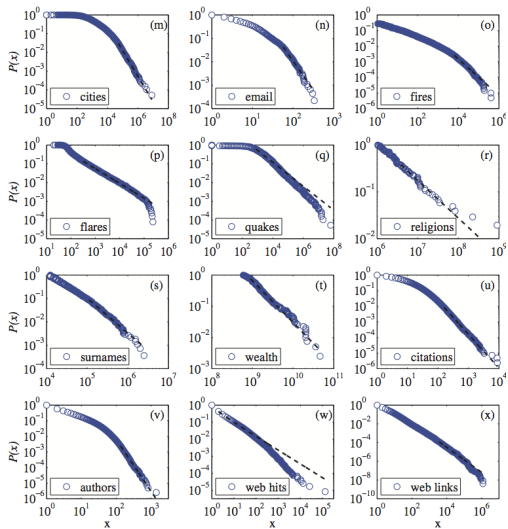
$$\sigma = \sqrt{n} \left[ \sum_{i=1}^n \ln \frac{x_i}{x_{\min}} \right]^{-1} = \frac{\alpha - 1}{\sqrt{n}}$$

- Kolmogorov-Smirnov test (compare model and experimental CDF)

$$D = \max_x |F(x|\alpha, x_{min}) - F_{exp}|$$



# Empirical models



# Example

Word count:

6187267 the

4239632 be

3093444 of

2687863 and

2186369 a

1924315 in

1620850 to

.....

801 incredibly

801 historically

801 decision-making

800 wildly

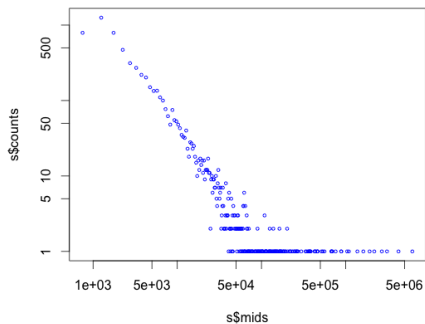
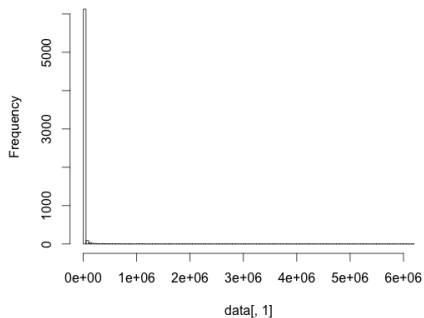
800 reformer

800 quantum

800 considering

# Example

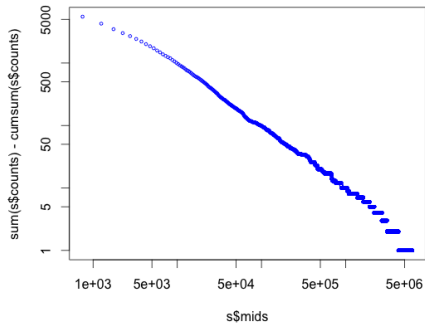
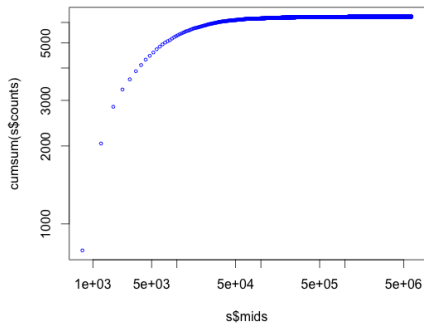
## Histogram (PDF)





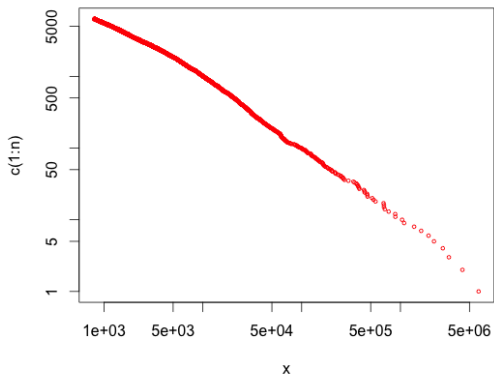
# Example

## CDF and cCDF



# Example

## Rank-Frequency plot



- Power laws, Pareto distributions and Zipf's law, M. E. J. Newman, Contemporary Physics, pages 323–351, 2005.
- Power-Law Distribution in Empirical Data, A. Clauset, C.R. Shalizi, M.E.J. Newman, SIAM Review, Vol 51, No 4, pp. 661-703, 2009.
- A Brief History of Generative Models for Power Law and Lognormal Distributions, M. Mitzenmacher, Internet Mathematics Vol 1, No 2, pp 226-251.