

Power laws

Definition, properties and how to plot them

Power-law degree distributions

- The degree distribution of many networks follows a *power-law distribution*:

- $$p(x) = C \cdot x^{-k}$$

- The proportion of nodes of degree x is (inversely) polynomial in x

Power-law distributions

- In general, a discrete power-law random variable X with parameter k is defined by
- $$Pr(X = x) = C \cdot x^{-k} \quad \text{for all } x > x_{\min}$$
- A continuous power-law random variable X with parameter k is defined by a probability density function
- $$p(x) = C \cdot x^{-k} \quad \text{for all } x > x_{\min}$$

“Scale-free” networks

- A power-law distribution is also called *scale-free*, since scaling the input does not change the form of the function:

- $$p(a \cdot x) = C \cdot a^{-k} \cdot x^{-k} = C' \cdot x^{-k}$$

The long tail

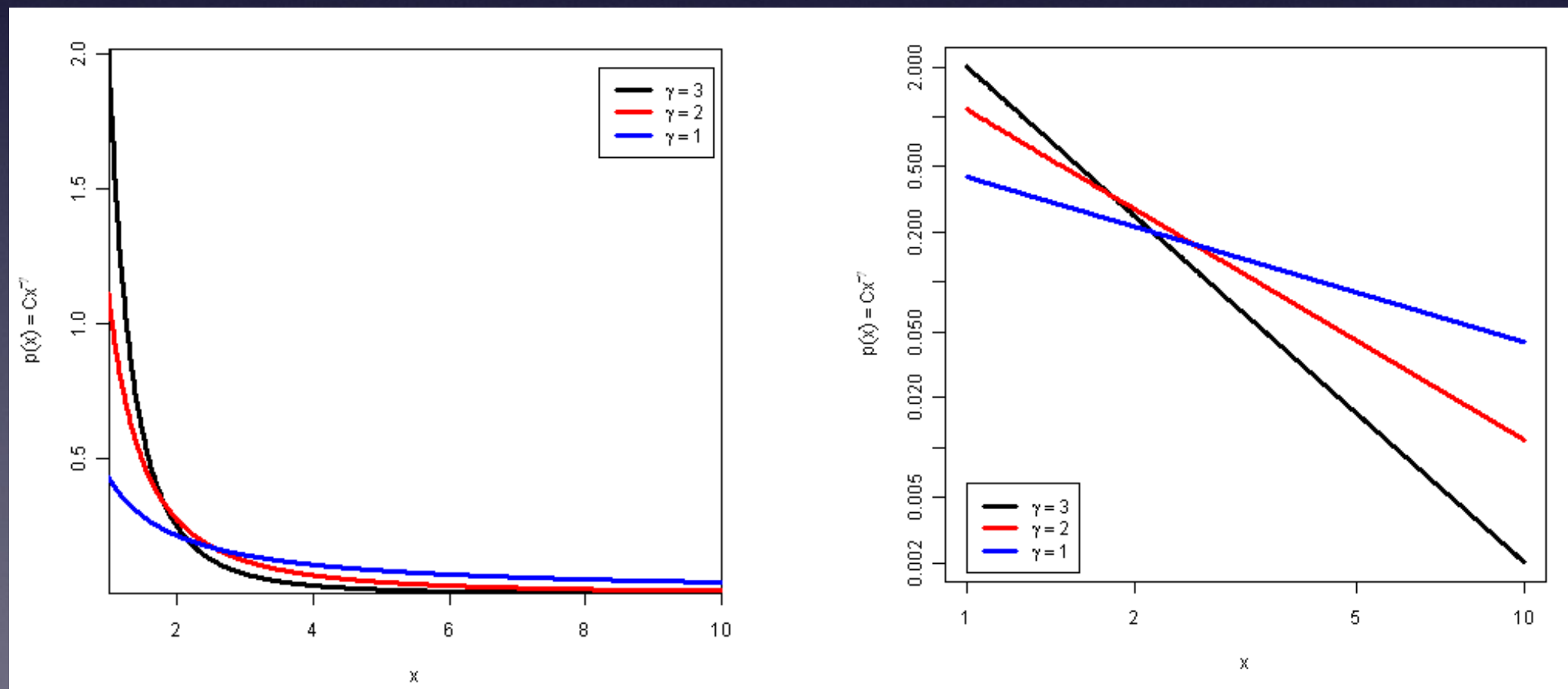
- A power-law distribution has a *long tail*: there is a relatively large number of samples of the population for which the variable has a large value



Power-law plotting: scales

- If we use a log-log scale, a power-law will show up as a *line*, because

- $$\log p(x) = \log C - k \cdot \log x$$



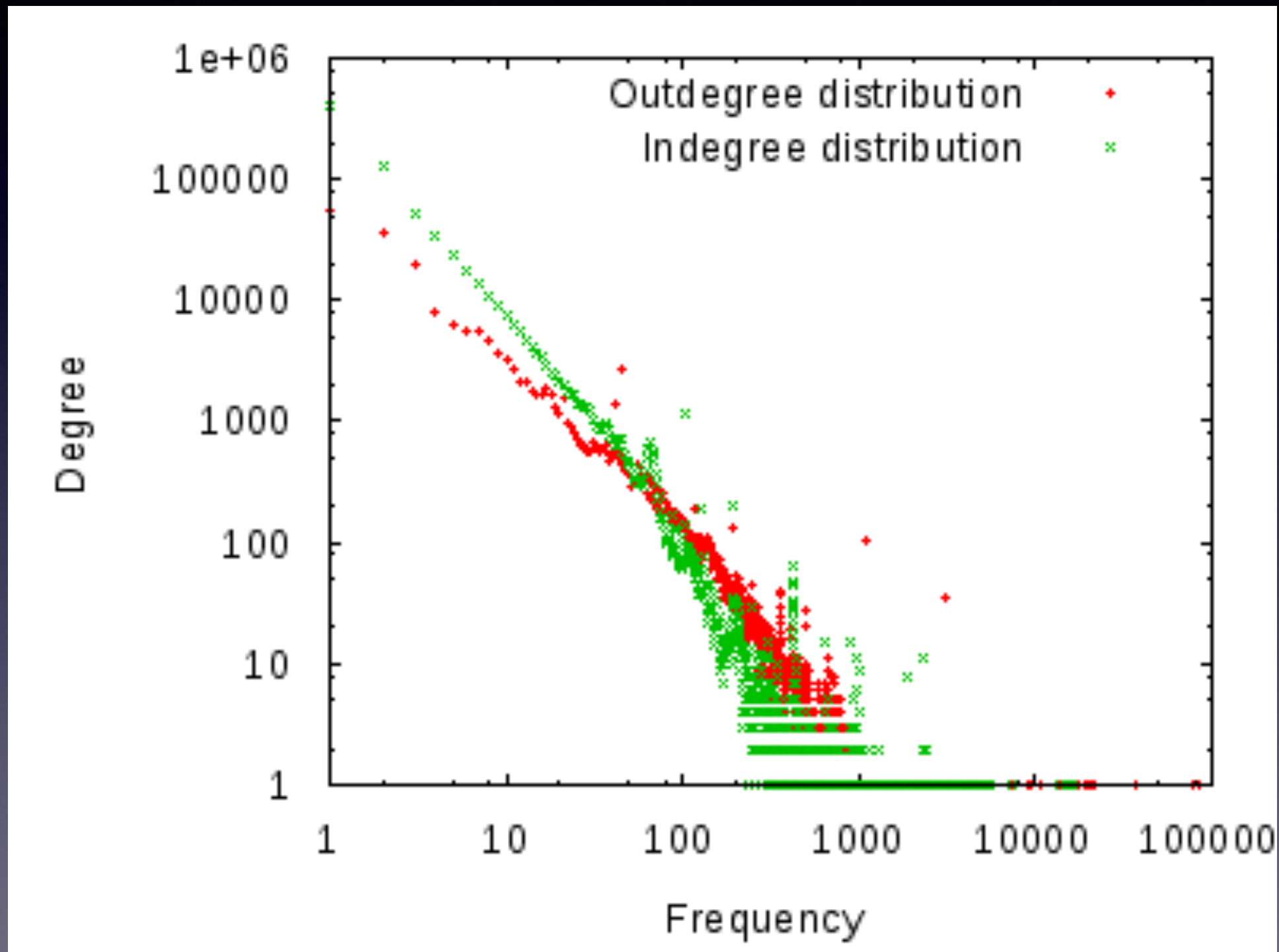
The P.D.F. and the C.D.F.

- Probability distribution function (P.D.F.): $p(x)$
- Cumulative distribution function (C.D.F.):

$$P(x) = \int_x^{\infty} p(x') dx'$$

- If $p(x) = C \cdot x^{-k}$, then $P(x) = C' \cdot x^{-(k-1)}$, and the C.D.F. is also a power-law
- Plotting $P(x)$ should be preferred

Example from web-graph.org



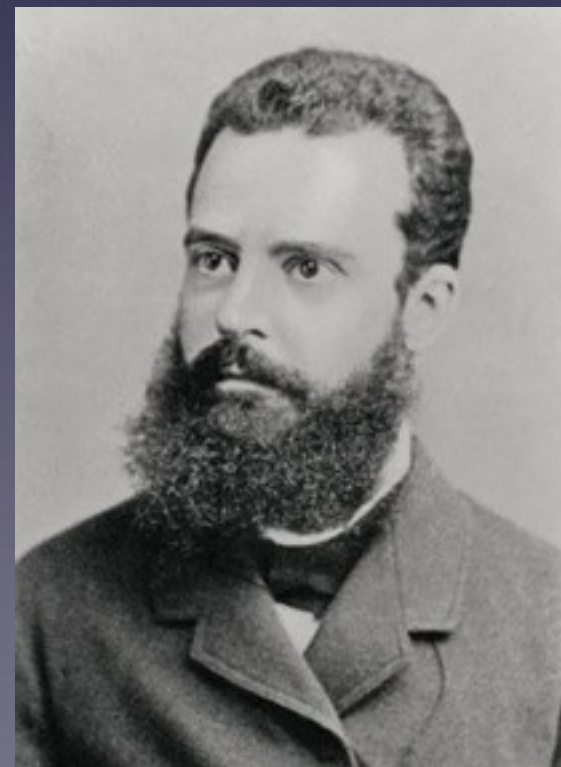
Power-laws everywhere?

- Power-laws don't appear just in degree distributions
- They characterize many very different phenomena
- Also known as *Zipf laws* or *Pareto distributions*

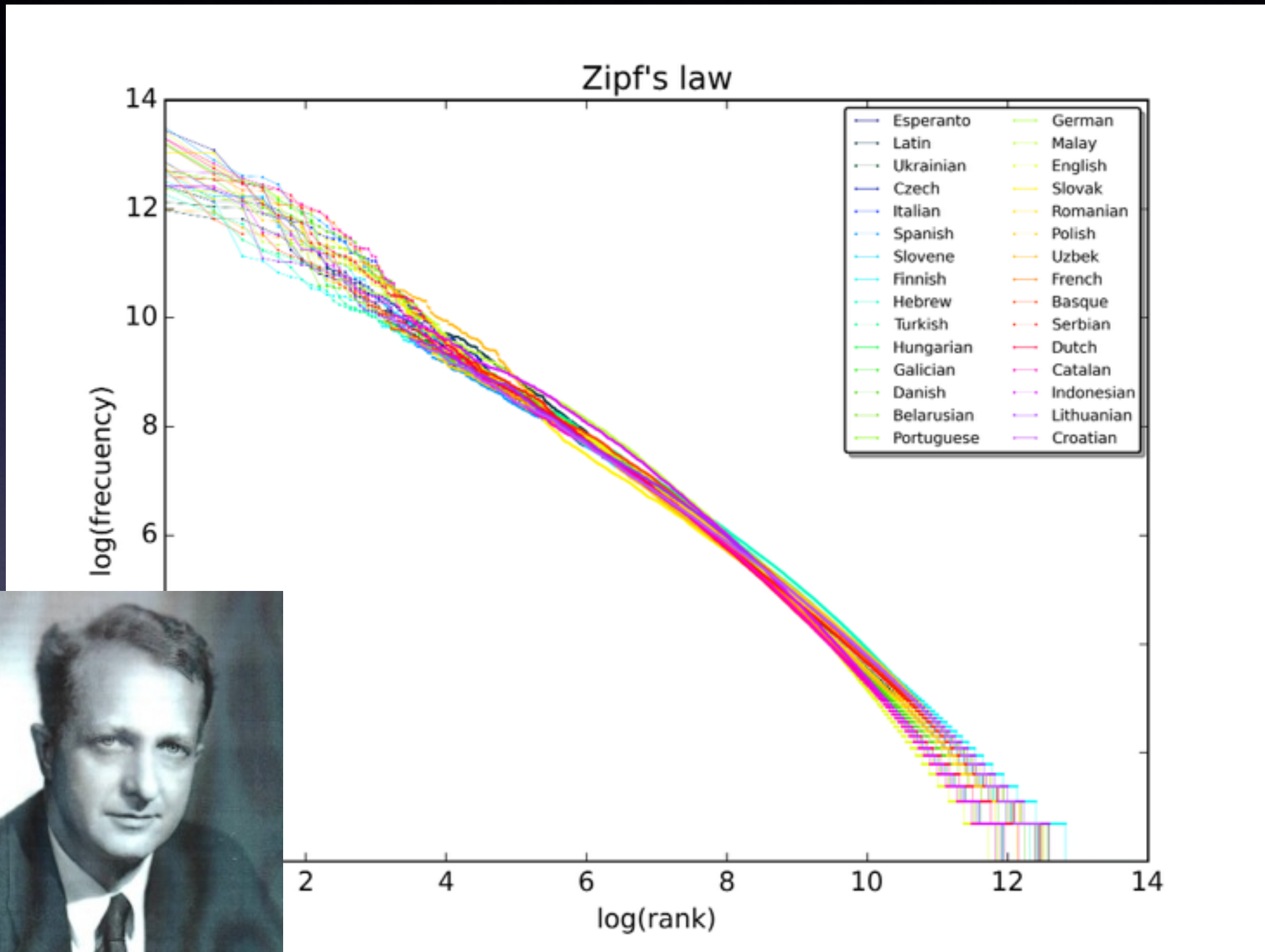
Pareto's distribution of income

- Pareto used available data on income distribution for several countries
- He found that income was well-approximated by a power-law with exponent between 2.32 and 2.72
- “30% of the population owns 70% of the land”

- Vilfredo Pareto (1848-1923),
Italian economist and engineer



Frequency of words - “Zipf’s Law”



George K. Zipf (1902-1950), US linguist

Another example: size of cities

- Let's re-discover by ourselves another power-law: size of Italian towns
- Get the data publicly available from ISTAT's website
- <http://www.istat.it/storage/codici-unita-amministrative/elenco-comuni-italiani.xls>

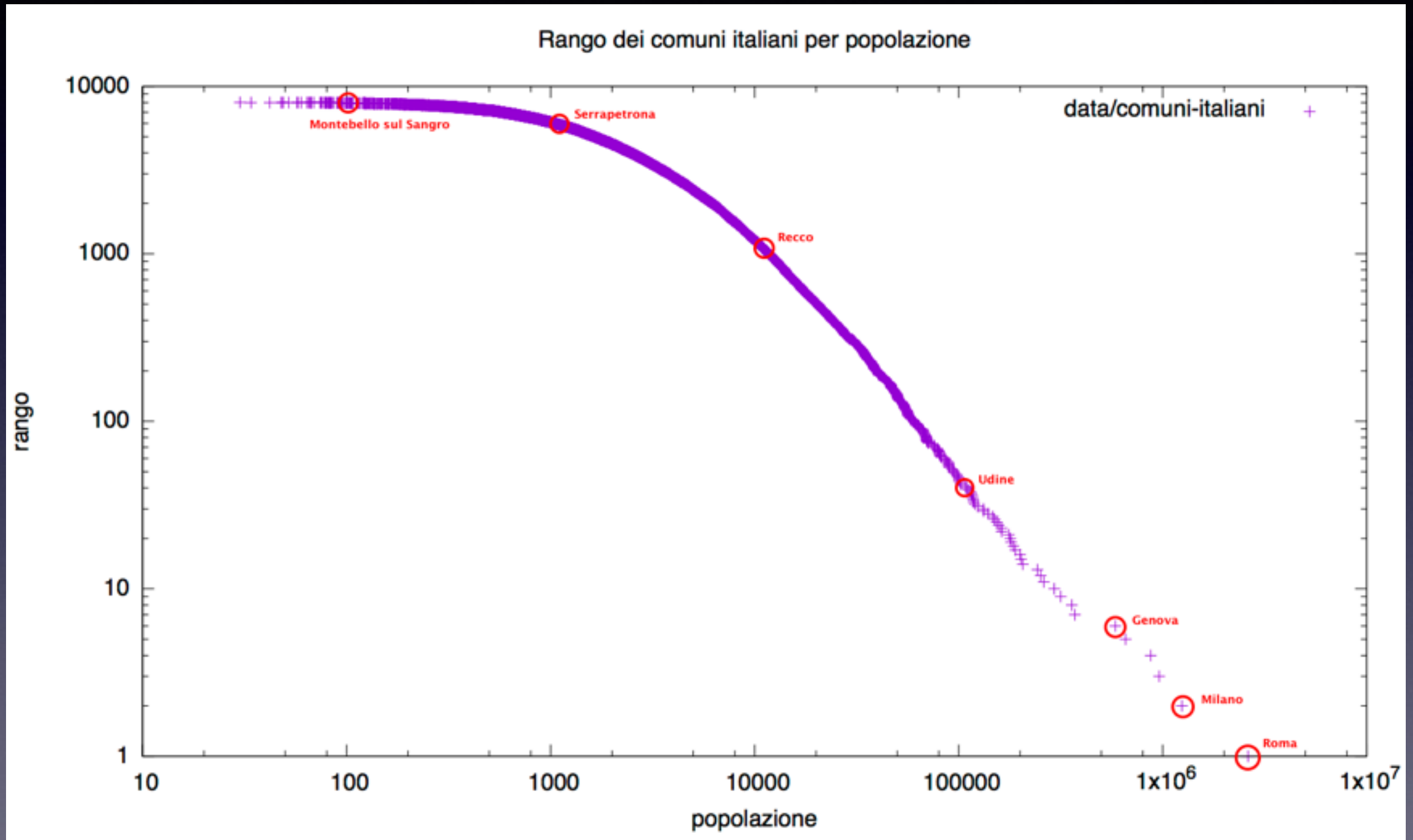
Another example: size of cities

- After processing ISTAT's .xls file, we arrive to a list of Italian city sizes, ordered by size:
 - 2617175 (Rome)
 - 1242123 (Milan)
 - etc

Another example: size of cities

- The *rank* of each item (city) is its *position* in the ordered list
- We plot the rank of each item over the quantity x of interest; in this case, $x =$ population size
- This is equivalent to plotting $P(x)$
- We use a log-log scale
- *Gnuplot* is recommended for best results

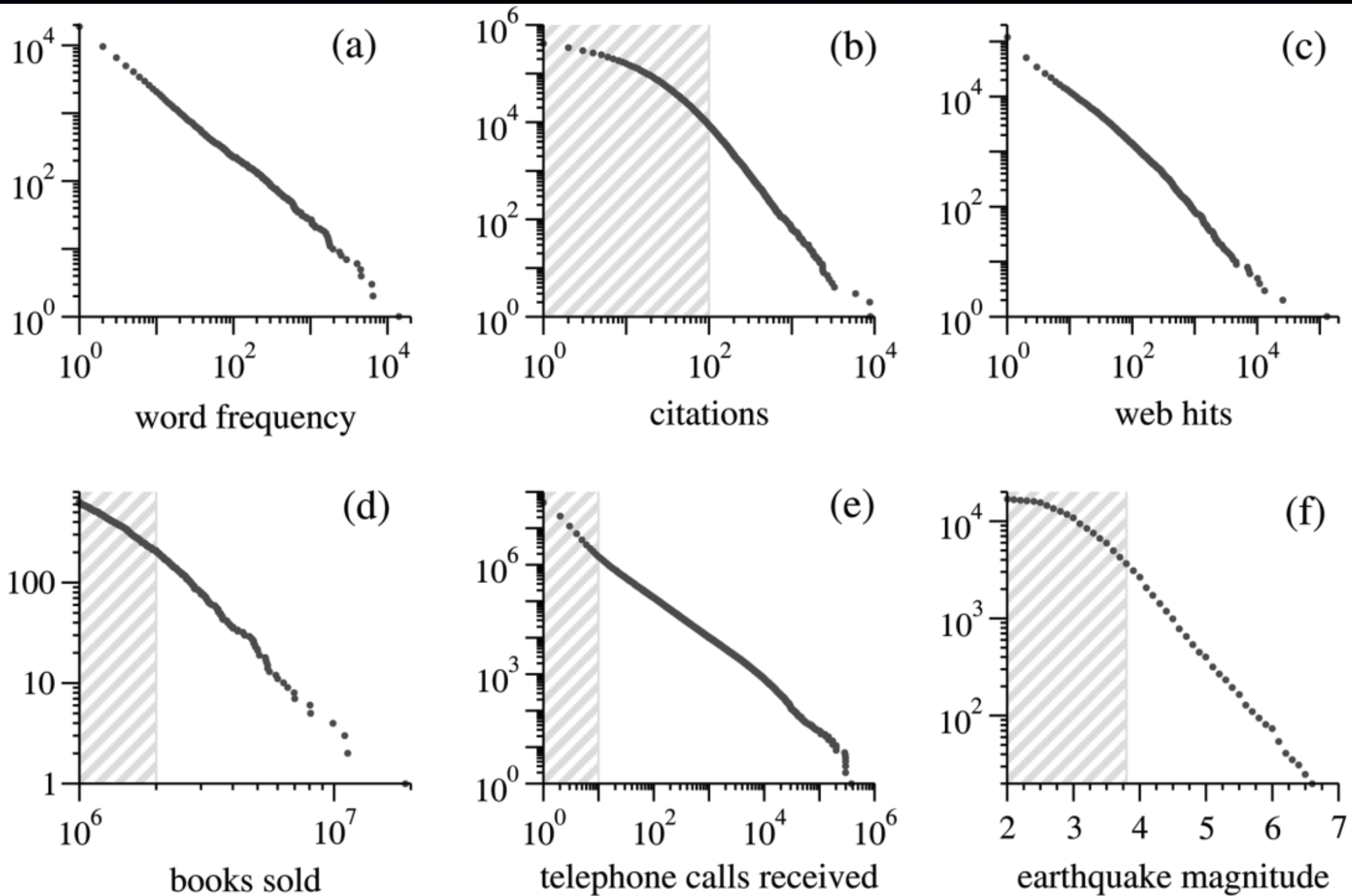
Another example: size of cities



Another example: size of cities

- We find out that the distribution $P(x)$ follows a power-law *only when x is large enough*
- This is a common phenomenon; the power-law relation holds for $x > x_{\min}$
- Thus, a power-law has two main parameters: the power-law exponent (k) and the threshold (x_{\min})

More examples



More examples

Quantity	Threshold (x)	Exponent (k)
(a) Frequency of use of words	1	2.20
(b) Number of citations to papers	100	3.04
(c) Number of hits on web sites	1	2.40
(d) Copies of books sold in the US	2000000	3.51
(e) Telephone calls received	10	2.22
(f) Magnitude of earthquakes	3.8	3.04

Estimating the power-law exponent

- Let x_1, x_2, \dots, x_n be the measured quantities (above x_{\min})
- A reliable estimator (for the discrete case) is

$$k = 1 + n \cdot \left[\sum_{i=1}^n \ln \frac{x_i}{x_{\min} - 1/2} \right]^{-1}$$

Estimating x_{\min}

- If we fix x_{\min} , we can estimate k , so do the following:
- Choose x_{\min} to minimize “distance” between the data and the fitted distribution (for that x_{\min})
- To measure the distance between two distributions, we can use the “Kolmogorov-Smirnov” distance:

$$D = \max_{x \geq x_{\min}} |S(x) - P(x)|$$

- $S(x)$ = cumulative distribution function of the data
- $P(x)$ = cumulative distribution function of the fitted power-law

Finally, an obvious remark:

- Not all right-skewed distributions are power-laws:
 - Lengths of relationships between couples (exponentially distributed)
 - Abundance of North-American bird species (log-normally distributed)
 - etcetera; see Newman (2007)