



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Multiplicative Models for Company Dynamics

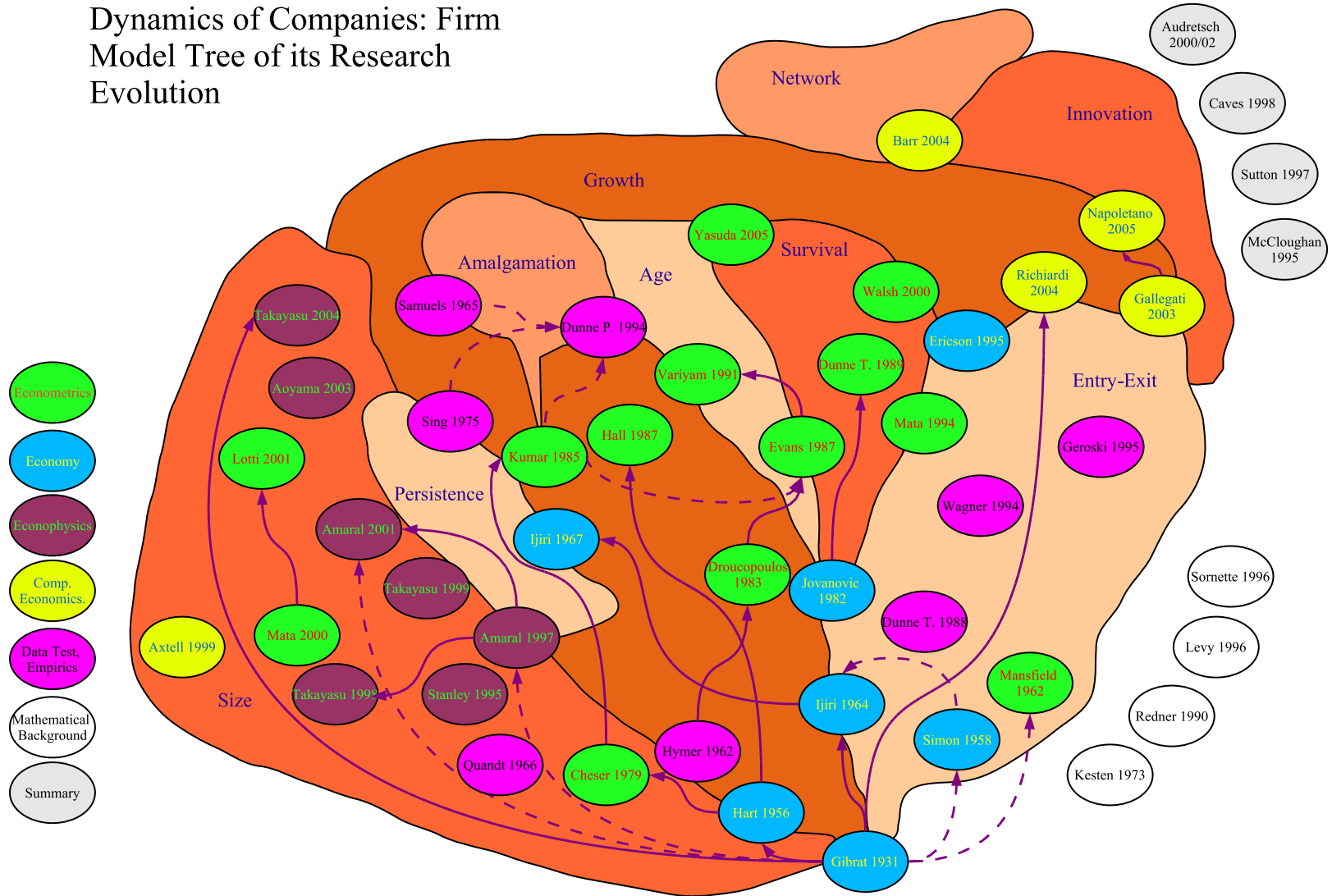
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Thanks to ...

S. Jain, L. Amaral, H. Takayasu, A. Seufert, N. Bürkler...

Dynamics of Companies: Firm Model Tree of its Research Evolution



Company Growth

- ▶ set of companies: $i = 1, \dots, N$
 - $x_i(t)$: company “size” (income, output, employees, ...)
 - growth rate: $dx_i/dt = \mathcal{F}_i(?)$
- ▶ $\mathcal{F}_i(t)$ with $\langle \mathcal{F}_i(t) \rangle = 0$, $\langle \mathcal{F}_i(t) \mathcal{F}_i(t') \rangle = S \delta_{ij} \delta(t - t')$
$$x_i(t + \Delta t) = x_i(t) + \sqrt{S \Delta t} \xi_i$$
 - growth as random walk (Bachelier, 1900)
- ▶ $\mathcal{F}_i = f(x_i) + \dots \Rightarrow$ rest of this talk
 - independent growth, proportional to size (Gibrat, 1931)
- ▶ $\mathcal{F}_i = f(x_j, x_k) + \dots$
 - growth through innovation networks

Growth through Network Effects

▶ $\dot{x}_i = \mathcal{F}_i = f(x_j, x_k) + \dots$

$$\frac{dx_i}{dt} = \sum_{j=1}^N c_{ij} x_j - \Phi x_i \quad (\text{Jain/Krishna '98, '01})$$

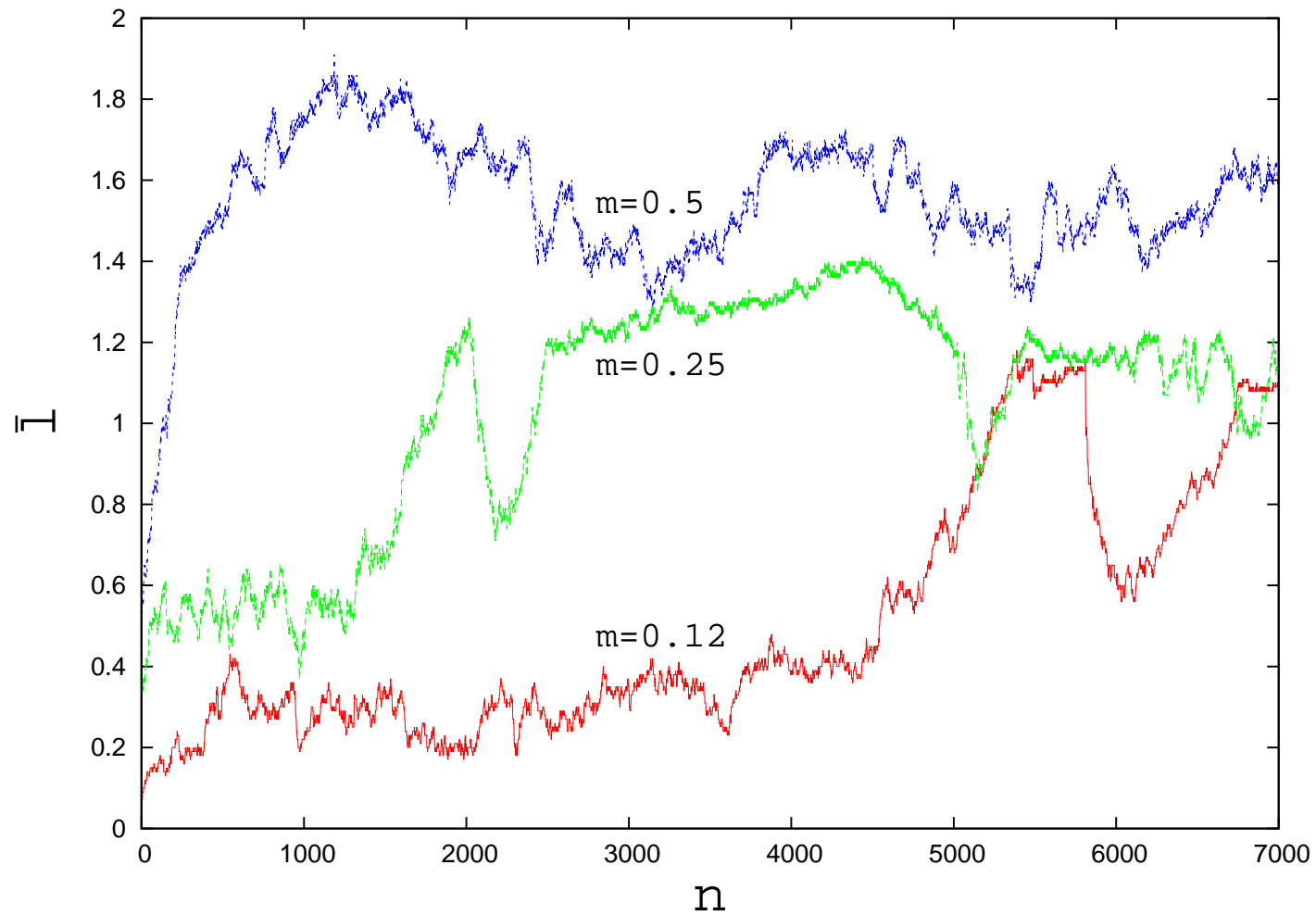
- $c_{ij} \in \{0, 1\} \Rightarrow$ represents a directed network
 - ★ j catalyzes the growth of i , link probability p
 - ★ i is connected to $m = p(N - 1)$ other companies (on average)

▶ *two time scales*: company growth (fast), network dynamics (slow)
assumption: extremal dynamics \Rightarrow minimum performance threshold

▶ questions:

- Under which conditions do companies survive?
- Which structures of innovation networks emerge?
- What happens if selection pressure is increased?

Result: Emergence of a core of *cooperative* companies, and a *parasitic* periphery, considerable crashes and recovery



Multiplicative Growth

▶ $\dot{x}_i = \mathcal{F}_i = f(x_i) + \dots = b x_i$

- “Law of proportionate growth” (Gibrat ’30, ’31; Sutton ’97)
- no interactions between firms

$$x_i(t + \Delta t) = x_i(t) \left[1 + b_i(t) \right]$$

▶ Assumptions:

- $b_i(t)$: independent of i , no temporal correlations (random noise)
- *growth “rates”*: $R(t) = x(t + 1)/x(t)$, $t \gg \Delta t$, $\ln(1 + b) \approx b$

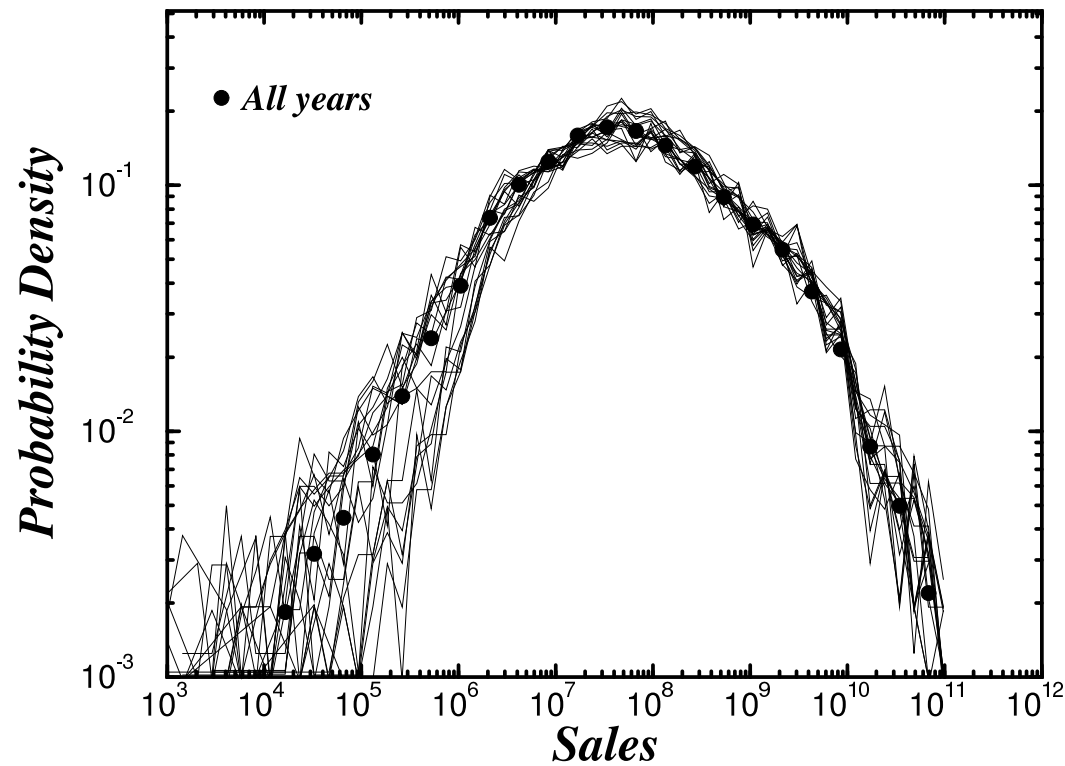
$$\ln R(t) = \sum_{n=1}^t b(n)$$

⇒ random walk for $\ln R(t)$ ⇒ log-normal distribution for $x_i(t)$

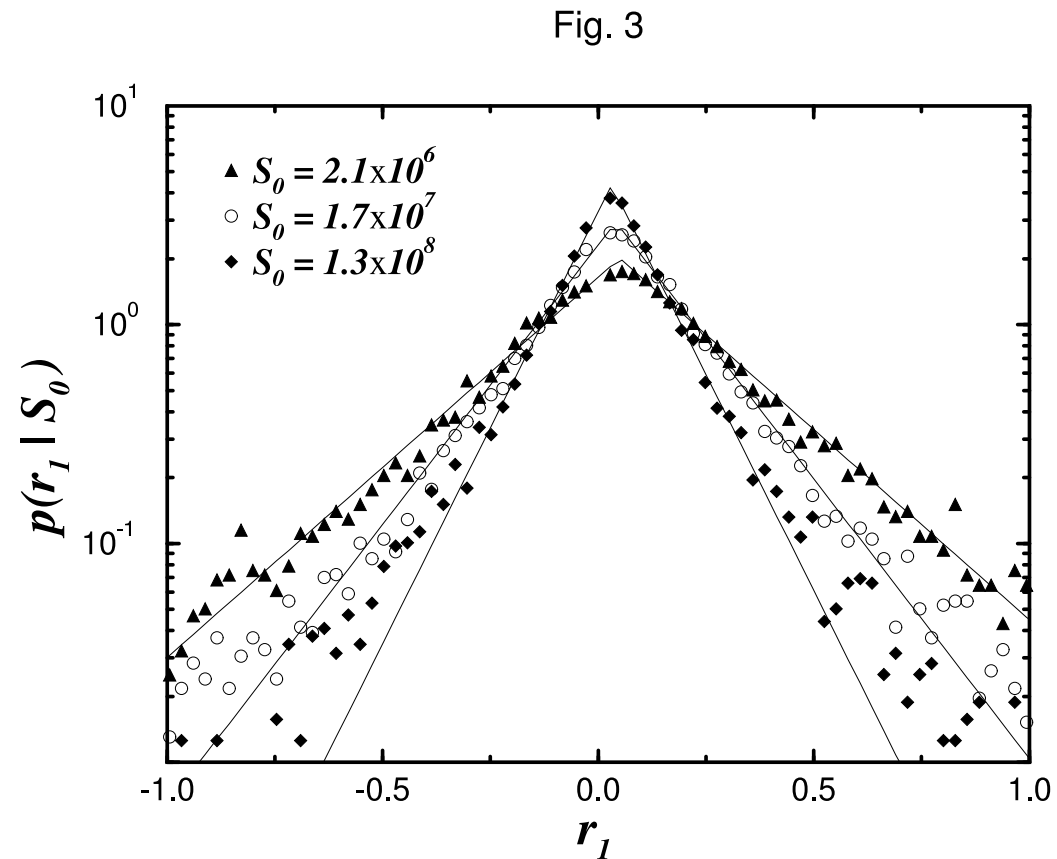
▶ empirical evidence?

- ▶ Empirical distribution of company sizes (1974-1993) (Amaral et al, 1997) \Rightarrow log-normal distribution

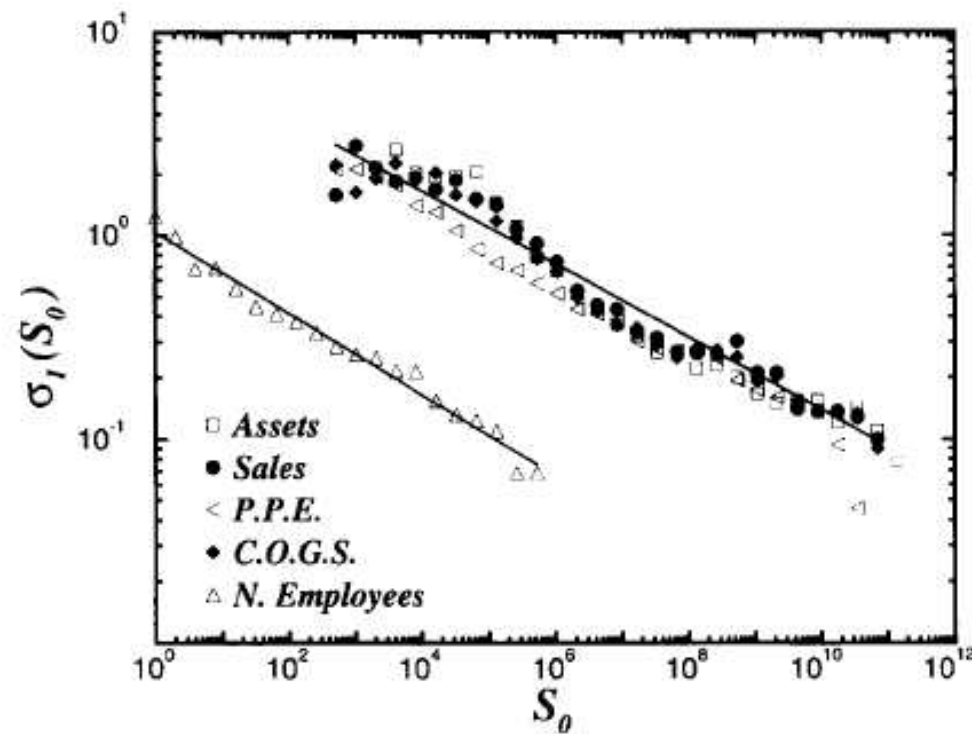
Fig. 2(a)



- ▶ Empirical distribution of growth rates (Amaral et al, 1997)
⇒ depend on size, tent-shape exponential distribution



- Empirical distribution of standard deviation of growth rates (Amaral et al, 1997) \Rightarrow depend on size, power-law distribution



Stylized facts:

- ▶ log-normal distribution of company sizes

$$P(x) = \frac{1}{\sqrt{2\pi} \sigma x} \exp \left[\frac{(-\ln x - \mu)^2}{2\sigma^2} \right]$$

- ▶ exponential growth ratio distribution

$$P(r_1|x_0) = \frac{1}{\sqrt{2} \sigma_1(x_0)} \exp - \left[\frac{\sqrt{2} |r_1 - \bar{r}_1(x_0)|}{\sigma_1(x_0)} \right]$$

- ▶ power-law distribution of the standard deviations

$$\sigma_1(x_0) \sim x_0^{-\beta} ; \quad \beta < 0.5$$

Explanations:

- ▶ correlations in the growth rates
company is attracted to an “optimal size”

$$\frac{x_{t+\Delta t}}{x_t} = \begin{cases} ke^{\varepsilon t}, & x_t < x^* \\ \frac{1}{k}e^{\varepsilon t}, & x_t > x^*, \end{cases}$$

- ▶ growth depends on properties of management hierarchies
 n levels, z units, decisions on higher level are followed with prob π

$$\beta = \begin{cases} -\ln(\pi)/\ln(z) & \text{if } \pi > z^{-1/2} \\ 1/2 & \text{if } \pi < z^{-1/2} \end{cases}$$

- β decreases in time \Leftrightarrow companies better coordinated

Further Improvements of Gibrat's Model

- *economic idea*: simple entry dynamics (Simon & Bonini '58)
- *mathematic idea*: add more noise! (Kesten '73)

$$x(t + 1) = x(t) [1 + b(t)] + a(t)$$

- b , a positive independent random variables
- $a(t)$ acts as “effective repulsion” from zero (Sornette & Cont '97)

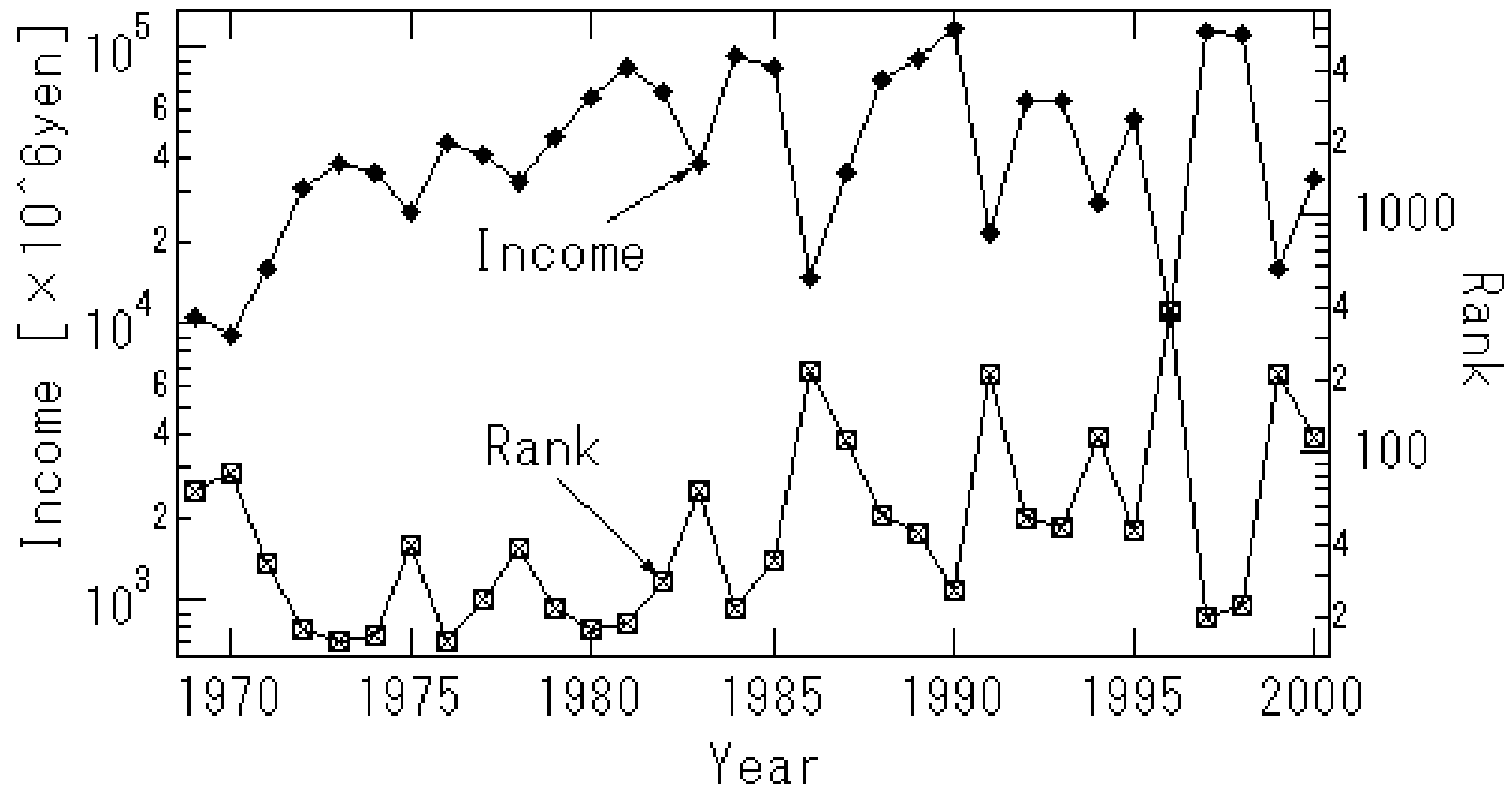
- *practical idea*: fit parameters (Takayasu et al '03)

$$x(t + 1) = \alpha(t)\lambda(t, x) x(t) + a(t)$$

- $[1 + b] \rightarrow \lambda(x, t)$: growth depends on size
- estimation from $\ln R(t) = \ln\{x(t + 1)/x(t)\}$ with standard deviation $\sigma(x)$

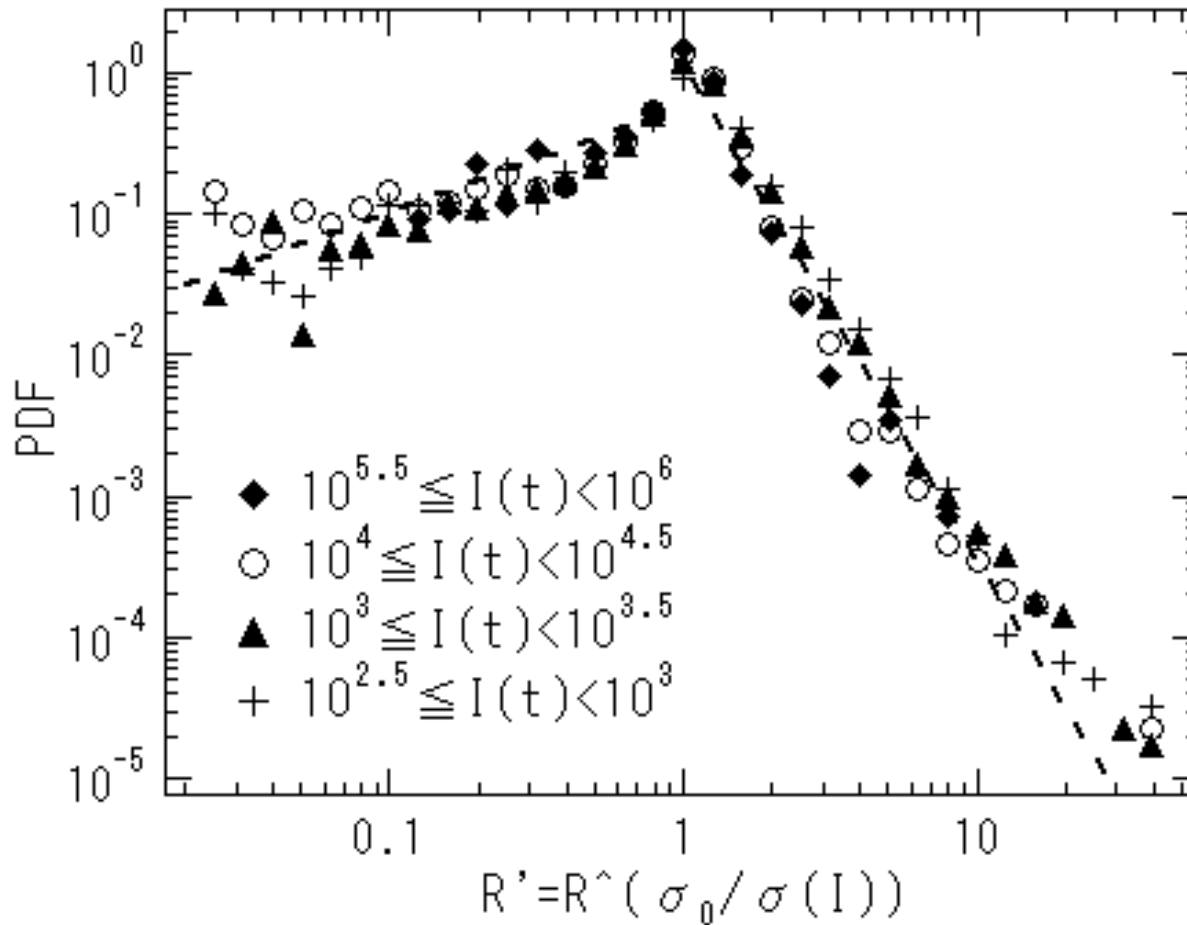
Comparison with real company data

- ▶ Takayasu et al '03: income of 15.000 US and 15.000 non-US comp., 80.000 Japanese comp. (income > 40 Mio Yen), before tax



Estimation of $\ln R(t)$, $\sigma(x)$

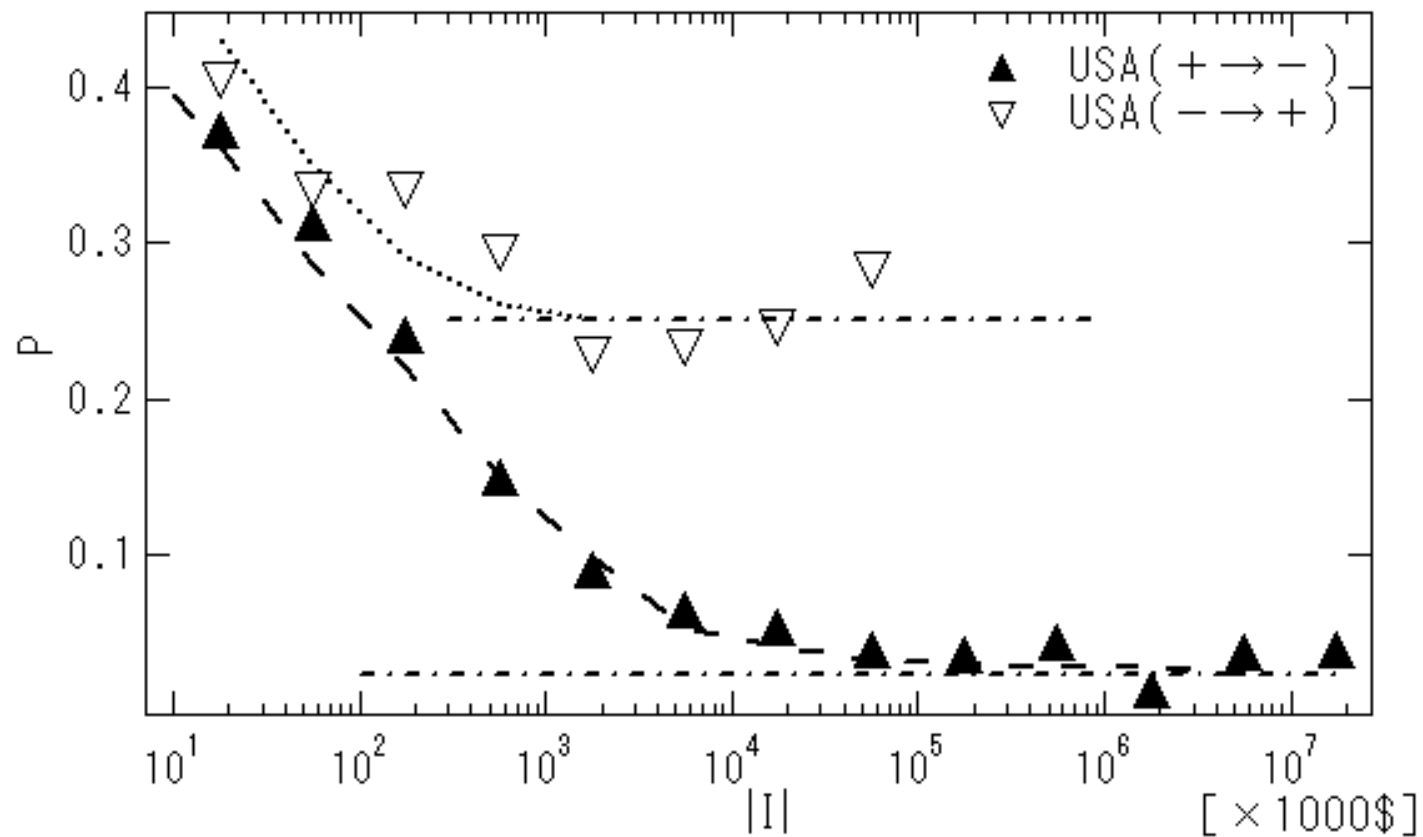
- for large x : σ_0 , $f(t)/x$ negligible
 scaling by means of normalized growth: $R^{\sigma(x)/\sigma_0}$



(Takayasu et al '03)

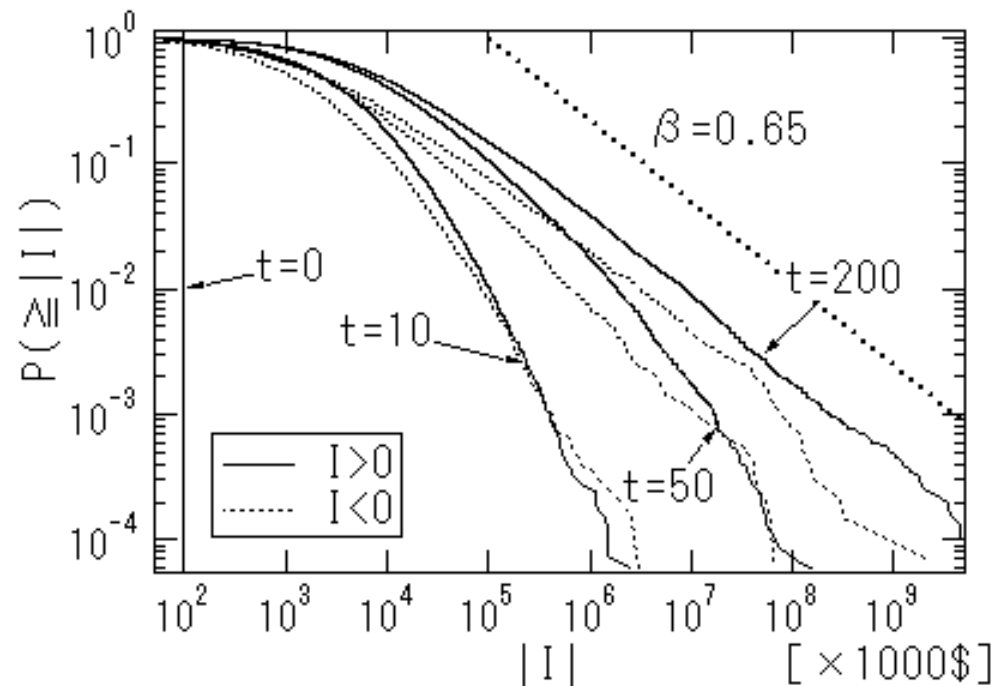
- $\alpha(t)$ either +1 (growth) or (-1) (slump)
 prob. determined empirically from large $|x(t)|$

$$\alpha(t) = \begin{cases} 1 & \text{with prob. 0.97 } (x(t) > 0), & 0.75 \text{ } (x(t) < 0) \\ -1 & \text{with prob. 0.03 } (x(t) > 0), & 0.25 \text{ } (x(t) < 0) \end{cases}$$



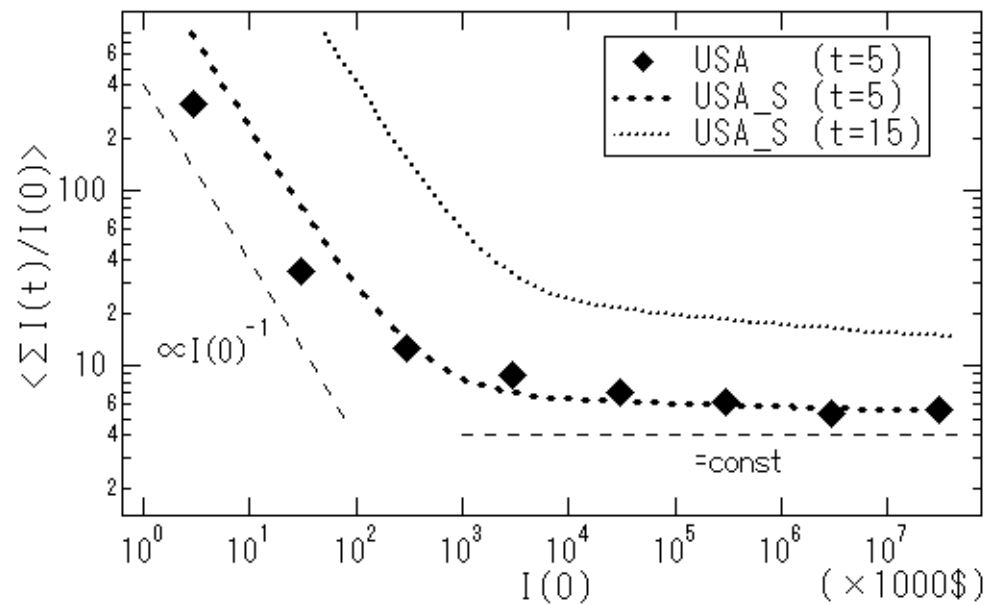
Forecast by means of Monte Carlo Simulations

- ▶ initial state: 6.000 companies, $x_i(0) = 100$
coefficients estimated from real data
- ▶ $t = 50$: qualitative agreement with real distribution (US)
with constant growth rate distribution: firms income will keep growing
for more than 100 years



Investment Strategies?

- ▶ normalized cumulative income for 5 years: $I = \sum_{n=1}^5 x(n)/x(0)$



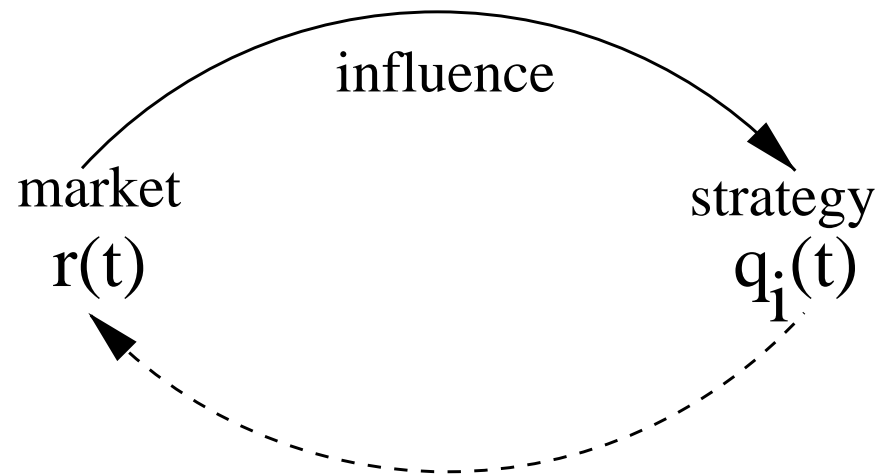
- ▶ for $x(0) > 10^6\$$: $I \propto x(0) \Rightarrow$ invest in small firms?

Topics for Future Investigations

1. Growth strategies

- ▶ companies with growth strategies: $b(t) \rightarrow r(t)q_i(t)$

$$x_i(t + \Delta t) = x_i(t) \left[1 + r(t)q_i(t) \right] + a$$



- ▶ $r(t) < 0 \Rightarrow q_i(t) \rightarrow q_{\min}$, $r(t) > 0 \Rightarrow q_i(t) \rightarrow q_{\max}$
 - Can company predict $r(t)$? Can it adjust $q_i(t)$ fast enough?

2. Stochastic Models with interactions

- ▶ combining multiplicative growth models and network models

$$x_i(t+1) = x_i(t) \left[1 + b_i(t) + \sum_{j=1}^N c_{ij} x_j \right] + \mathcal{G}(t) - \mathcal{L}(t)$$

- network dynamics: $c_{ij} \Rightarrow c_{ij}(t, \mathbf{r})$
represents spatial interaction, cooperation, spread of innovations, ...
- “catalytic” growth: $b_i(t) = r(t)q_i(t)$
represents strategic decisions, organizational efficiency, ...
- $\mathcal{G}(t)$, $\mathcal{L}(t)$ represent external conditions (gov, environment, ...),
global couplings (limited resources, ...)

Conclusions

- ▶ *company growth*: field of applications for stochastic processes
 - network effects? \Rightarrow innovation economics, input from physics?
- ▶ *multiplicative growth models*: are successfully applied to real data
 - stylized facts reproduced
 - conceptual drawbacks: random growth \Leftrightarrow firms as profit maximizers respond similarly to changing market conditions
- ▶ *micro models ??*
 - hierarchical-tree model (Amaral et al, 1997)
 - multi-agent models (Axtell, ...)
 - heterogeneity? market ecology of different organizational forms