

Epidemic SIR dynamics on scale-free nets

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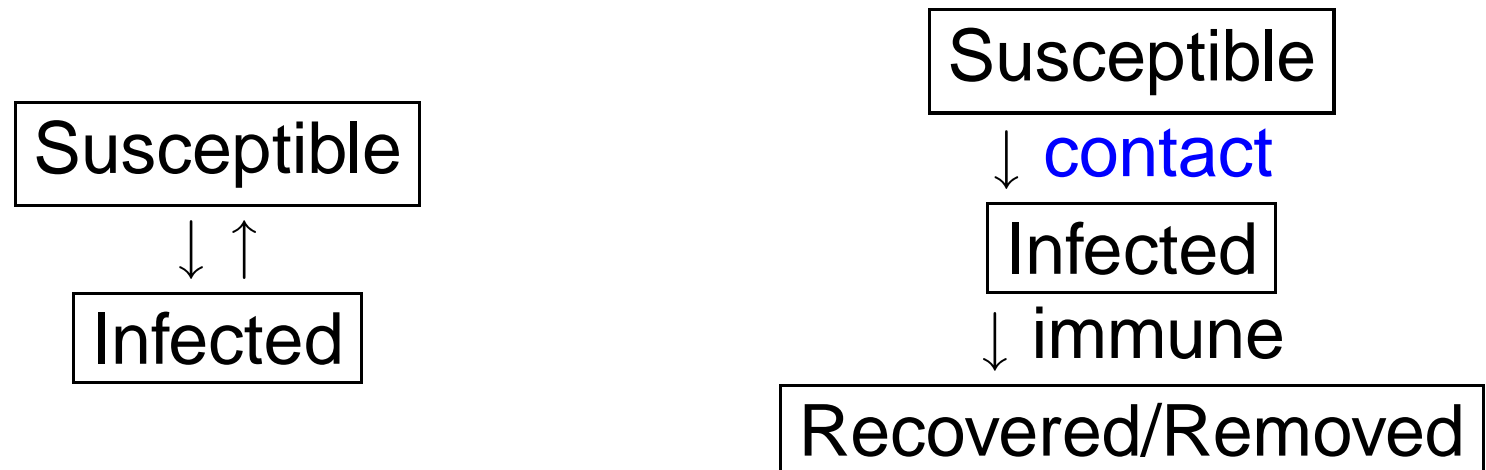
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⇒ Refer to **Y.Hayashi et al., “Oscillatory epidemic prevalence in growing scale-free networks,”** Physical Review E 69, 016112, 2004,

& Virtual Journal of Biological Physics Research, STATISTICAL AND NONLINEAR PHYSICS, Feb.1, 2004, <http://www.vjbio.org>

Conventional Epidemic Model

State transition in Kermack-McKendrick models



assuming lattice or random graphs, and fixed size:
equal birth and death rates in a const. population

⇒ Today, we are traveling in world-wide, and communicating on the growing Internet !

Common Struct. in Many Real Nets

Many real nets \neq regular(e.g lattice), uniformly random



Recently('98-'02), a surprisingly common structure has been found in many real nets

Social: actor-collabo., world trading, acquaintance, citation, language

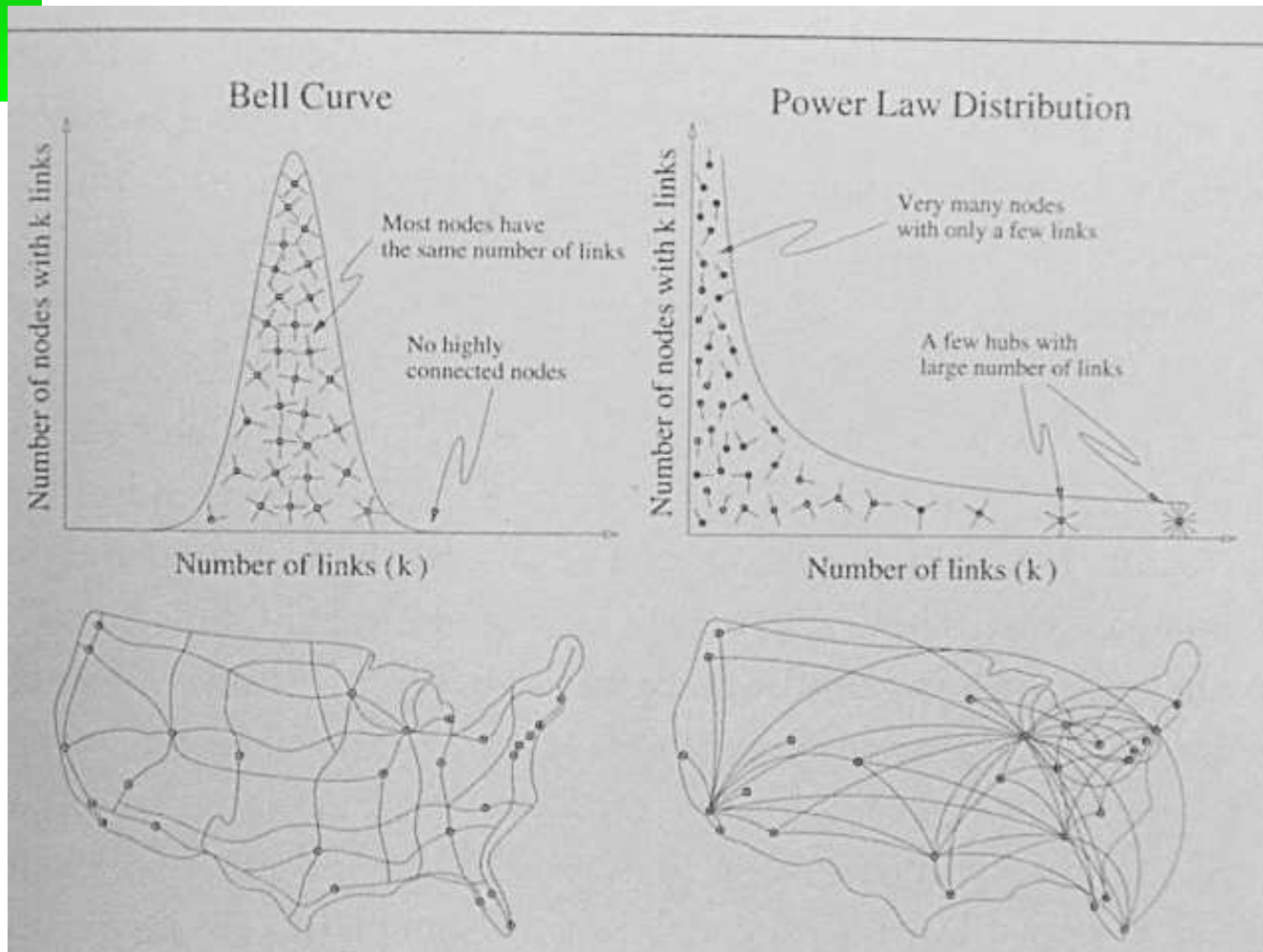
Technological: power grid, Internet, WWW, email

Biological: neural net, genome, metabolic pathway, foodweb

The topological struct. are deeply related to the damage of prevalence.

Scale-Free Network

The heterogeneous structure with hubs are called **SF net**, whose degree dist. exhibits $P(k) \sim k^{-\gamma}$, $2 < \gamma < 3$.



Different Spreading Properties

To be fully connected, more than average

Random: 3 links are necessary
the delivery time \propto the size N

Realistic SF: 5 links are necessary
almost const. time, due to the spread from hubs
(larger net. consts of nodes with larger degrees)

K. Ohkubo, Y. Hayashi, and S. Ninagawa, Trans. of IEICE
Vol. J85-D-I, No. 2, 2002

\Rightarrow Many researchers are attracted to SF networks and
the evolution mechanisms.

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Assortative: social

Disassortative: technological or biological

M.E.J Newman, PRE 67, 026126, 2003, A. Vázquez, PRE 67, 056104, 2003.

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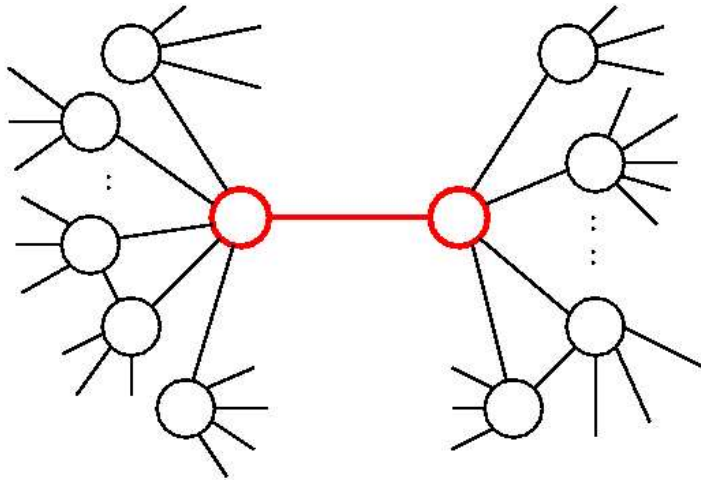
Disassortative: technological or biological

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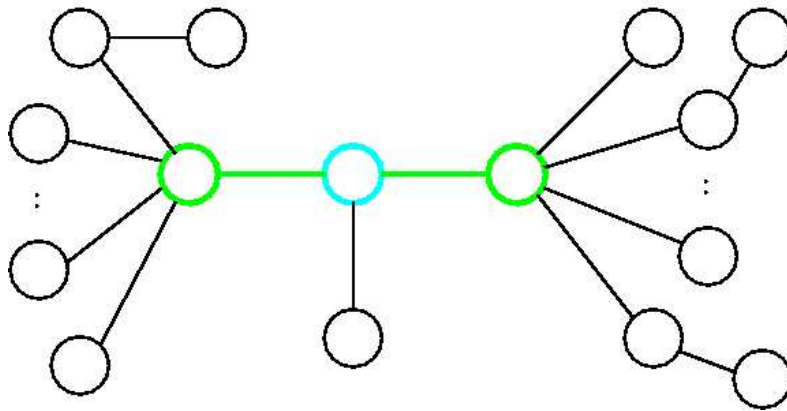
The struct. are crucial for epidemic spreading

Note that our contact relations (email, world trading, etc.) are supported by both **social** and **tech.** networks, today !

Assortative and Disassortative Correlations



Ass: tend to have connections between similar peers

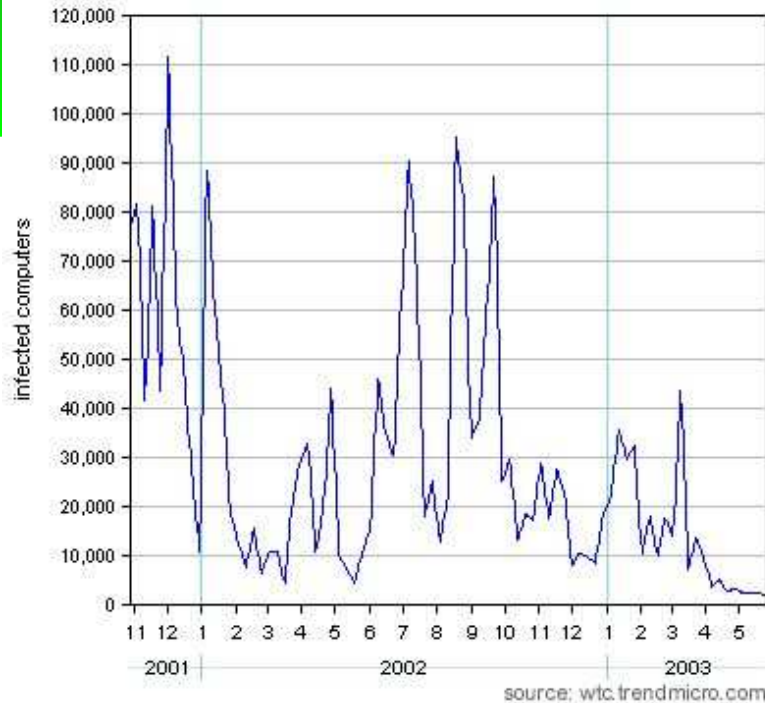


Dis: hub and peripheral nodes with low degrees

Let us consider the conditional probability $P(k|l)$ of connection of nodes with deg. k, l for each type.

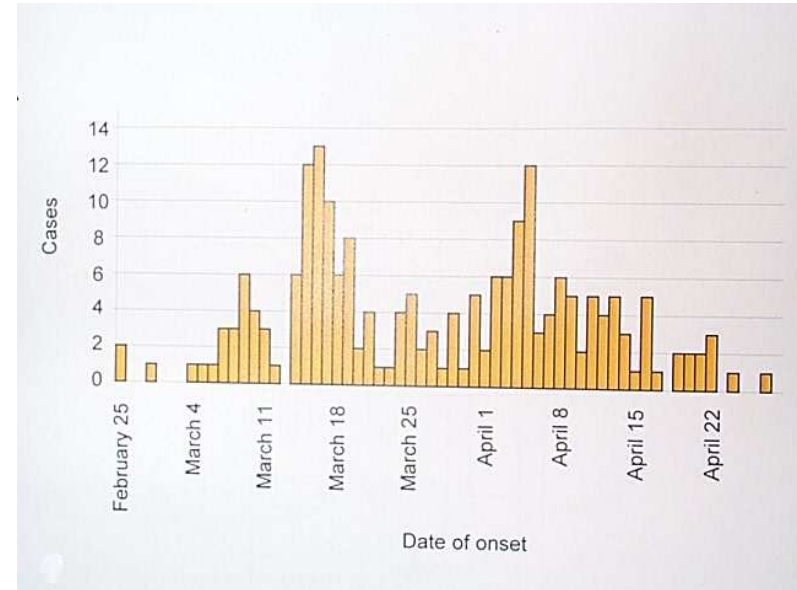
Oscillatory Epidemic Prevalence

Typically observed, but unknown the mechanism



computer virus NIMDA

⇒ SIR (susceptible-infected-recovered/removed state transition) model on SF nets for epidemic spreading



SARS in Singapore,
Scienceexpress May 23,
2003

Heterogeneous SIR Model on Linearly Growing SF Nets

Epidemic dynamics for the macro. eq. at the MF level

$$\frac{dS_k}{dt} = -bk \underbrace{S_k \Theta_k}_{\text{contact}} + a_k, \quad \frac{dI_k}{dt} = -\delta I_k + bk \underbrace{S_k \Theta_k}_{\text{contact}},$$

where b and δ denote the infection and immune rate, $a_k = Ak^{-\gamma}$, $A > 0$, provides a constant increasing of S_k ,

Mean-Field infection: $\Theta_k(t) \stackrel{\text{def}}{=} \sum_l \frac{l-1}{l} P(k|l) \frac{I_l(t)}{N_l(t)}$.

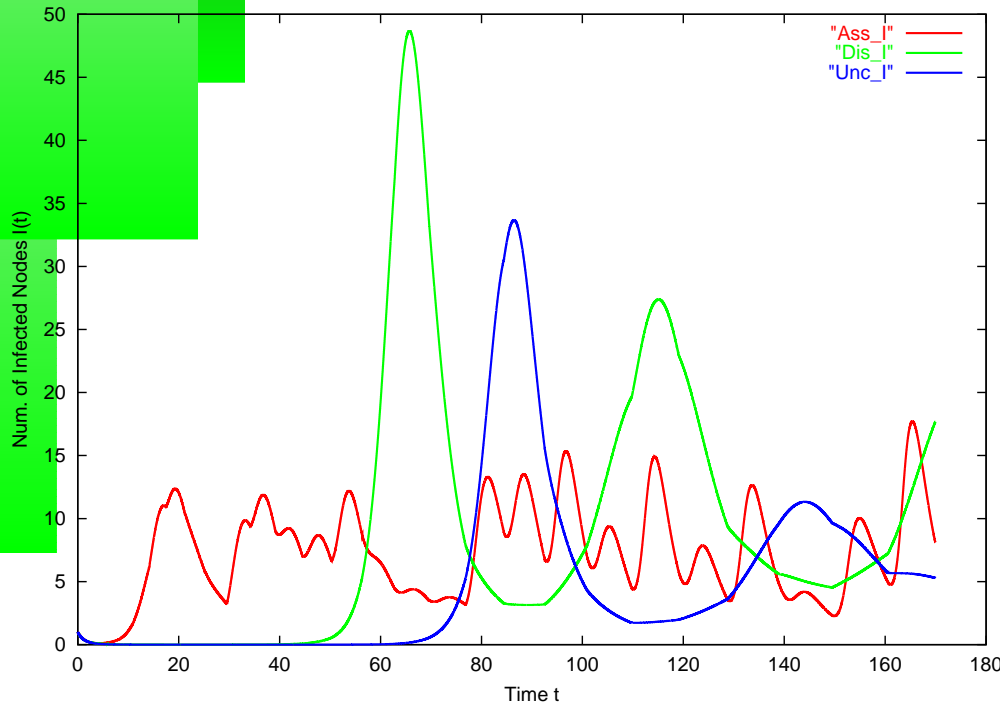
From $N_k = S_k + I_k + R_k$ and $\frac{dR_k}{dt} = \delta I_k$, $\frac{dN_k}{dt} = a_k$, **the growing** $N_k(t) \sim a_k t$ gives asymptotic $P(k) \sim k^{-\gamma}$,

$$N(t) = \sum_k N_k \sim \left(\sum_k a_k \right) t.$$

\Rightarrow **linearly growing SF net**, and simultaneously progress of epidemic spreading

Simulation Results

Different behavior depend on the correlation types
Trade-off: persistency and breaking size



Ass: persistently survived with fluctuation

Dis: later outbreaks

Unc: corresponded to the conventional SF models without correlations

Time evolution of $I(t) = \sum_k I_k(t)$

⇒ the behavior on Dis or Unc is also consistent with a stochastic SIR model, **but it on Ass has'nt been found**

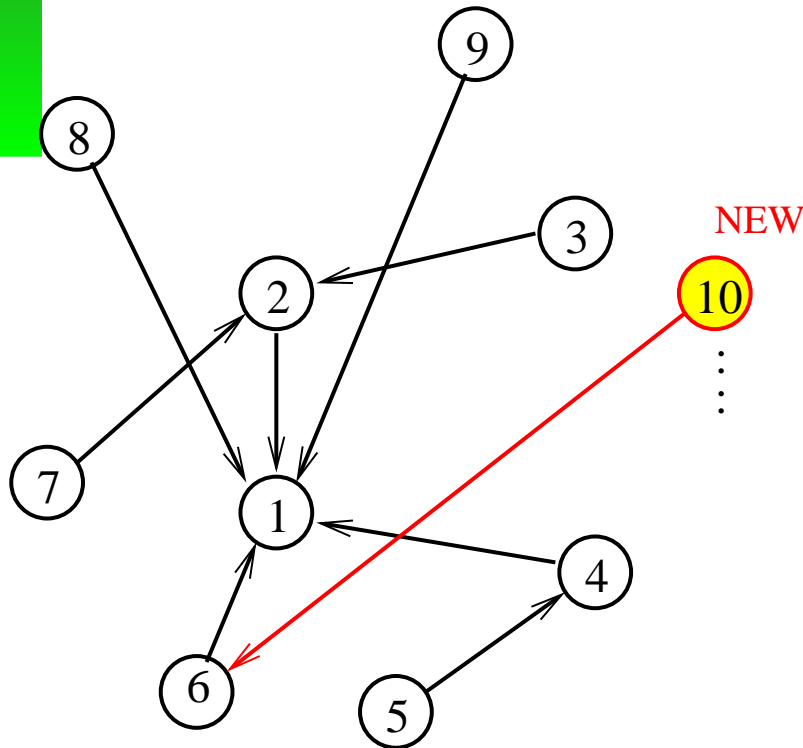
Summary

- Instead of the conventional homogeneous contacts in a const. population, we've considered a more realistic SIR model on growing SF nets with new participants.
 - Moreover, we've extended it with Ass (social, between peers) and Dis (tech. or bio., hub-periph.) connectivity correlations.
 - In our simulation of the SIR dynamics, the correlations cause quite different behavior for epidemic spreading.
- ⇒ a good struct. will be used for preventing the spread with huge damage in distributed network evolutions.

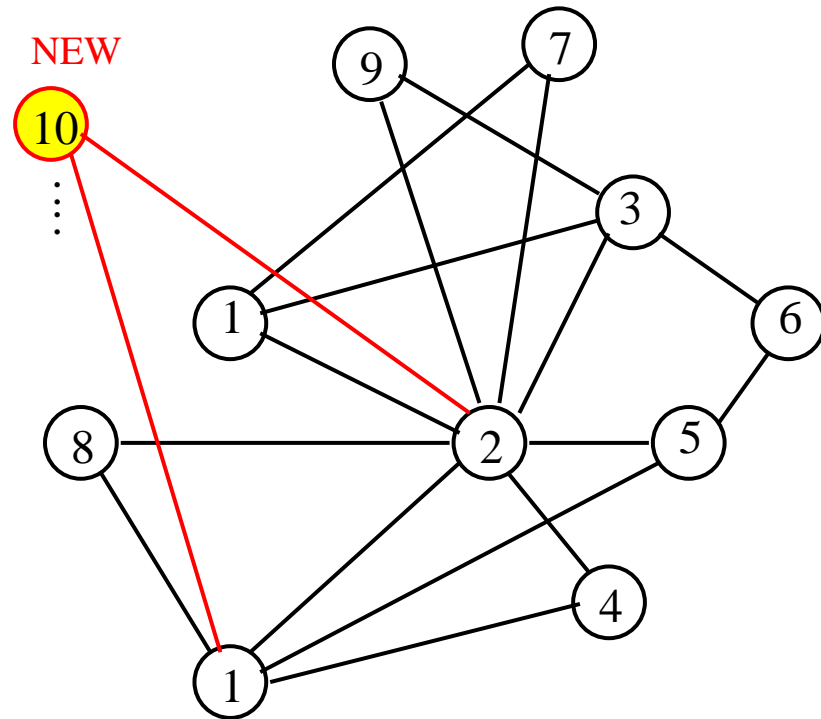
Appendix 1: Evolutional Net. Models

$$P(k) \sim k^{-\gamma}, \gamma \sim 3, m = 1$$

$$\gamma = 3, m = 2$$



Krapivsky-Render's GN model

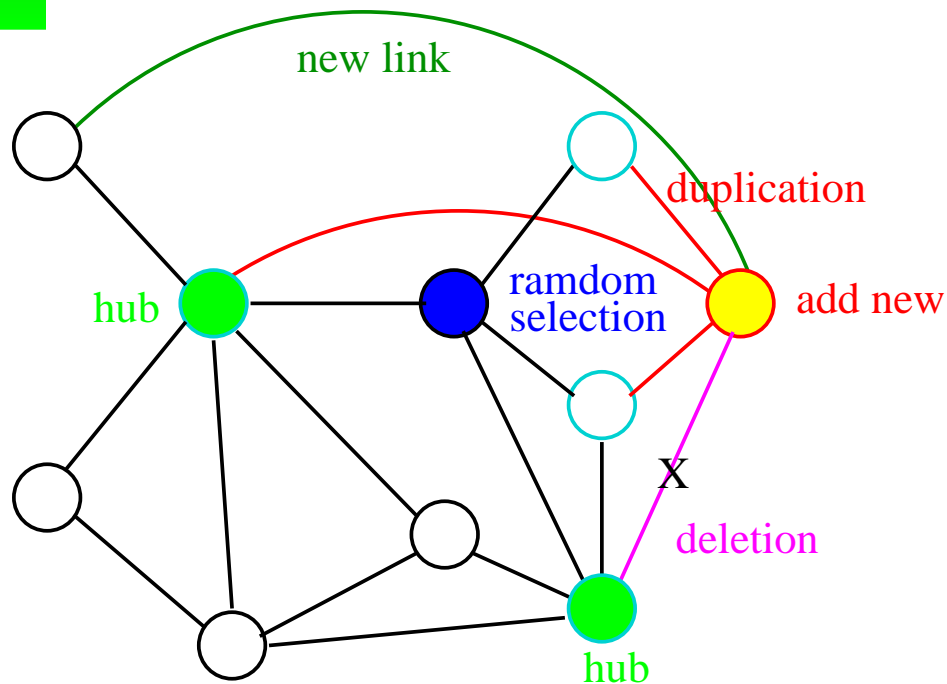


BA model

⇒ age-effect (S.N. Dorogovtsev et al. PRL 85, 2000), rewire(A. Albert, PRL 85, 2000), fitness(G. Bianconi, PRL 86, 2001)

Appendix 2: Duplication Model

In spite of random node selection, the neighbor hub node has many chance to get duplicate connections (proportional to the degree).



⇒ Biologically plausible networks realize Preferential Attachment in a local rule !

Appendix 3: Robust and Vulnerable Connectivity

Robust: for random failure, remaining the connectivity

Vulnerable: for **targeted attack against hubs**, disconnecting into many isolated parts



Appendix 4: Variety of Correlations

$$P(k|l) \sim \alpha k^{1-\gamma} + \frac{1-\alpha}{|k-l|^{\nu+1}},$$

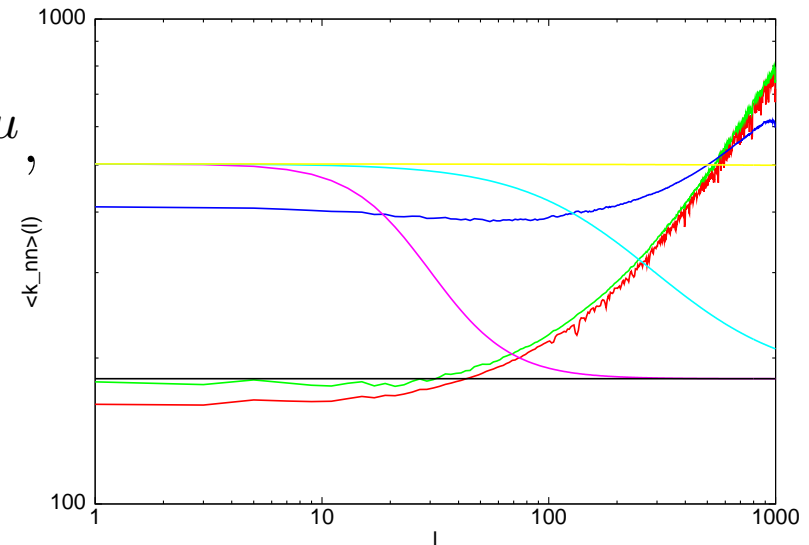
$$\sim \alpha k^{1-\gamma} + (1-\alpha)l^{-\mu},$$

and adding random noise as arbitrary correlations

$$\langle k_{nn} \rangle (l) \stackrel{\text{def}}{=} \sum_{k=1}^K kP(k|l)$$

$\alpha = 0.2$, $\gamma = 2.5$, **Ass**: $\nu = 2.5, 1.5, 0.5$, **Dis**: $\mu = 2.5, 1.5, 0.5$

estimated from real data of actor-collabo. (Ass), Internet-AS level-, and email (Dis).



ave. correl. $\langle k_{nn} \rangle$ vs. degree l