

BIO 365 ecological networks

coordinators: Jordi Bascompte and Alessandro Vindigni

co-teachers: Matt Barbour, Rodrigo Cámara-Leret, Marília Gaiarsa,
Klementyna Gawecka, Matt Hutchinson, Eva Knop, Fernando
Pedraza

Introduction to course

outline of course

From	To		Thursday March 17	Friday March 18
10:15	12:00	LECTURE	Outline and Intro	Food webs
	Speaker		Bascompte	Gaiarsa
12:00	13:00		Lunch	
	Lead	EXERCISE	Vindigni	Gaiarsa
13:00	17:00		Toolkit for network analysis	Measuring modularity

Tuesday March 22	Wednesday March 23	Thursday March 24	Friday March 25
	Mutualistic networks	Null models	Statistical approaches
	Bascompte	Bascompte	Barbour
Lunch	Lunch	Lunch	Lunch
Knop	Gawecka	Pedraza	Barbour
Sampling an ecological network	Measuring nestedness	Null models	Statistical models

Tuesday March 29	Wednesday March 30	Thursday March 31	Friday April 1
	Spatial networks	Animal behaviour	Evolution in networks
	Gawecka	Hutchinson	Pedraza
Lunch	Lunch	Lunch	Lunch
Vindigni	Gawecka	Hutchinson	Pedraza
Assessing topological robustness	Comparing networks in space	Simulating foraging	Models of evolution in networks

Tuesday April 5	Wednesday April 6	Thursday April 7
	Socio-ecological networks	Open time
	Cámara-Leret	office hours (with appointment)
	Lunch	Lunch
	Cámara-Leret	Vindigni
Open time	Socio-ecological networks	Exam

general readings

general readings

- Barabási, A.-L. (2002). *Linked: The New Science of Networks*. Perseus Books Group
- Bascompte, J. and Jordano, P. (2013). *Mutualistic Networks*. Princeton University Press
- Pascual, M. and Dunne, J.A. (2006). *Ecological Networks: Linking Structure to Dynamics in Food Webs*. Oxford University Press
- Pimm, S.L. (1982). *Food Webs*. Chicago University Press

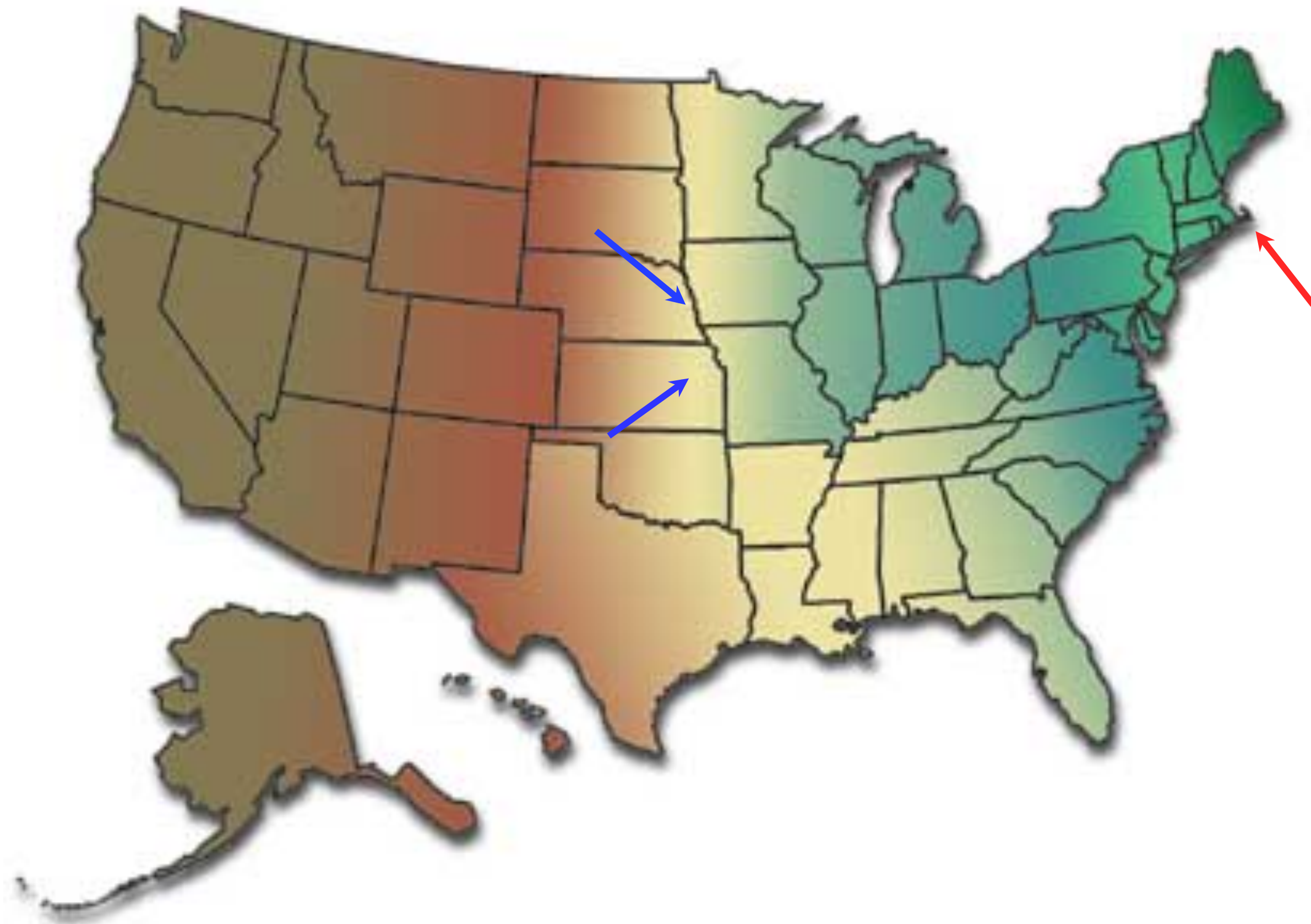
course grading

- Practicals with web-based RStudio environment, up to three points.
- Written exam: multiple-choice test, up to two points.

Introduction to network theory

social networks

160 letters from Wichita (Kansas) and Omaha (Nebraska) to Sharon (Mass)



Milgram (1967)

social networks

In the Nebraska study, the chains varied between 2 and 10 intermediate acquaintances, with the median at 5

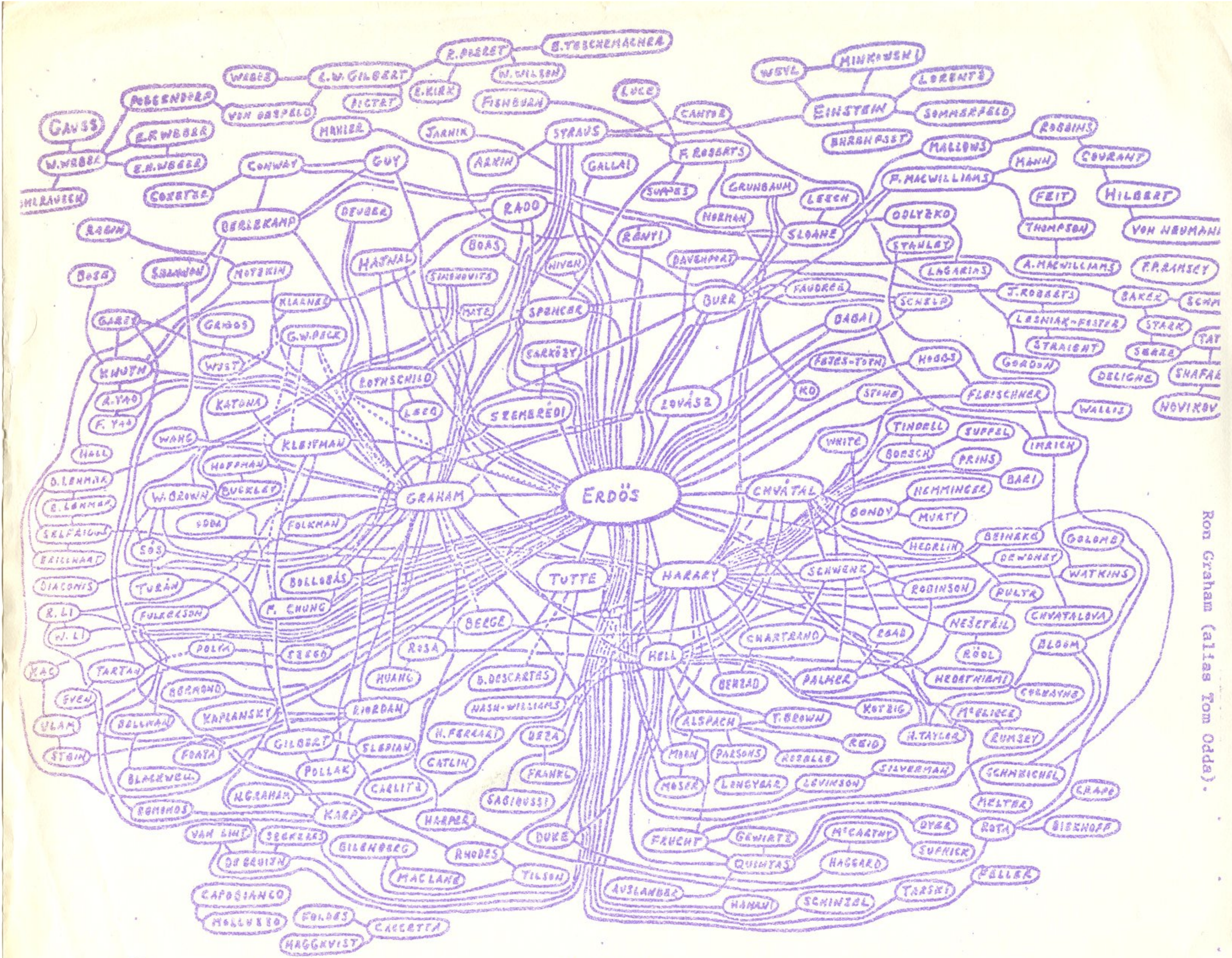
What a small world!

El mundo es un pañuelo!

C'est petit le monde!

Die Welt ist klein!

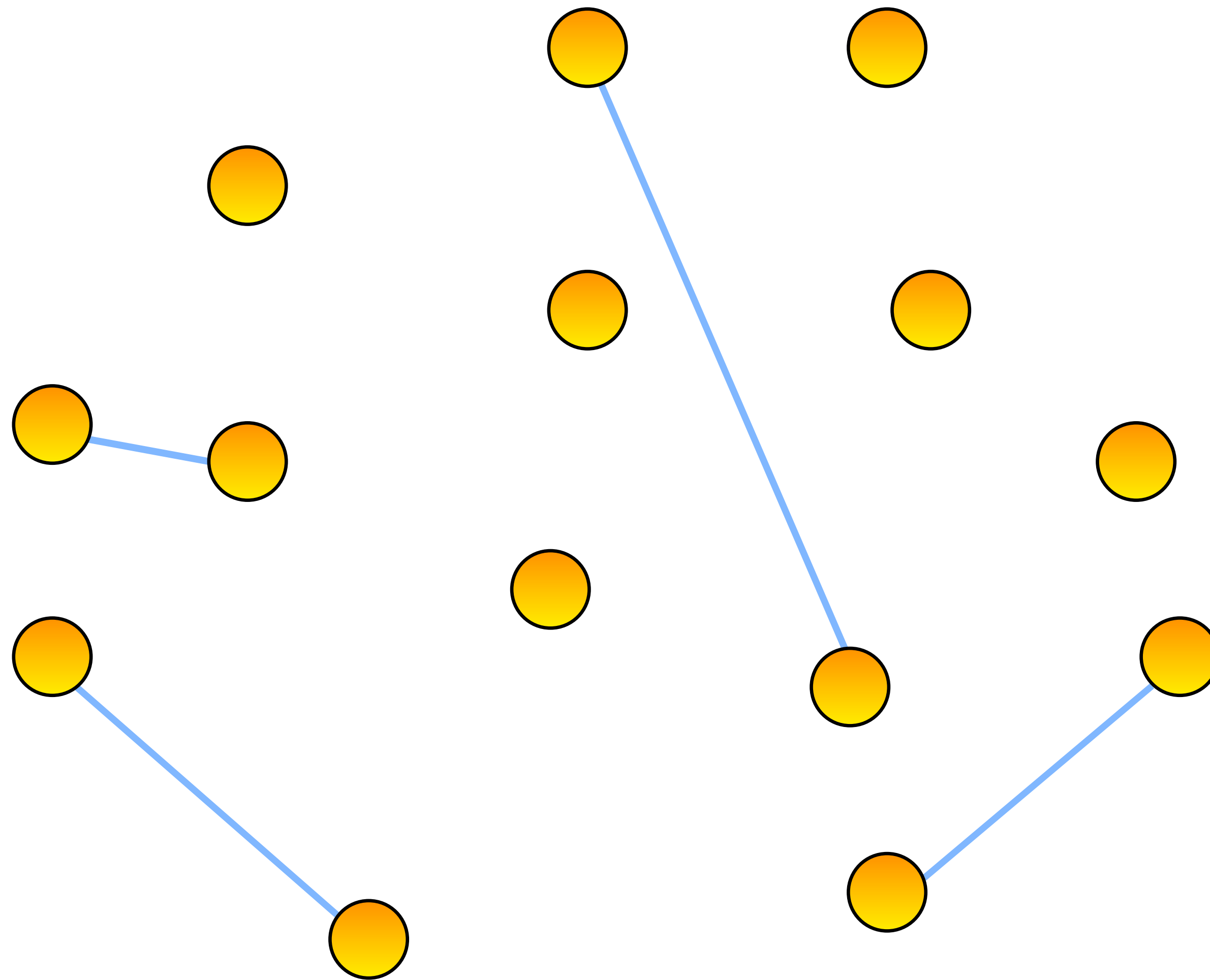
Milgram (1967)



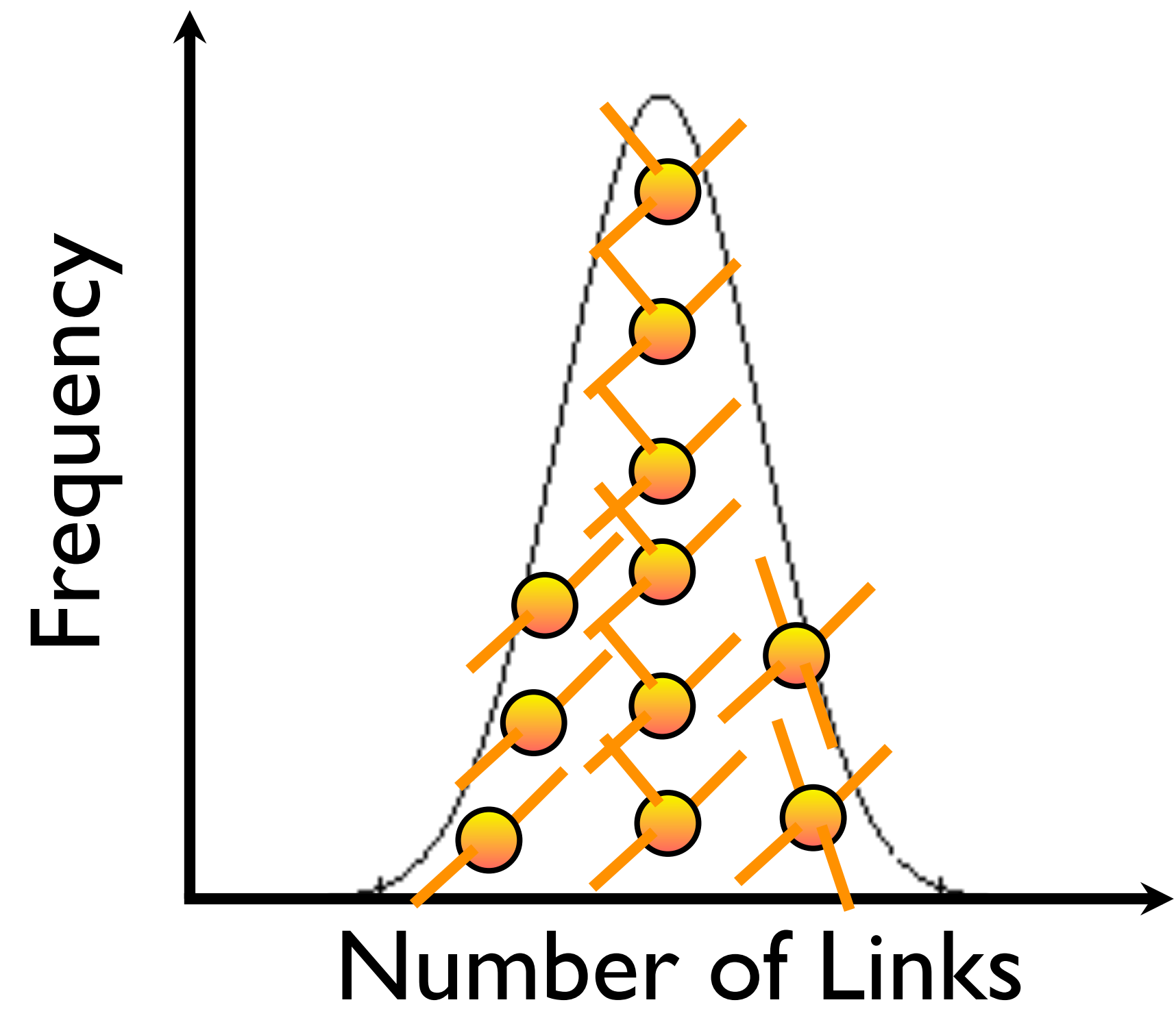
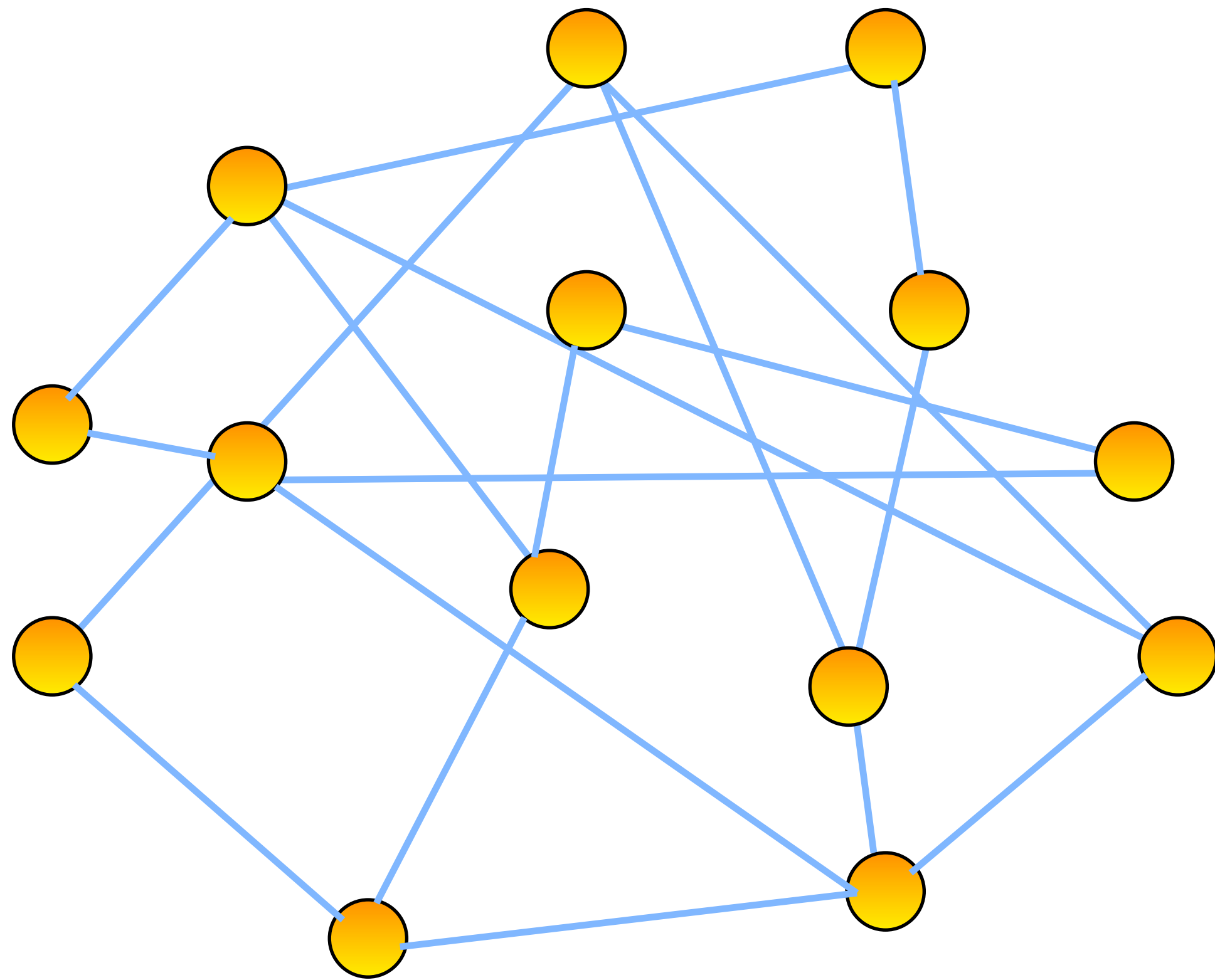
Ron Graham (alias Tom Oda).

Figure 1
 To appear in Topics in Graph Theory (F. Harary, ed.), New York Academy of Sciences (1979).

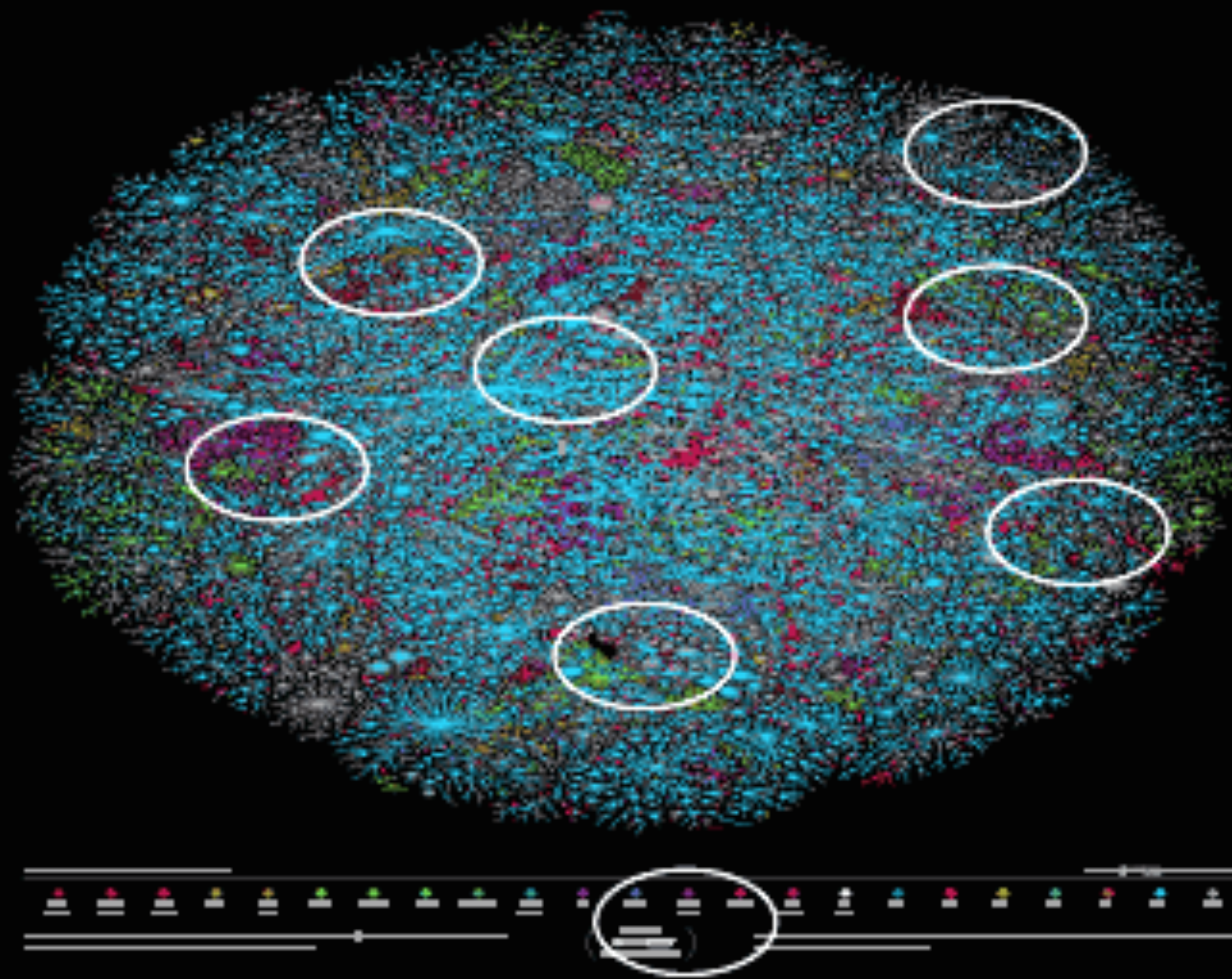
Erdős-Rényi model



Erdős-Rényi model



THE INTERNET: 2001



VARIG
Brasil
LINHAS NACIONAIS
 DOMESTIC ROUTES



GRUPO VARIG

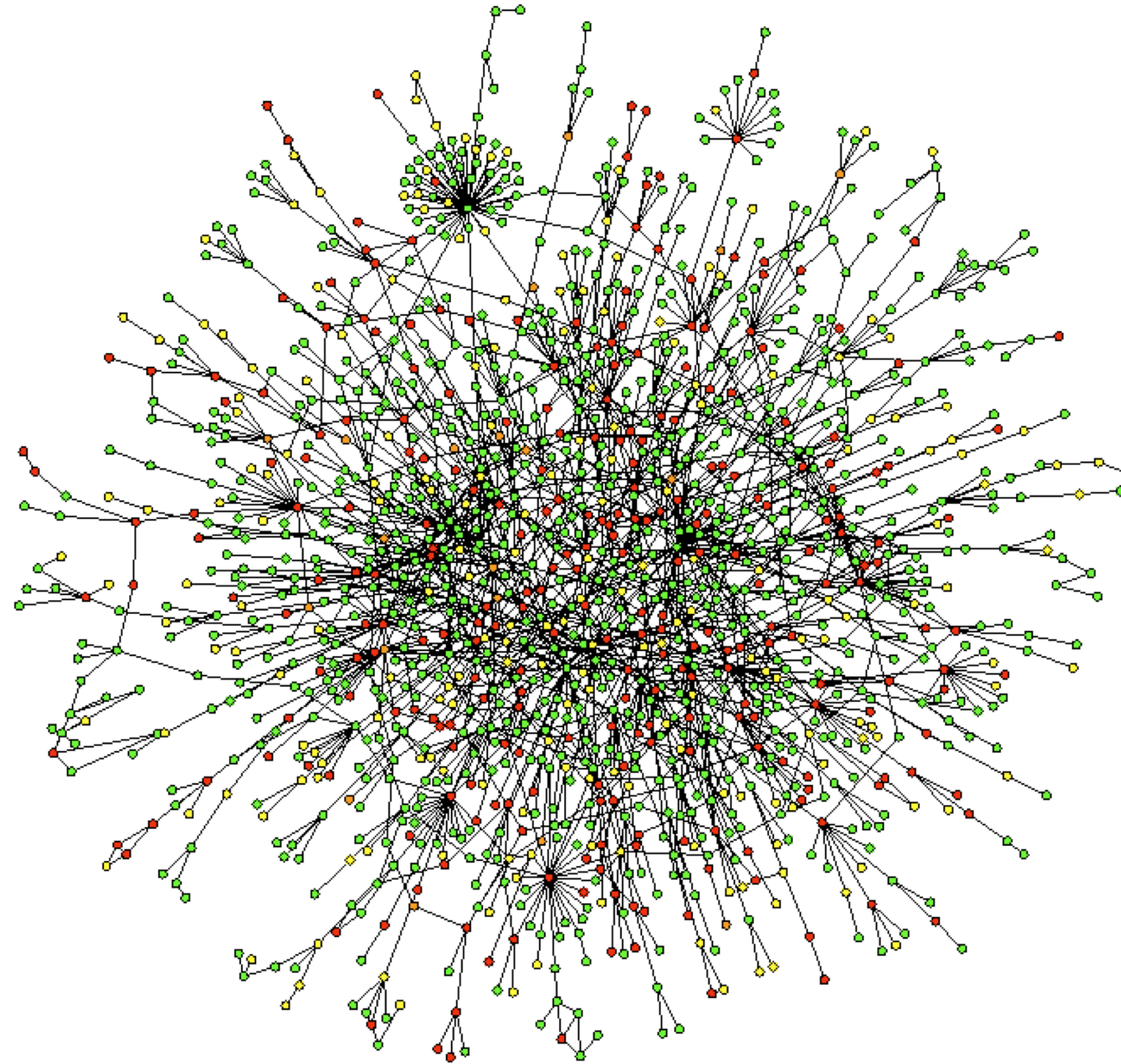
RESERVAS BRASIL
 Reservation 0300-7887000*

* O custo por minuto para ligações feitas de telefone fixo é R\$ 0,29 e de telefone celular, R\$ 0,63.

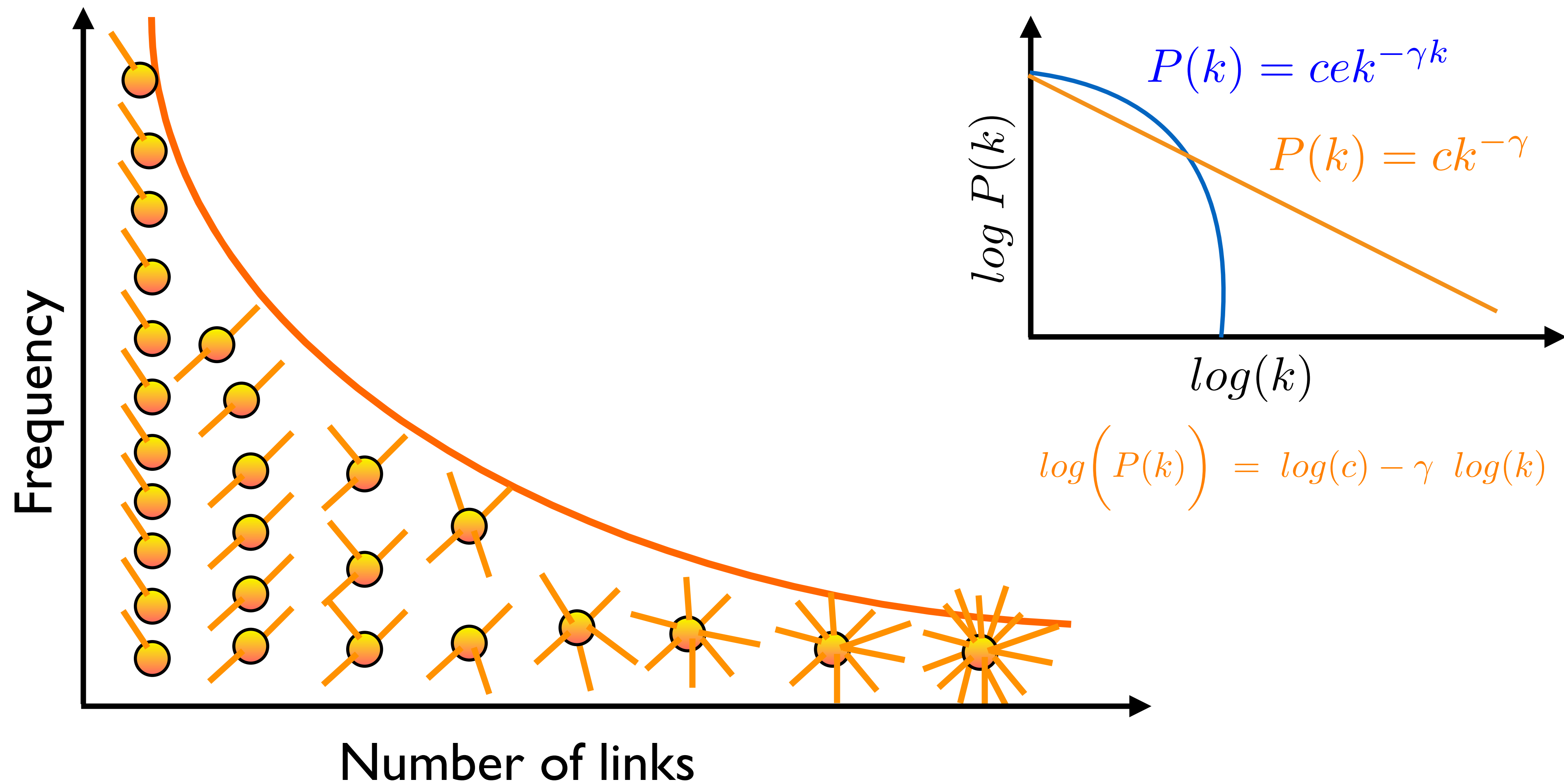


MAPAS: JUS FERNANDO MARTINI

protein networks



complex networks are heterogeneous

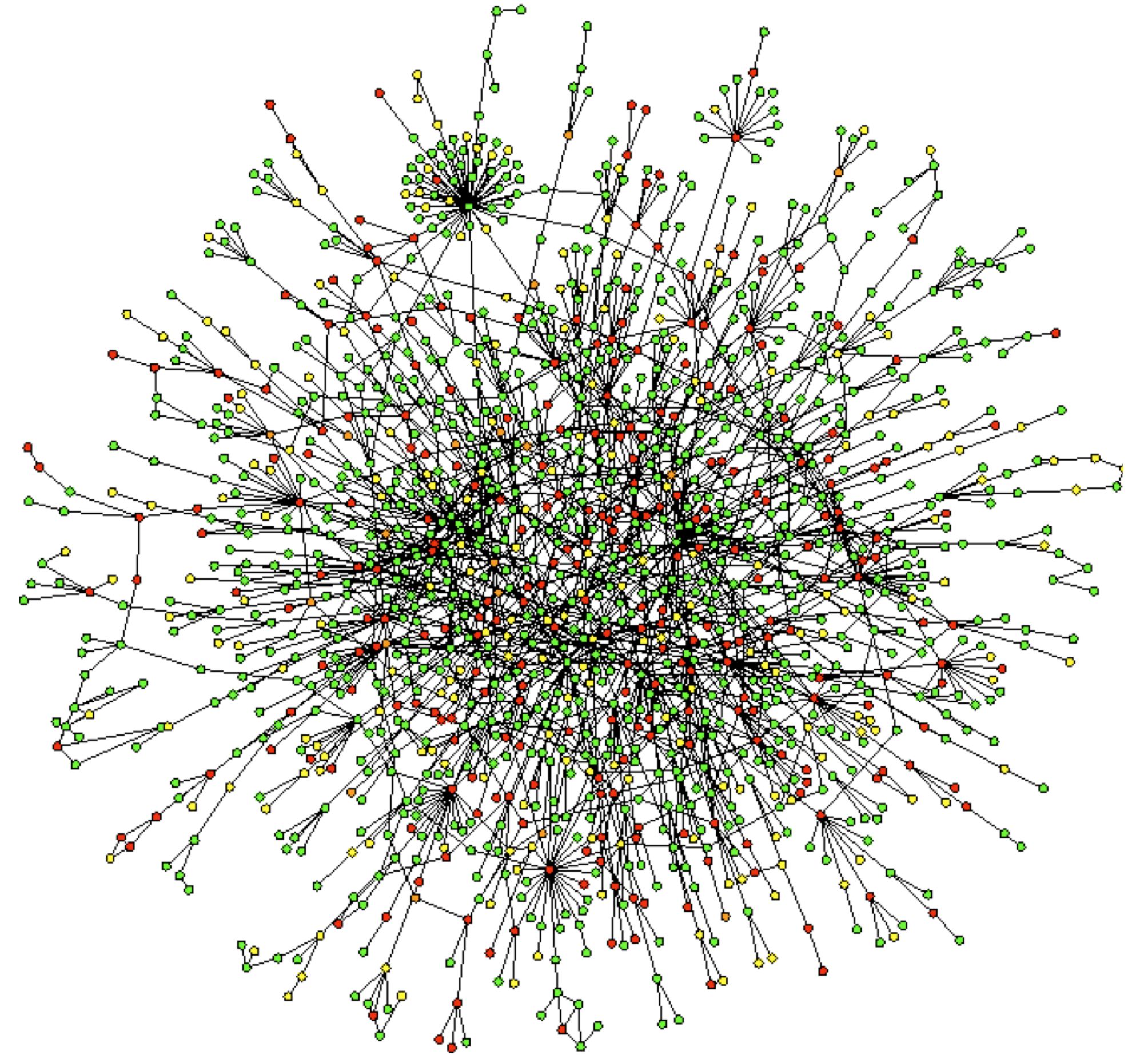
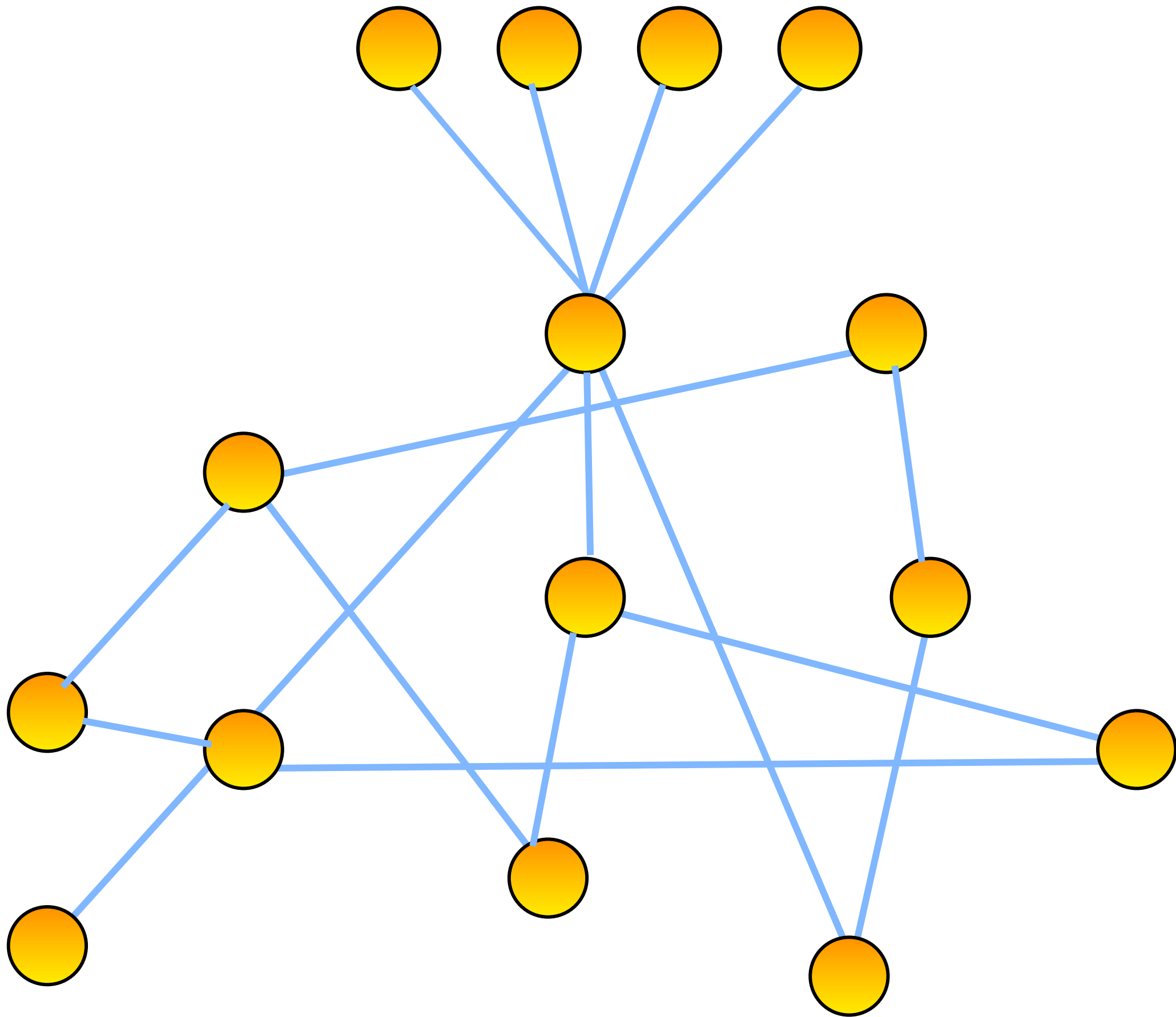


Most real networks have the same internal structure

Why?

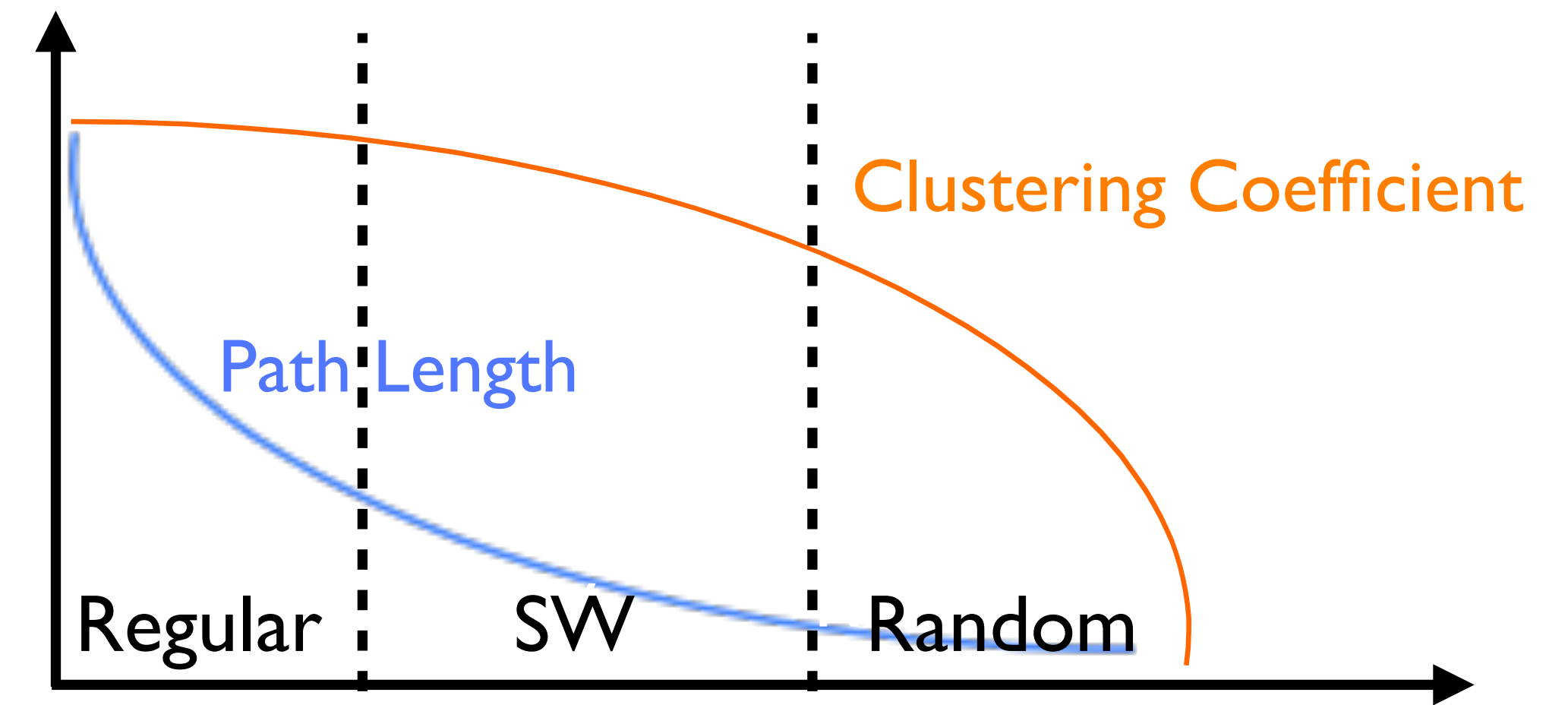
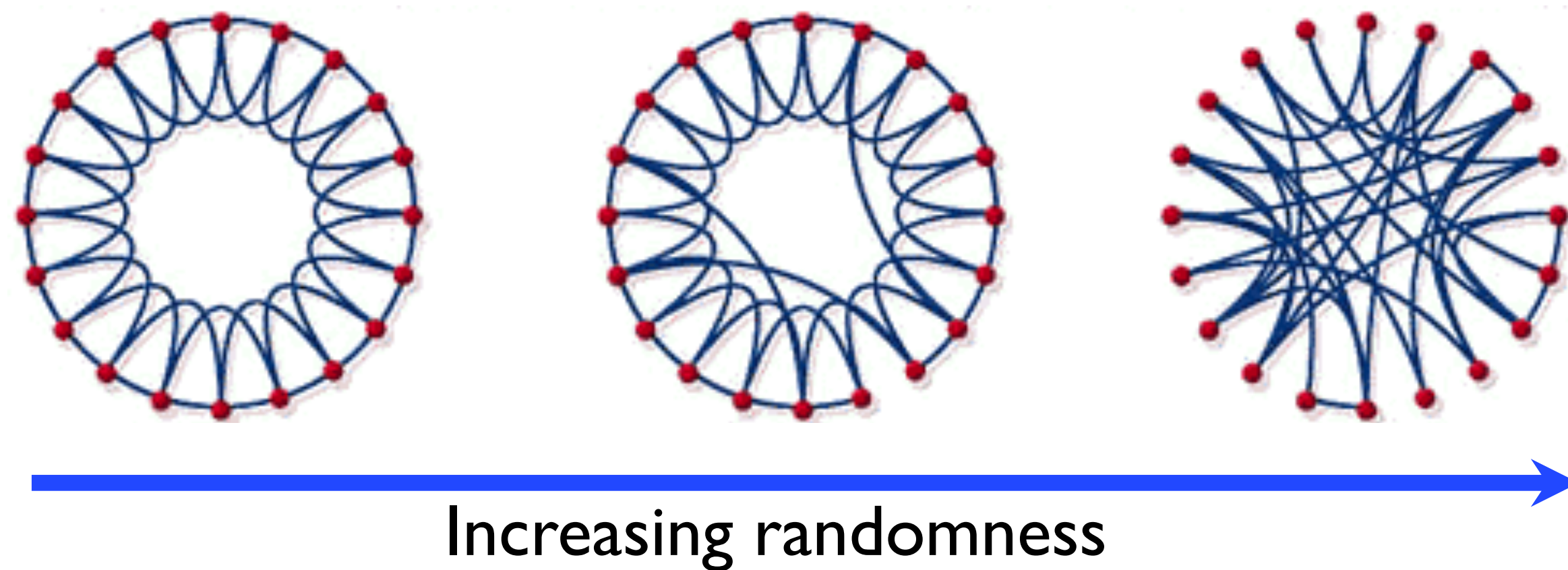
What are the implications?

preferential attachment



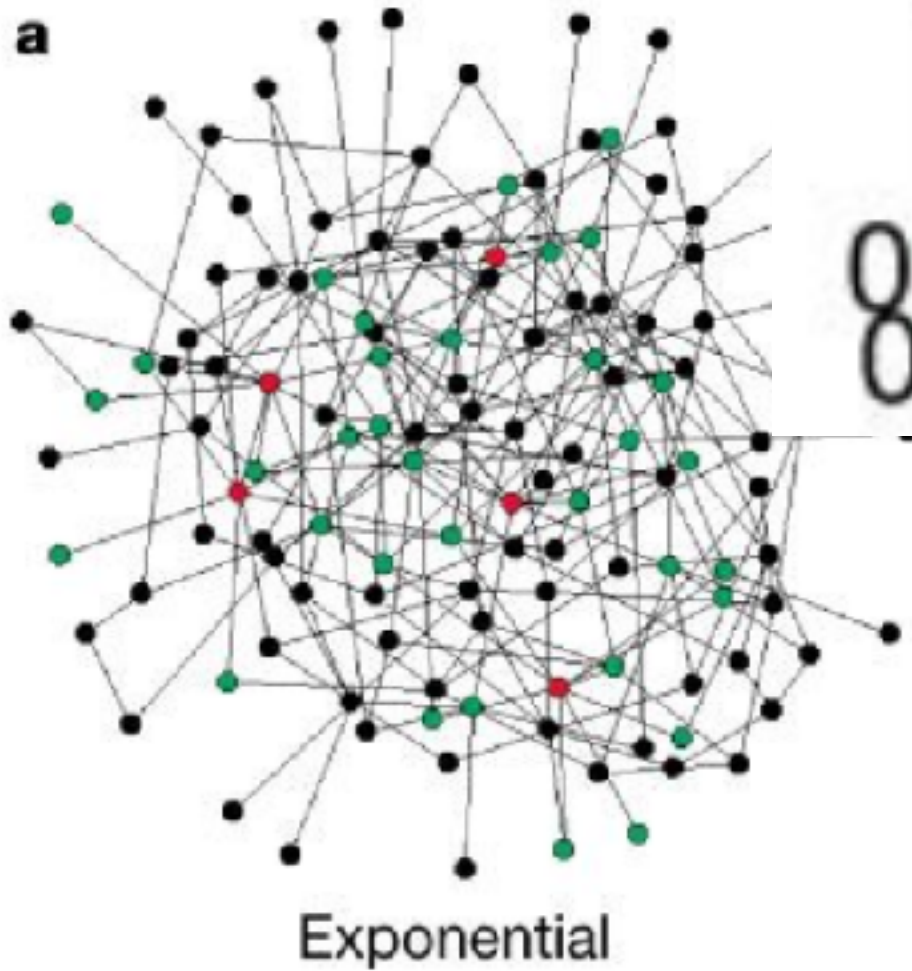
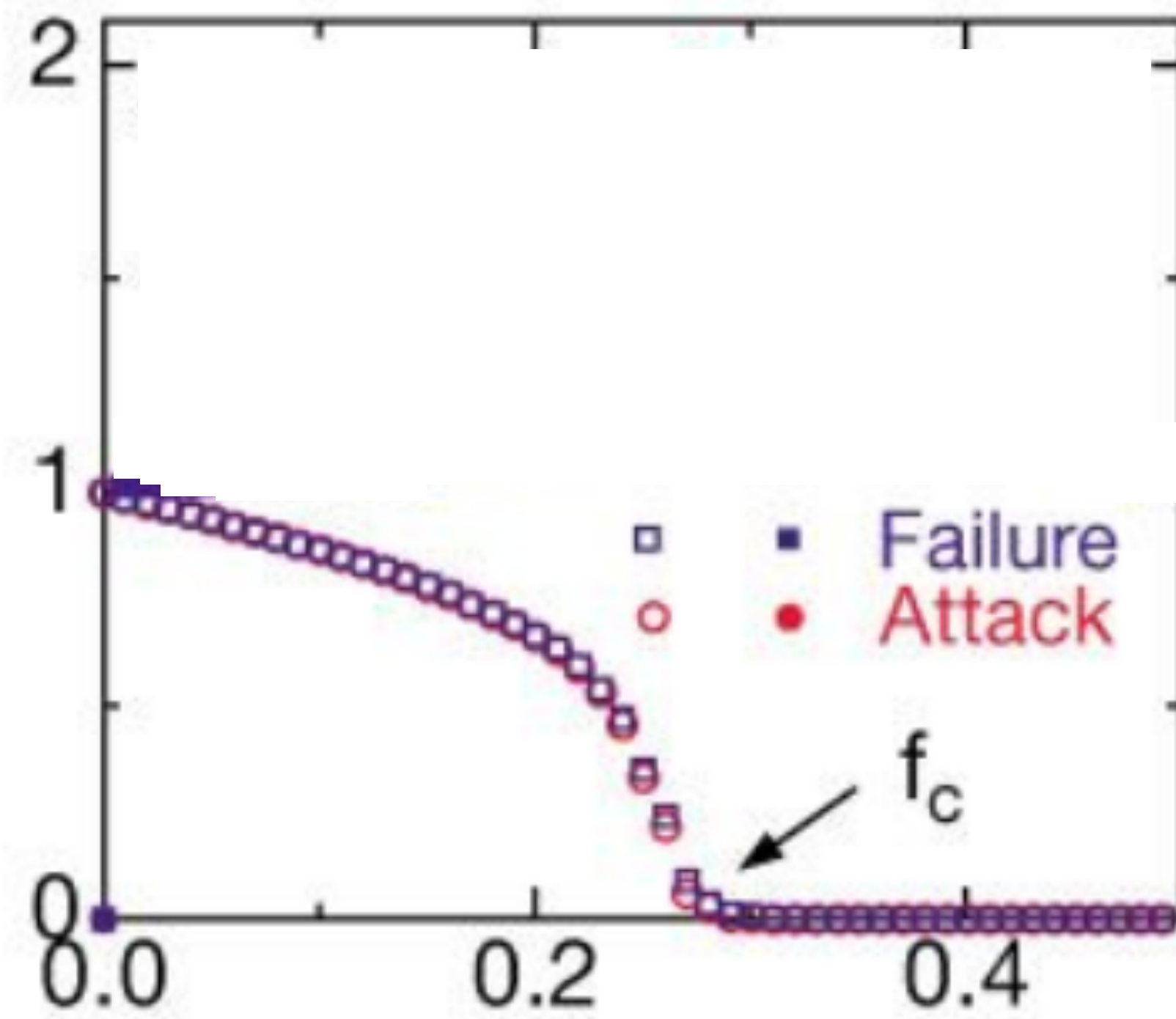
Rich get richer!

back to the small world



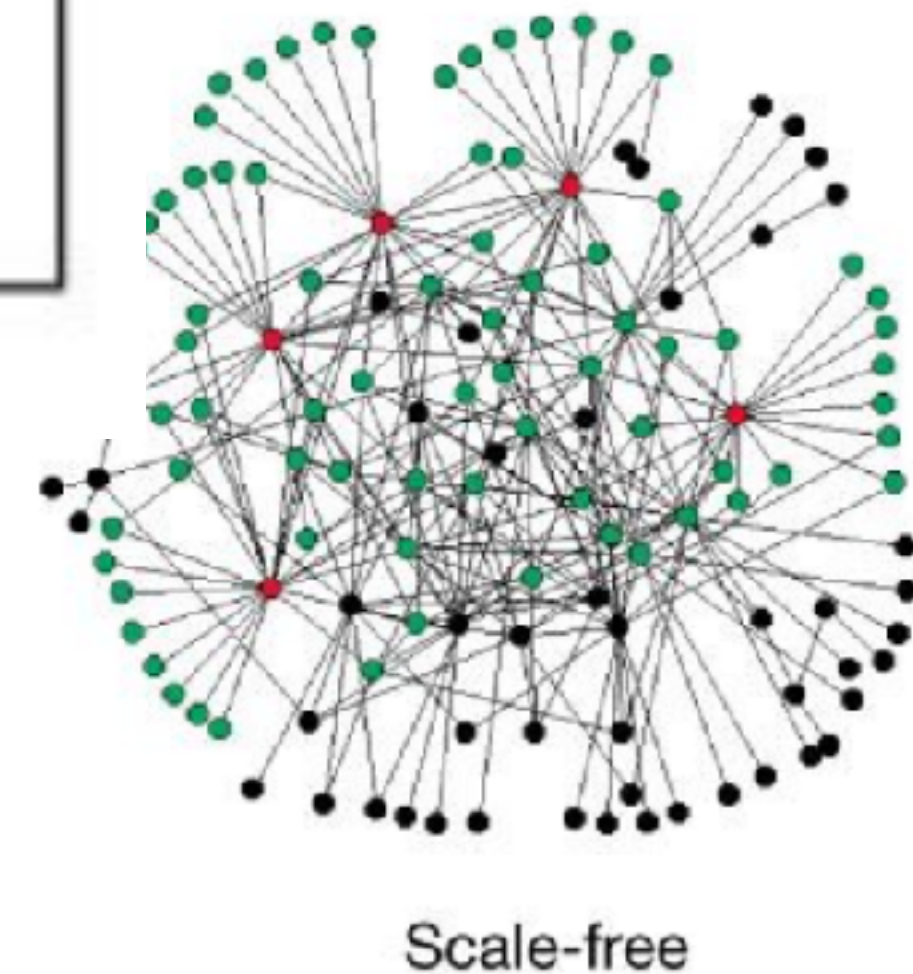
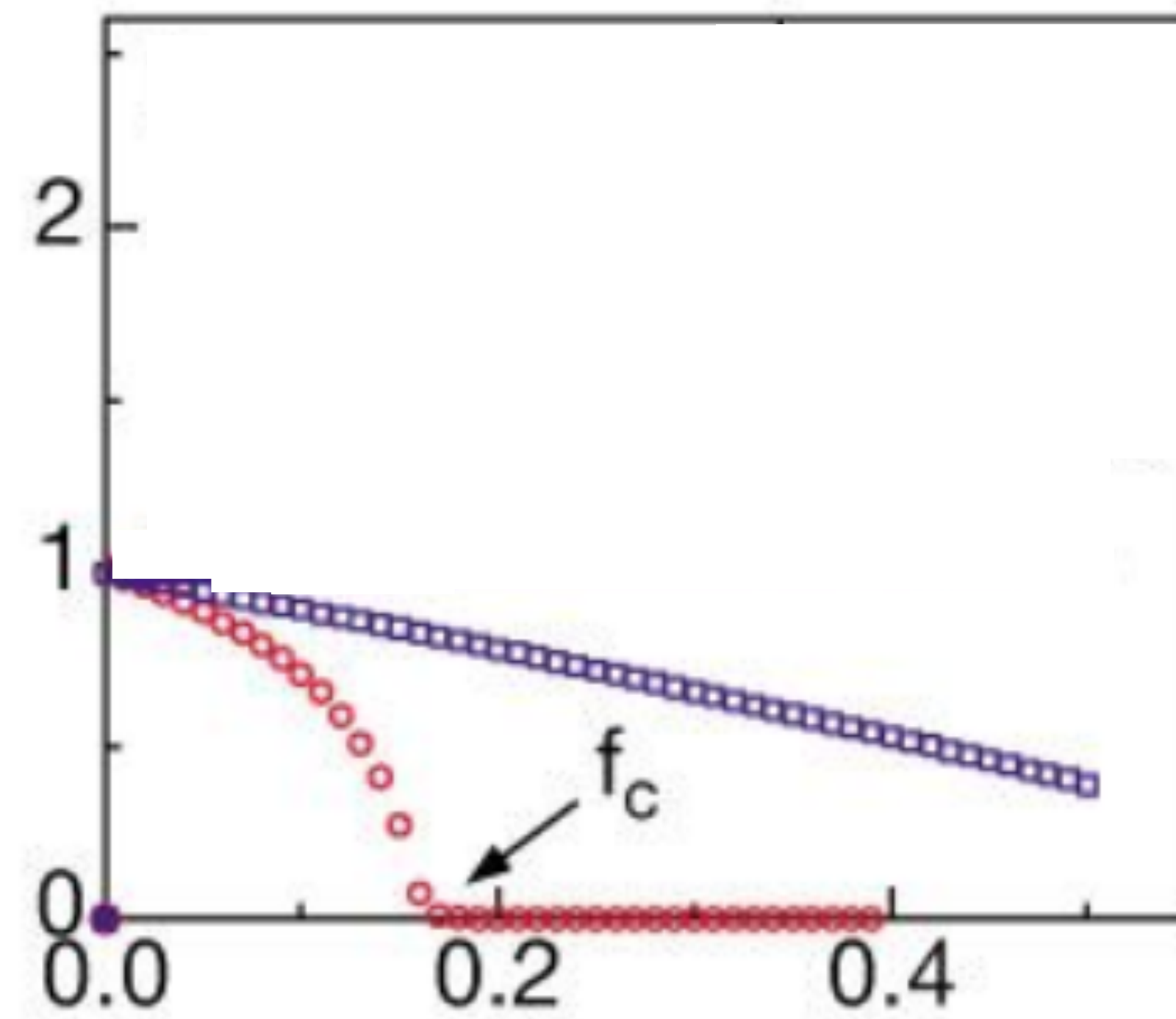
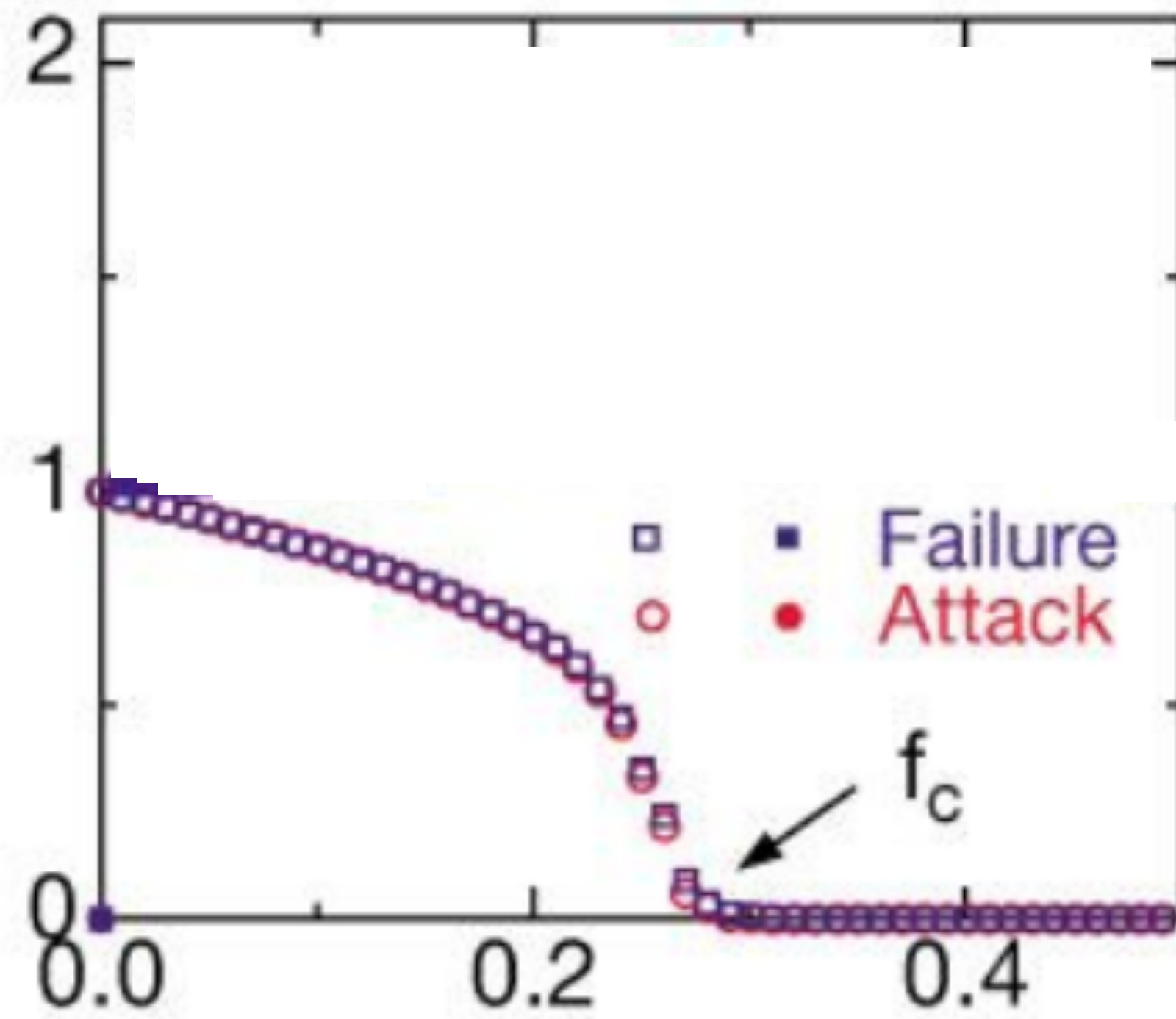
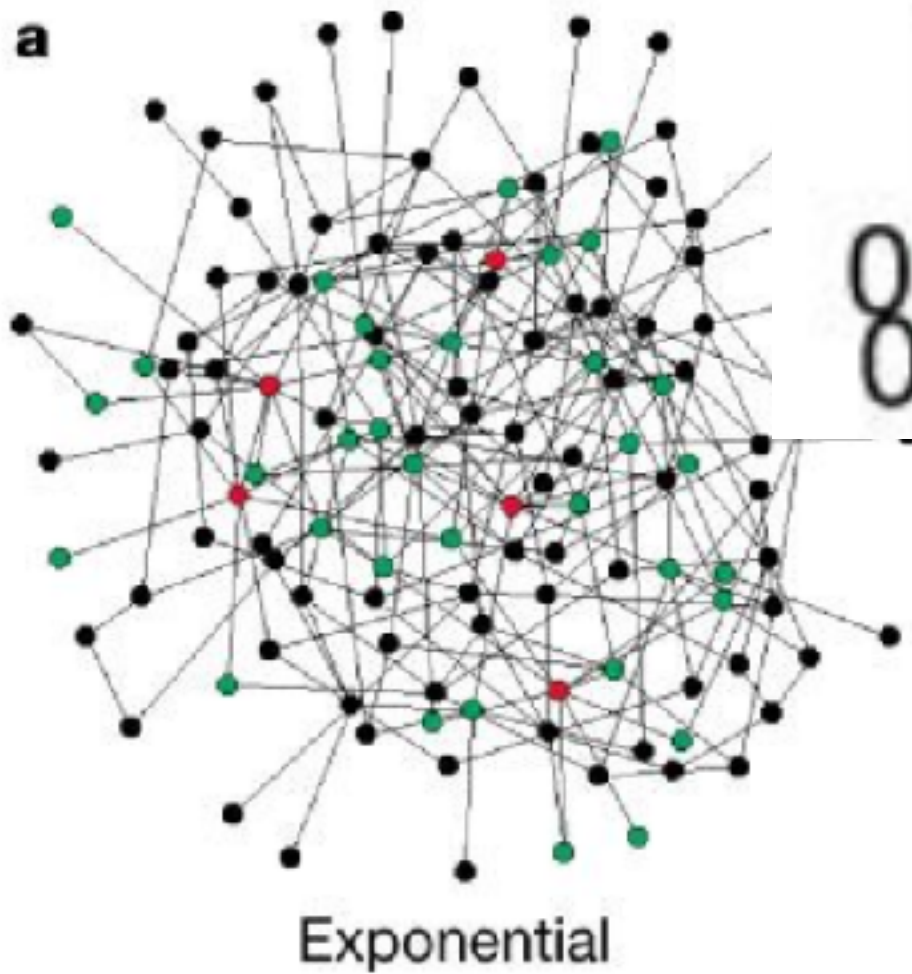
Watts and Strogatz (1998)

network robustness



Albert *et al.* (2000)

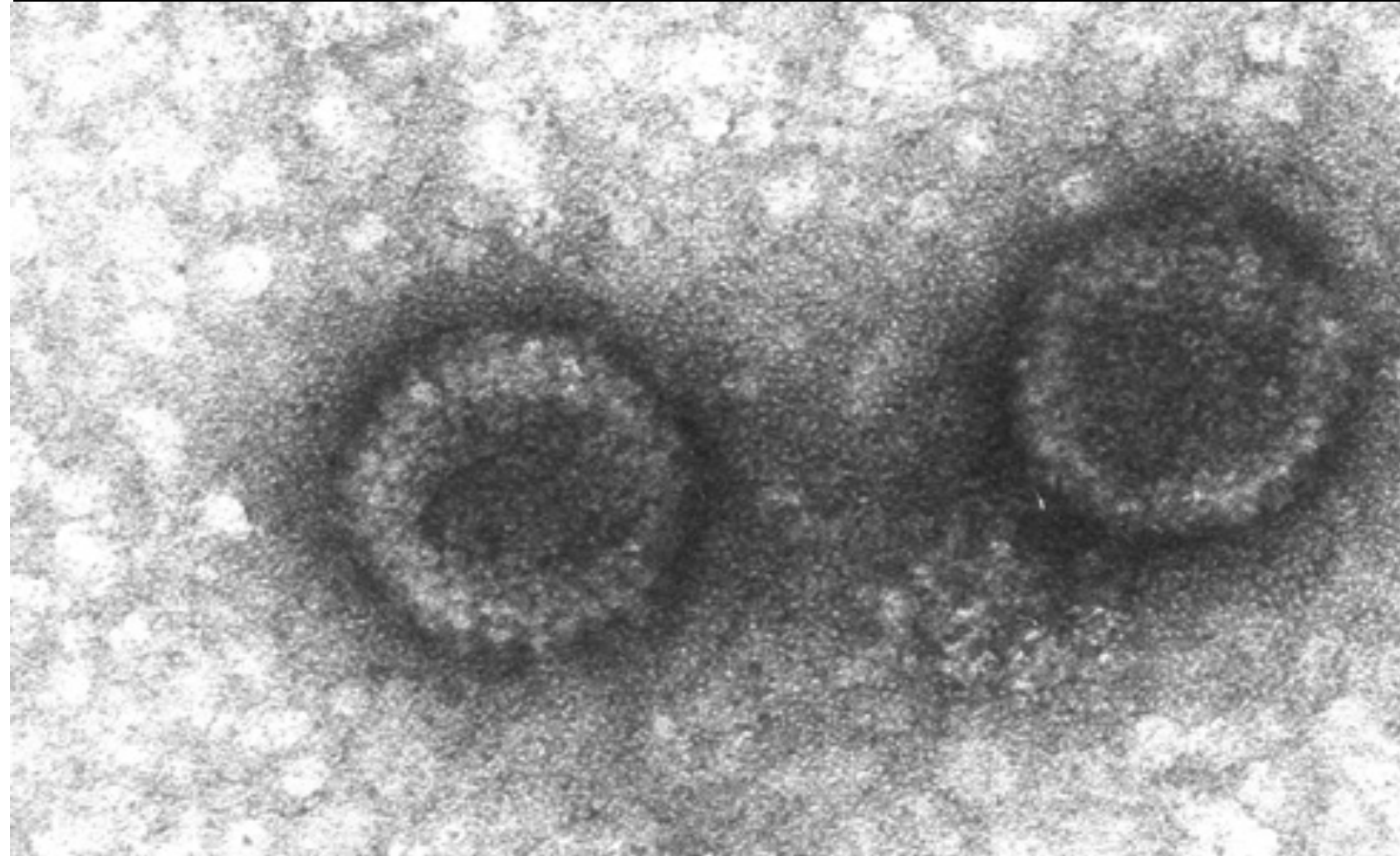
network robustness



Albert *et al.* (2000)

eradication in viruses

```
00 40 E0 n 0xjw05a000 00
00 1B E5 0`çþ± áβ*óábL-←õ
00 1B E5 ♦ ←õ*óáb<-←õ □ ←õ
58 69 73 *óáb<-←õ3çþþThis
72 6F 6D code arose from
50 65 72 the dust of Per
00 00 00 mutation City
3F FF 1A /// 1βL†fõ⊙ Pβþ →
L2 9F E5 < òõ± éó/// Êβ8†fõ
1D A0 E3 ⊙ Pβ□ → Êβ•↔án
```



infection rate

spreading rate

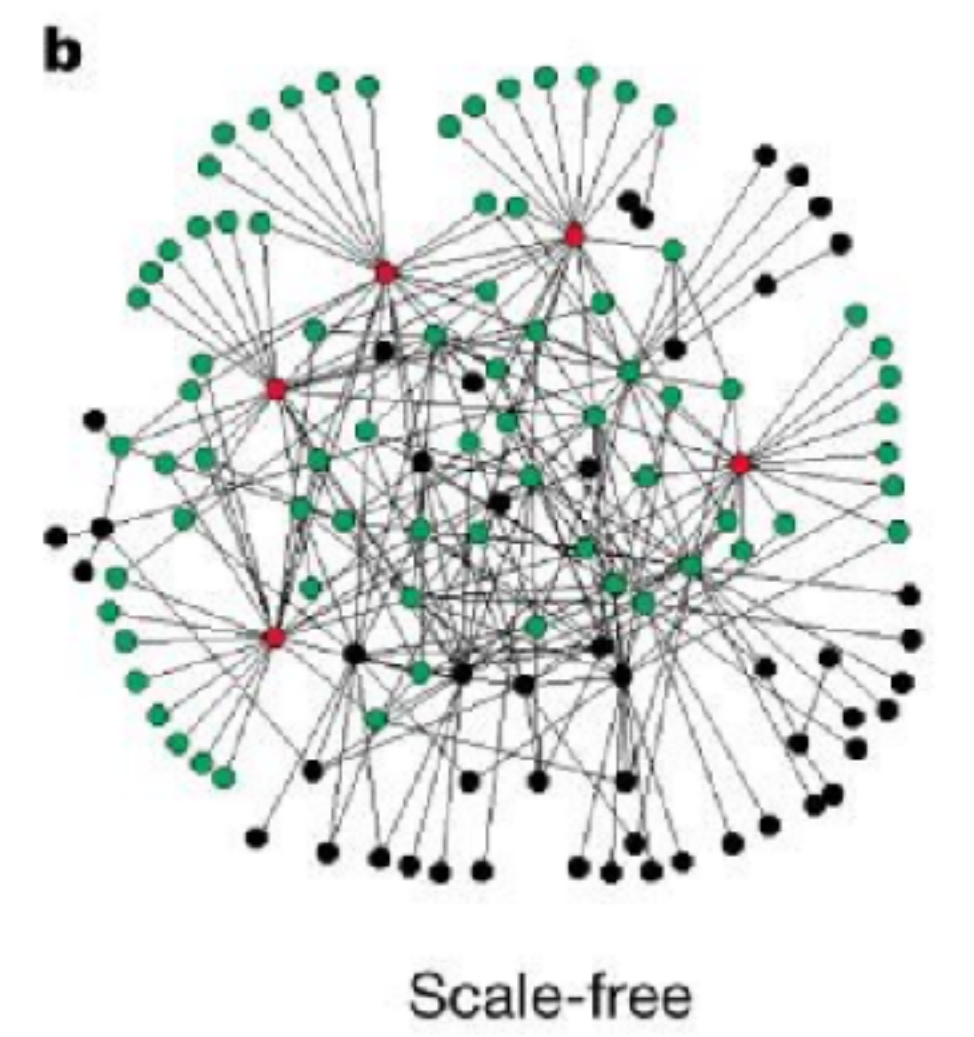
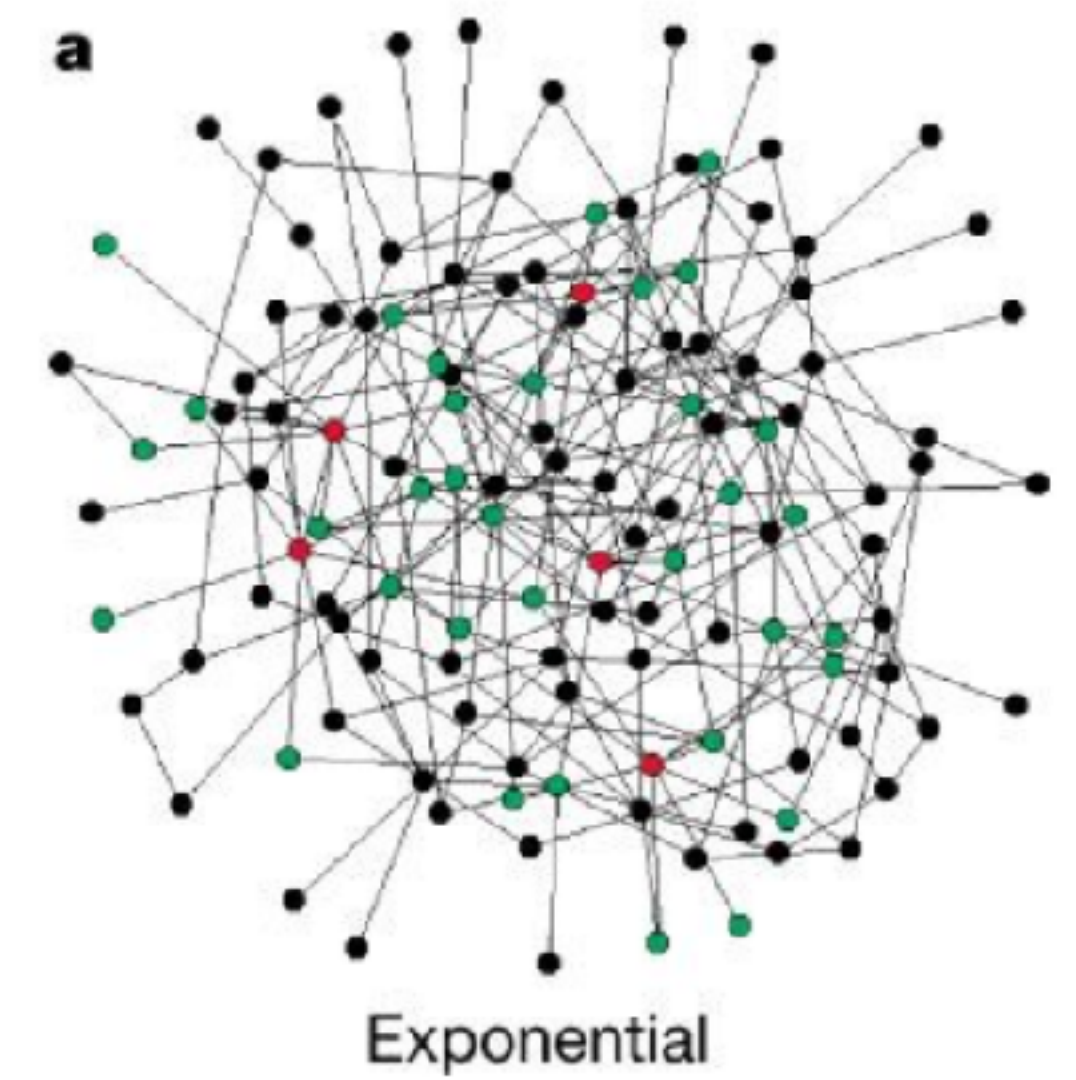
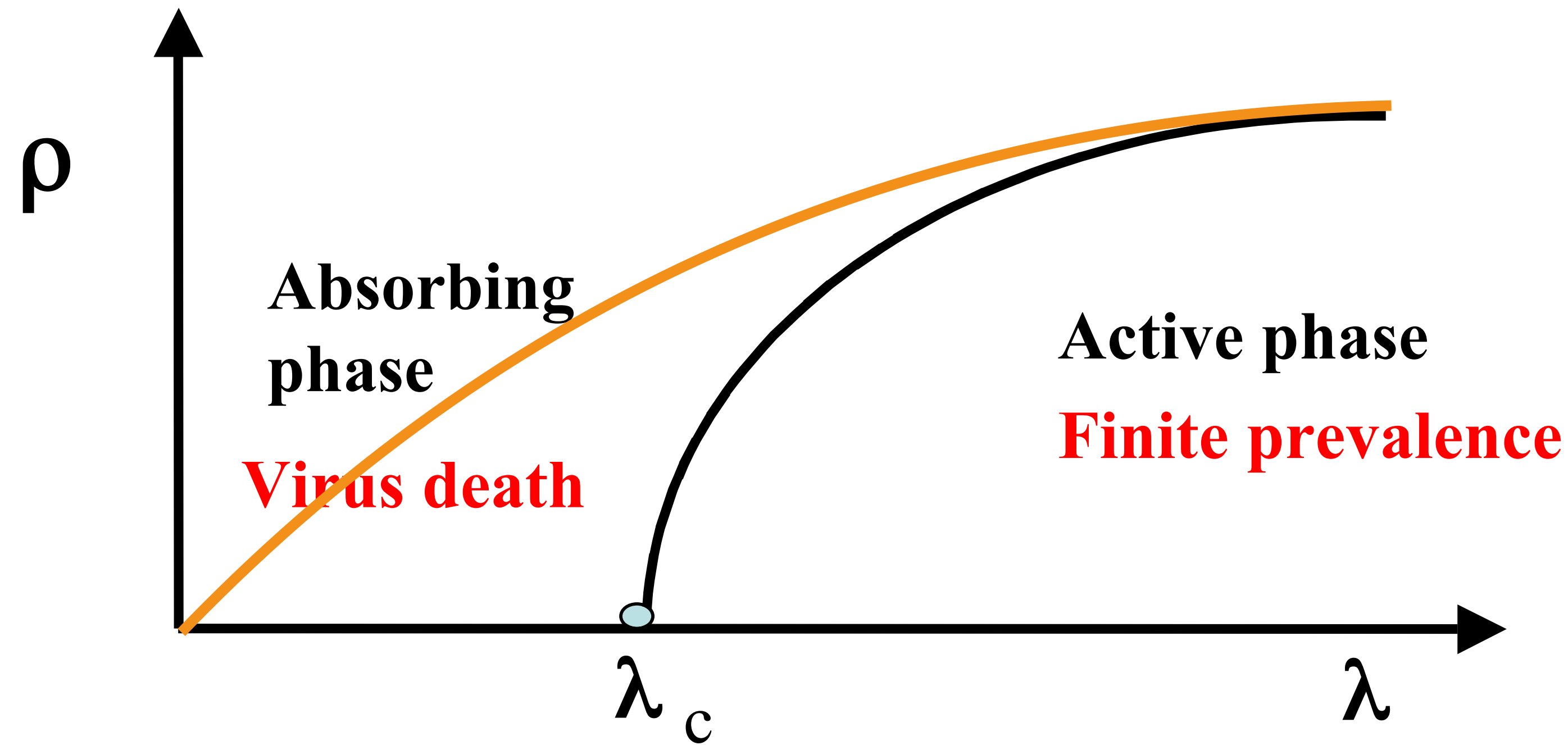
$$\lambda = \frac{\mu}{\delta}$$

recovery rate

prevalence

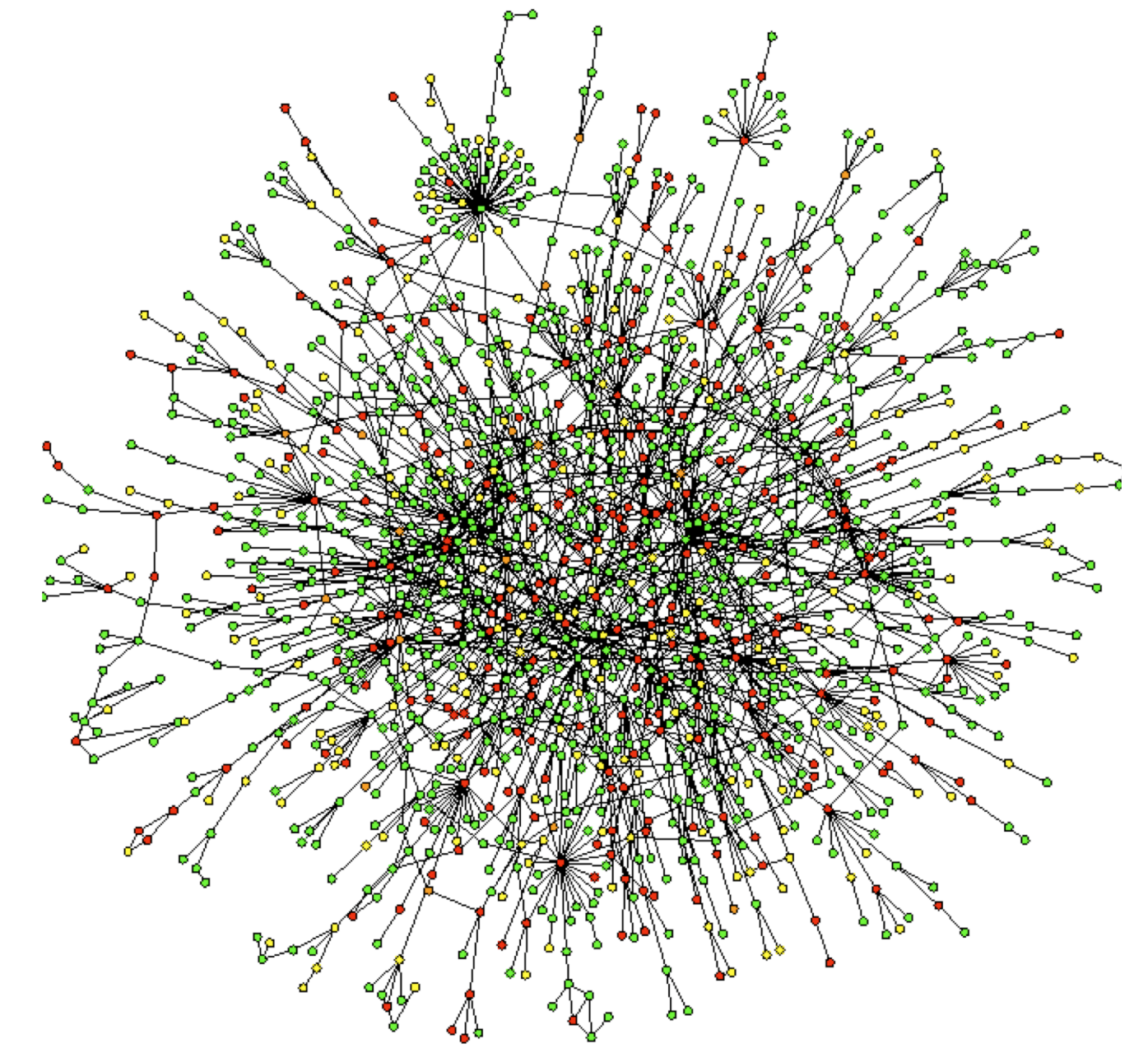
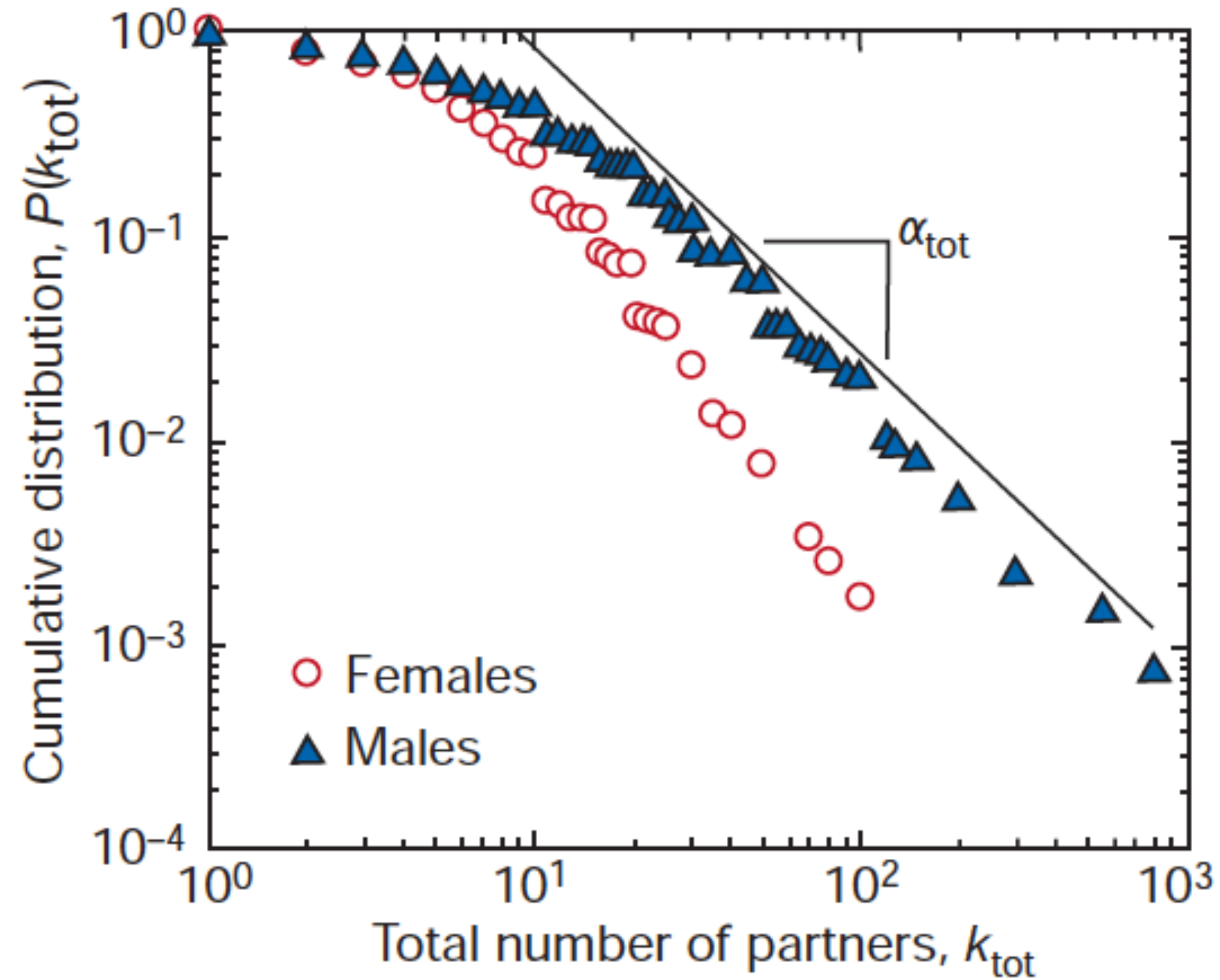
$$\rho = 1 - \frac{\delta}{\mu}$$

eradication in viruses



Pastor Satorras and Vespignani (2001)

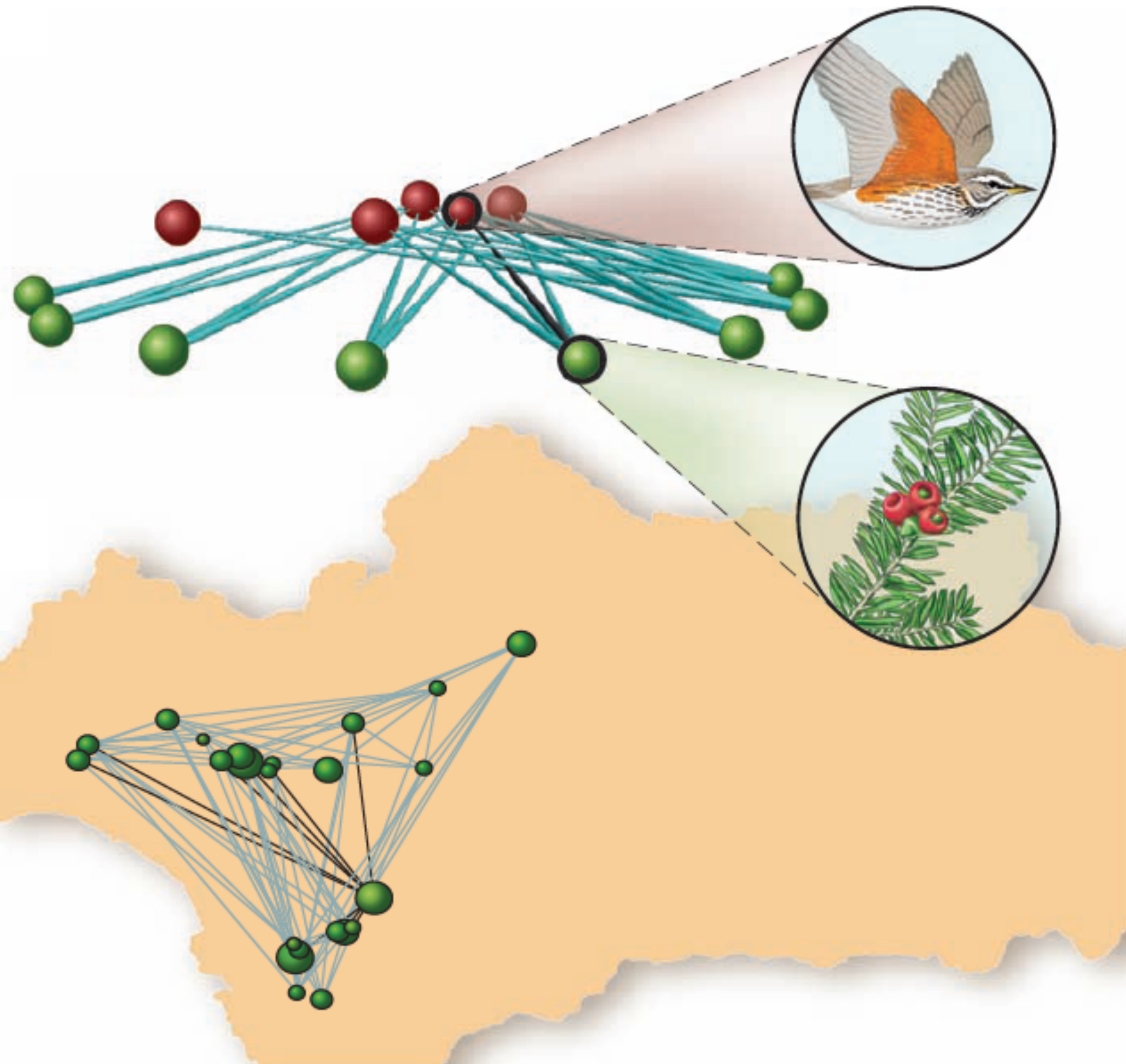
sexually transmitted diseases



Absence of eradication thresholds in scale-free networks

Lijeros *et al.* (2001)

wrapping up: why networks?



- networks allow introducing heterogeneity into our previous homogeneous theories.
- networks put the focus on the patterns of interactions among elements.
- networks allow searching for commonalities among disparate systems.

current/future applications of network theory

Systemic risk in banking ecosystems

Andrew G. Haldane¹ & Robert M. May²

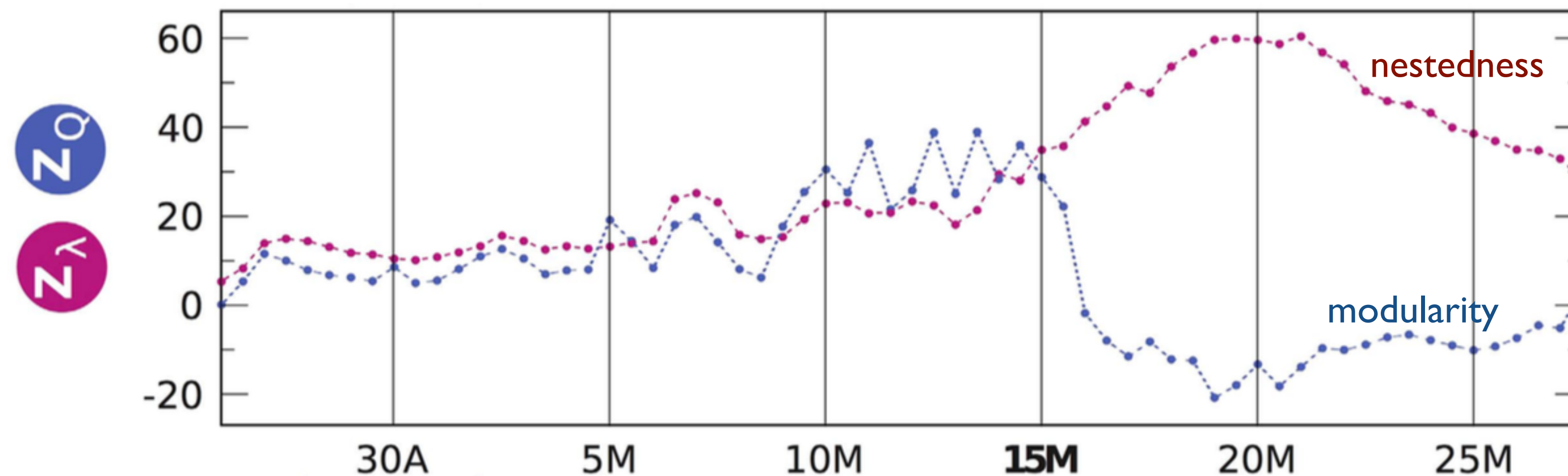
In the run-up to the recent financial crisis, an increasingly elaborate set of financial instruments emerged, intended to optimize returns to individual institutions with seemingly minimal risk. Essentially no attention was given to their possible effects on the stability of the system as a whole. Drawing analogies with the dynamics of ecological food webs and with networks within which infectious diseases spread, we explore the interplay between complexity and stability in deliberately simplified models of financial networks. We suggest some policy lessons that can be drawn from such models, with the explicit aim of minimizing systemic risk.

- *“Looking at financial risk through a network lens indicates a fundamentally different rationale for prudential regulation.”*
- *“In the United Kingdom, the new government have recently set up a Royal Commission to investigate the case for encouraging modularity and diversity in banking ecosystems, as a means of buttressing systemic resilience.”*

OPEN Emergence of consensus as a modular-to-nested transition in communication dynamics

Received: 01 March 2016
Accepted: 28 December 2016

Javier Borge-Holthoefer^{1,2,3}, Raquel A. Baños³, Carlos Gracia-Lázaro³ & Yamir Moreno^{3,4,5}



“Our results show that collective attention around a topic is reached when the user-meme network self-adapts from a modular to a nested structure, which ultimately allows minimizing competition and attaining consensus.”

OPINION

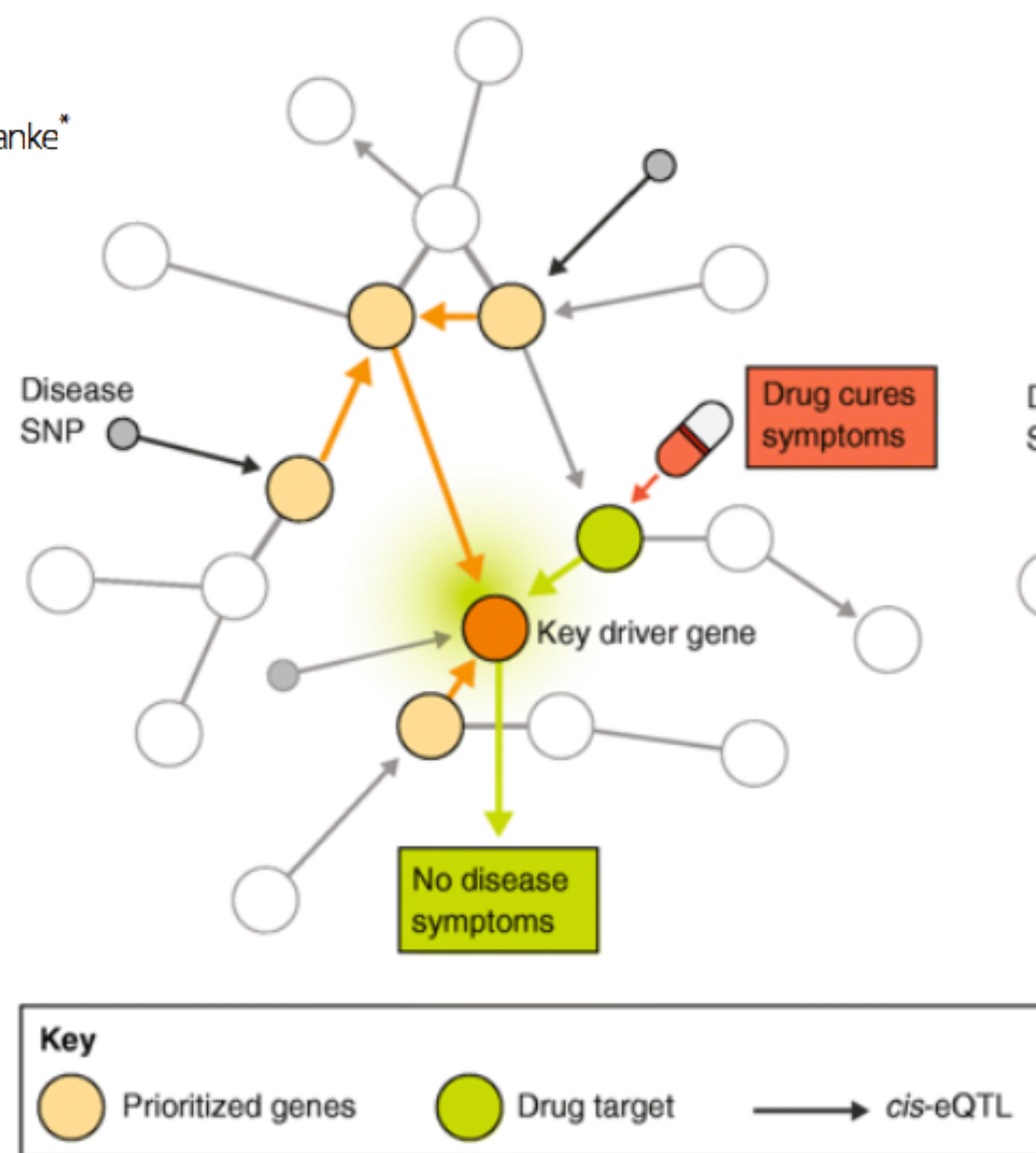
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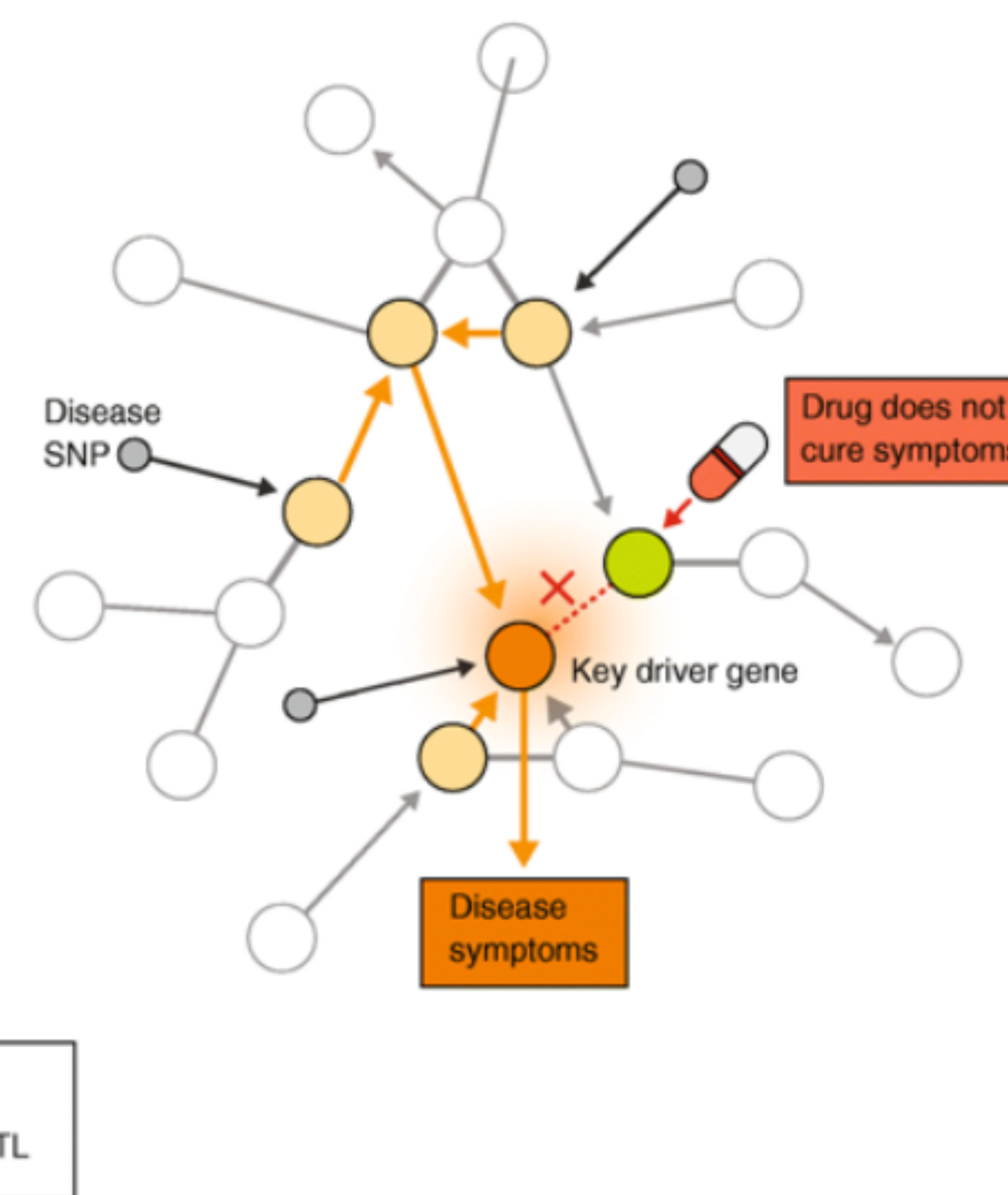
An integrative approach for building personalized gene regulatory networks for precision medicine

Monique G. P. van der Wijst[†], Dylan H. de Vries[†], Harm Brugge, Harm-Jan Westra and Lude Franke^{*}

inter-individual variation in drug response driven by differences in each patient's gene regulatory networks



the drug target gene activates the key driver gene



interaction between the drug target gene and the key driver gene is absent

(3) Personalized medicine