

Lecture 4

In the short run, there are plenty of good problems about the nonlinear dynamics of systems coupled according to small-world, scale-free or generalized random connectivity. The speculations that these architectures are dynamically advantageous need to be sharpened, then confirmed or refuted mathematically for specific examples. Other ripe topics include the design of self-healing networks, and the relationships among optimization principles, network growth rules and network...

In the longer run, network thinking will become essential to all branches of science as we struggle to interpret the data pouring in from neurobiology, genomics, ecology, finance and the World-Wide Web.

Steven Strogatz
Exploring complex networks
Nature 2001

Review

- Common Properties:
- small-world effect
- transitivity or clustering
- network resilience
- mixing patterns

Can have a profound effect on the structural properties of a network!

Today

- Mixing patterns (scalar properties)
- degree correlation
- community structure

Mixing patterns

<i>E</i>		woman			
		black	hispanic	white	other
men	black	506	32	69	26
	hispanic	23	308	114	38
	white	26	46	599	68
	other	10	14	47	32

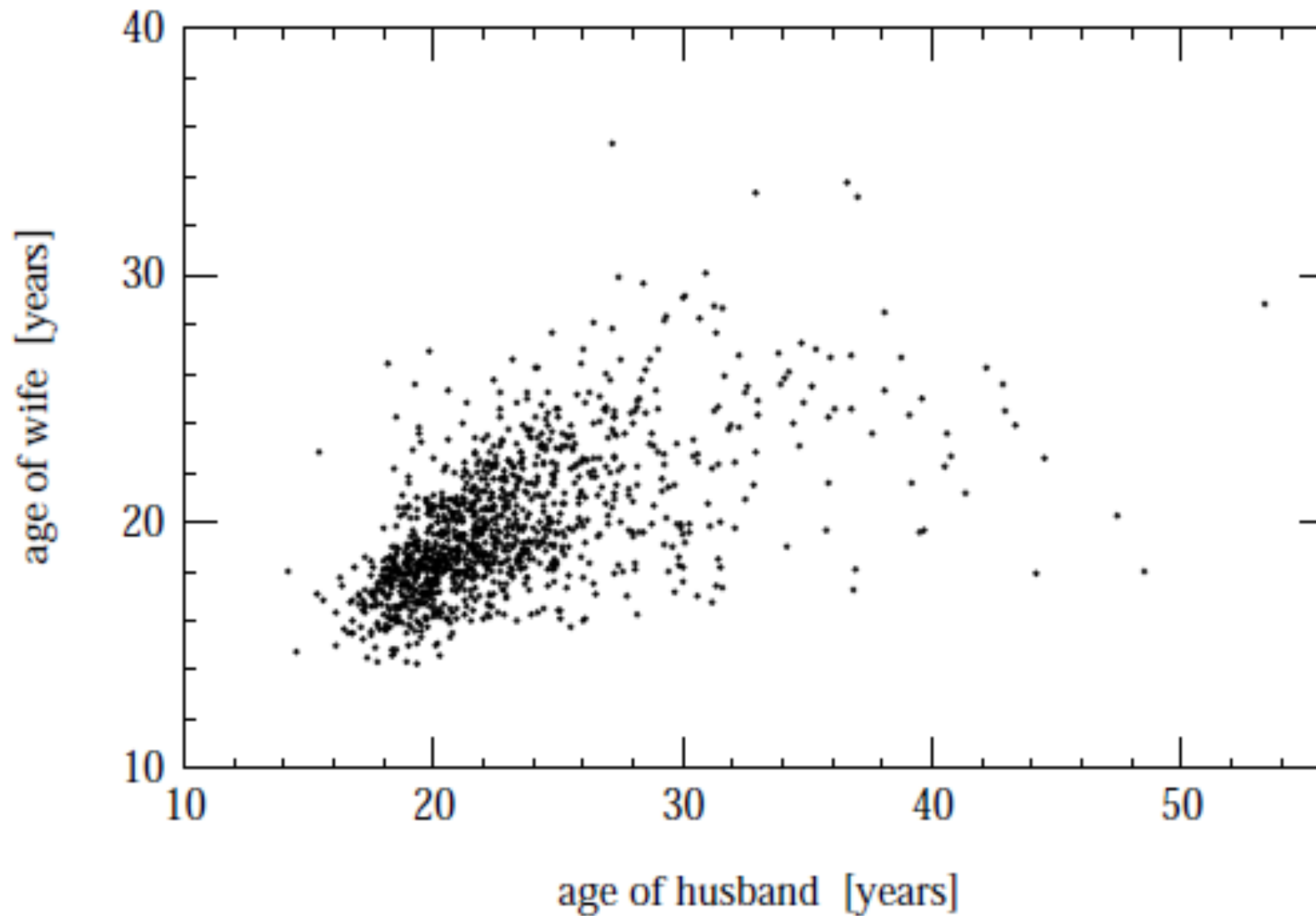
		woman			
		black	hispanic	white	other
men	black	0.258	0.016	0.035	0.013
	hispanic	0.012	0.157	0.058	0.019
	white	0.013	0.023	0.306	0.035
	other	0.005	0.007	0.024	0.016

$$\begin{aligned}
 P(j|i) &= \frac{e_{ij}}{\sum_j e_{ij}} \\
 &= \frac{0.258}{0.258 + 0.016 + 0.035 + 0.013} \\
 &= 0.79
 \end{aligned}$$

Mixing by Scalar Properties

e.g.

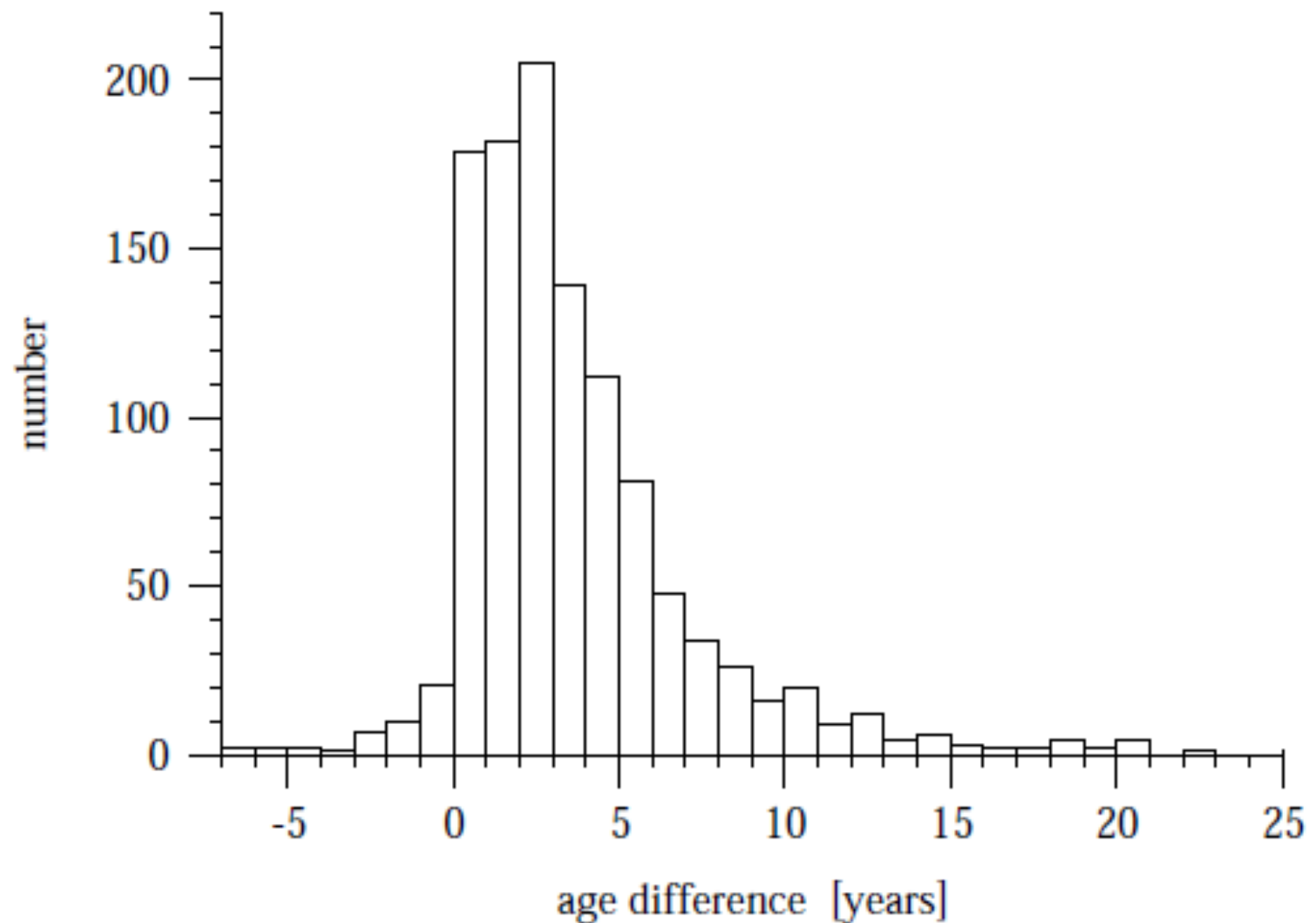
1141 married couples at time of marriage



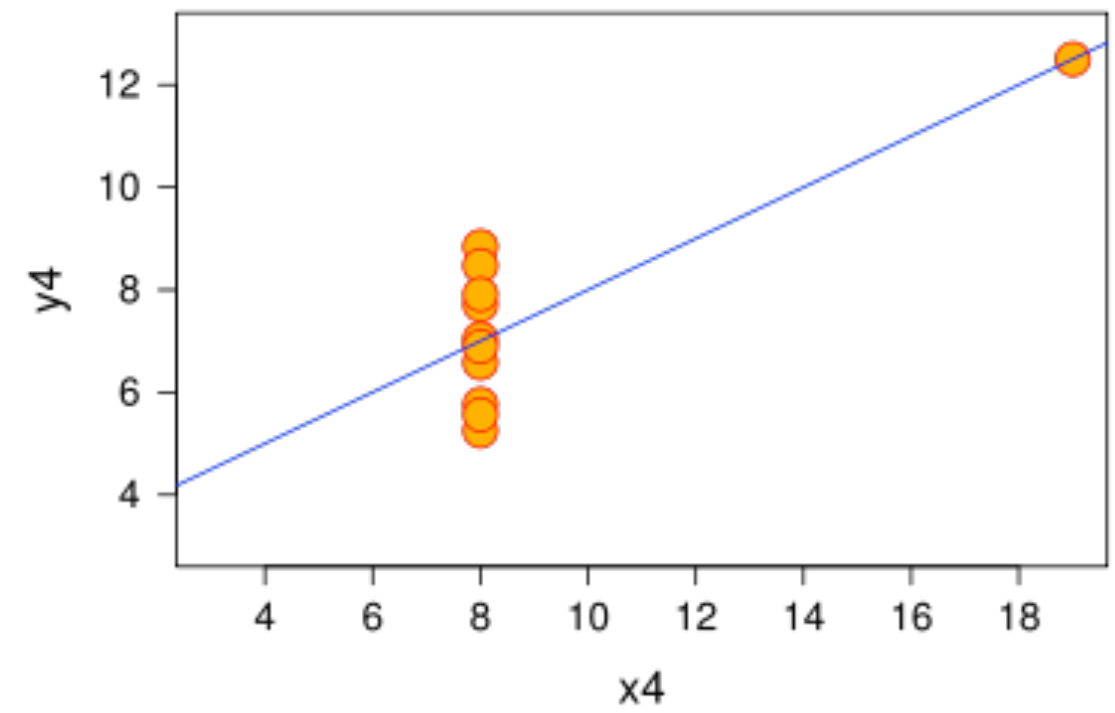
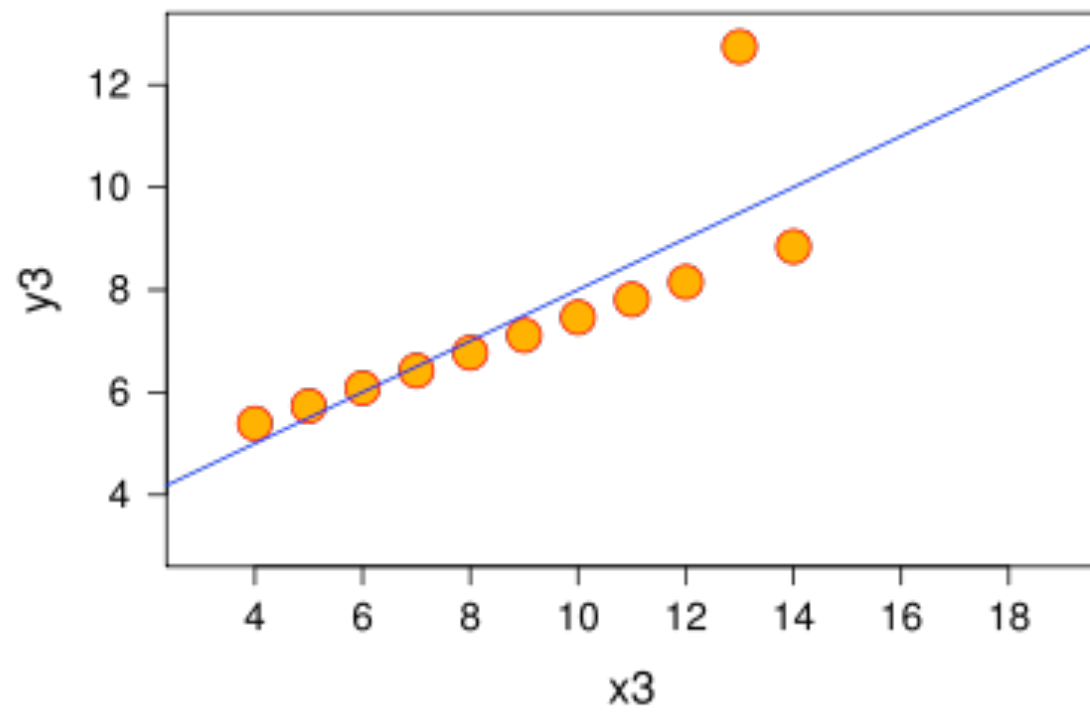
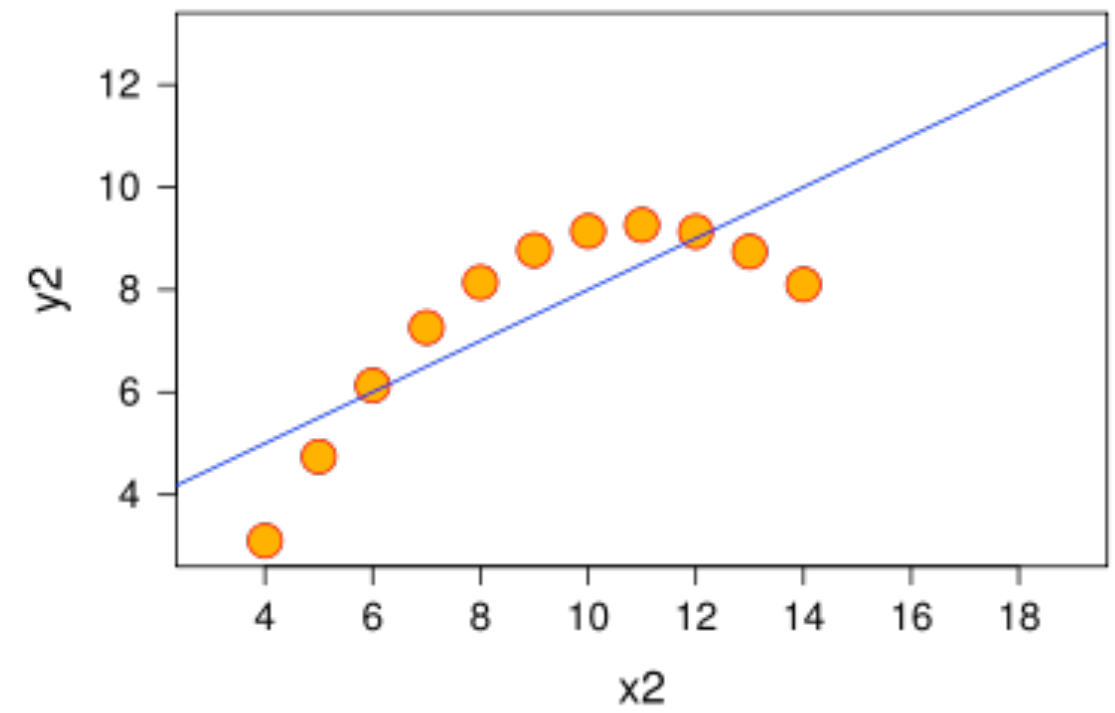
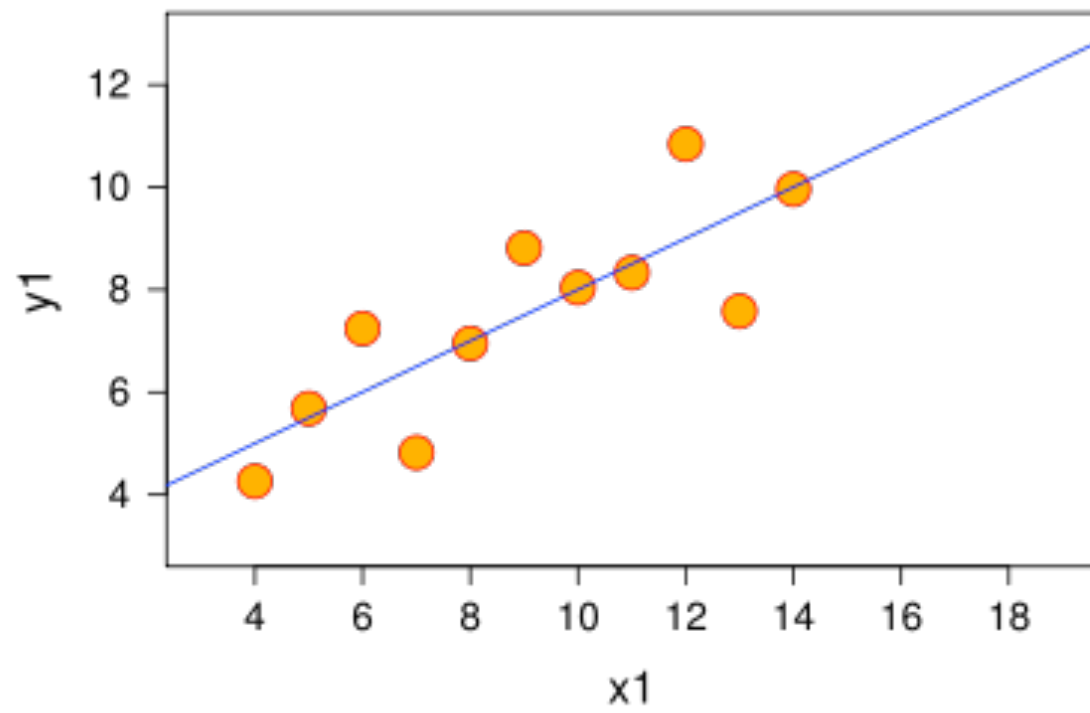
strong positive correlation between the ages

M. E. J. Newman, Mixing patterns in networks, Phys. Rev. E, 67, 016128, 2003

1141 married couples at time of marriage



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The four y variables have the same mean (7.5), standard deviation (4.12), correlation (0.816) and regression line ($y = 3 + 0.5x$)

Degree Correlation

network	type	n	m	z	ℓ	α	$C^{(1)}$	$C^{(2)}$	r
film actors	undirected	449 913	25 516 482	113.43	3.48	2.3	0.20	0.78	0.208
company directors	undirected	7 673	55 392	14.44	4.60	–	0.59	0.88	0.276
math coauthorship	undirected	253 339	496 489	3.92	7.57	–	0.15	0.34	0.120
physics coauthorship	undirected	52 909	245 300	9.27	6.19	–	0.45	0.56	0.363
biology coauthorship	undirected	1 520 251	11 803 064	15.53	4.92	–	0.088	0.60	0.127
telephone call graph	undirected	47 000 000	80 000 000	3.16		2.1			
email messages	directed	59 912	86 300	1.44	4.95	1.5/2.0		0.16	
email address books	directed	16 881	57 029	3.38	5.22	–	0.17	0.13	0.092
student relationships	undirected	573	477	1.66	16.01	–	0.005	0.001	−0.029
sexual contacts	undirected	2 810				3.2			
WWW nd.edu	directed	269 504	1 497 135	5.55	11.27	2.1/2.4	0.11	0.29	−0.067
WWW Altavista	directed	203 549 046	2 130 000 000	10.46	16.18	2.1/2.7			
citation network	directed	783 339	6 716 198	8.57		3.0/–			
Roget's Thesaurus	directed	1 022	5 103	4.99	4.87	–	0.13	0.15	0.157
word co-occurrence	undirected	460 902	17 000 000	70.13		2.7		0.44	
Internet	undirected	10 697	31 992	5.98	3.31	2.5	0.035	0.39	−0.189
power grid	undirected	4 941	6 594	2.67	18.99	–	0.10	0.080	−0.003
train routes	undirected	587	19 603	66.79	2.16	–		0.69	−0.033
software packages	directed	1 439	1 723	1.20	2.42	1.6/1.4	0.070	0.082	−0.016
software classes	directed	1 377	2 213	1.61	1.51	–	0.033	0.012	−0.119
electronic circuits	undirected	24 097	53 248	4.34	11.05	3.0	0.010	0.030	−0.154
peer-to-peer network	undirected	880	1 296	1.47	4.28	2.1	0.012	0.011	−0.366
metabolic network	undirected	765	3 686	9.64	2.56	2.2	0.090	0.67	−0.240
protein interactions	undirected	2 115	2 240	2.12	6.80	2.4	0.072	0.071	−0.156
marine food web	directed	135	598	4.43	2.05	–	0.16	0.23	−0.263
freshwater food web	directed	92	997	10.84	1.90	–	0.20	0.087	−0.326
neural network	directed	307	2 359	7.68	3.97	–	0.18	0.28	−0.226
									+

Possible explanations

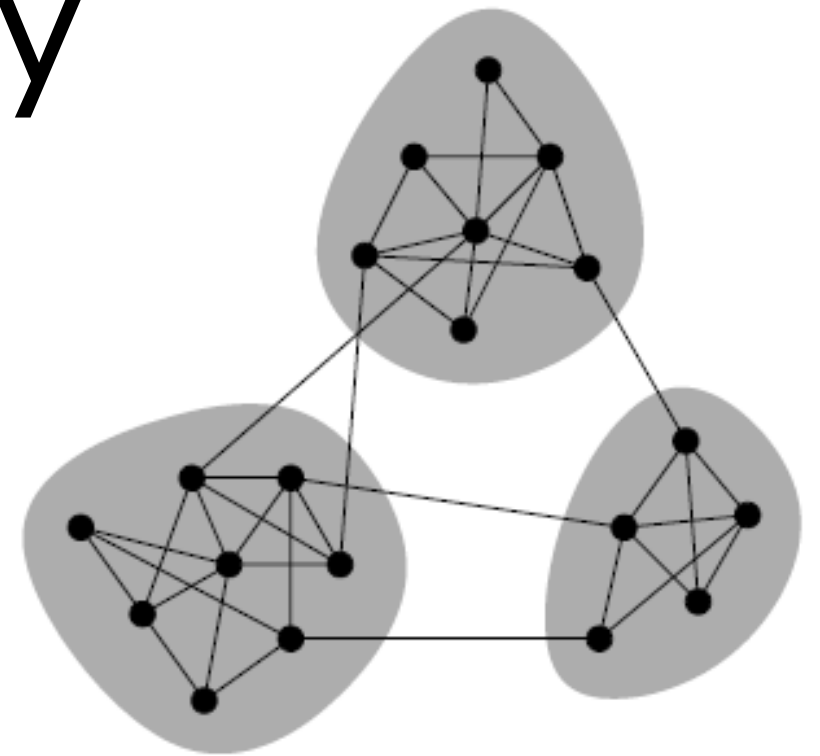
“In the social networks it is entirely possible, and is often assumed in the sociological literature, that similar people do attract one another, and therefore that there could be a real preference among gregarious people for association with other gregarious people, and similarly for hermits.”

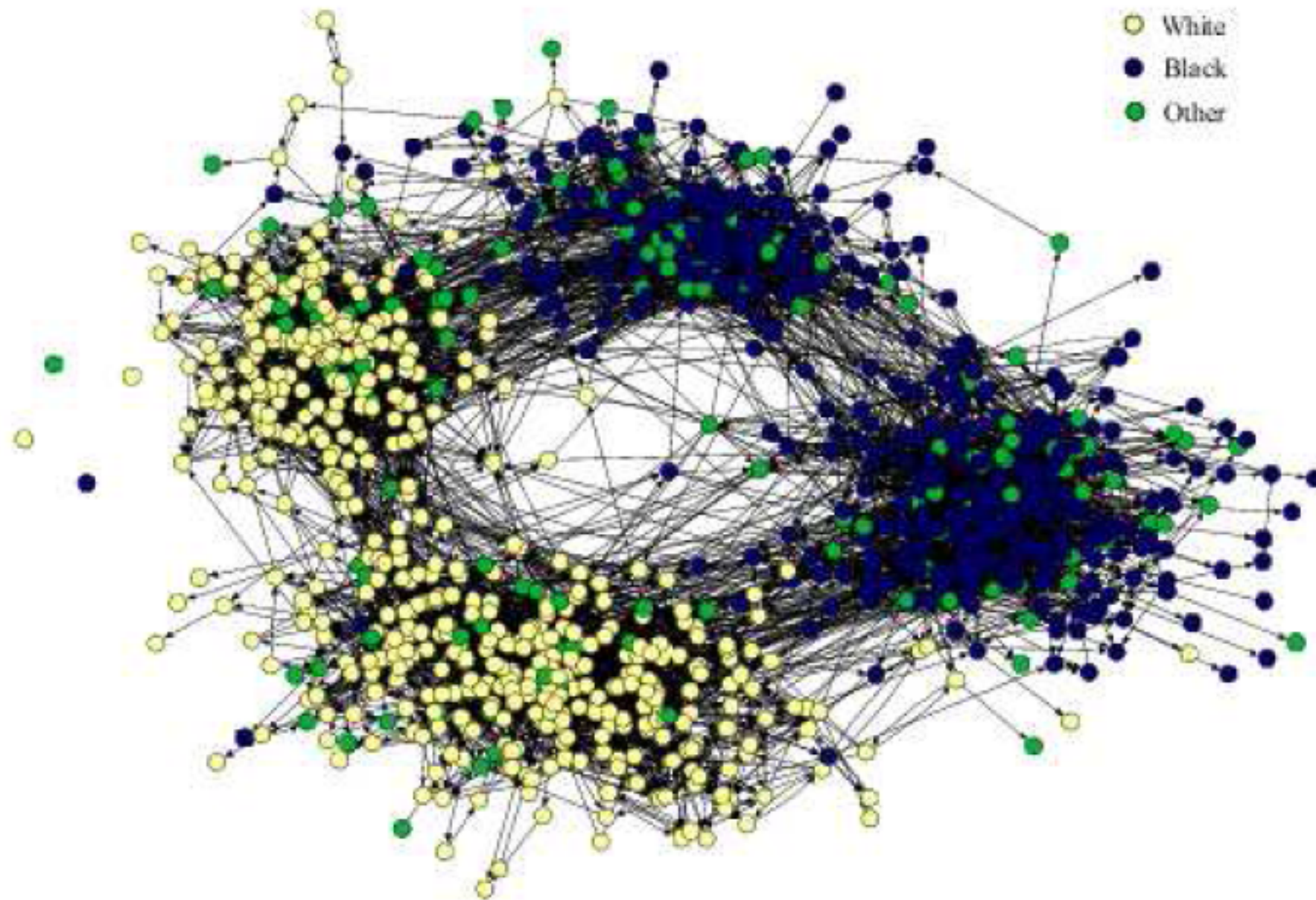
“In the Internet and the World-Wide Web there may be organizational reasons for degree anti-correlation between vertices. The high-degree vertices in these networks are often connectivity providers (Internet) or directories (Web), which by definition tend to be connected to the “little people”—the individual service subscribers in the case of the Internet or the individual web-pages on the Web.”

M. E. J. Newman, Mixing patterns in networks, Phys. Rev. E, 67, 016128, 2003

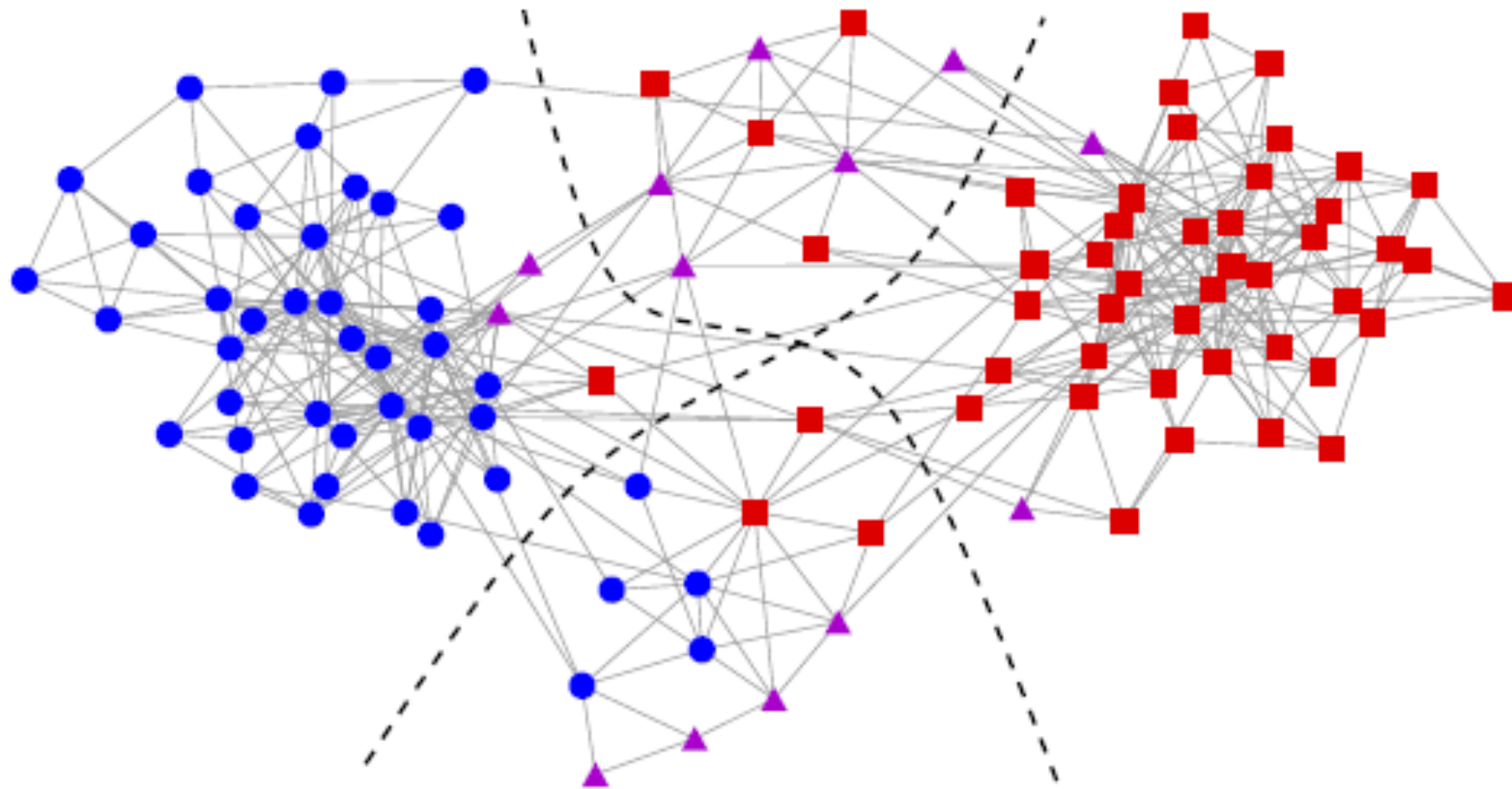
some of the degree correlations we see in our networks could have real social or organizational origins, while others may be artifacts of the types of networks we are looking at and the constraints that are placed on their structure.

Community





Friend ship network in a US school (Spring embedding algorithm)



circles (blue) are liberal, squares (red) are conservative, triangles (purple) are centrist or unaligned.

M. E. J. Newman, Modularity and community structure in networks, Proc. Natl. Acad. Sci. USA 103, 8577-8582 (2006)

insight review articles

Exploring complex networks

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The study of networks pervades all of science, from neurobiology to statistical physics. The most basic issues are structural: how does one characterize the wiring diagram of a food web or the Internet or the metabolic network of the bacterium *Escherichia coli*? Are there any unifying principles underlying their topology? From the perspective of nonlinear dynamics, we would also like to understand how an enormous network of interacting dynamical systems — be they neurons, power stations or lasers — will behave collectively, given their individual dynamics and coupling architecture. Researchers are only now beginning to unravel the structure and dynamics of complex networks.

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Pajek - Program for Large Network Analysis

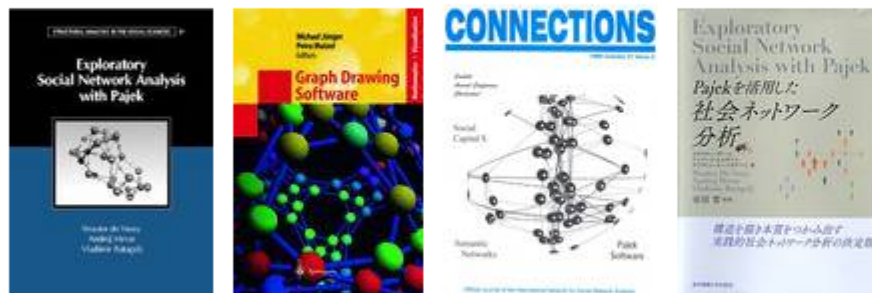
Pajek (Slovene word for Spider) is a program, for Windows, for analysis and visualization of large networks. It is freely available, for noncommercial use, at its [download](#) page. See also a [reference manual](#) for Pajek (in [PDF](#)). The development of Pajek is traced in its [History](#).

Pajek is developed by [Vladimir Batagelj](#) and [Andrej Mrvar](#). Some procedures were contributed also by [Matjaž Zaveršnik](#).

A detailed introduction to Pajek is given in the book W. de Nooy, A. Mrvar, V. Batagelj: *Exploratory Social Network Analysis with Pajek*, [CUP](#), 2005. [Appendix 4](#). ESNA in Japanese, [TDU](#), 2010.

For a short description see Batagelj V., Mrvar A.: Pajek - Analysis and Visualization of Large Networks. in Jünger, M., Mutzel, P., (Eds.) *Graph Drawing Software*. [Springer](#), Berlin 2003. p. 77-103.

An improved version of the paper Batagelj V., Mrvar A.: *Pajek - Program for Large Network Analysis*, presented at Sunbelt'97, was published in [Connections](#) 21(1998)2, 47-57 - ([preprint](#) [Prison Kinemage](#)).



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Next class

Mathematical models of networks

- Random graph model (generalized)
- Markov graphs

Then...

- Small-worlds models
- Models of growing networks