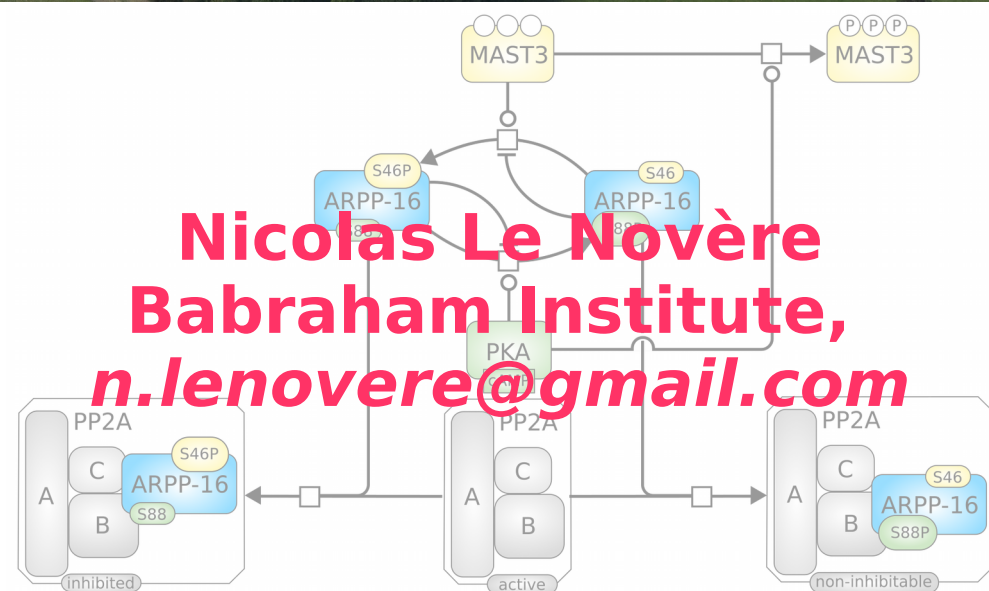
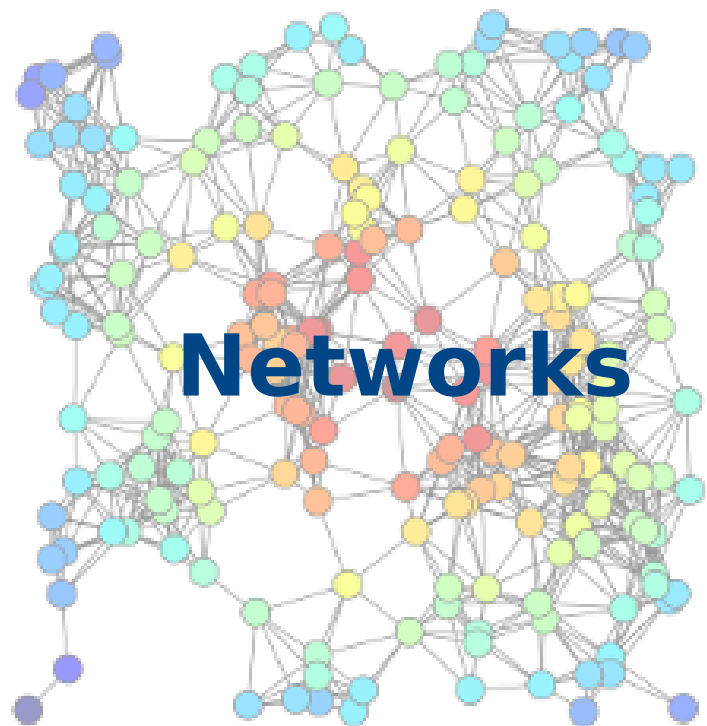




Who am I? - Education



Scientific baccalaureate at the *Prytanée National Militaire*

"I was in one of the most famous schools of Europe"
René Descartes – Discours de la méthode

Preparatory classes for the French *Grandes Écoles*

Mathematics

Physics
Chemistry

Natural
Sciences



Bachelor in Physiology and Cellular Biology
Master of Biology-Biochemistry

UNIVERSITÉ
PARIS
SUD
Evolution

Molecular
Biophysics



Neuroendocrinology

Paris

Who am I? - Research

“A tale of two cities”

1992 1999



PhD
Nicotinic receptors:
Bioinformatics,
Neuroanatomy
Behaviour

2001 2003



CNRS Researcher
Nicotinic receptors:
Bioinformatics

2014-2015
20%



Chief data officer

1999 2001



EMBO post-doc
Bacterial chemotaxis:
Bioinformatics,
Mathematical Modelling

2003 2012



Group leader
Systems Biology
Synaptic signalling:
Modelling, knowledge
Representation, databases

2012-



Senior group leader

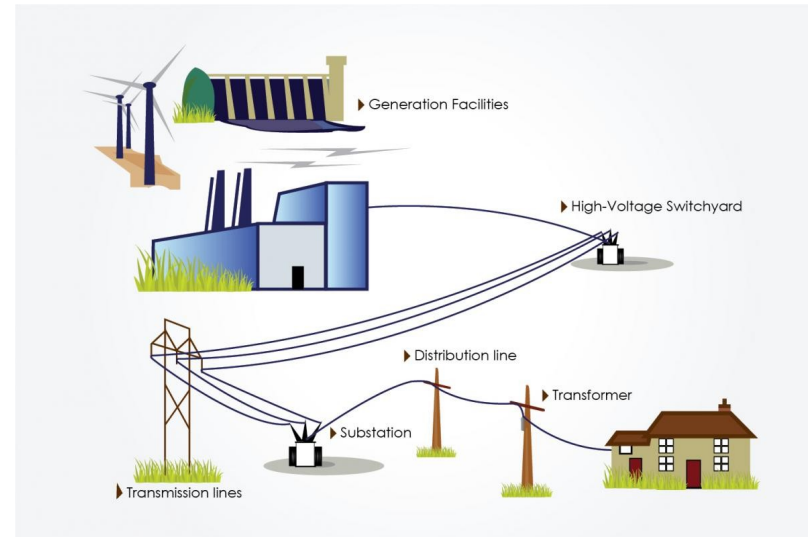
Signalling	50%
Epigenetic	30%
Lymphocytes	15%
Nuclear Dynamics	5%

Cambridge

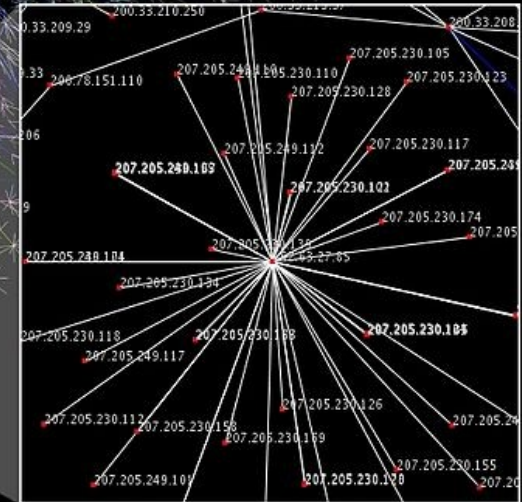
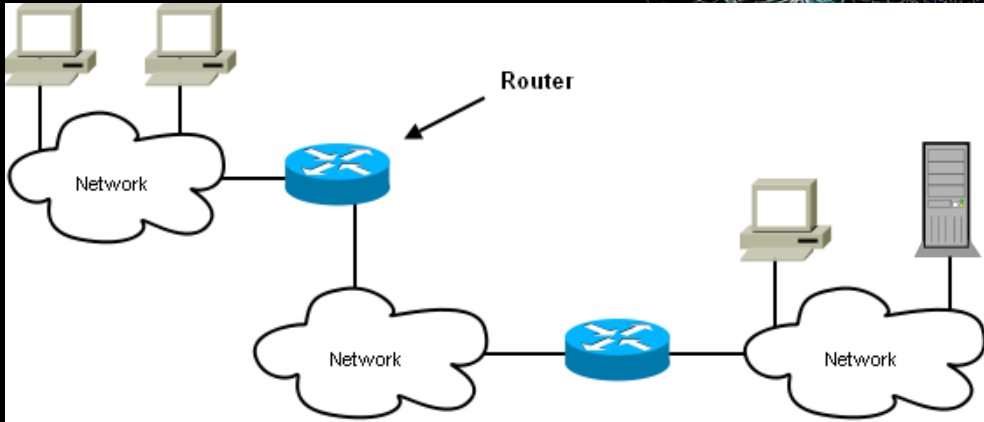
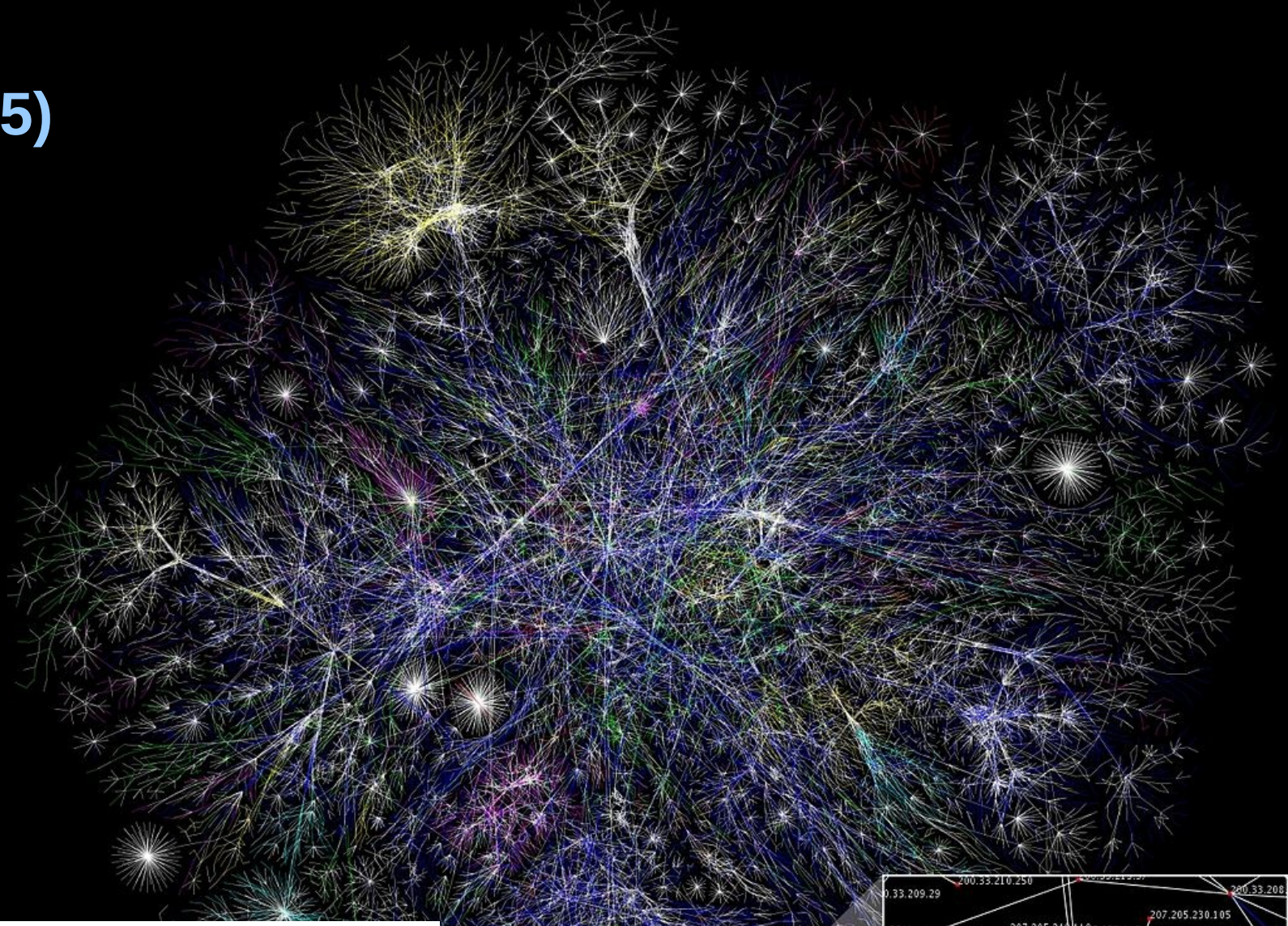


Networks are everywhere

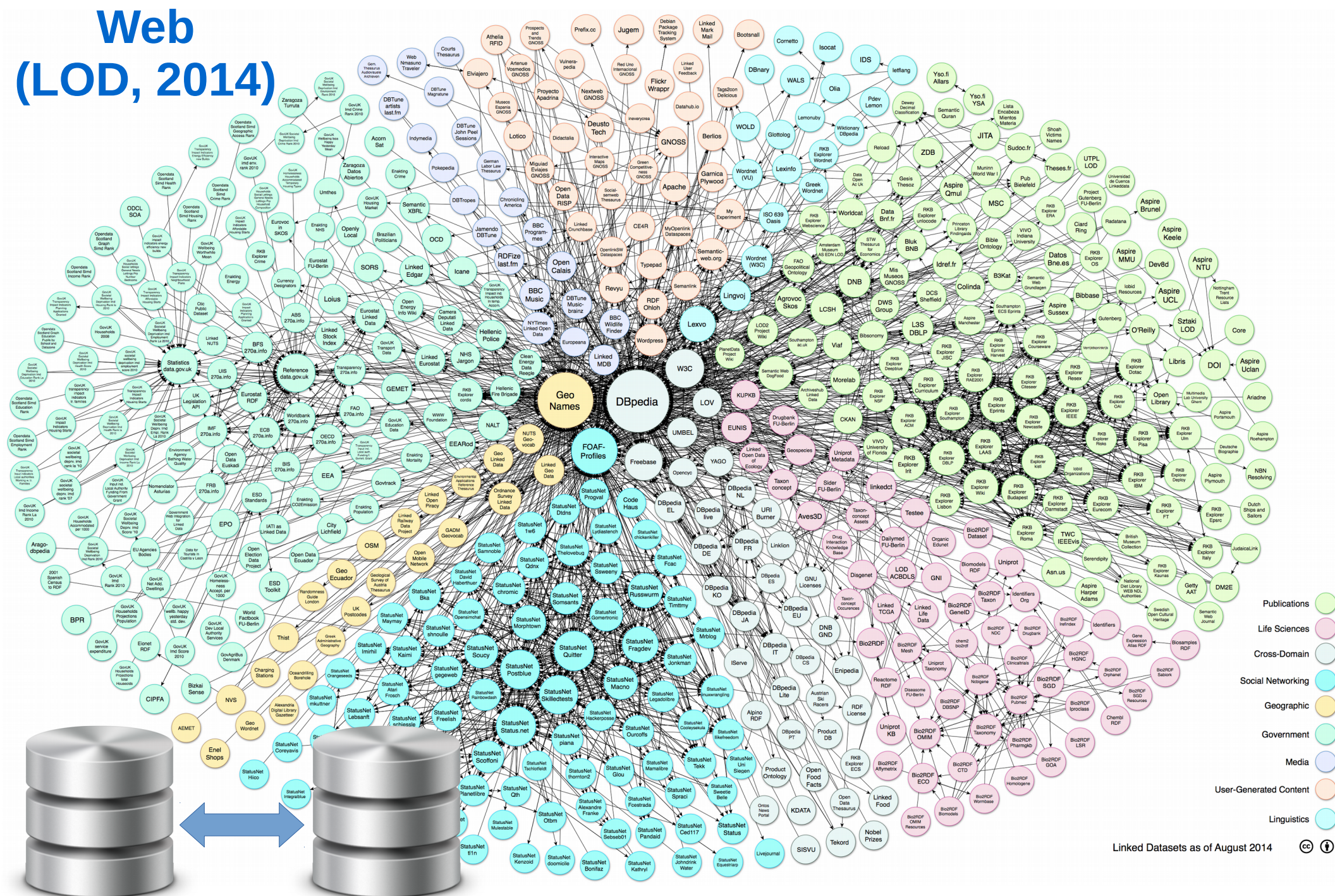
Power grid (detail, US)



Internet (portion, 2005)

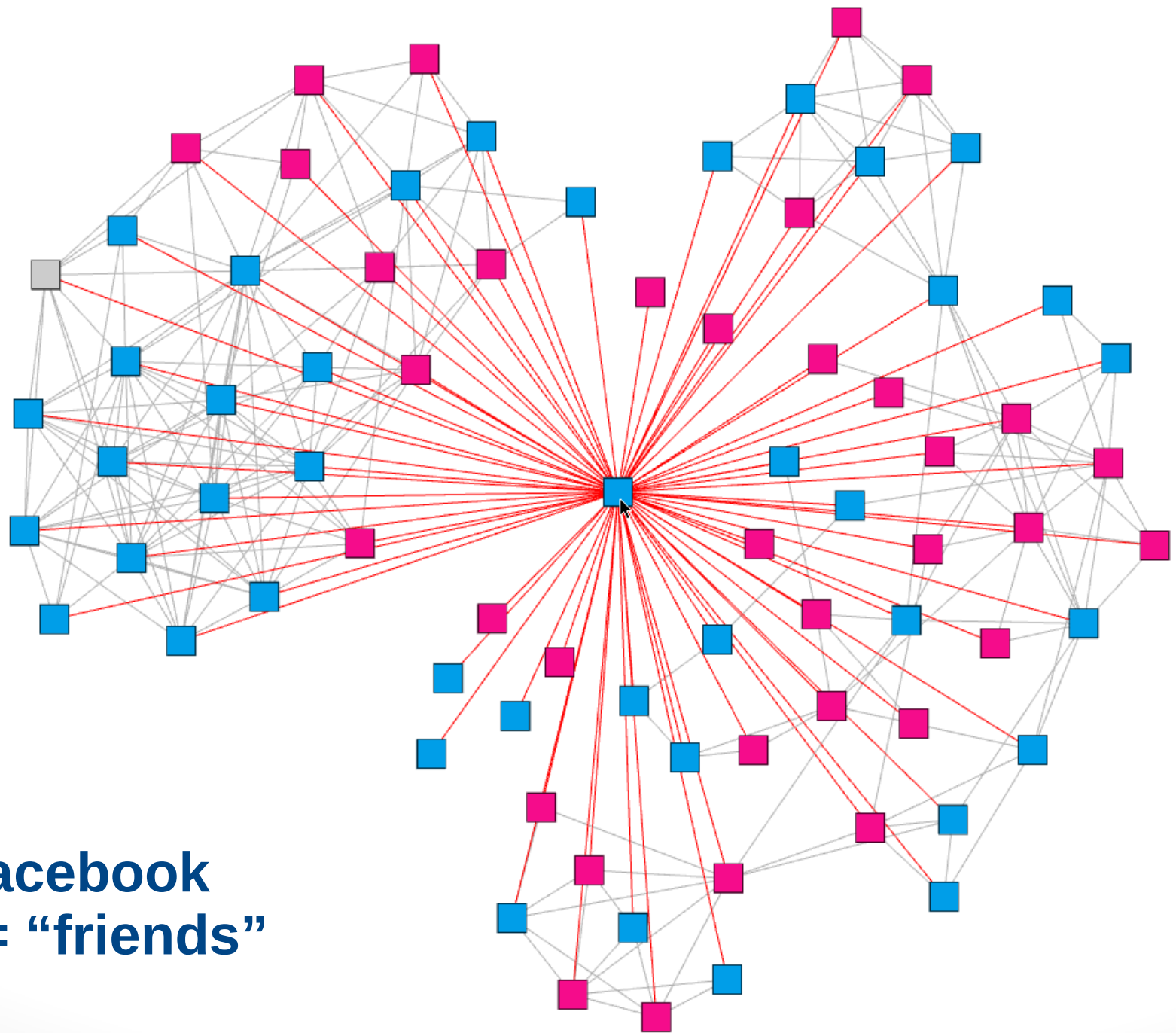


Web (LOD, 2014)

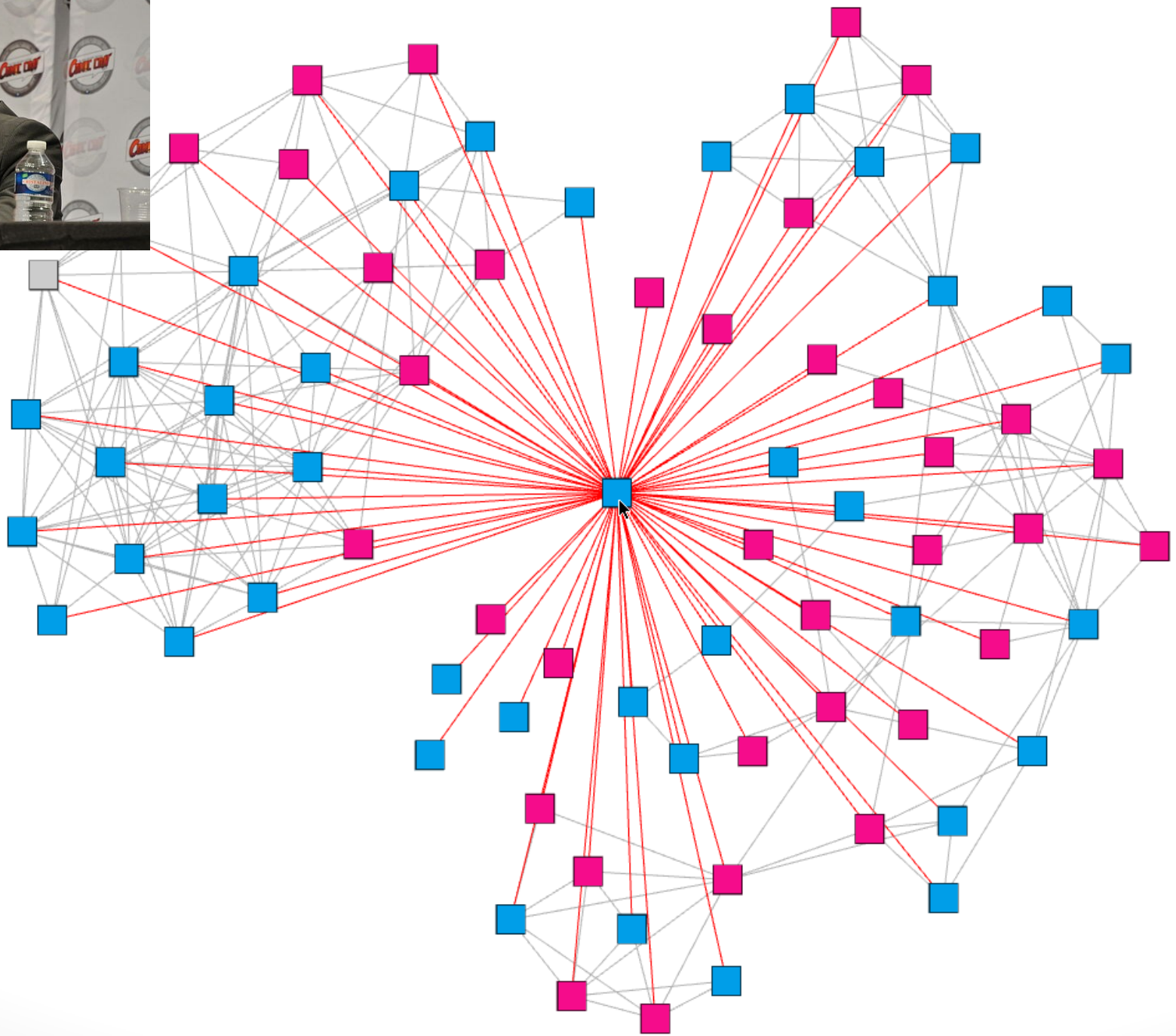


UTC, Cambridge, 10 January 2017

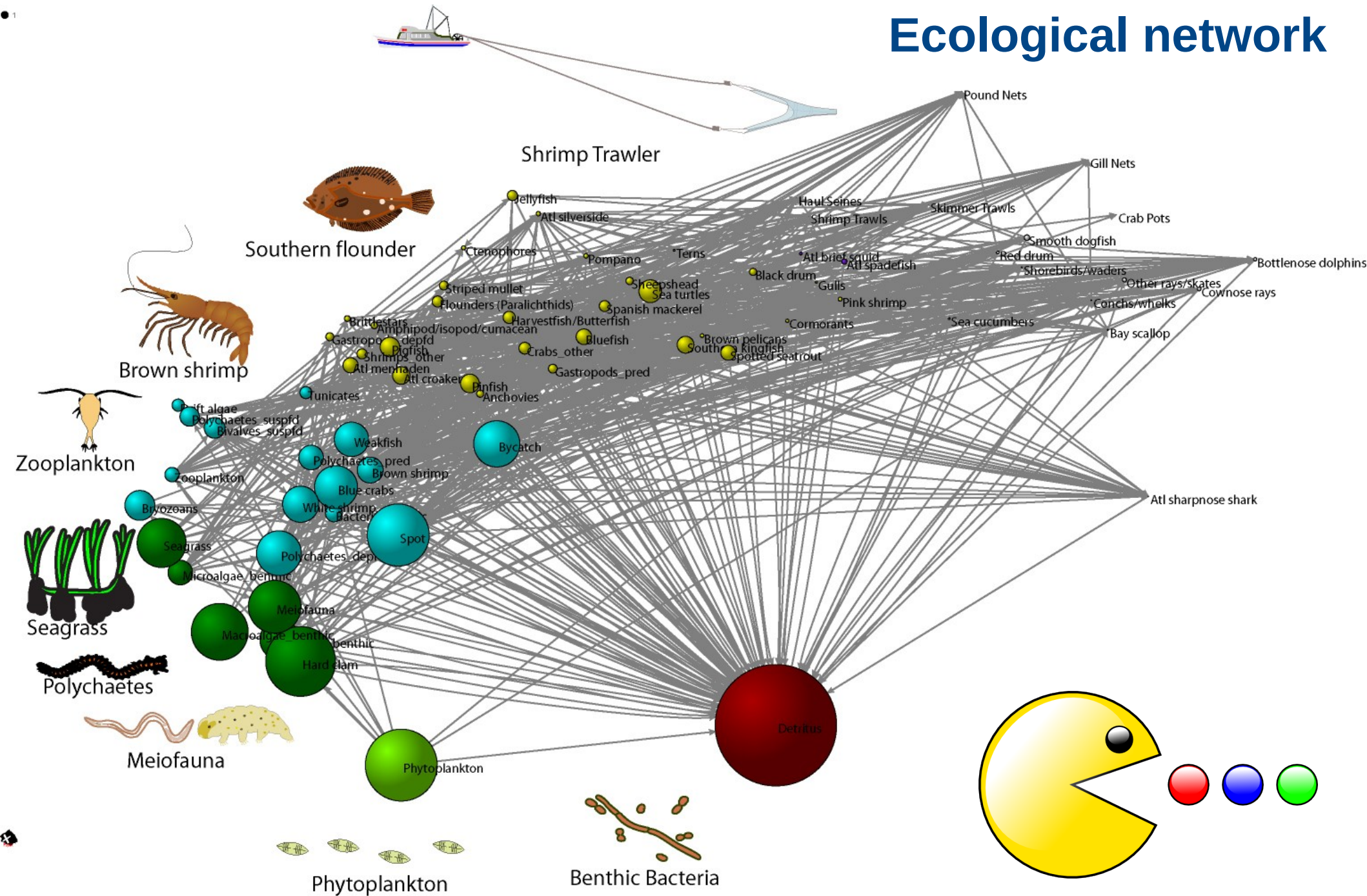


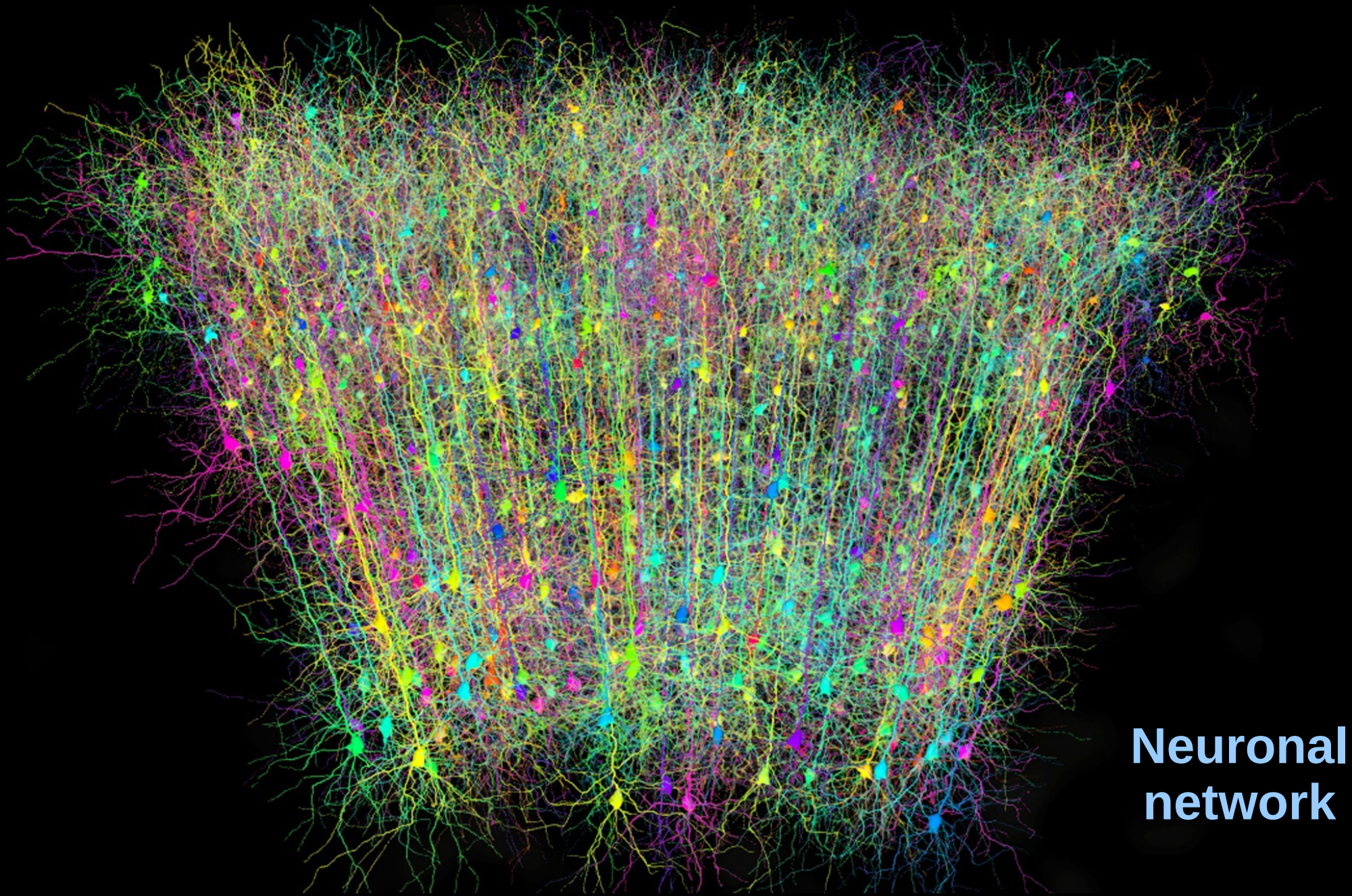


**My Facebook
nodes = “friends”**



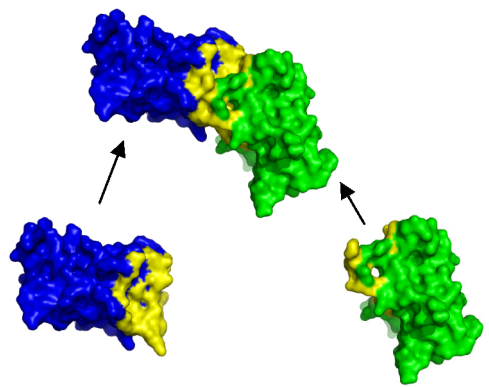
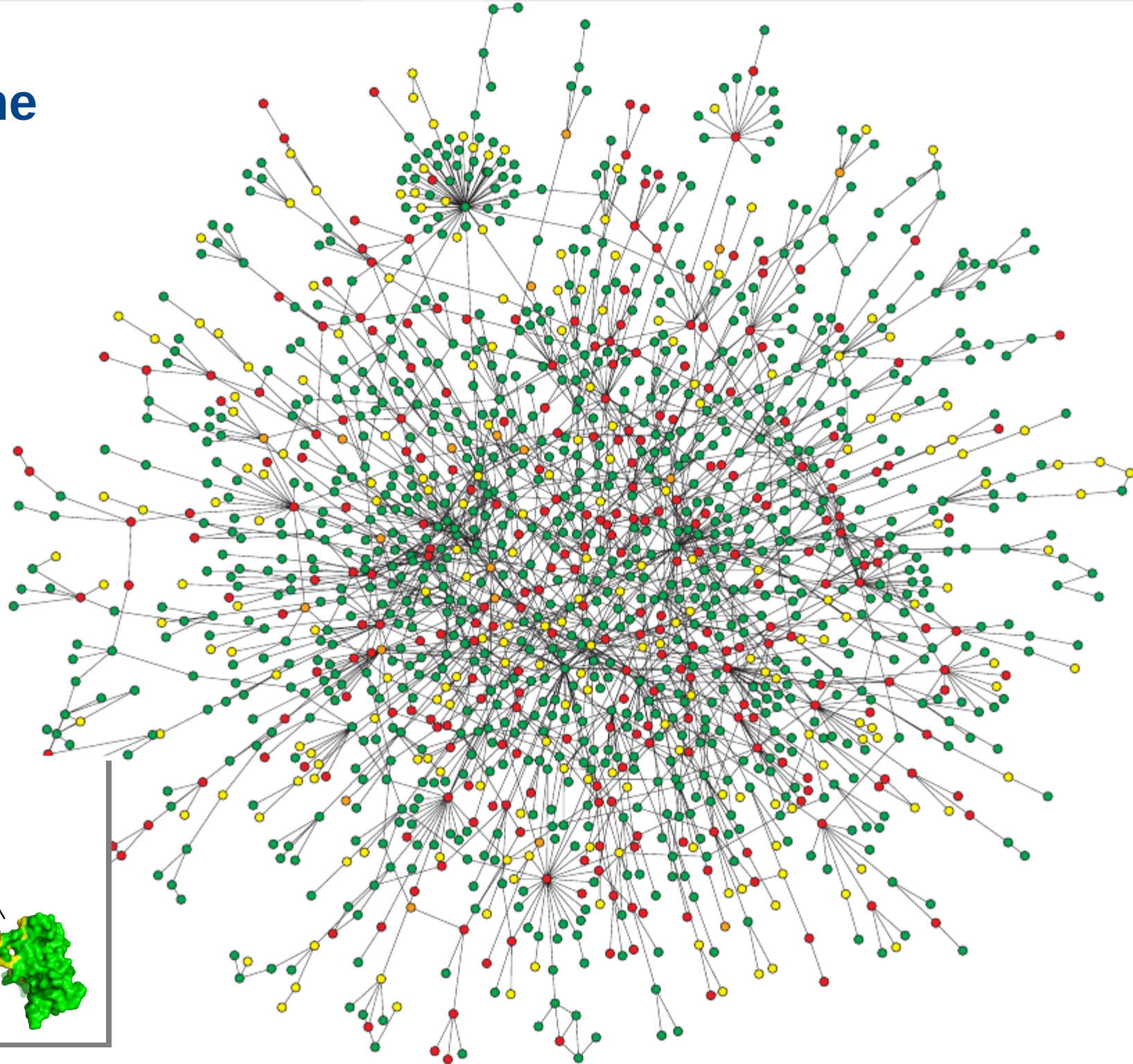
Ecological network

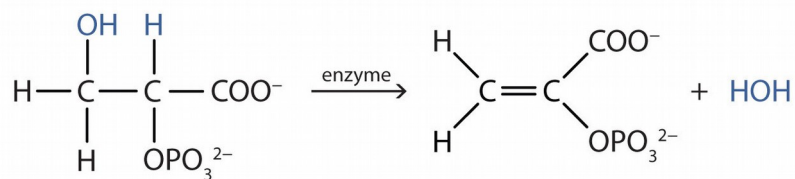




Neuronal network

Protein interactome (yeast)

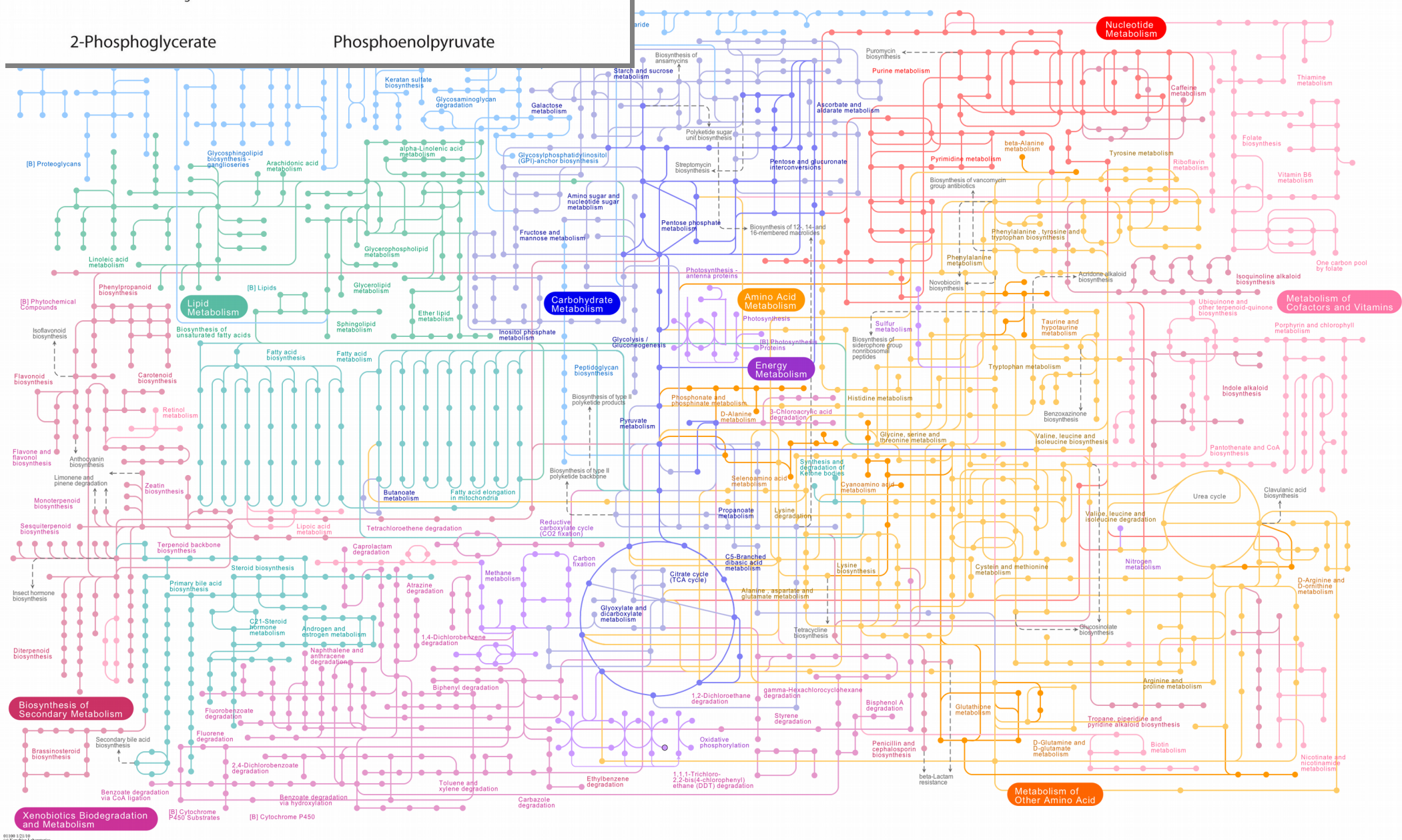




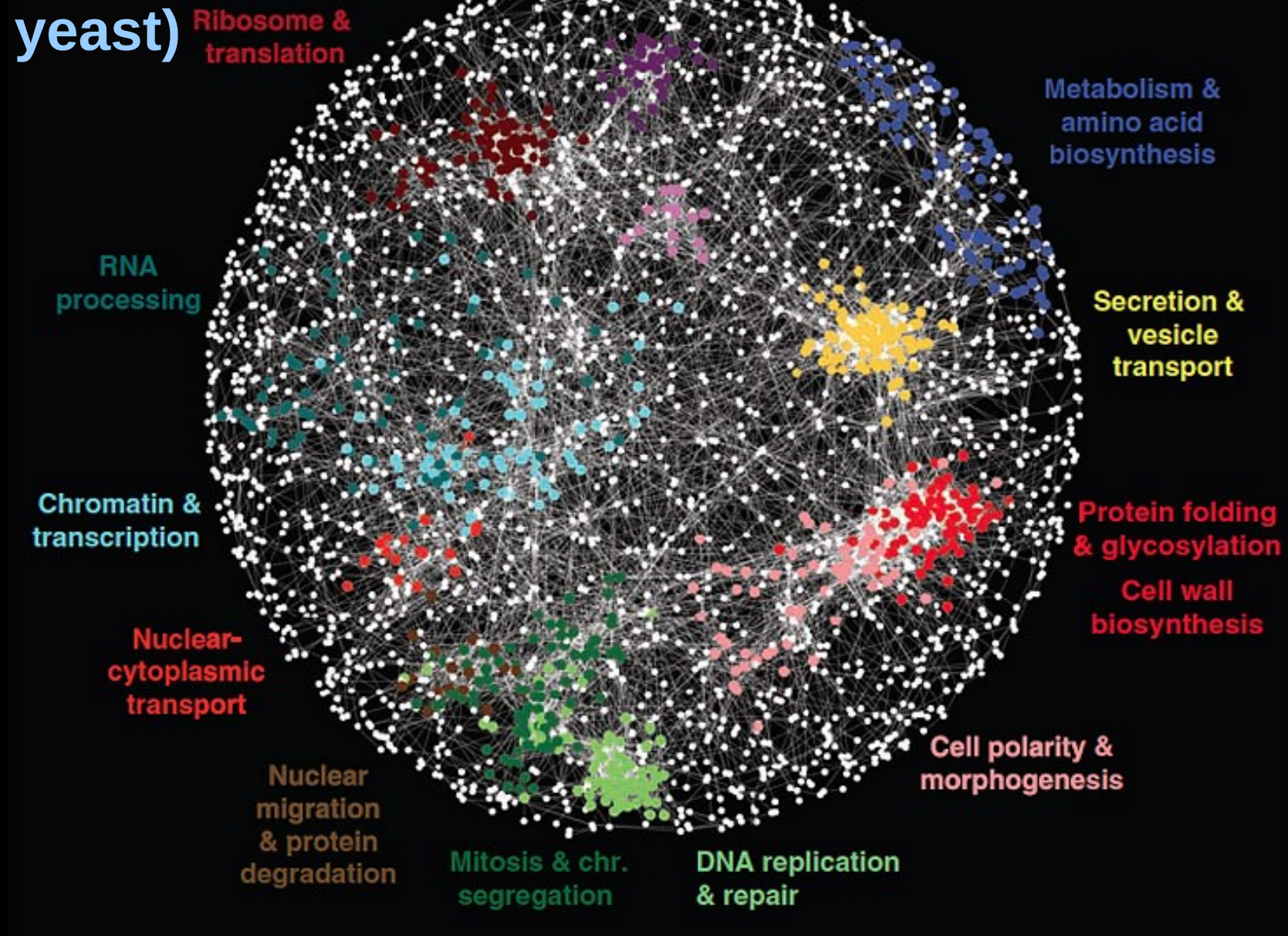
2-Phosphoglycerate

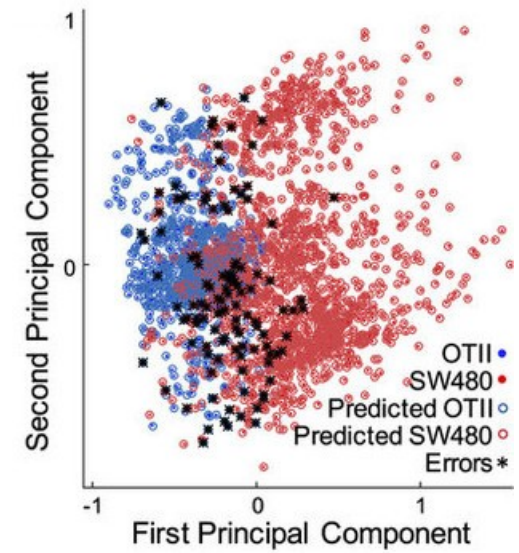
Phosphoenolpyruvate

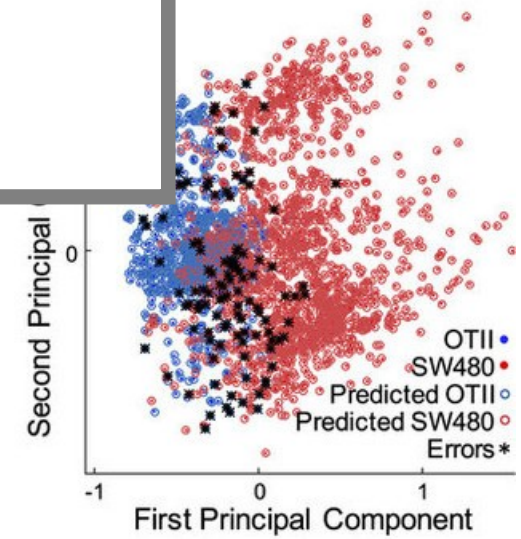
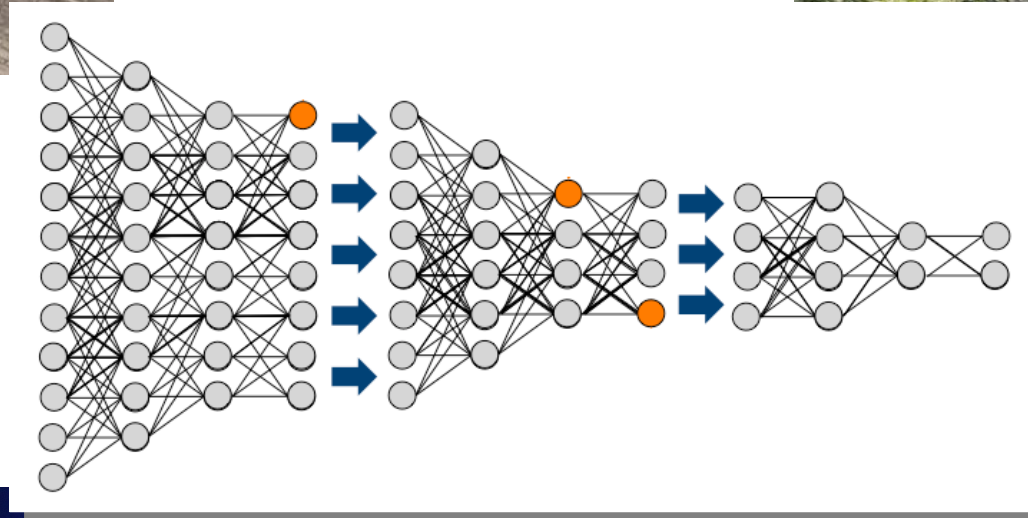
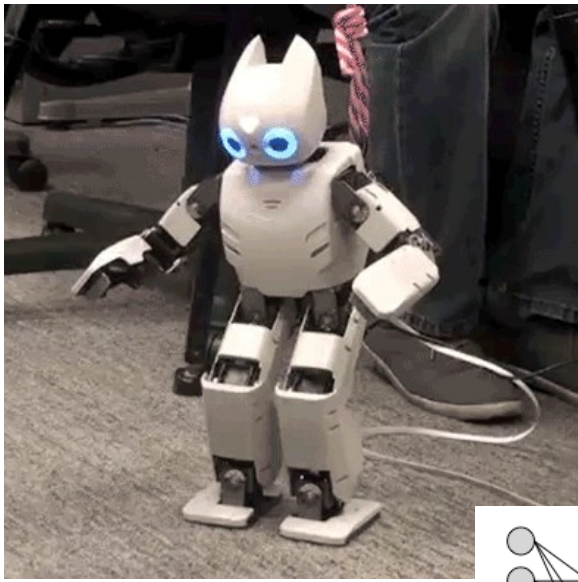
Metabolic network



Genetic network (synthetic lethality yeast)







Reconstructing networks

Pre-existing knowledge



Search: Q01097

Search

Clear

[Show Advanced Fields »](#) [MIQL syntax reference](#)

- Free text search will look by default for interactor identifier, species, interaction id, detection method, interaction type, publication identifier or author
- For a more specific search, use MIQL syntax or advanced search
- Search based on exact word matches eg. BRCA2 will not match BRCA2B
- Search for isoforms of 'P12345' by using 'P12345'

Home

Search

Interactions (254)

Browse

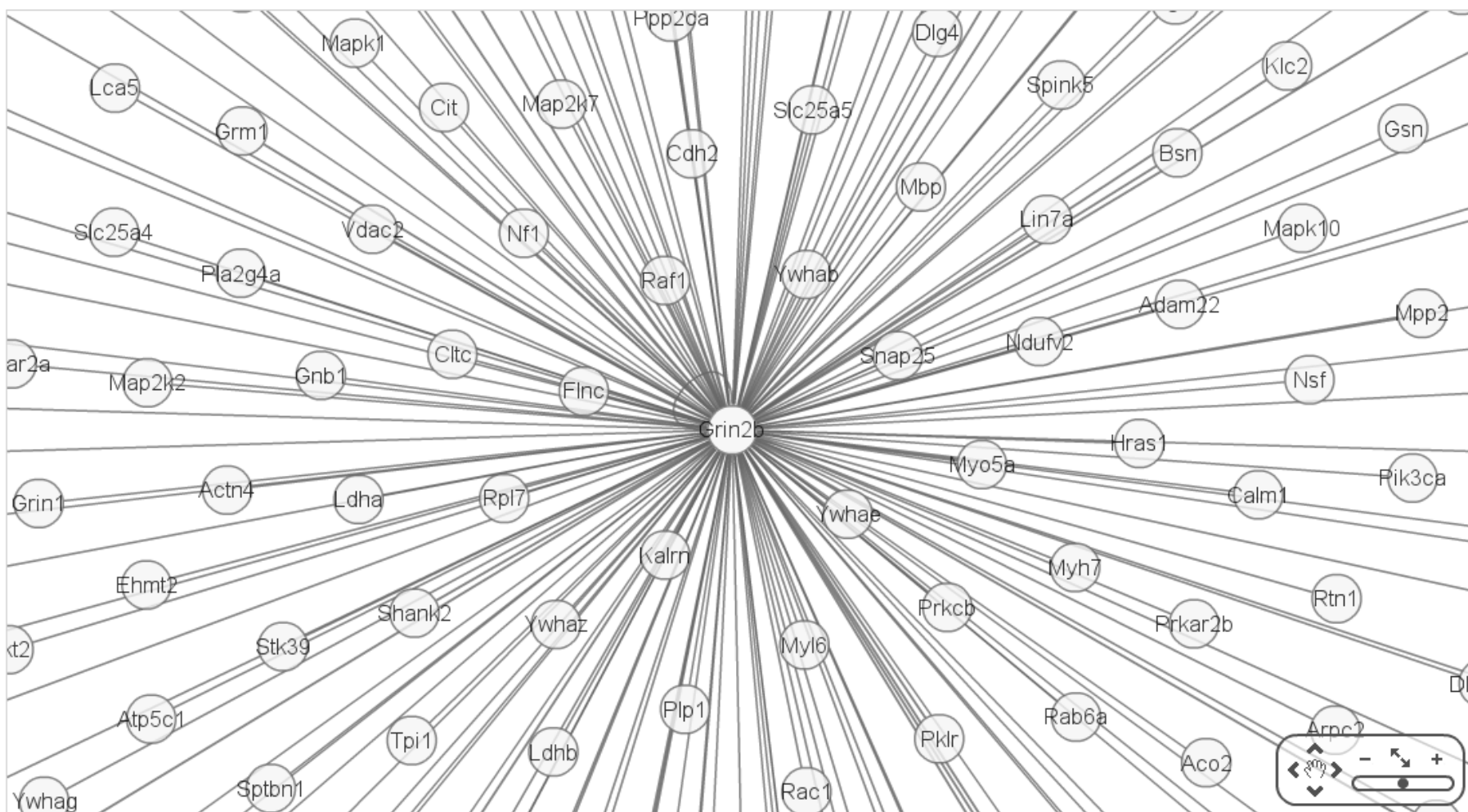
Lists

Interaction Details

Molecule View

Graph

Network visualisation



- Home
- Advanced Search
- Tools
- Data Submission
- Downloads
- Documentation
- Acknowledgements
- Contact Us

FOLLOW US ON
twitter

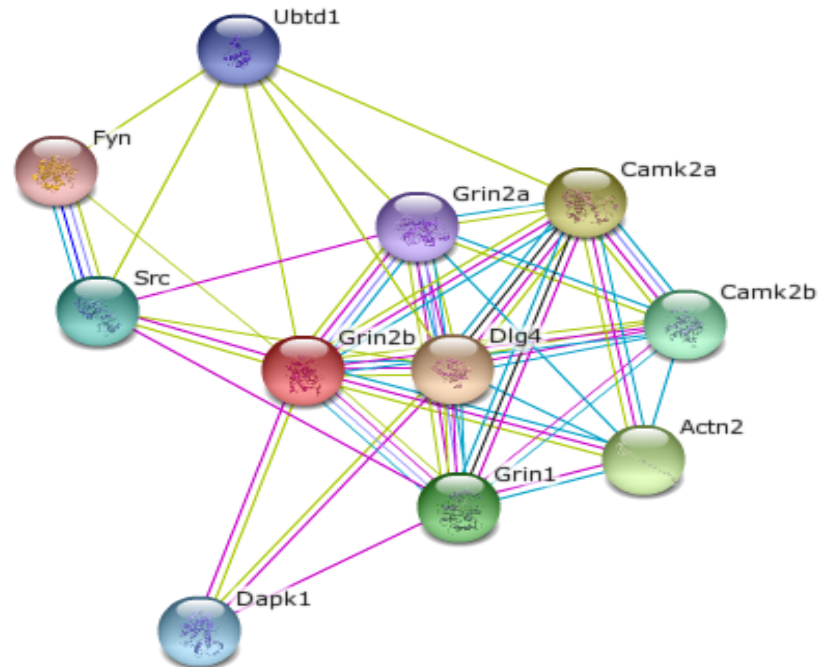
[IntAct release](#)
about 8 days ago
158/159! 302465 binary
interactions from 5787
manually curated
publications
<http://ftp.ebi.ac.uk/pub/databases/intact/current>

[IntAct is now available for download in PSI MITAB2.7 in addition to PSI-MI XML, MITAB2.5 and MITAB2.6.](#)
about 13 days ago

[Tarassov et al \(PMID:18467557\) low confidence interactors withdrawn in consultation with the authors](#)
about 13 days ago

[Dataset of the month \(November\). Blandin et al. A Human Skeletal Muscle Interactome Centered on Proteins Involved in Muscular Dystrophies.](#)
about 21 days ago

[Now IntAct can visualise dynamic](#)
about 34 days ago



This is the **evidence view**. Different line colors represent the types of evidence for the association.



(requires Flash player 10 or better)

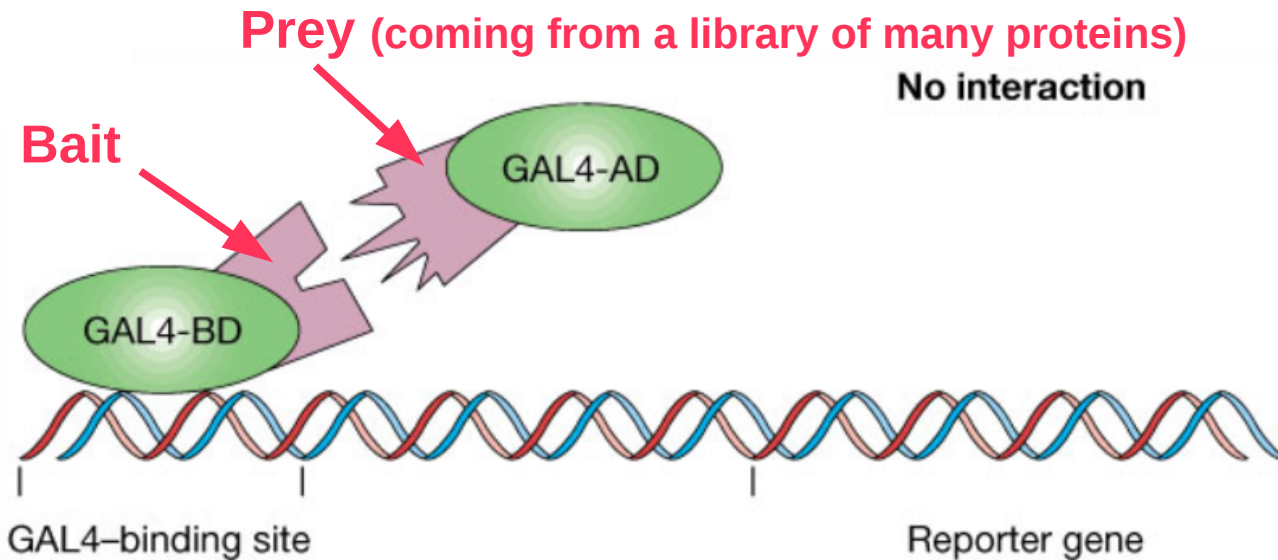
Your Input:

- Grin2b glutamate receptor, ionotropic, NMDA2B (epsilon 2) Gene; NMDA receptor subtype of glutamate-gated ion channels with high calcium permeability and voltage-dependent sensitivity to magnesium. Mediated by glycine (1482 aa) (*Mus musculus*)

Predicted Functional Partners:

Neighborhood
 Gene Fusion
 Cooccurrence
 Coexpression
 Experiments
 Databases
 Textmining
 [Homology]
Score

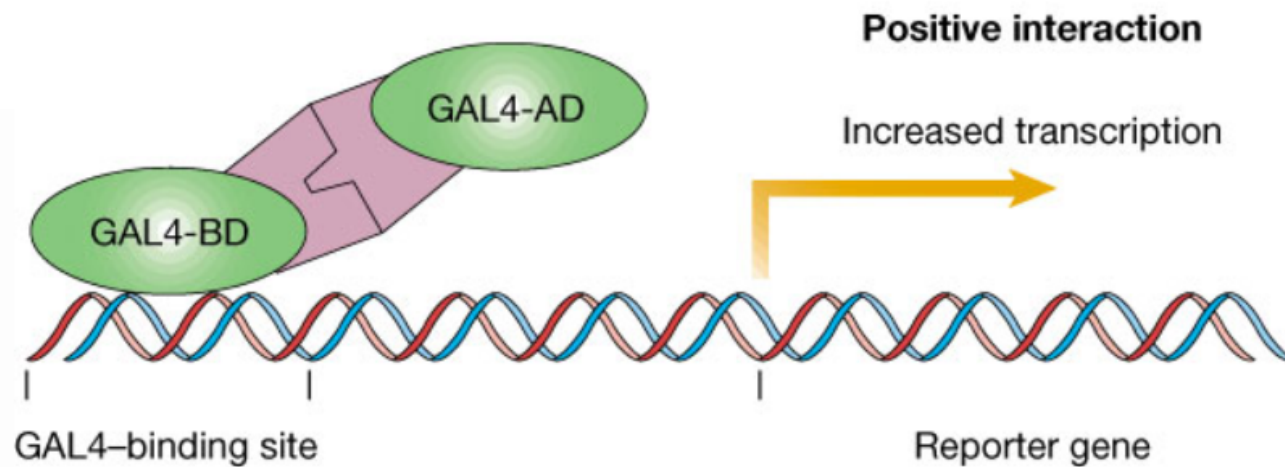
observations: yeast two hybrid system



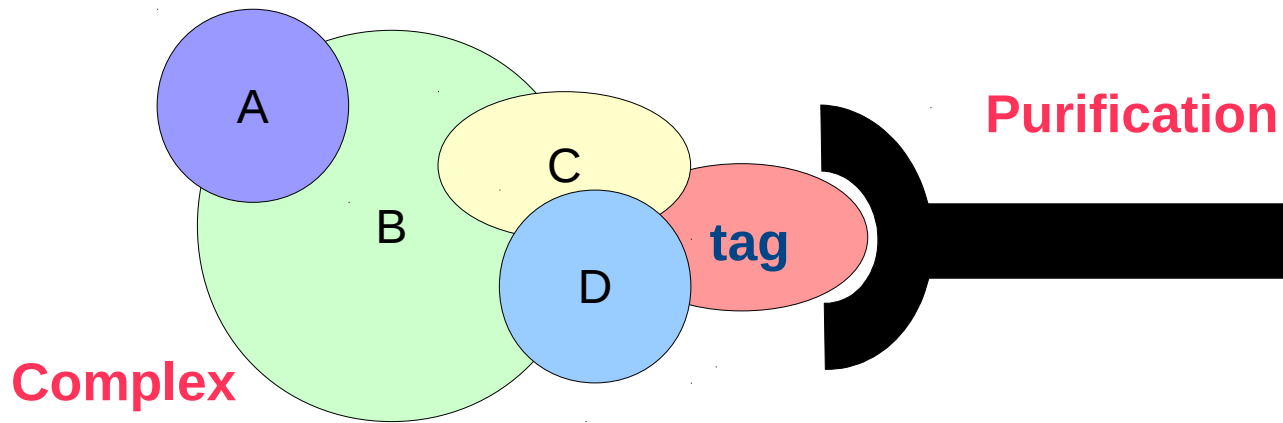
Direct physical interactions



All possible interactions (may never actually occur)

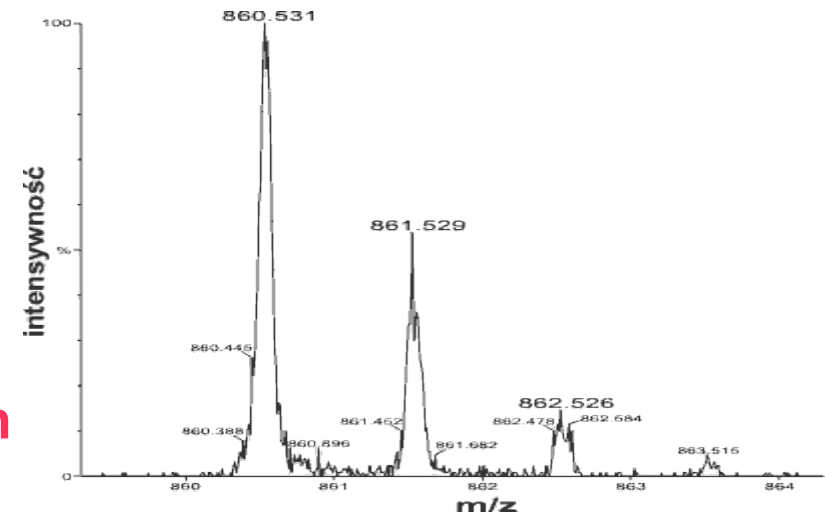
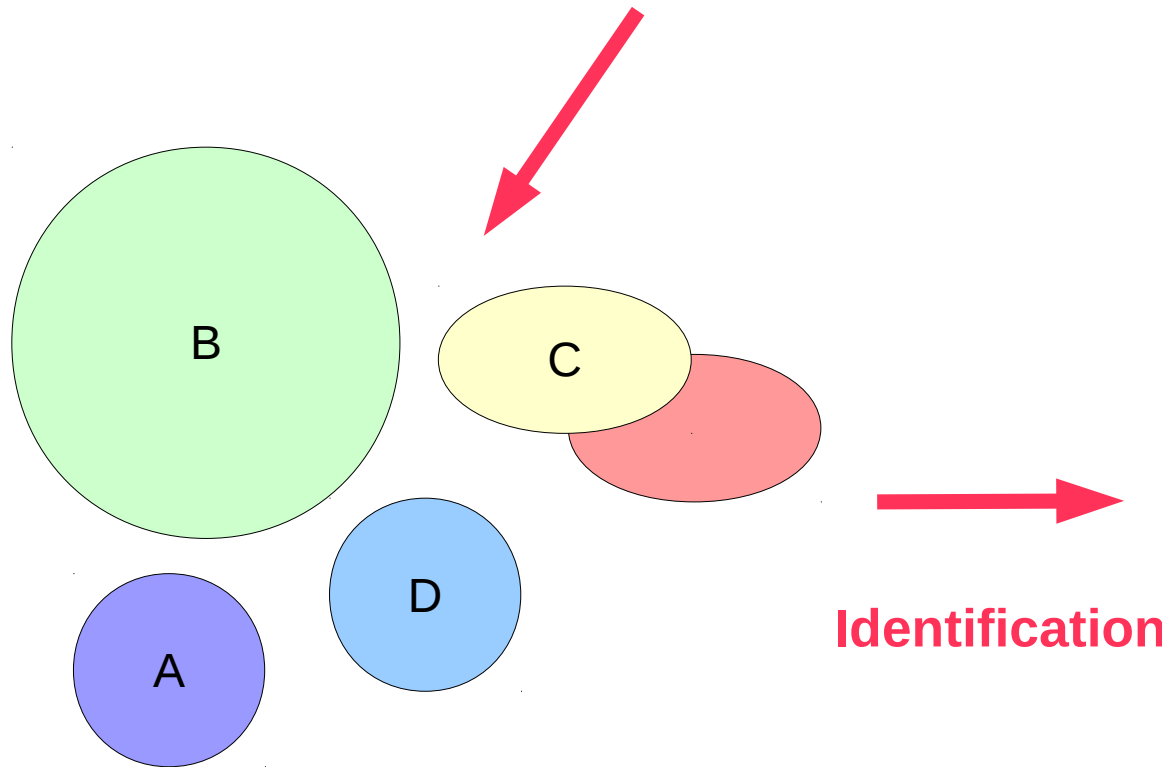


observations: protein complex precipitation



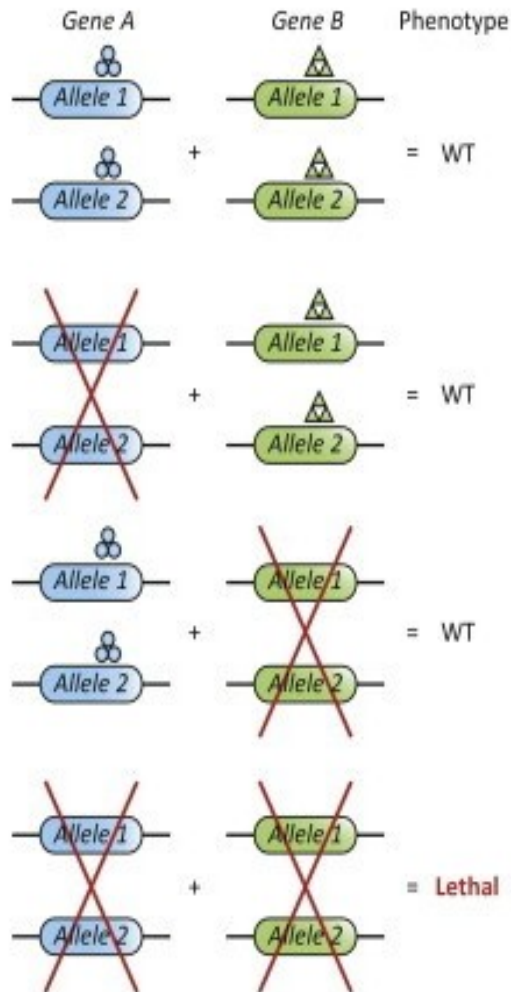
✓
actual complex

✗
not direct
interactions
(A-C and A-D
wrongly predicted)

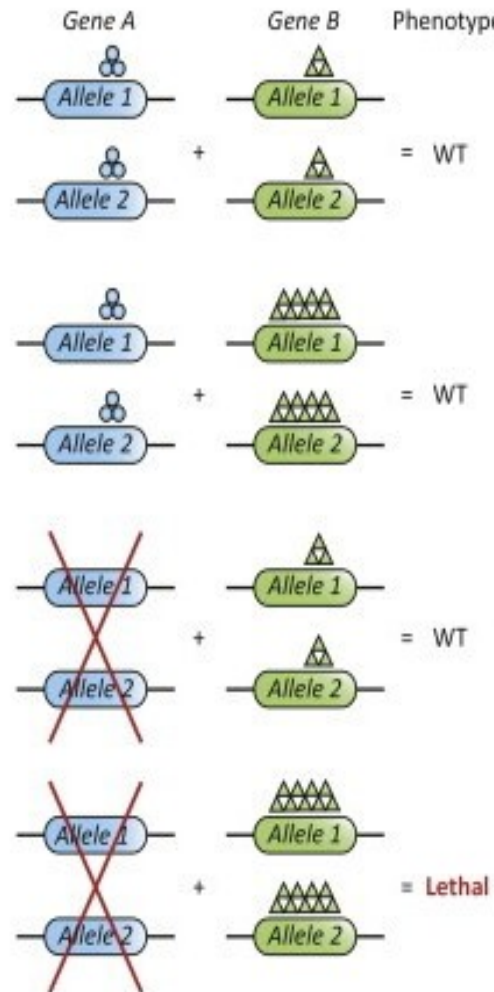


Synthetic lethality

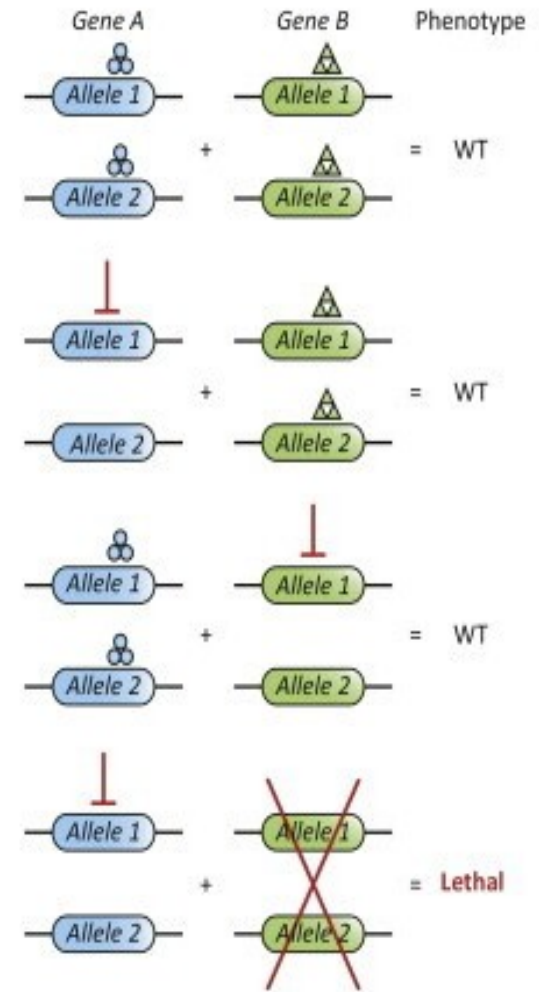
(i) Synthetic lethality



(ii) Dosage lethality

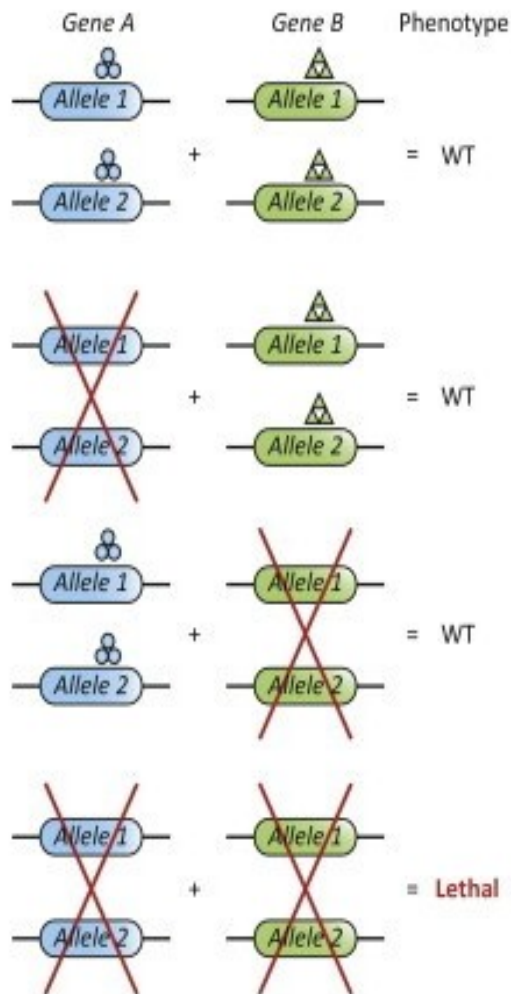


(iii) Chemical genetics

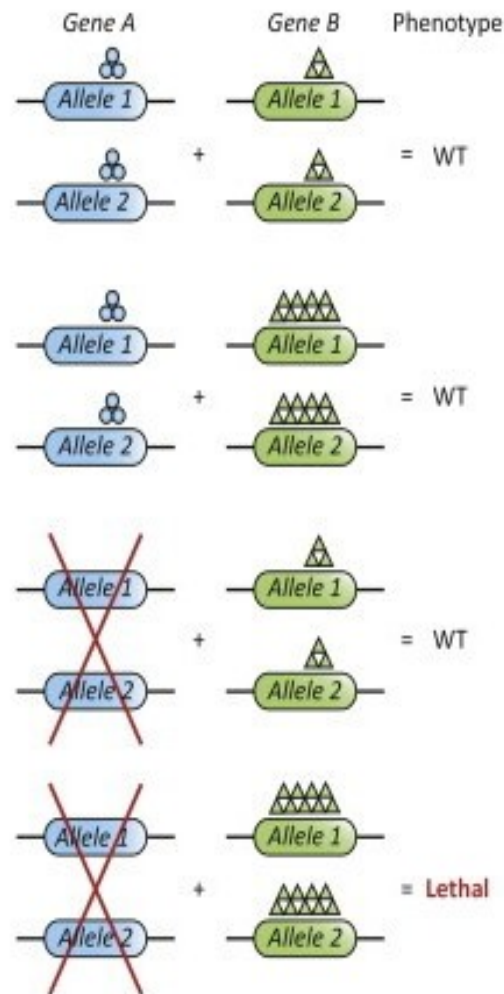


Synthetic lethality

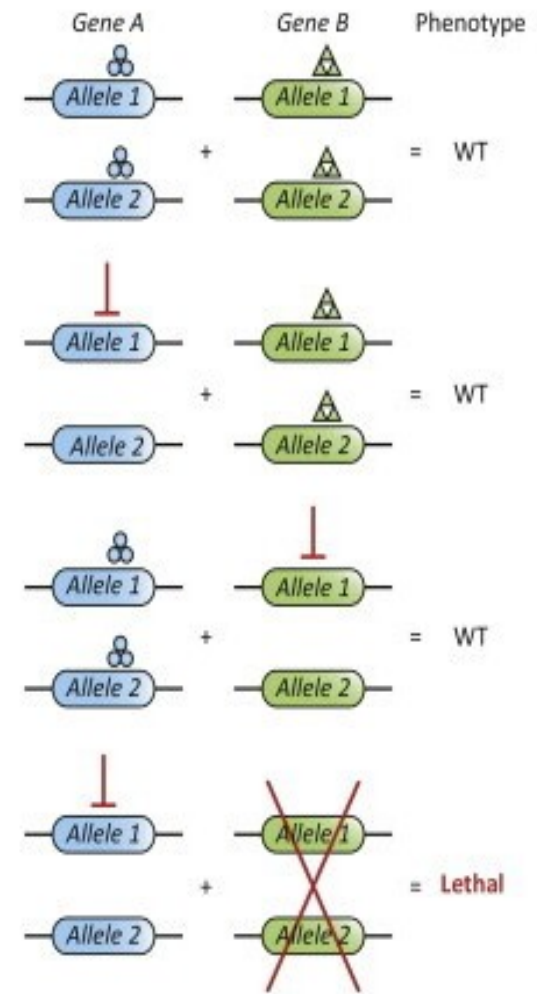
(i) Synthetic lethality



(ii) Dosage lethality

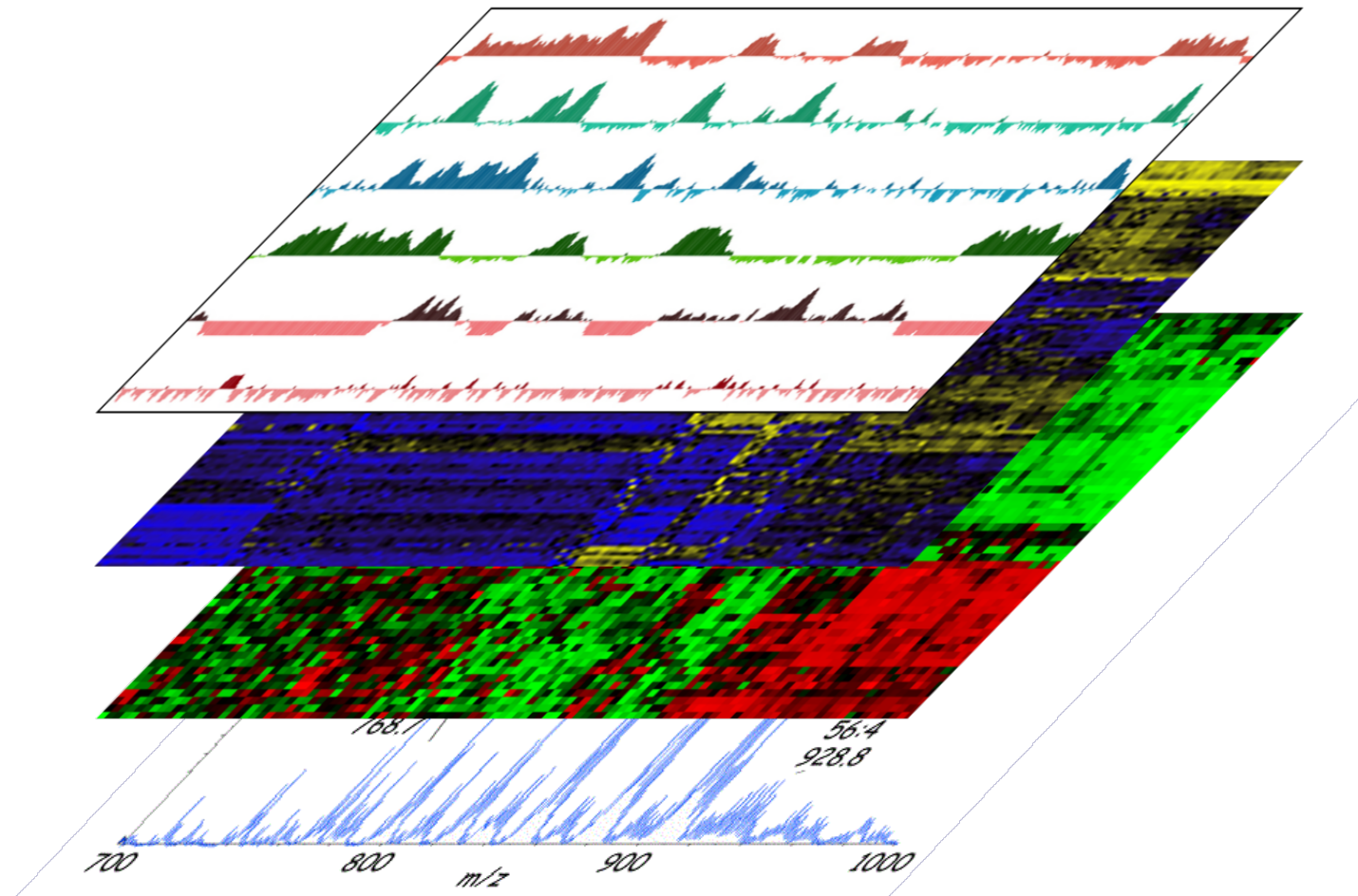


(iii) Chemical genetics

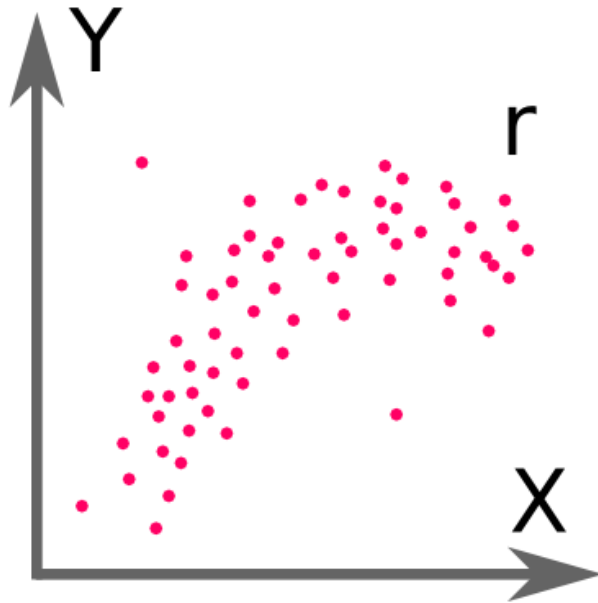


Can you think of a practical use of synthetic lethality in medicine?

network inference from data

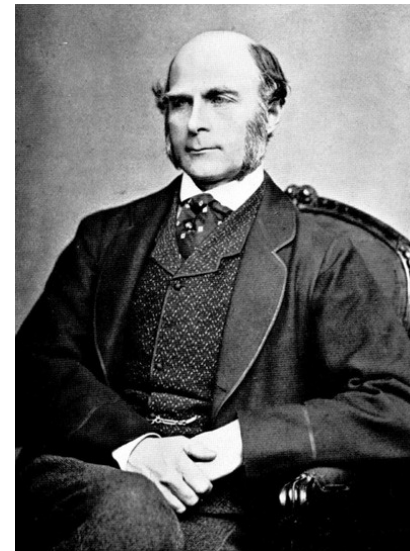


Network inference: Correlation

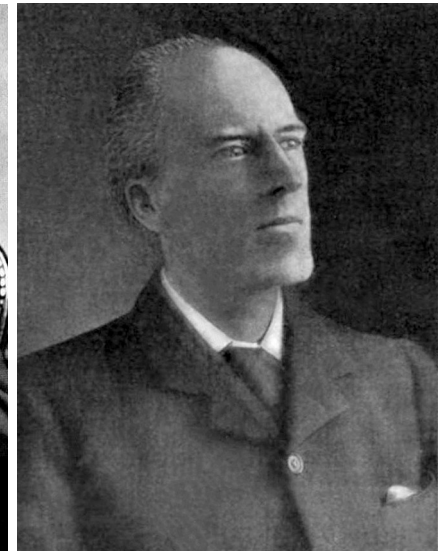


Values of X are somewhat related to values of Y

$$r = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum_{i=1}^n (X_i - \bar{X})^2} \sqrt{\sum_{i=1}^n (Y_i - \bar{Y})^2}}$$

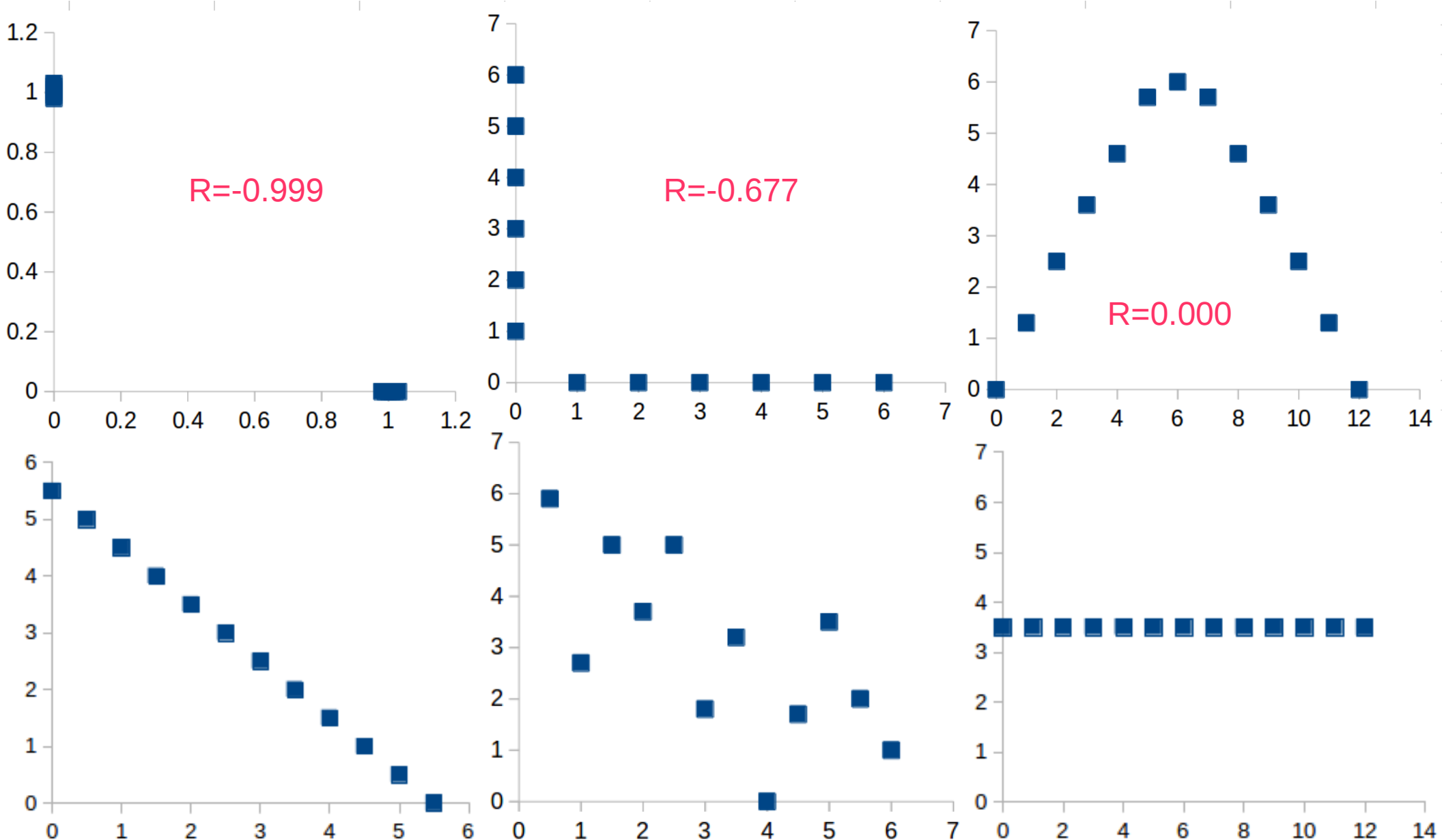


Francis Galton
(studied at Trinity
College, Camb)



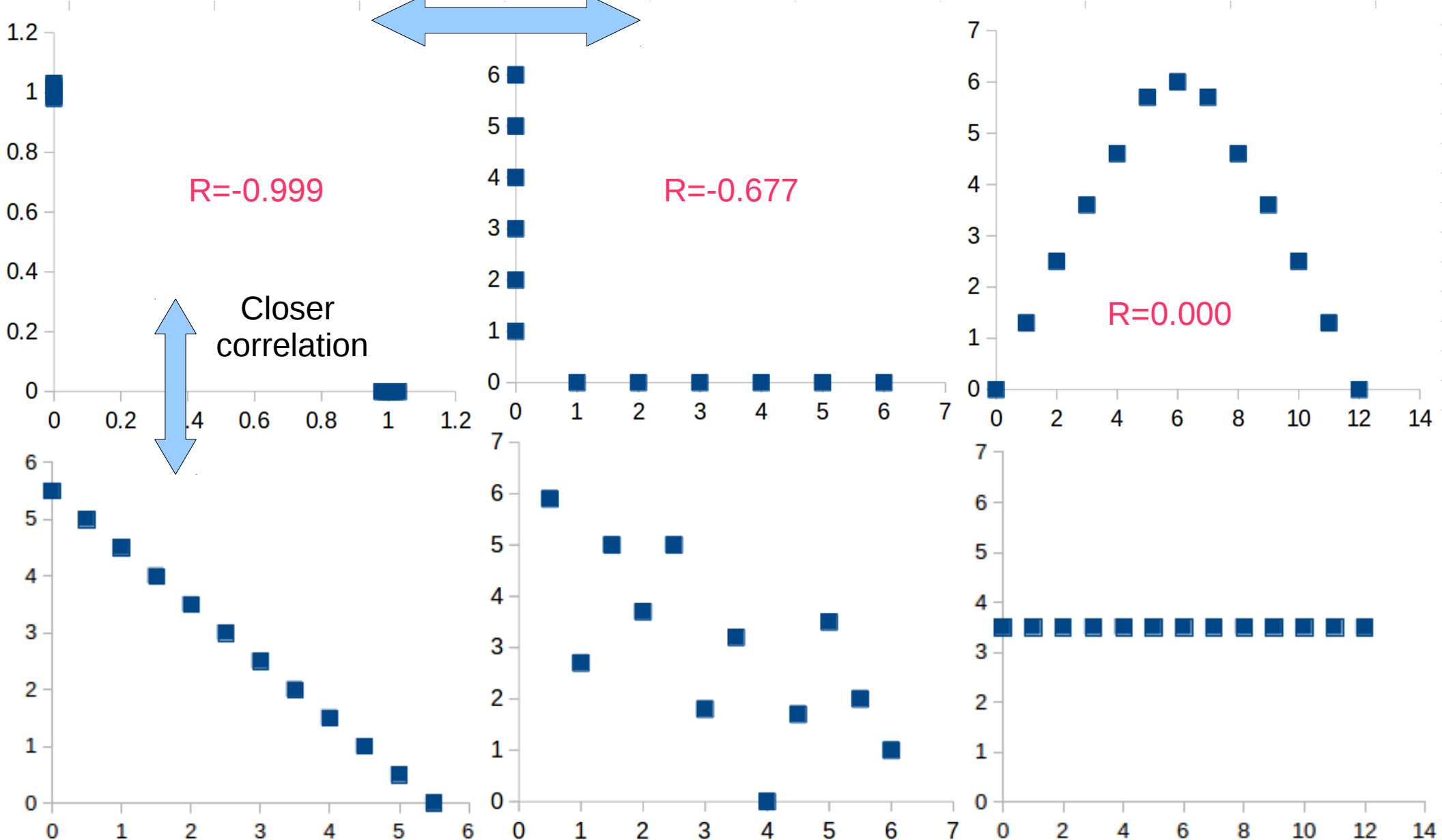
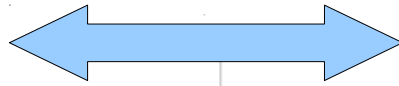
Karl Pearson
(studied at King's
College, Camb)

Correlations coefficients might be misleading

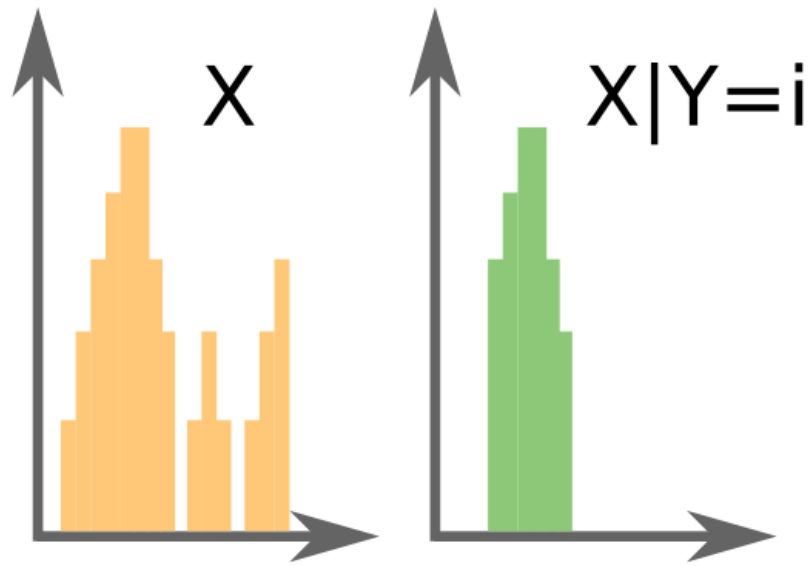


Correlations coefficients might be misleading

Closer biological meaning

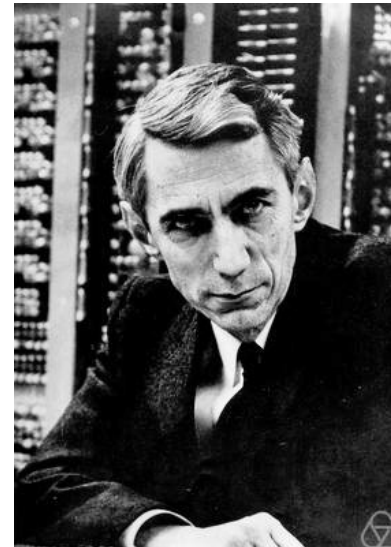


Network inference: Information Theory



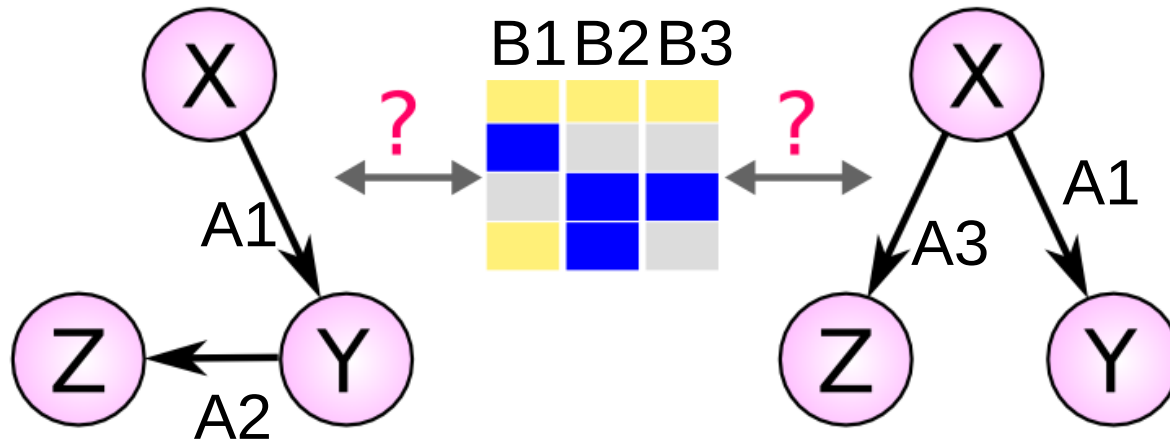
$$I(X, Y) = H(X) - H(X|Y)$$

E.g. mutual information:
knowledge of the value of Y reduces
the uncertainty on the values of X



Claude
Shannon

Network inference: Bayesian inference

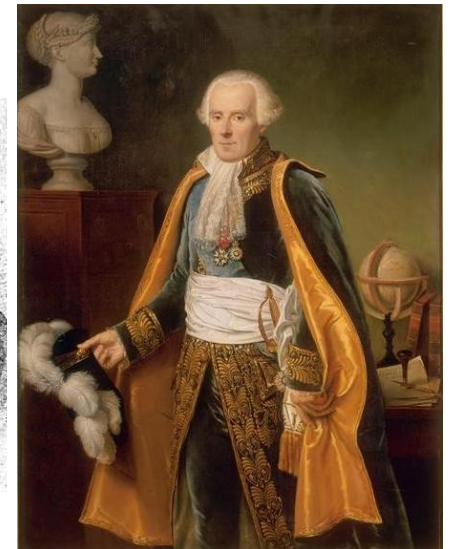


Which network most probably generated this dataset?

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

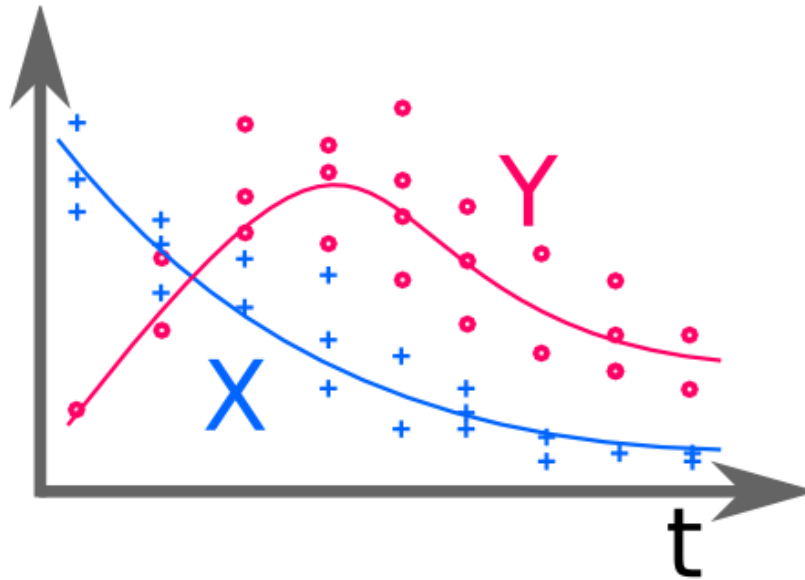


Thomas Bayes



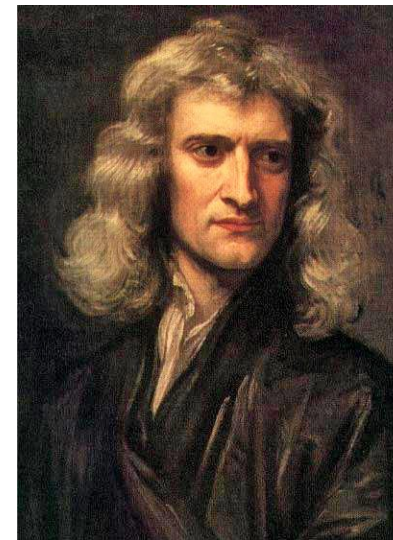
Pierre-Simon
De Laplace

Network inference: Ordinary differential equations



The rate of change of X depends on the value of Y

$$\frac{dx_i}{dt} = \sum_{j=1}^n a_{i,j} x_j$$



Isaac Newton
(Trinity College,
Camb)

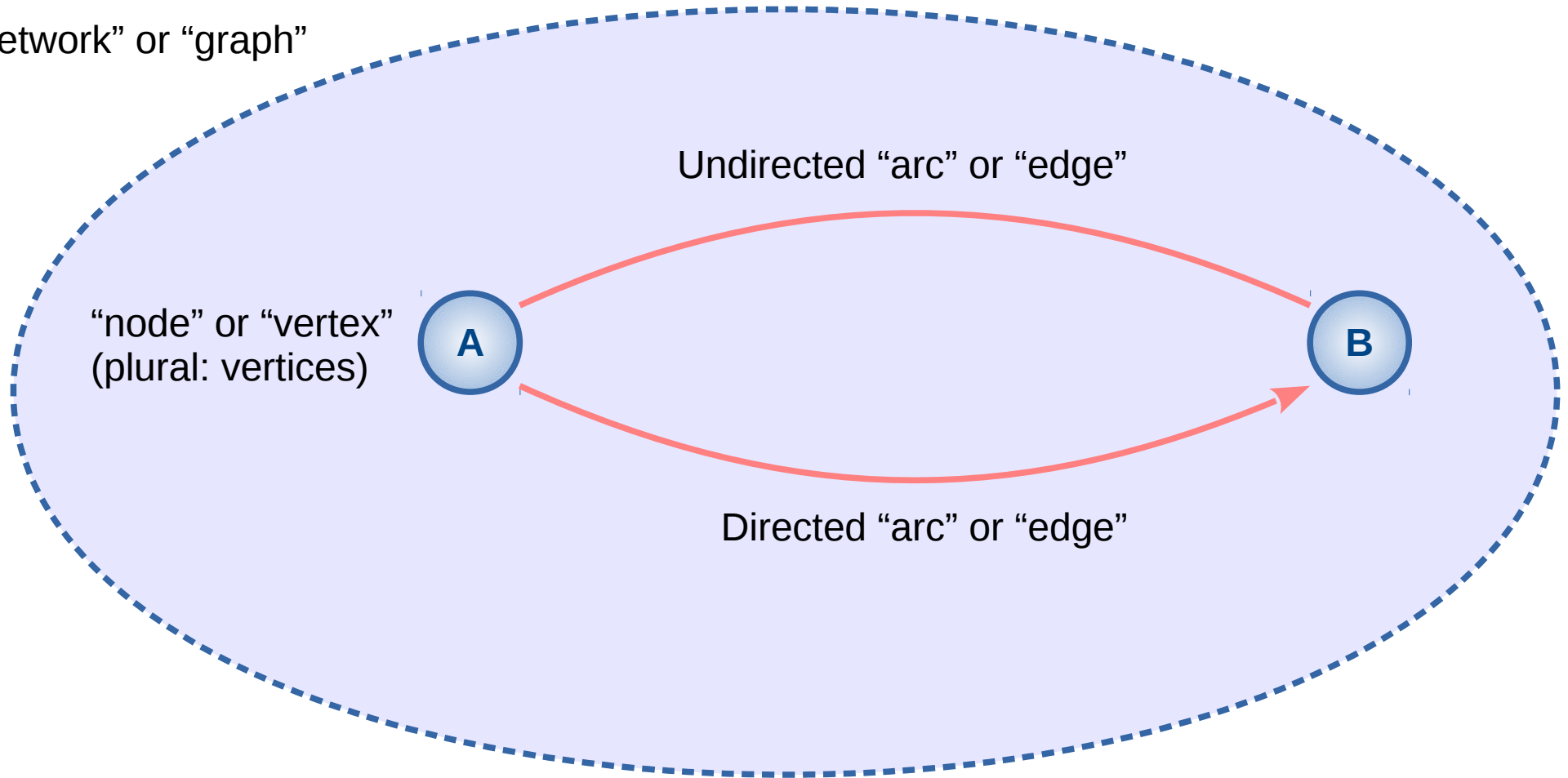


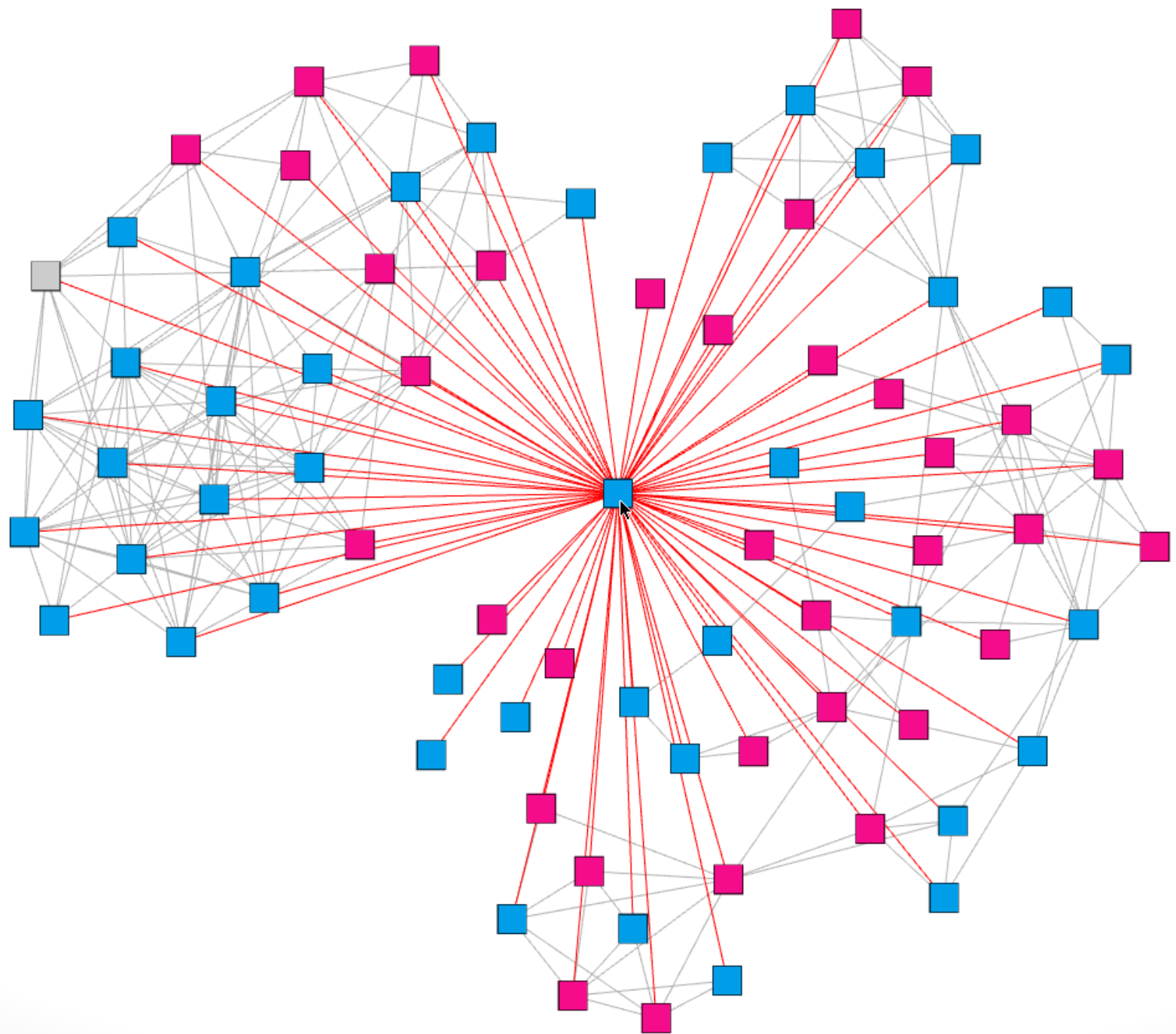
Gottfried Wilhelm
Leibniz

Analysing the structure of networks

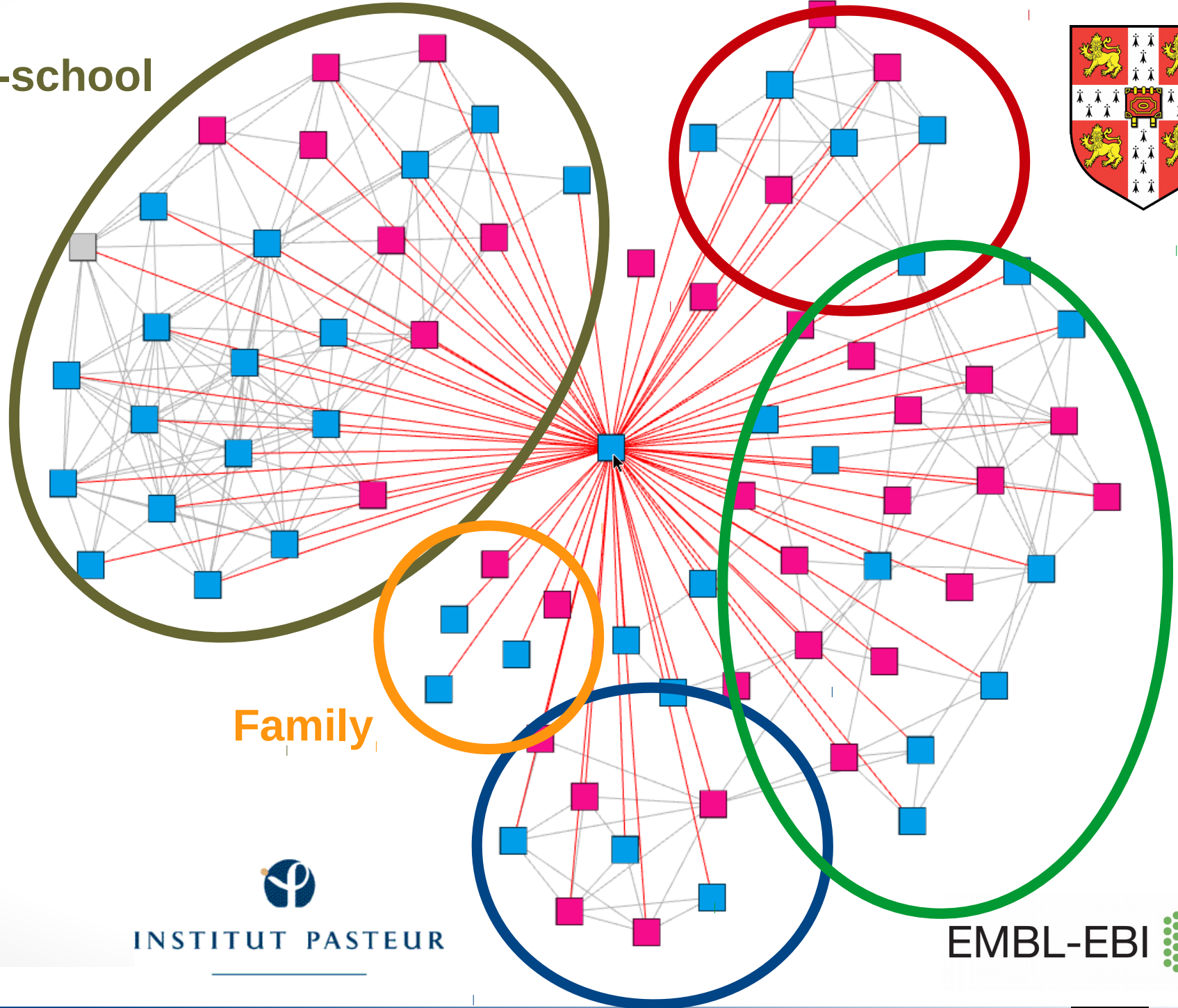
Definitions

“network” or “graph”





High-school

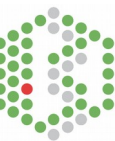


Family

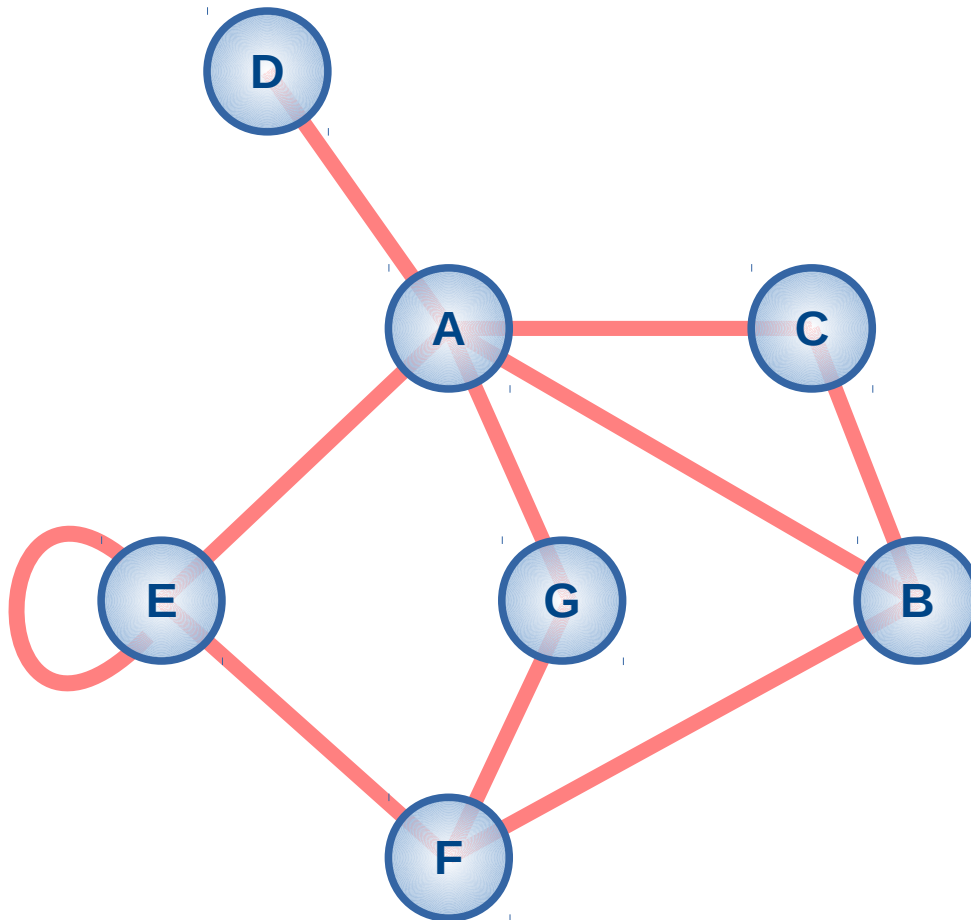


INSTITUT PASTEUR

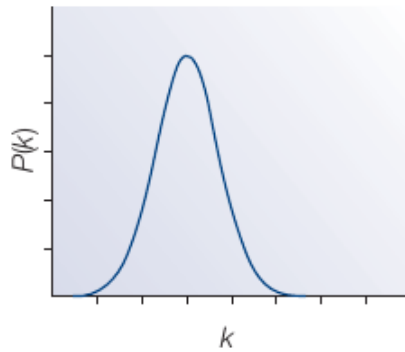
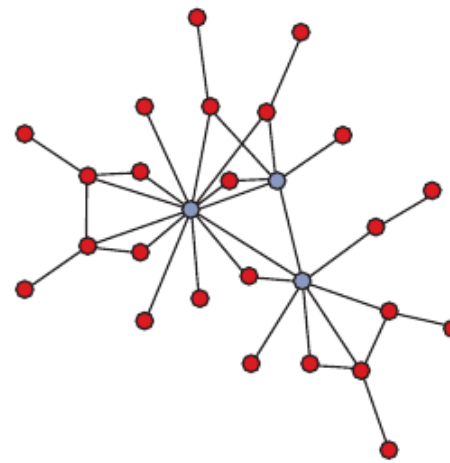
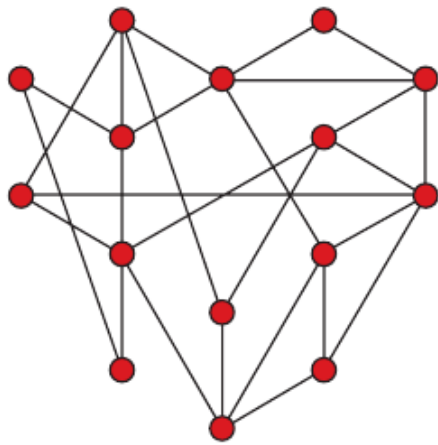
EMBL-EBI



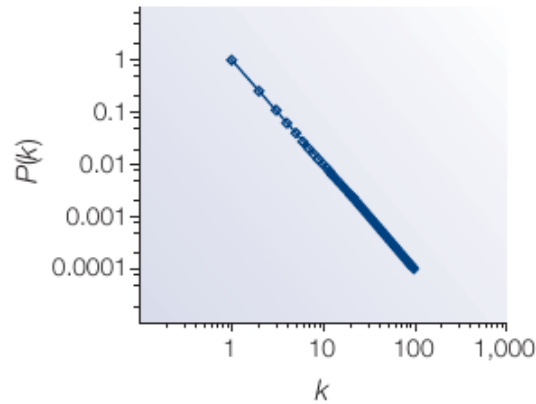
Node degree (k): number of ending edges



node	k
A	5
B	3
C	2
D	1
E	4
F	3
G	2

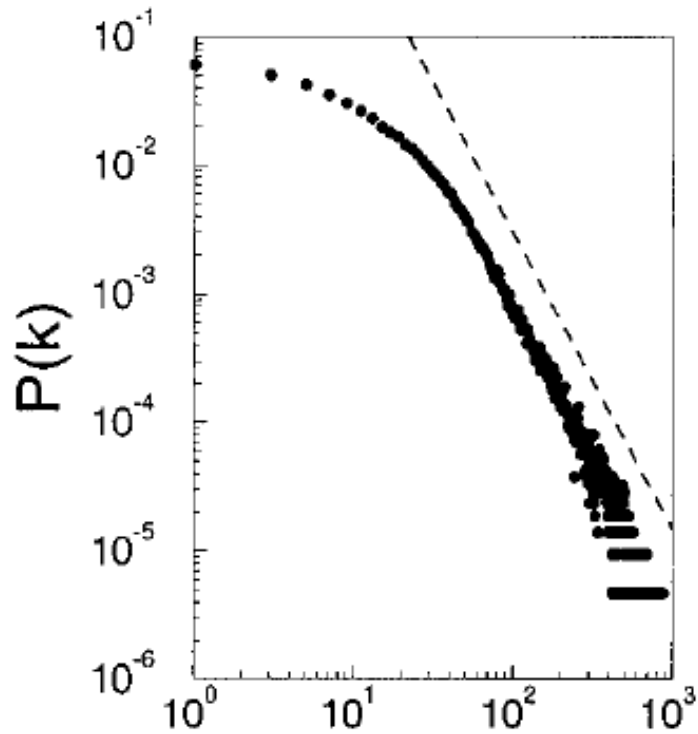


random
networks
(Erdős & Rényi)

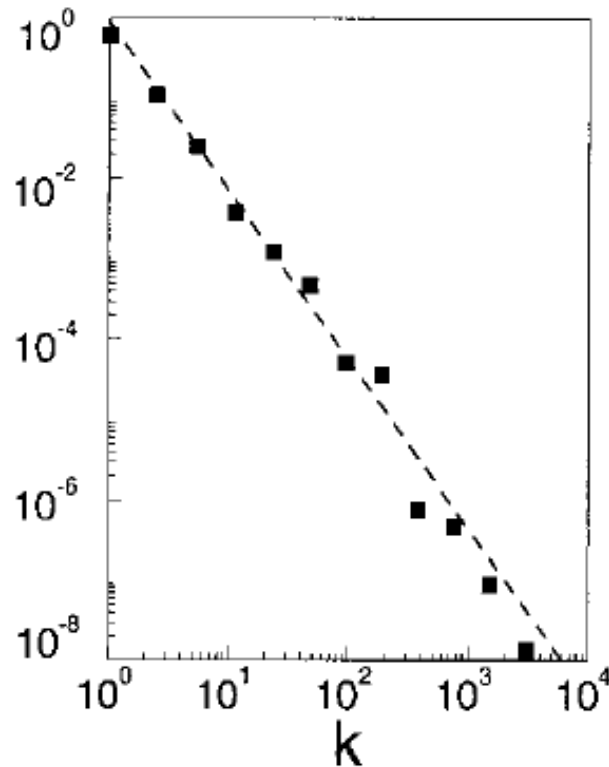


Power law
networks
(Barabási & Albert)

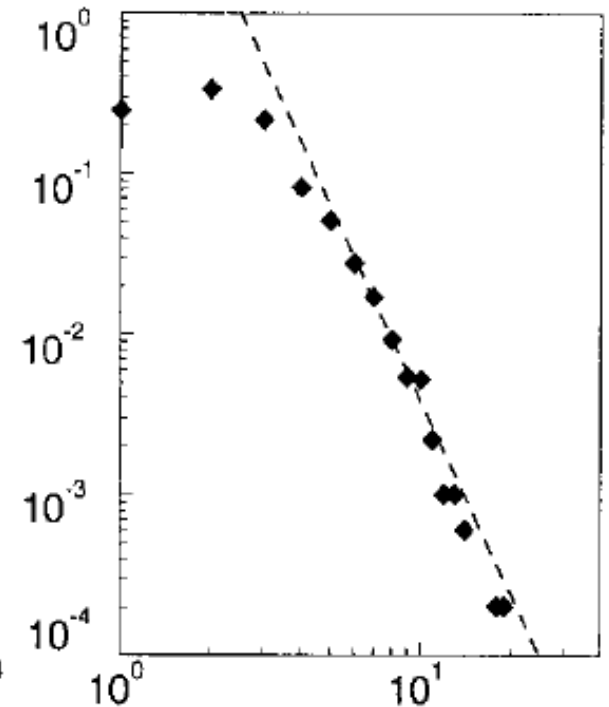
Examples of power-law networks



Actor
collaboration

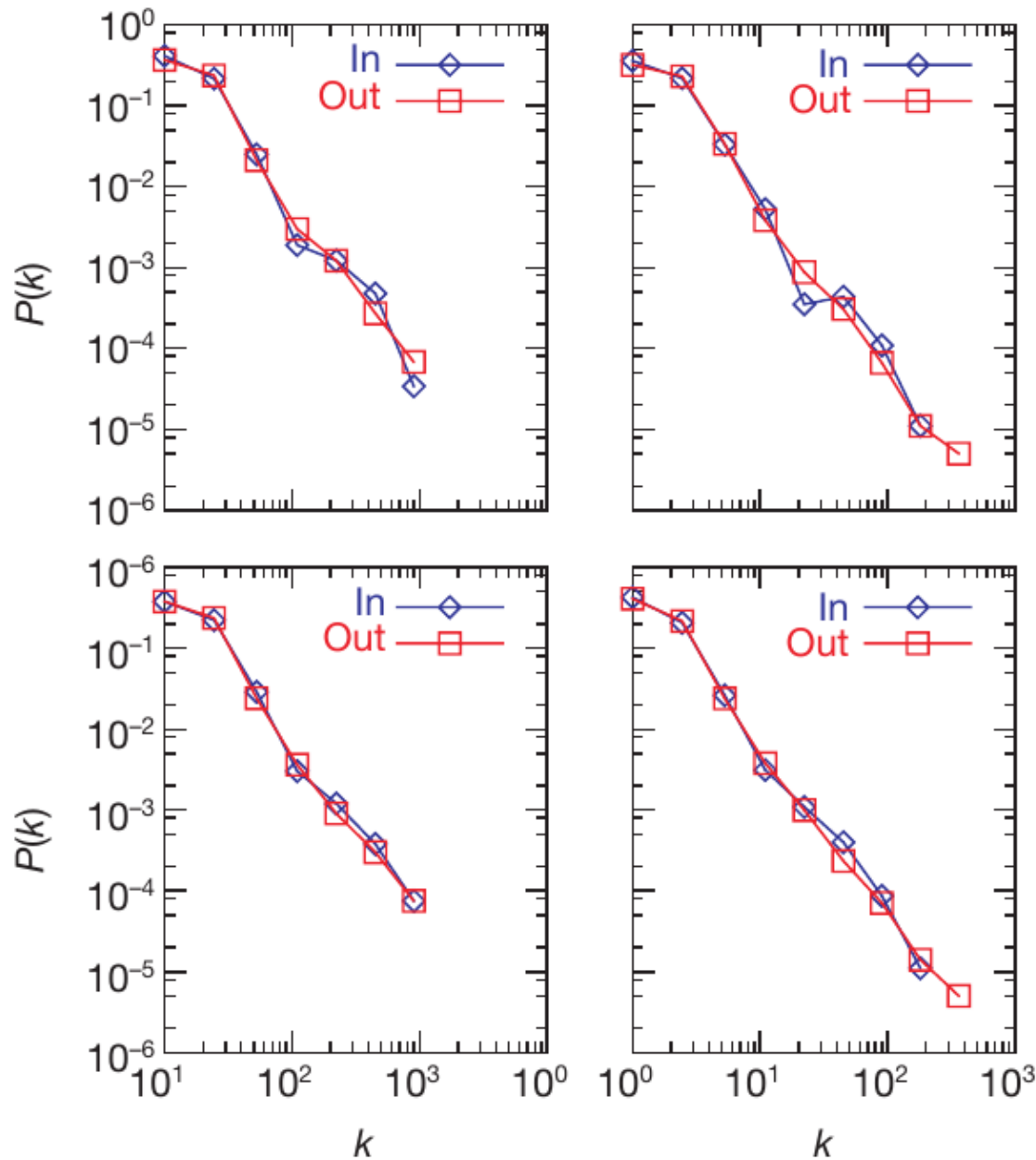


WorldWide Web



Power grid

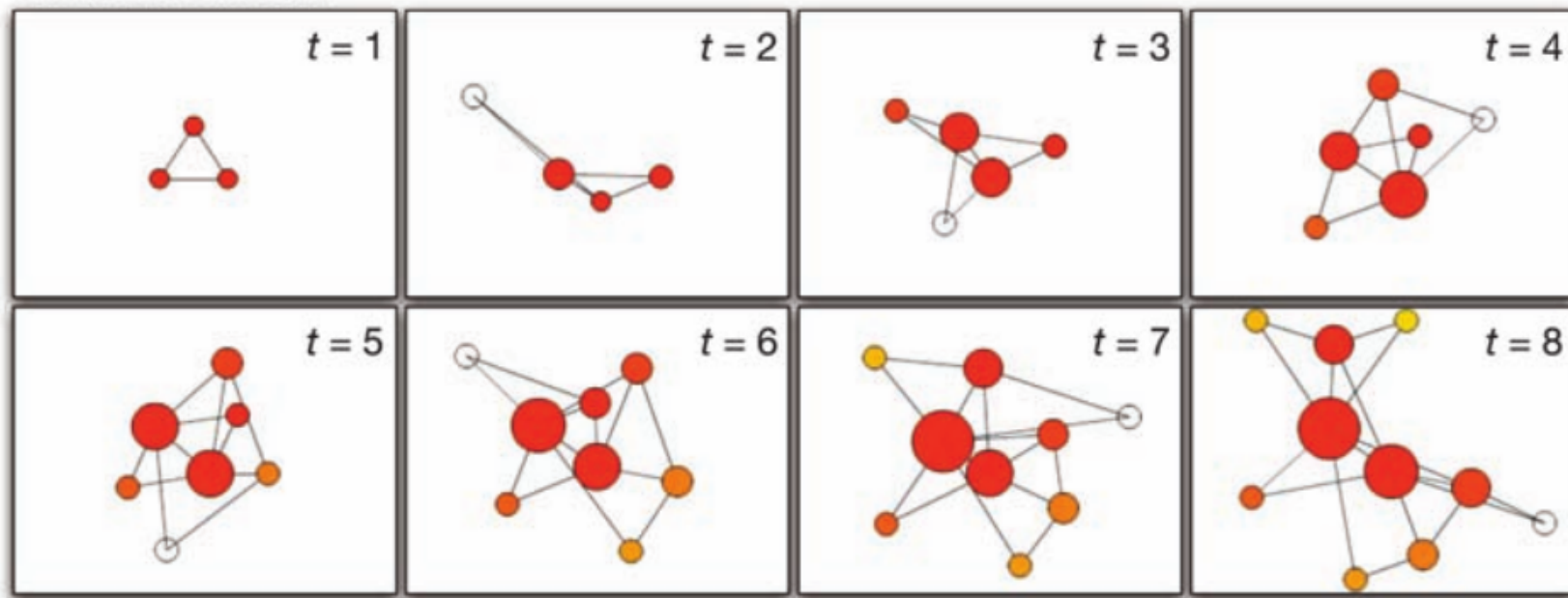
Metabolic networks are power-law



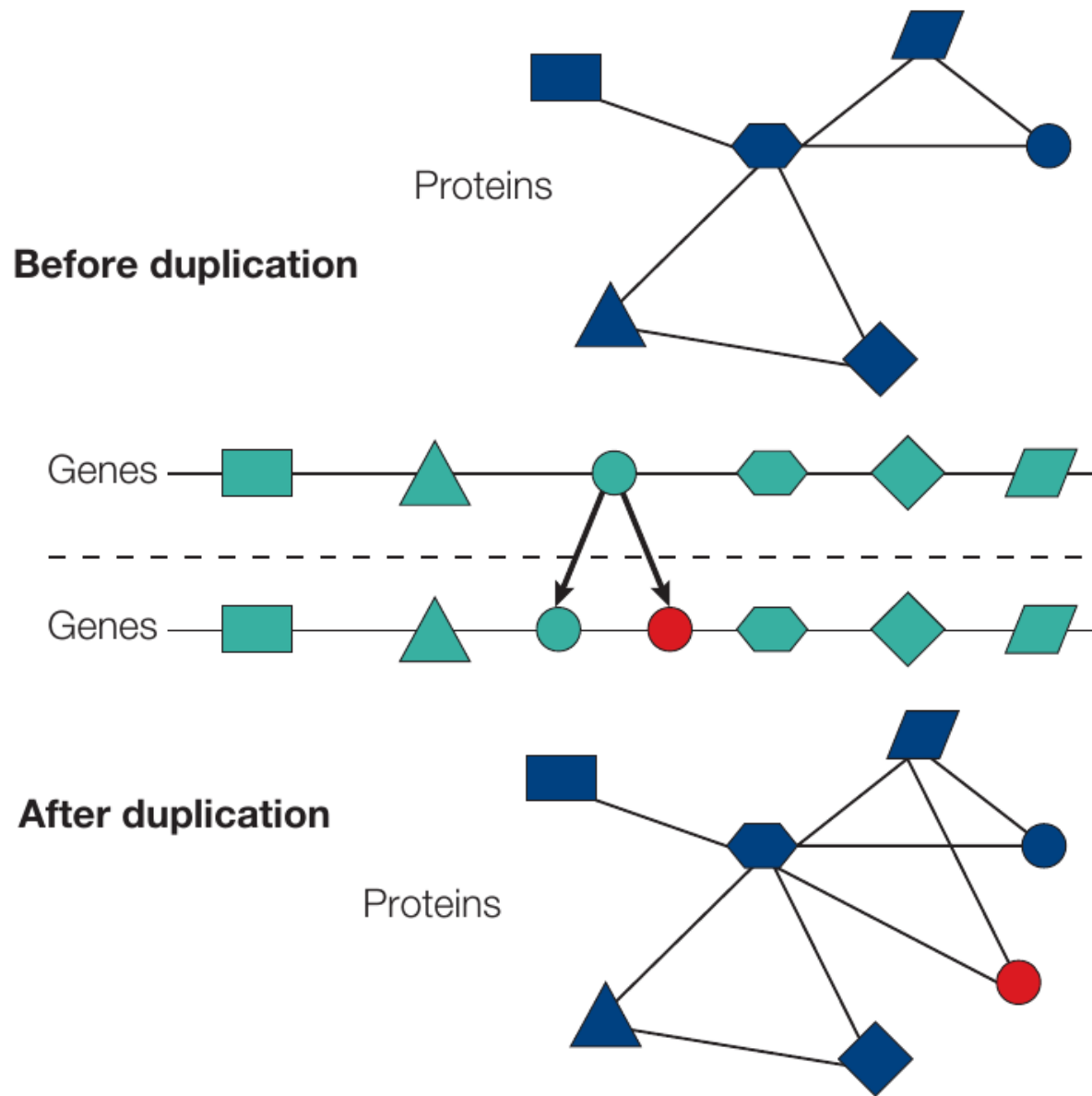
Metabolism of
different organisms

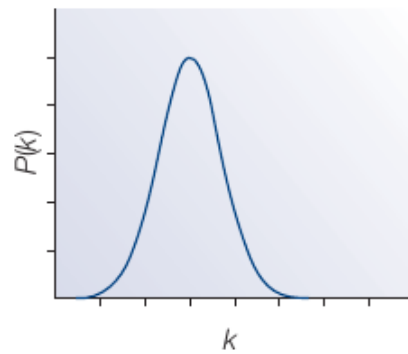
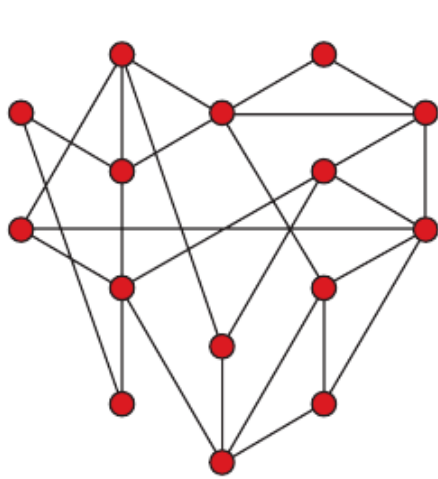
Creation = random, Evolution = power-law

- Random creation at once:
Do v times: pick a node among n , pick another among the same
→ random network
- Evolution:
Repeat until n nodes: add a new node, link it to existing nodes with a probability depending on their existing degree
→ power-law network

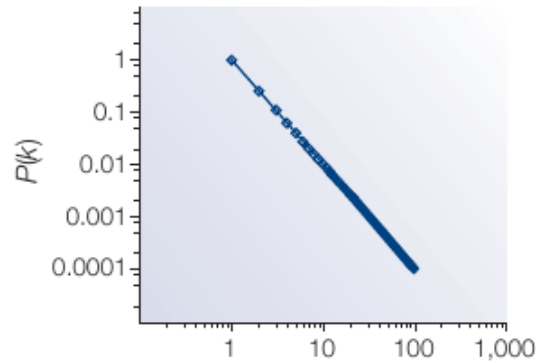
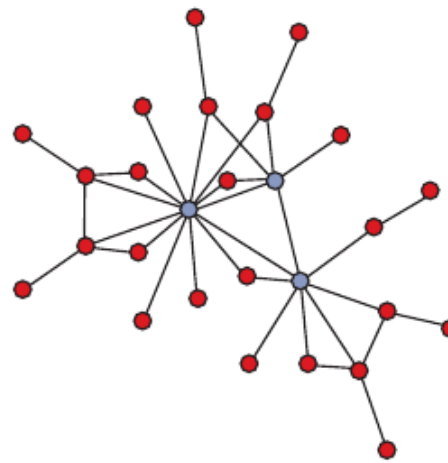


Gene duplications and power-law interactomes

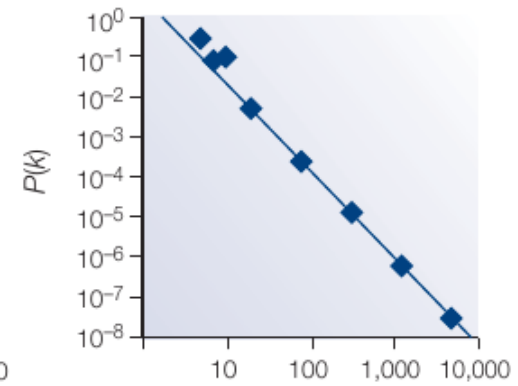
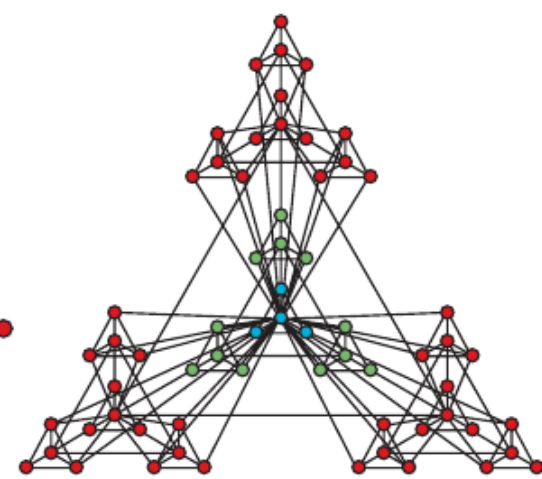




Random networks

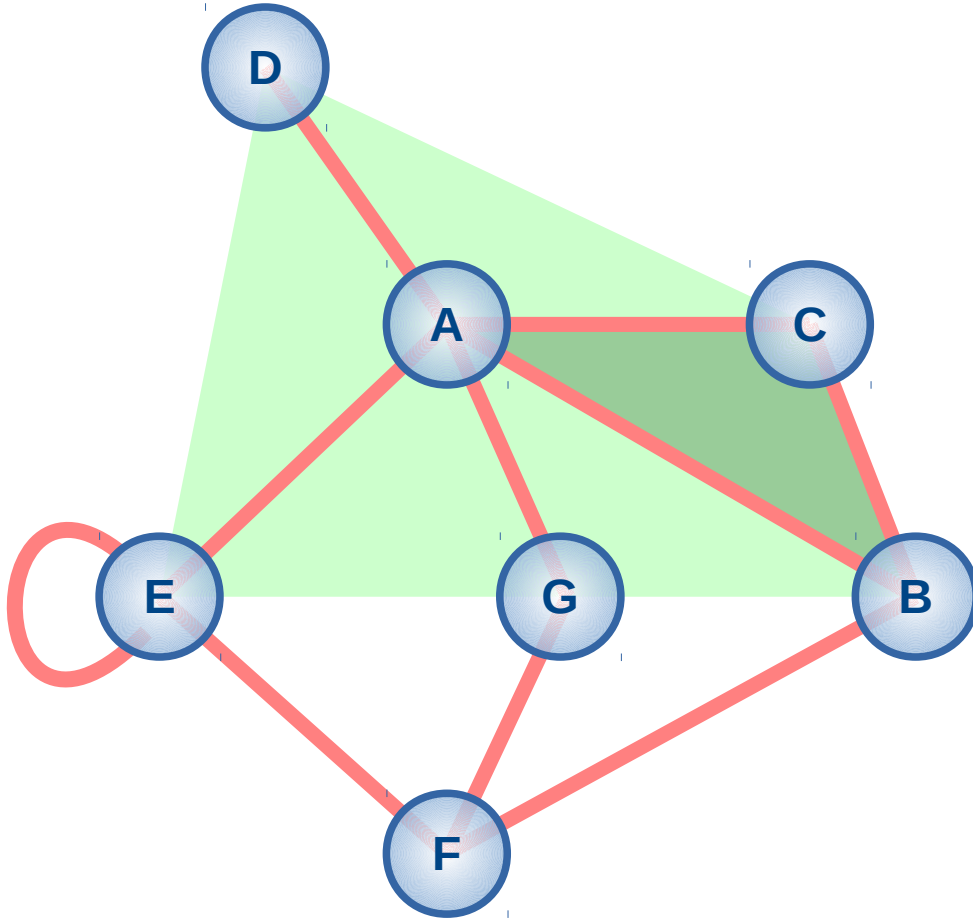


Power law networks



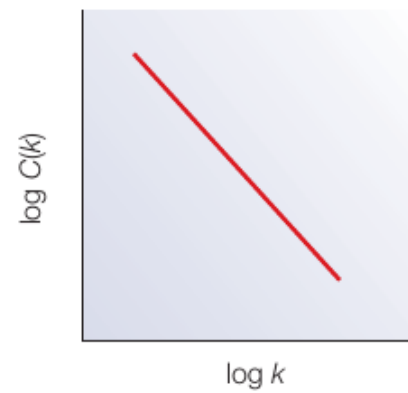
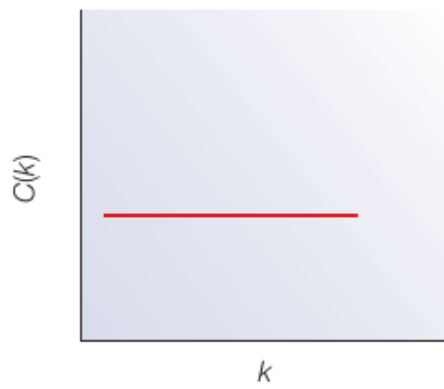
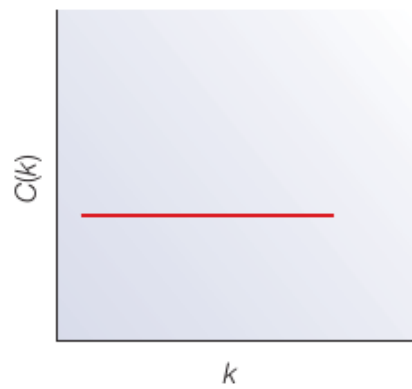
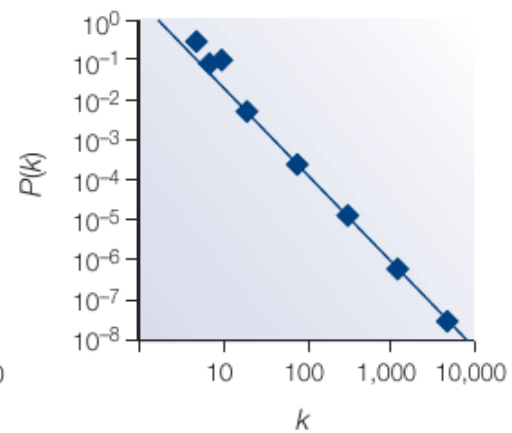
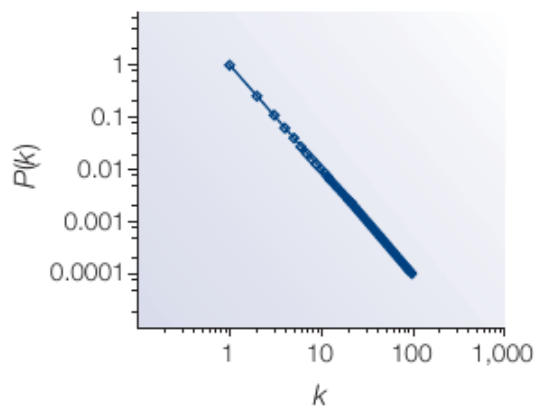
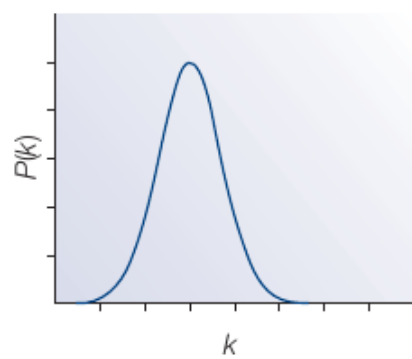
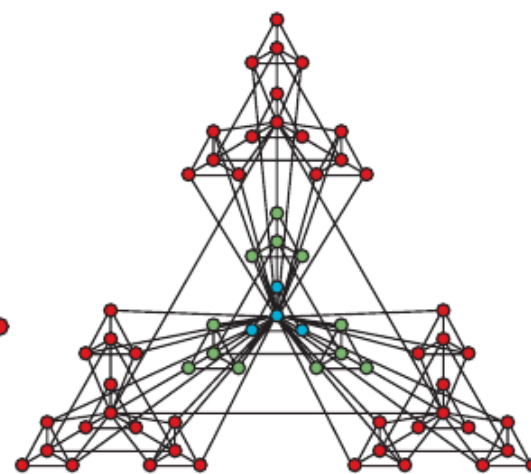
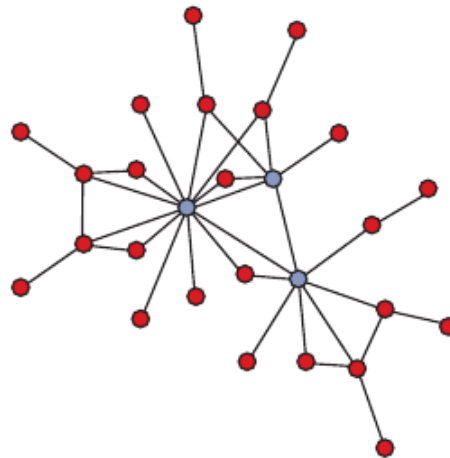
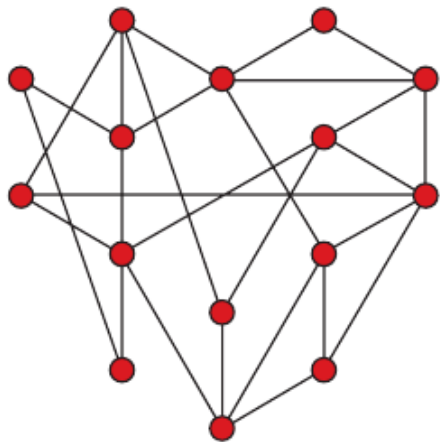
Hierarchical networks

Clustering (C): Proportion of possible cliques

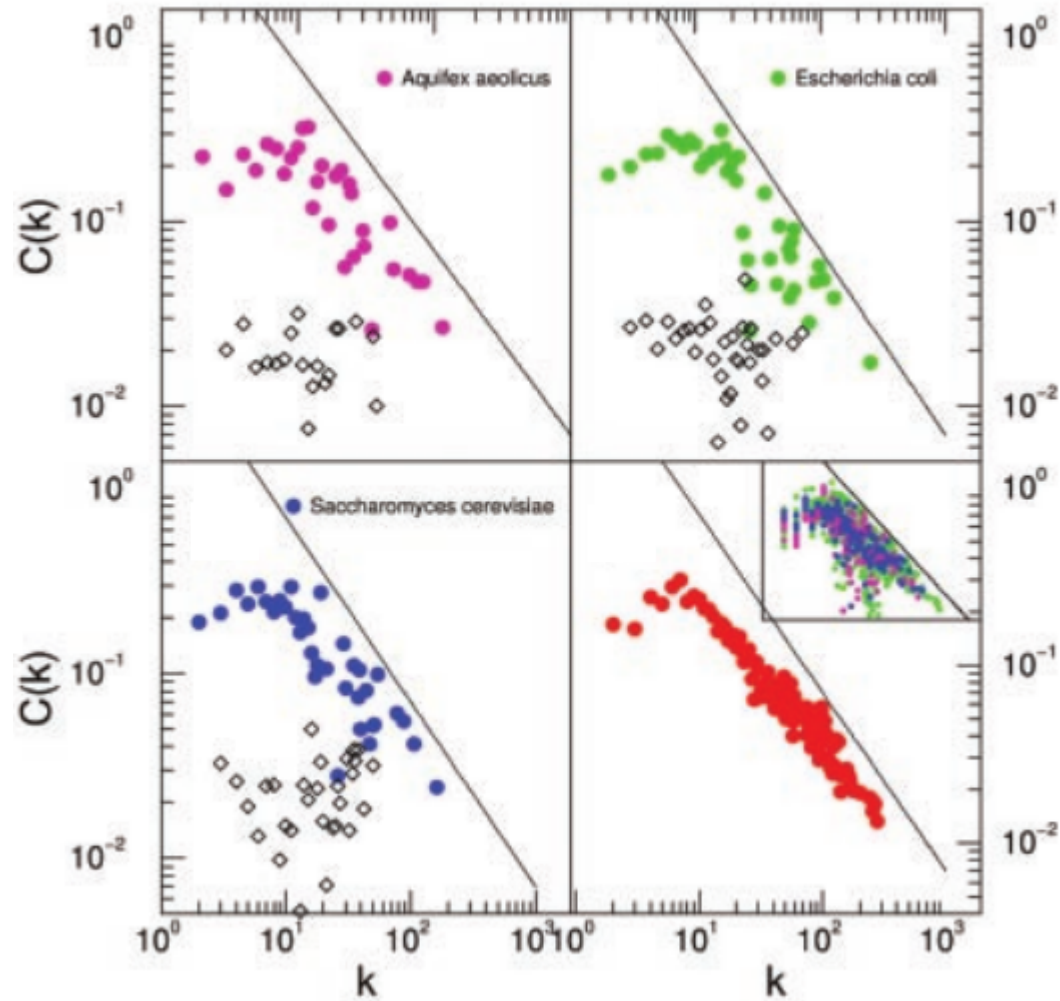


node	k	C
A	5	0.1
B	3	0.33
C	2	1
D	1	0
E	4	0
F	3	0
G	2	0

NB: nb of possible cliques = $\frac{k!}{2!(k-2)!}$

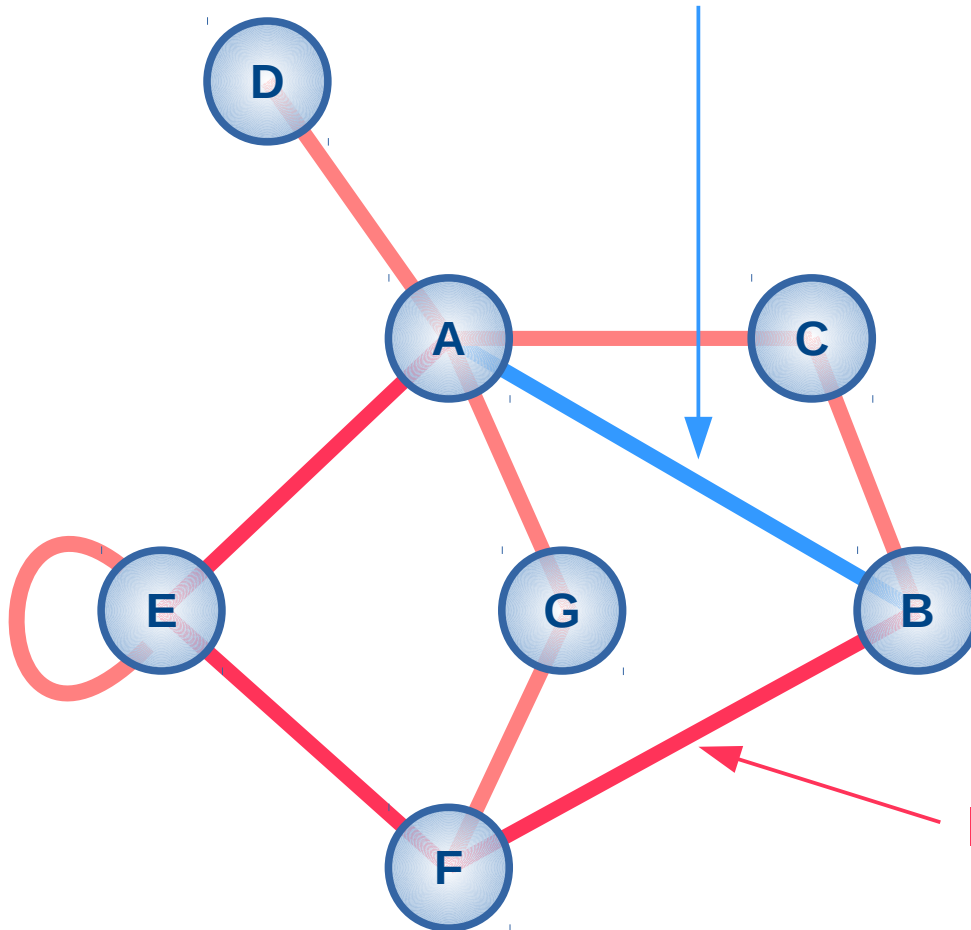


Metabolic networks are hierarchical



Shortest path (<SP>)

shortest path between A and B



node	k	C	<SP>
A	5	0.1	1.167
B	3	0.33	1.5
C	2	1	1.67
D	1	0	2
E	4	0	1.67
F	3	0	1.67
G	2	0	1.67

longest path between A and B

Example of shortest path: Erdős number

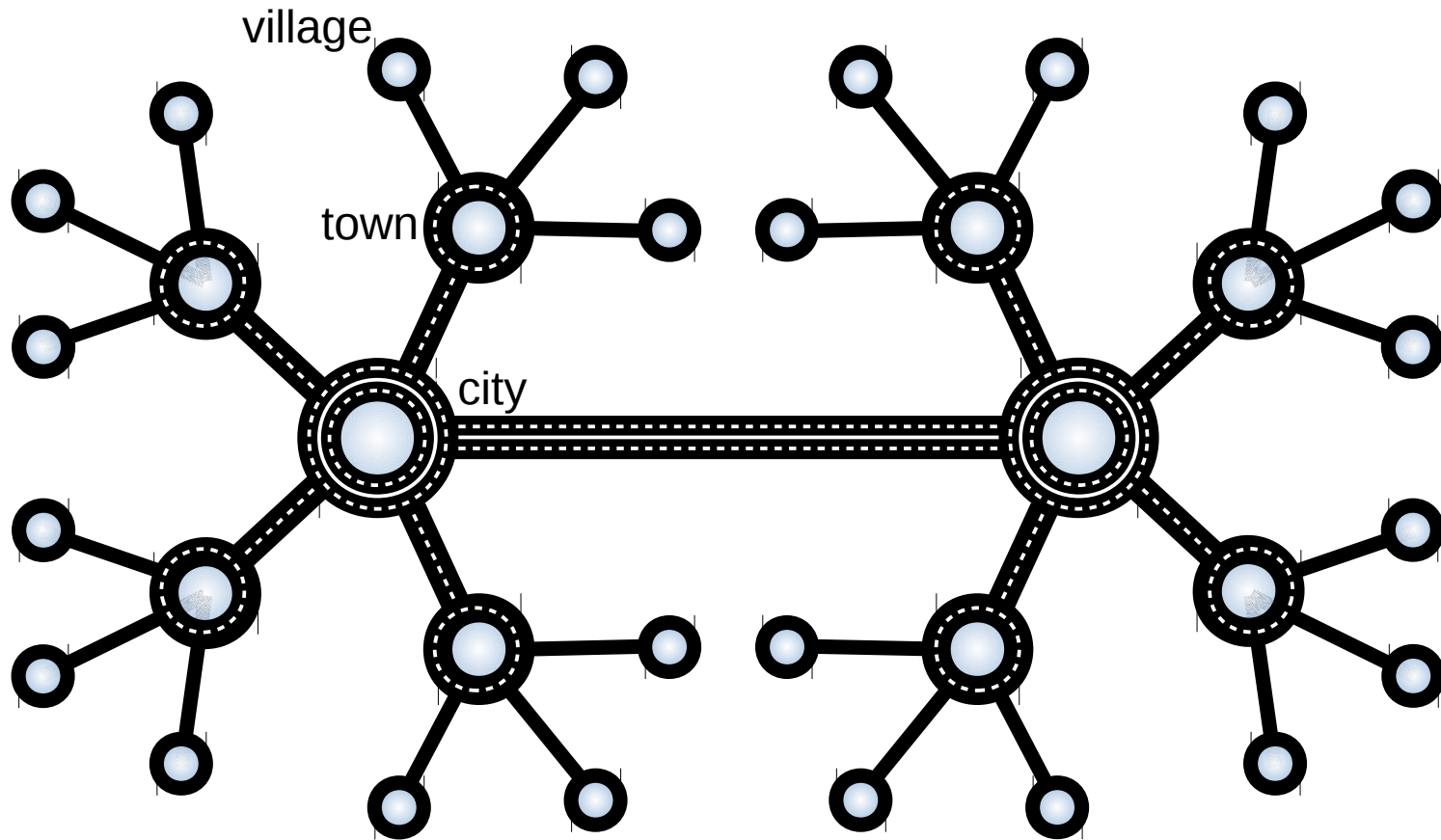


- 13 Field medals with $E_n=2$
- 2 Nobel Physics with $E_n=2$, 14 with $E_n=3$
- 4 Nobel Chemistry with $E_n=3$
- 3 Nobel Medicine with $E_n=3$
- 1 Nobel Economy with $E_n=2$. 8 with $E_n=3$
- My Erdős number is 4

(Le Novère – Doyle – Boyd – Diaconis – Erdős
Le Novère – Sauro – Nadim – Salamon - Erdős)

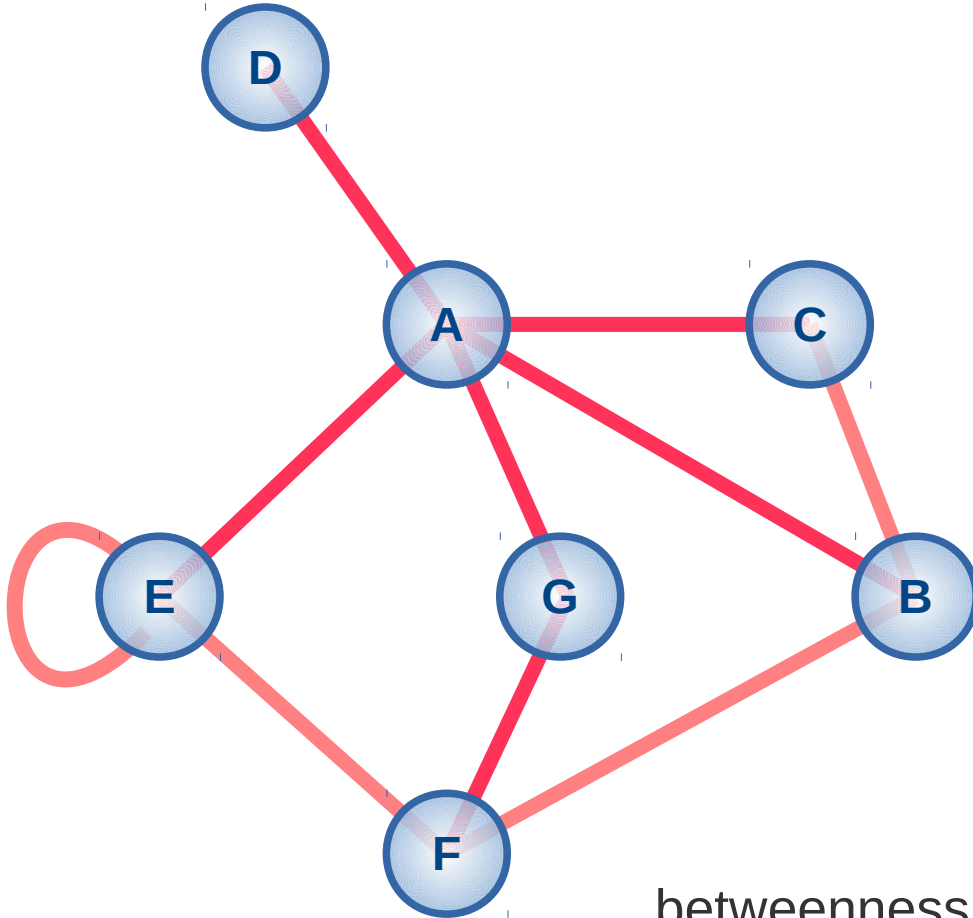


Power law often = small-world



The perfect road system=shortest average distance between two villages

Betweenness centrality (Cb)



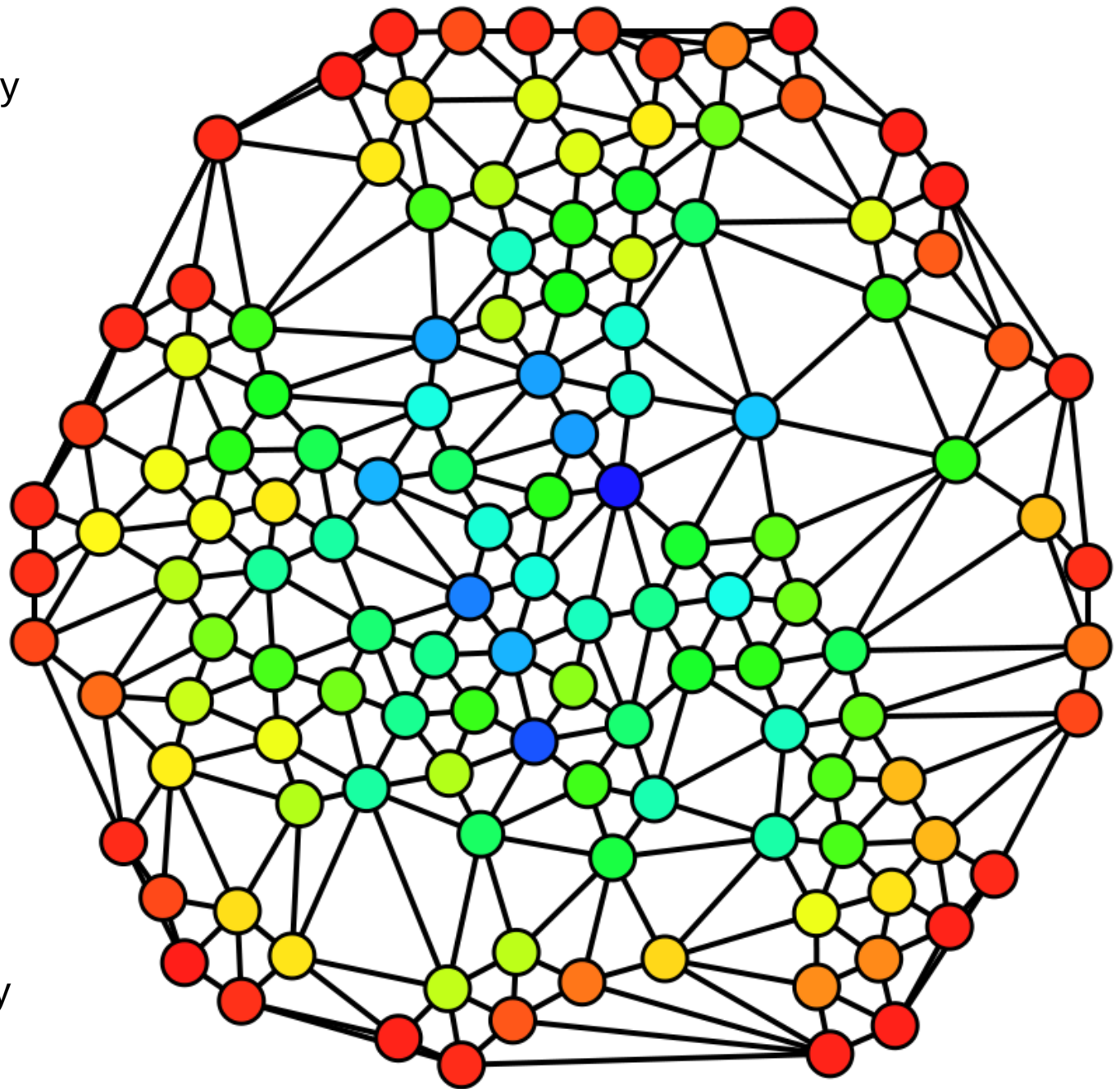
node	k	C	<SP>	Cb
A	5	0.1	1.167	0.567
B	3	0.33	1.5	0.111
C	2	1	1.67	0
D	1	0	2	0
E	4	0	1.67	0.044
F	3	0	1.67	0.1
G	2	0	1.67	0.044

betweenness = fraction of encountered shortest paths

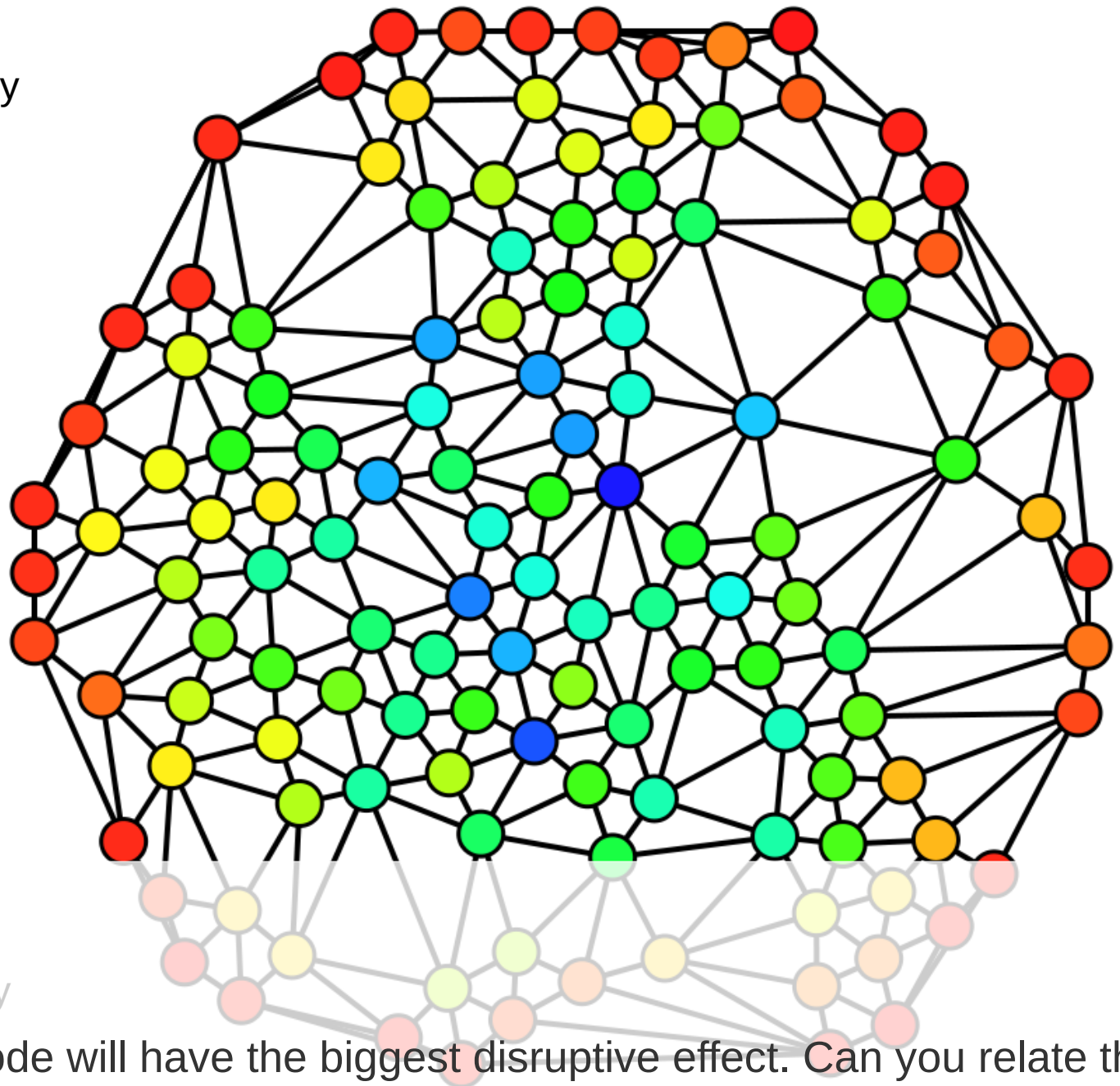
High betweenness centrality



Low betweenness centrality



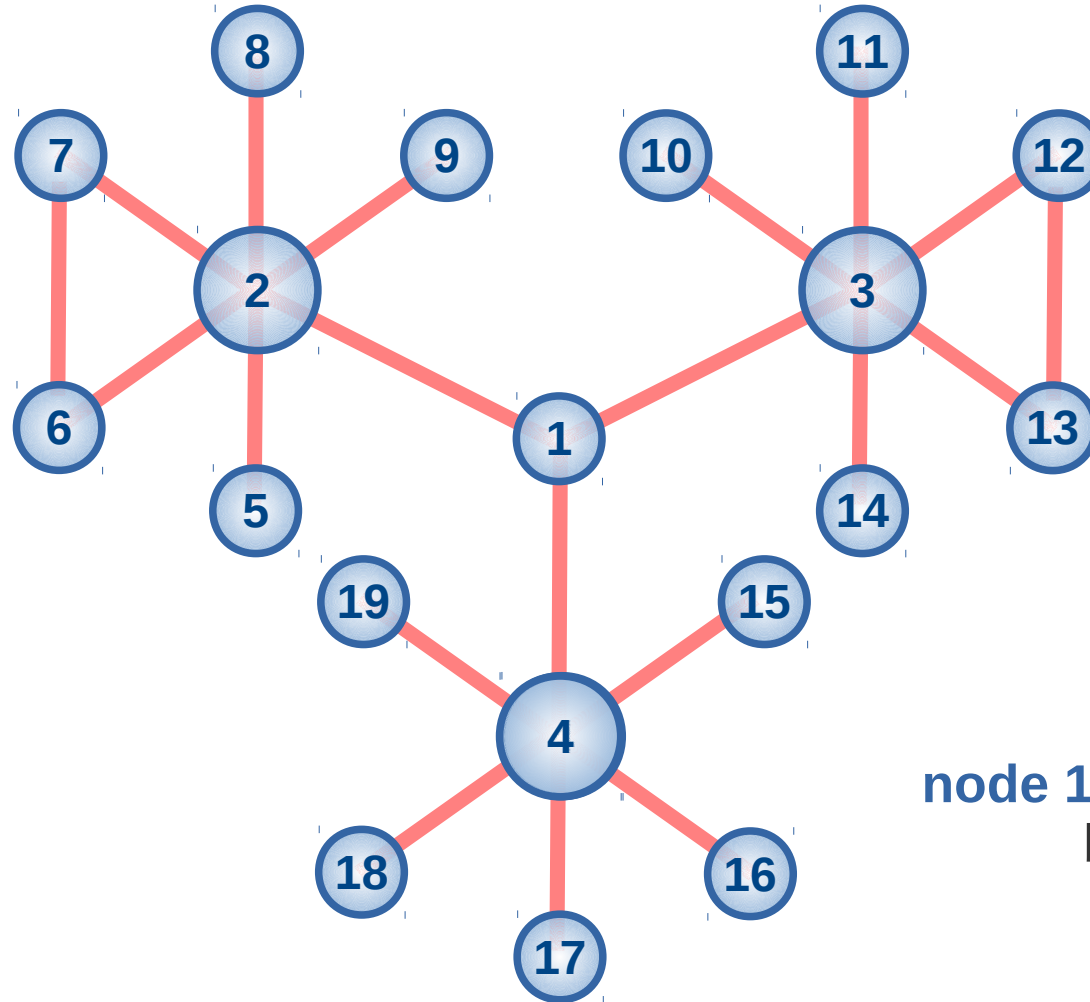
High betweenness centrality



Low betweenness centrality

Hitting the dark blue node will have the biggest disruptive effect. Can you relate that to causes and treatments of cancer?

No need to be highly connected to be important

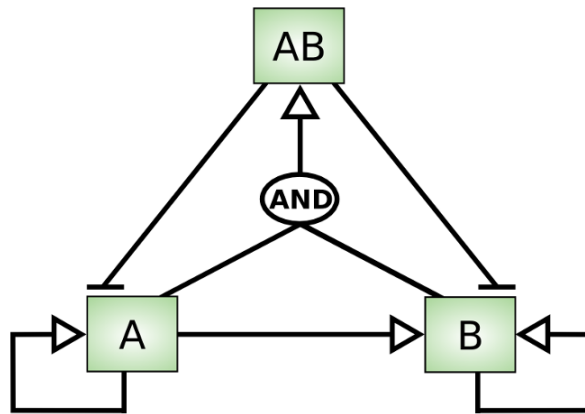


node 1 has low degree, low clustering
but largest betweenness

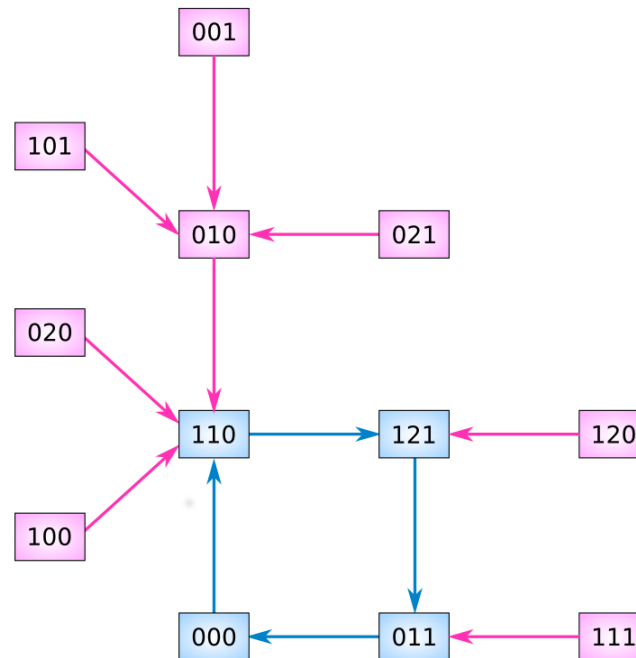
What's next?

Logic models

- Variables can take a discrete number of values, at least 2
- Transitions of output are expressed as logical combinations of input values
- Simulations can be:
 - synchronous: all the nodes are updated at once
 - asynchronous: nodes are updated one after the other
- One can add delays, inputs etc.



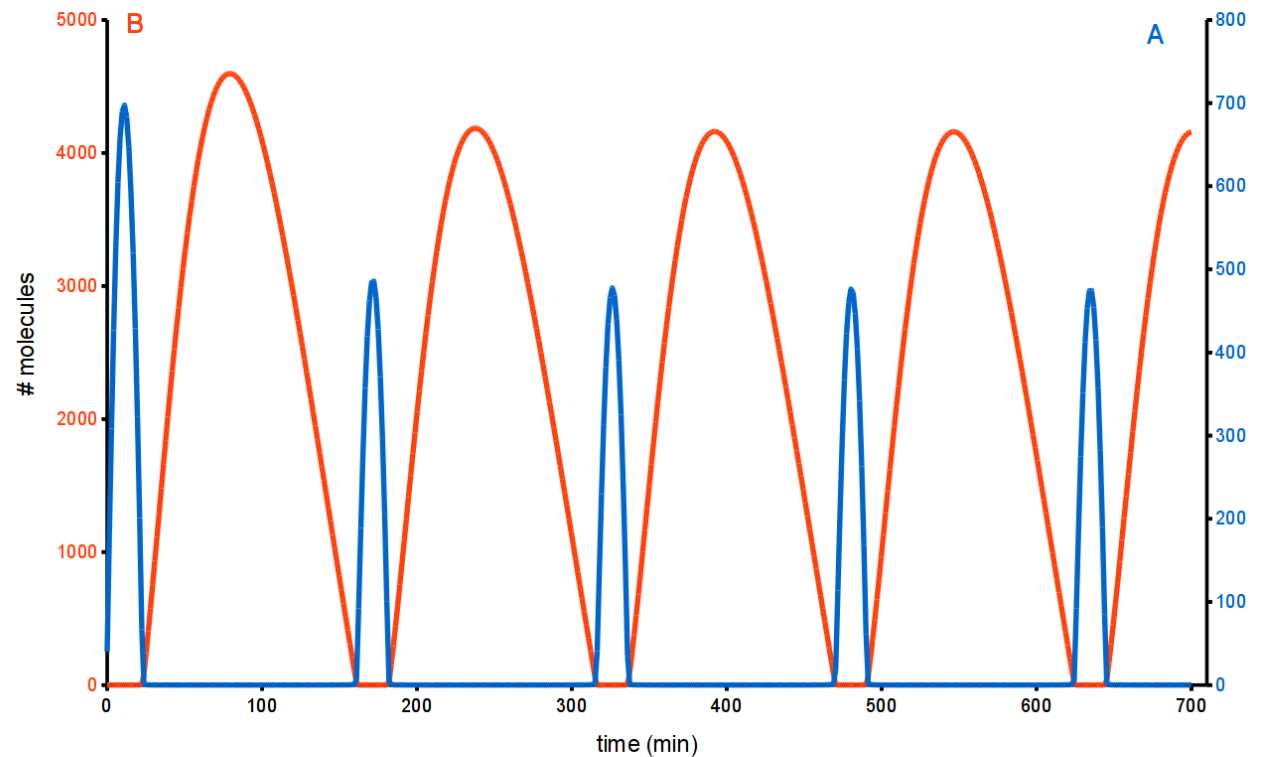
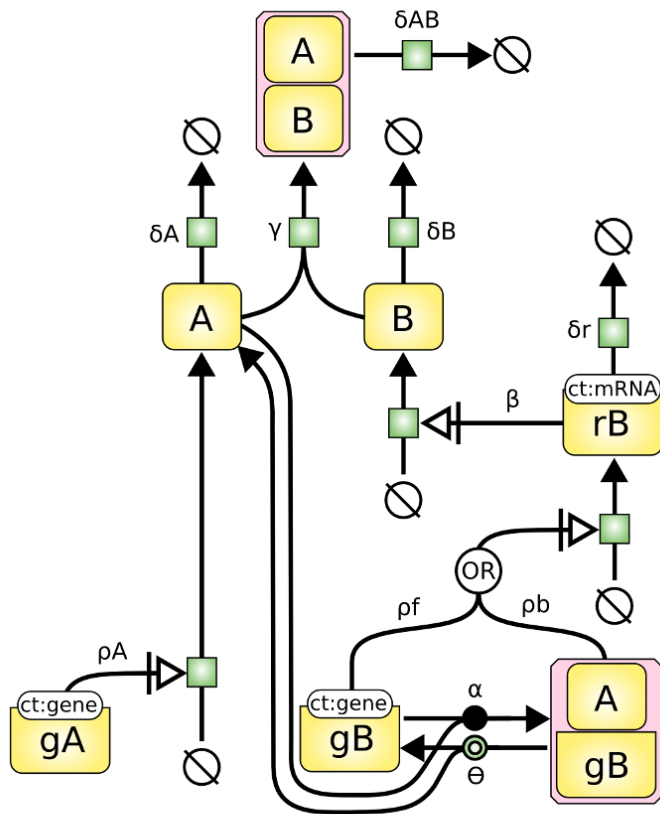
Influence diagram

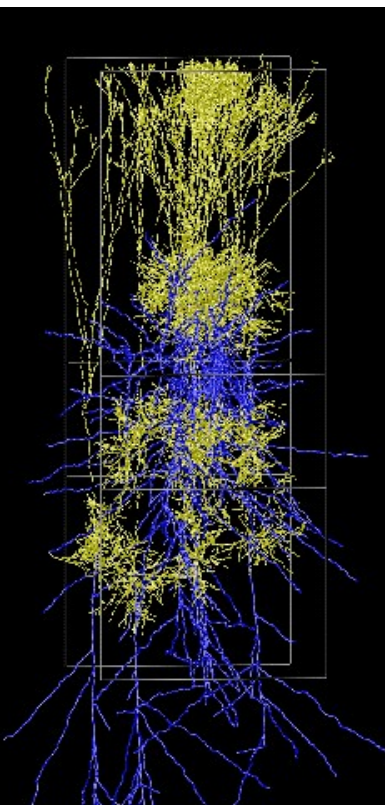


state diagram

Quantitative models

- Variables are concentrations or amounts
- Reactions rates are calculated based on quantitative values
- Simulations deterministic or stochastic





Any question: n.lenovere@gmail.com

Further information:

Barabási & Zoltán. Network biology: understanding the cell's functional organization.

Nature Reviews Genetics (2004) **5**: 101-113

Le Novère. Quantitative and logic modelling of molecular and gene networks.

Nature Reviews Genetics (2015) **16**: 146–158

