

# Structure of biological networks

Presentation by  
Atanas Kamburov

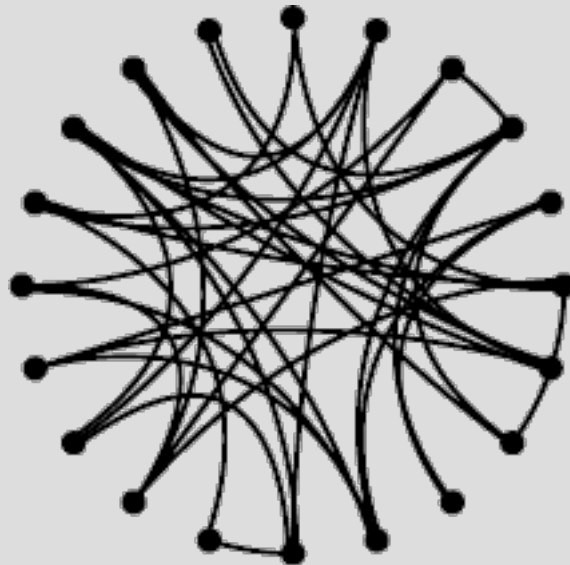
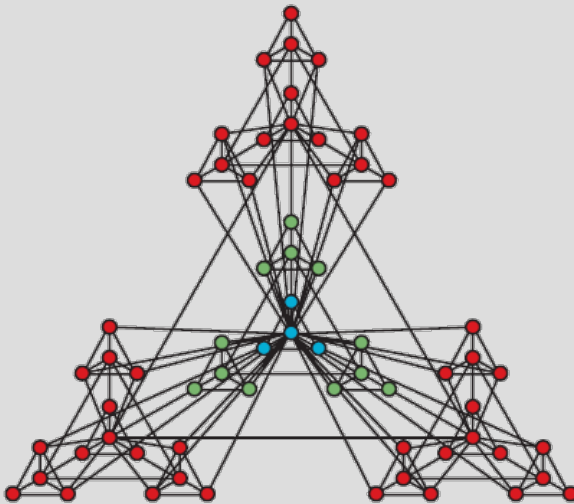
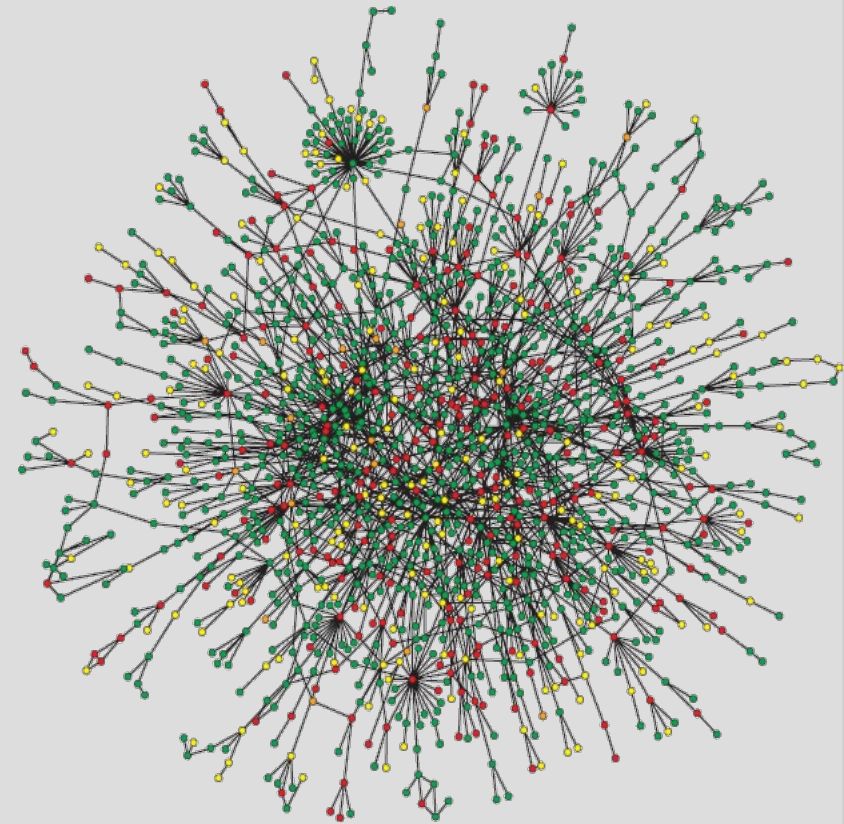
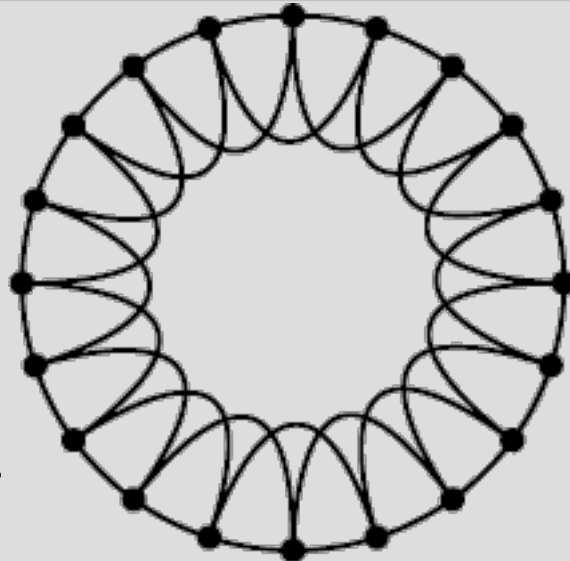
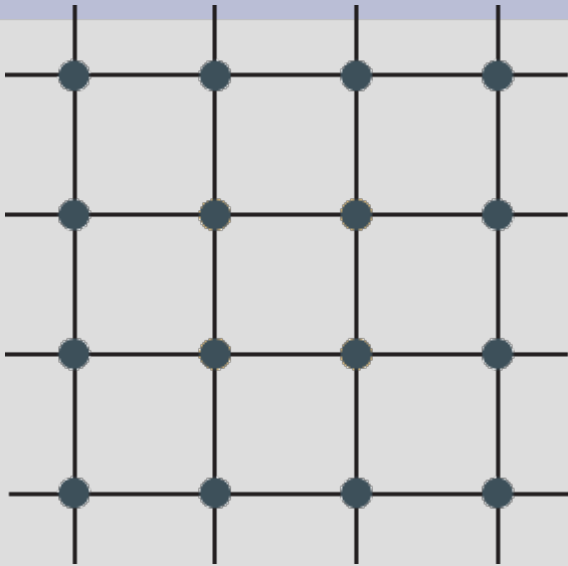
# Overview

- Motivation
- Definitions
- Large-scale properties of cellular networks
  - the scale-free property
  - the small-world effect
  - disassortativity
  - modularity
- Bottom-up characterization of cellular networks:  
network motifs
- Summary
- Literature

# Motivation

- Biological functions can rarely be related to *individual molecules*
- Instead, they arise from the **complex molecular interactions** that take place in the cell
- The numerous interactions form **interaction networks** where nodes represent molecules and edges represent functional interactions
- understanding the **structure** and dynamics of such networks is a central challenge for biology
  - network biology, systems biology

# Network shapes



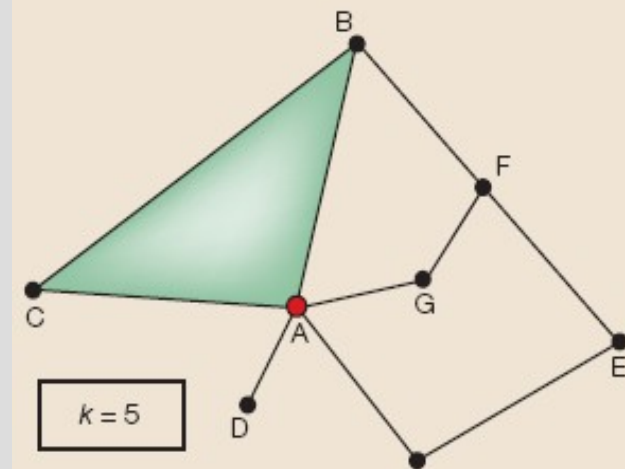
# Network characterization

- Networks can be characterized using global and/or local measures
- global properties
  - node connectivity distribution
  - path lengths
- local properties
  - node clustering ratio
  - frequent patterns (motifs)

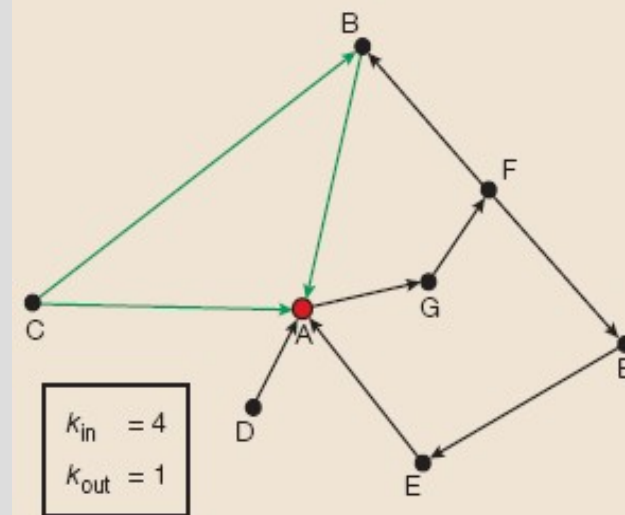
# Some definitions

- Networks can be **directed** or **undirected** depending on the nature of the interactions
  - *directed*: material flow, information flow
  - *undirected*: physical interactions
- **Degree (connectivity)  $k_i$  of a node** is the number of links that the node has to other nodes
  - directed networks have two types of degree: **indegree** and **outdegree**
- **Average degree  $\langle k \rangle = \text{sum}(k_i) / \text{\#nodes}$**
- **Degree distribution  $P(k)$**  is the probability that a randomly picked node has  $k$  links

a Undirected network

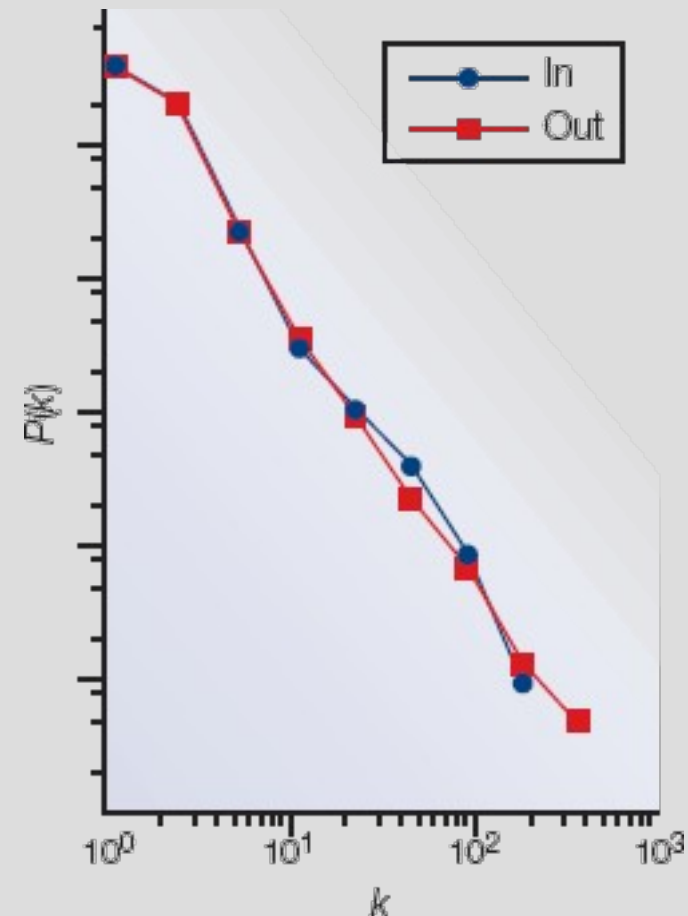


b Directed network



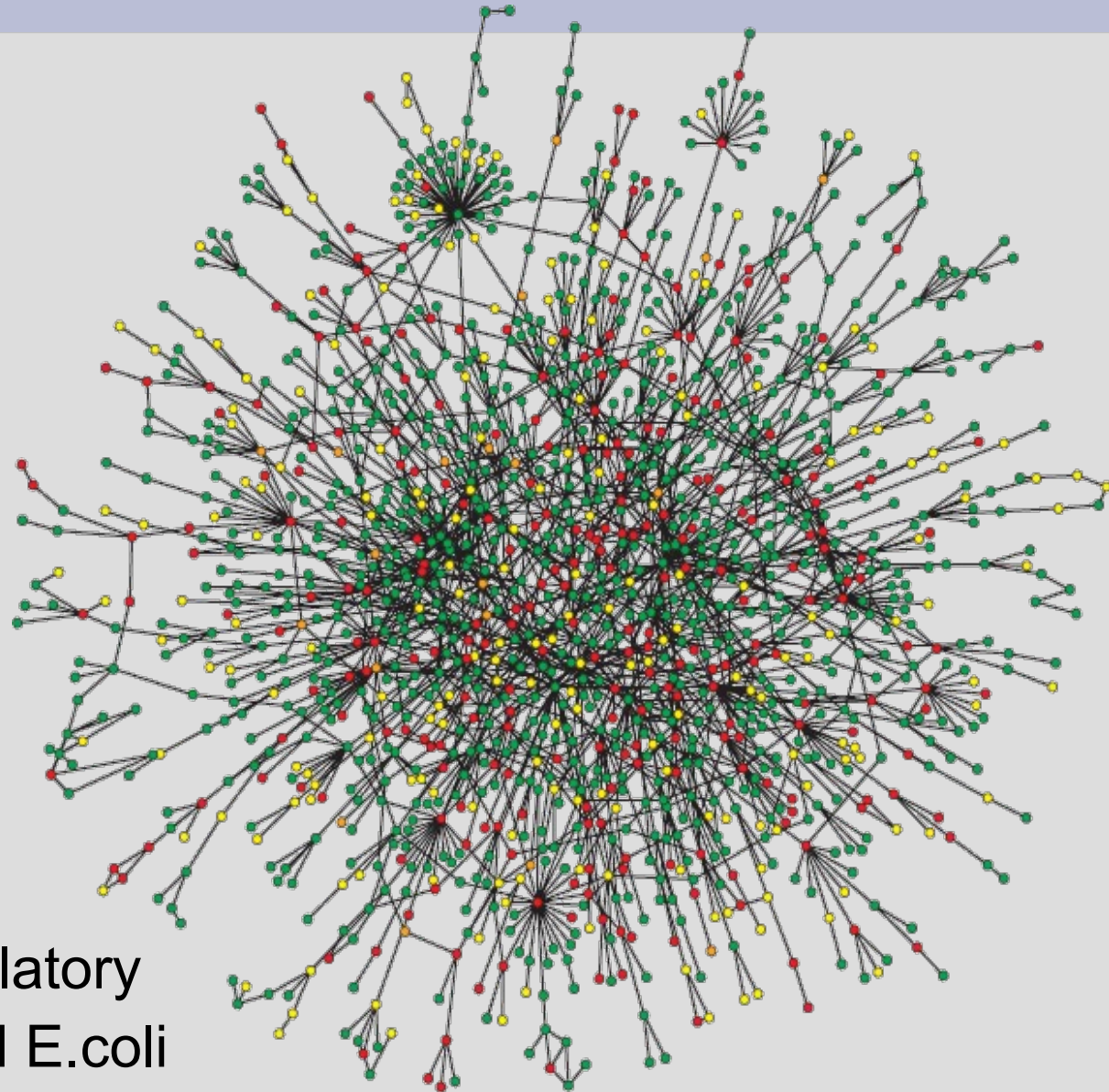
# The scale-free property

- For decades, cellular networks were considered to be either completely regular or completely random
- Both models are unable to explain the existence of nodes with high connectivity
- Recent studies show that in many biological networks,  $P(k)$  follows a power law distribution  $P(k) \sim k^{-\gamma}$  with  $2 < \gamma < 3$
- **Scale-free** network
  - many nodes with few links
  - few nodes with many links (**hubs**)



# The scale-free property

- Metabolic networks have scale-free nature
- Most protein interaction networks and gene regulatory networks also have scale-free nature
- **Not every** cellular network is scale-free
  - e.g. transcription regulatory networks of *S.cer.* and *E.coli*



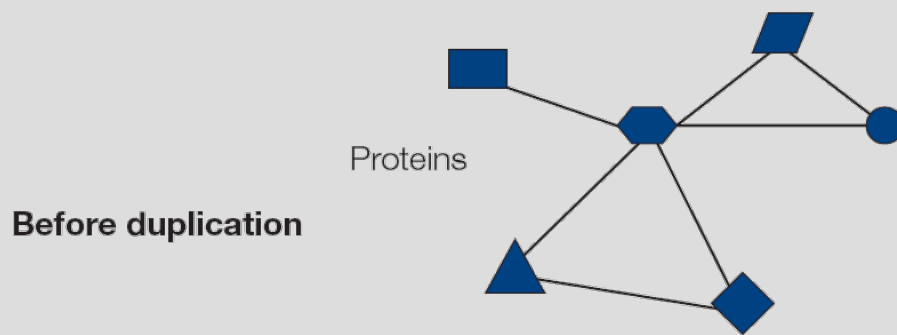
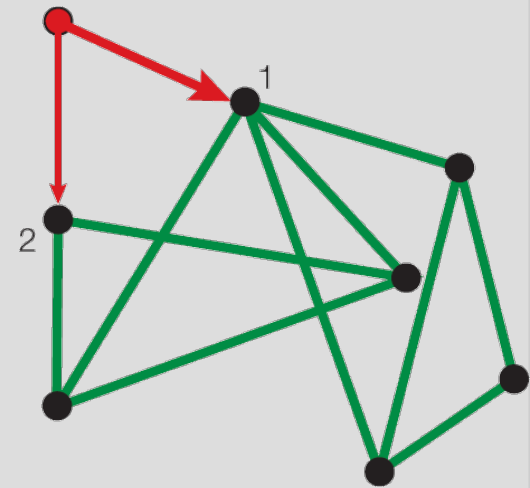


# The scale-free property

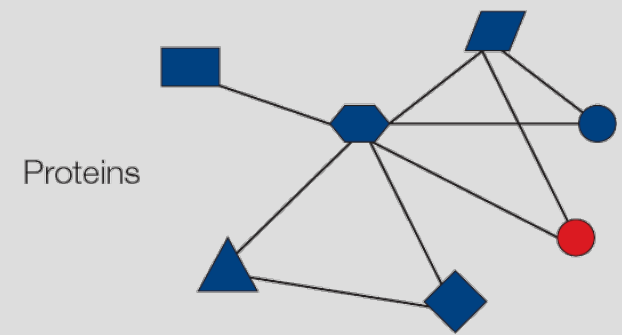
- Effects of the scale-free nature of cellular networks
  - Robustness: random removal of nodes often does not affect the network integrity and function
  - Attack vulnerability of hubs: if any hubs are removed, the networks is disintegrated
    - In yeast, only 10% of the proteins with less than 5 links are essential; from proteins with more than 15 links, over 60% are essential

# The scale-free property

- Evolutionary origin of the scale-free nature of cellular networks:  
**two fundamental processes**
  - growth process: new nodes join the system over an extended time period
  - preferential attachment: new nodes prefer to connect to nodes with many connections
- The scale-free property of protein interaction networks is probably rooted in gene duplication
  - products of duplicated genes have same interaction partners
  - the duplication of a hub's interaction partner is more probable



After duplication

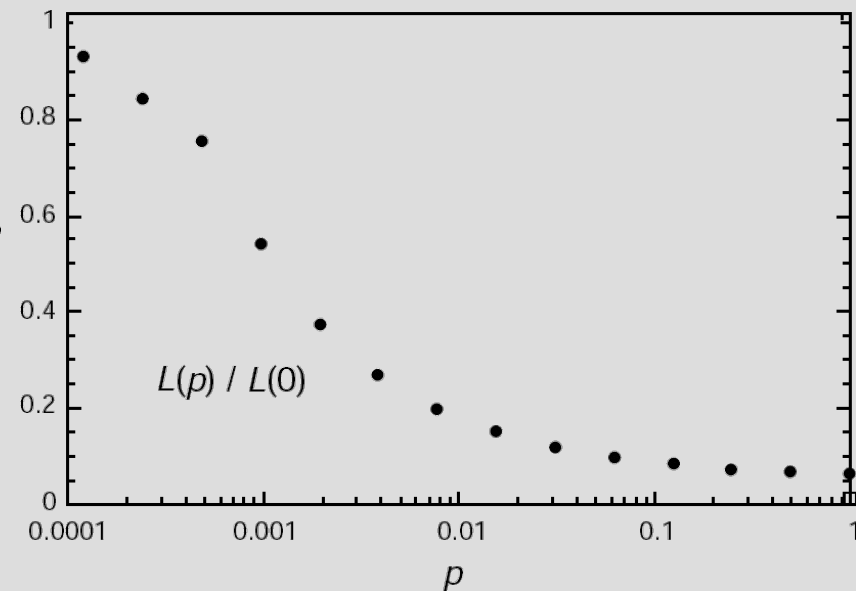
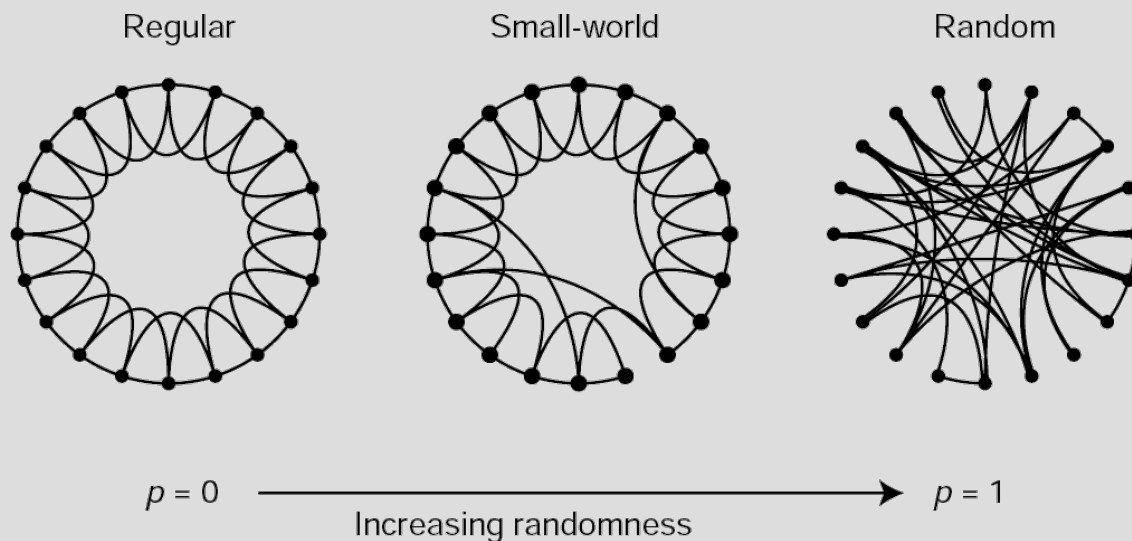


# Disassortativity of biological networks

- An interesting observation: hubs avoid connecting to each other (**disassortativity**)
- To this point, disassortativity cannot be explained by network biologists

# The small-world property

- In regular networks, the average path length  $L$  between couples of nodes is relatively long
- $L$  is shorter for random networks: the **small-world** property
- Even a few rewiring procedures of a regular network suffice to achieve a small-world effect

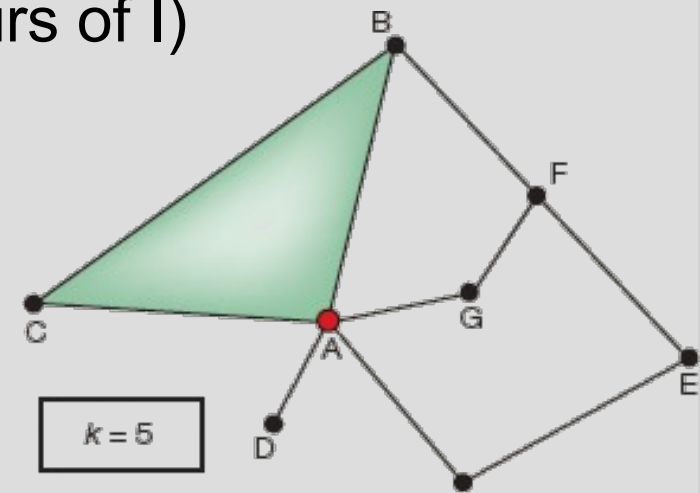


# The small-world property

- Biological networks have even smaller average path lengths than random networks: **ultra-small-world** effect
  - first documented for metabolism
    - often, 3-4 reactions suffice to convert one metabolite to another
    - local concentration perturbations quickly reach the whole system
    - ultra-small-world effect underlies selection: mean path length of metabolic networks is the same in bacteria and complex multicellular species

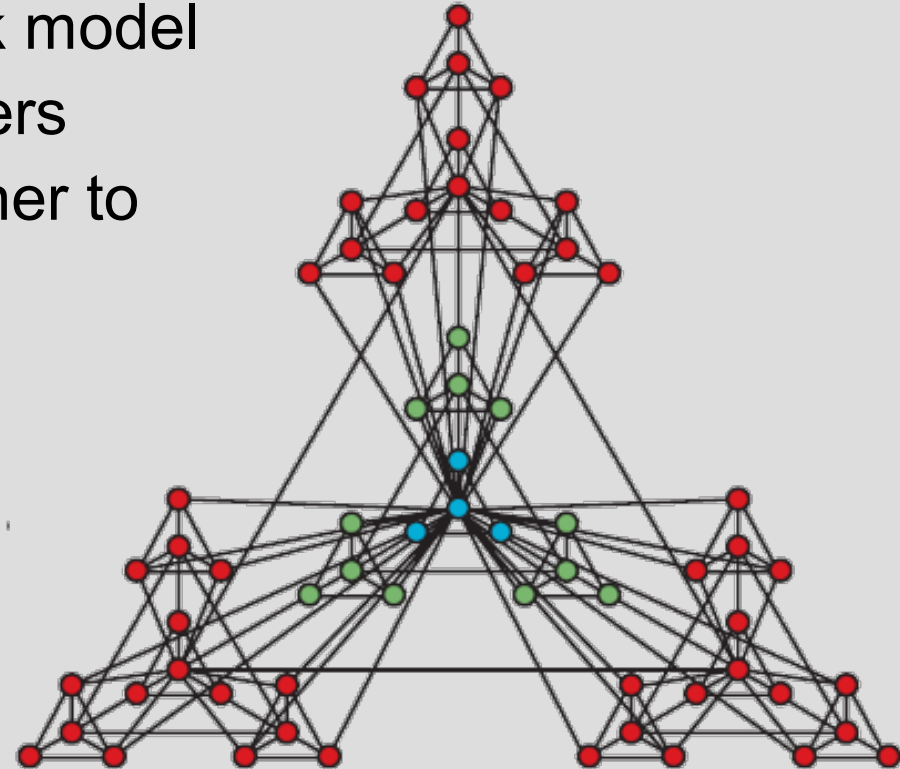
# Modularity of cellular networks

- Cellular functions often carried out in a highly **modular** manner
  - protein complexes
  - temporally coregulated molecules that control cell cycle
- A sign for this is the significantly high **average clustering coefficient  $\langle C \rangle$**  of most cellular networks
  - **The clustering coefficient  $C_i = 2n_i / k_i(k_i - 1)$**  shows the fraction of neighbours of node  $i$  that are directly linked to each other ( $n_i$  is the number of links between neighbours of  $i$ )
  - $\langle C \rangle = \text{sum}(C_i) / \text{number of nodes}$



# Modularity of cellular networks

- Functional modules are interconnected in the cell to form a “Network of networks”
- The scale-free property and modularity coexist in the hierarchical network model (Ravasz et al.,2002), where clusters are combined in an iterative manner to generate a hierarchical network



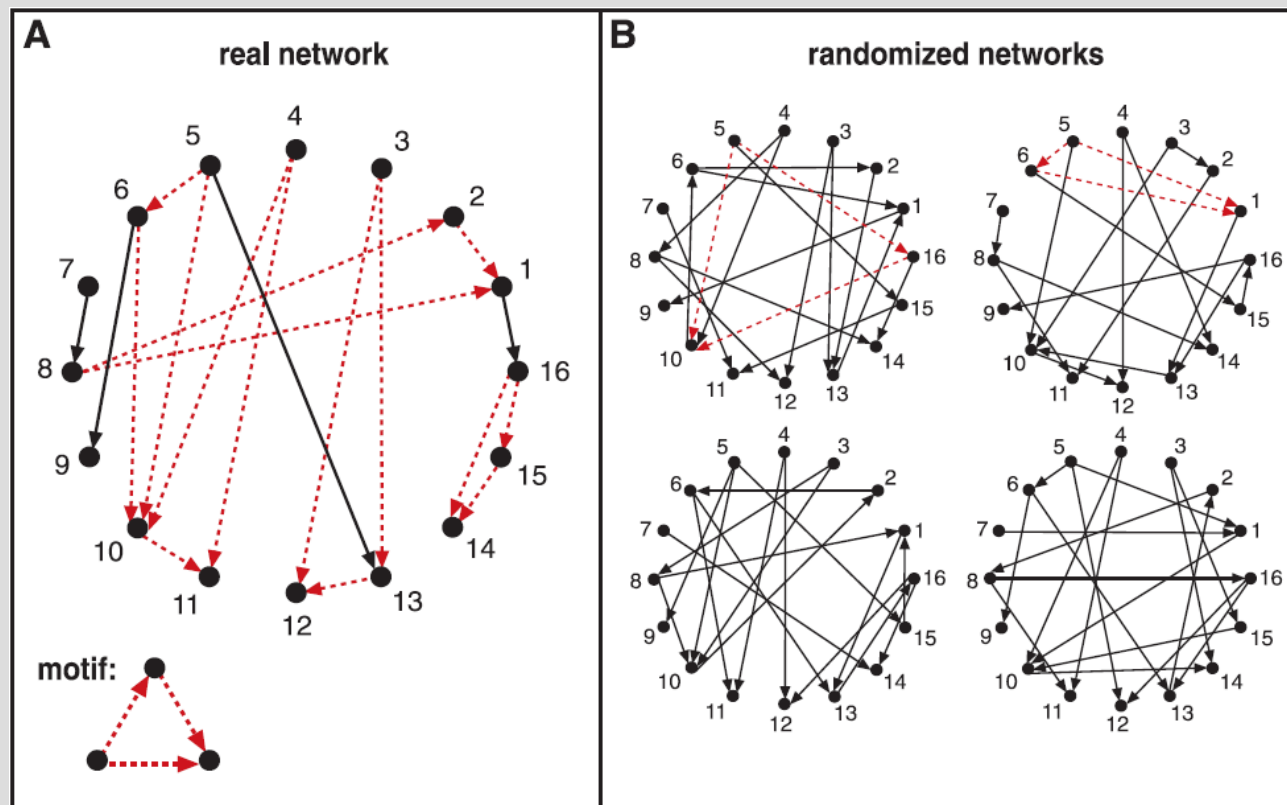
# Architectural features of cellular networks



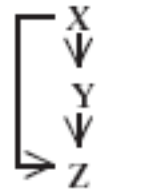


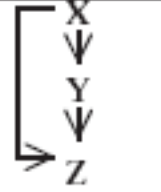





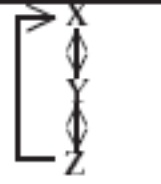


- Topological properties of most biological networks
  - **scale-free**
    - **robustness**
    - **attack vulnerability** of hubs
  - **disassortativity**
  - **ultra-small-world** property
  - **modularity**



# Network motifs

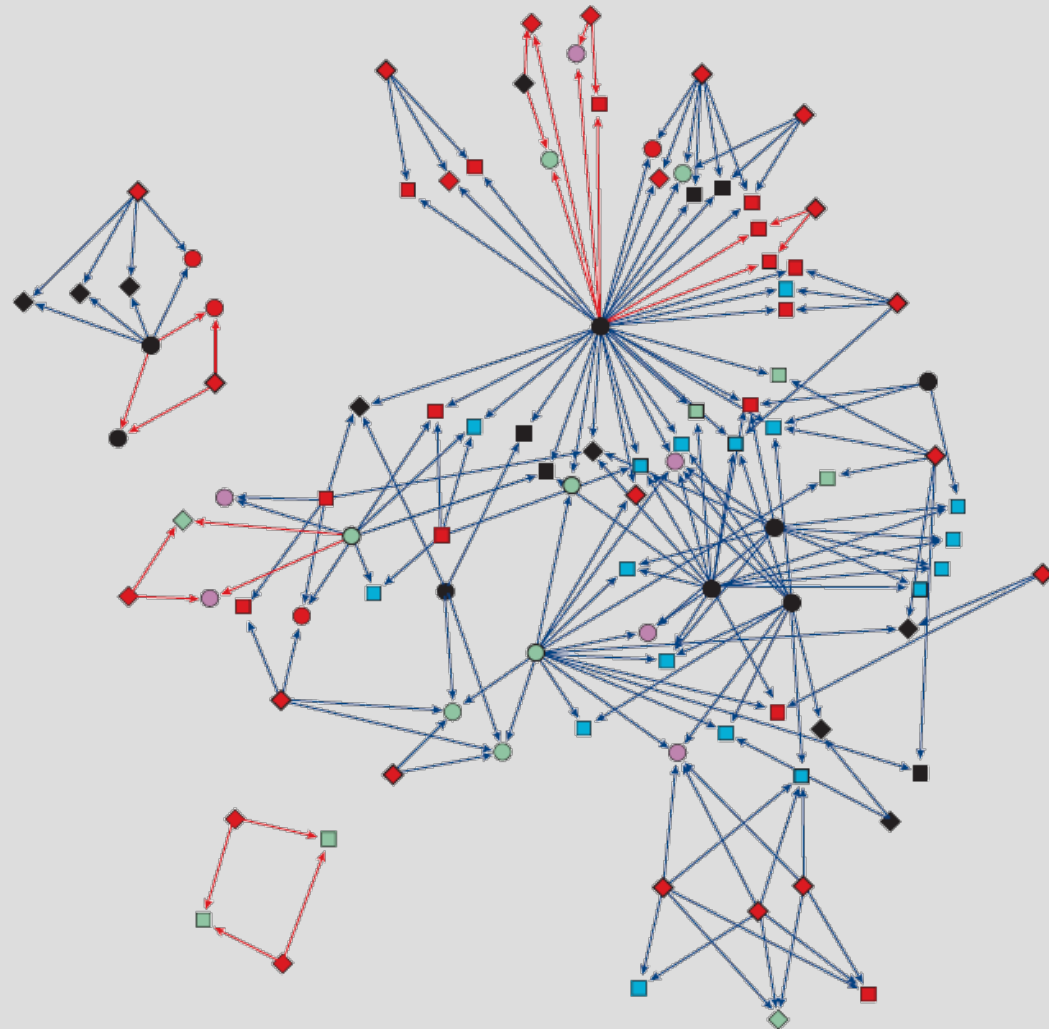
- Scale-free and hierarchical features determine the large-scale feature of networks
- An alternative approach for network analysis: finding subgraphs that are significantly overrepresented (motifs)



| Network   | Nodes   | Edges  | $N_{real}$  | $N_{rand} \pm SD$              | Z score   | $N_{real}$            | $N_{rand} \pm SD$   | Z score                 | $N_{real}$ | $N_{rand} \pm SD$ | Z score |
|---|---------|--------|---|--------------------------------|---|-----------------------|---|-------------------------|------------|-------------------|---------|
| <b>Gene regulation (transcription)</b>                      |         |        |     | Feed-forward loop              |     | Bi-fan                |   |                         |            |                   |         |
| <i>E. coli</i>  | 424     | 519    | 40  | 7 ± 3                          | 10  | 203                   | 47 ± 12   | 13                      |            |                   |         |
| <i>S. cerevisiae</i> *                                      | 685     | 1,052  | 70  | 11 ± 4                         | 14  | 1812                  | 300 ± 40  | 41                      |            |                   |         |
| <b>Neurons</b>  |         |        |    | Feed-forward loop              |    | Bi-fan                |    | Bi-parallel             |            |                   |         |
| <i>C. elegans</i> †   | 252     | 509    | 125   | 90 ± 10                        | 3.7   | 127                   | 55 ± 13   | 5.3                     | 227        | 35 ± 10           | 20      |
| <b>Electronic circuits (forward logic chips)</b>            |         |        |    | Feed-forward loop              |    | Bi-fan                |    | Bi-parallel             |            |                   |         |
| s15850  | 10,383  | 14,240 | 424   | 2 ± 2                          | 285   | 1040                  | 1 ± 1   | 1200                    | 480        | 2 ± 1             | 335     |
| s38584  | 20,717  | 34,204 | 413   | 10 ± 3                         | 120   | 1739                  | 6 ± 2   | 800                     | 711        | 9 ± 2             | 320     |
| s38417  | 23,843  | 33,661 | 612   | 3 ± 2                          | 400   | 2404                  | 1 ± 1   | 2550                    | 531        | 2 ± 2             | 340     |
| s9234   | 5,844   | 8,197  | 211   | 2 ± 1                          | 140   | 754                   | 1 ± 1   | 1050                    | 209        | 1 ± 1             | 200     |
| s13207  | 8,651   | 11,831 | 403   | 2 ± 1                          | 225   | 4445                  | 1 ± 1   | 4950                    | 264        | 2 ± 1             | 200     |
| <b>Electronic circuits (digital fractional multipliers)</b> |         |        |  | Three-node feedback loop       |  | Bi-fan                |  | Four-node feedback loop |            |                   |         |
| s208  | 122     | 189    | 10  | 1 ± 1                          | 9   | 4                     | 1 ± 1   | 3.8                     | 5          | 1 ± 1             | 5       |
| s420  | 252     | 399    | 20  | 1 ± 1                          | 18  | 10                    | 1 ± 1   | 10                      | 11         | 1 ± 1             | 11      |
| s838‡   | 512     | 819    | 40  | 1 ± 1                          | 38  | 22                    | 1 ± 1   | 20                      | 23         | 1 ± 1             | 25      |
| <b>World Wide Web</b>                                       |         |        |  | Feedback with two mutual dyads |  | Fully connected triad |  | Uplinked mutual dyad    |            |                   |         |
| nd.edu§   | 325,729 | 1.46e6 | 1.1e5   | 2e3 ± 1e2                      | 800   | 6.8e6                 | 5e4±4e2   | 15,000                  | 1.2e6      | 1e4 ± 2e2         | 5000    |

# Network motifs

Motifs are rarely independent:  
**motif clusters**



# Summary

- Most (**but not all!**) cellular networks are scale-free, ultra-small-world networks organized most probably in a hierarchical manner
- To characterize the large-scale properties of the network, measures like node connectivity, path length, clustering coefficient are used
- Motifs are subgraphs that are significantly overrepresented in real networks. Each class of complex networks has own characteristic motifs

# Discussion

- Network biology still at its infancy
- Topology does not determine the function of cellular networks
  - System dynamics

Thanks!

# Sources

- Barabasi A, Oltvai Z. **Network biology: Understanding the cell's functional organization** (2004) *Nature Genetics* 5 (101-113)
- Watts D, Strogatz S. **Collective dynamics of “small-world” networks** (1998) *Nature* 393 (440-442)
- Milo R, Shen-Orr S, Itzkovitz S, Kashtan N, Chklovskii D, Alon U. **Network motifs: Simple building blocks of complex networks** (2002) *Science* 298 (824-27)
- Ravasz E, Somera A, Mongru D, Oltvai Z, Barabasi A. **Hierarchical organization of modularity in metabolic networks** (2002) *Science* 297 (1551-1555)
- The above publications are sources for the pictures used in this presentation.