

# Large-scale organization of metabolic networks

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CS 466

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# Complex network in cell

- Cellular processes generating mass & energy, transferring information, specifying cell fate
- Integrated through complex network of several constituents and reactions
- Can we look at such networks and learn something about biology and evolution?

# Key findings

- “Metabolic” networks analyzed
- Compared among 43 species from all three domains of life (archaea, bacteria, eukarya)
- Noticed the same topologic scaling properties across all species
- Metabolic organization identical for living organisms, and is “robust”
- May extrapolate to other cellular networks

# Introduction

- Fundamental design principles of cellular networks?
- Example: Dynamic interactions of various constituents impart “robustness” to cellular processes
  - If one reaction did not happen optimally, does not necessarily mess up the whole process

# Introduction

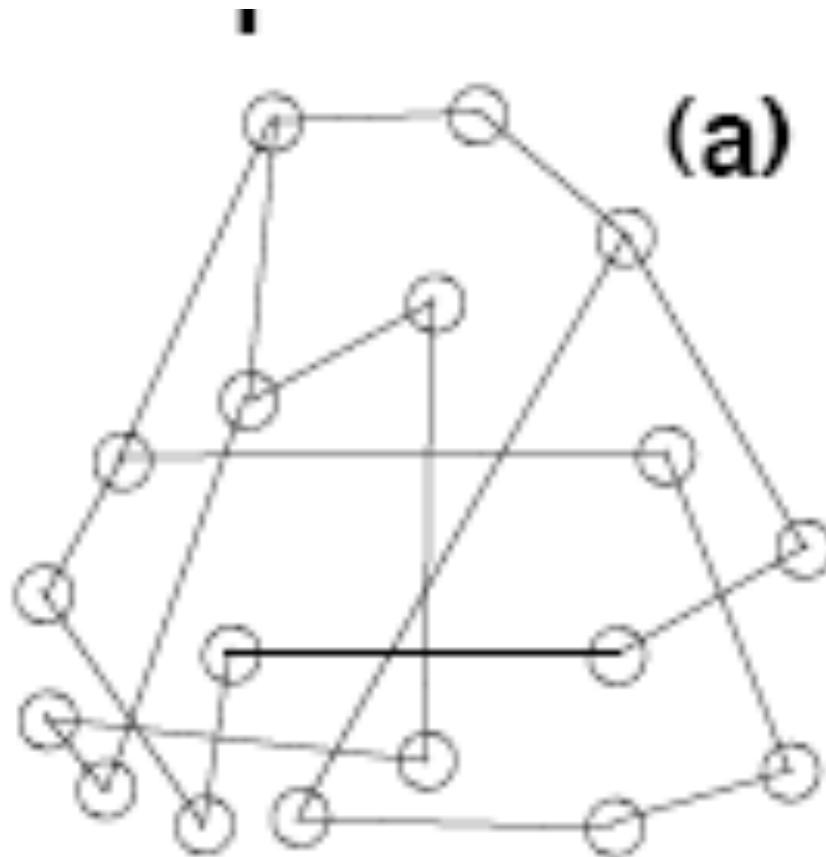
- Constituents of networks: DNA, RNA, proteins, small molecules
- High throughput biology has helped develop databases of networks, e.g., metabolic networks
- Maps are extremely complex
- Fundamental features of network topology?

# Network models

# Erdos-Renyi random graph

- Start with a fixed number of nodes, no edges
- Each pair of nodes connected by an edge with probability  $p$
- This leads to a “statistically homogeneous” graph or network
- Most nodes have same degree

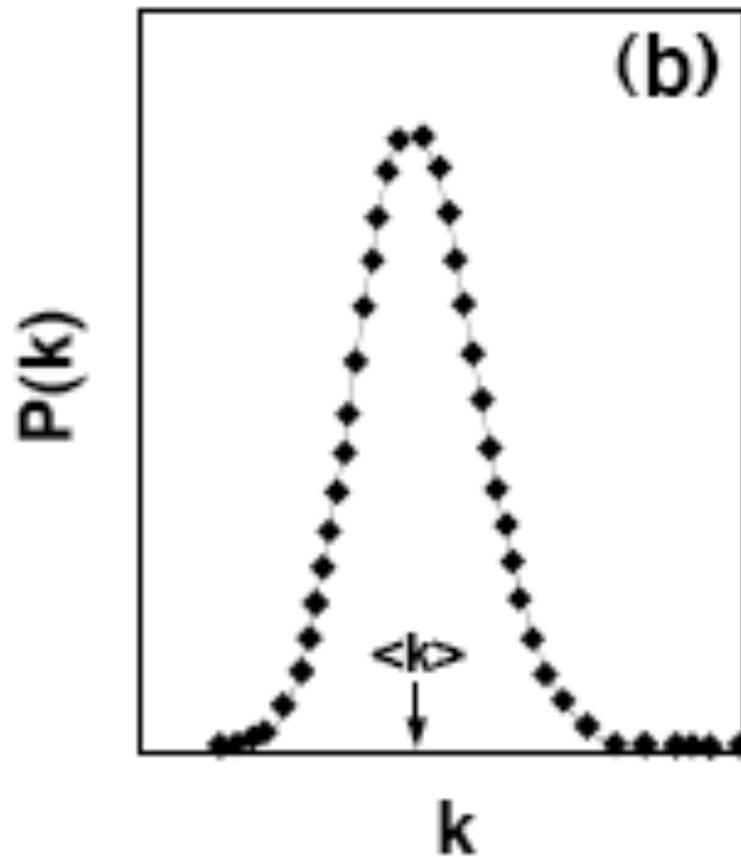
<http://arxiv.org/pdf/cond-mat/0010278>



# Erdos-Renyi random graph

- Degree distribution is Poisson with strong peak at mean  $\langle k \rangle$
- Therefore, probability of finding a highly connected node decays exponentially

<http://arxiv.org/pdf/cond-mat/0010278>



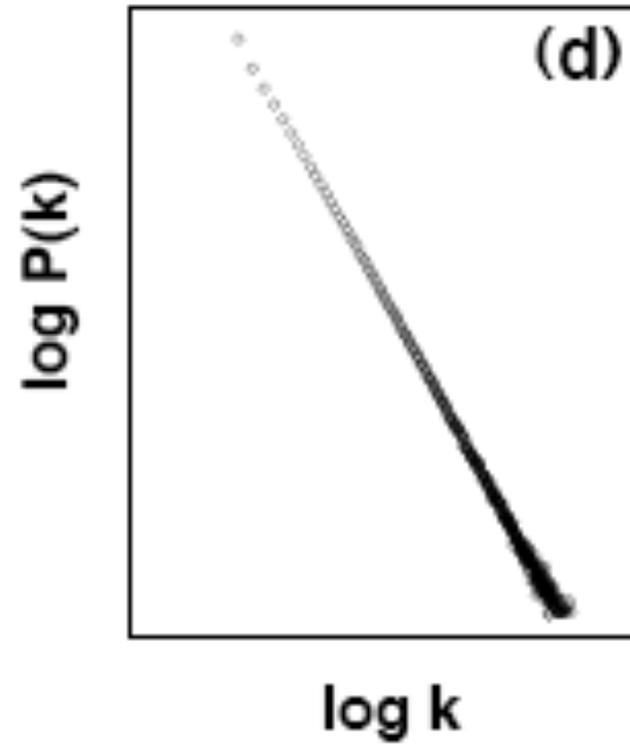
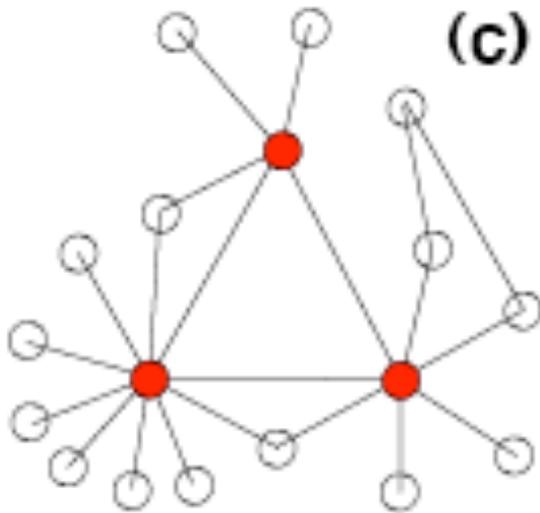
# Empirical graphs

- World-wide web, internet, social networks have been studied
- Serious deviations from random structure of E-R model
- Better described by “scale-free” networks

# Scale-free networks

- Degree distribution  $P(k)$  follows power law distribution
  - $P(k) \sim k^{-x}$
- Scale-free networks are extremely heterogeneous:
  - a few highly connected nodes (hubs)
  - rest of the (less connected) nodes connect to hubs
- Generated by a process where new nodes are preferentially attached to already high-degree nodes

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# What does this tell us?

- Difference between Erdos-Renyi and scale-free graphs arise from simple principles of how the graphs were created
- Therefore, understanding topological properties can tell us how the cellular networks were created

Data

# Metabolic networks

- Core metabolic network of 43 different organisms (WIT database)
- 6 archaea, 32 bacteria, 5 eukarya
- Nodes = substrates, edges = metabolic reactions, additional nodes = enzymes
- Based on firmly established data from biochemical literature
- Sufficient data for statistical analysis



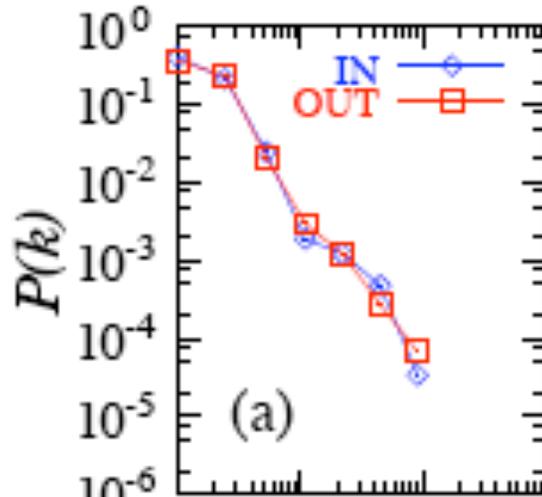
# Results

# Topology

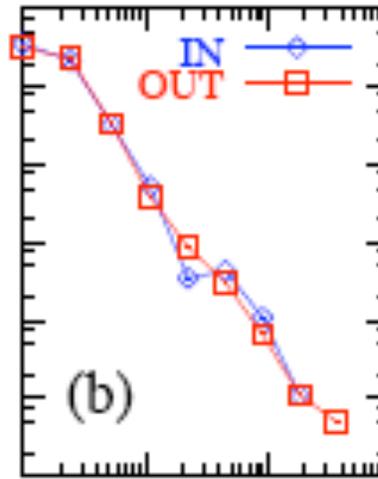
- Is the topology described by E-R model or by scale-free model?
- Observed that degree distribution follows a power law. Therefore, scale-free networks

<http://arxiv.org/pdf/cond-mat/0010278>

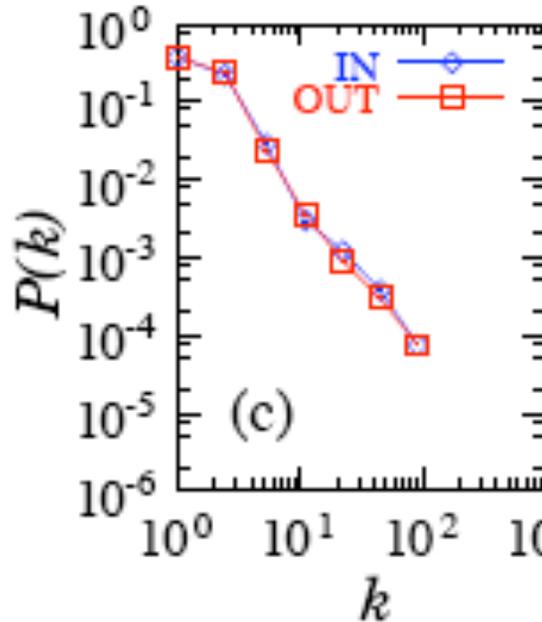
A. fulgidus



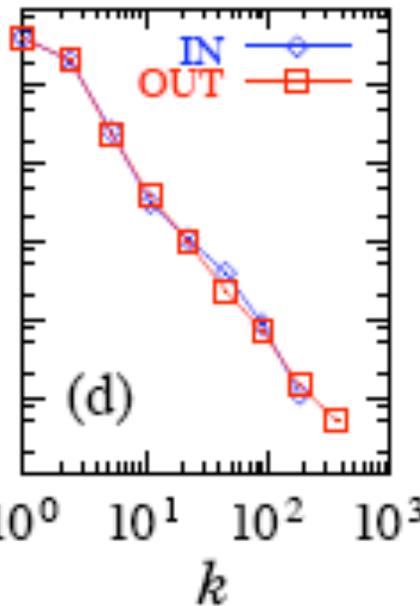
E. coli



C. elegans



Average



# Small-world property

- General feature of many complex networks: any two nodes can be connected by relatively short paths
- In metabolic network, a path is the biological “pathway” connecting two substrates
- Characterized by “network diameter”
  - Shortest path, over all pairs of nodes

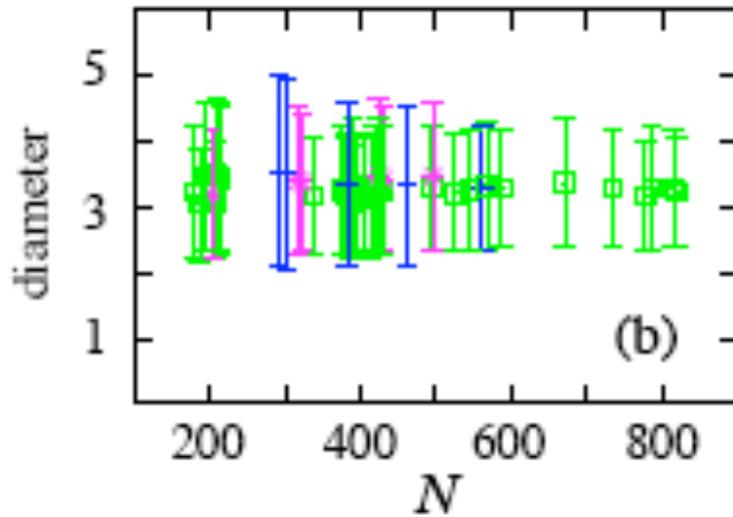
# Small-world property

- For non-biological networks, the average degree is usually fixed
- This implies that network diameter increases logarithmically with new nodes being added
- Is this true of metabolic networks ?
- That is, more complex bacterium (more substrates and enzymes) will have larger diameter ?

# Small-world property

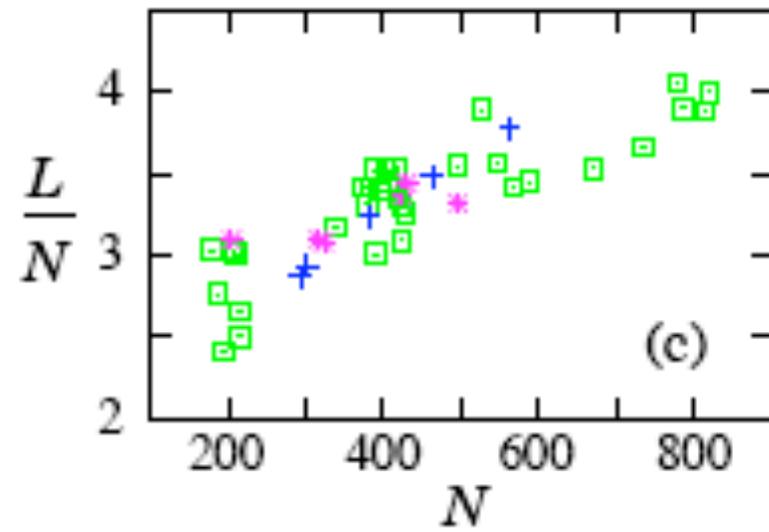
- More complex bacterium (more substrates and enzymes) will have larger diameter ?
- Observed: diameter same across all 43 species !
- A possible explanation: average degree must be higher for more complex organisms
- This is also verified.

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network diameter for different organisms

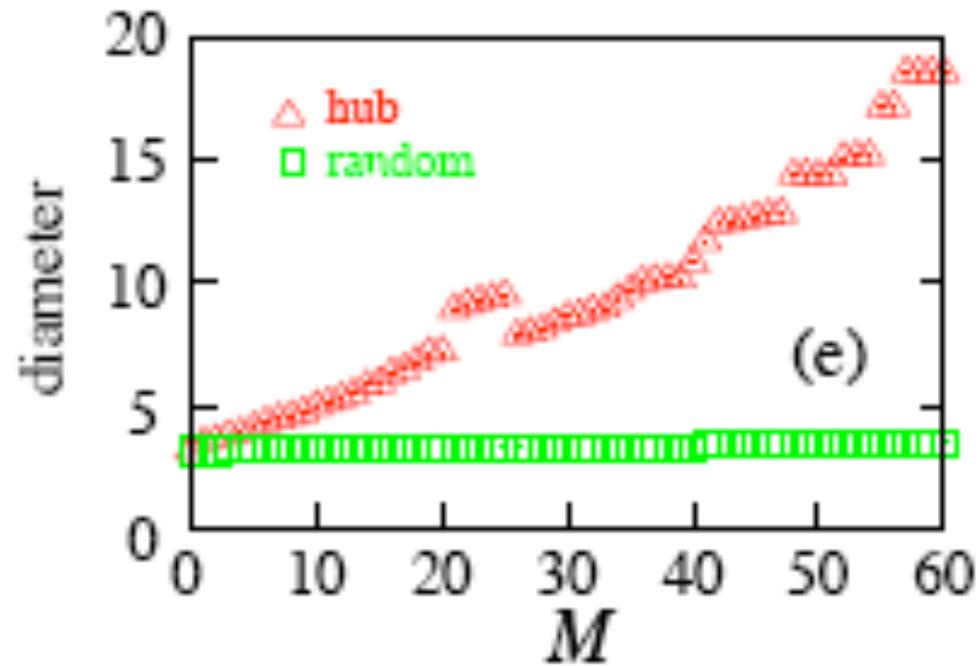
average degree over different organisms



# Hubs in network

- Power-law connectivity implies that a few “hub” nodes dominate the overall connectivity
- Sequential removal of hubs => diameter rises sharply
- Observed: metabolic networks show this phenomenon too

<http://arxiv.org/pdf/cond-mat/0010278>



Diameter after removing  $M$  substrates

# Hubs in network

- At the same time, scale-free networks are robust to random errors
- In metabolic network, removal of randomly chosen substrates did not affect average distance between remaining nodes
- Fault tolerance to removal of metabolic enzymes also demonstrated through biological experiments

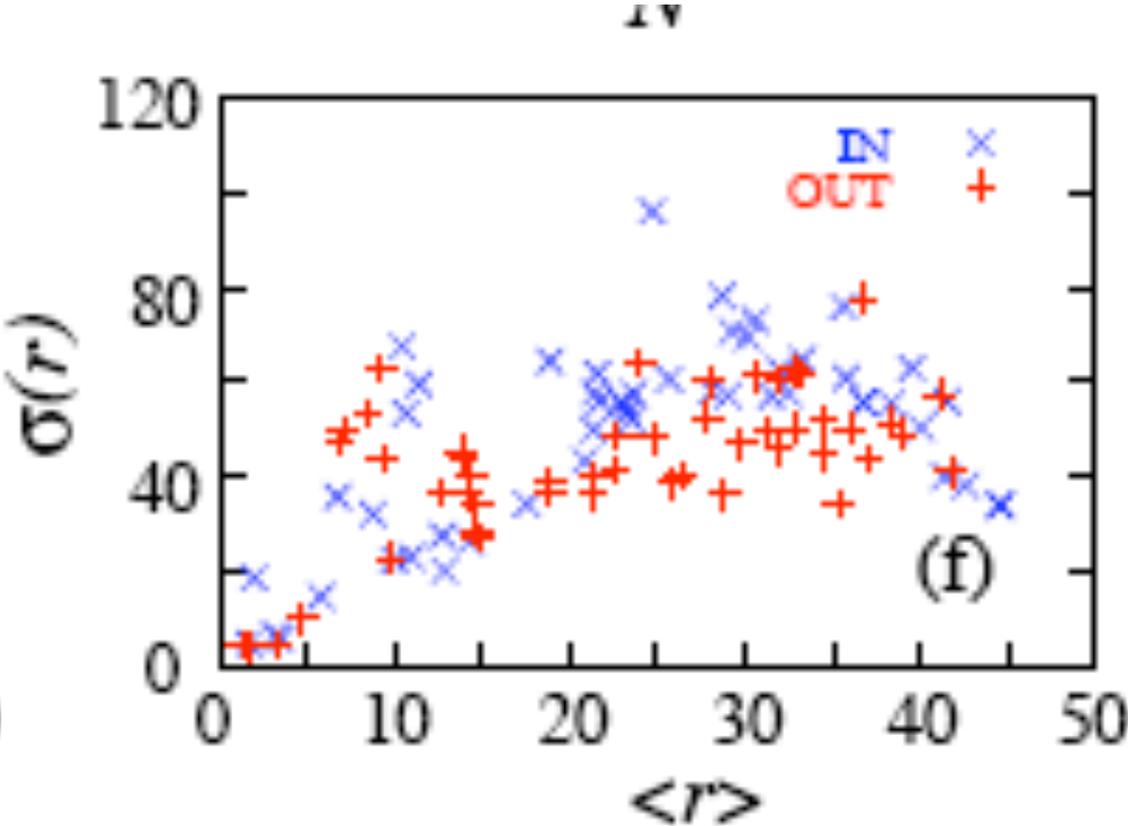
# Hubs across networks

- Do the same substrates act as hubs in all organisms?
- Rank all substrates by their degrees
- Ranking of the top substrates is practically same across all species
- For every substrate present in all species, compute rank “ $r$ ” (in terms of degree) in each species

# Hubs across networks

- Compute mean  $\langle r \rangle$  and standard deviation  $\sigma_r$  of rank of each substrate
- Observed:  $\sigma_r$  increases with  $\langle r \rangle$
- The top-ranking nodes (hubs) have relatively little variance across species

<http://arxiv.org/pdf/cond-mat/0010278>



# Summary

- Other biological networks also hypothesized to be scale-free.
- Evolutionary selection of a robust and error tolerant architecture