

Theory and Applications of Complex Networks: Advances and Challenges

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Outline

- Introduction
- Recent Theoretical Advances
- Recent Applications
- Conclusions

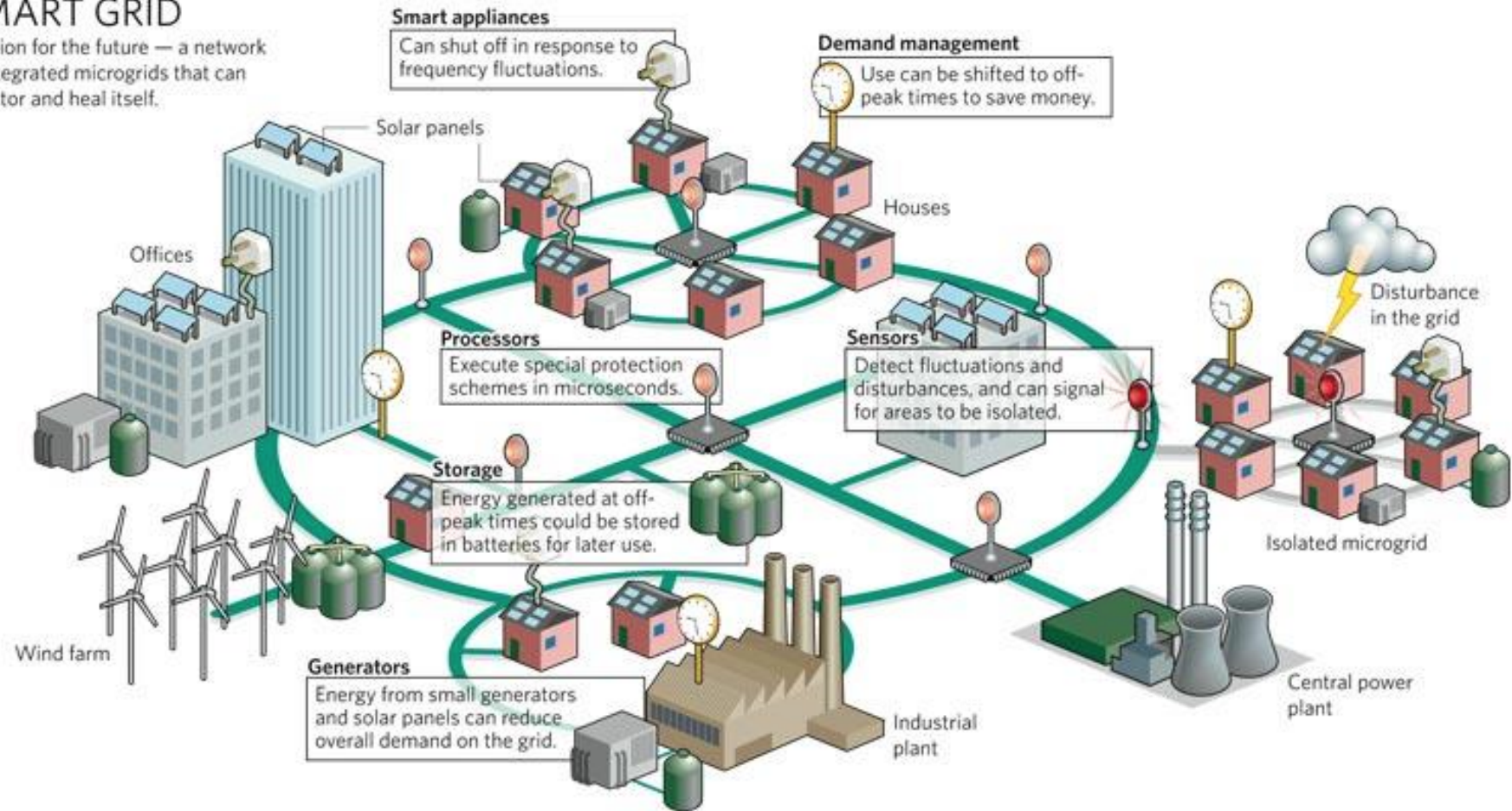
Complex networks are everywhere!



Smart Grids

SMART GRID

A vision for the future — a network of integrated microgrids that can monitor and heal itself.



Transportation Networks



Social Networks



Financial Networks



Network Graphs

➤ Random Graphs

- Nodes and edges are generated via a random process
- Erdős and Rényi model (1960)

➤ Small World Graphs

- Nodes and edges are generated so that most of the nodes are connected through a small number of nodes in between
- Watts and Strogatz model (1998)

➤ Scale-Free Graphs

- Graphs whose node degree distributions follow a power-law
- Rich gets richer
- Barabási and Albert model (1999)

Recent Theoretical Advances

For example:

- **Controllability** of complex networks
- **Observability** of complex networks
- **Pinning Control** of complex networks

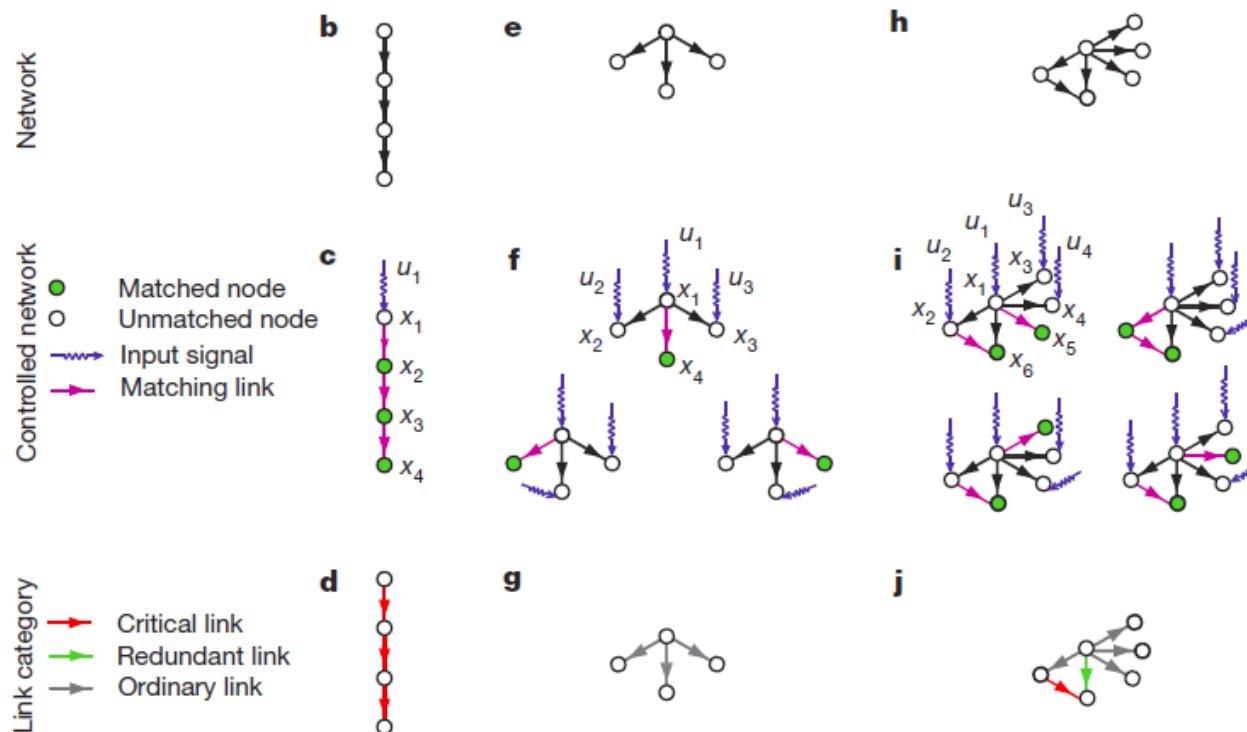
Controllability of Complex Networks

ARTICLE

doi:10.1038/nature10011

Controllability of complex networks

Yang-Yu Liu^{1,2}, Jean-Jacques Slotine^{3,4} & Albert-László Barabási^{1,2,5}



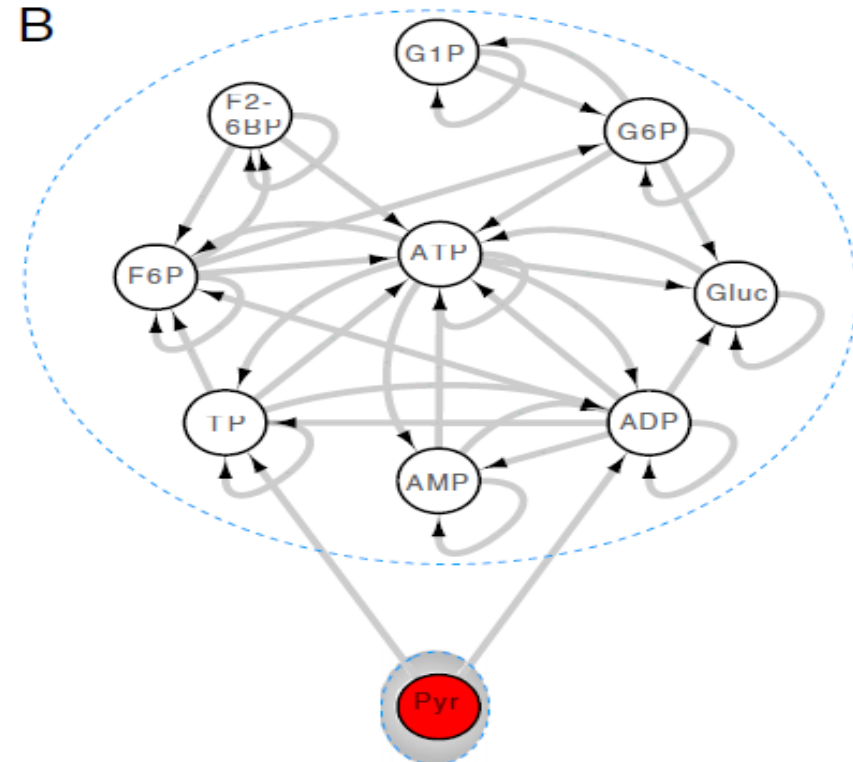
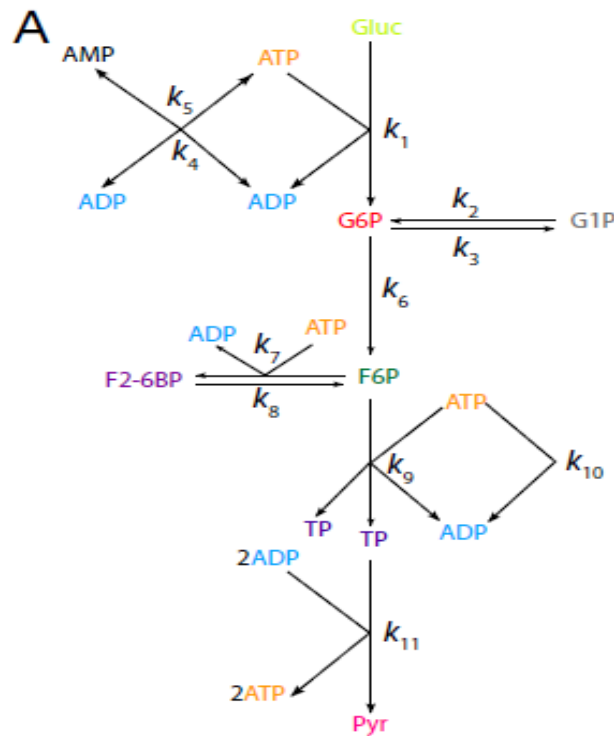
Observability of Complex Networks

PNAS

Observability of complex systems

Yang-Yu Liu^{a,b,c,d,e}, Jean-Jacques Slotine^{f,g,h}, and Albert-László Barabási^{a,b,c,d,e,i,1}

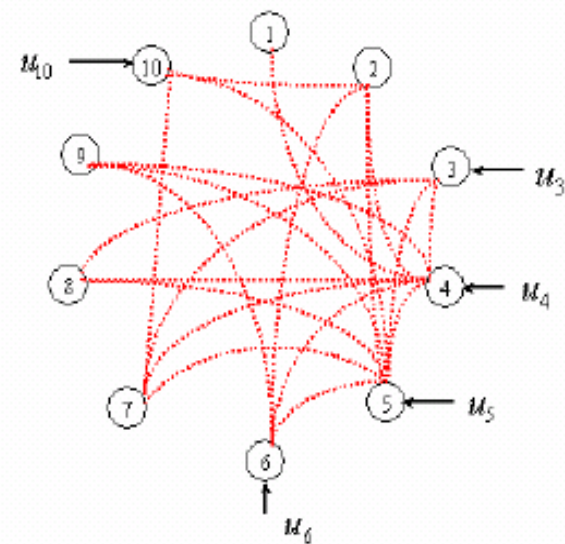
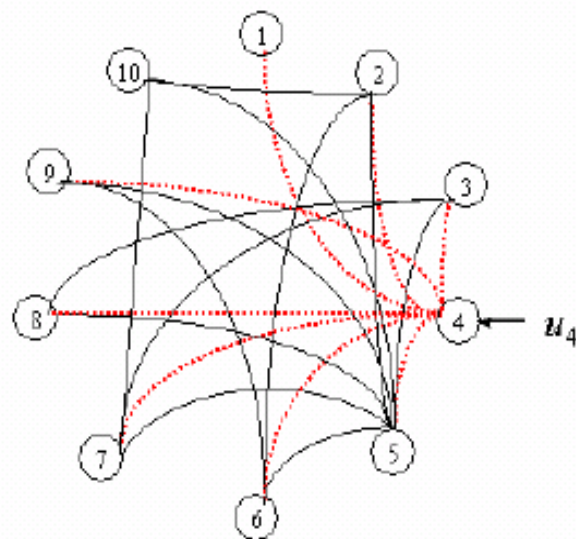
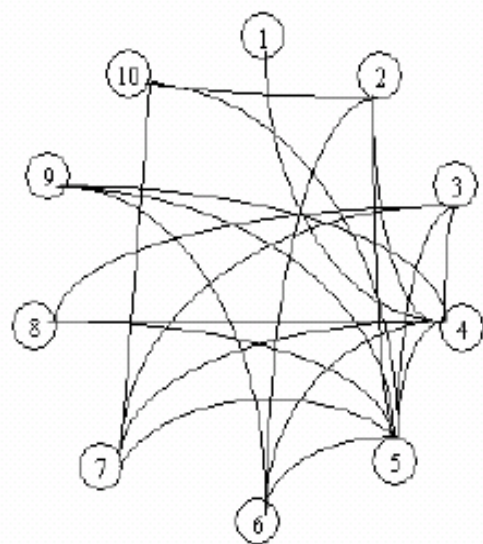
^aCenter for Complex Network Research and Departments of ^bPhysics, ^cComputer Science, and ^dBiology, Northeastern University, Boston, MA 02115; ^eCenter for Cancer Systems Biology, Dana-Farber Cancer Institute, Boston, MA 02115; ^fNonlinear Systems Laboratory, Massachusetts Institute of Technology, Cambridge, MA 02139; Departments of ^gMechanical Engineering and ^hBrain and Cognitive Sciences, Massachusetts Institute of Technology, Cambridge, MA 02139; and ⁱDepartment of Medicine, Brigham and Women's Hospital, Harvard Medical School, Boston, MA 02115



Pinning Control of Complex Networks

- It is practically **impossible** to **control every node** in a large complex network
- Is it **possible** to control **a small fraction of nodes** (e.g., 5%) to achieve the same effect?
- **Pinning control**

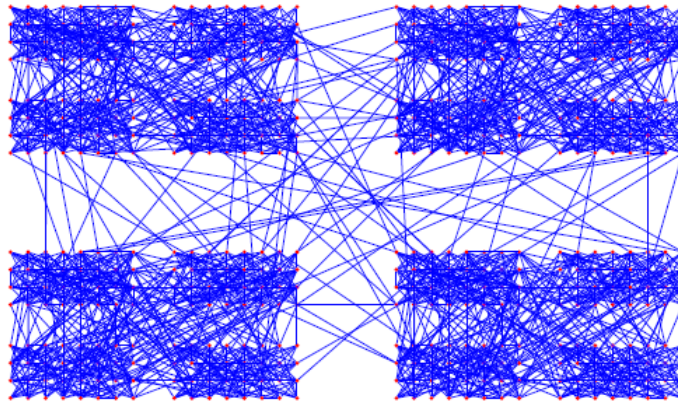
Pinning Control: A Simple Example



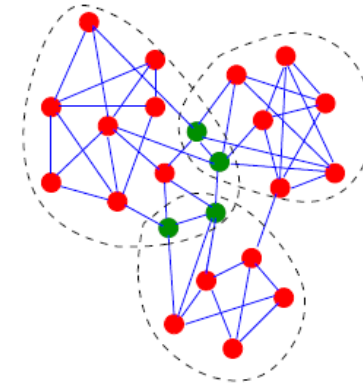
A network with 10-nodes
generated by the B-A scale-free model ($N=10$, $m=m_0=3$)

X. F. Wang, G. Chen, Physica A (2002)
X. Li, X. F. Wang, G. Chen, IEEE T-CAS (2003)

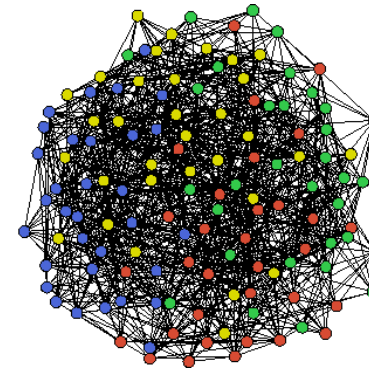
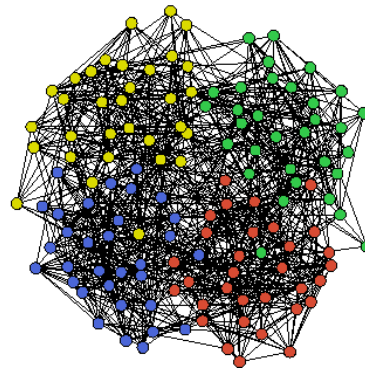
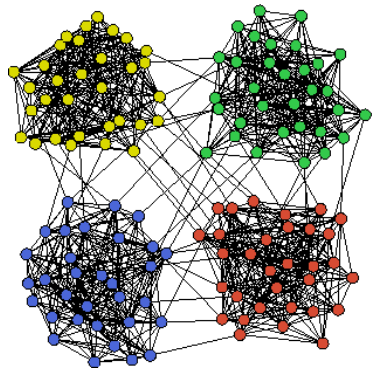
Detecting Community Structures: Challenges



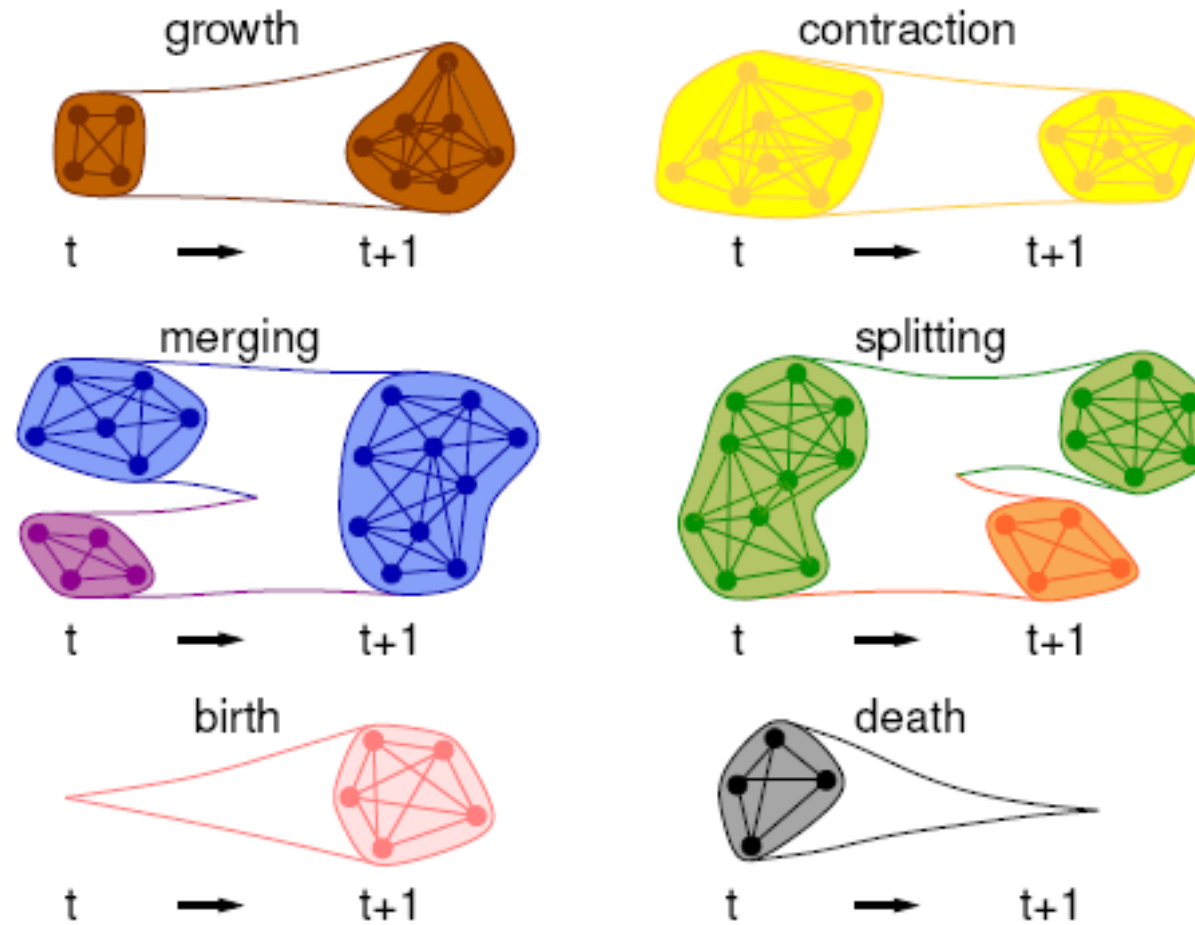
Hierarchical



Overlapping



Detecting Community Structures: Challenges



Two Fundamental Challenging Problems

- How many and which nodes should a network with fixed structure and coupling strength be pinned in order to reach network synchronization?
- How large the coupling strength should be applied to a network with fixed structure and pinning nodes in order to reach network synchronization?

- J. Zhou, J. Lu, J. Lü, Automatica, 44(4): 996–1003, 2008.
- H. Liu, J. Lu, J. Lü, D. Hill, Automatica, 45(8): 1799–1807, 2009.
- W. Yu, G. Chen, J. Lü, Automatica, 45(2): 429–435, 2009.

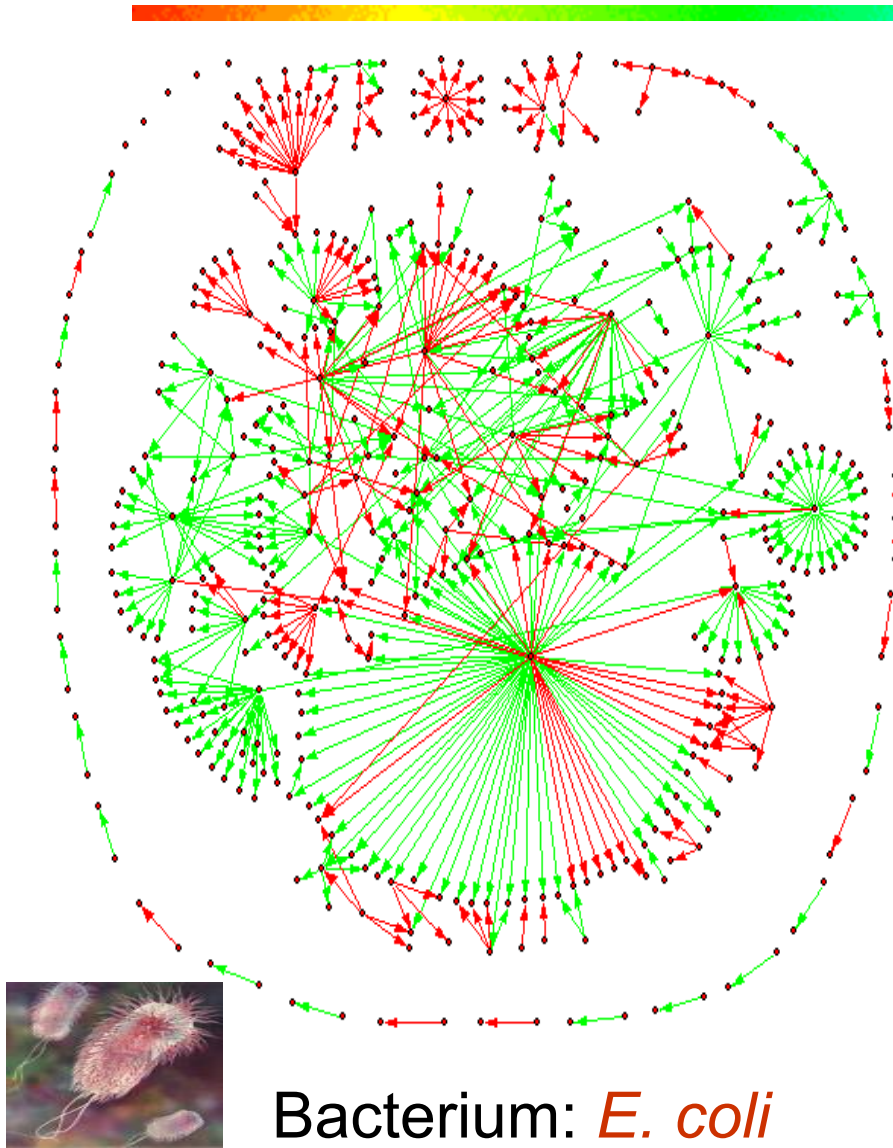
Application Examples



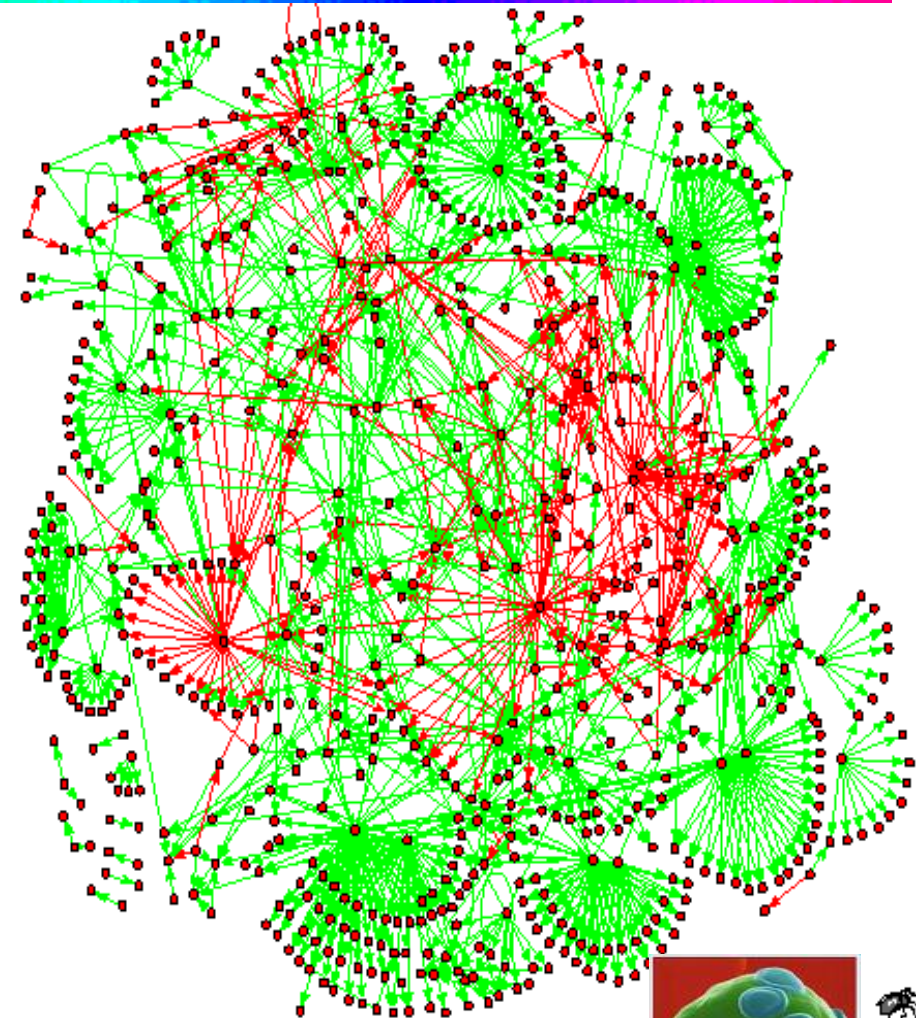
Numerous real-world applications:

- Applications in **Systems Biology**
- Applications in **Engineering Networks**
- P. Wang, J. Lü, and M. J. Ogorzalek, “Global relative parameter sensitivities of the feed-forward loops in genetic networks,” *Neurocomputing*, vol. 78, no. 1, pp. 155-165, Feb. 2012.

Genetic Regulatory Networks

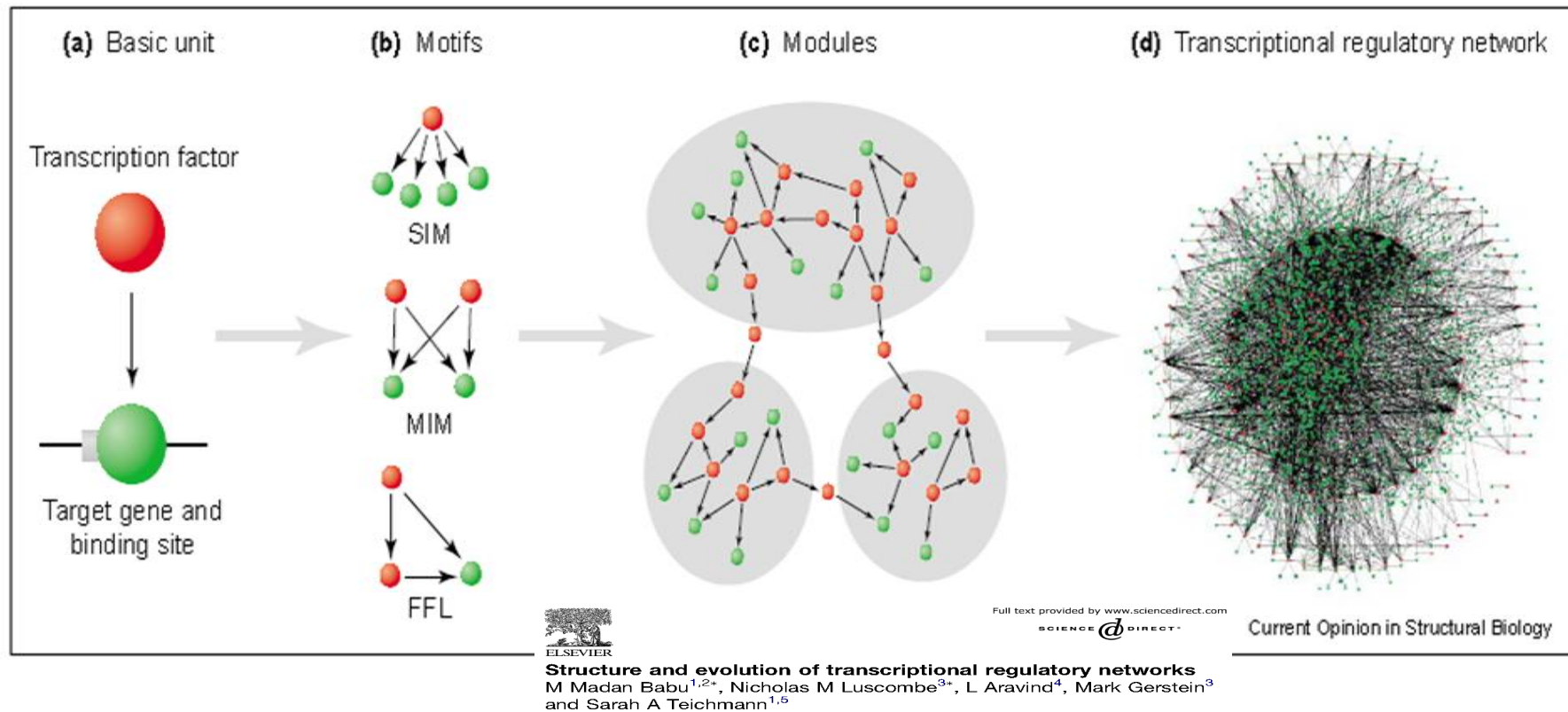


Bacterium: *E. coli*



Single-celled eukaryote:
S. cerevisiae

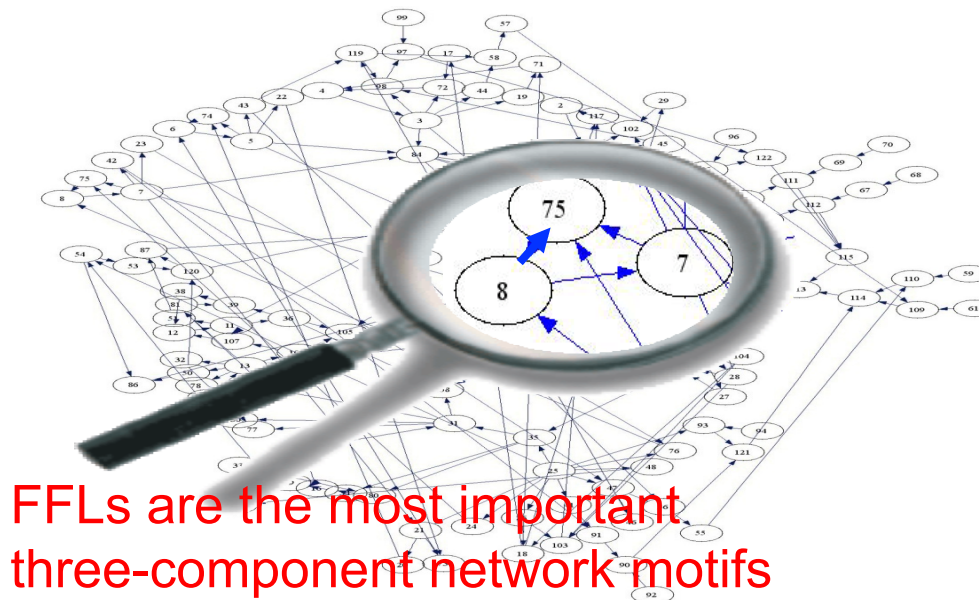
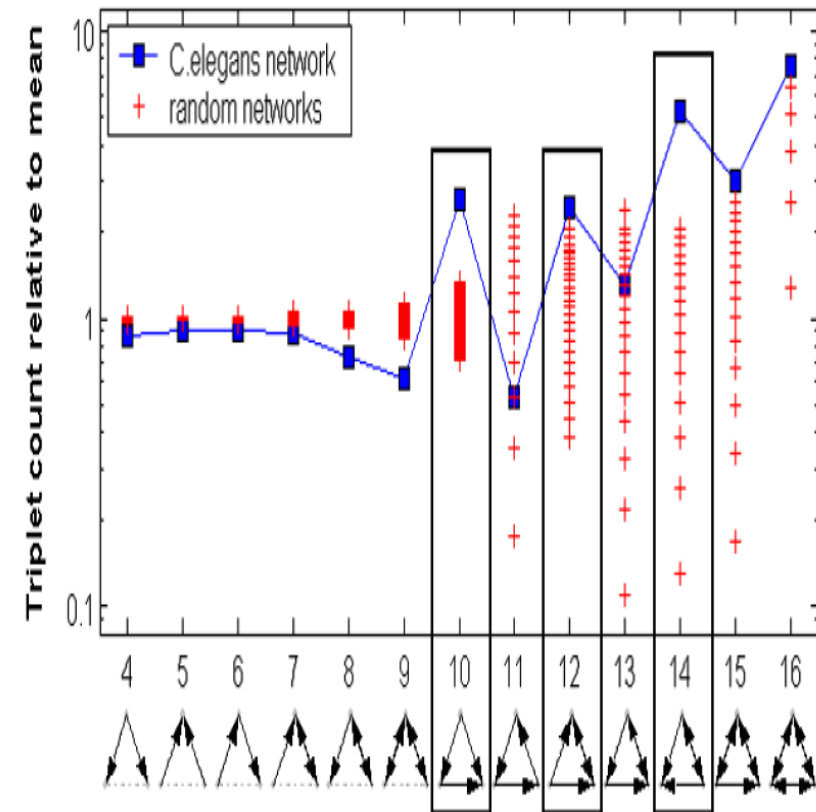
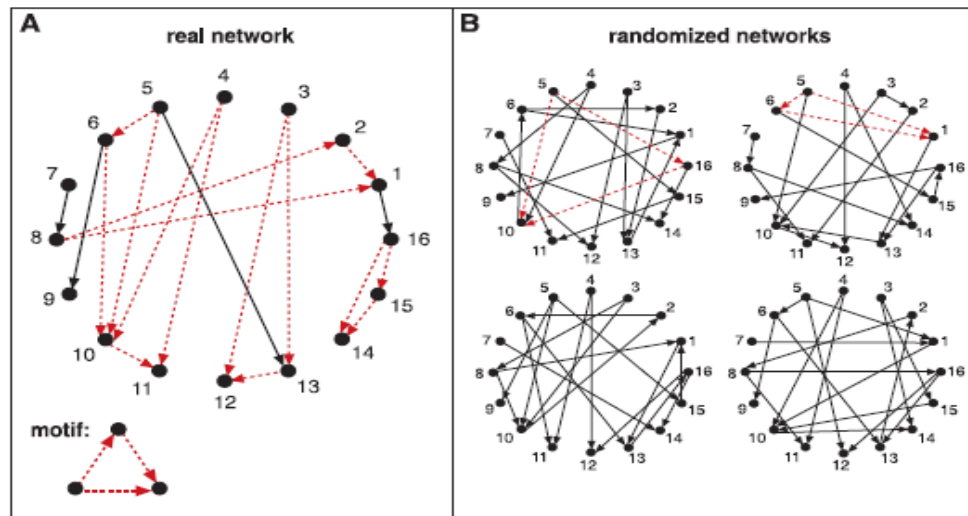
Real-World GRNs Are Too Complex



- Have **hindered** the understanding of real-world GRNs
- Complex GRNs have many **simple building blocks**
- Investigation of simple networks is **a first step to understand real complex GRNs**

Network Motif

Network motif: frequent patterns in real networks



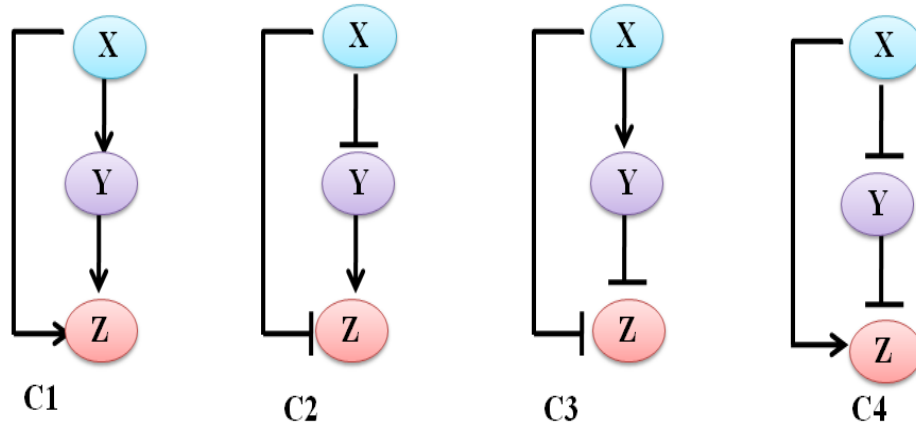
FFLs are the most important three-component network motifs

Science, 298 (2002) 824-827

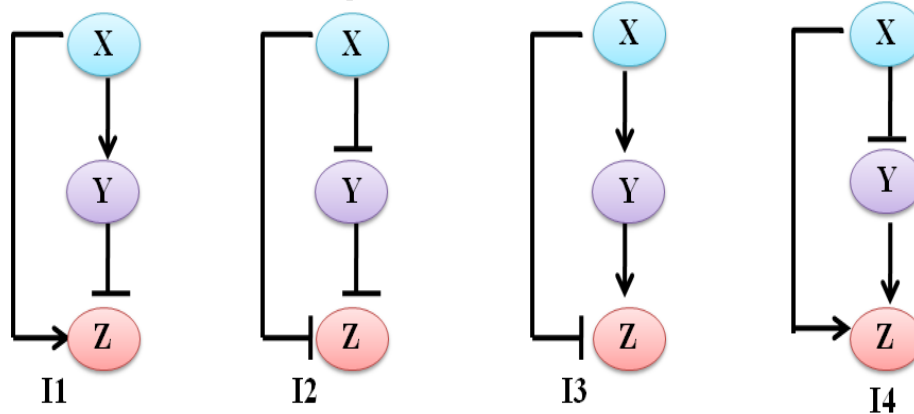
Nat. Genet., 31 (2002) 64-68

Eight Configurations of FFLs

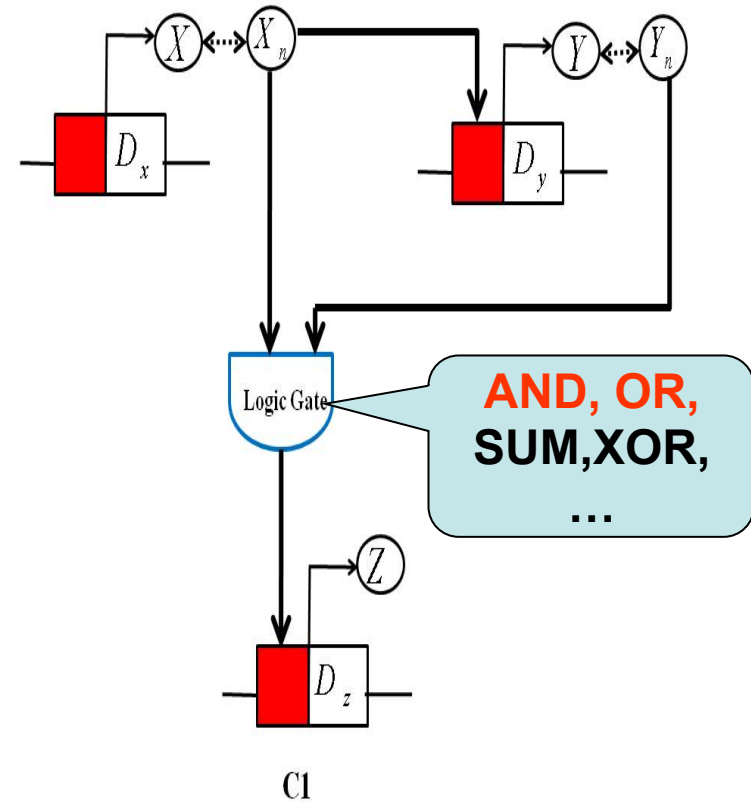
Coherent Feed-forward Loops (CFFLs)



Incoherent Feed-forward Loops (IFFLs)

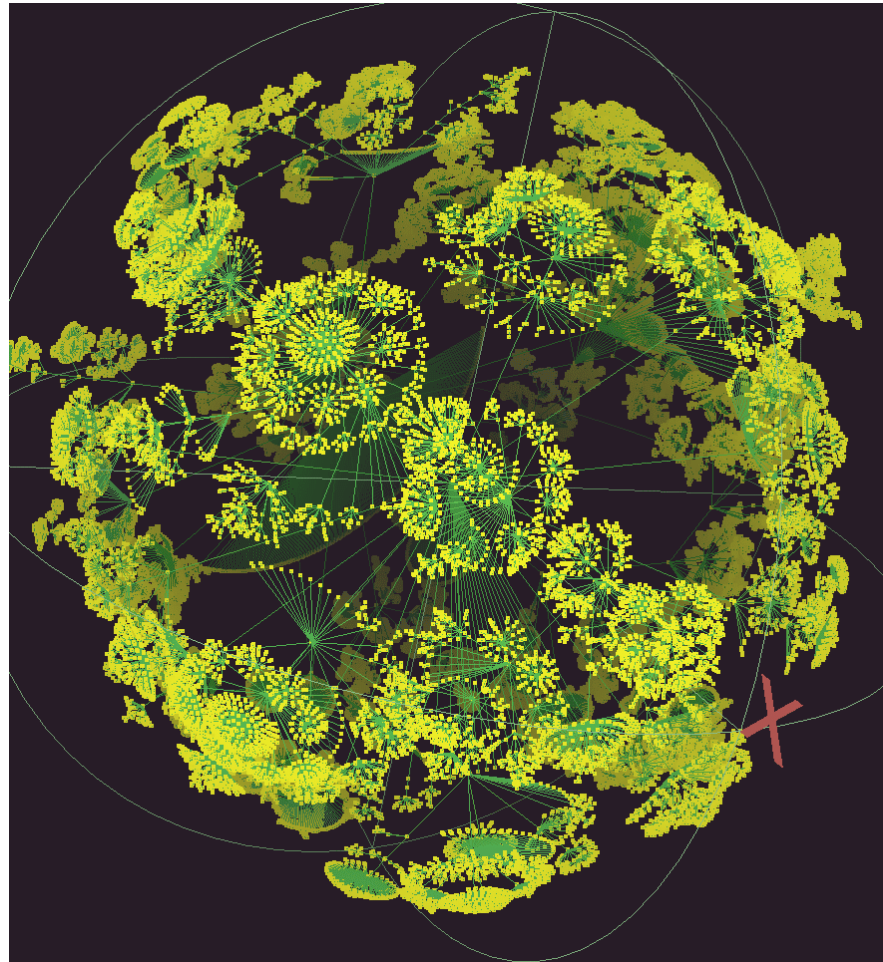


8 configurations



C1 detailed regulation

Applications in Engineering Networks



Internet Graphs

- Internet is a network of Autonomous Systems (AS):
 - groups of networks sharing the same routing policy
 - identified with Autonomous System Numbers
- Autonomous System Numbers:
<http://www.iana.org/assignments/as-numbers>
- Internet topology on **AS-level**:
 - the arrangement of ASes and their interconnections
- Analyzing the Internet topology and finding properties of associated graphs rely on mining data and capturing information about Autonomous Systems (ASes)

Internet AS-Level Data

Source of data are routing tables:

- **Route Views:** <http://www.routeviews.org>
 - most participating ASes reside in North America
- **RIPE (Réseaux IP européens):** <http://www.ripe.net/ris>
 - most participating ASes reside in Europe

Analyzed Datasets

➤ Sample datasets:

■ Route Views:

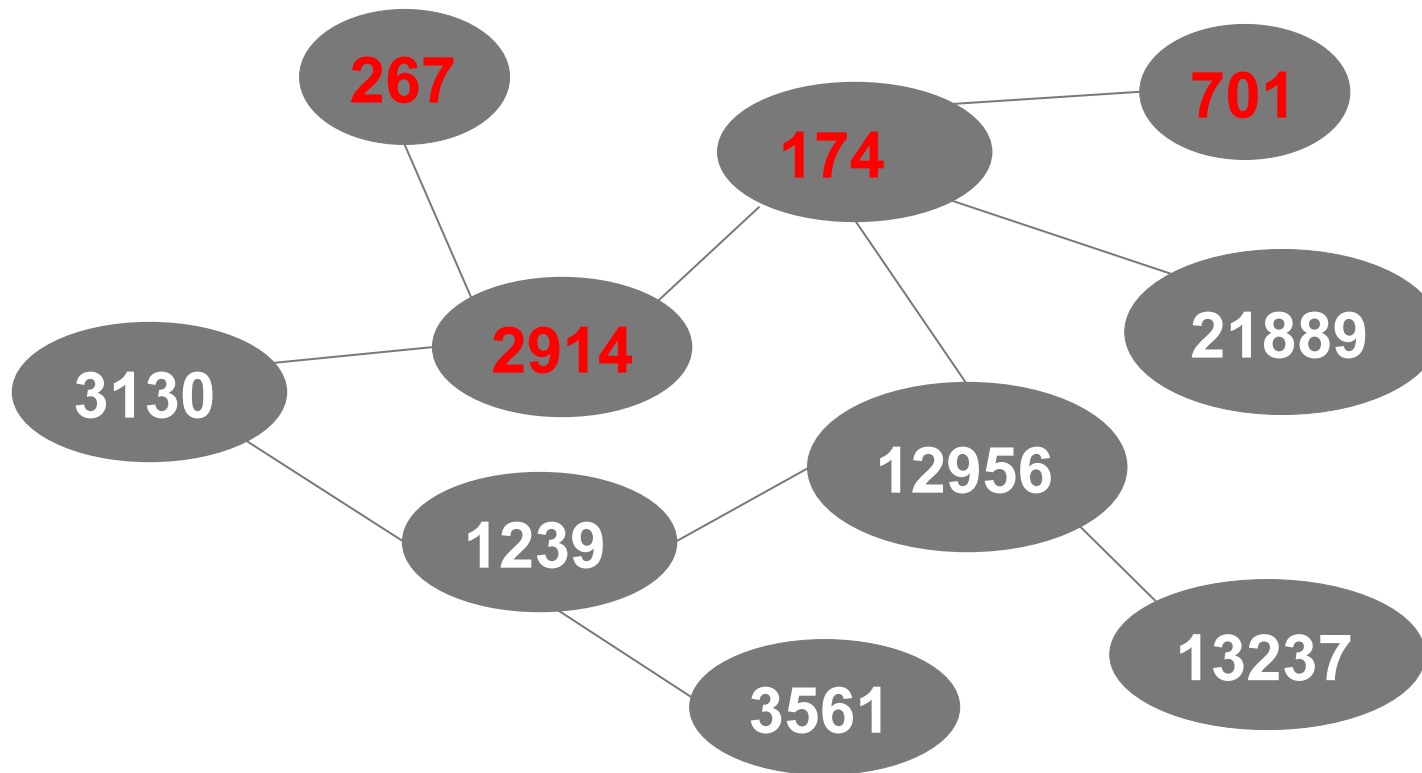
```
TABLE_DUMP| 1050122432| B| 204.42.253.253|  
267| 3.0.0.0/8| 267 2914 174 701| IGP|  
204.42.253.253| 0| 0| 267:2914 2914:420  
2914:2000 2914:3000| NAG| |
```

■ RIPE:

```
TABLE_DUMP| 1041811200| B| 212.20.151.234|  
13129| 3.0.0.0/8| 13129 6461 7018 | IGP|  
212.20.151.234| 0| 0| 6461:5997 13129:3010|  
NAG| |
```

Internet Topology at AS Level

- Datasets collected from Border Gateway Protocols (BGP) routing tables are used to infer the Internet topology at AS-level



Internet Topology



- The Internet topology is characterized by the presence of various power-laws observed when considering:
 - node degree vs. node rank
 - node degree frequency vs. degree
 - number of nodes within a number of hops vs. number of hops
 - eigenvalues of the adjacency matrix and the normalized Laplacian matrix vs. the order of the eigenvalues.

- M. Faloutsos, P. Faloutsos, and C. Faloutsos, 1999
- G. Siganos, M. Faloutsos, P. Faloutsos, and C. Faloutsos, 2003

Internet Matrices

- Adjacency matrix $A(G)$:

$$A(i, j) = \begin{cases} 1 & \text{if } i \text{ and } j \text{ are adjacent} \\ 0 & \text{otherwise,} \end{cases}$$

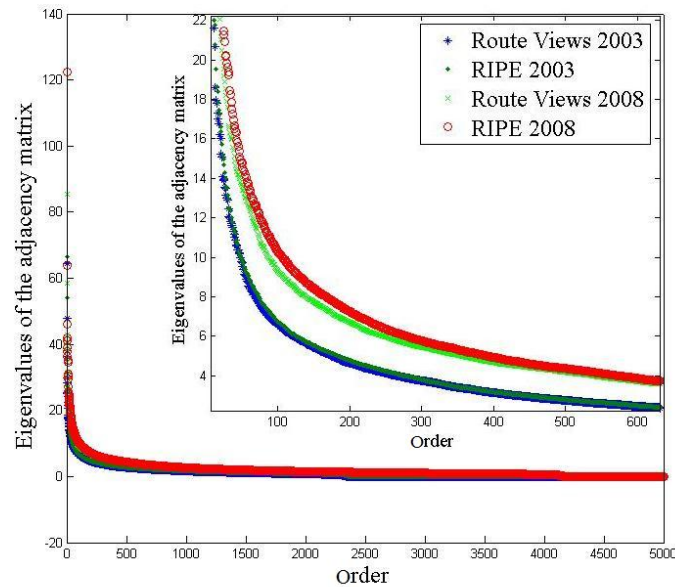
where i and j are the graph nodes

- Normalized Laplacian matrix $NL(G)$:

$$NL(i, j) = \begin{cases} 1 & \text{if } i = j \text{ and } d_i \neq 0 \\ -\frac{1}{\sqrt{d_i d_j}} & \text{if } i \text{ and } j \text{ are adjacent} \\ 0 & \text{otherwise,} \end{cases}$$

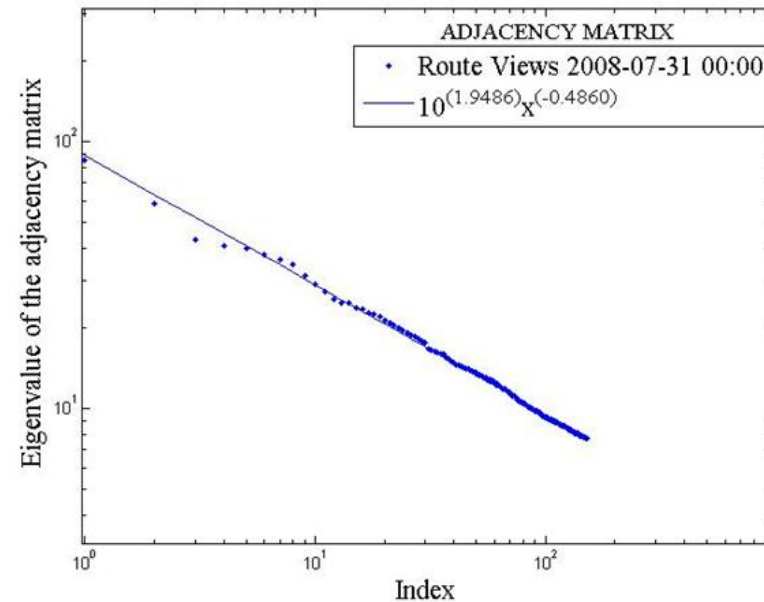
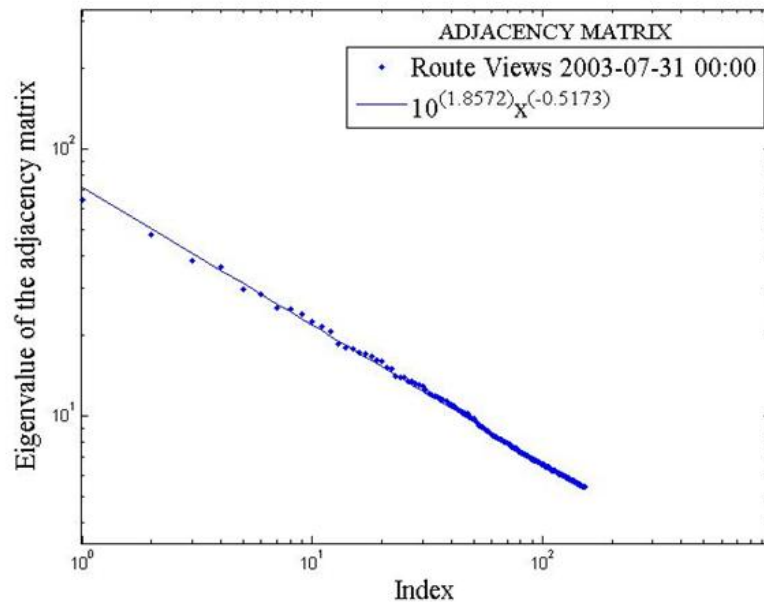
where d_i and d_j are degrees of node i and j , respectively

Eigenvalues of the Adjacency Matrix



| order | Route Views 2003 | Route Views 2008 | RIPE 2003 | RIPE 2008 |
|-------|------------------|------------------|-----------|-----------|
| 1 | 64.30 | 85.43 | 66.65 | 122.28 |
| 2 | 47.75 | 58.56 | 54.19 | 63.94 |
| 3 | 38.15 | 42.77 | 38.24 | 46.14 |
| 4 | 36.23 | 40.85 | 36.14 | 41.98 |
| 5 | 29.88 | 39.69 | 31.21 | 41.08 |
| 6 | 28.50 | 37.85 | 27.38 | 38.93 |
| 7 | 25.47 | 36.21 | 26.41 | 37.94 |
| 8 | 25.06 | 34.66 | 25.06 | 36.47 |
| 9 | 24.13 | 31.58 | 23.86 | 35.08 |
| 10 | 22.51 | 29.34 | 23.32 | 34.47 |
| 11 | 21.61 | 27.40 | 22.02 | 30.97 |
| 12 | 20.69 | 25.69 | 21.77 | 30.54 |
| 13 | 18.58 | 25.00 | 20.75 | 29.68 |
| 14 | 17.94 | 24.82 | 19.55 | 27.03 |
| 15 | 17.78 | 23.89 | 18.67 | 25.74 |
| 16 | 17.31 | 23.69 | 18.42 | 25.35 |
| 17 | 16.99 | 22.81 | 17.85 | 24.83 |
| 18 | 16.75 | 22.46 | 17.44 | 24.30 |
| 19 | 16.22 | 22.04 | 17.24 | 24.06 |
| 20 | 16.01 | 21.36 | 16.63 | 24.00 |

Power Laws: Eigenvalues vs. Index

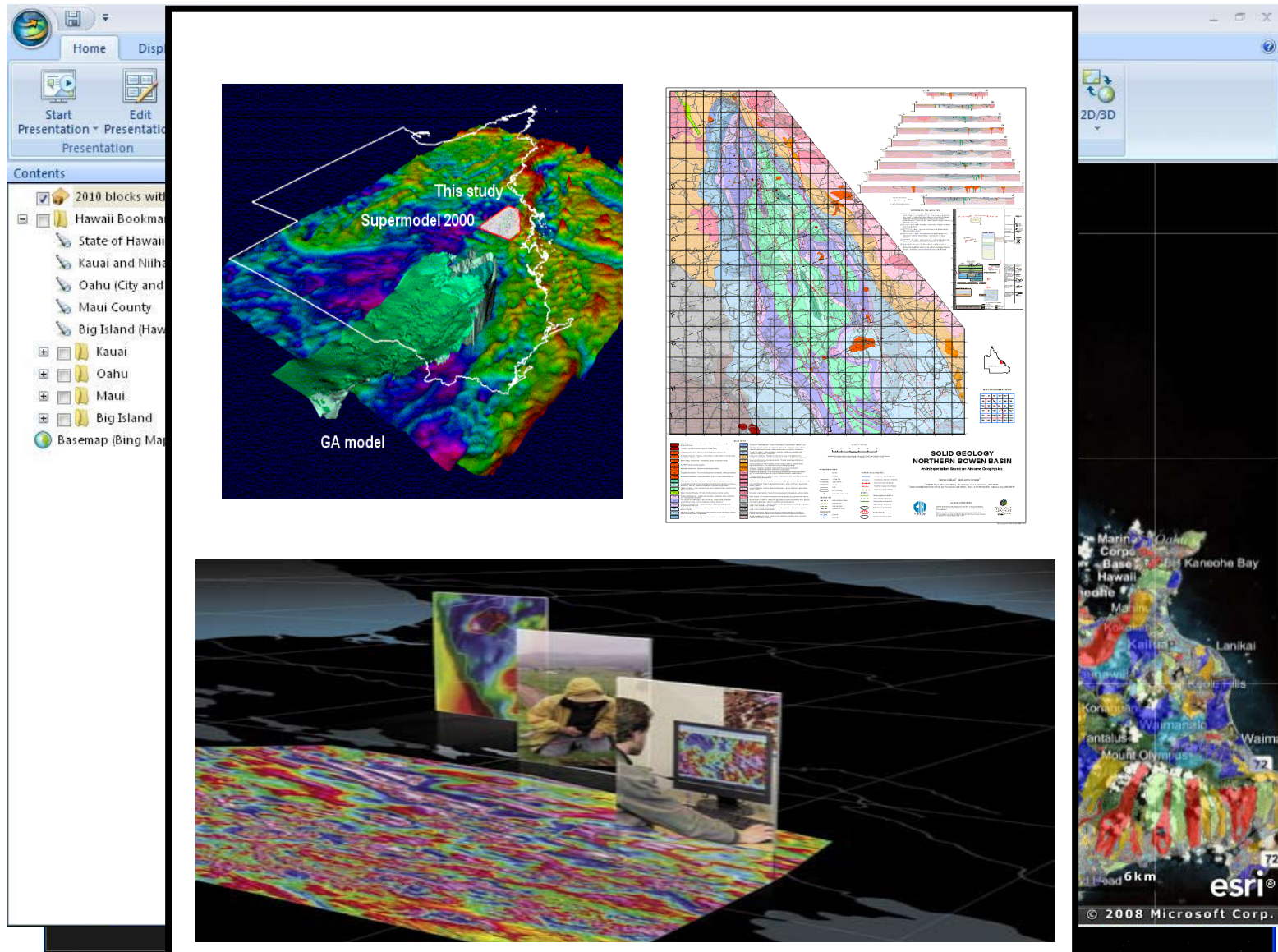


Adjacency matrix:

- Route Views 2003 datasets: $\epsilon = -0.5713$ and $r = -0.9990$
- Route Views 2008 datasets: $\epsilon = -0.4860$ and $r = -0.9982$

ϵ = power-law exponent; r = correlation coefficient

Typical Example



Conclusions: Opportunities and Challenges



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and Information
computer Science

Network Science and Engineering (NetSE) Research Agenda



A Report of the Network Science and Engineering Council
Release Version 1.1

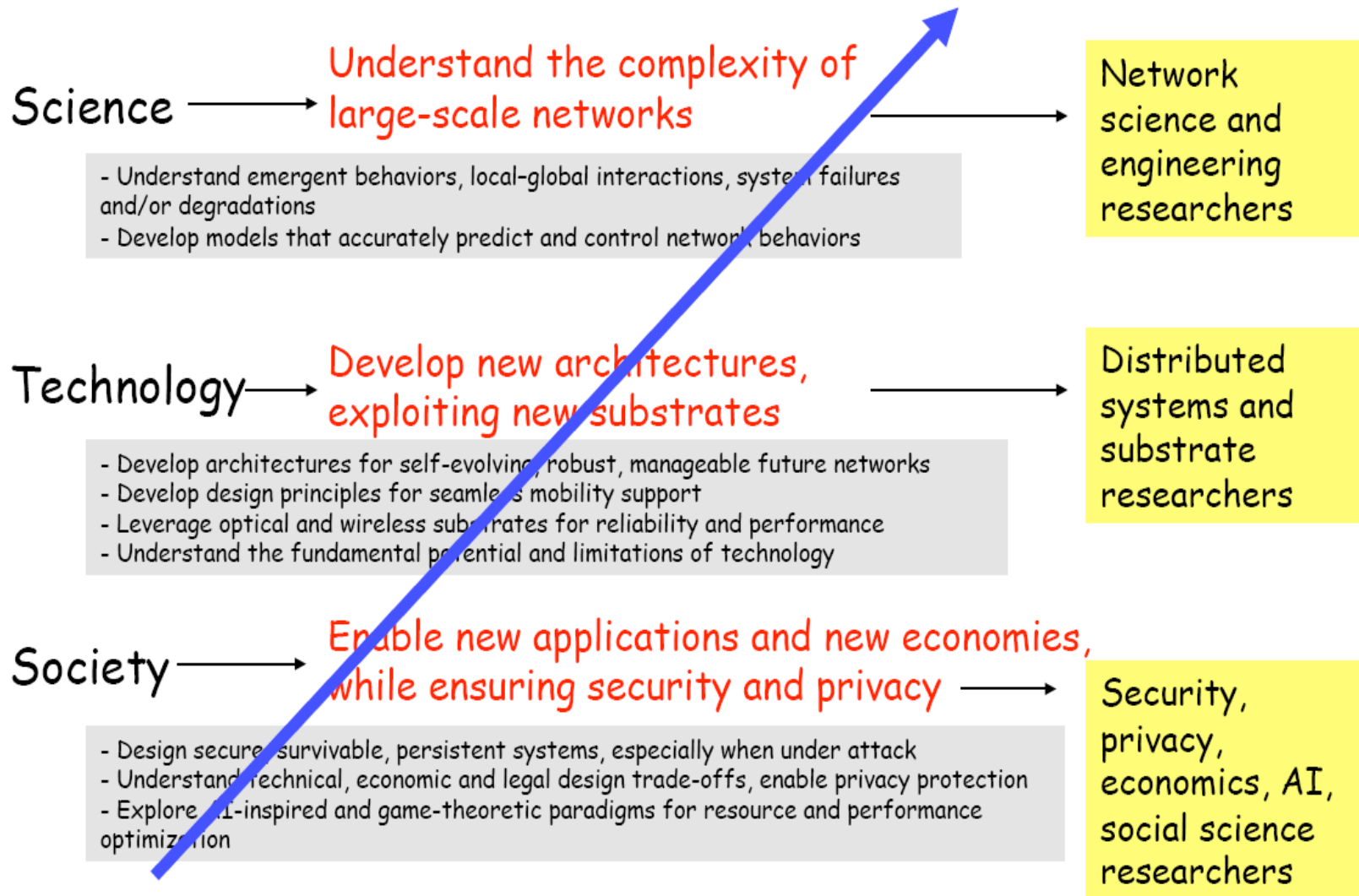
NetSE Research Agenda: Executive Summary

- Over the past forty years, the Internet has changed the way we live, work and play, and altered our notions of democracy, education, healthcare, entertainment and commerce
- The Internet has been a powerful engine for technological innovation and societal evolution
- However, and the Internet in particular, will need to improve:
more secure, more accessible, more predictable, and more reliable

NetSE Research Agenda: Recommendations

- The funding agencies of the United States government must increase investment in research that will lead to **a better Internet or risk a marginal future role**
- Funding agencies should support a broad array of **interdisciplinary research activities** related to understanding the current Internet and designing future networks to include the Internet
- **New mathematical tools and frameworks.** Realizing better networks requires a foundational mathematical theory of network architecture and design

Network Science and Engineering: Fundamental Challenges



Emerging New Opportunities

- Big Data
- Cloud Computing
- Mobile Internet
- Social Networks

Thank You!



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