

# Understanding Protein Function on a Genome-scale through the Analysis of Molecular Networks



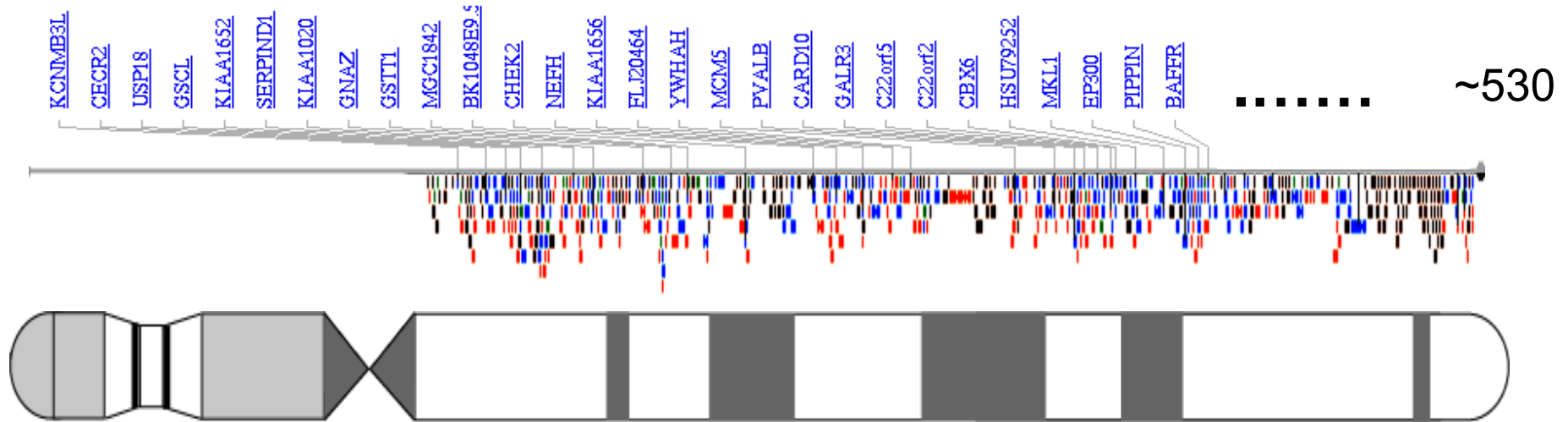
Mark B Gerstein  
Yale

**Slides at**  
[Lectures.GersteinLab.org](http://Lectures.GersteinLab.org)

(See Last Slide for References  
& More Info.)



# The problem: Grappling with Function on a Genome Scale?



- 250 of ~530  
originally characterized on chr. 22  
[Dunham et al. Nature (1999)]
- >25K Proteins in Entire Human Genome  
(with alt. splicing)



# UniProt

# EF2\_YEAST

**Descriptive Name:**  
Elongation Factor 2

**Lots of references**  
to papers

**Summary sentence describing function:**  
This protein promotes the GTP-dependent translocation of the nascent protein chain from the A-site to the P-site of the ribosome.

EF2\_YEAST | Basic UniProt Protein Viewer - UniProt | [the universal protein knowledgebase]

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# UniProt

the universal protein knowledgebase

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Home > Database > UniProt Protein Viewer

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## General information about the UniProt/Swiss-Prot entry

Entry name	EF2_YEAST
Primary accession number	P32324
Entered in Swiss-Prot	Release 27, 01-OCT-1993
Sequence was last modified	Release 27, 01-OCT-1993
Annotations were last modified	Release 47, 01-MAY-2005

## Protein description

Protein name	Elongation factor 2
Synonyms	EF-2

## References

[1]	NUCLEOTIDE SEQUENCE (EF1 AND EF2). MEDLINE=92112760; PubMed=1730643; [NCBI, ExPASy, EBI, Israel, Japan] Perentesis J.P., Phan L.D., Laporte D.C., Livingston D.M., Bodley J.W., "Saccharomyces cerevisiae elongation factor 2. Genetic cloning, characterization of expression, and G-domain modeling.";
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## Comments

FUNCTION	This protein promotes the GTP-dependent translocation of the nascent protein chain from the A-site to the P-site of the ribosome.
SUBCELLULAR LOCATION	Cytoplasmic.

EF2\_YEAST

U26373, A46042.1, [EMBL] GenBank DDBJ [CodingSequence]

DA1778- DA1778

Internet



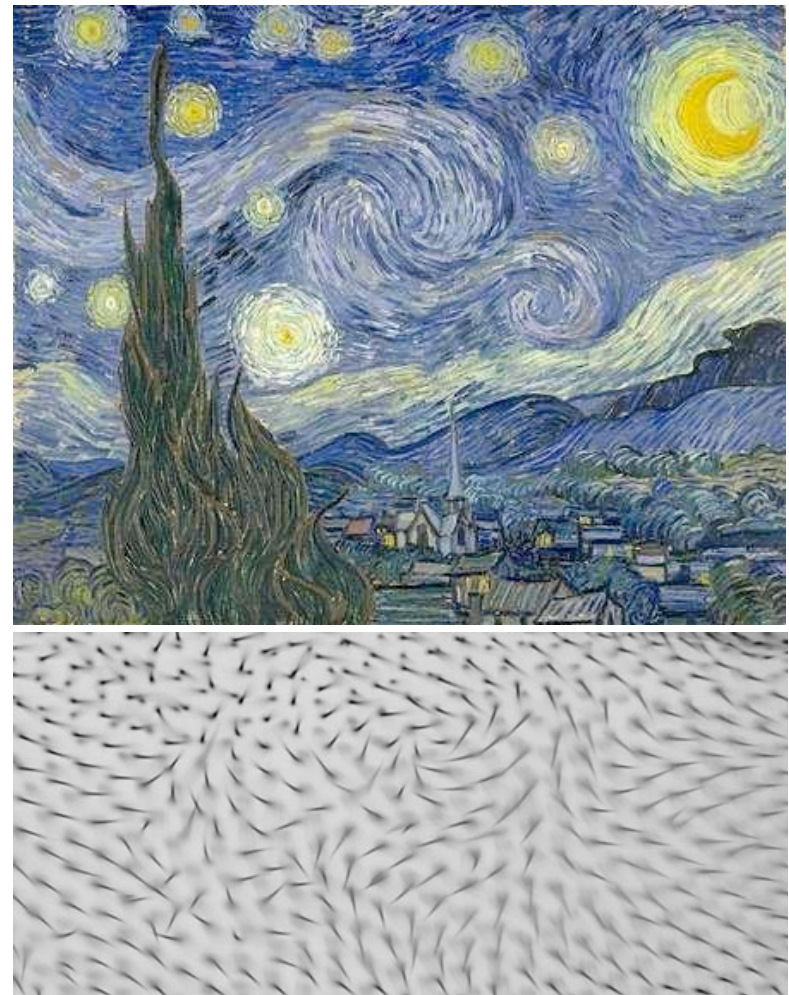
# Some obvious issues in scaling single molecule definition to a genomic scale

- Fundamental complexities
  - ◇ Often >2 proteins/function
  - ◇ Multi-functionality:  
2 functions/protein
  - ◇ Role Conflation:  
molecular, cellular, phenotypic



# Some obvious issues in scaling single molecule definition to a genomic scale

- Fundamental complexities
  - ◇ Often >2 proteins/function
  - ◇ Multi-functionality:  
2 functions/protein
  - ◇ Role Conflation:  
molecular, cellular, phenotypic
- Fun terms... but do they scale?....
  - ◇ **Starry night** (P Adler, '94)





# An Ontology of Naming Pathologies

Single

**M**

Explicit meaning

M-scientific SEMA5A<sup>a</sup>

Not "funny"; usually acronym or concatenation of long descriptive scientific name

M-literal drop dead<sup>b</sup>

Inherent meaning of words is sufficient to describe gene function in some way; no cultural knowledge is required

M-embed

Clever reference or allusion. Cultural savvy or other knowledge required to make sense

Literary malvolio<sup>c</sup>

Acronym LOV<sup>d</sup>

Historical yuri<sup>e</sup>

Pop culture tribbles<sup>f</sup>

**~M**

No explicit meaning

~M-outside kuzbanian<sup>g</sup>

Some outside, non-obvious reason for name

~M-irrel ring<sup>h</sup>

Irrelevant acronym; not tied to gene function

~M-nr yippee<sup>i</sup>

Silly or funny names. No relevance to underlying gene function

Multi

**T**

Transferred naming system

T-relation kryptonite and superman

Naming ceases to make sense if names are shuffled among genes

T-norelation arleekin  
valiet  
tungus...<sup>k</sup>

Names could be shuffled among genes with no loss of meaning

**P**

Problematic relationships

P-clash PKD1 and lov-1<sup>l</sup>

Analogous genes with very different names

P-confusion MT-1<sup>m</sup>

Many genes with same name, or many names for one gene

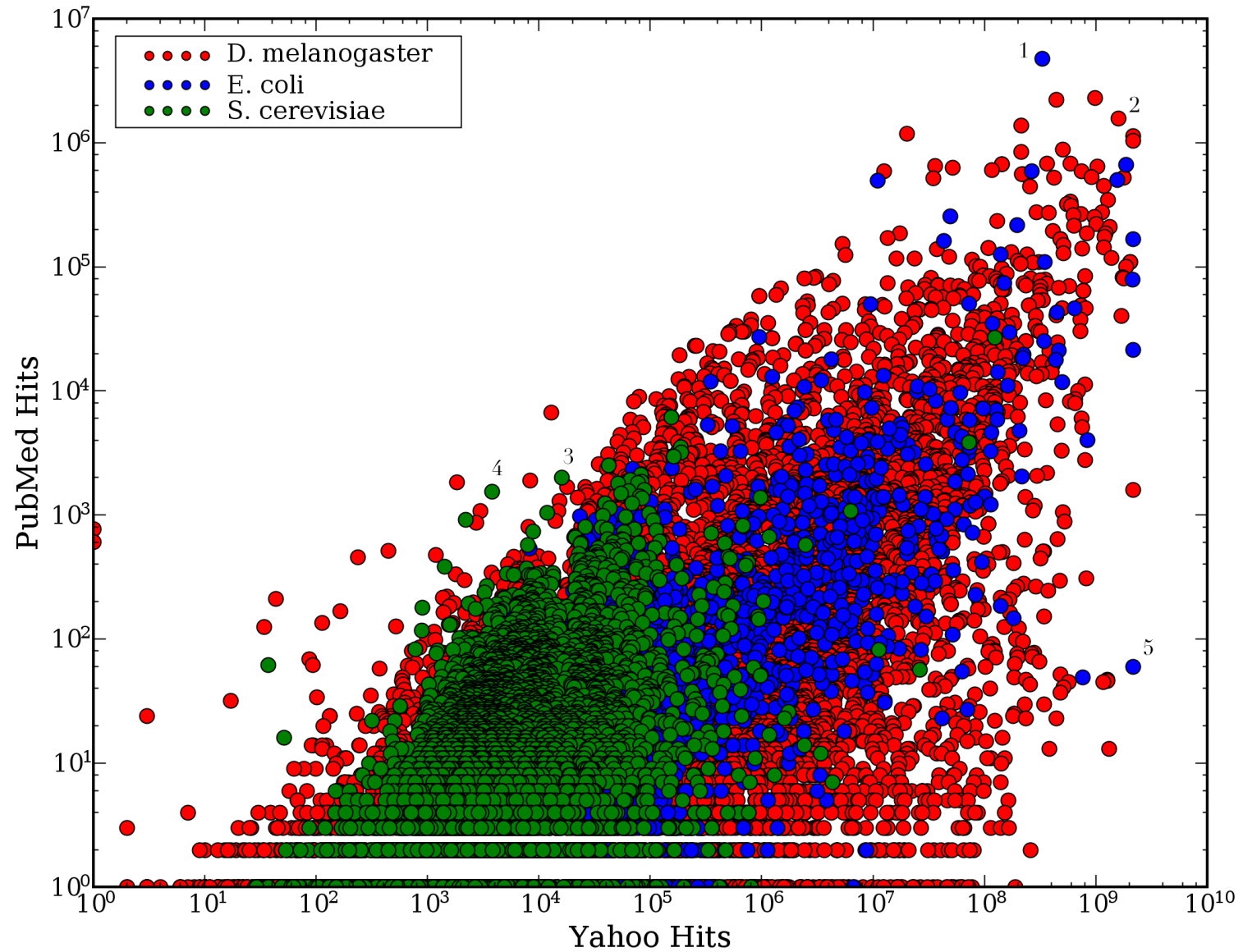
P-defunct BAF45 and BAF47<sup>n</sup>

Gene named to reflect information later shown to be inaccurate or untrue

[Seringhaus et al. GenomeBiology (2008)]



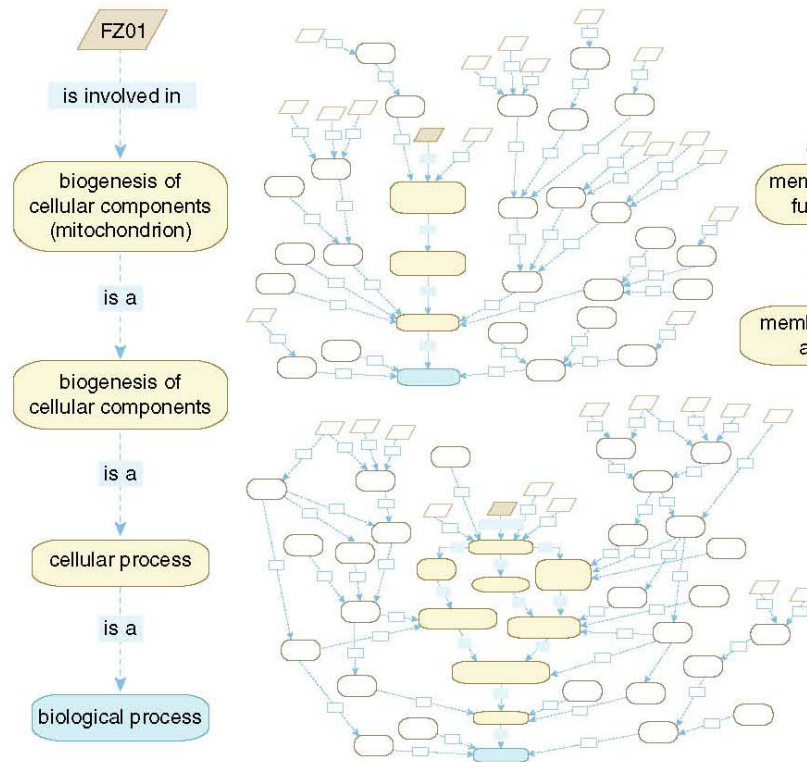
# Gene Name Skew



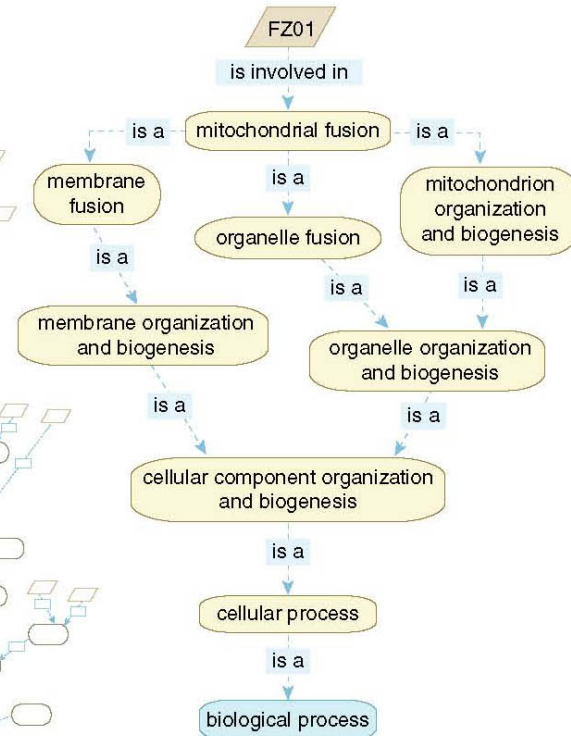
[Seringhaus et al. GenomeBiology (2008)]



# Hierarchies & DAGs of controlled-vocab terms but still have issues...



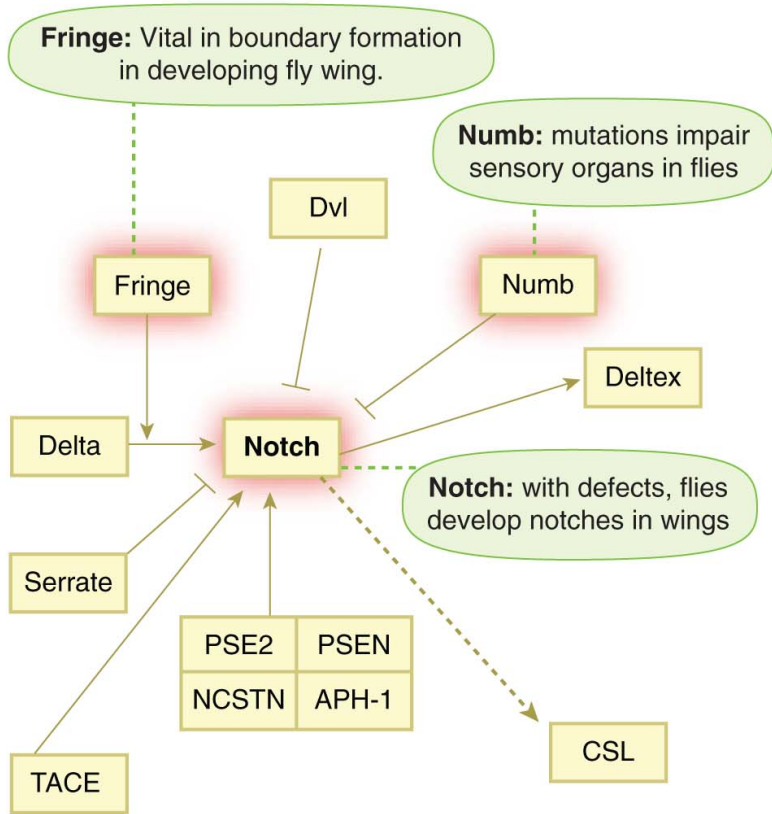
**MIPS (Mewes et al.)**



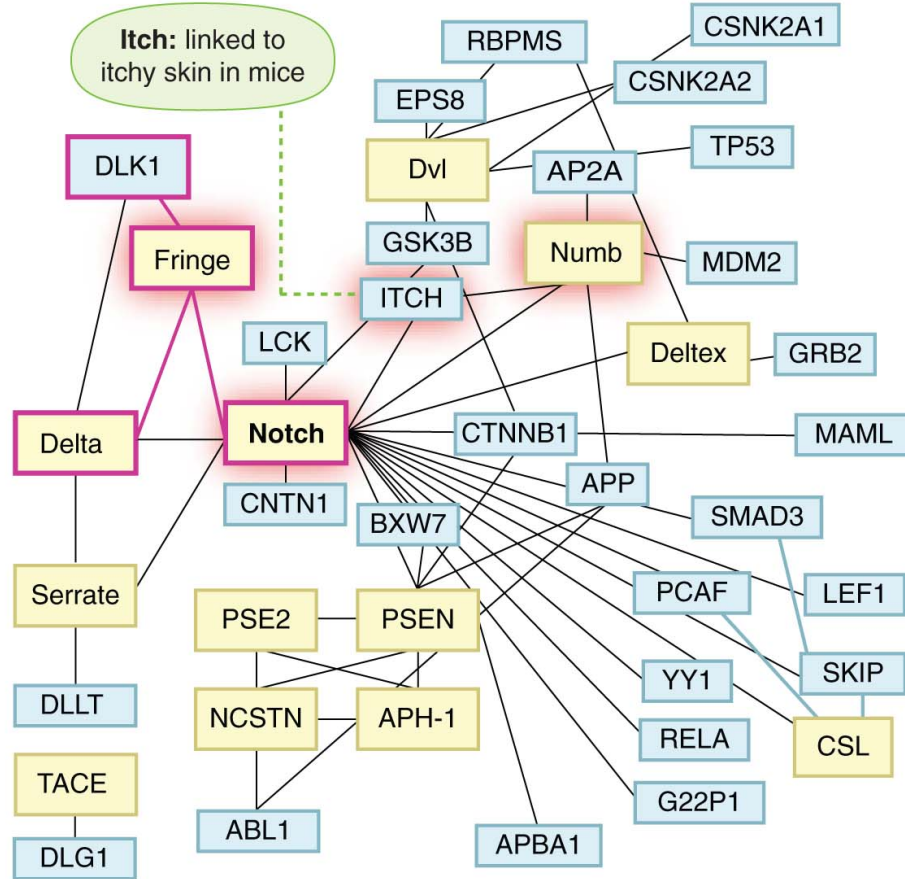
**GO (Ashburner et al.)**



# Networks (Old & New)



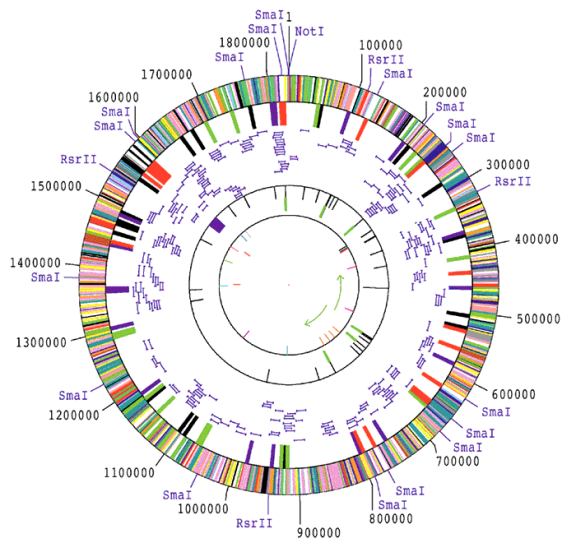
### Classical KEGG pathway



## Same Genes in High-throughput Network

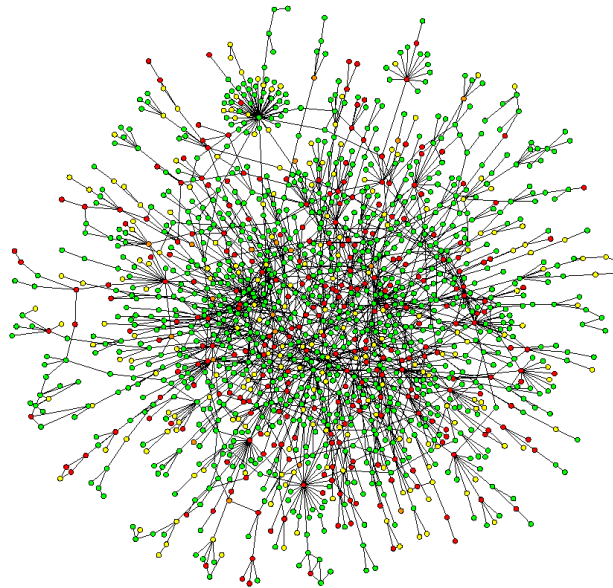


# Networks occupy a midway point in terms of level of understanding



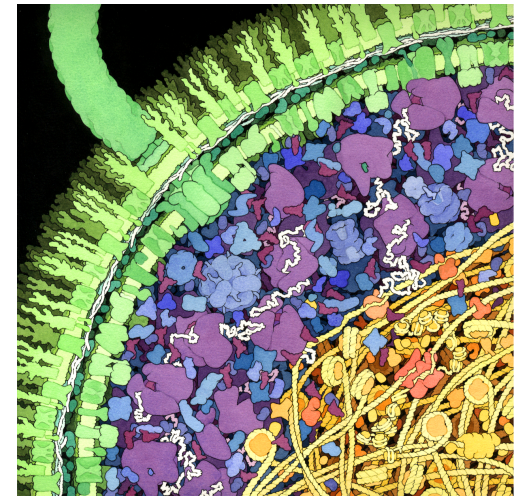
1D: Complete  
Genetic Partslist

[Fleischmann et al., Science, 269 :496]



~2D: Bio-molecular  
Network  
Wiring Diagram

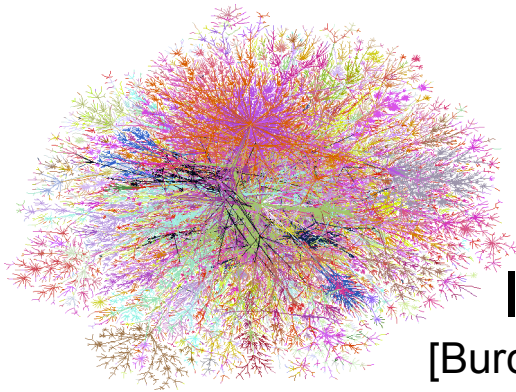
[Jeong et al. Nature, 41:411]



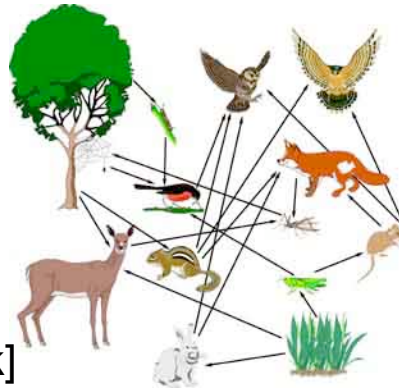
3D: Detailed  
structural  
understanding of  
cellular machinery



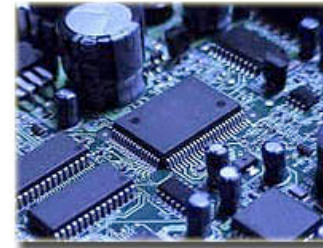
# Networks as a universal language



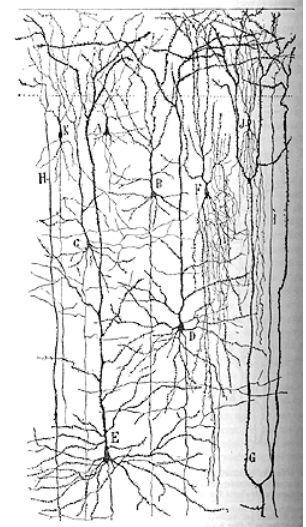
Internet  
[Burch & Cheswick]



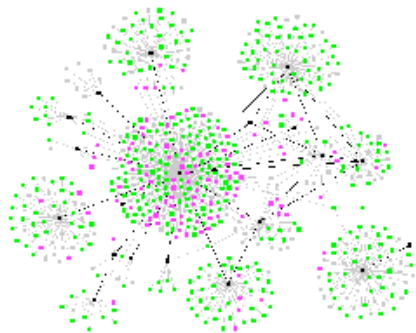
Food Web



Electronic  
Circuit



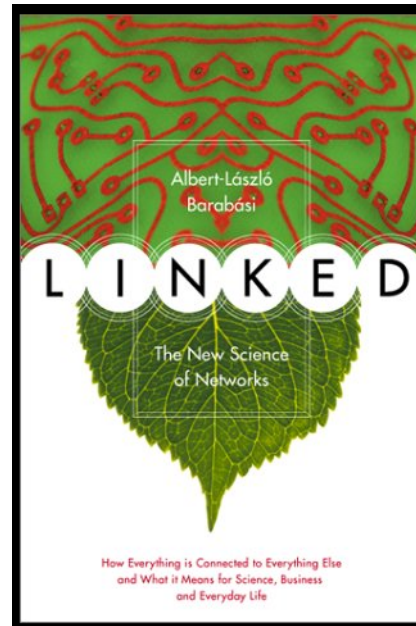
Neural Network  
[Cajal]



Disease  
Spread  
[Krebs]



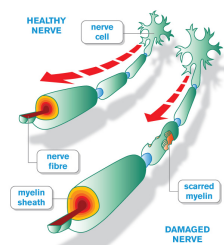
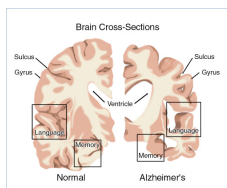
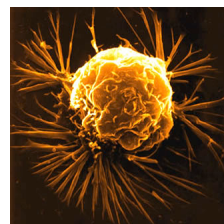
Protein  
Interactions  
[Barabasi]



Social Network



# Network pathology & pharmacology

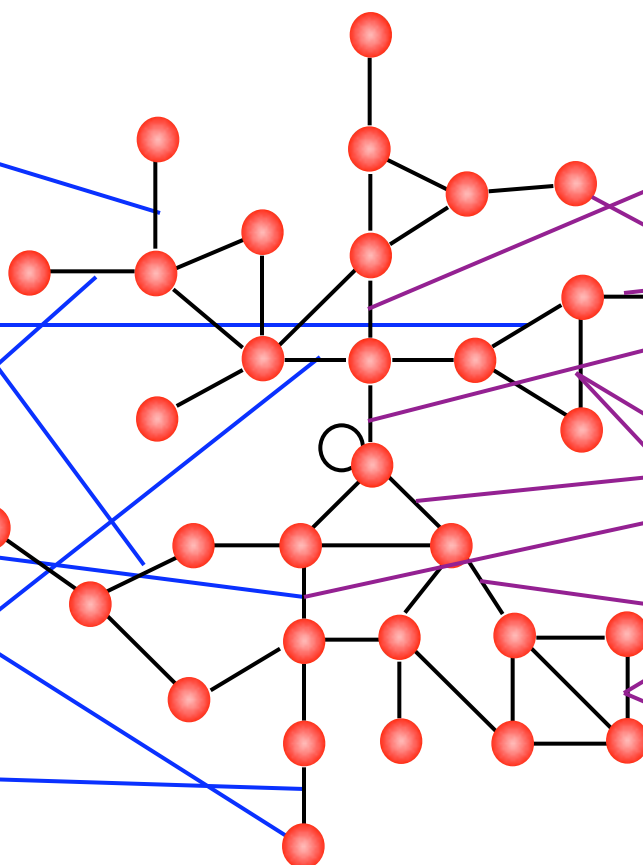


**Breast Cancer**

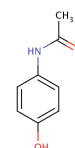
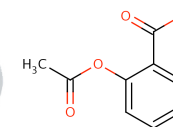
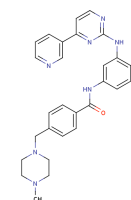
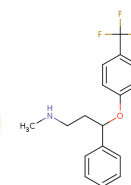
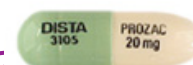
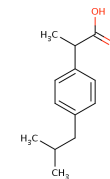
**Alzheimer's Disease**

**Parkinson's Disease**

**Multiple Sclerosis**



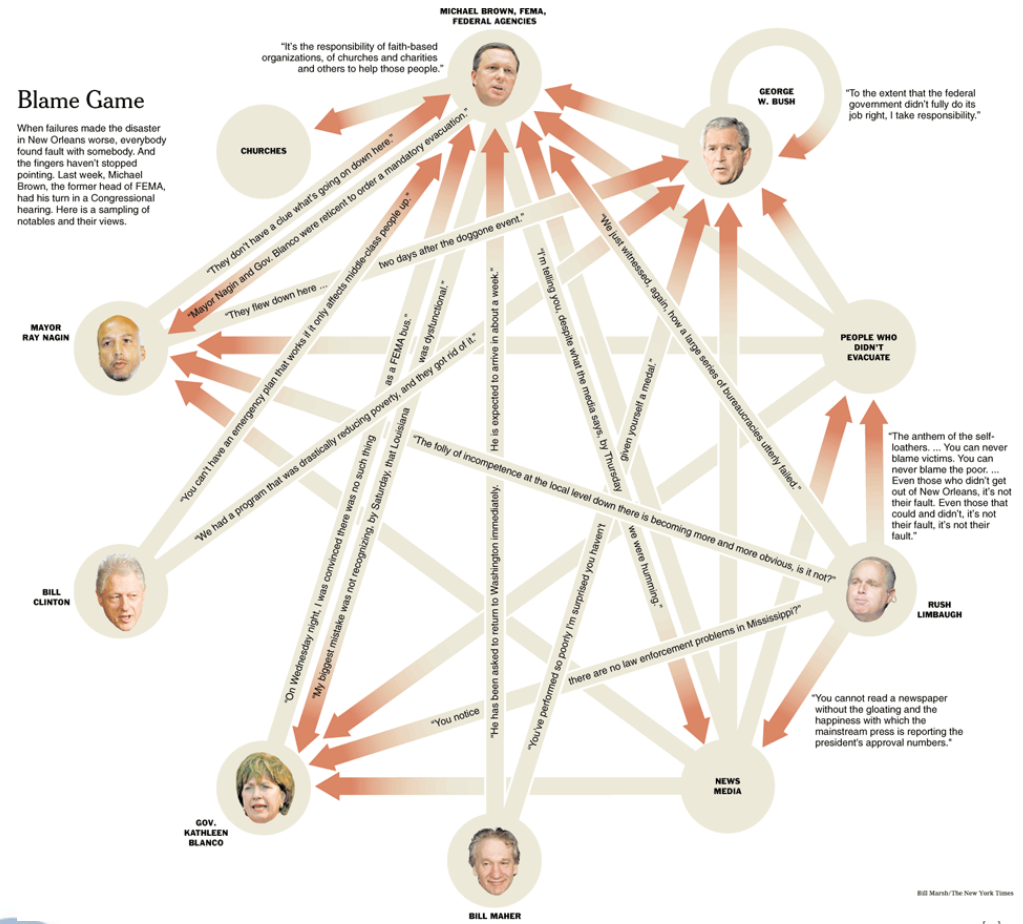
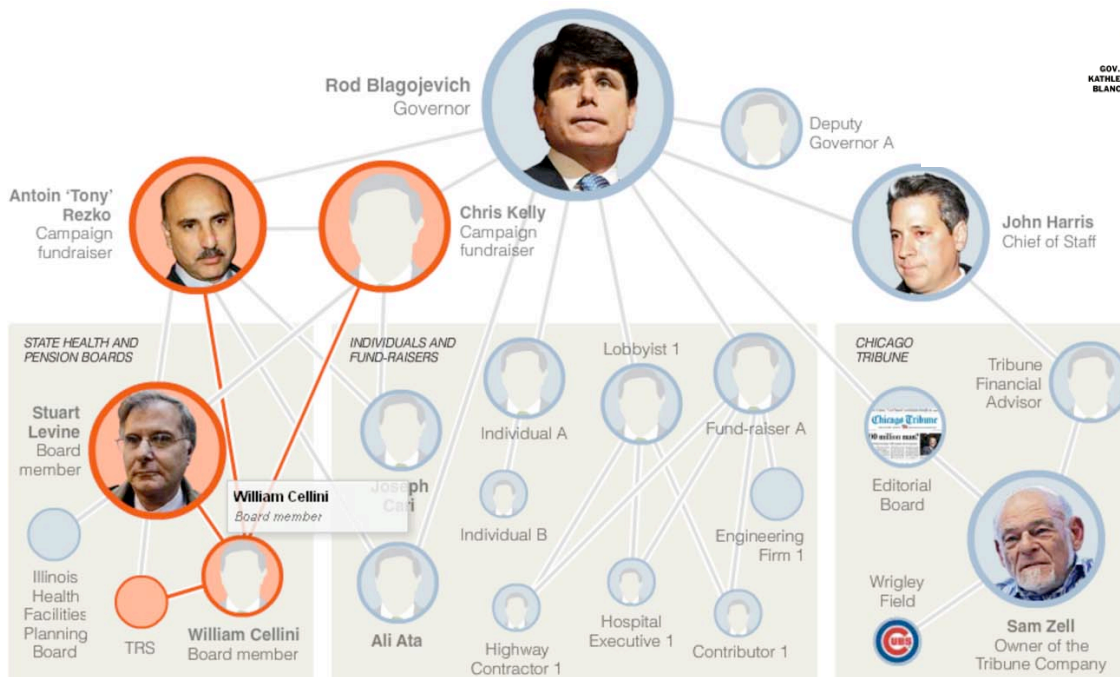
Interactome networks



[Adapted from H Yu]



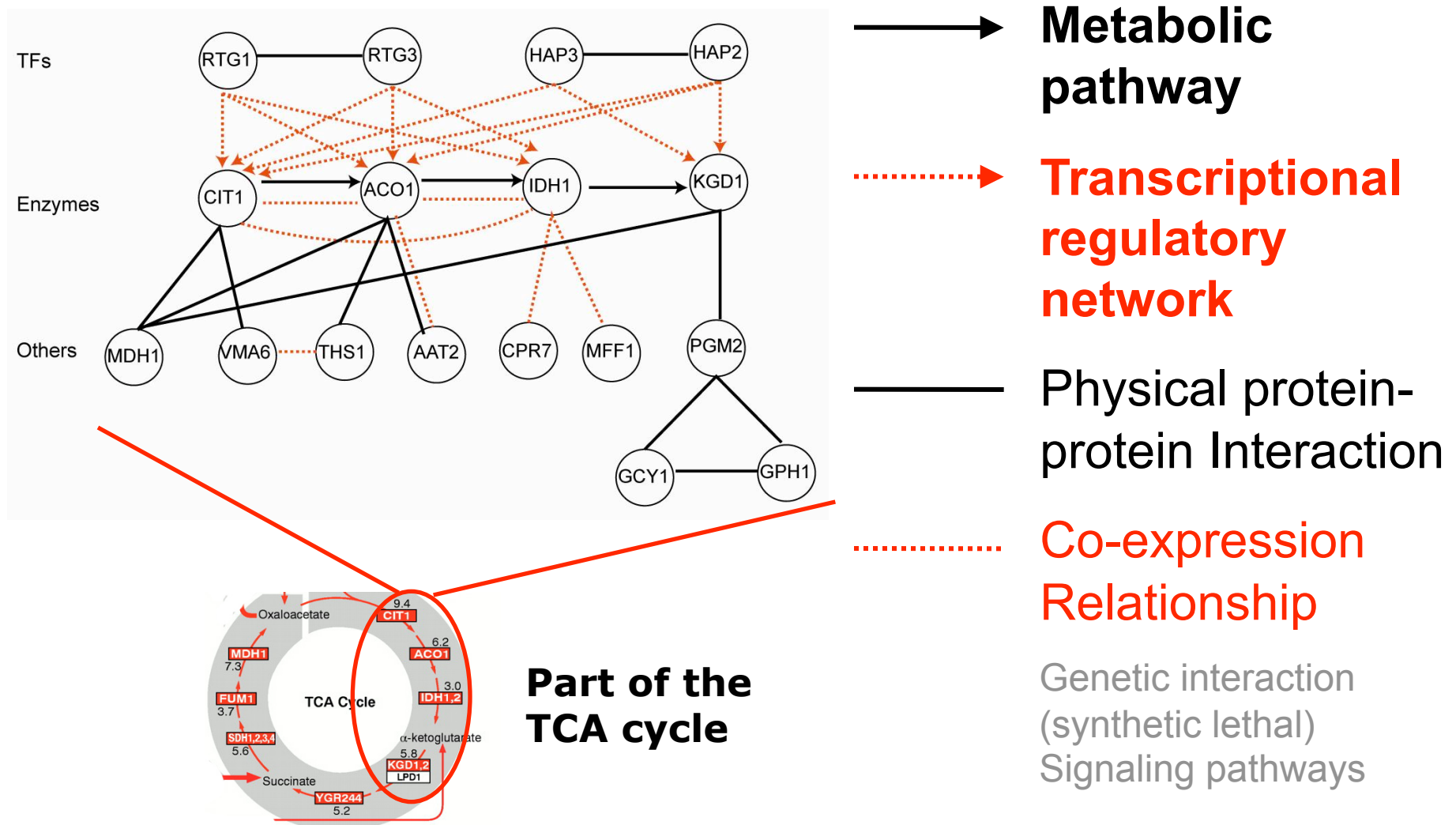
# Using the position in networks to describe function



[NY Times, 2-Oct-05, 9-Dec-08]



# Combining networks forms an ideal way of integrating diverse information





# Outline: Molecular Networks

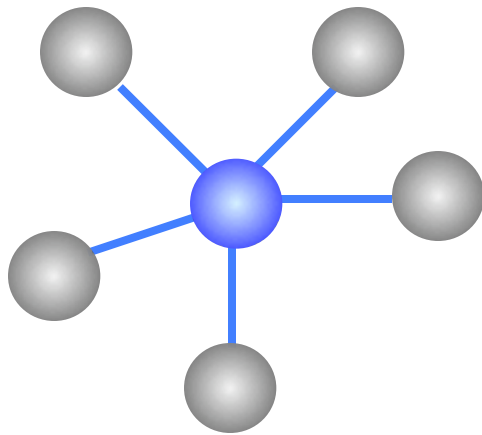
- Why Networks?
- Network Structure:  
Key Positions
  - ◊ Hubs & Bottlenecks
  - ◊ Tops of a Hierachy
- Networks, Variation & the Environment
  - ◊ Which pathways change most with the environment



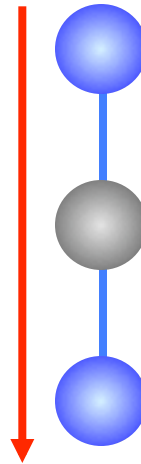


# Global topological measures

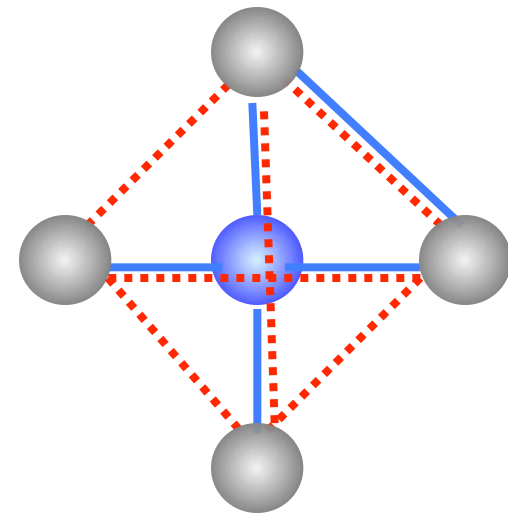
Indicate the gross topological structure of the network



Degree ( $K$ )  
5



Path length ( $L$ )  
2



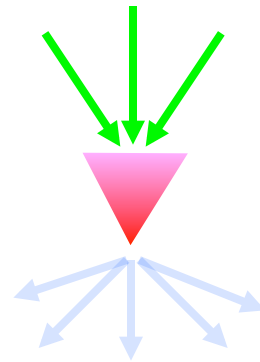
Clustering coefficient ( $C$ )  
 $1/6$

Interaction and expression networks are ***undirected***

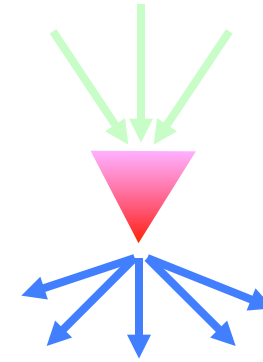
[Barabasi]



# Global topological measures for directed networks



In-degree  
3



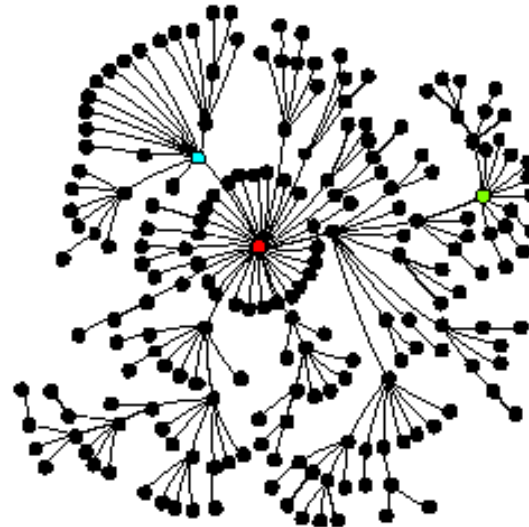
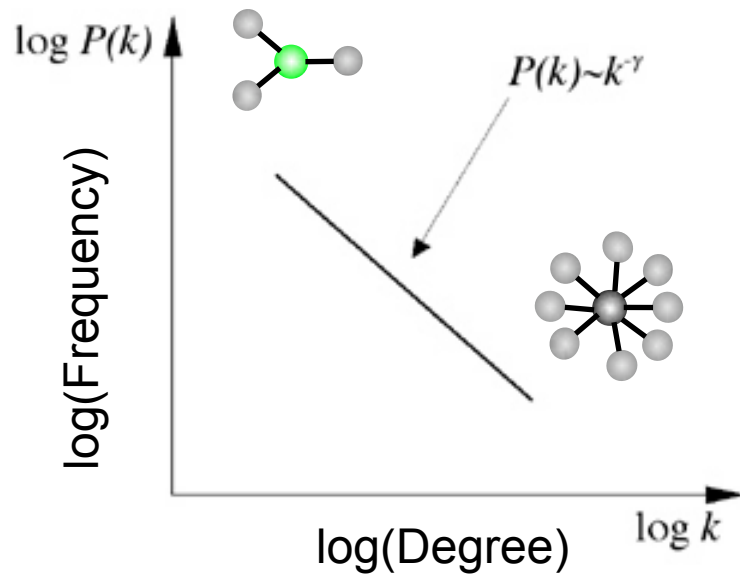
Out-degree  
5

Regulatory and metabolic networks are ***directed***



# Scale-free networks

Power-law distribution



**Hubs** dictate the structure of the network

[Barabasi]

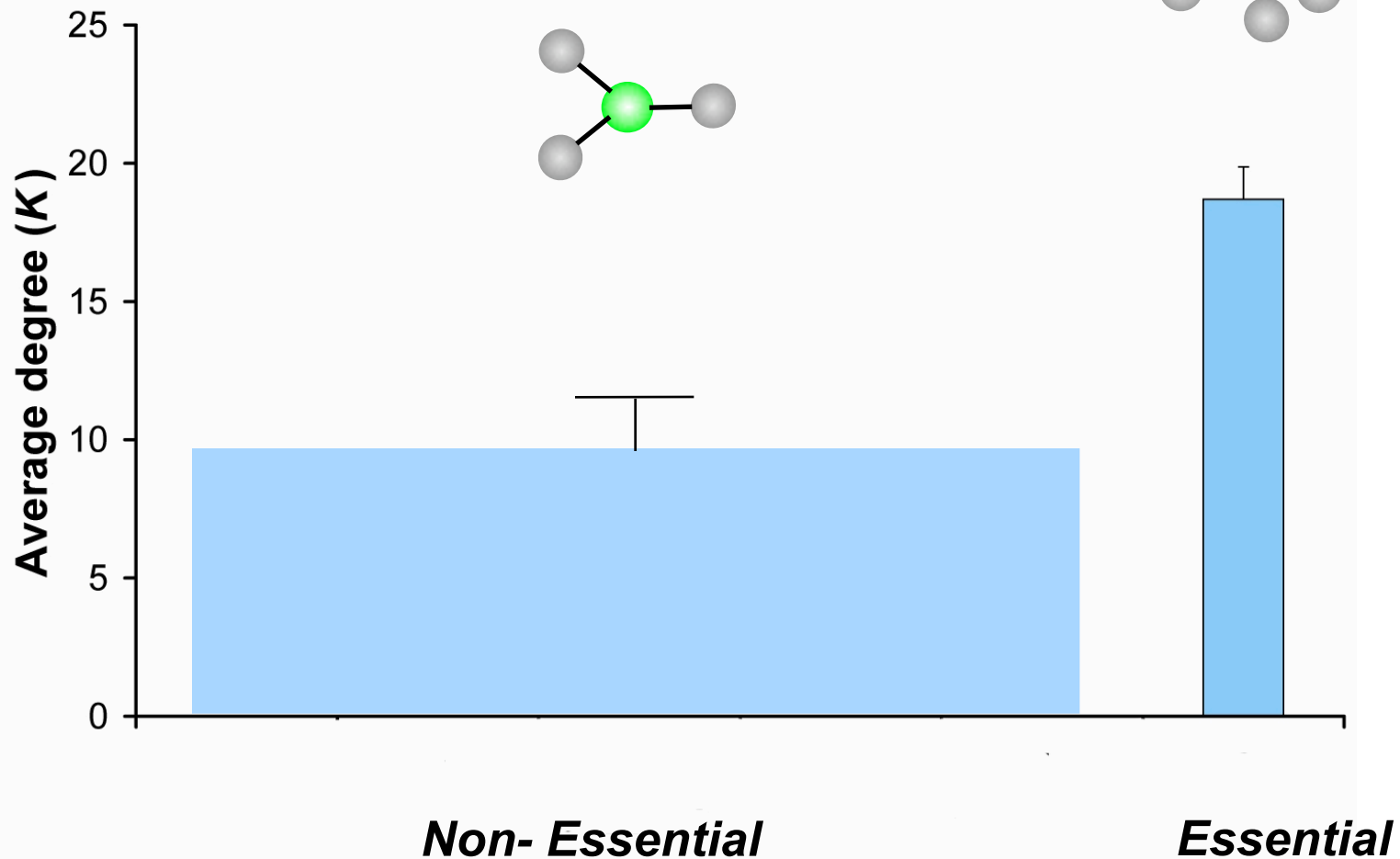


# Hubs tend to be Essential

Integrate gene essentiality data with protein interaction network. Perhaps hubs represent vulnerable points?

[Lauffenburger, Barabasi]

"hubbiness"

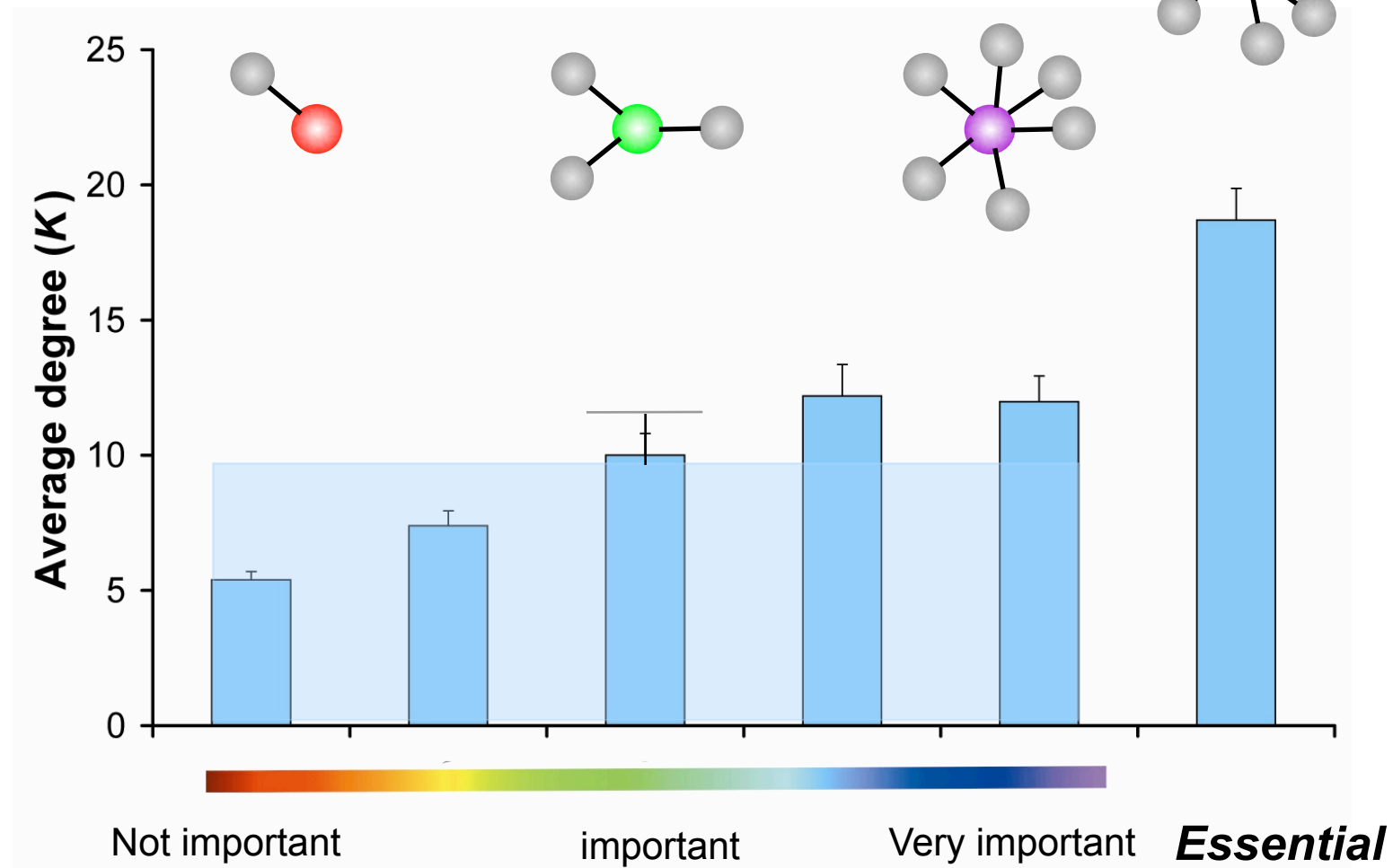




# Relationships extends to "Marginal Essentiality"

Marginal essentiality measures relative importance of each gene (e.g. in growth-rate and condition-specific essentiality experiments) and scales continuously with "hubbiness"

"hubbiness"



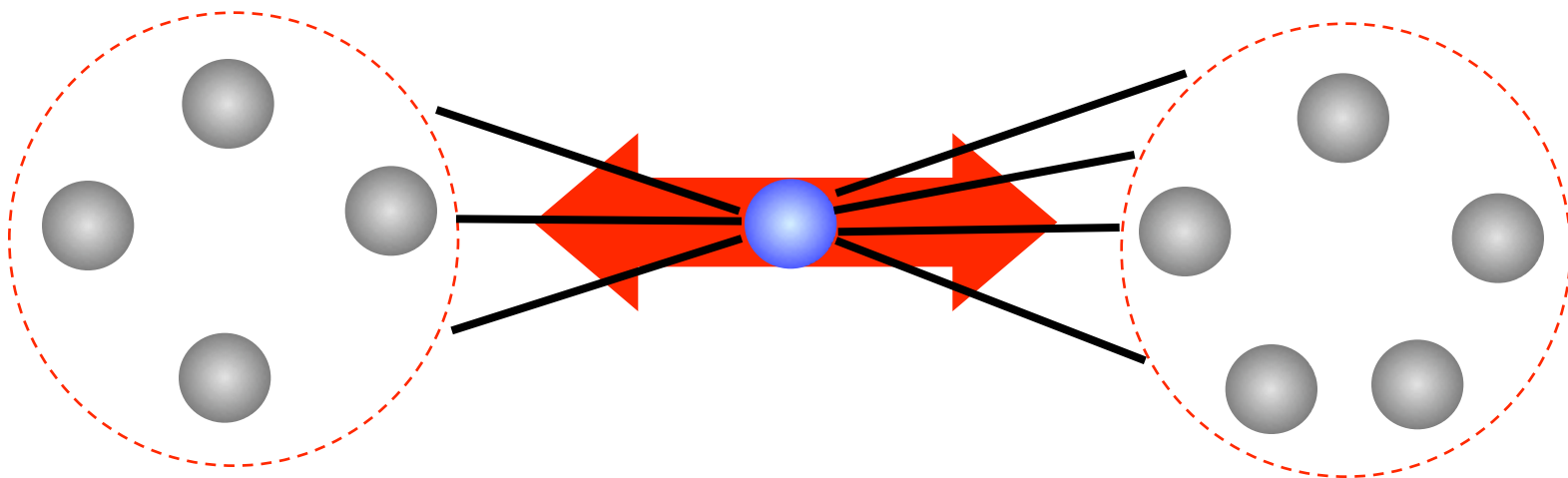


# Another measure of Centrality: Betweenness centrality

Betweenness of a node is the number of shortest paths of pairs of vertices that run through it -- a measure of information flow.

Freeman LC (1977) Set of measures of centrality based on betweenness.  
Sociometry 40: 35–41.

**Girvan & Newman (2002) PNAS 99: 7821.**





# Betweenness centrality -- Bottlenecks

Proteins with high betweenness are defined as *Bottlenecks* (top 20%), in analogy to the traffic system

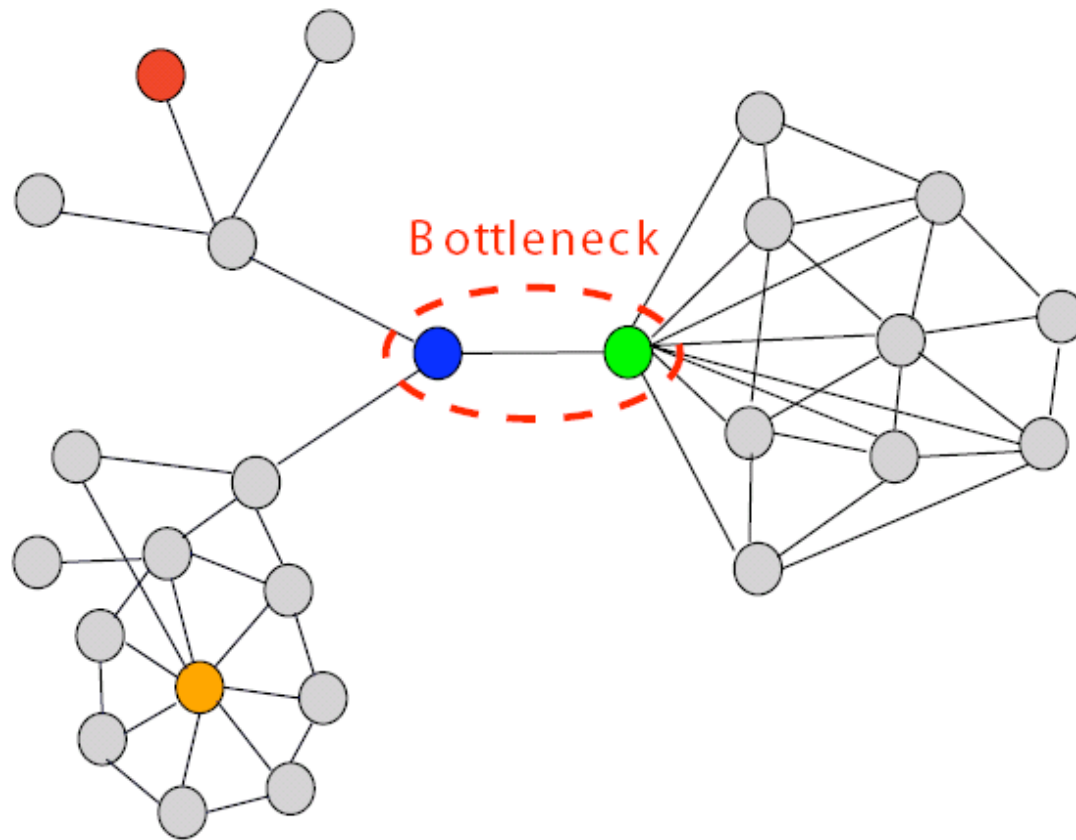






George Washington  
Bridge





# Bottlenecks & Hubs

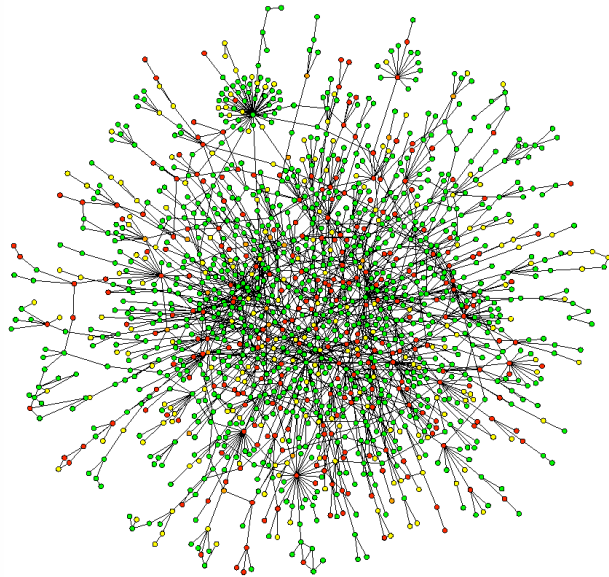


-  Hub-bottleneck **node**
-  Non-hub-bottleneck **node**
-  Hub-non-bottleneck **node**
-  Non-hub-non-bottleneck **node**

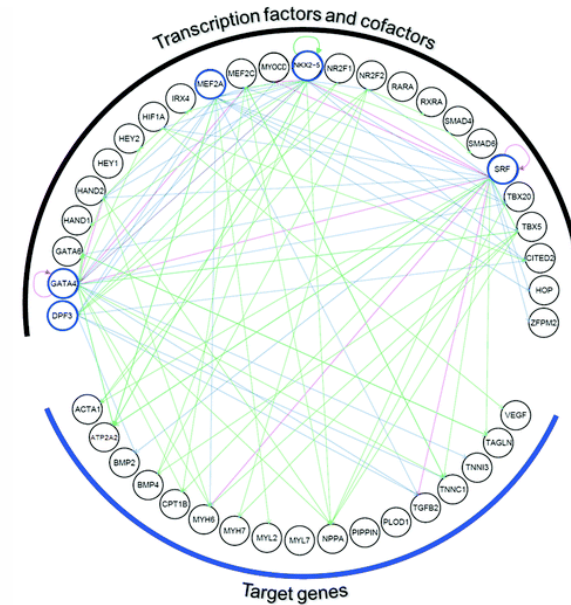
[Yu et al., PLOS CB (2007)]



# Different Interactome networks

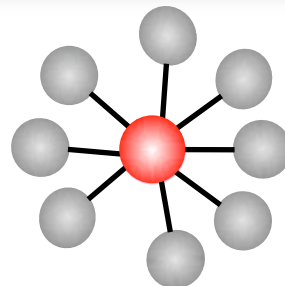


Interaction networks

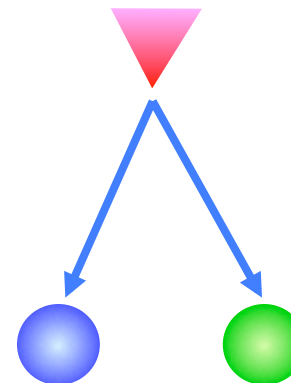


Regulatory networks

Undirected



Directed

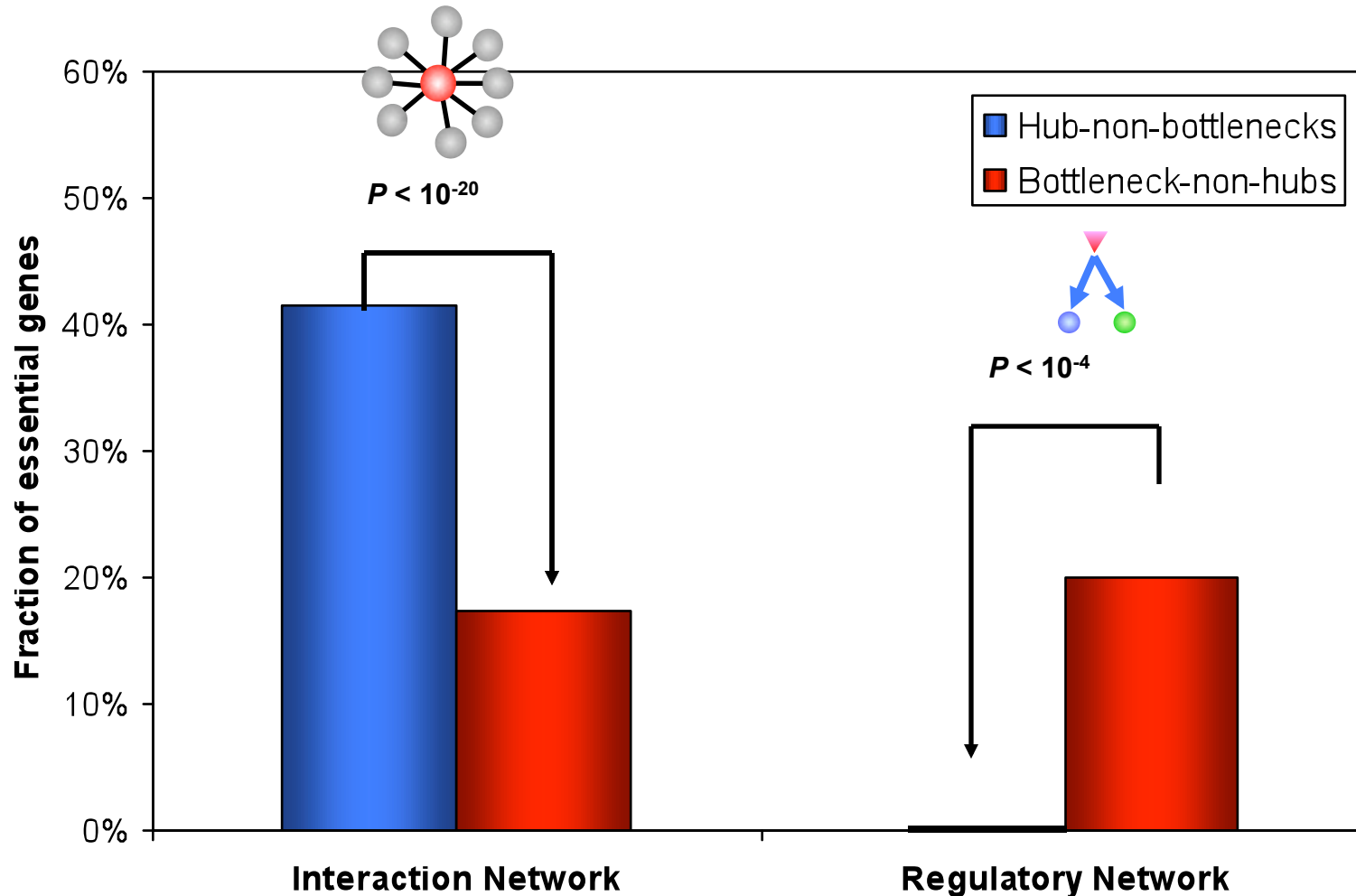


[Toenjes, *et al*, *Mol. BioSyst.* (2008)]

[Jeong *et al*, *Nature* (2001)]

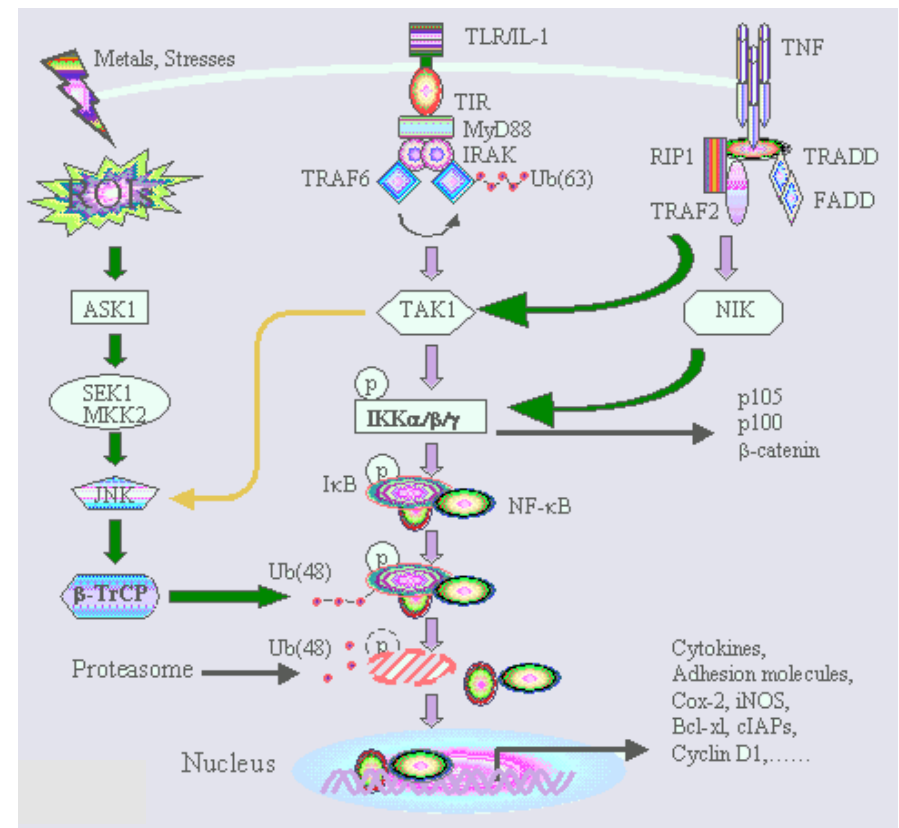


# Bottlenecks are what matters in regulatory networks





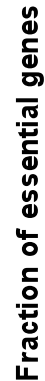
# Signaling transduction pathways are directed



[Xianglin  
Shi ]



## are important





# Outline: Molecular Networks

- Why Networks?
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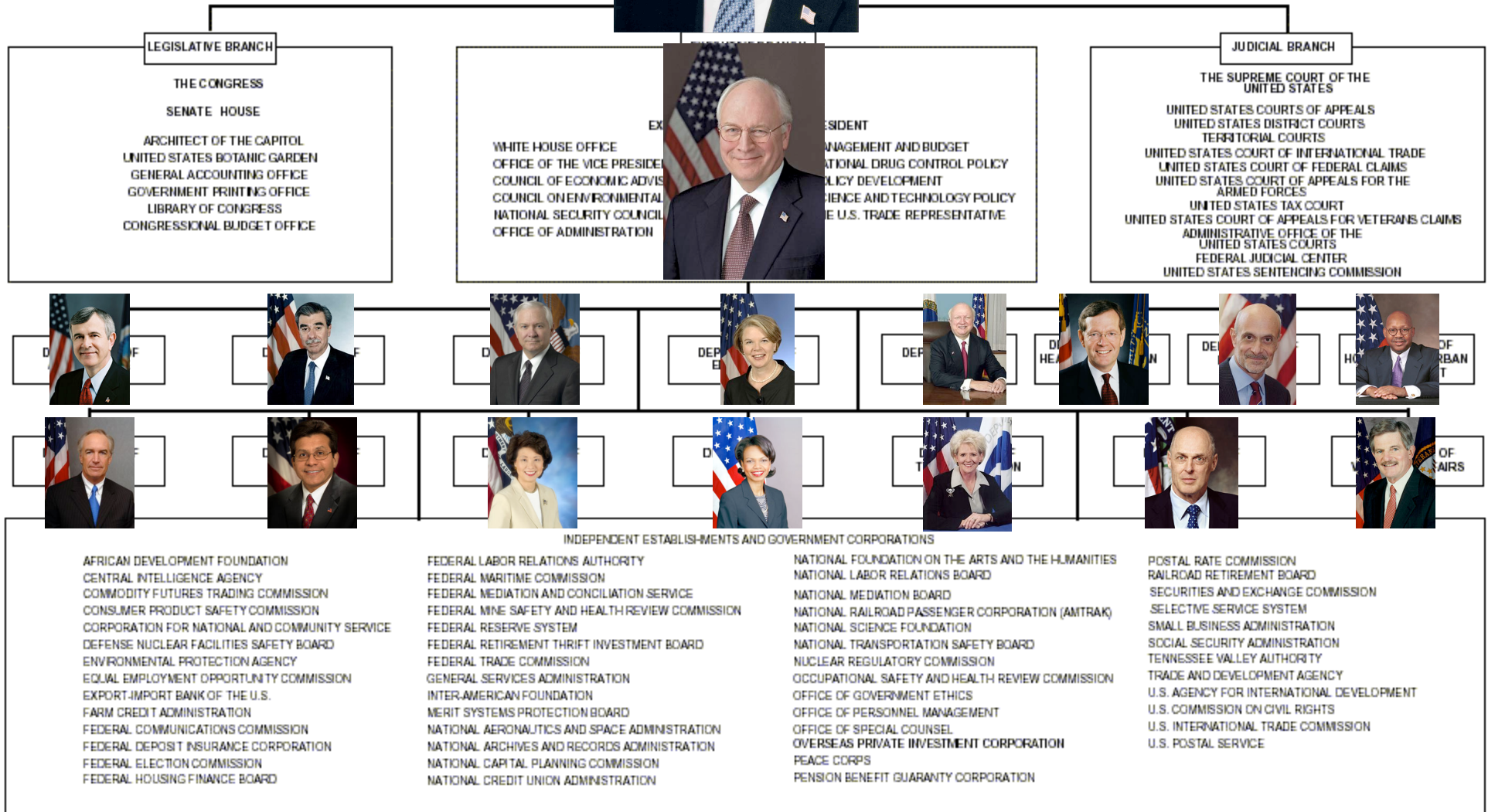
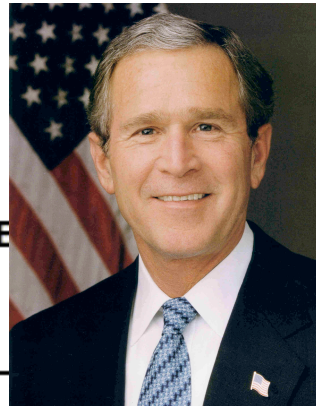




# Social Hierarchy

THE GOVERNMENT

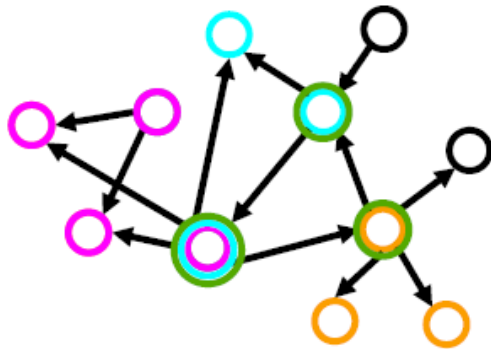
UNITED STATES



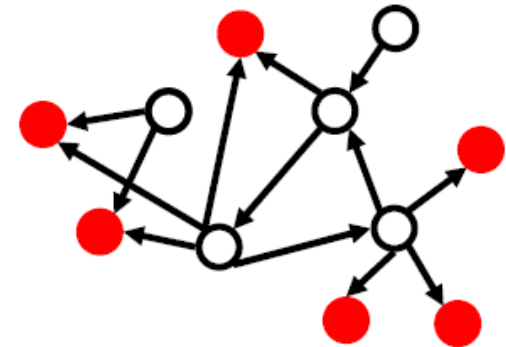


# Determination of "Level" in Regulatory Network Hierarchy with Breadth-first Search

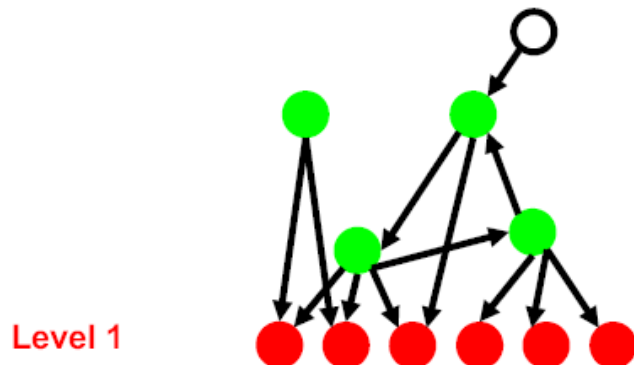
I. Example network with all 4 motifs



II. Finding terminal nodes (Red)

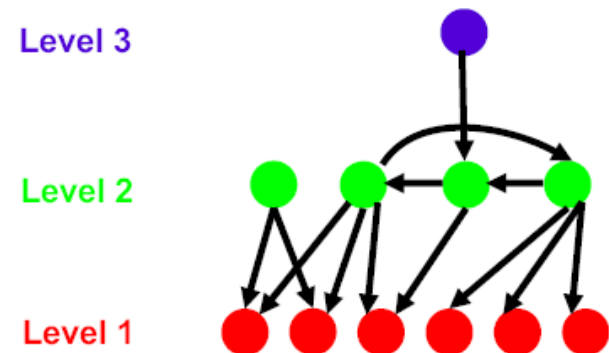


III. Finding mid-level nodes (Green)



Level 1

IV. Finding top-most nodes (Blue)



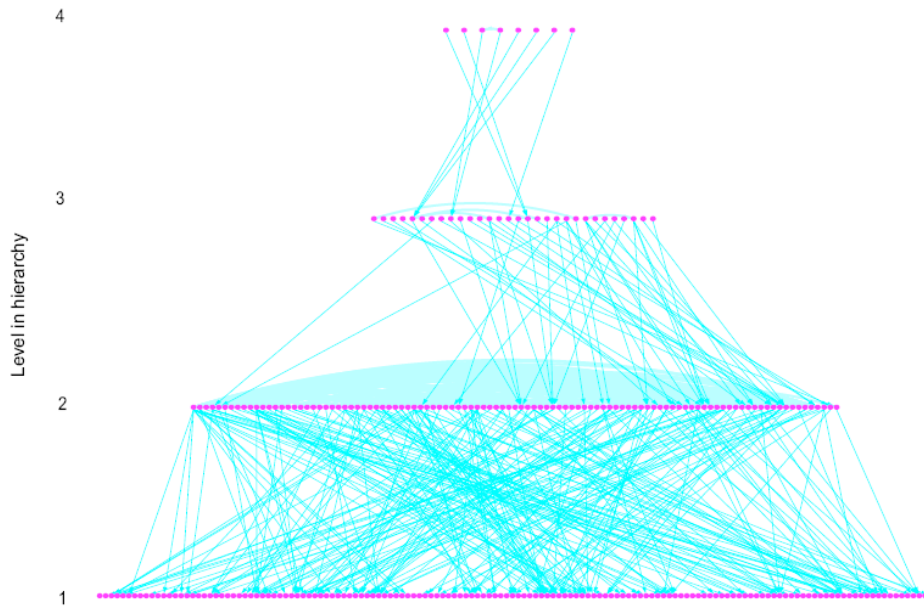
Level 3

Level 2

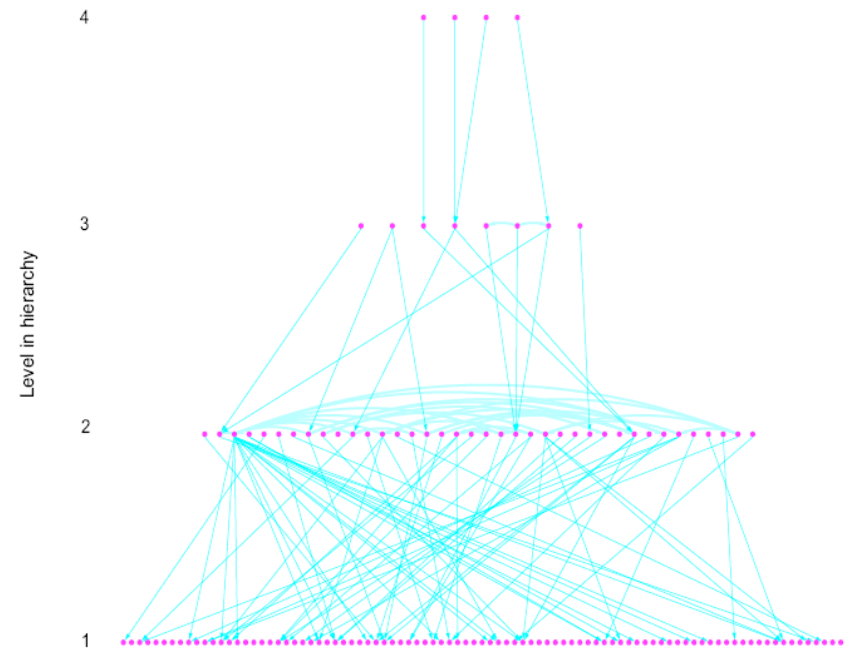
Level 1



# Regulatory Networks have similar hierarchical structures



*S. cerevisiae*

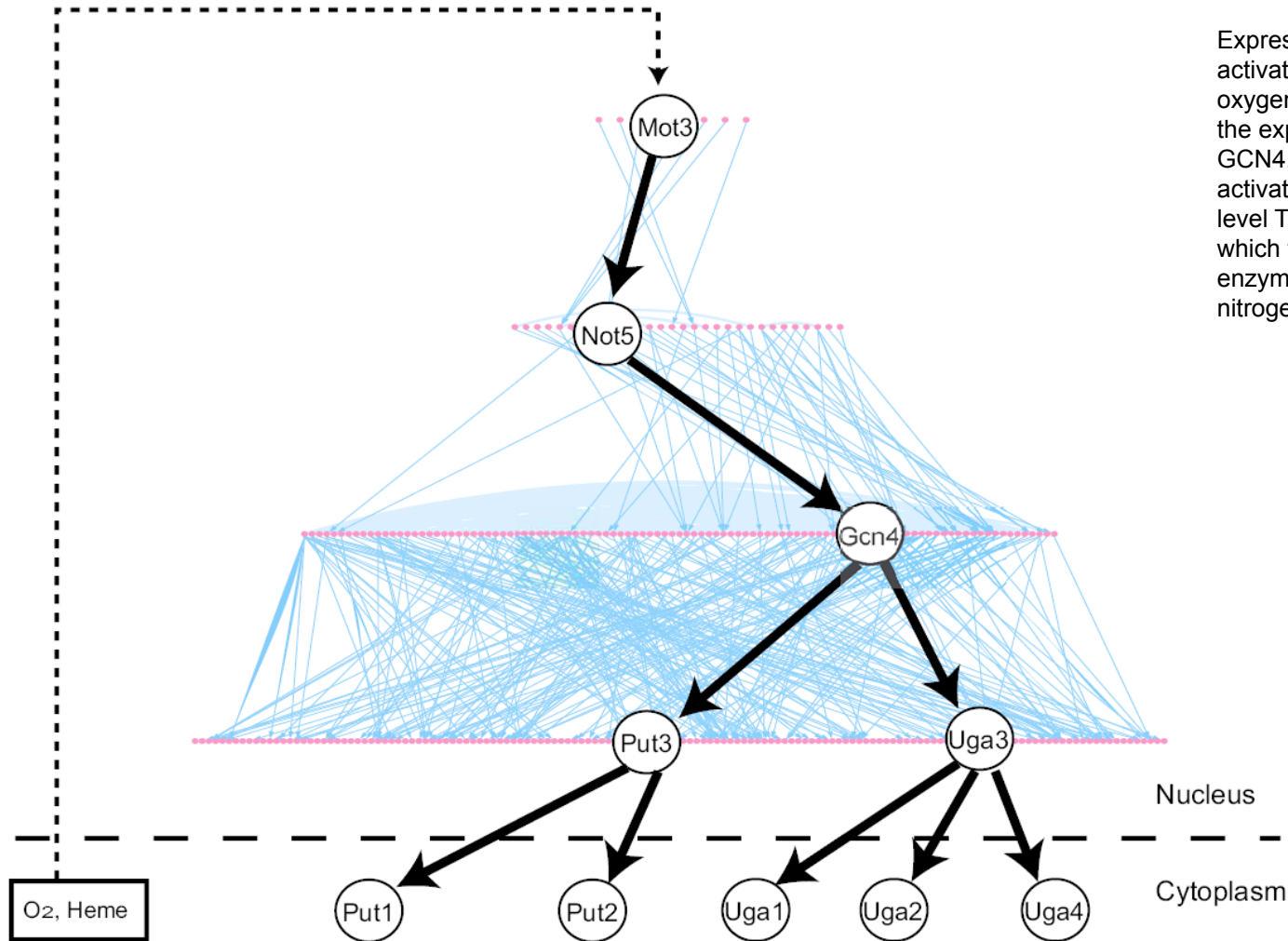


*E. coli*

[Yu et al., Proc Natl Acad Sci U S A (2006)]



# Example of Path Through Regulatory Network

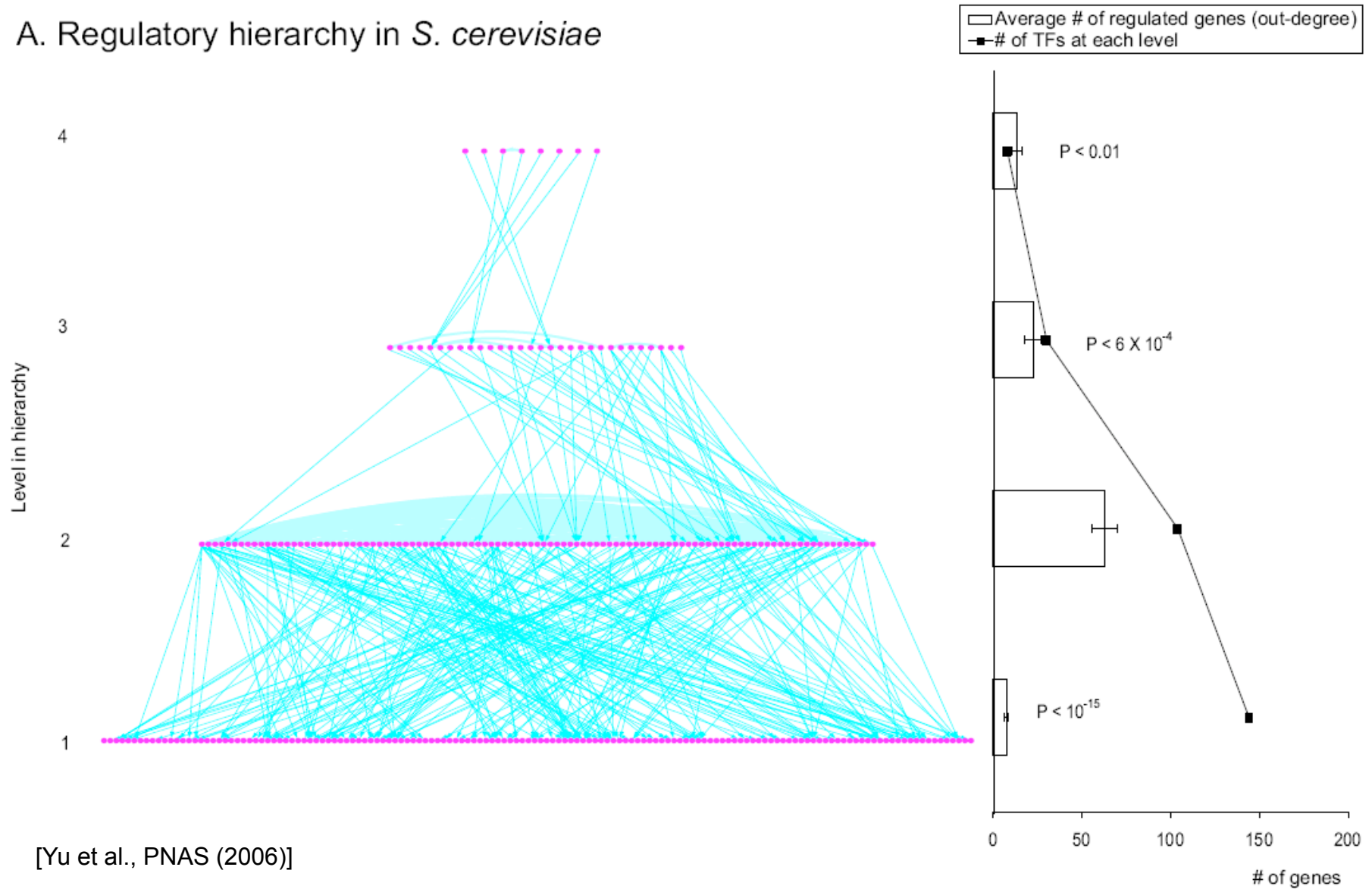


Expression of MOT3 is activated by heme and oxygen. Mot3 in turn activates the expression of NOT5 and GCN4, mid-level hubs. GCN4 activates two specific bottom-level TFs, Put3 and Uga3, which trigger the expression of enzymes in proline and nitrogen utilization.



# Yeast Regulatory Hierarchy: the Middle-managers Rule

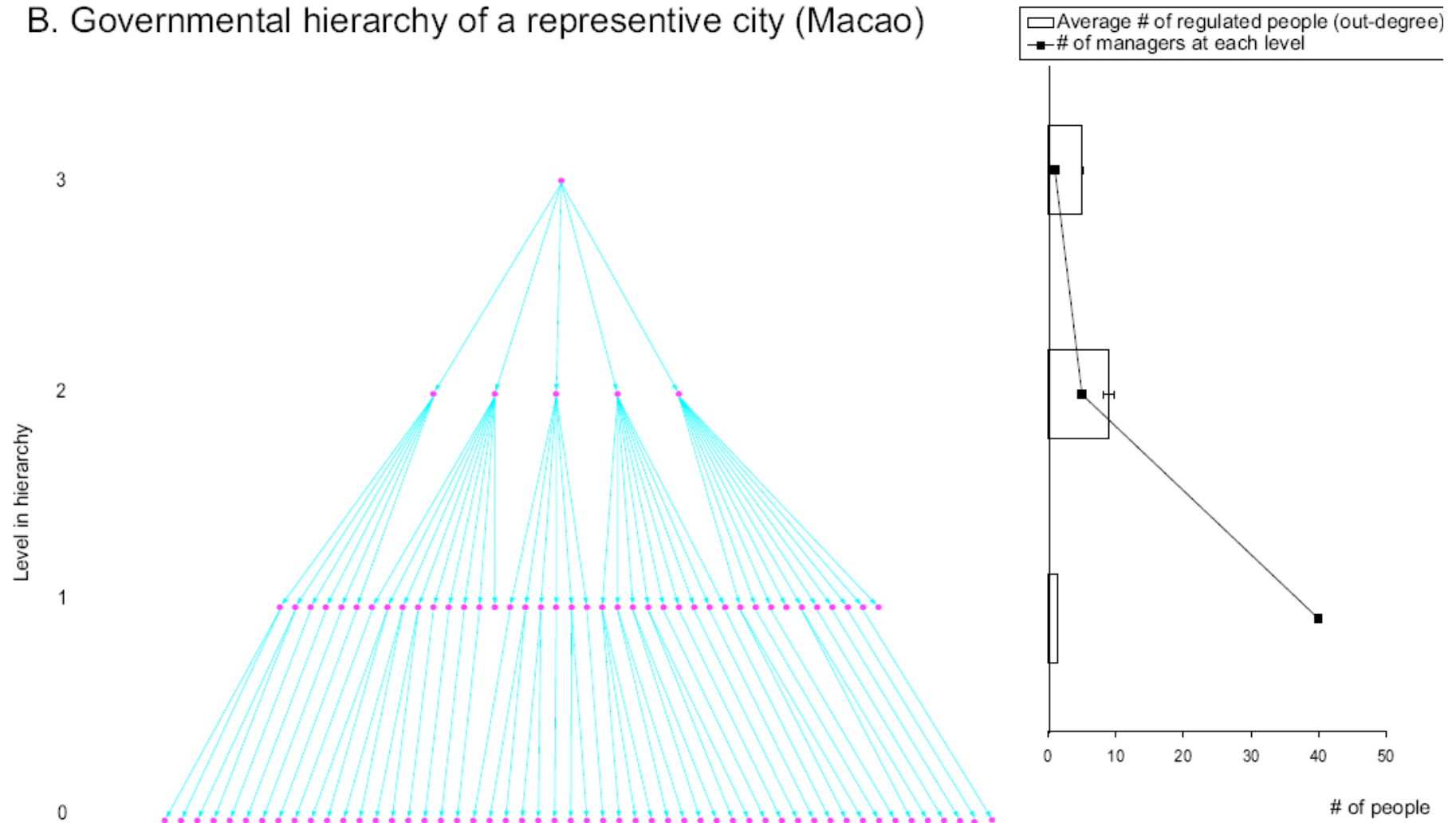
A. Regulatory hierarchy in *S. cerevisiae*





# Yeast Network Similar in Structure to Government Hierarchy with Respect to Middle-managers

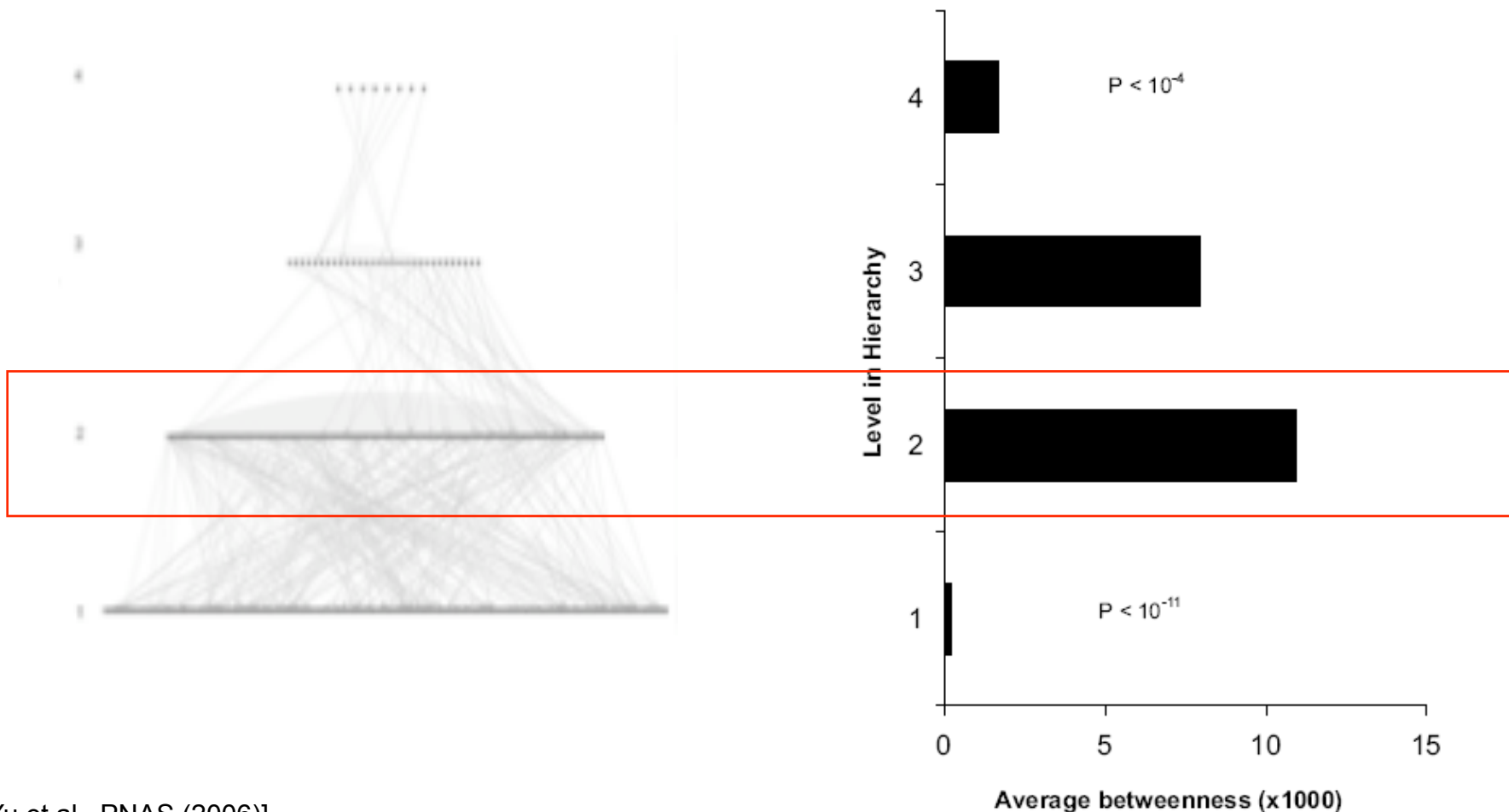
B. Governmental hierarchy of a representative city (Macao)





# Characteristics of Regulatory Hierarchy: Middle Managers are Information Flow Bottlenecks

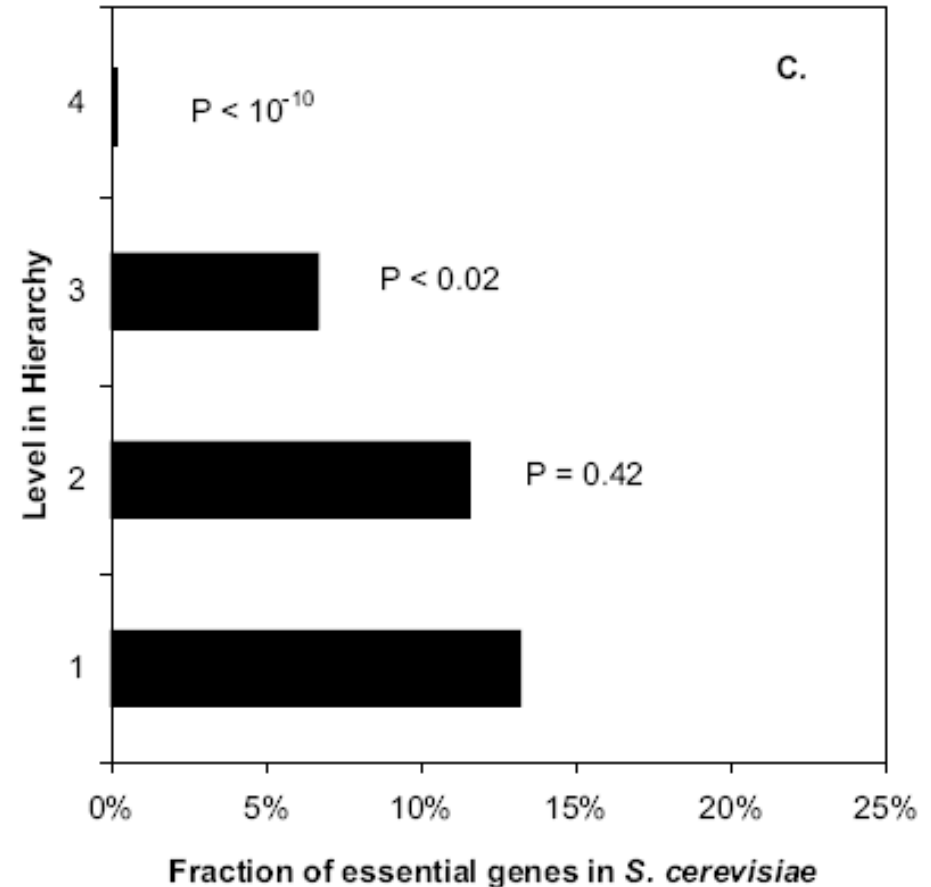
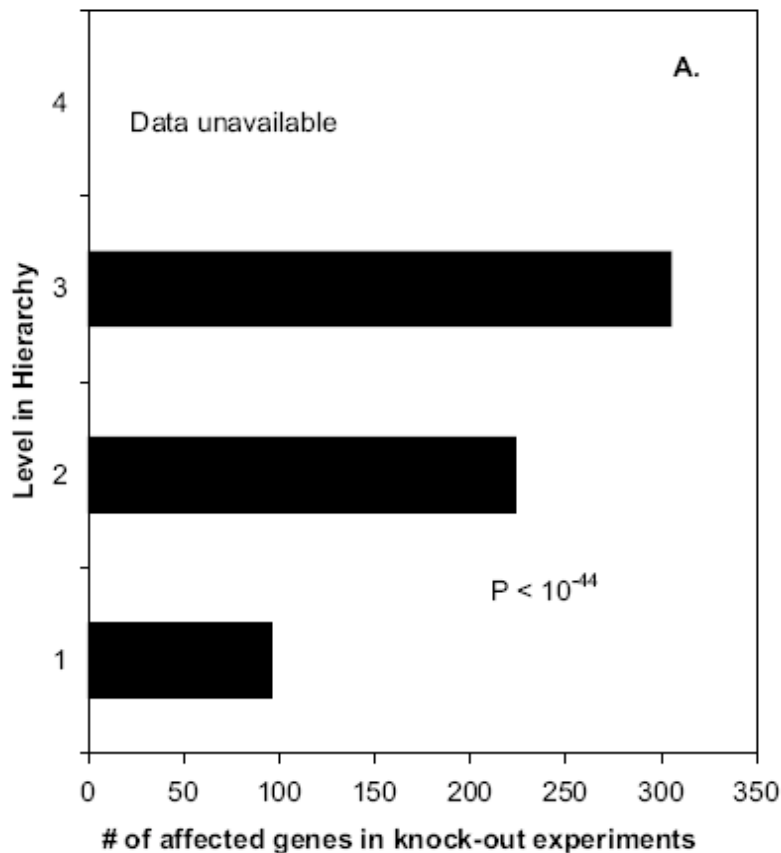
Average betweenness at each level



[Yu et al., PNAS (2006)]



# Characteristics of Regulatory Hierarchy: The Paradox of Influence and Essentiality



[Yu et al., PNAS (2006)]



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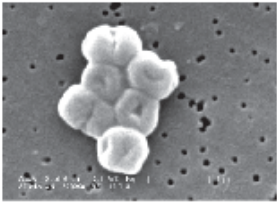




# What is metagenomics?

## Genomics Approach

Culture Microbes



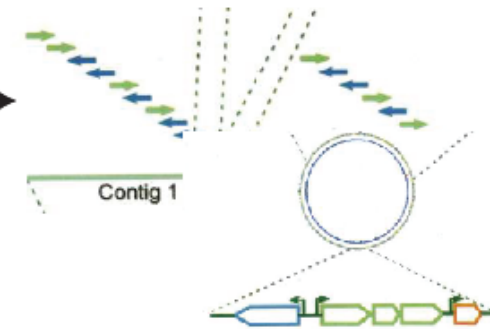
Extract DNA



Sequence

```
ATCGTATA
CGCGAAG
ACGTCTGA
AGTGCTGCT
```

Assemble and Annotate



PROBLEM: Estimated that less than 1% can be cultured in the lab

## Metagenomics Approach

Collect Sample



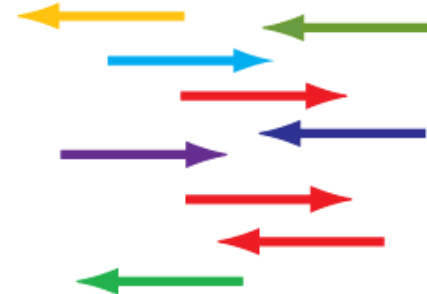
Extract DNA



Sequence

```
ATCGTGATAGATGATAGTAGA
ATGCTGCATGCATCTAGCACT
ACAGTAGCTAGCTACGTAATA
CAGCTGACTAGCTAGCTAGCT
ACGTAGCATGCTAGCTAGCAG
ACGTACGTAGCTAGCTAGTAG
ACGTACGTACGTAGCTAGCATC
AGTCGACTGAGCCAGTGATGAT
ACGATGCATGAGCAGATGCTAC
AGATCGTAGCATGCTAGCATGCT
ACGTACGTAGCTAGCTAGCTAAG
AGCTAGCATGCTAGTAGCATGAG
ACGATGCTAGCTAGCTAGCTGATA
TCGATCAGCATGCTACGATGCAAG
ACGATCGATGCTAGCTAGCAT
AGCTAGCTAGTCAGCTAGCTAGTG
```

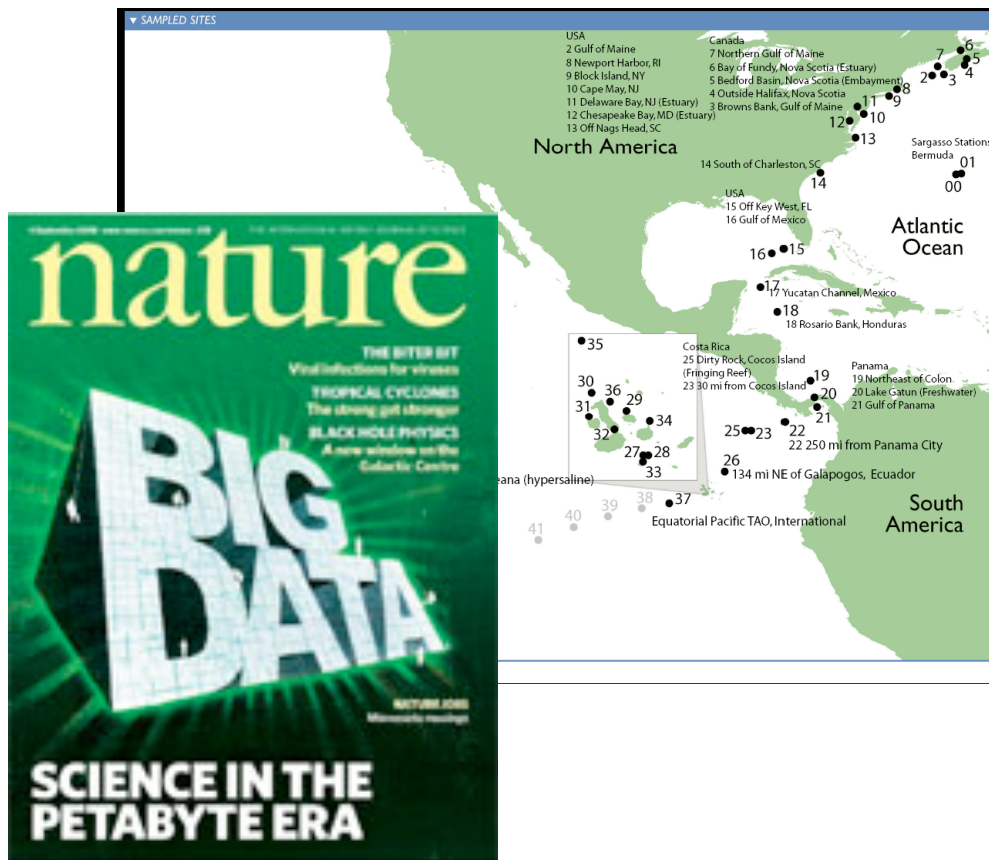
Partially Assemble and Annotate



PROBLEM: Lose information about which gene belongs to which microbe.



# Global Ocean Survey Statistics (GOS)



6.25 GB of data  
7.7M Reads  
1 million CPU hours  
to process

Rusch, et al., PLOS Biology 2007



## Pathway Sequences (Community Function)

## Environmental Features

Metabolic Pathways

	P1	P2	P3		
Sites B1	3800	1400	1000		
B2	2200	100	400		
↓	---	---	---		



Environmental Metadata

	Temp	NaCl	Depth		
Sites B1	15°C	27.2	10 m		
B2	23°C	36.6	5 m		
↓	---	---	---		

READS → PROTEIN FAMILIES → PATHWAYS

CCGTGAGCACGATGCGC-----  
 ATGCTCATGCT-----  
 ATCGTGACGCGATGC-----  
 CCGTGAGCACGATGCGC-----  
 ATGCTCATGCT-----  
 ATCGTGACGCGATGC-----  
 ATGCTCATGCT-----  
 GCGATCGATCGATCGTAGC-----  
 TGCTGCTAGCATGCT-----  
 GCGATCGATCGATCGTAGC-----  
 TGCTGCTAGCATGCT-----  
 CCGTGAGCACGATGCGC-----  
 GTATCGTAGCATGCTT-----  
 CCGTGAGCACGATGCGC-----  
 GCGATCGATCGATCGTAGC-----



$$P_1 = f_1 + f_2 + f_3$$

$$P_2 = f_4 + f_5 + f_6$$

PATHWAYS

SITES	$P_{1,1} = 2 + 1 + 3$ $P_{1,2} = 5 + 2 + 6$	$P_{2,1} = 2 + 4 + 3$ $P_{2,1} = 5 + 7 + 6$

**Expressing**  
**data as**  
**matrices**  
**indexed by**  
**site, env. var.,**  
**and pathway**  
**usage**

[Rusch et. al., (2007) PLOS Biology;  
 Gianoulis et al., PNAS (in press, 2009)]

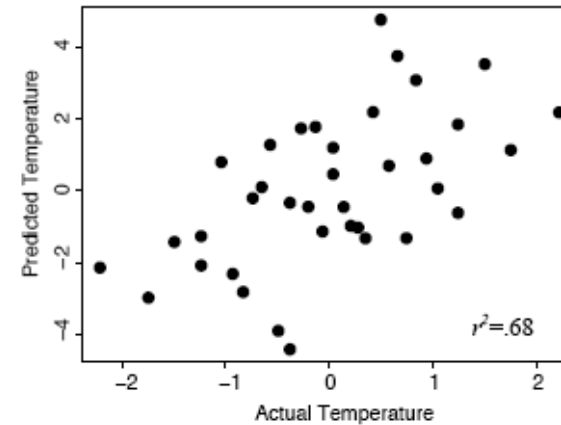
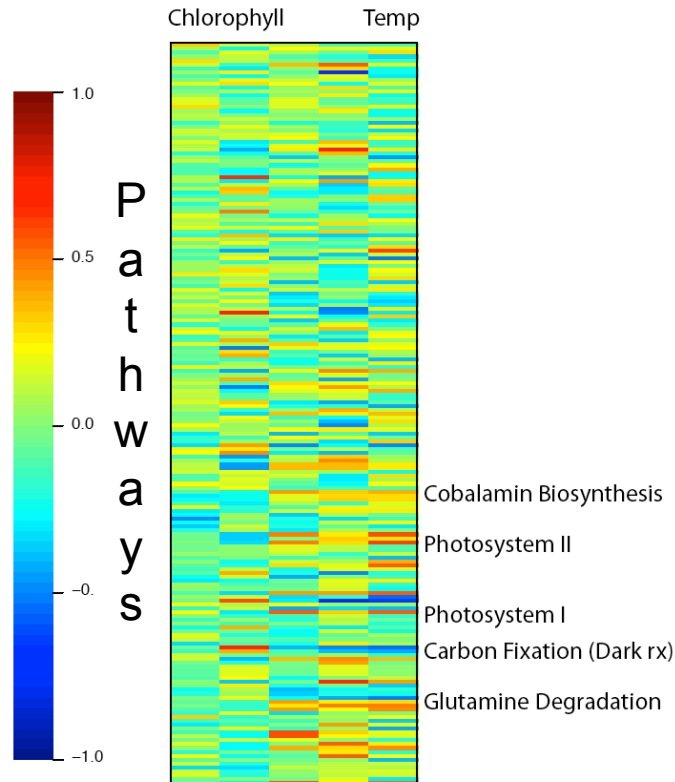


# Simple Relationships: Pairwise Correlations




Environmental Features



[Gianoulis et al., PNAS (in press, 2009)]





# Canonical Correlation Analysis: Simultaneous weighting

Score	# of papers published
GRE	

Undergraduate Performance Index (UPI)	Graduate School Performance Index (GPI)
GRE 	

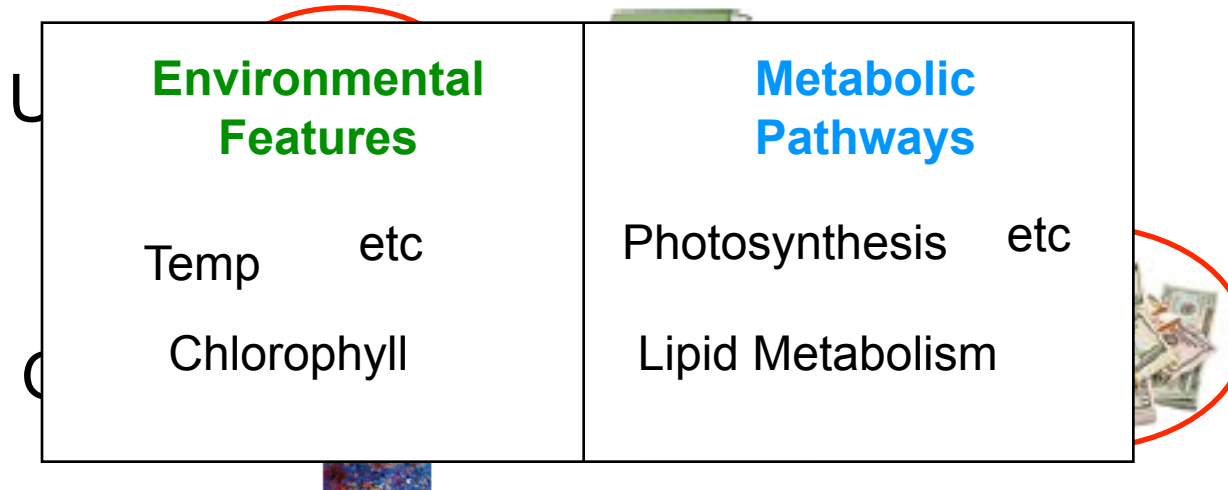
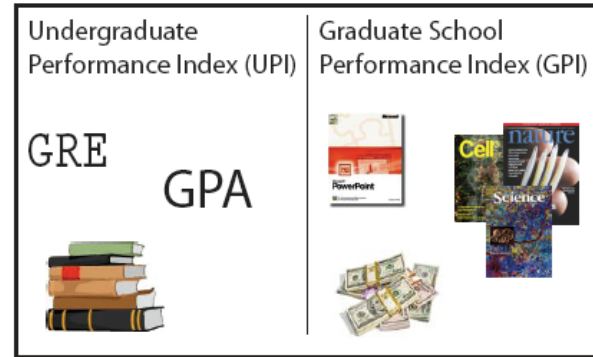
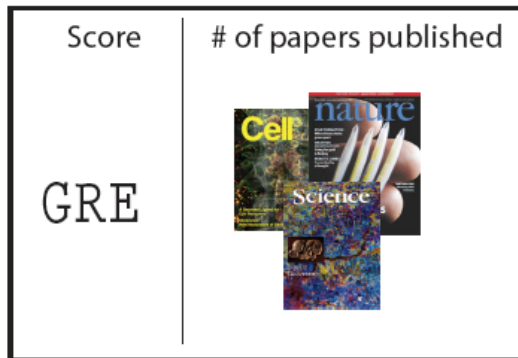
$$\text{UPI} = a \text{ GRE} + b \text{ GPA}$$

$$\text{GPI} = a' \text{ GRE} + b' \text{ GPA} + c' \text{ Money}$$

[ Gianoulis et al., PNAS (in press, 2009) ]



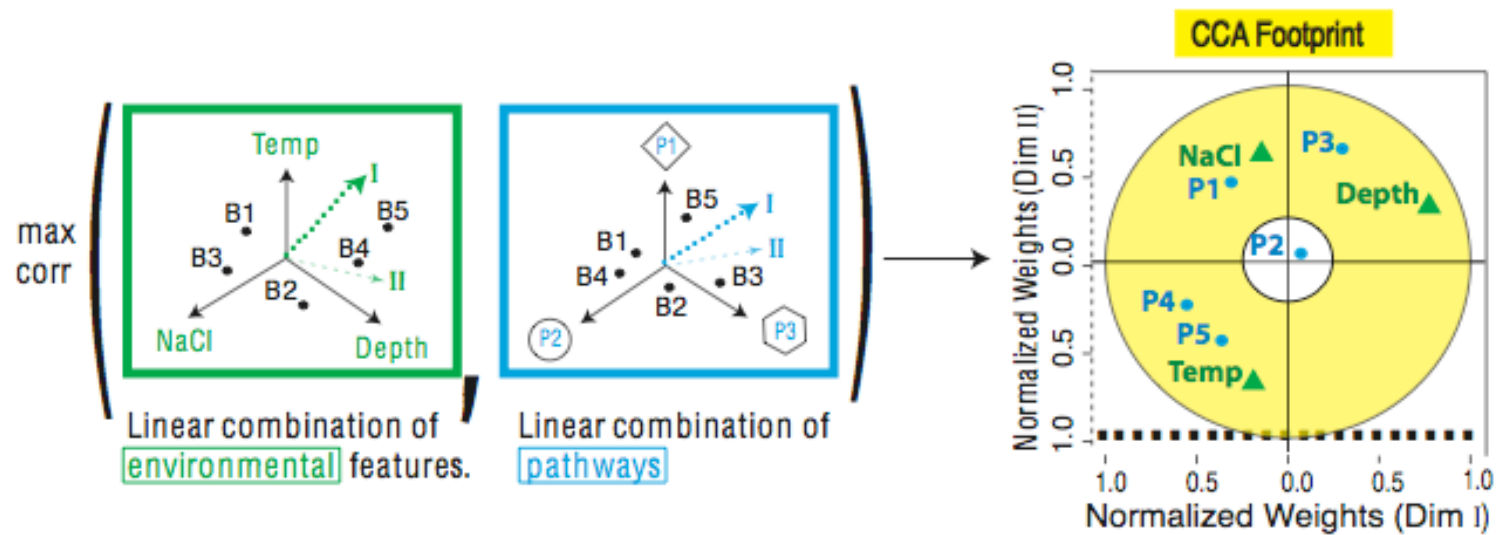
# Canonical Correlation Analysis: Simultaneous weighting



[ Gianoulis et al., PNAS (in press, 2009) ]



# Environmental-Metabolic Space



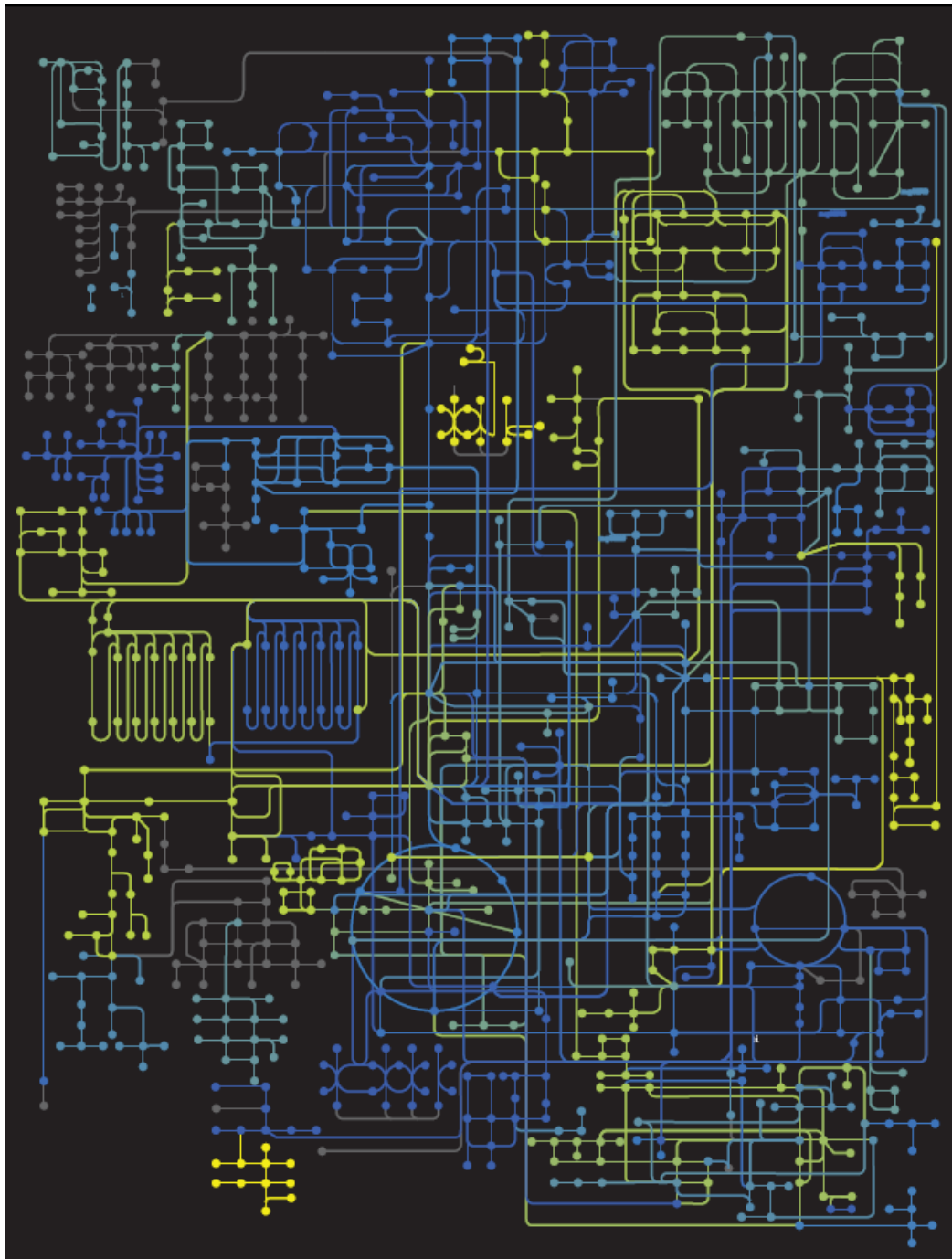
The goal of this technique is to interpret cross-variance matrices  
We do this by defining a change of basis.

Given  $X = \{x_1, x_2, \dots, x_n\}$  and  $Y = \{y_1, y_2, \dots, y_m\}$

$$C = \begin{matrix} \sum_X & \sum_{X,Y} \\ \sum_Y & \sum_{Y,X} \end{matrix} \quad \max_{a,b} \text{Corr}(U,V) = \frac{a' \sum_{12} b}{\sqrt{a' \sum_{11} a} \sqrt{b' \sum_{22} b}}$$

[ Gianoulis et al., PNAS (in press, 2009) ]



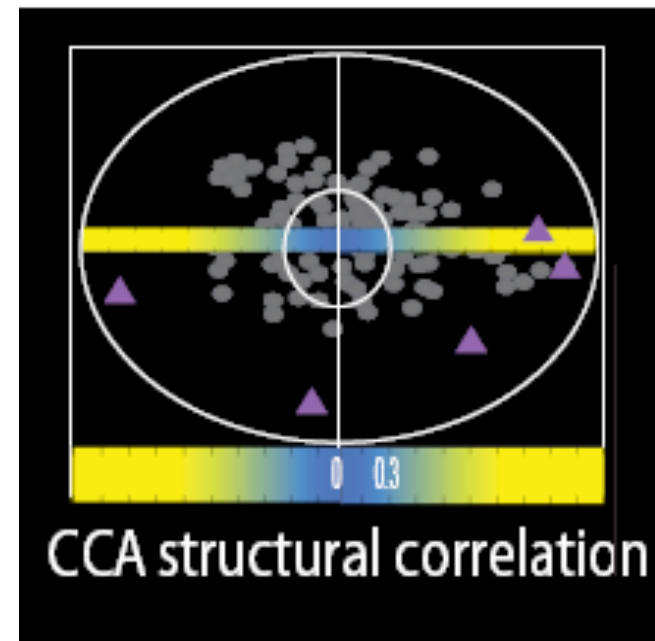


## Strength of Pathway co-variation with environment



Environmentally  
invariant

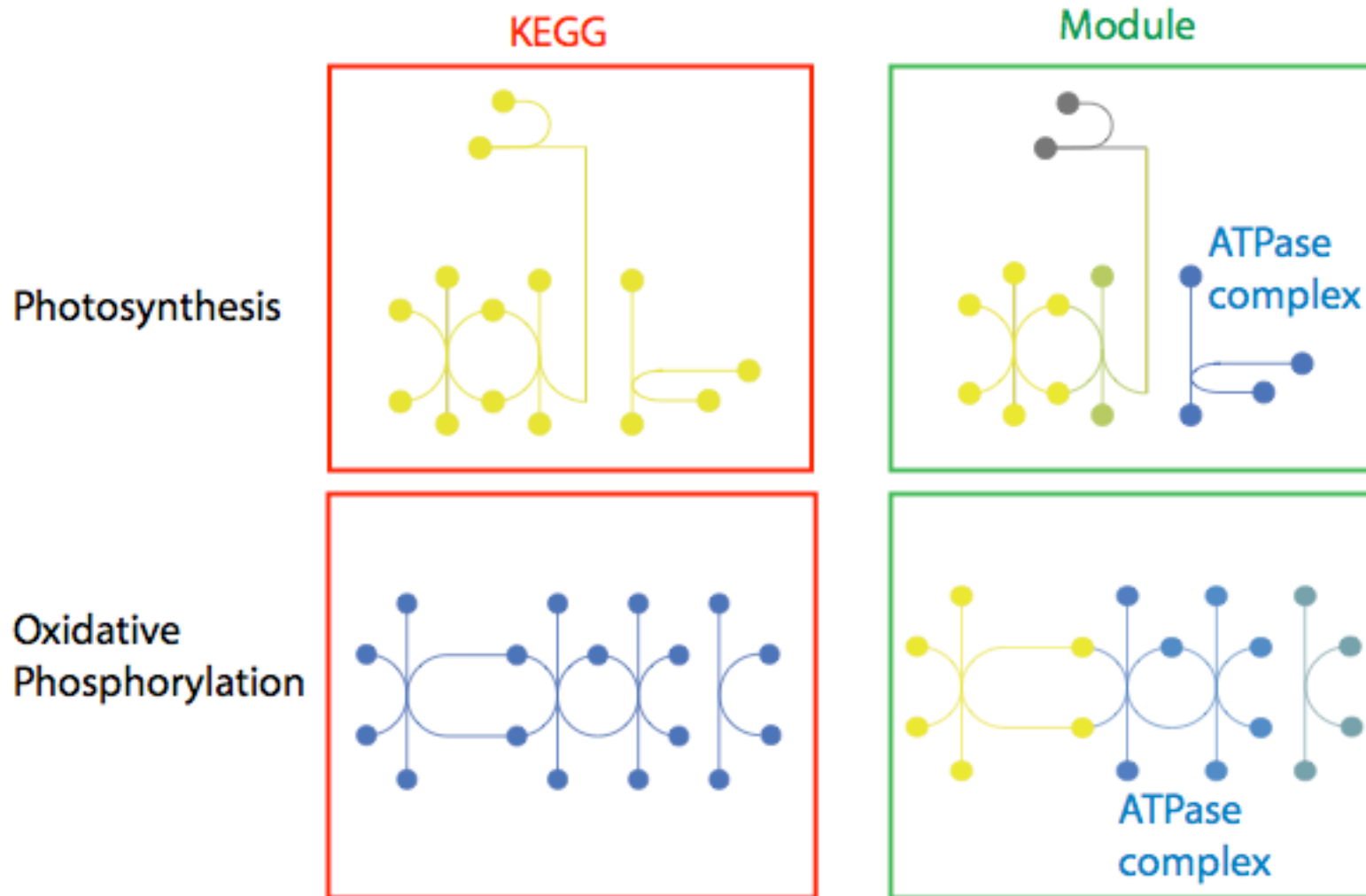
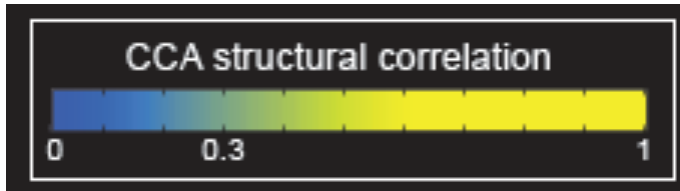
Environmentally  
variant



[ Gianoulis et al., PNAS (in press, 2009) ]

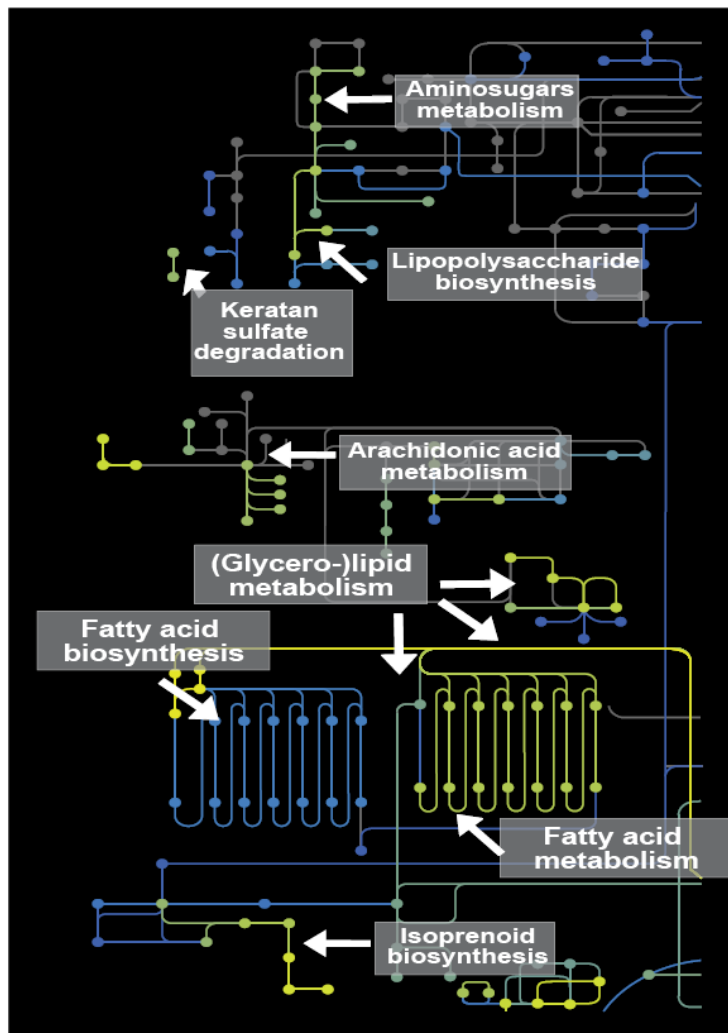


# Conclusion #1: energy conversion strategy, temp and depth





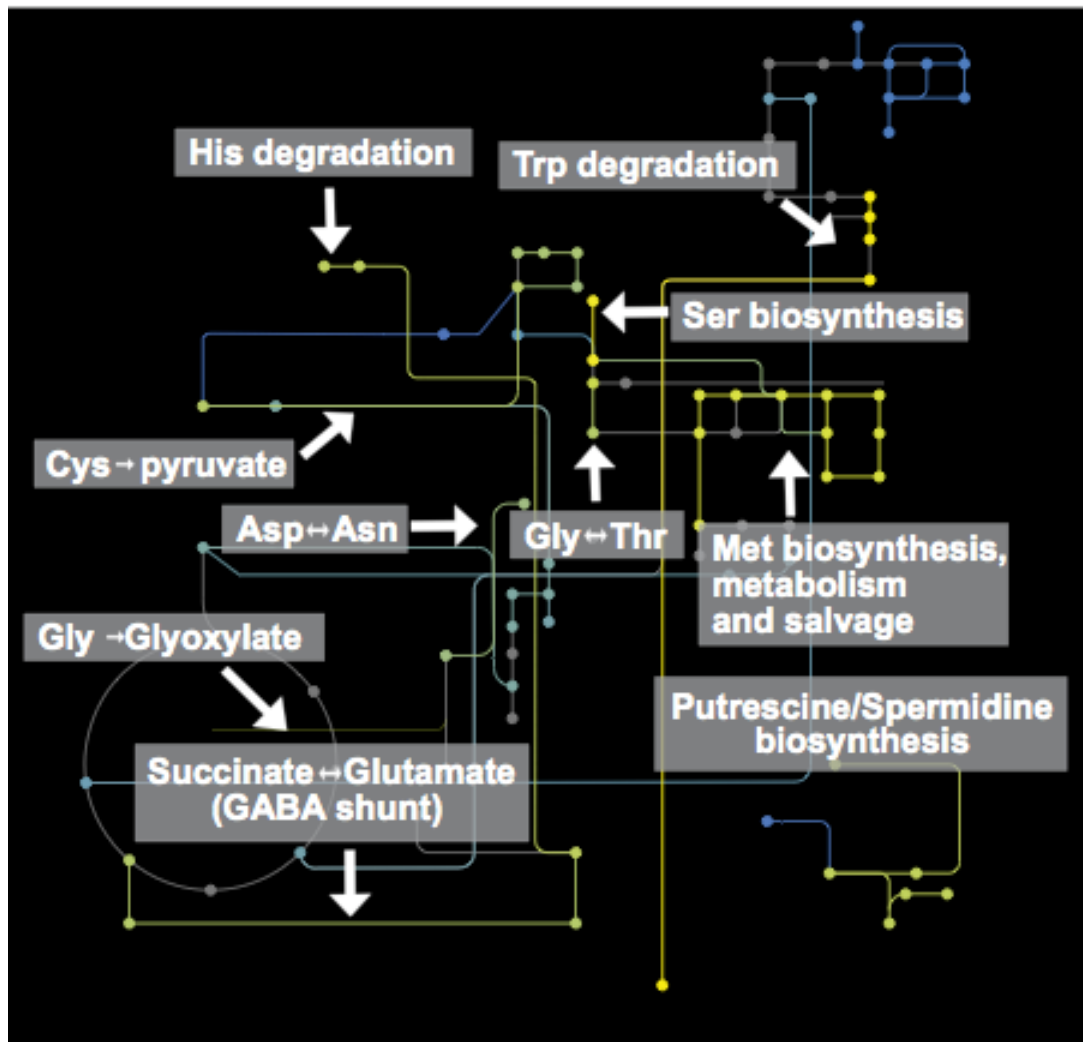
## Conclusion #2: Outer Membrane components vary the environment



[ Gianoulis et al., PNAS (in press, 2009) ]



# Conclusion #3: Covariation of AA biosynthesis and Import

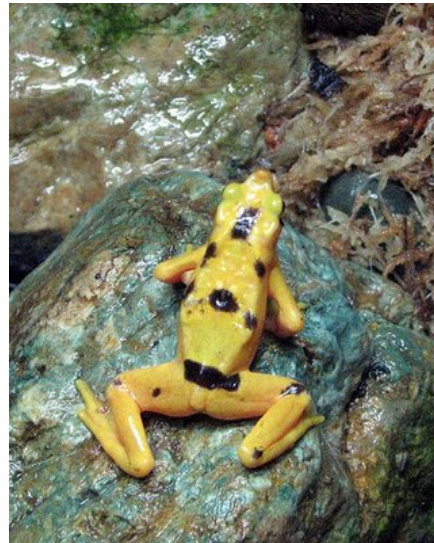


Why is their fluctuation in amino acid metabolism? Is there a feature(s) that underlies those that are environmentally-variant as opposed to those which are not?

[ Gianoulis et al., PNAS (in press, 2009) ]



# Biosensors: Beyond Canaries in a Coal Mine



[ Gianoulis et al., PNAS (in press, 2009) ]



# Conclusions



- Developing Standardized Descriptions of Protein Function
- Gene Naming
- Betweenness is an important global network statistic
  - ◇ Bottlenecks are more correlated with essentiality than hubs in regulatory networks
- Regulatory Network Hierarchies
  - ◇ Middle managers dominate, sitting at info. flow bottlenecks
  - ◇ Paradox of influence and essentiality
  - ◇ Topmost proteins sit at center of interaction network



# Conclusions: Networks Dynamics across Environments



- Developed and adapted techniques to connect quantitative features of environment to metabolism.
- Applied to available aquatic datasets, we identified footprints that were predictive of their environment (potentially could be used as biosensor).
- Strong correlation exists between a community's energy conversion strategies and its environmental parameters (e.g. temperature and chlorophyll).
- Suggest that limiting amounts of cofactor can (partially) explain increased import of amino acids in nutrient-limited conditions.





- an automated web tool

**tYNA**

(vers. 2 :

**"TopNet-like**

**Yale Network Analyzer")**

**tYNA**

Getting started API WSDL Download tYNA Installation guide Plugins for Cytoscape Contact Known problems

You are logged in as kevin. [Logout](#) View: Simple Advanced

List Owned Biological networks with (Attribute name) = (Attribute value) List

**Workspace manager**

Load an existing network

Load: [14. Uetz 2000 yeast two ...]  
 Into: [workspace 0]  
 Categorized by: [Nil]  
 Load

Current working networks in your workspaces:

Workspace 0: statFilter(degrees, geq, 1, value, neighbors=false, intersection("Uetz 2000 yeast two hybrid", "Ito 2001 yeast two hybrid"))  
 Workspace 1: (empty)  
 Workspace 2: (empty)  
 Workspace 3: (empty)

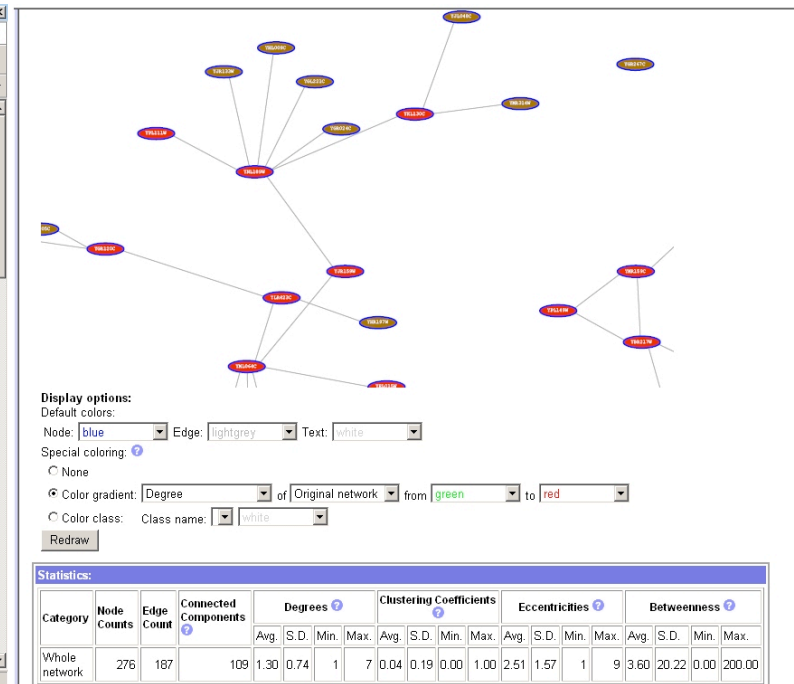
**Multiple network analysis**

**Networks in database (upload download)**

ID	Name	Creator	Creation date	
14	Uetz 2000 yeast two hybrid	kevin	21-Feb-06	<a href="#">Delete</a>
15	Ito 2001 yeast two hybrid	kevin	21-Feb-06	<a href="#">Delete</a>
16	Ho 2002 pull down	kevin	21-Feb-06	<a href="#">Delete</a>
17	Gavin 2002 pull down	kevin	21-Feb-06	<a href="#">Delete</a>
18	Jansen 2003 PIT	kevin	21-Feb-06	<a href="#">Delete</a>
19	MIPS yeast PPI	kevin	21-Feb-06	<a href="#">Delete</a>
21	BIND yeast data	kevin	21-Feb-06	<a href="#">Delete</a>
22	DIP yeast data	kevin	21-Feb-06	<a href="#">Delete</a>
23	Kim 2006 structural interaction	kevin	21-Feb-06	<a href="#">Delete</a>
24	Han 2004 FYI data	kevin	21-Feb-06	<a href="#">Delete</a>
25	Luscombe 2004 regulatory	kevin	21-Feb-06	<a href="#">Delete</a>

**Categories in database (upload download)**

ID	Name	Creator	Creation date
----	------	---------	---------------



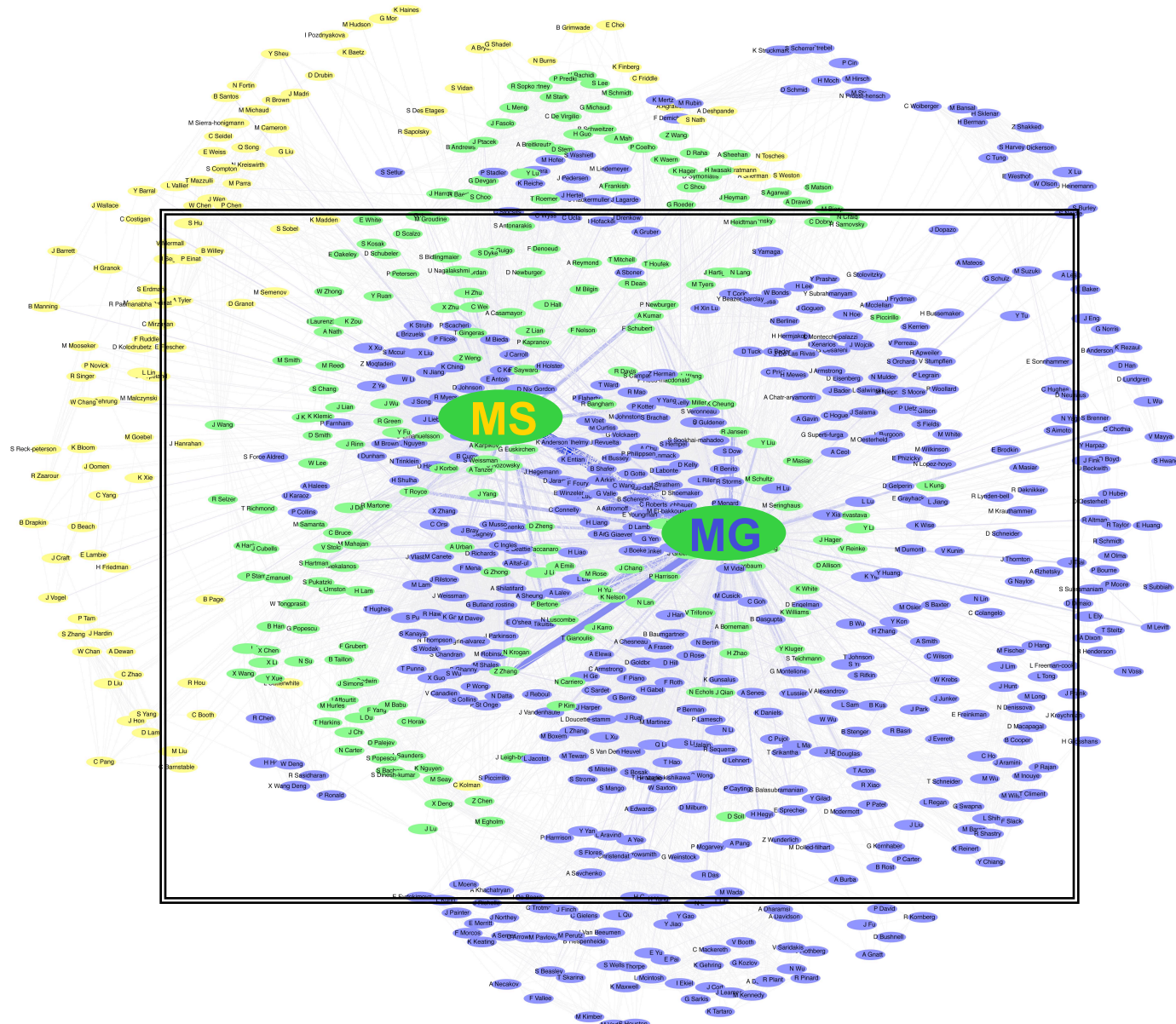
Normal website + Downloaded code (JAVA)  
 + Web service (SOAP) with Cytoscape plugin

[Yu et al., NAR (2004); Yip et al. Bioinfo. (2006);  
 Similar tools include Cytoscape.org, Idekar, Sander et al]



# Acknowledgements

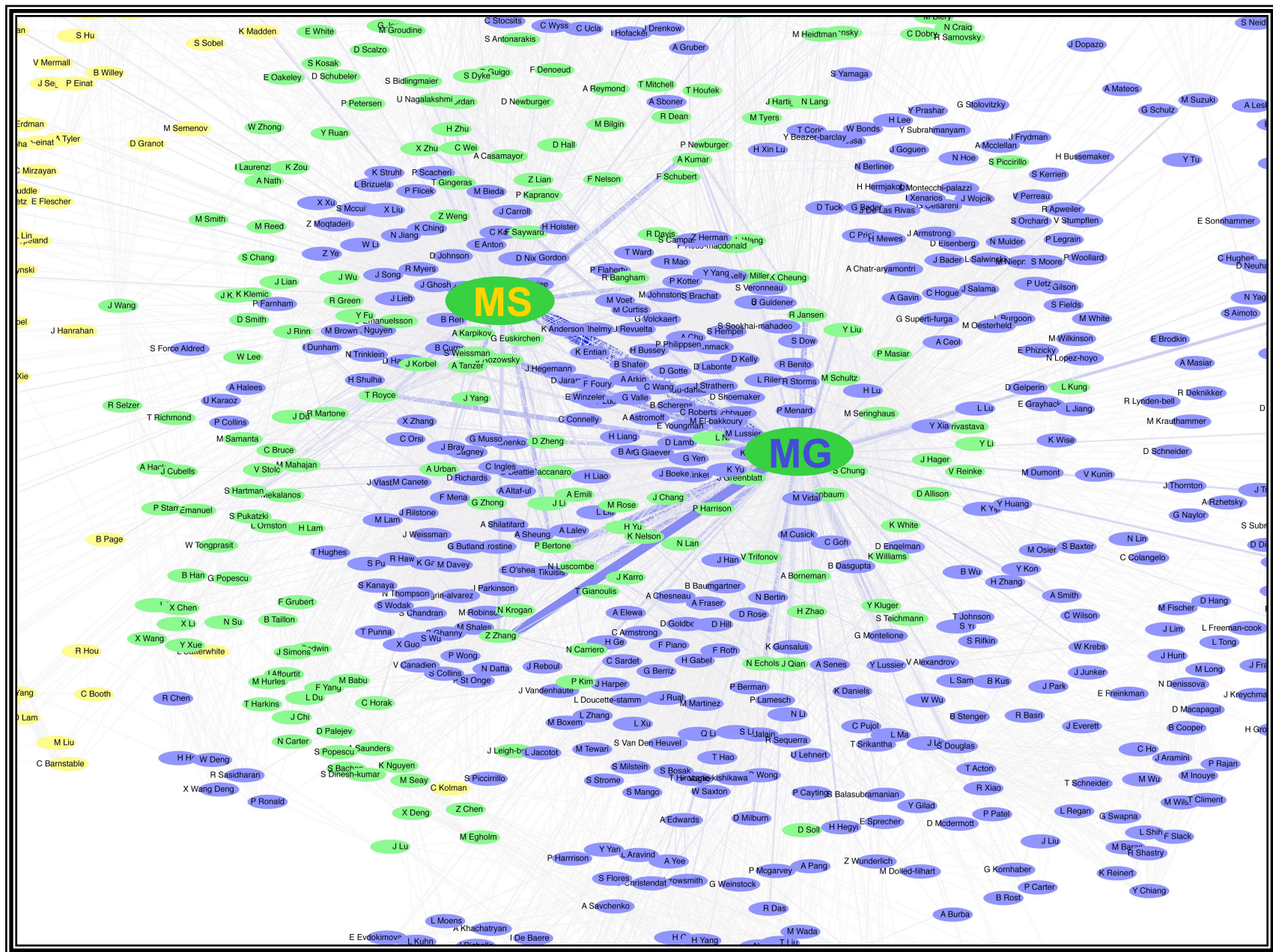
## TopNet.GersteinLab.org





# Acknowledgements

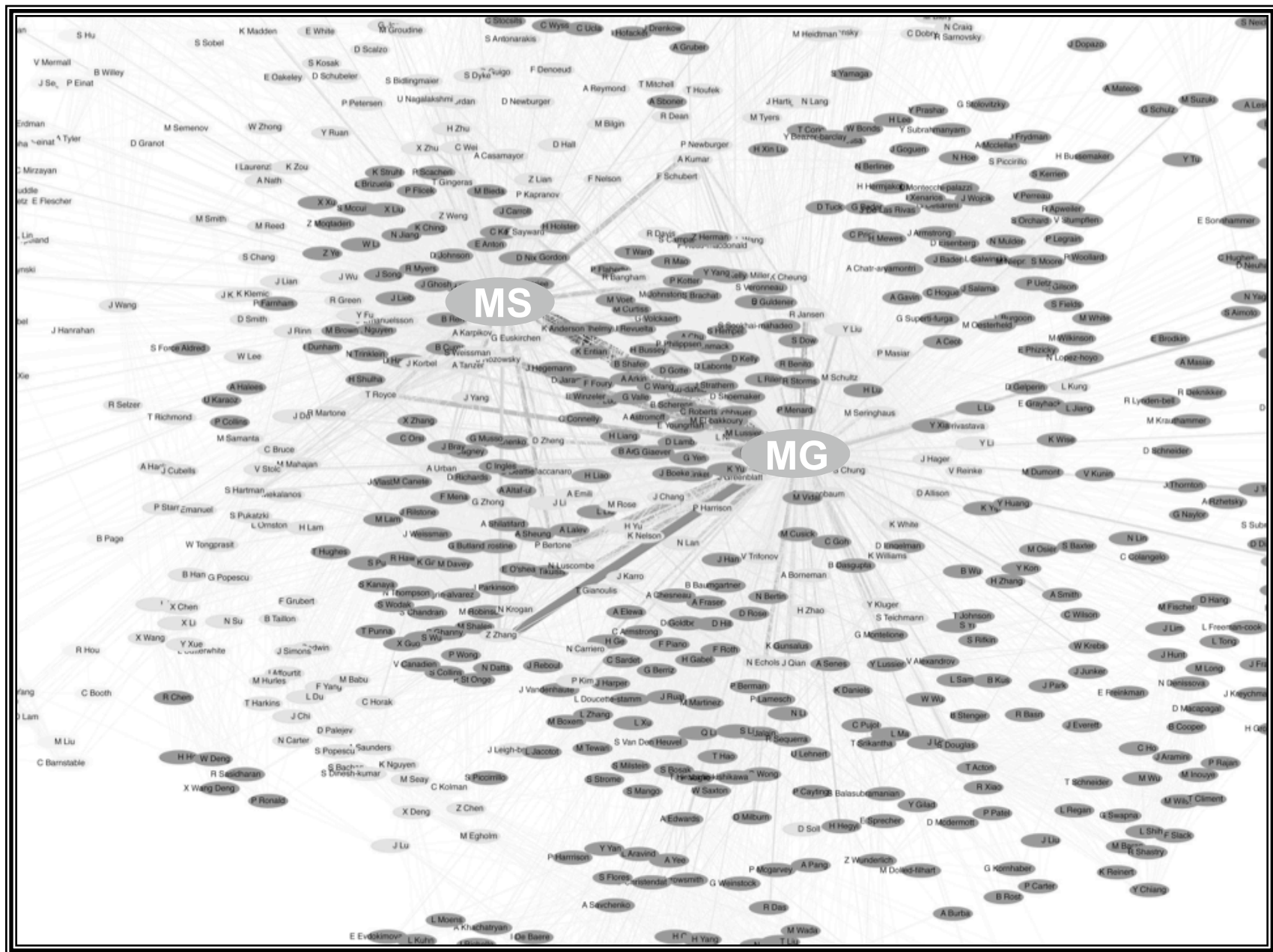
## TopNet.GersteinLab.org





# Acknowledgements

## TopNet.GersteinLab.org

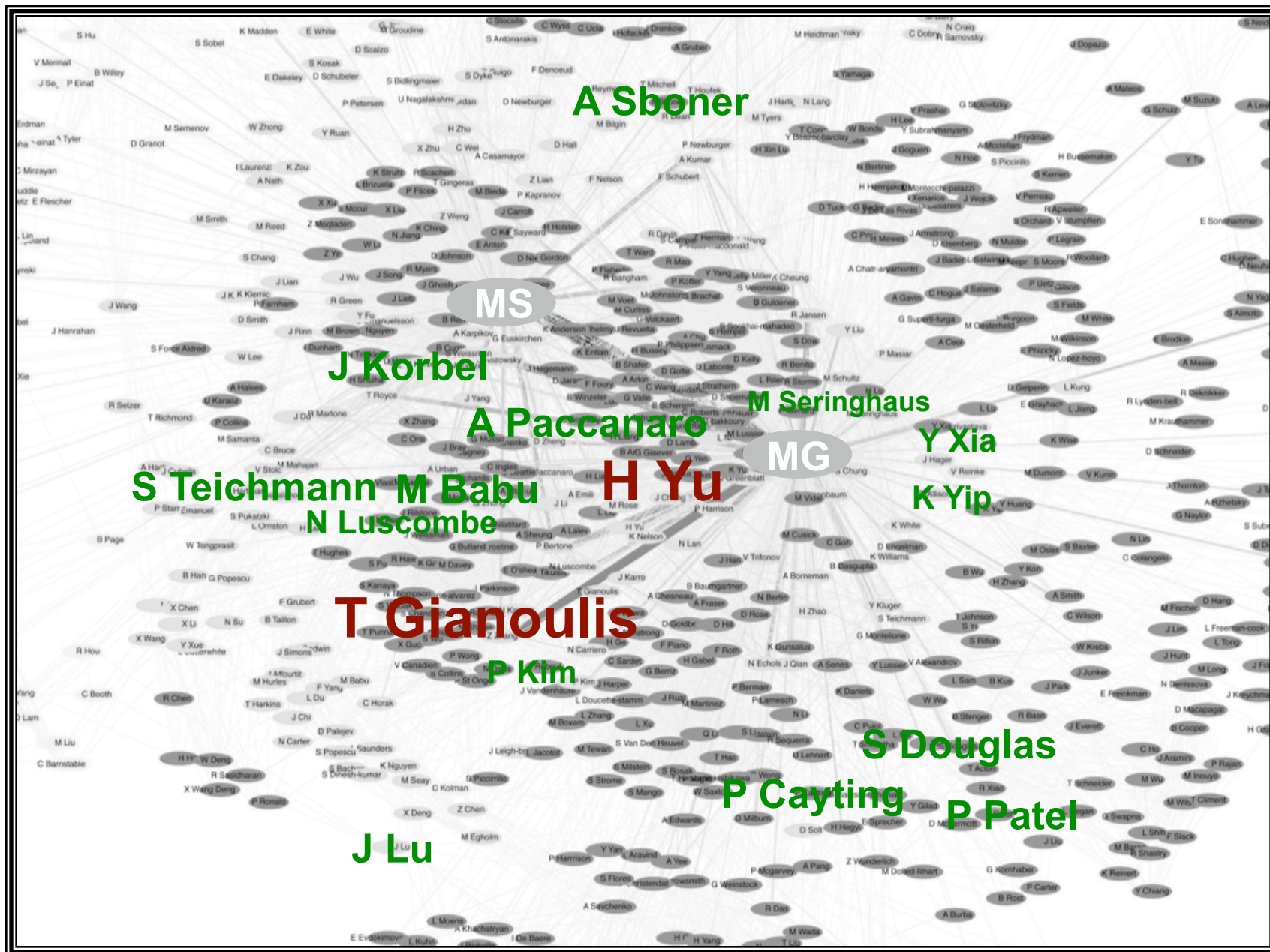




P Bork, J Raes

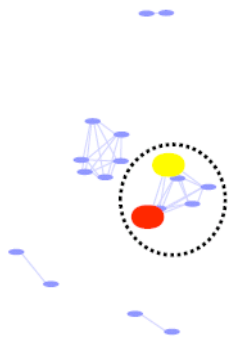
# Acknowledgements TopNet.GersteinLab.org

Job opportunities currently  
for postdocs & students

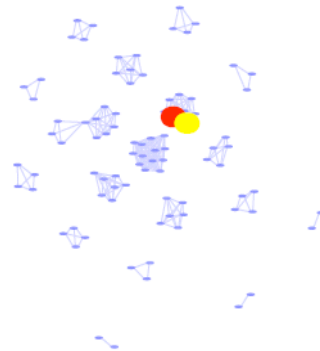




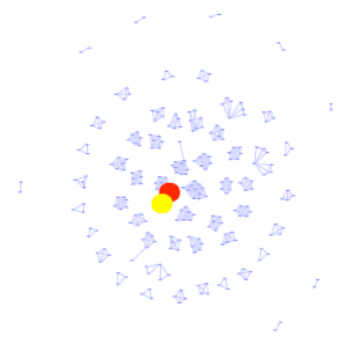
# RNAi: Birth of a Field in the Literature Culmin -ating in the 2006 Nobel



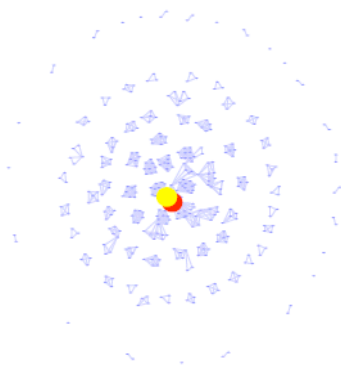
1998



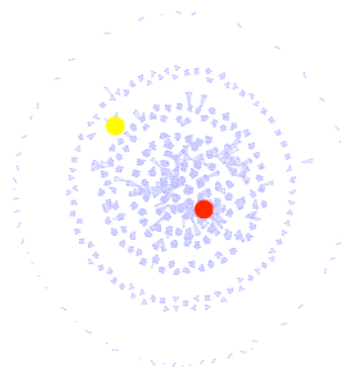
1999



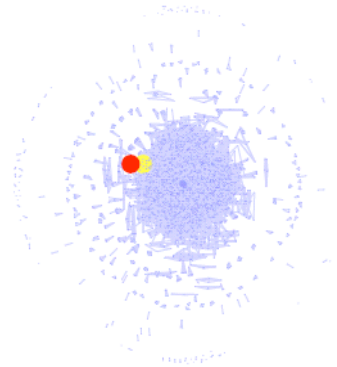
2000



2001



2002



2003

● Andrew Fire      ● Craig Mello

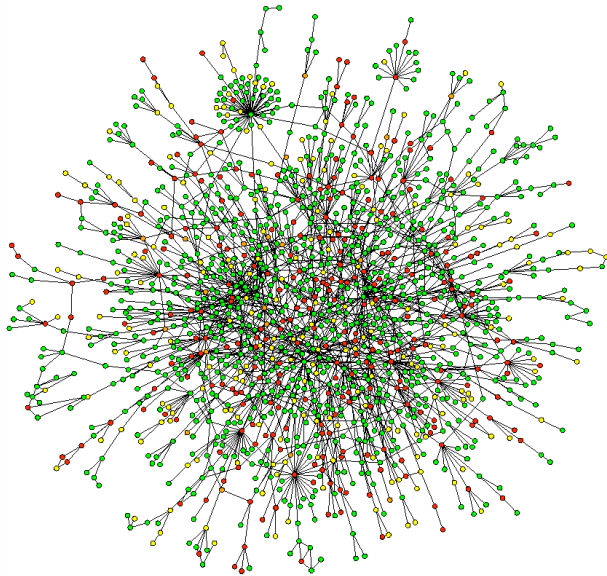
Source:  
Gerstein & Douglas.  
PLoS Comp. Bio. 3:e80  
(2007)  
PubNet.GersteinLab.org



# Extra



# Types of Networks



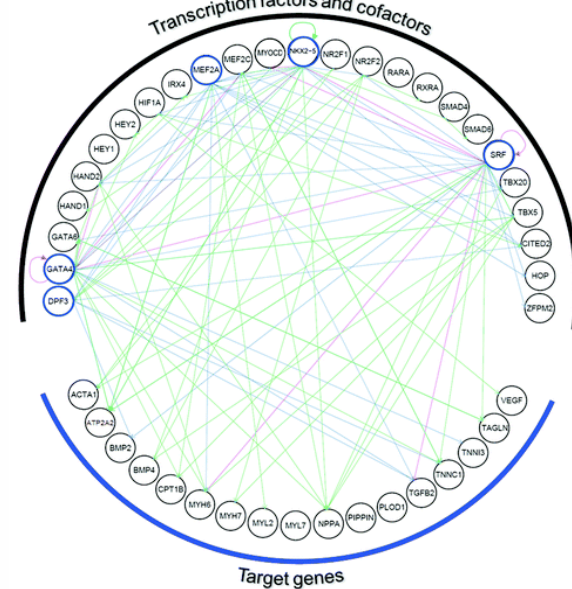
## Interaction networks

**Nodes: proteins or genes**  
**Edges: interactions**

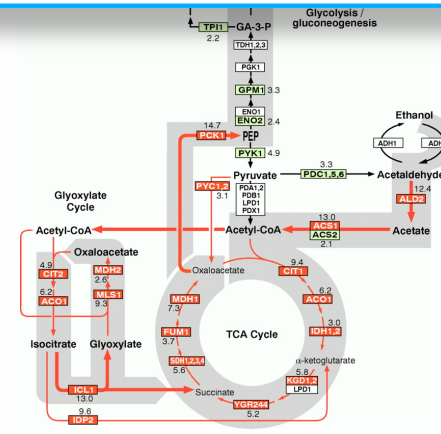
[Horak, et al, Genes & Development, 16:3017-3033]

[DeRisi, Iyer, and Brown, *Science*, 278:680-686]

[Jeong et al, Nature, 41:411]



## Regulatory networks



## Metabolic networks



# More Information on this Talk

**TITLE:** Understanding Protein Function on a Genome-scale through the Analysis of Molecular Networks

**SUBJECT:** Networks

**DESCRIPTION:**

National Academy of Engineering, Meeting at Columbia U, 2009.04.14, 14:00-14:30; [I:**NAECU**] (Short networks talk, incl. the following topics:

why networks w. **amsci\***, **funnygene\***, **bottleneck\***, **nethierarchy\***, **metagenomics\***, **tyna\* + topnet\***, & **pubnet\*** . Fits into 30' w. 5' questions. PPT works on mac & PC and has many photos w. EXIF tag **kwtimewemet** .)

(Paper references in the talk were mostly from Papers.GersteinLab.org. The above topic list can be easily cross-referenced against this website. Each topic abbrev. which is starred is actually a papers "ID" on the site. For instance,

the topic **pubnet\*** can be looked up at <http://papers.gersteinlab.org/papers/pubnet> )

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