

COMPRESSIBLE COMPONENTS REVEAL NETWORK ARCHITECTURES

CABDyN Seminar, University of Oxford
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Cavendish Laboratory

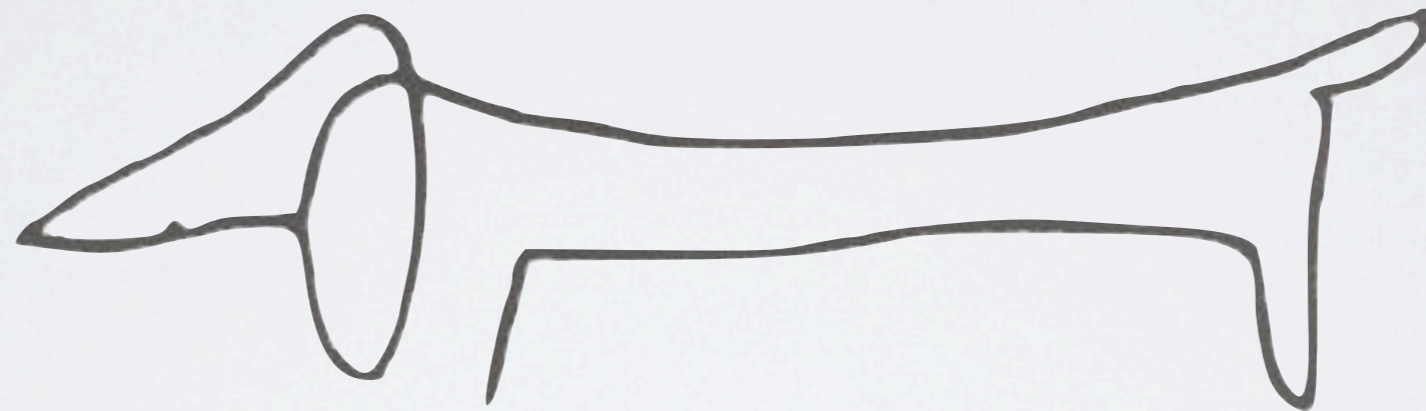


**UNIVERSITY OF
CAMBRIDGE**

WHAT IS A SKETCH?

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A sketch concisely captures dominant features.



1. 2. 3.

A NETWORK SKETCH?

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How can we draw a sketch of a network?

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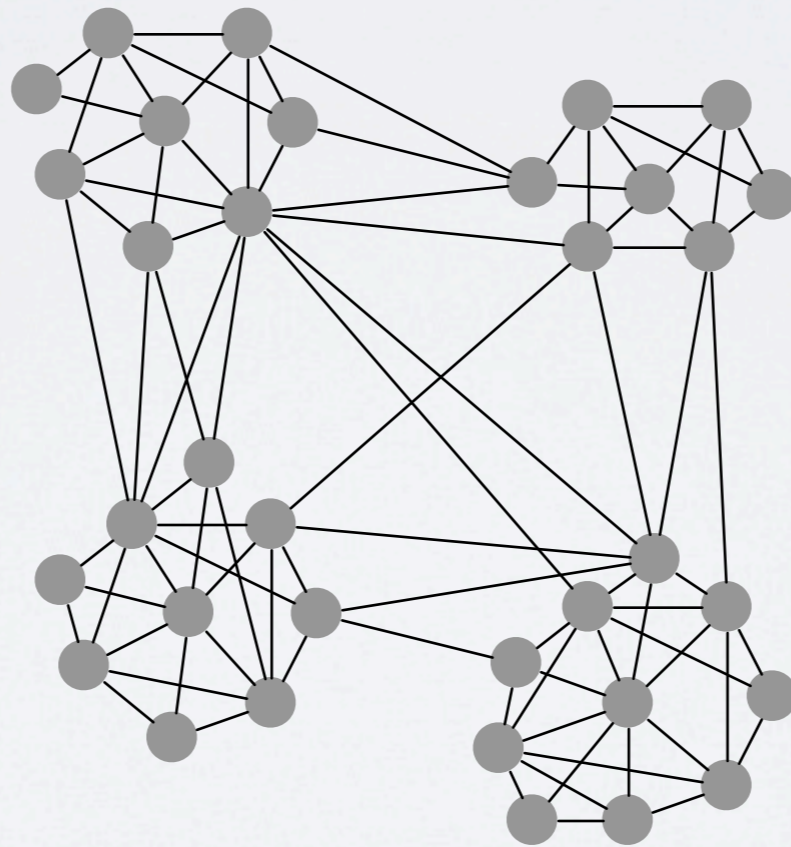
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What are dominant structural features?

Communities?

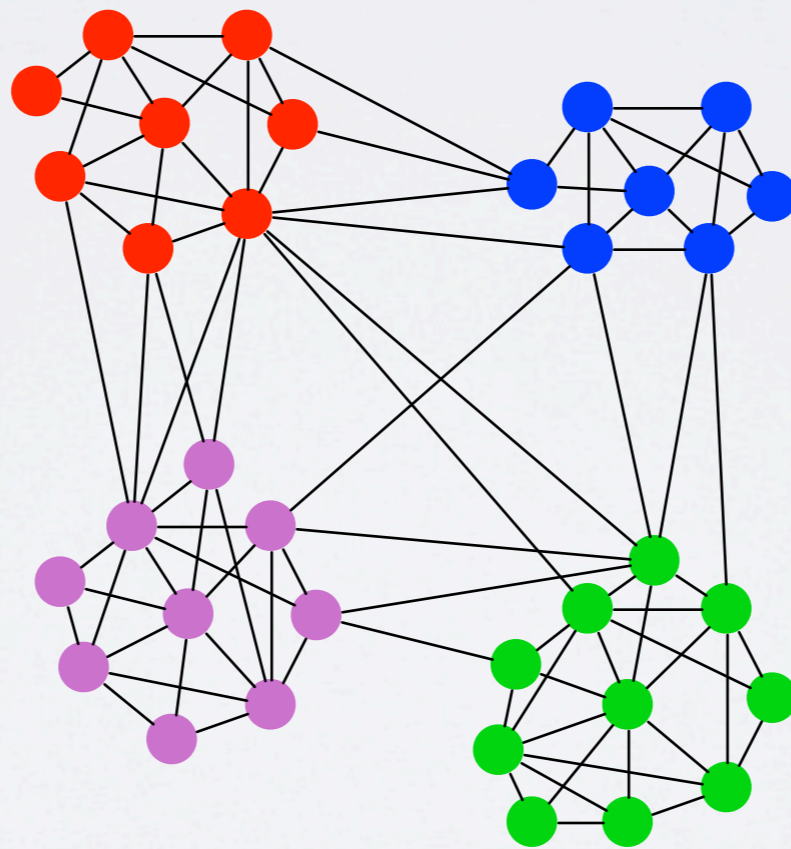
COMMUNITY DETECTION

Communities in networks are typically defined as densely interconnected subsets of nodes.



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COMMUNITY DETECTION

One of the best known methods of this kind is Newman & Girvan's [PRE 69, 026113 (2004)]:

- 1) calculate betweenness for all edges
- 2) remove highest betweenness edge
- 3) recalculate betweenness for all edges and go to 2)

This fragments the network and thus gives rise to a series of partitions.

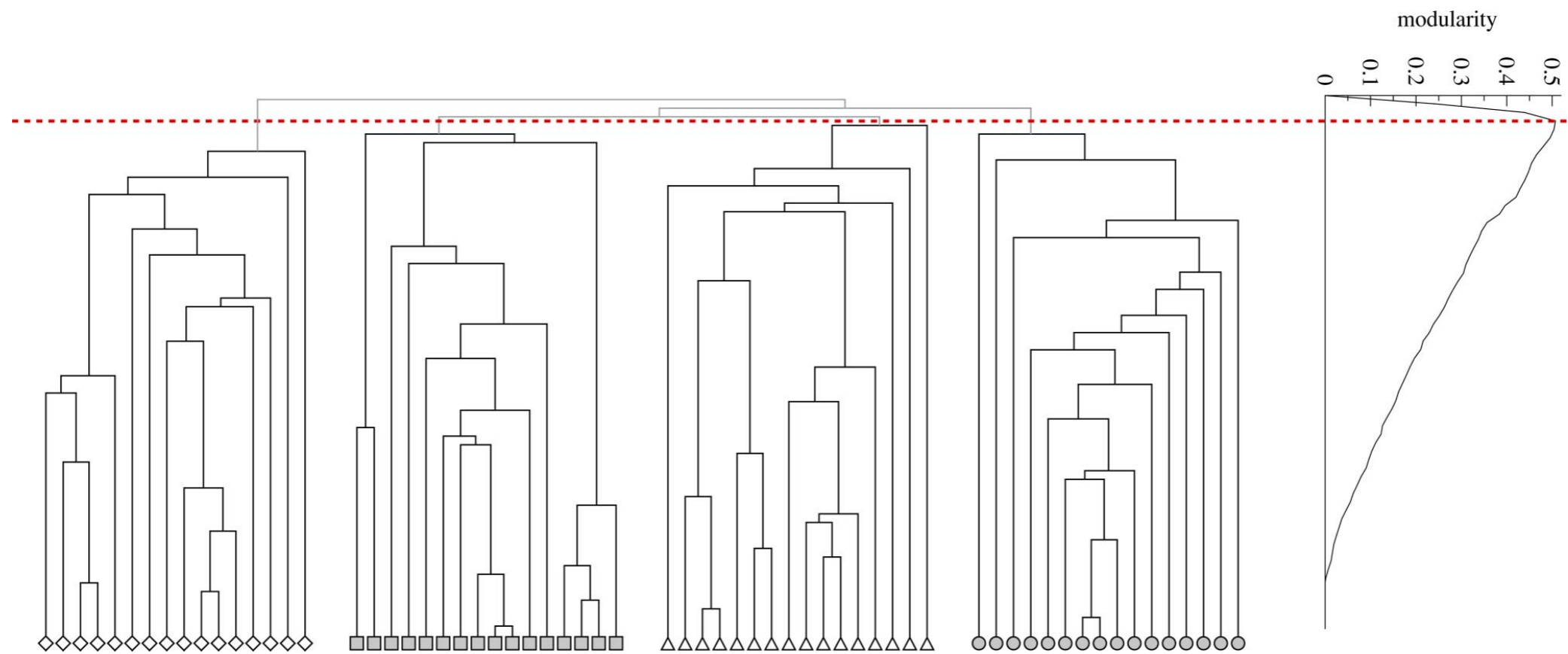
COMMUNITY DETECTION

One then chooses the optimal partition by maximising *modularity*, which is defined as:

$$Q = \sum_i (e_{ii} - a_i^2)$$

where e_{ii} is the fraction of network edges *within* partition i , and a_i is the fraction of edges *connecting to* partition i .

COMMUNITY DETECTION



THE PROBLEMS

Most existing community detection methods:

- are **node-based**
- attempt to **partition** the nodes
- look at connectivity **within** a community.

THE PROBLEMS

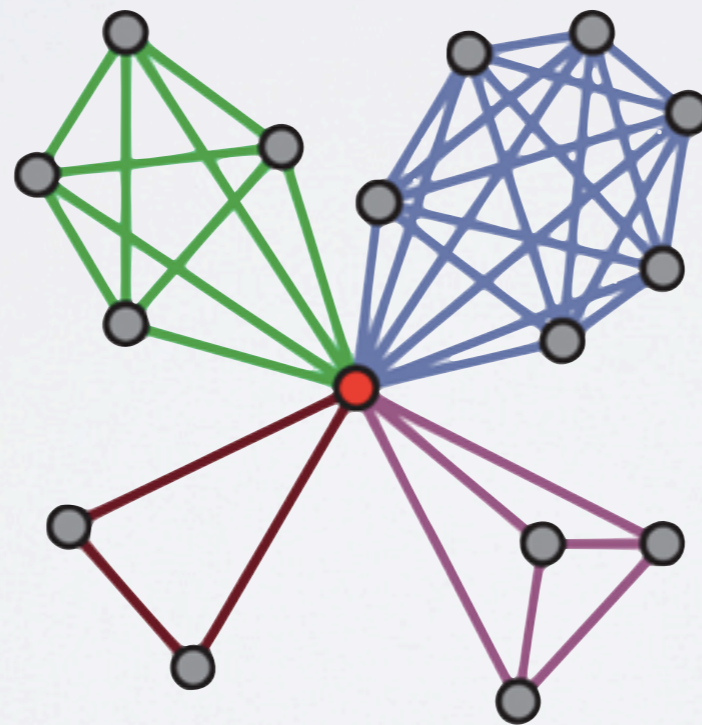
Most existing community detection methods:

- are **node-based**
- attempt to **partition** the nodes
- look at connectivity **within** a community.

This means that they are likely to overlook highly overlapping communities, and communities that are defined by relationships of one node set to another.

EDGE-BASED COMMUNITIES

Recent work by Ahn, Bagrow & Lehmann introduced link communities to deal with the problem of overlapping communities.



Nature 466, 761-764 (2010)
(figure from paper)

EDGE-BASED COMMUNITIES

Their approach is to compare pairs of edges that share a node. If the two nodes at the other ends of these edges share many neighbours, the edges are similar.



$$S(e_{ik}, e_{jk}) = \frac{|n_+(i) \cap n_+(j)|}{|n_+(i) \cup n_+(j)|}$$

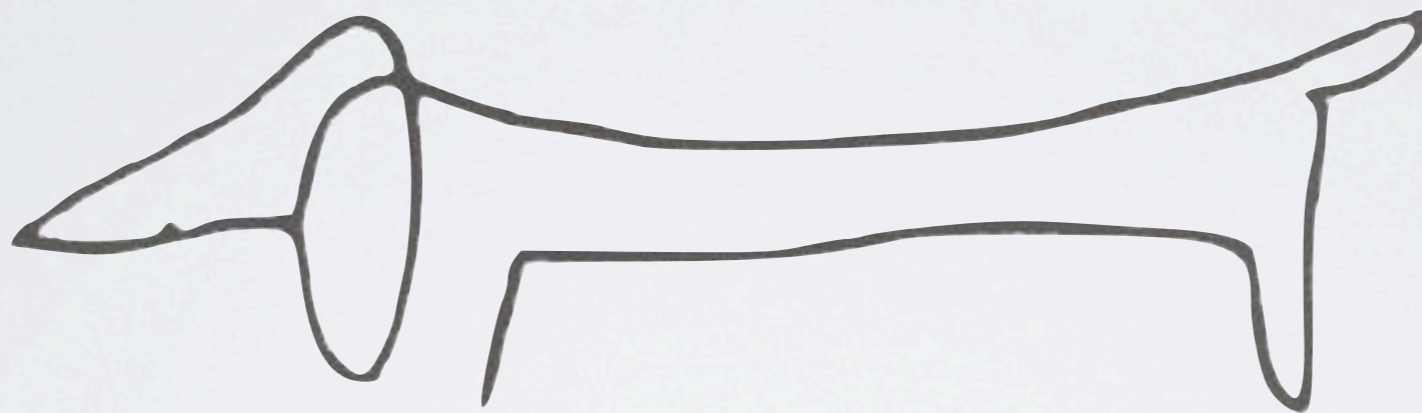
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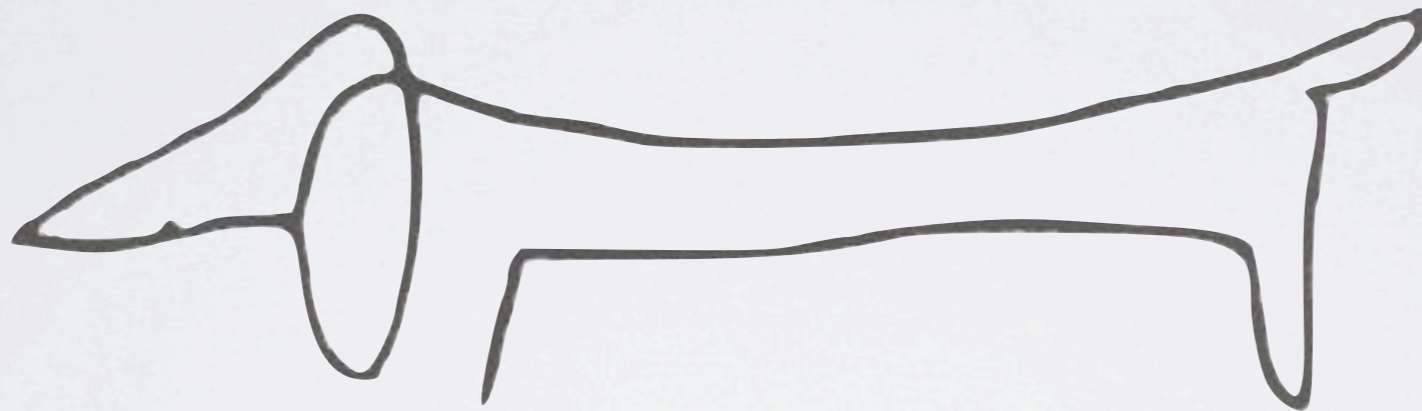
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WHAT IS A SKETCH?



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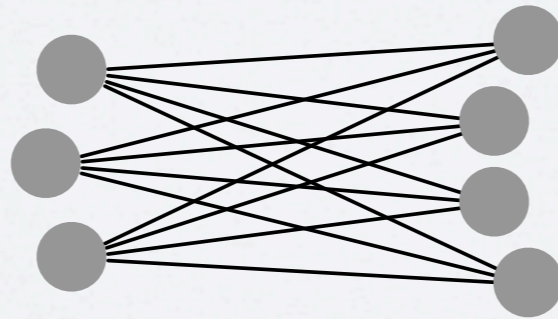
A sketch is a **compression**.

Sketch

POWER-EDGES

A power-edge is an edge between two sets of nodes, known as power-nodes. It signifies that all the nodes in one set are connected to all the nodes in the other set.

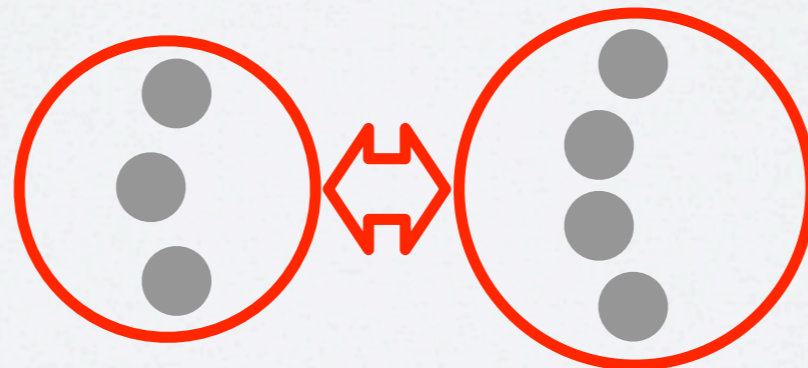
This does not mean that the nodes within a set have to be connected to each other.



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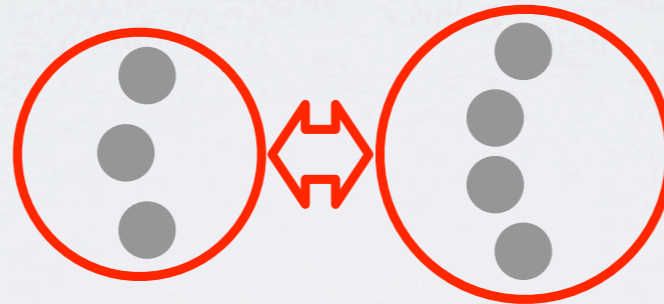
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COMPRESSING NETWORKS

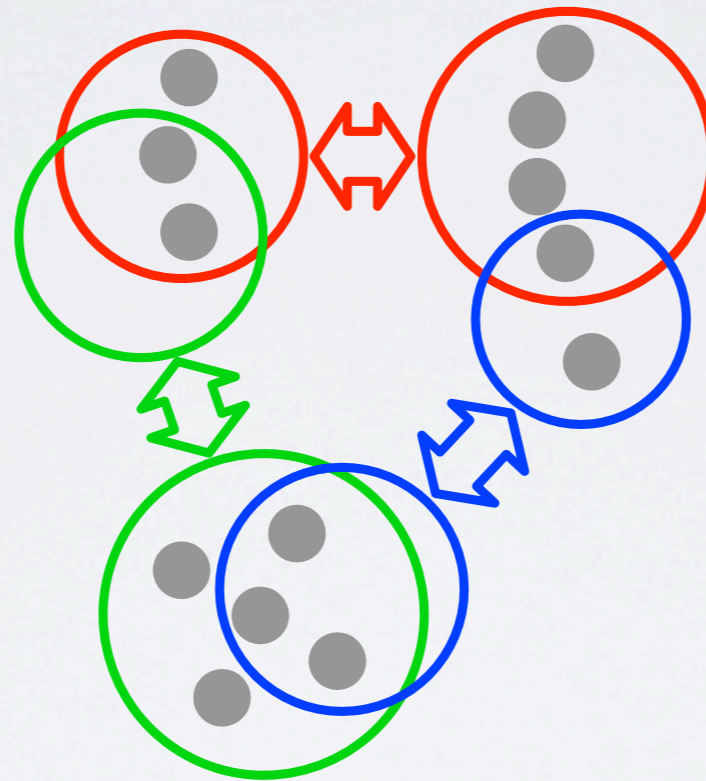
These power-nodes can overlap. Note that self-interactions are also possible.



How do we choose a suitable set of power-edges to represent the graph?

COMPRESSING NETWORKS

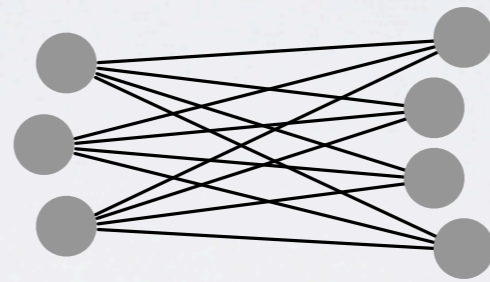
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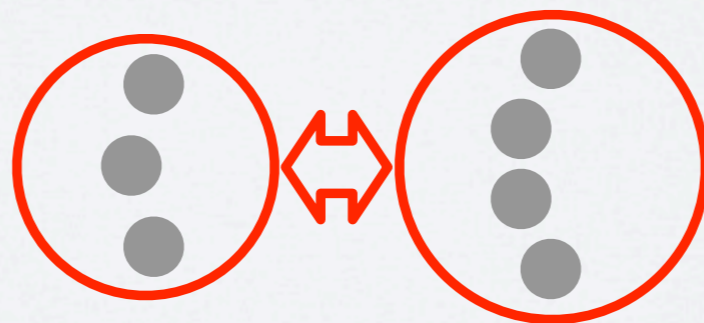
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COMPRESSIBLE COMPONENTS

Let us compare the original set of edges

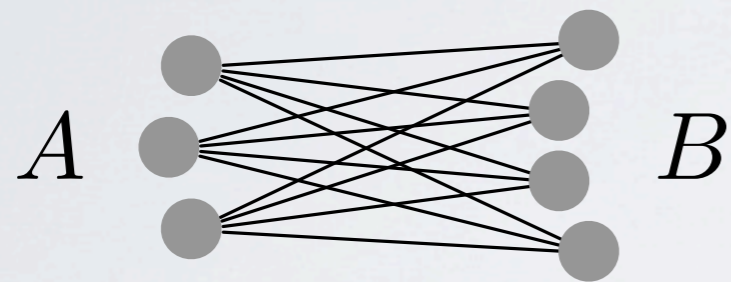


with the power-edge:

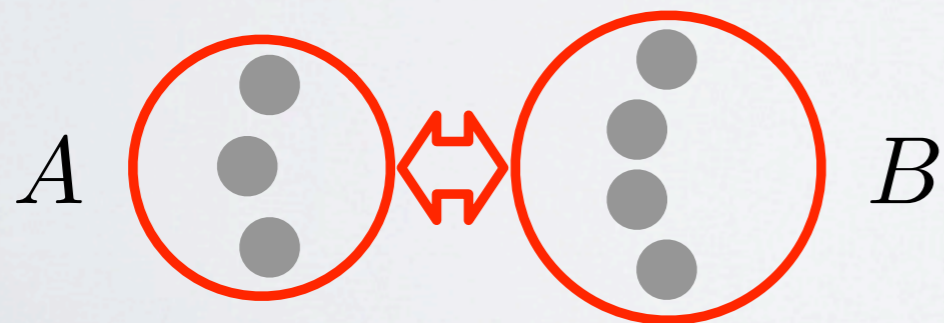


COMPRESSIBLE COMPONENTS

The information required to describe these is:



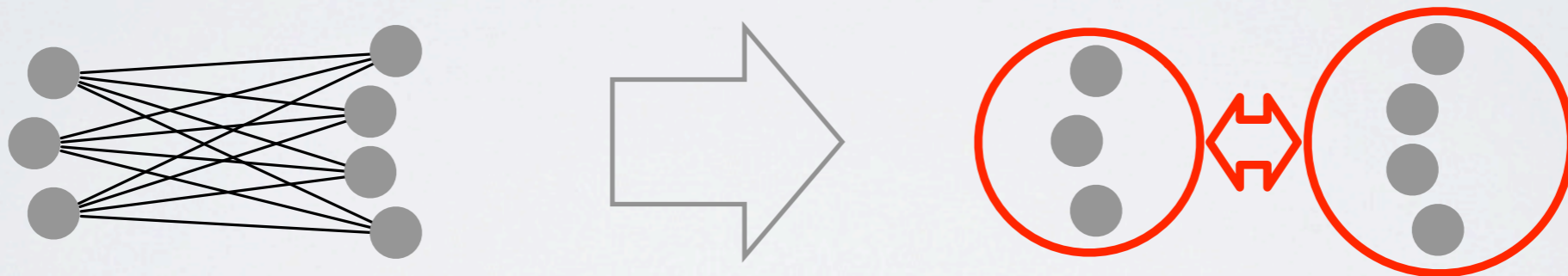
$$\begin{aligned} I_{AB} &= 2n_A n_B \log_2 N \\ &= 24 \log_2 7 \end{aligned}$$



$$\begin{aligned} I_{AB}^{(p)} &= 2(n_A + n_B) \log_2 N \\ &= 14 \log_2 7 \end{aligned}$$

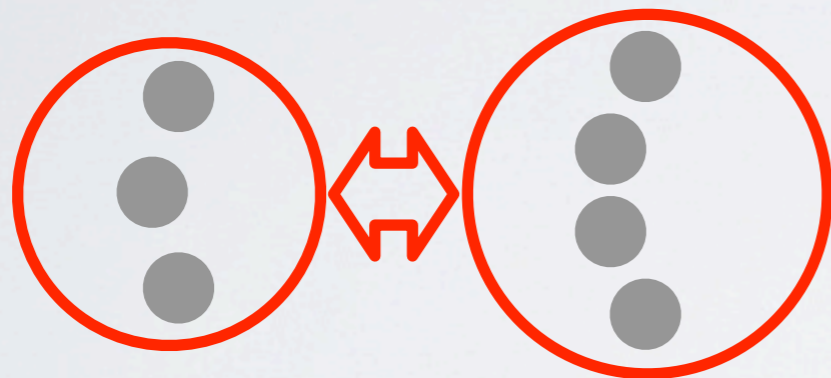
COMPRESSIBLE COMPONENTS

Hence the compression is given by:

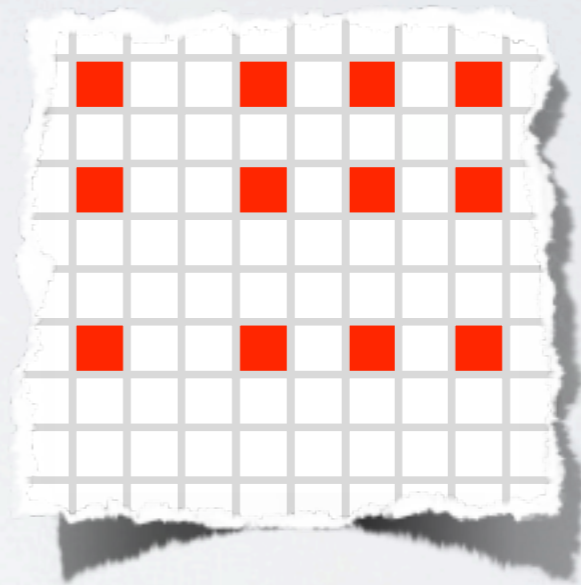
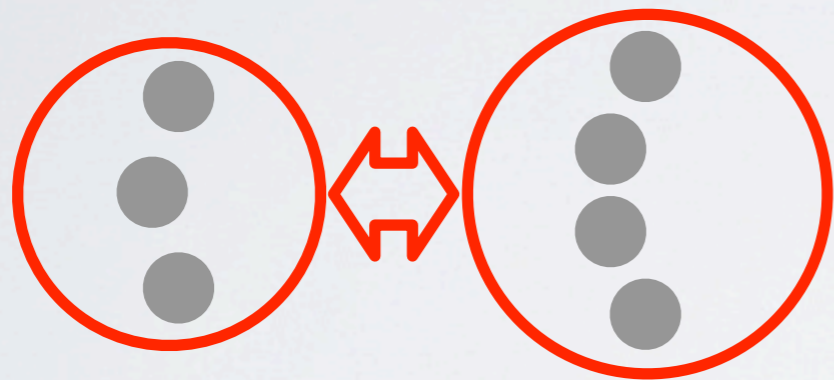


$$\begin{aligned}\Delta I_{AB} &= I_{AB}^{(p)} - I_{AB} = 2(n_A n_B - n_A - n_B) \log_2 N \\ &= (24 - 14) \log_2 7 = 10 \log_2 7\end{aligned}$$

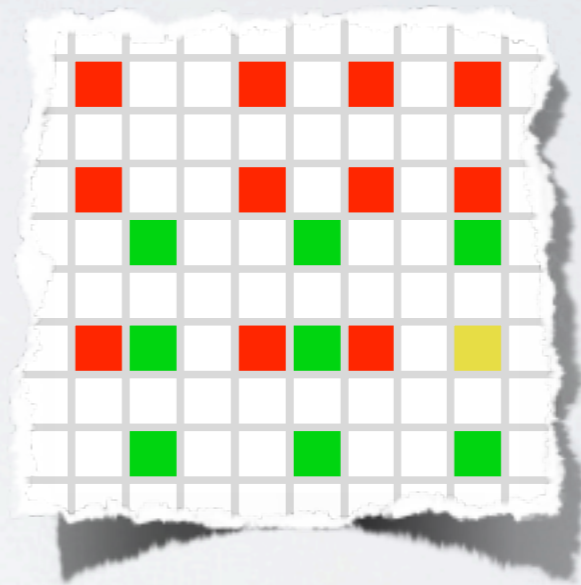
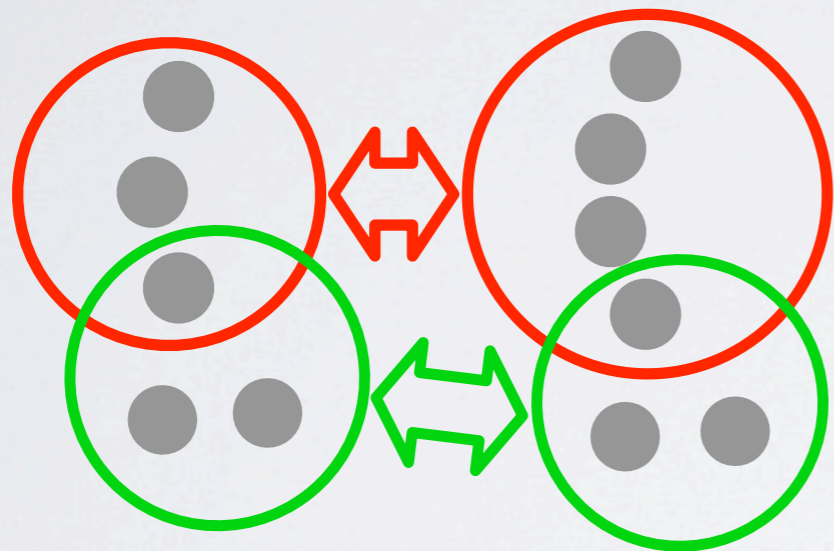
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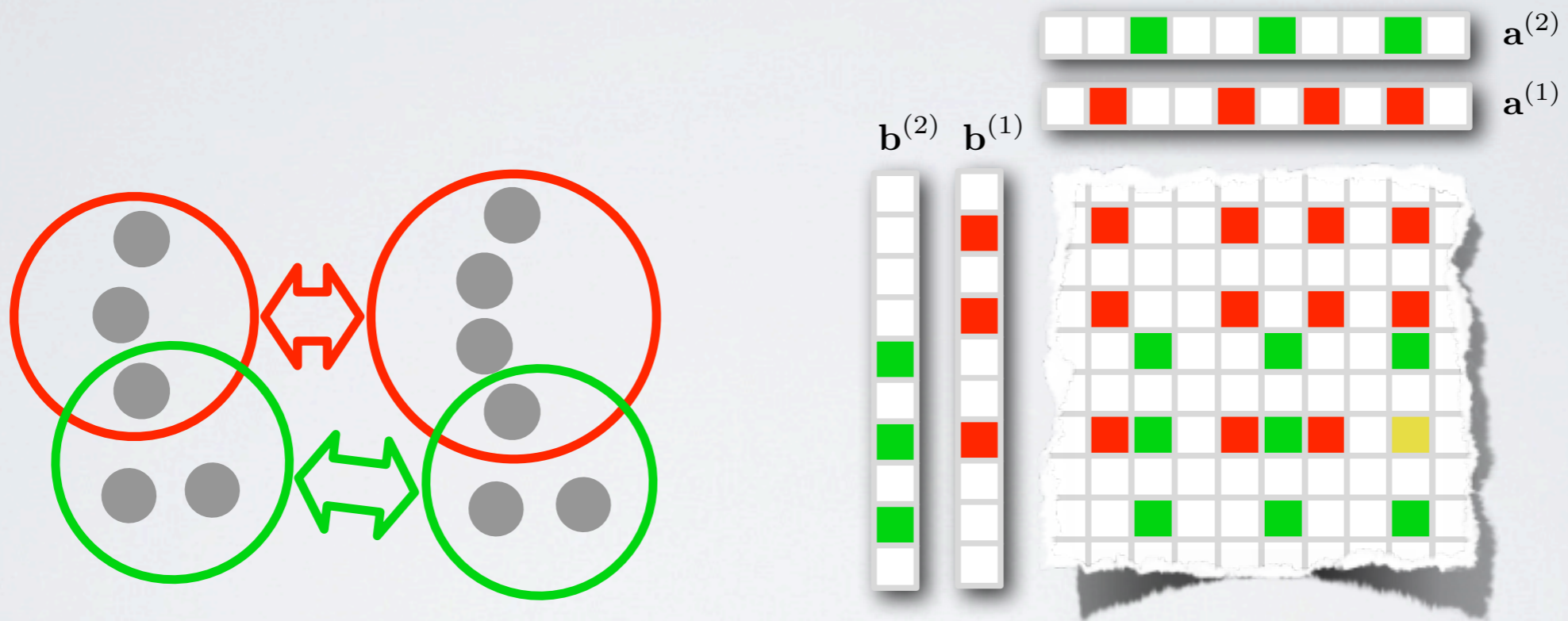
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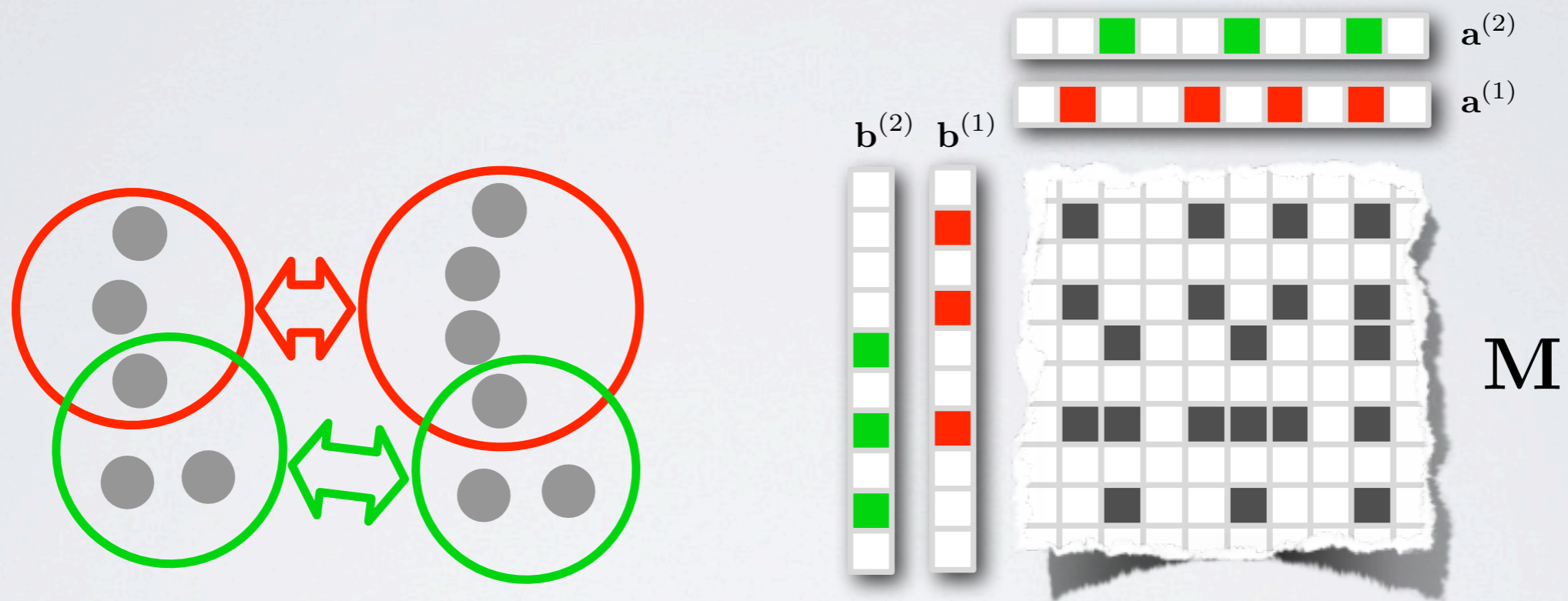


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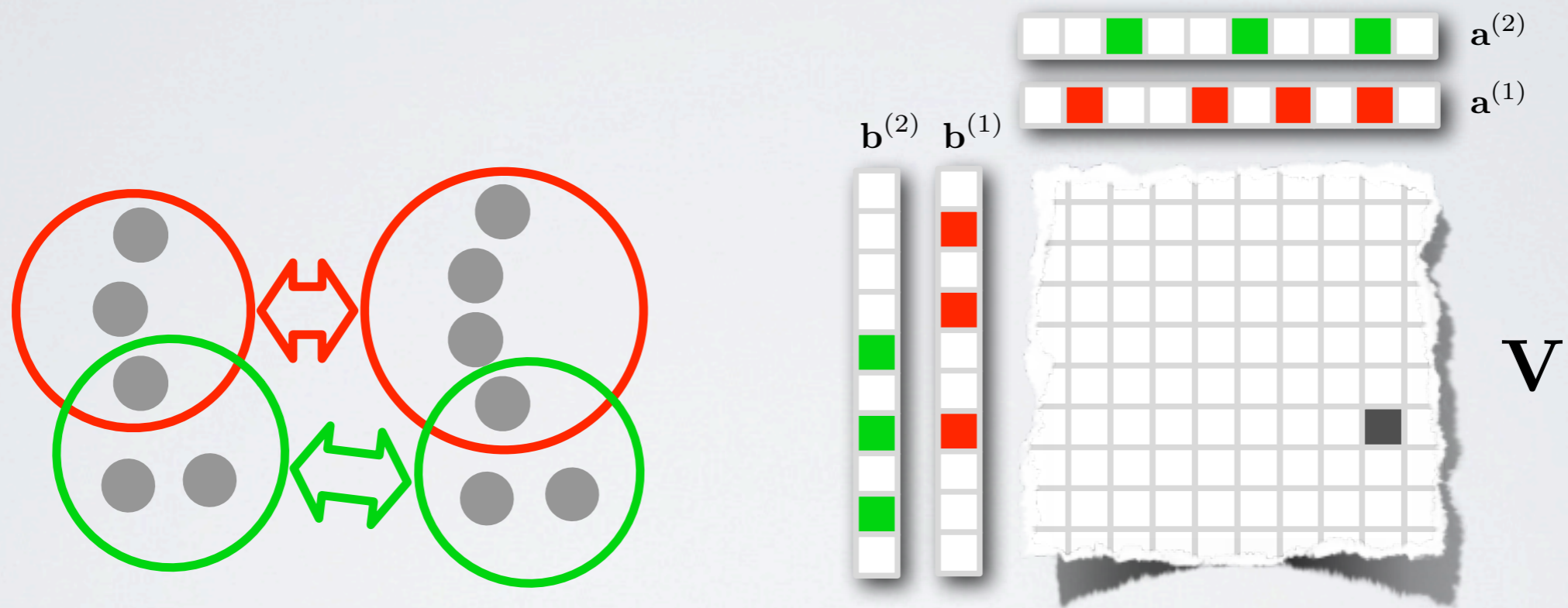
$$v_{ij} = \frac{1}{2} \left(\sum_{k \in S} a_i^{(k)} b_j^{(k)} - m_{ij} + \left| \sum_{k \in S} a_i^{(k)} b_j^{(k)} - m_{ij} \right| \right)$$

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$$\Delta I_{total} = \sum_{k \in S} \Delta I_k - (\mathbf{1}^T \mathbf{V} \mathbf{1}) 2 \log_2 N$$

COMPRESSIBLE COMPONENTS

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We then maximise this quantity using a greedy algorithm:

- 1) Rank all possible compressible components in order of compressibility.
- 2) Select most compressible component.
- 3) Recalculate compressibilities of all others in light of overlaps with selected component.
- 4) Repeat from step 1) for all compressible components apart from selected one(s).

SOCIAL NETWORKS

A classical social network dataset for community detection is Zachary's karate club network. The shown communities were calculated by Newman & Girvan.

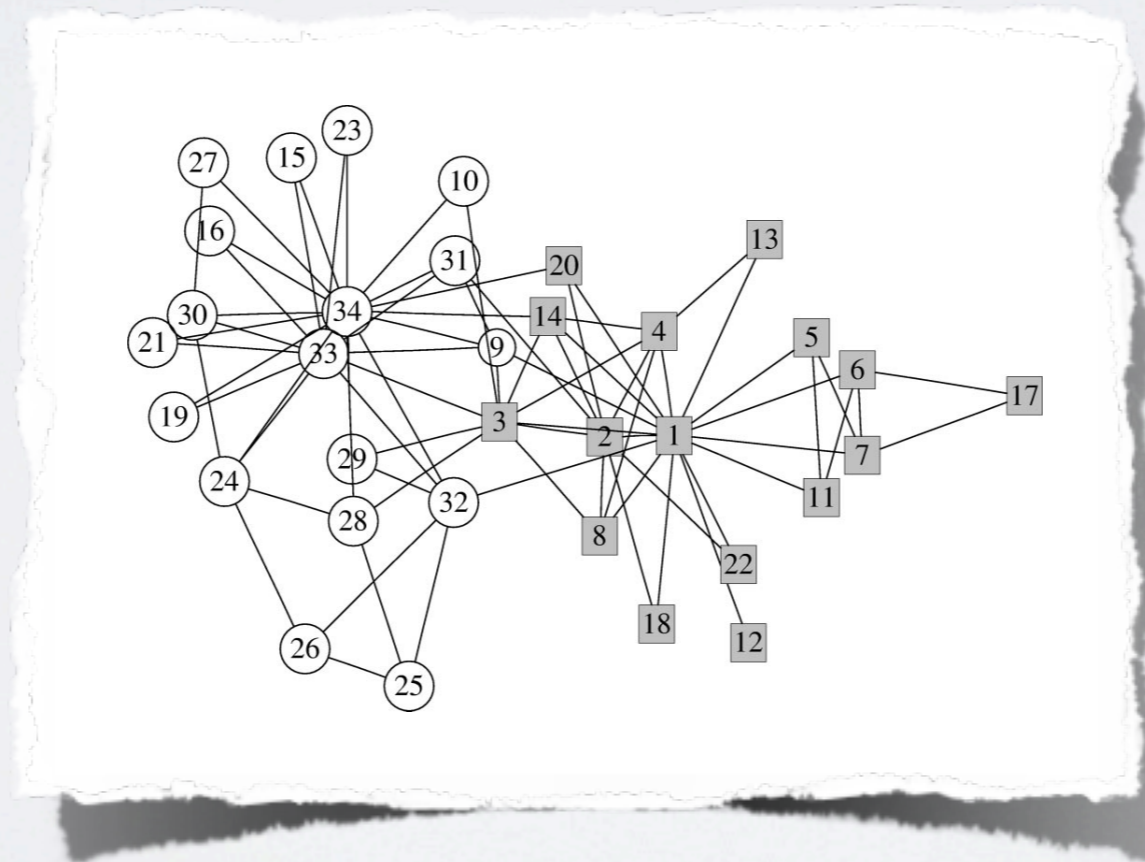


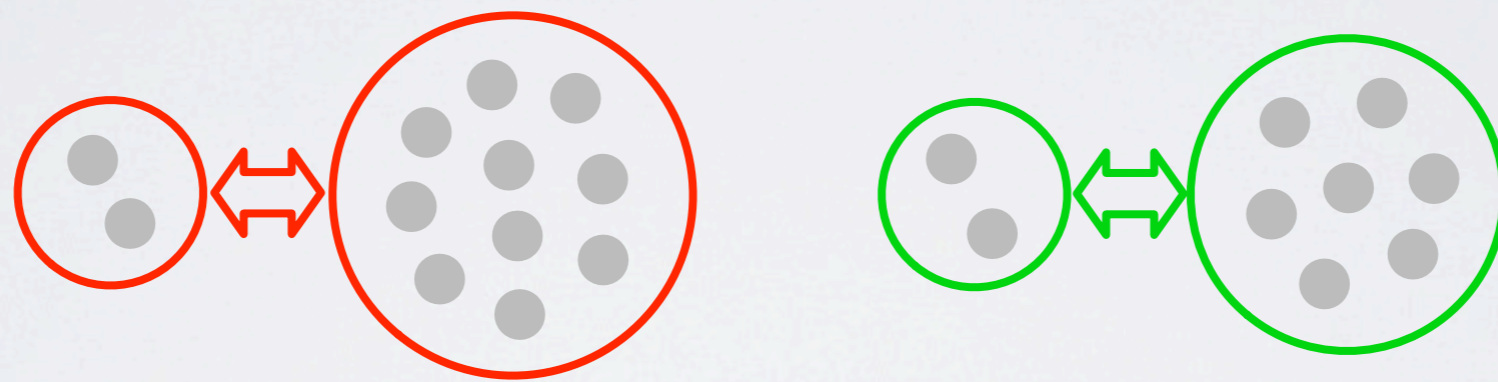
Figure from M. E. J. Newman & M. Girvan, PRE 69, 026113 (2004)

SOCIAL NETWORKS

What are the most compressible components in the karate club network?

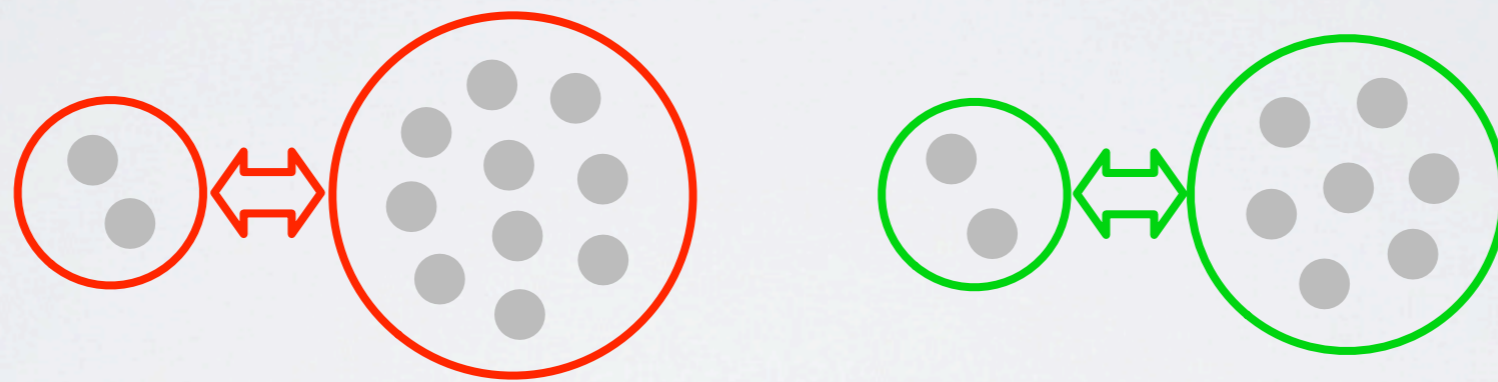
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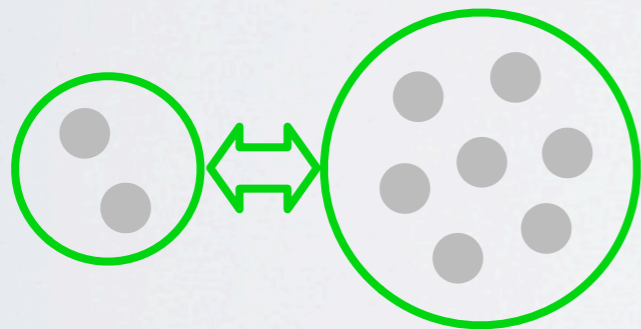
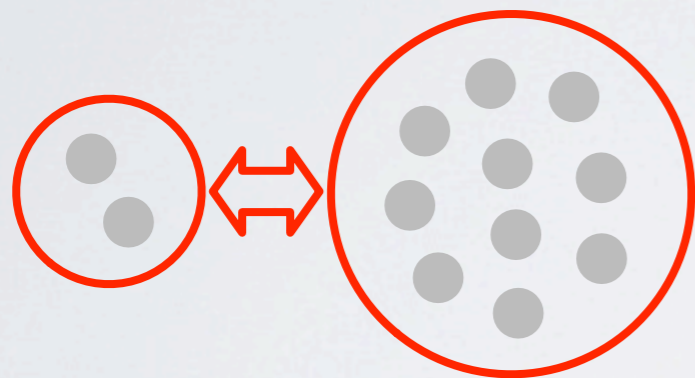
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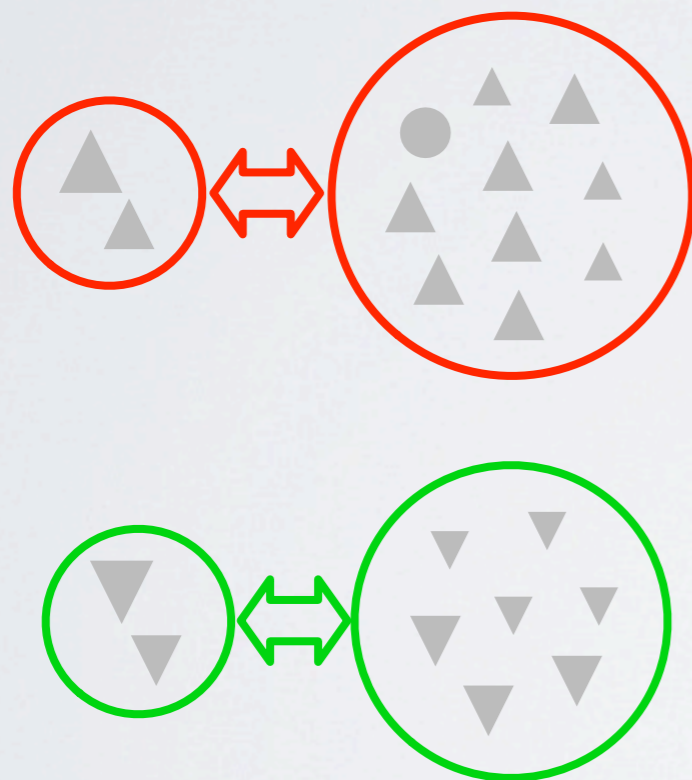
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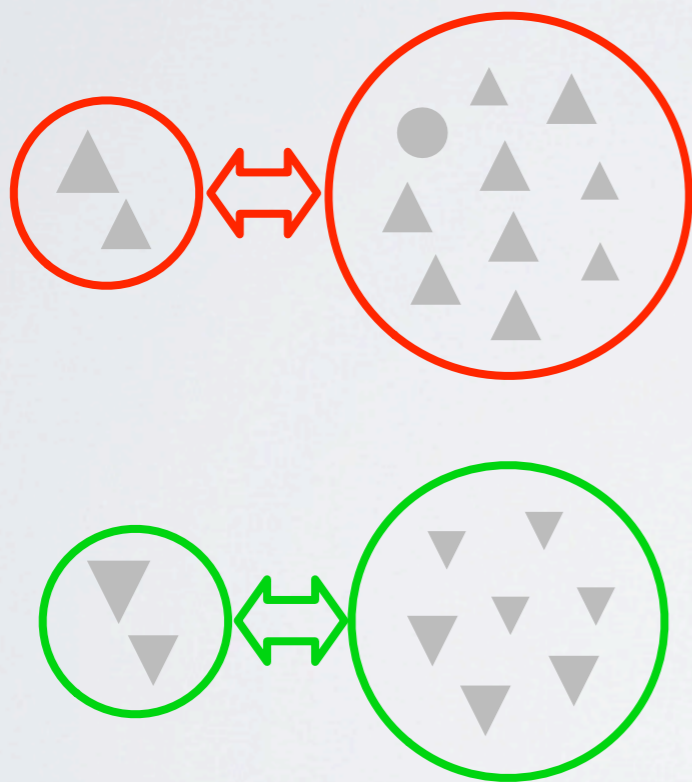


The two factions and their leaders.

INDIVIDUAL NUMBER	SIDE OF CUT	FACTION	CLUB AFTER FISSION
1	Source	Mr. Hi - Strong	Mr. Hi's
2	Source	Mr. Hi - Strong	Mr. Hi's
3	Source	Mr. Hi - Strong	Mr. Hi's
4	Source	Mr. Hi - Strong	Mr. Hi's
5	Source	Mr. Hi - Strong	Mr. Hi's
6	Source	Mr. Hi - Strong	Mr. Hi's
7	Source	Mr. Hi - Strong	Mr. Hi's
8	Source	Mr. Hi - Strong	Mr. Hi's
9	Sink	John - Weak	Mr. Hi's
10	Sink	None	Officers'
11	Source	Mr. Hi - Strong	Mr. Hi's
12	Source	Mr. Hi - Strong	Mr. Hi's
13	Source	Mr. Hi - Weak	Mr. Hi's
14	Source	Mr. Hi - Weak	Mr. Hi's
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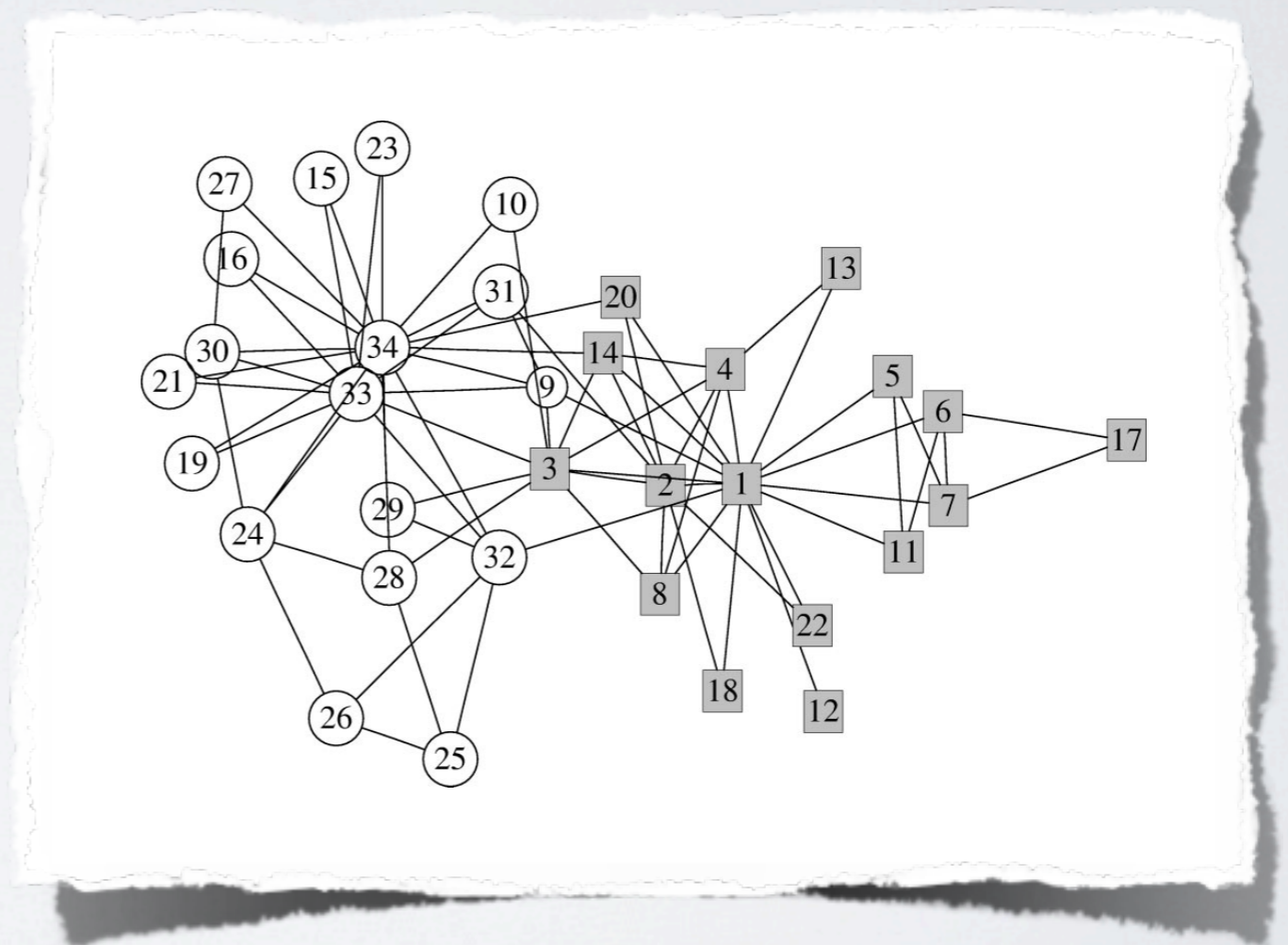


Figure from M. E. J. Newman & M. Girvan, PRE 69, 026113 (2004)

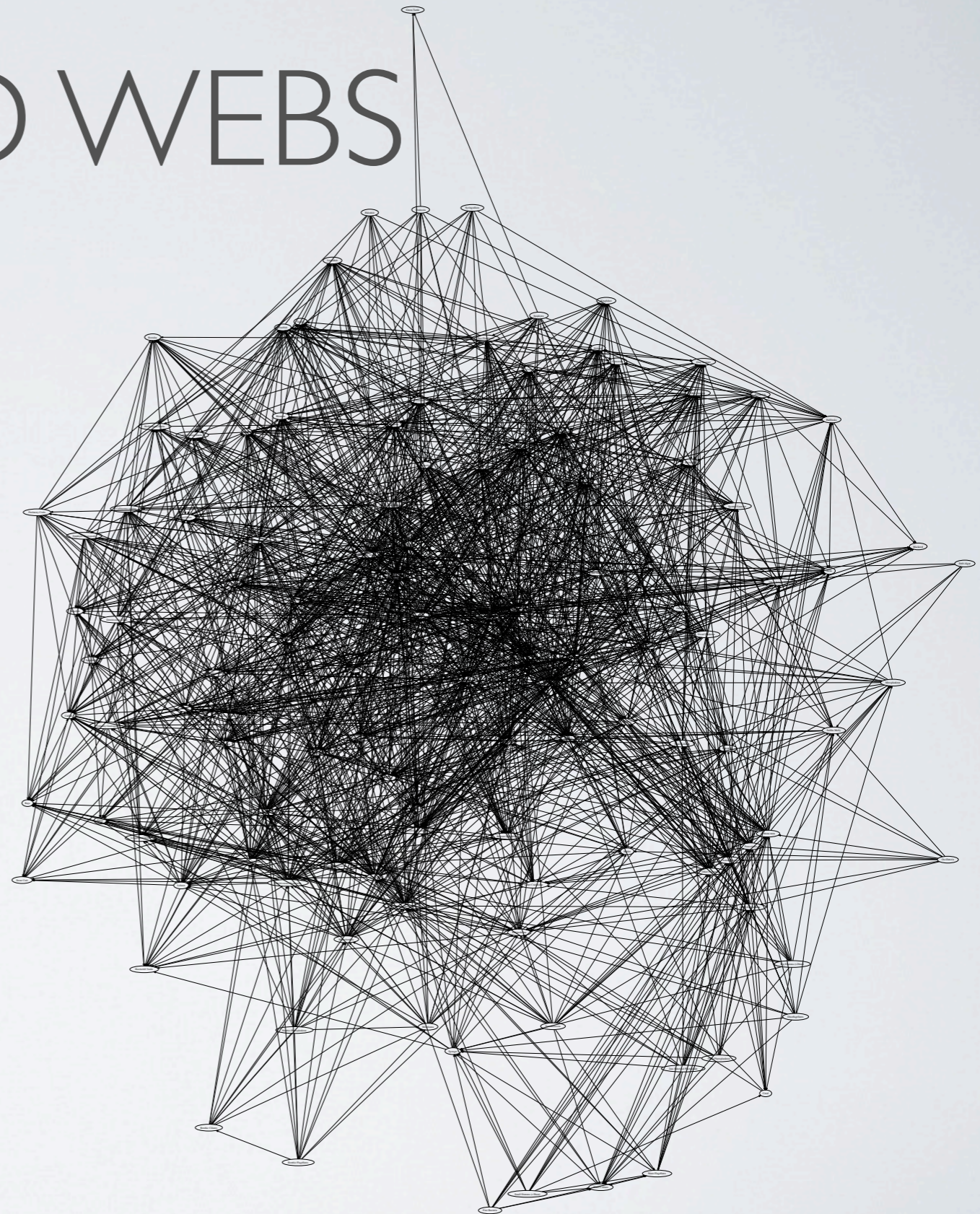
FOOD WEBS

Food webs are ecological networks of predator-prey relationships between organisms.

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Here we will show results for a food web of 122 species in a Florida ecosystem, with 1767 edges.

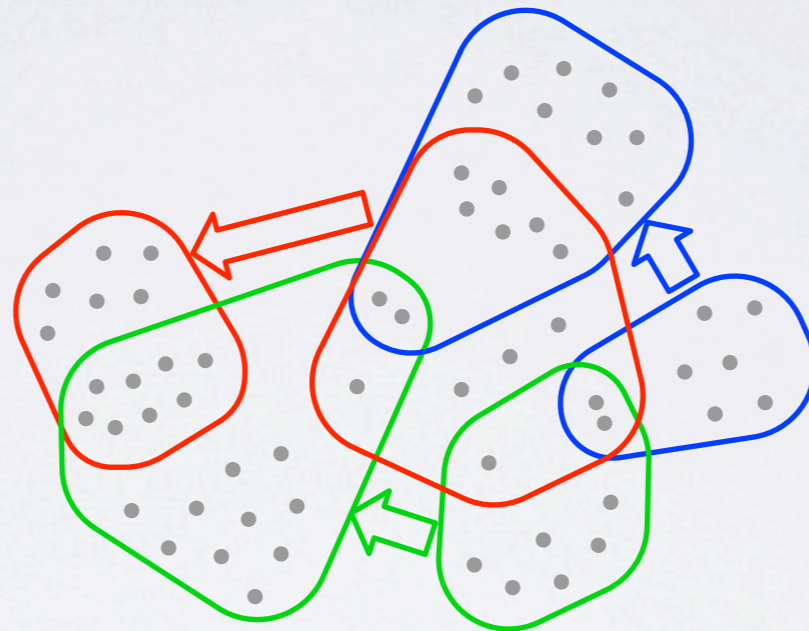


FOOD WEBS

The most compressible components of this food web:

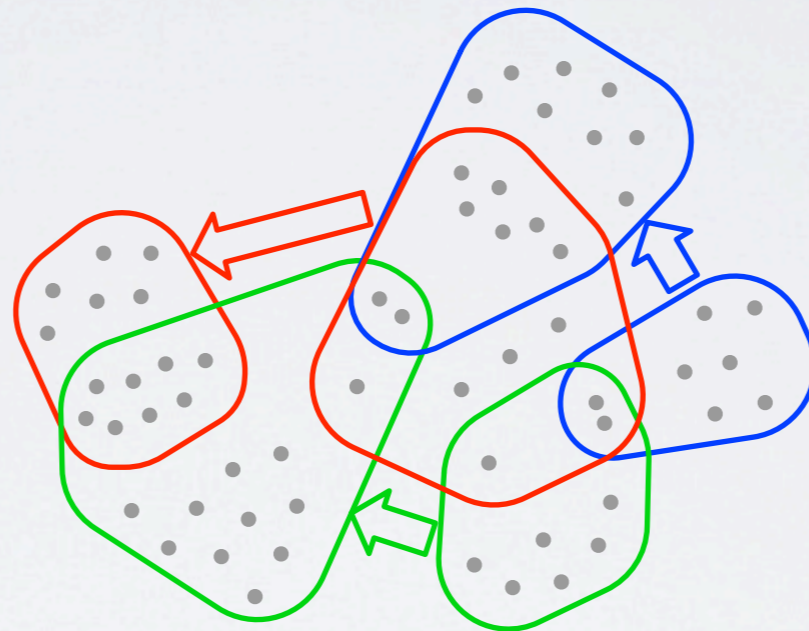
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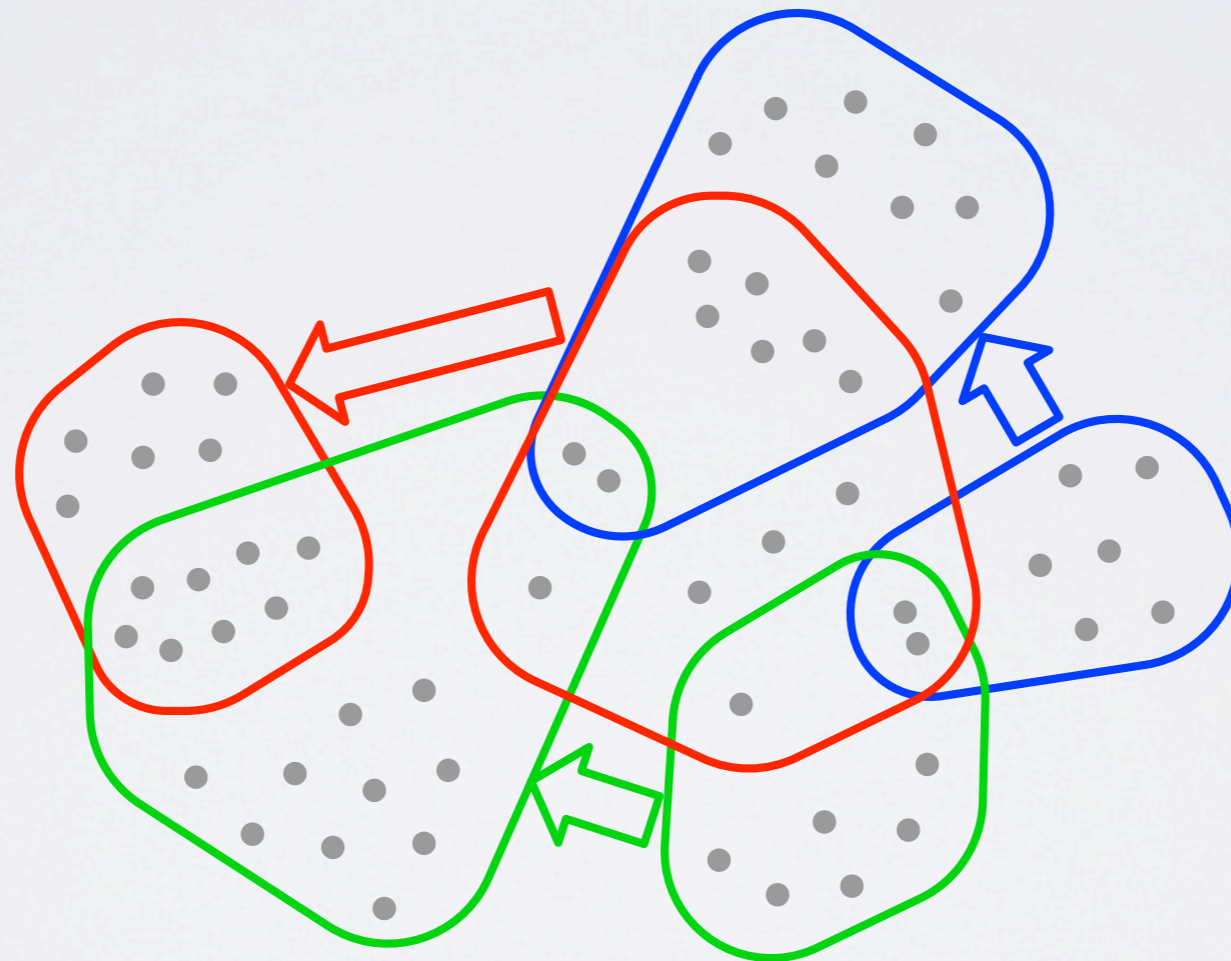
The most compressible components of this food web:



501 edges
28% of network

FOOD WEBS

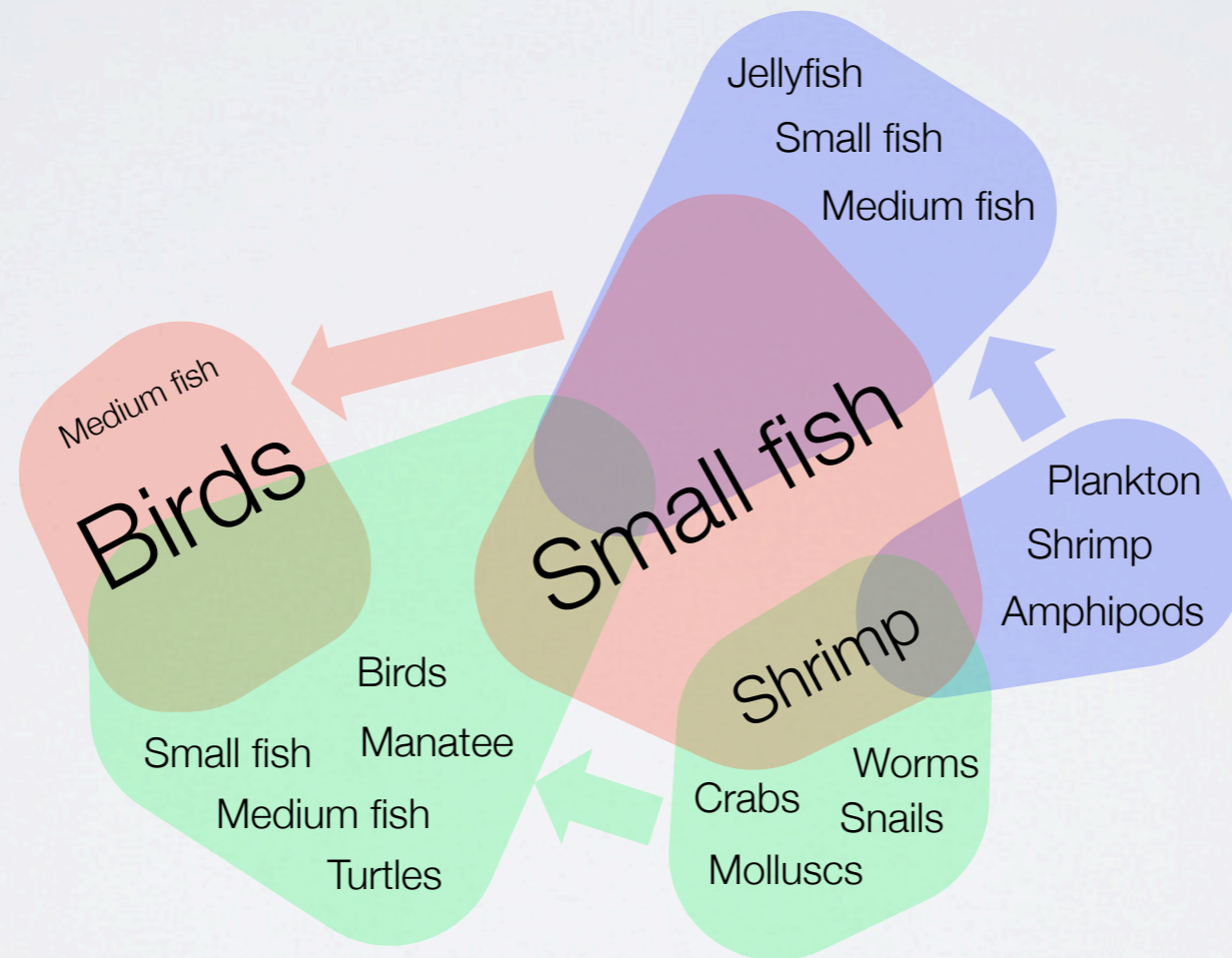
What do these sets mean?



Can we assign meaning in a more rigorous way?

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TERM ENRICHMENT

Given a set of terms associated with the nodes we can calculate the enrichment of these relative to a null model.

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The probability of finding term i in set s exactly k times, if the term appears n_i times among all N nodes, is

$$p_i(k, s) = \frac{n_i C_k^{N-n_i} C_{s-k}^{n_i}}{N C_s}$$

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$$\text{p-value: } P_i(k^*, s) = 1 - \sum_{k=0}^{k^*-1} p_i(k, s)$$

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$$P_i(k^*, s) = 1 - \sum_{k=0}^{k^*-1} p_i(k, s) \quad P_i^{(B)}(k^*, s) = T P_i(k^*, s)$$

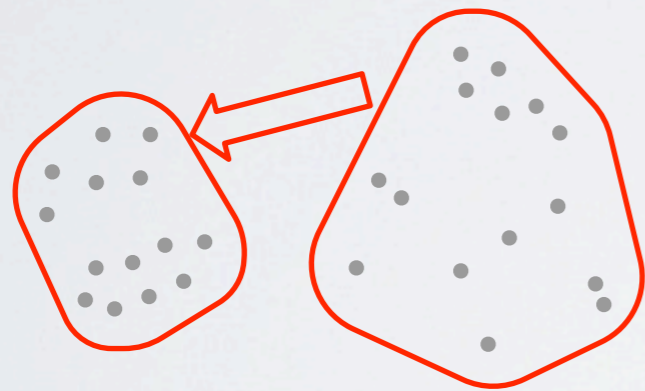
Bonferroni

FOOD WEB TERM ENRICHMENT

Some of the enriched terms in the compressible components of the food web:

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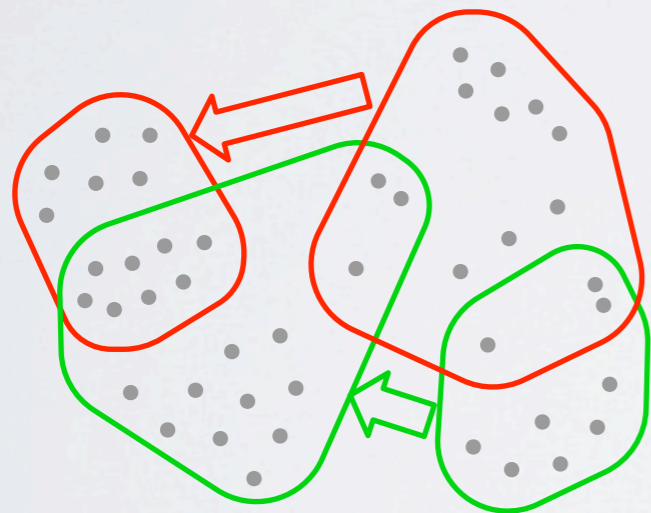
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p-value: $8e-06$



birds
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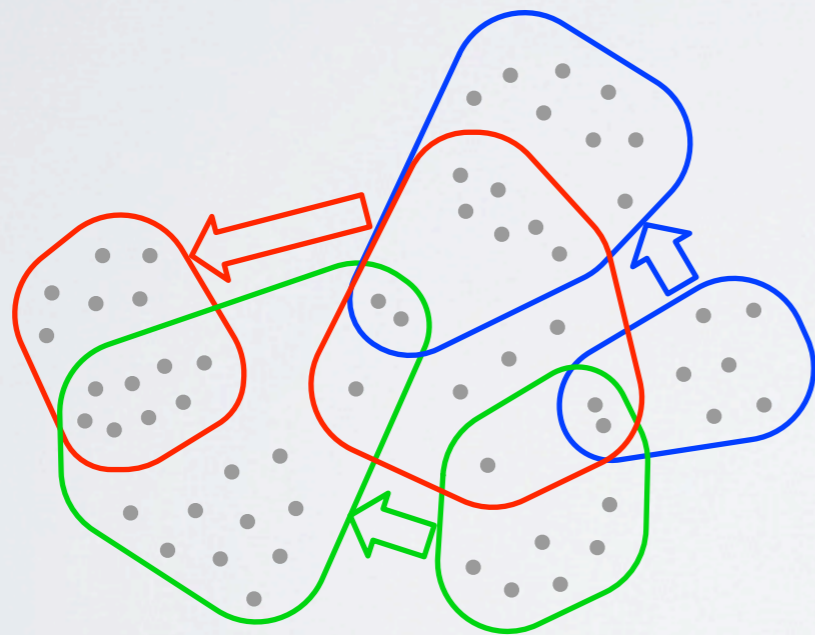
worms
p-value: $8e-03$



birds
p-value: $4e-04$

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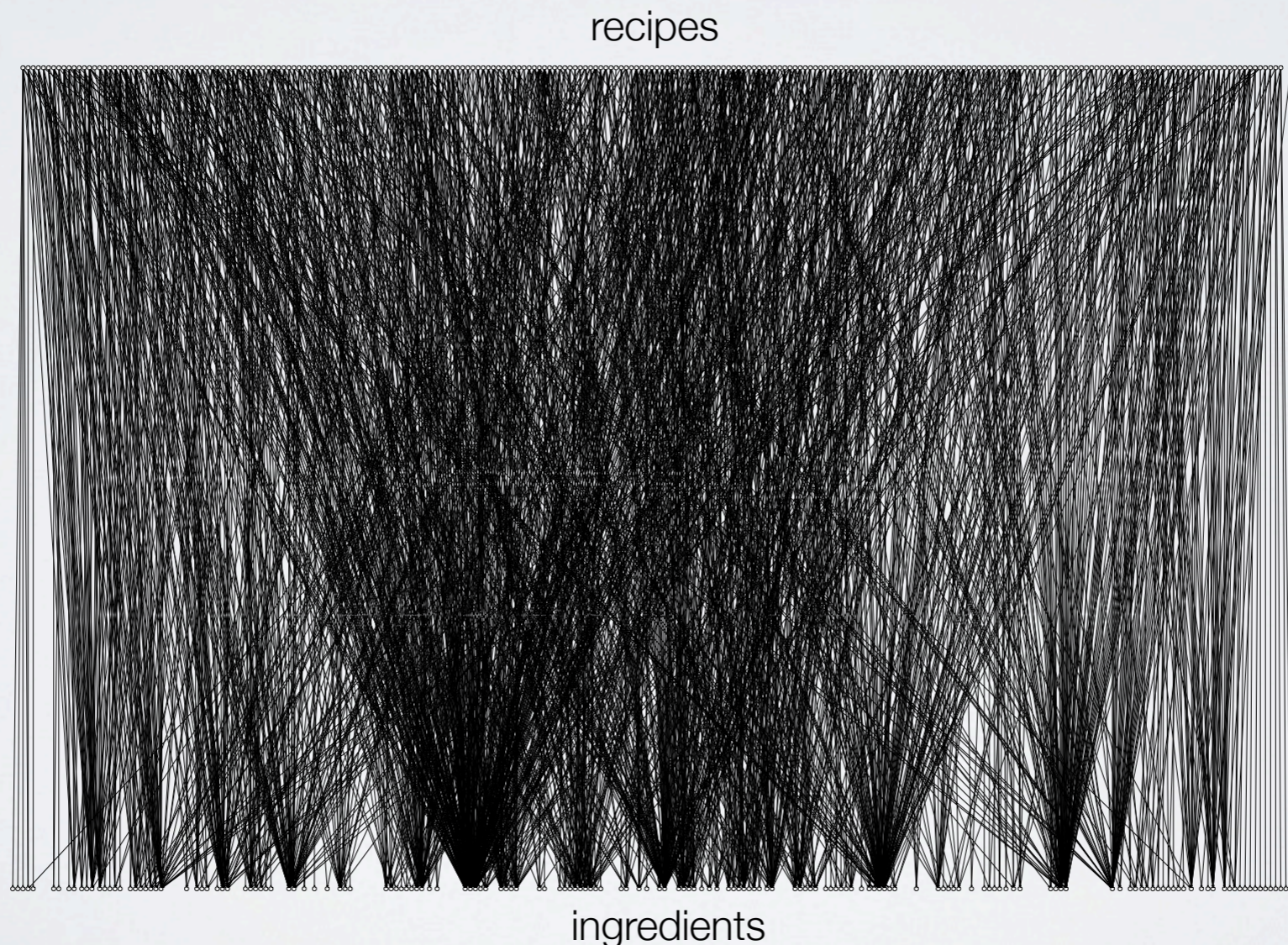
copepods
p-value: $4e-04$



small fish
p-value: $2e-07$

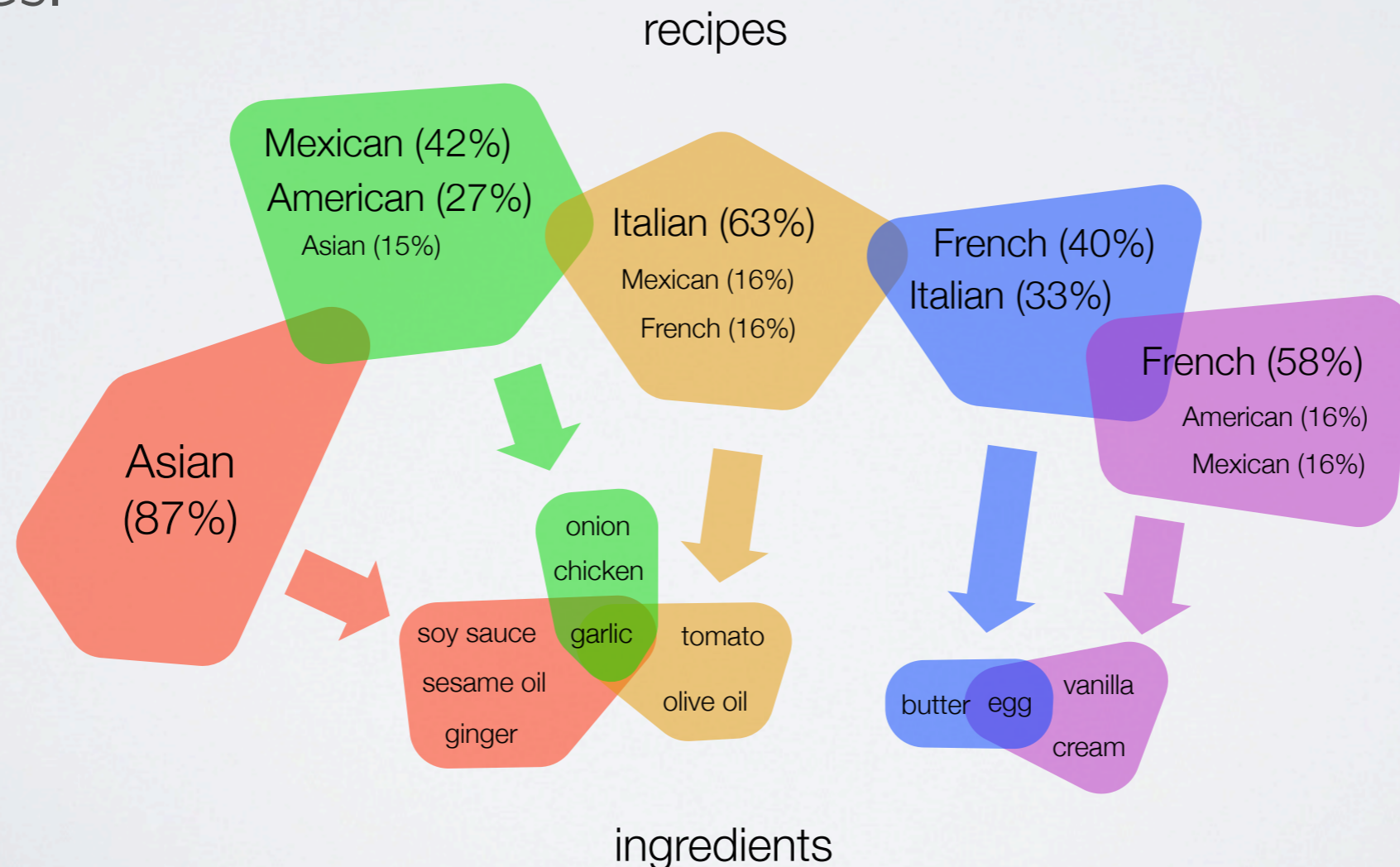
RECIPE NETWORK

Food ingredients and recipes form a bipartite network. Here we study an example with 415 nodes and 1748 edges.



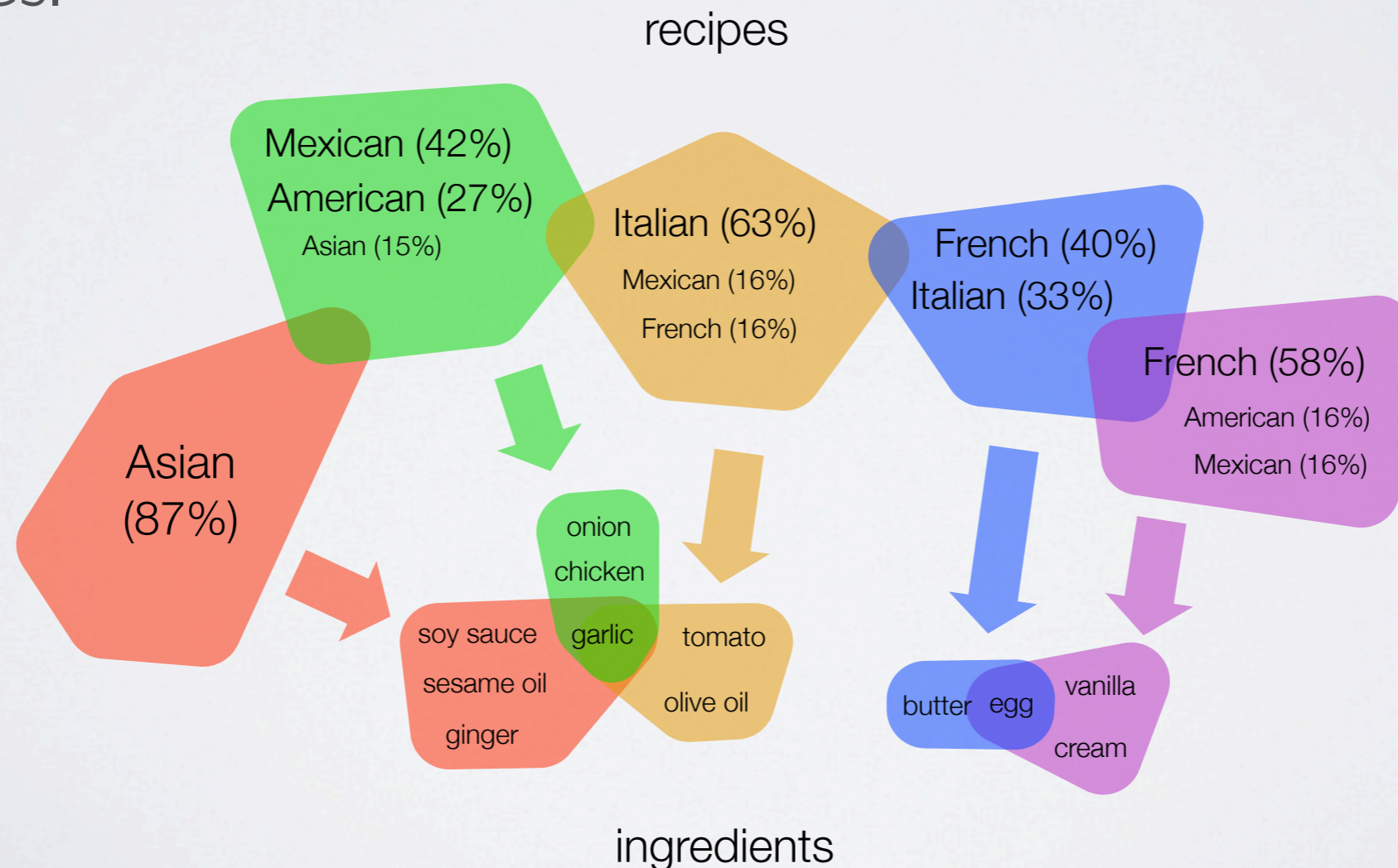
RECIPE NETWORK

The five most compressible components exhibit highly significant enrichment for certain cuisines. The overlap between the power nodes shows the proximity of cuisines.



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YEAST TRANSCRIPTION

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In a yeast transcription network the nodes are yeast genes and the edges are regulatory interactions between them.

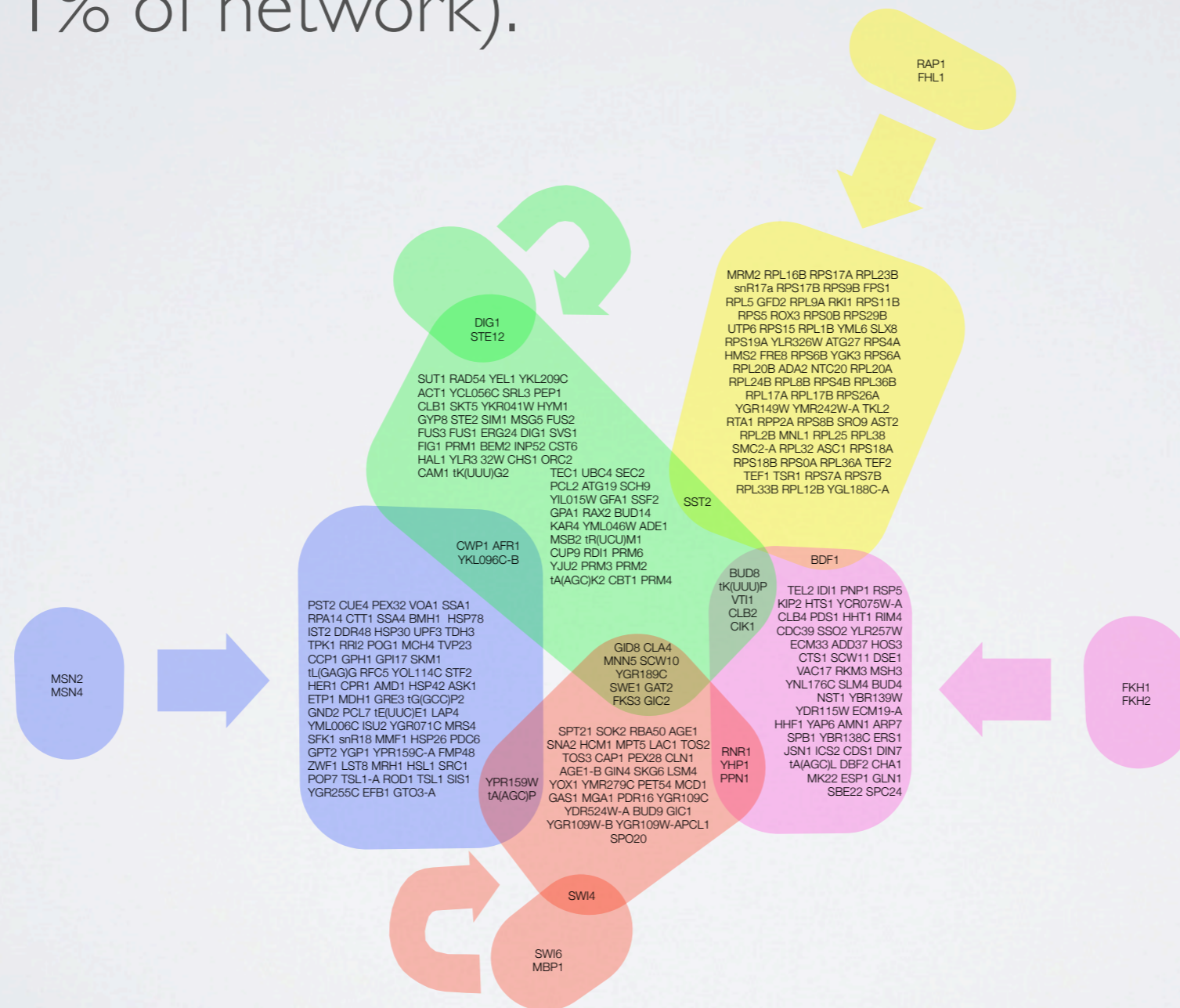
Our dataset here has 2534 nodes and 6071 edges.

YEAST TRANSCRIPTION

The 5 most compressible components cover 699 edges (11% of network).

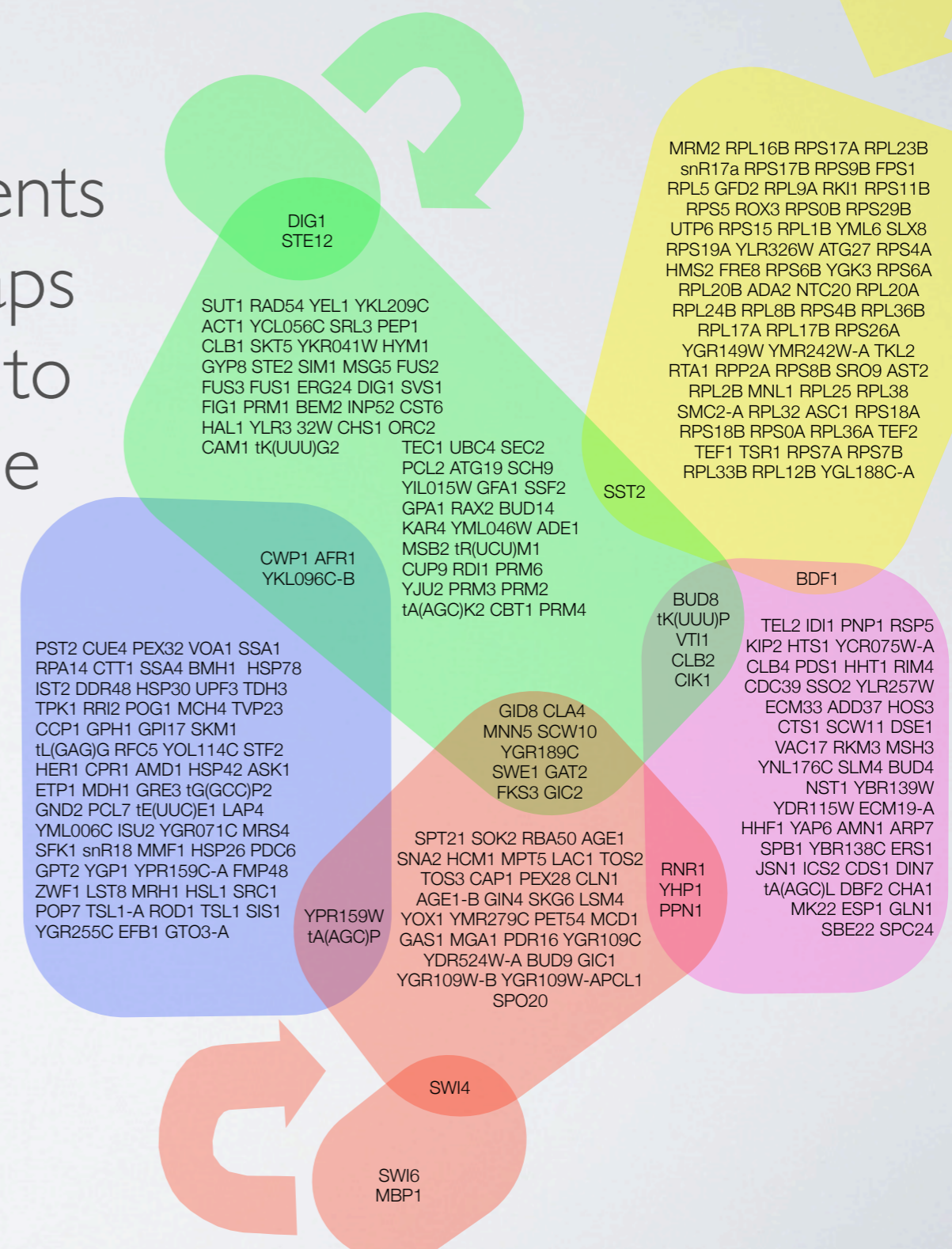
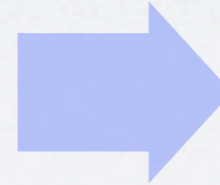
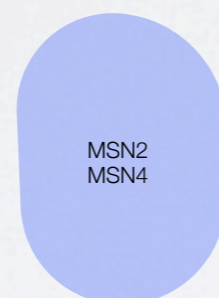
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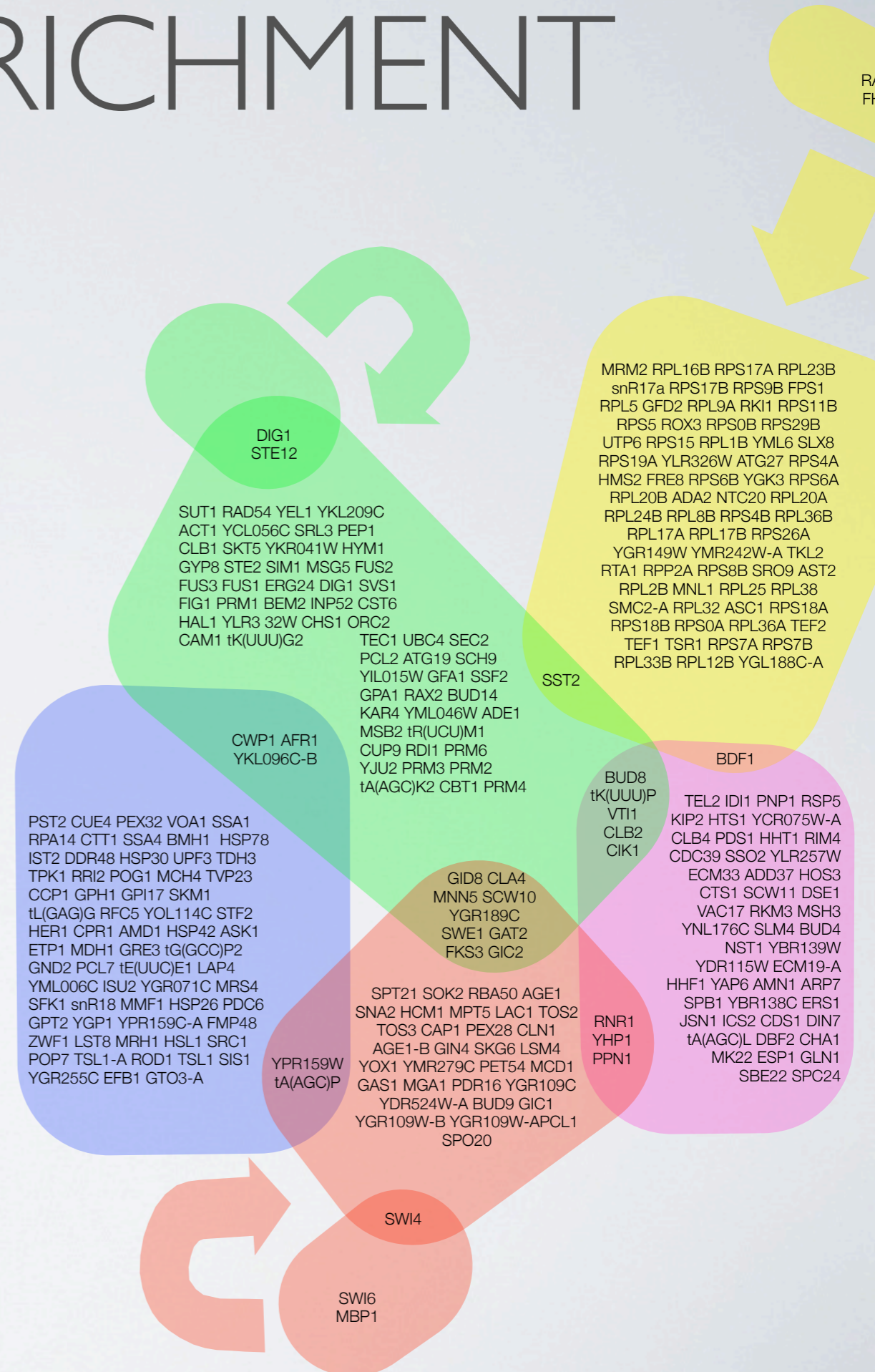
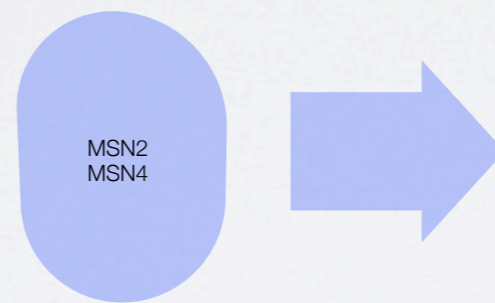
Note that compressible components have no restrictions on the overlaps between power nodes belonging to the same or different compressible components.



GO TERM ENRICHMENT

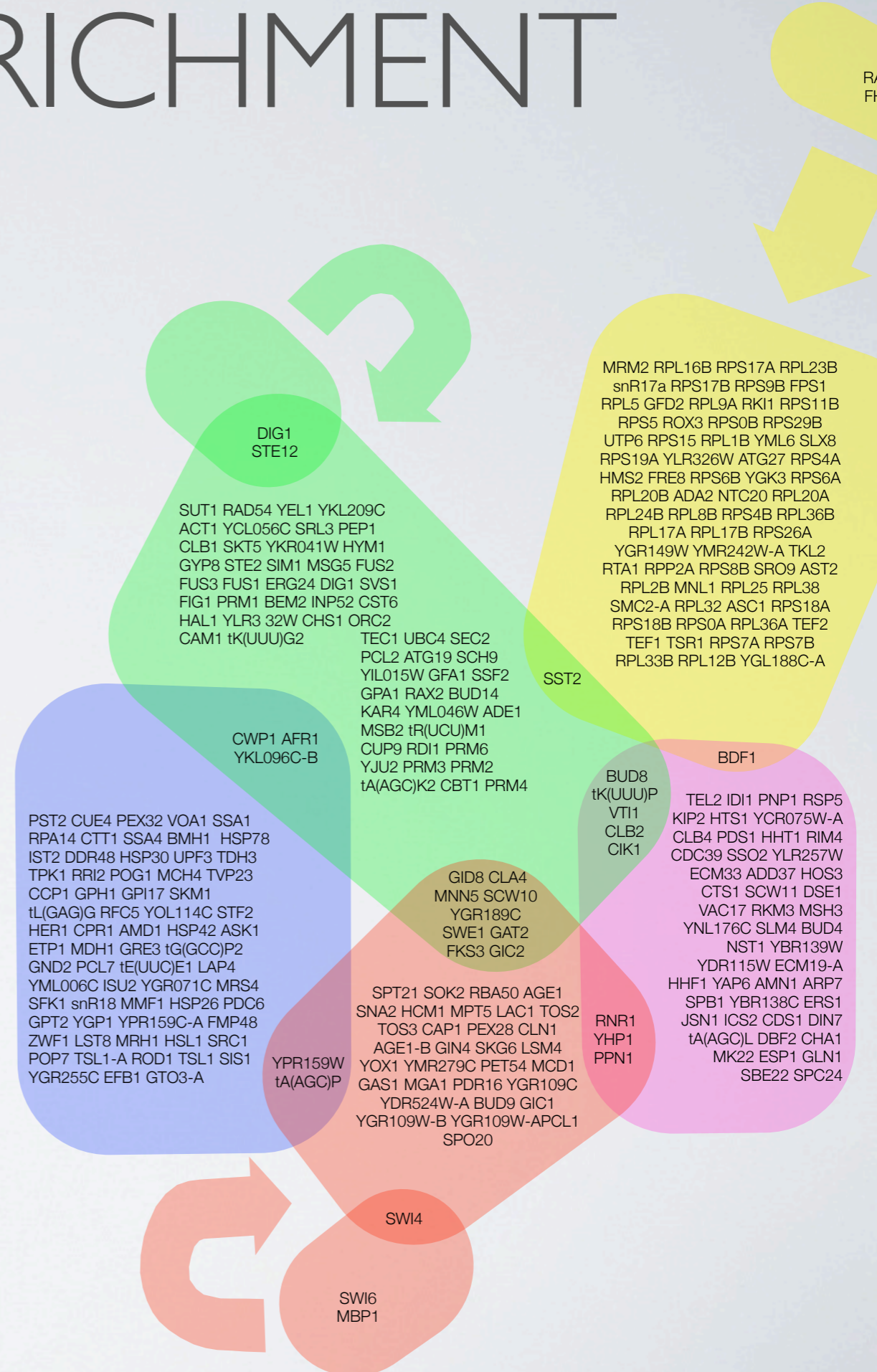
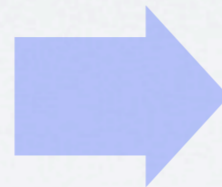
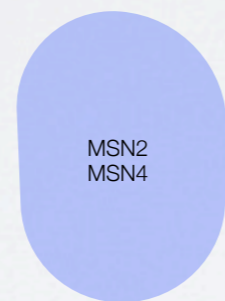
In the yeast transcription network 67% of power nodes exhibit Gene Ontology (GO) term enrichment.

Among the top 100 compressible components, this rises to 91%.



GO TERM ENRICHMENT

The most compressible component represents regulation of the G1/S phase of cell cycle by two heterodimers: SBF (Swi4/Swi6) and MBF (Mbp1/Swi6)



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The most compressible component represents regulation of the G1/S phase of cell cycle by two heterodimers: SBF (Swi4/Swi6) and MBF (Mbp1/Swi6)



Set A	GO terms	Set B	GO terms
MBP1 SWI4 SWI6	regulation of transcription involved in G1/S phase of mitotic cell cycle (2.05E-04)	47 genes	cell cycle process (1.69E-04) cell division (1.87E-04) incipient cellular bud site (3.03E-04)

GO TERM ENRICHMENT

The second most compressible component represents regulation of two developmental processes, mating and filamentation, regulated by the **Ste12/Dig1/Dig2** and **Tec1/Ste12/Dig1** complexes.

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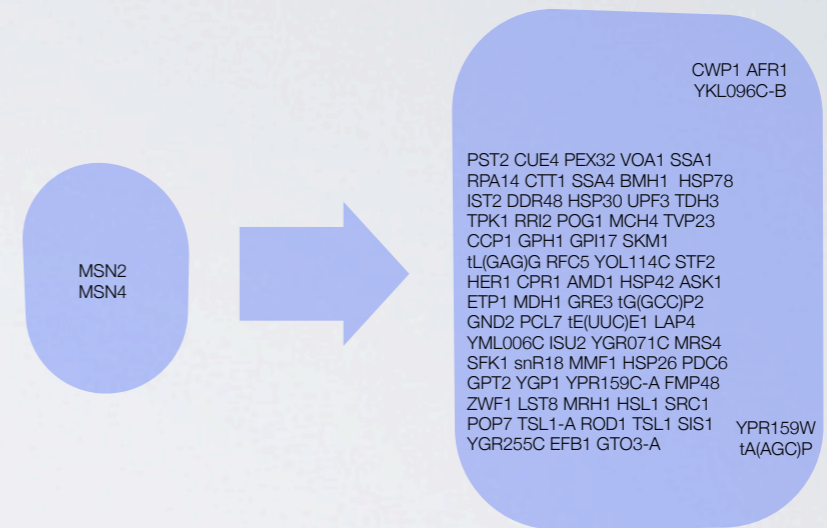
Set A	GO terms	Set B	GO terms
STE12 DIG1	<p>regulation of transcription by pheromones (4.93E-04)</p> <p>cell growth (7.54E-04)</p> <p>regulation of transcription from RNAP II promoter by pheromones (9.45E-04)</p>	78 genes	<p>multi-organism process (6.35E-14)</p> <p>site of polarized growth (7.59E-11)</p> <p>reproduction (2.65E-10)</p>

GO TERM ENRICHMENT

The third most compressible component represents regulation of a broad stress response by **Msn2** and **Msn4**.

GO TERM ENRICHMENT

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Set A	GO terms	Set B	GO terms
MSN2 MSN4	response to hydrostatic pressure (6.96E-05) response to nitrosative stress (6.96E-05) response to water (8.60E-05)	72 genes	unfolded protein binding (2.06E-03) chaperonin-containing T-complex (2.28E-03) hydrogen peroxide catabolic process (5.38E-03)

GO TERM ENRICHMENT

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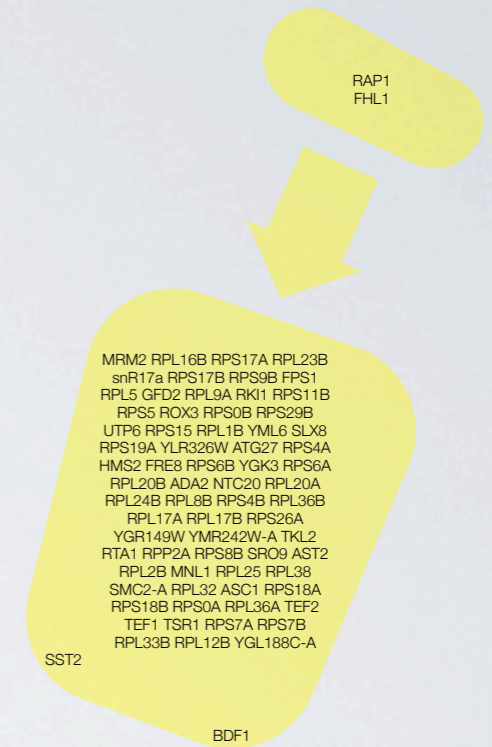
...can also be seen from the p-values of the GO term enrichment.

GO TERM ENRICHMENT

The fourth most compressible component represents highly specific regulation of the ribosome.

GO TERM ENRICHMENT

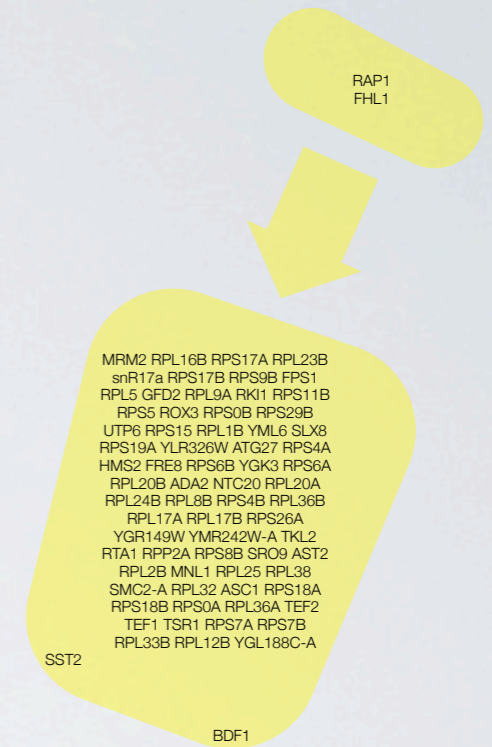
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Set A	GO terms	Set B	GO terms
RAP1 FHL1	CURI complex (6.08E-03) nucleic acid binding transcription factor activity (7.90E-03)	71 genes	ribonucleoprotein complex (1.40E-37) ribosome (5.46E-37) ribosomal subunit (4.93E-36)

GO TERM ENRICHMENT

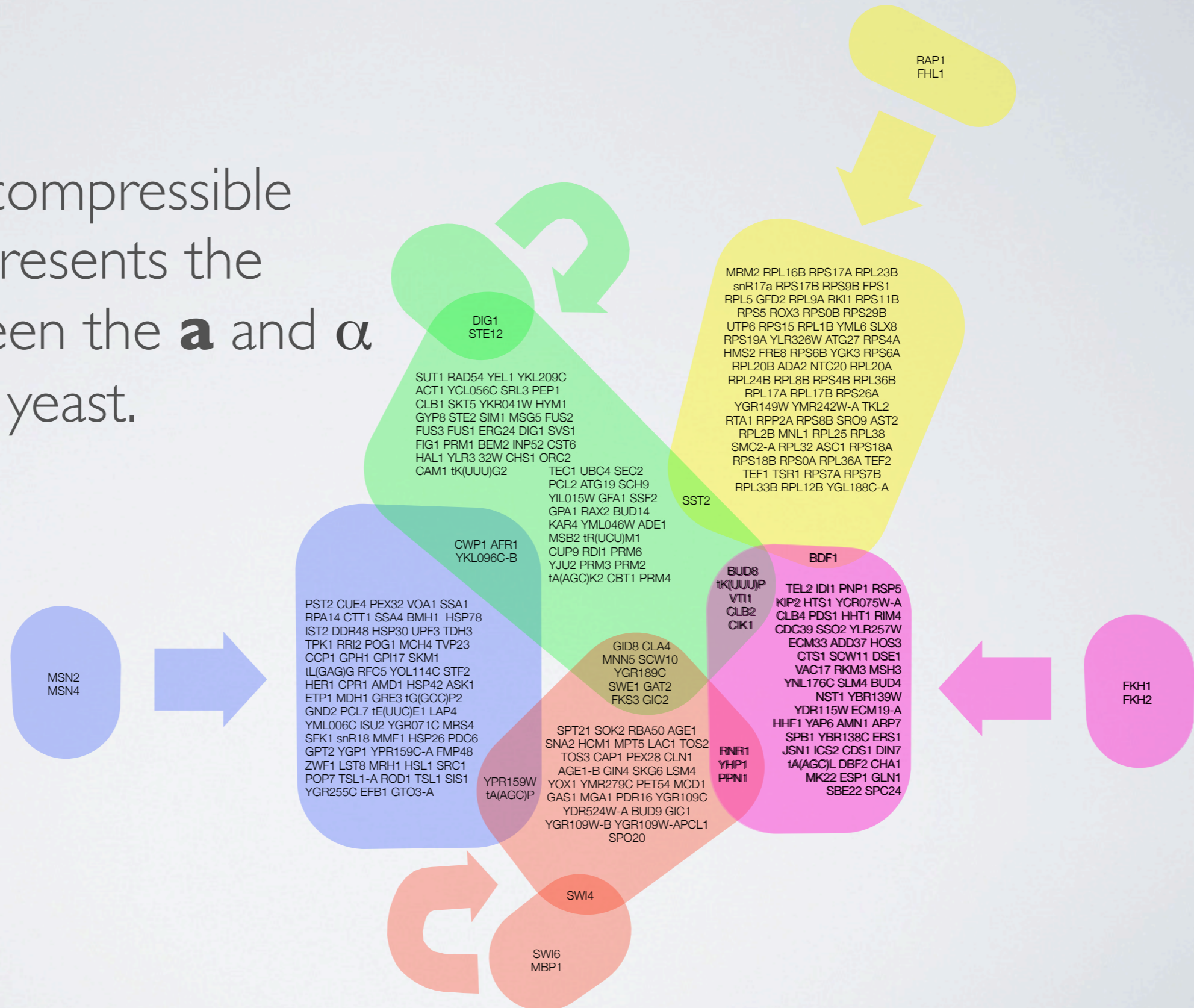
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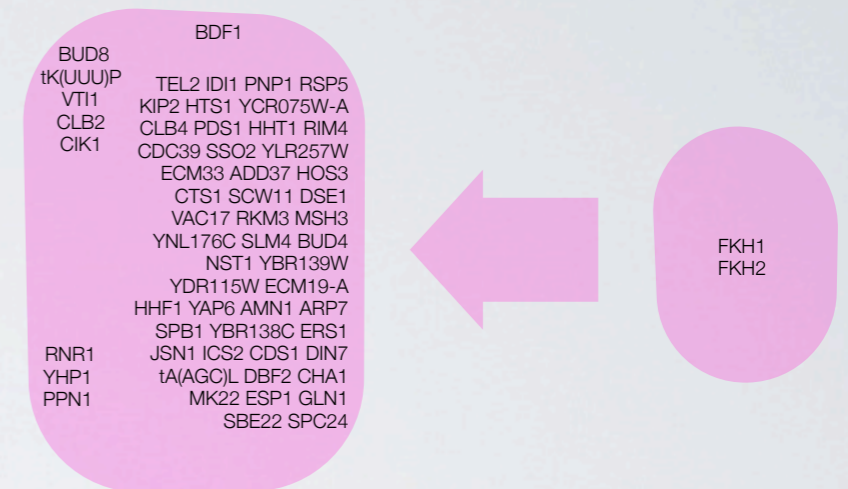
GO TERM ENRICHMENT

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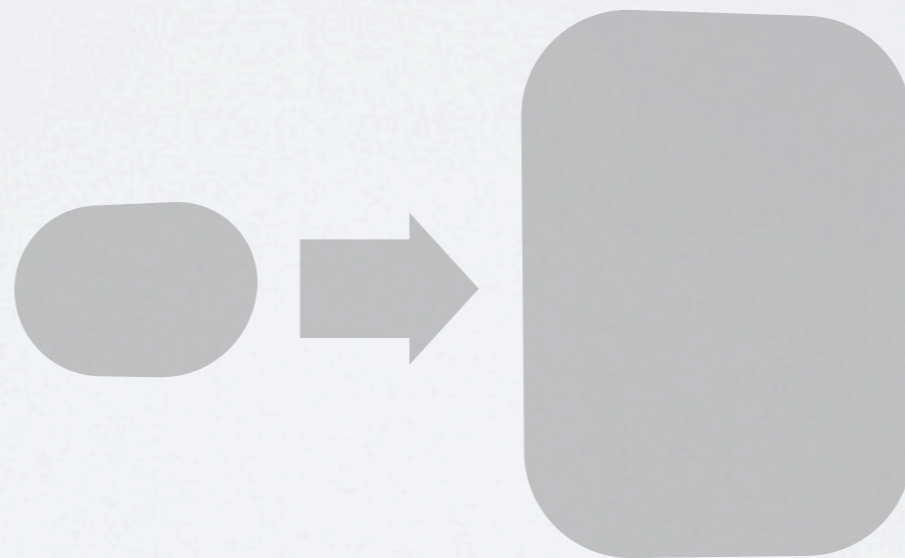


Set A	GO terms	Set B	GO terms
FKH1 FKH2	<p>regulation of transcription involved in G2/M-phase of mitotic cell cycle (1.03E-04)</p> <p>regulation of transcription elongation from RNA polymerase II promoter (2.80E-04)</p> <p>regulation of transcription elongation, DNA-dependent (2.93E-04)</p>	58 genes	<p>cell cycle (1.64E-05)</p> <p>cell cycle process (9.68E-05)</p> <p>establishment of organelle localization (4.81E-04)</p>

GO TERM ENRICHMENT

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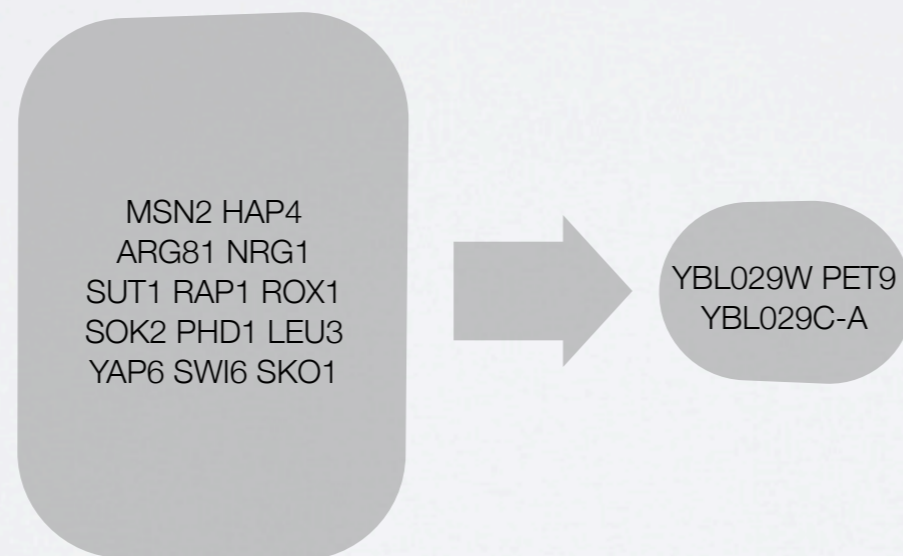
Most compressible components in transcription networks have few genes in Set A and many genes in Set B.



GO TERM ENRICHMENT

Most compressible components in transcription networks have few genes in Set A and many genes in Set B.

A notable exception is the 18th compressible component.



GO TERM ENRICHMENT



GO TERM ENRICHMENT



Set A	GO terms	Set B	GO terms
MSN2 HAP4 ARG81 NRG1 SUT1 RAP1 ROX1 SOK2 PHD1 LEU3 YAP6 SWI6 SKO1	nucleic acid binding transcription factor activity (1.6E-12) protein binding transcription factor activity (1.2E-7)	PET9 YBL029W YBL029C-A	cell death (8.9E-03) (8.9E-03)

This compressible component represents processes that lead to cell death, forming a bottleneck of regulatory information as no downstream processing is required - quite literally a regulatory dead end.

GO TERM ENRICHMENT

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YBL029W has been speculatively linked to apoptosis as one of several genes that are over-expressed in apoptotic temperature-sensitive yeast cells.

YBL029C-A has never been linked to apoptosis. Its membership in this set suggests that it might be.

E. COLI TRANSCRIPTION NETWORK

Another widely-studied transcription network is that of the bacterium *E. coli*. The one we use here has 889 nodes and 1465 edges.

In *E. coli* many of the genes are organised into overlapping DNA regions, known as operons. These groups of genes correspond to functional modules.

E. COLI TRANSCRIPTION NETWORK

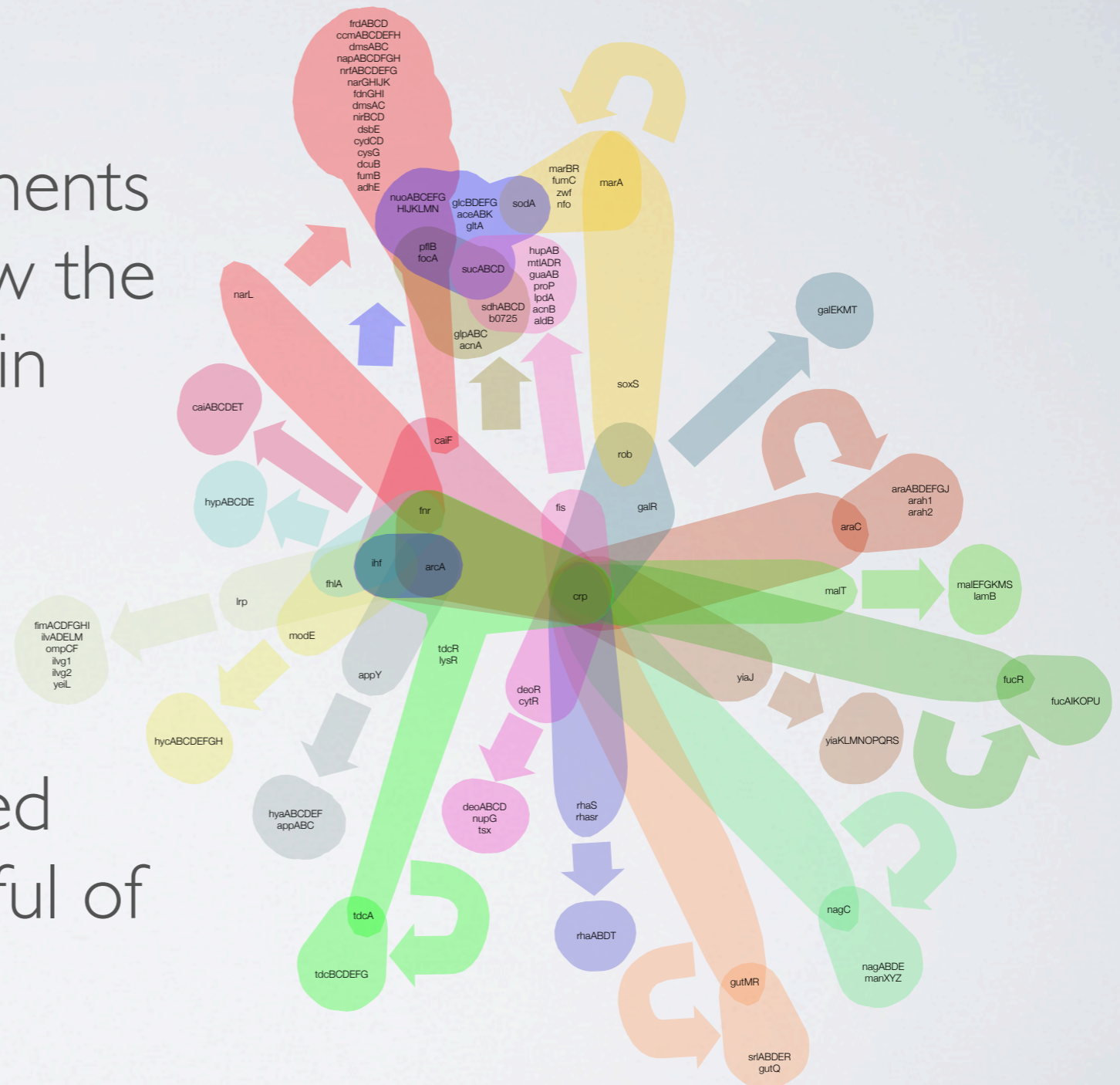
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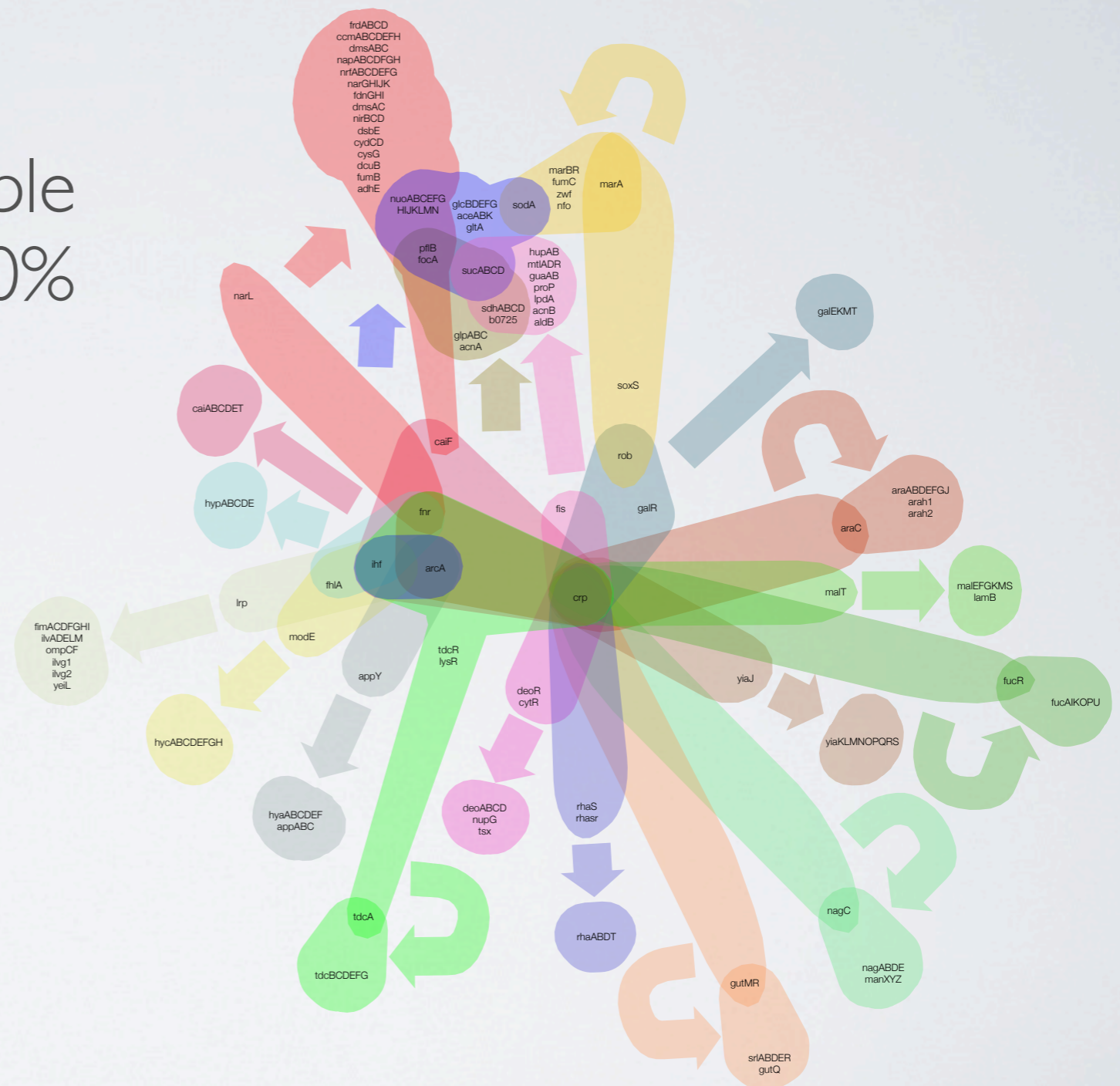
The compressible components of the *E. coli* network show the organisation of the genes in terms of operons.

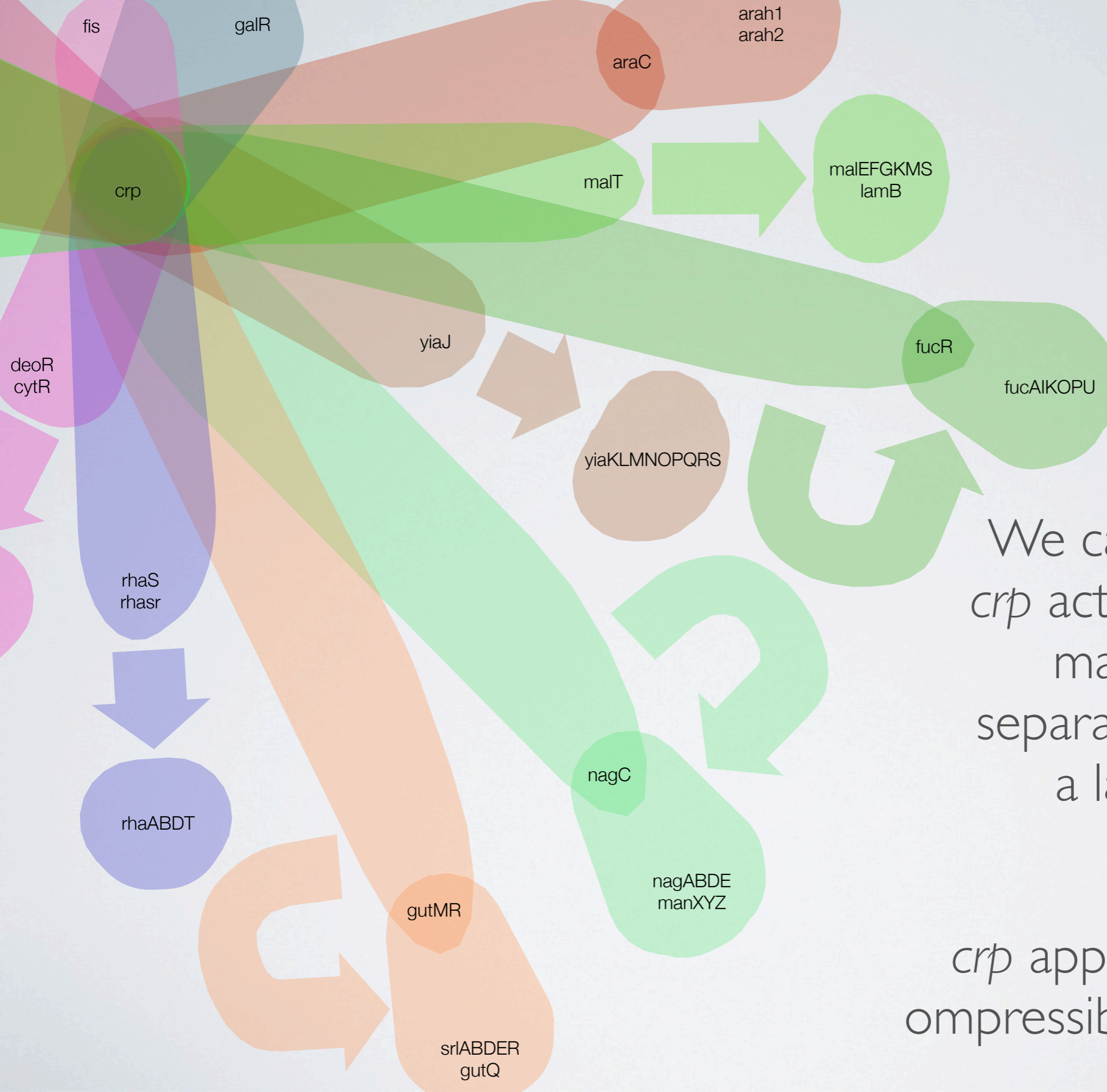
The operons are controlled combinatorially by a handful of key regulatory genes.



E. COLI TRANSCRIPTION NETWORK

In the 20 most compressible components (pictured), 70% of power nodes exhibited significant GO term enrichment.





We can also see that *crp* acts together with many other genes separately to regulate a large number of operons.

crp appears in 29 of 62 compressible components.

GLOBAL EFFICIENCY

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Global efficiency of a network is defined as the inverse of the harmonic mean of the shortest path length across all possible node pairs in the network.

$$E_{global} = \frac{1}{N(N-1)} \sum_{i,j} \frac{1}{L_{ij}}$$

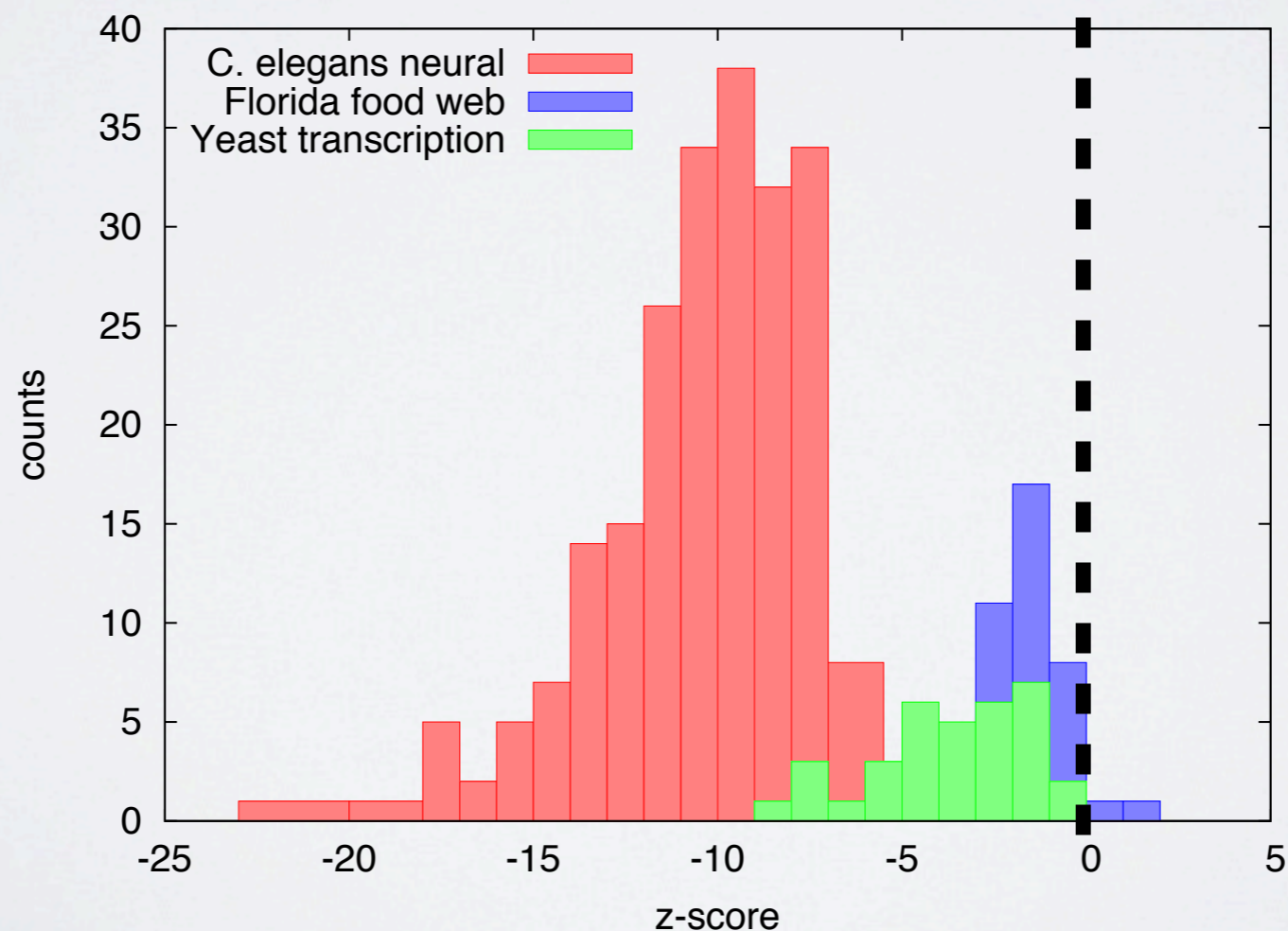
What role do compressible components play in terms of global efficiency?

GLOBAL EFFICIENCY

If we remove compressible components, the global efficiency is almost always reduced more than for an equally sized set of randomly chosen edges.

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FUTURE DIRECTIONS

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One possible future direction would be to introduce the possibility of imperfect power edges, in which the compression information would include small numbers of edges that are missing from an even more compressible component.

Another would be to construct inference approaches which allow the classification of unknown nodes based on term enrichment and their membership in compressible components.

CONCLUSIONS

Compressible components provide a way to identify and visualise dominant connection structures in networks.

This allows the identification of node sets that are defined by their relationships with other node sets, rather than internal connectivity of sets.

The potentially highly overlapping node sets that emerge from this approach can be classified using term enrichment analysis.

CONCLUSIONS

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In transcription networks this approach identifies sets of genes that regulate other sets of genes in combination.

GO term enrichment can be used to assign biological meaning to these overlapping sets.

The partial overlaps of power nodes are a crucial characteristic for the description of transcription networks.

PUBLICATIONS

PUBLICATIONS

The work on transcription networks
has been published in:

S. E. Ahnert, Mol. Biosyst. 9, 2681 (2013)

A more general publication is currently under review.

ACKNOWLEDGMENTS

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(Cambridge University)

Siobhan Brady and Mallorie Taylor-Teeple
(University of California at Davis)

THANK YOU

for your attention!