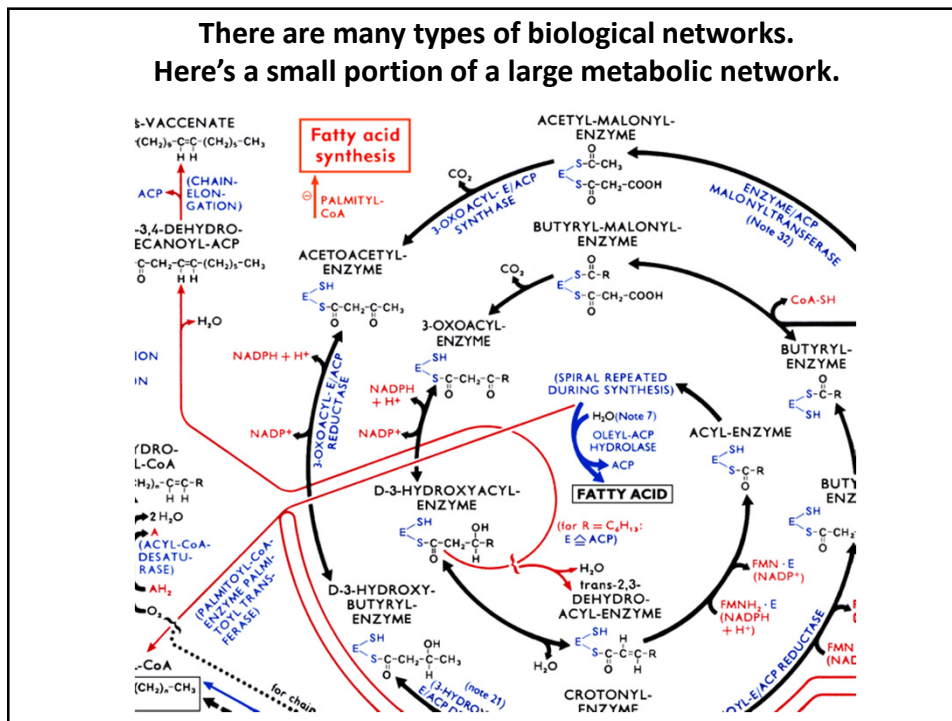
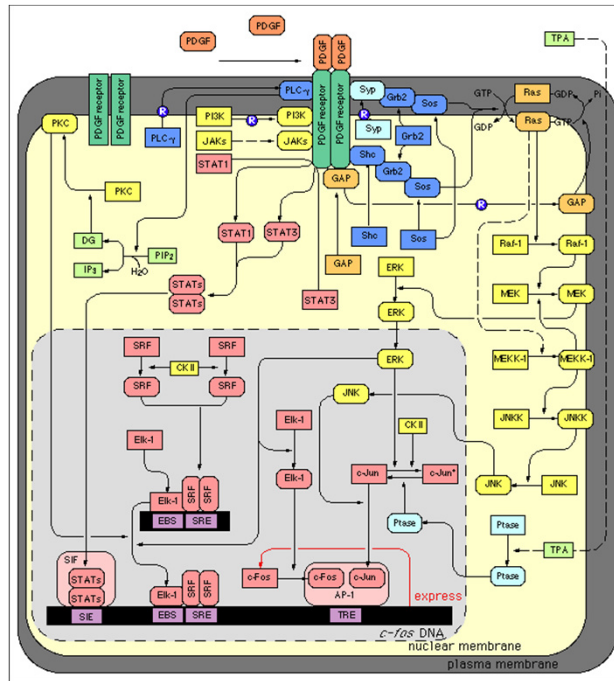


Network biology (& predicting gene function)

BCH394P/364C Systems Biology / Bioinformatics
Edward Marcotte, Univ of Texas at Austin



A typical
genetic
network

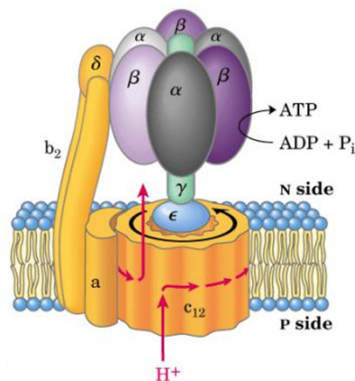


Contacts between proteins define protein interaction networks

X-ray structure
of ATP synthase

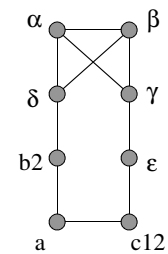


Schematic
version



Total set = protein complex
Sum of **direct** + **indirect**
interactions

Network
representation

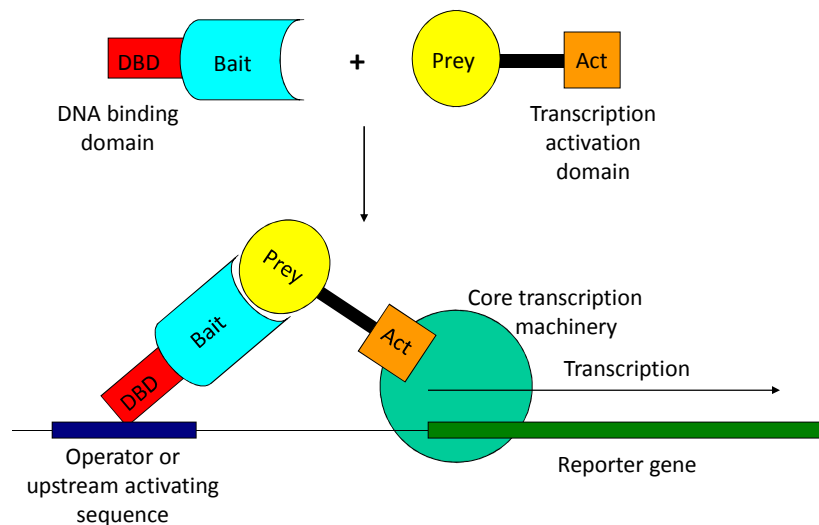


Let's look at some of the types of interaction data in more detail.

Some of these capture physical interactions, some genetic, some informational or logical.

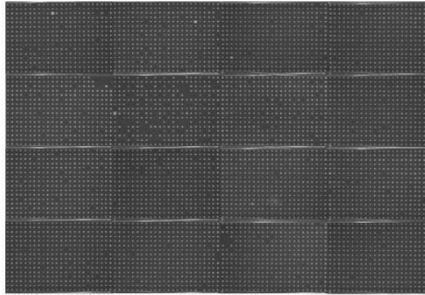
Pairwise protein interactions

In general, purifying proteins one at a time, mixing them, and assaying for interactions is far too slow & laborious. We need something faster! Hence, high-throughput screens, e.g. yeast two-hybrid assays

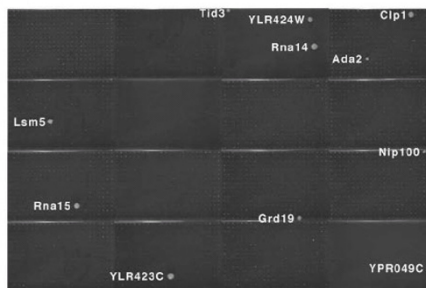


High-throughput yeast two-hybrid assays

Haploid yeast cells expressing activation domain-prey fusion proteins



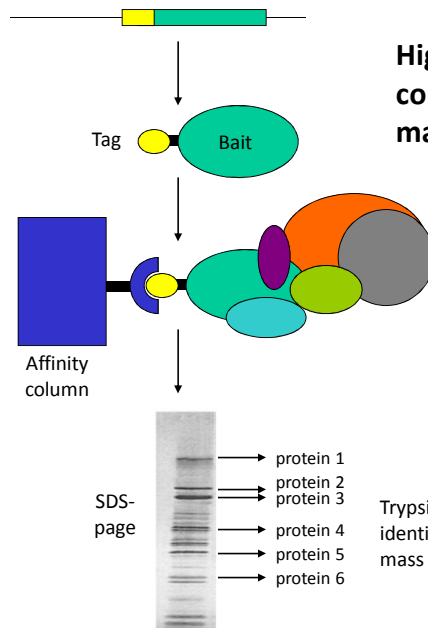
Diploid yeast probed with DNA-binding domain-Pcf11 bait fusion protein

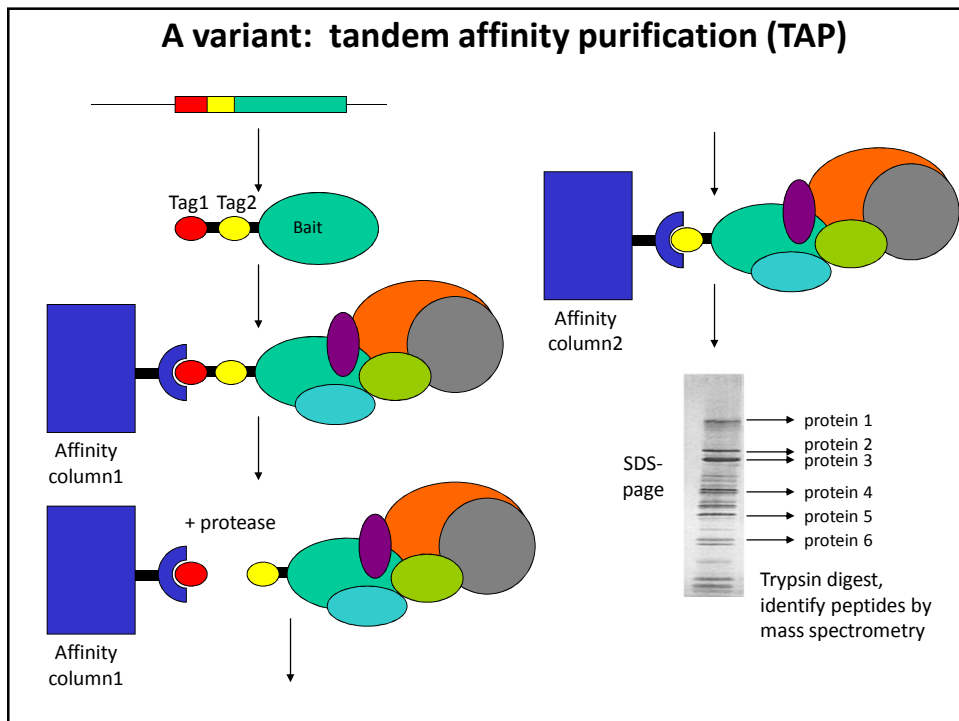
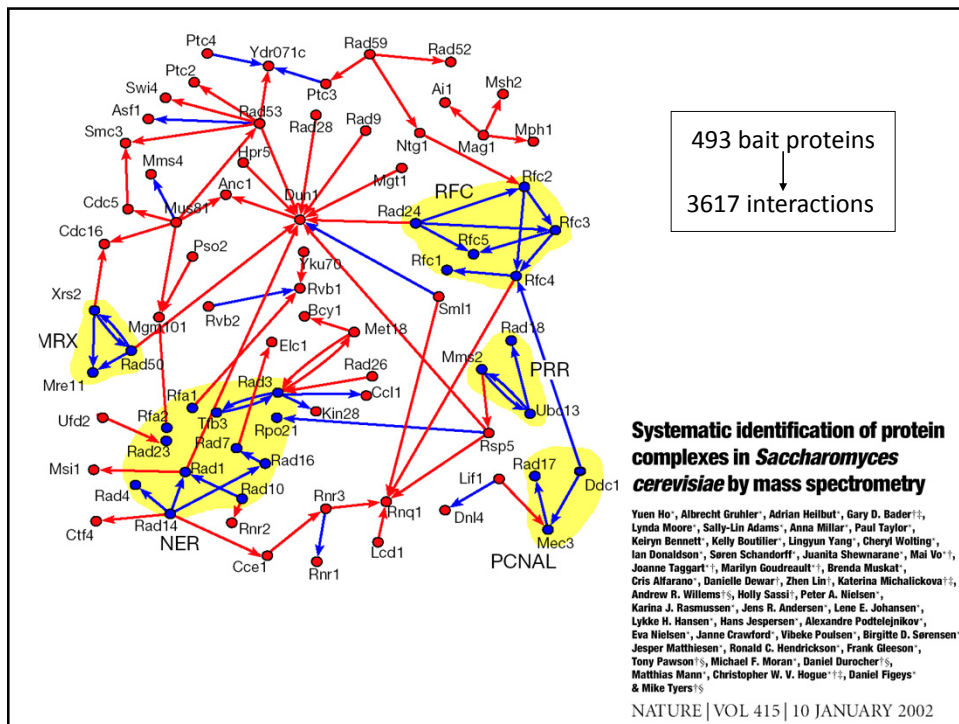


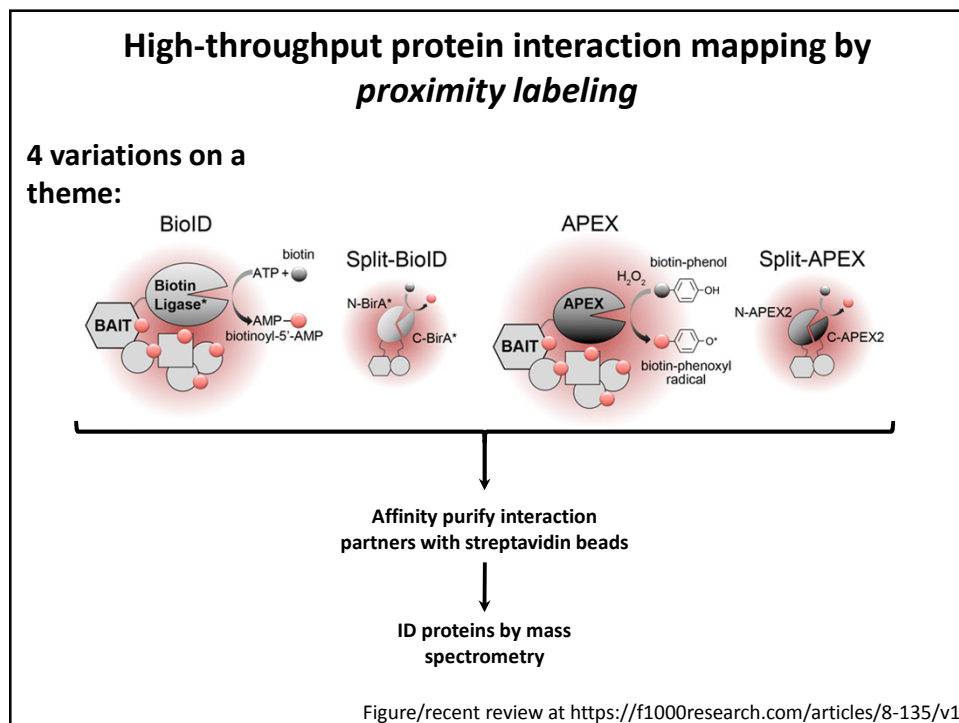
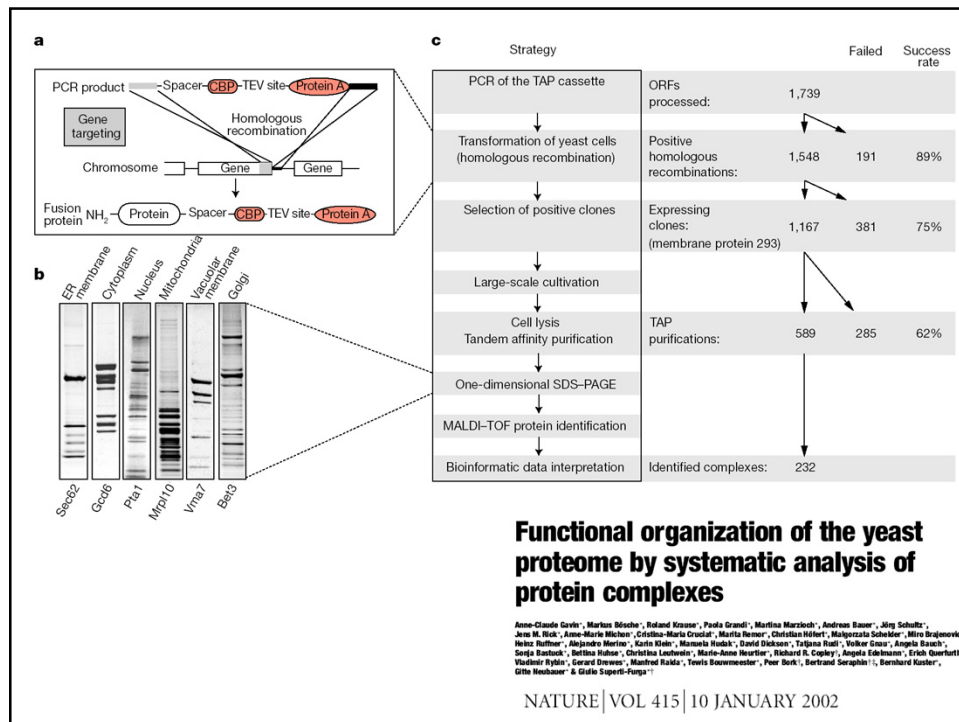
Uetz, Giot, *et al. Nature* (2000)

Protein complexes

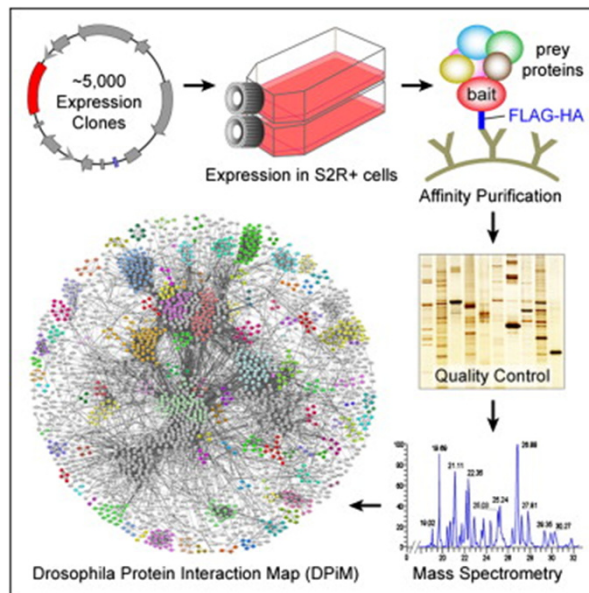
High-throughput complex mapping by mass spectrometry







The current state-of-the-art in animal PPI maps – AP/MS



~3,500 affinity purification experiments

~11K interactions /
~2.3K proteins

→ spans 556 complexes

Still daunting for the
human proteome, but...

Guruharsha *et al.* (2011) *Cell* 147, 690–703

The current state-of-the-art in human PPI maps – Y2H (< 1 week old!)

Human ORFeome (v9.1) → now ~90% of the protein-coding genes!

Screened *all x all* (150M pairs!) in 9 Y2H assays

52,569 PPIs involving 8,275 proteins

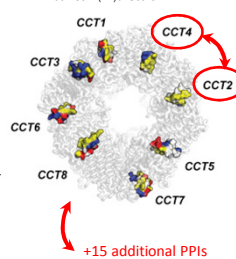
<https://www.nature.com/articles/s41594-017-0016-2>



Y2H captures pairwise PPIs that can form when the proteins are expressed out of biological context (e.g., as fusion proteins in a yeast cell nucleus). It can reveal directly contacting proteins but often misses those that require additional molecular context or higher order assemblies.

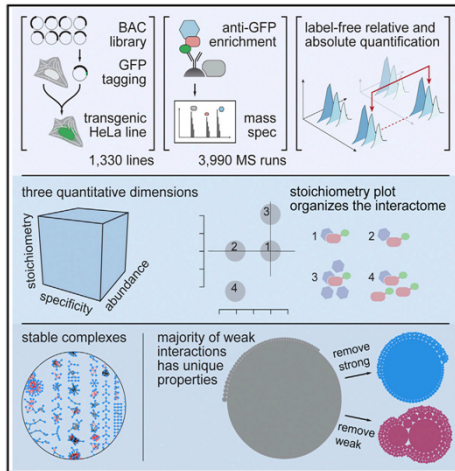
← the exocyst e.g. the CCT complex →

[https://www.cell.com/cell/fulltext/S0092-8674\(14\)01369-5](https://www.cell.com/cell/fulltext/S0092-8674(14)01369-5)

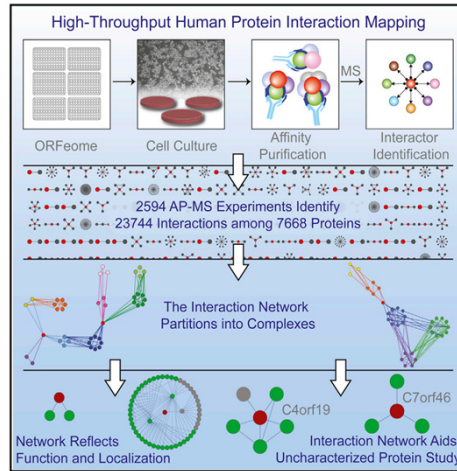


Luck *et al.*, A reference map of the human protein interactome, *bioRxiv*, posted April 10, 2019
<https://www.biorxiv.org/content/10.1101/605451v1>

The current state-of-the-art in human PPI maps – large scale AP/MS



Hein *et al.*, *Cell* (2015) 163:712-23.

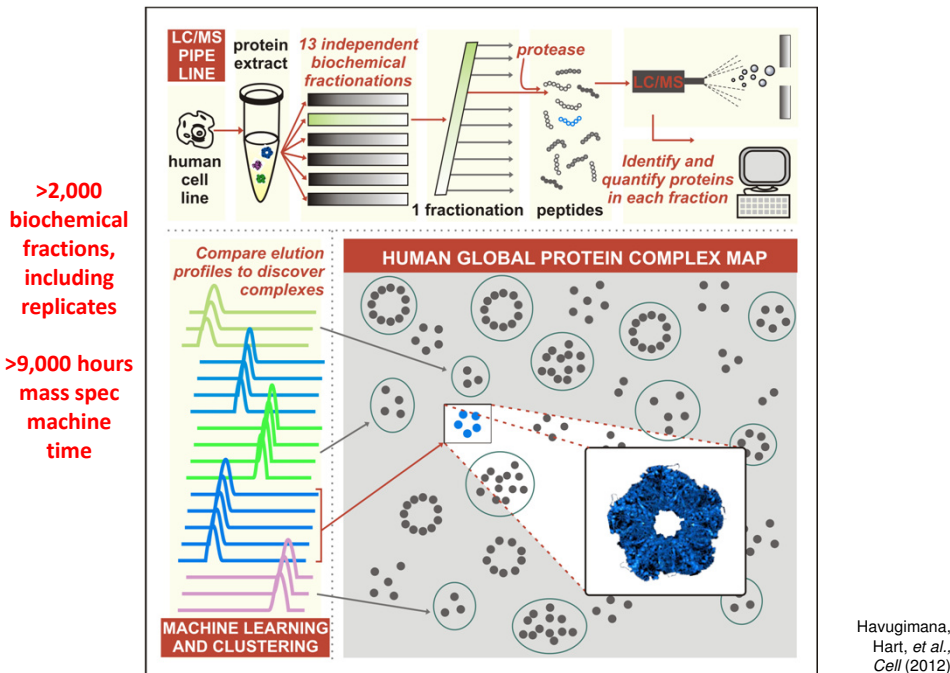


Huttlin *et al.*, *Cell* (2015) 162:425-440

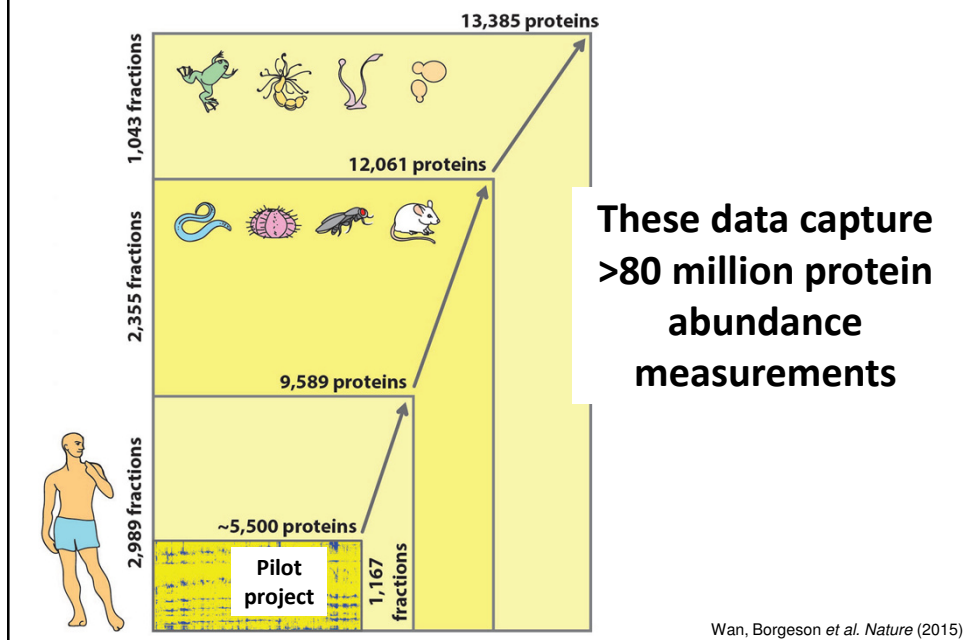
Huttlin *et al.*, *Nature* (2017) 545:505-509

Just in the past 3 years, nearly 6K affinity purification experiments on tagged human proteins expressed in cell lines

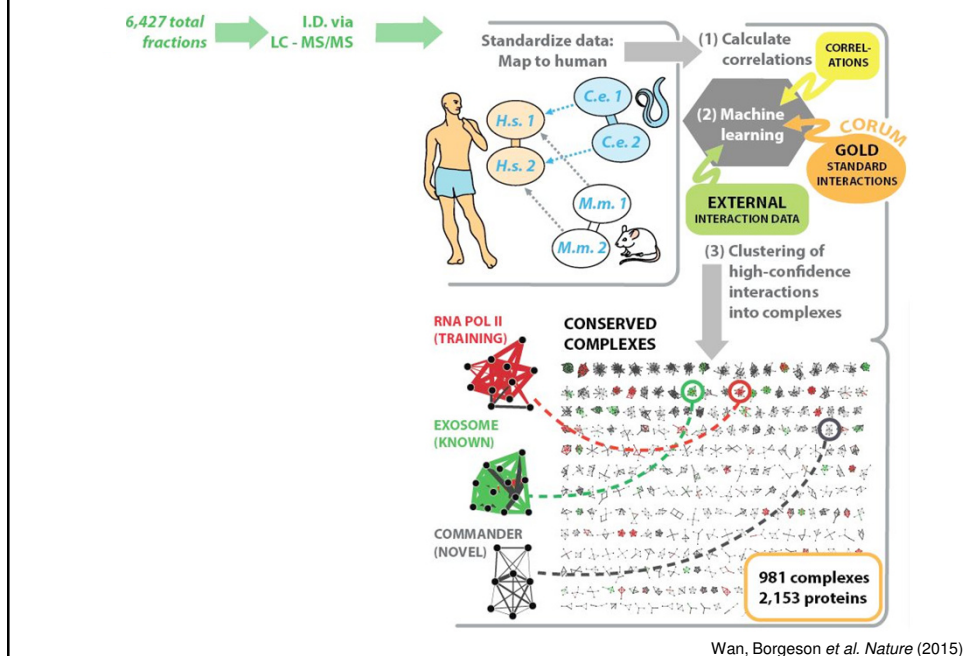
The current state-of-the-art in animal PPI maps – co-fractionation/MS



Now >6,400 CF/MS experiments across animals

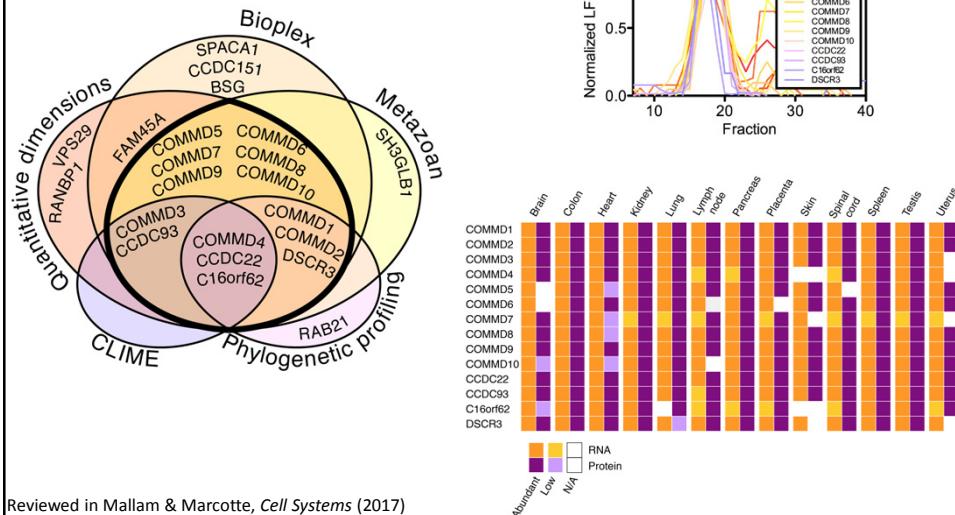


Extending the map across animals...



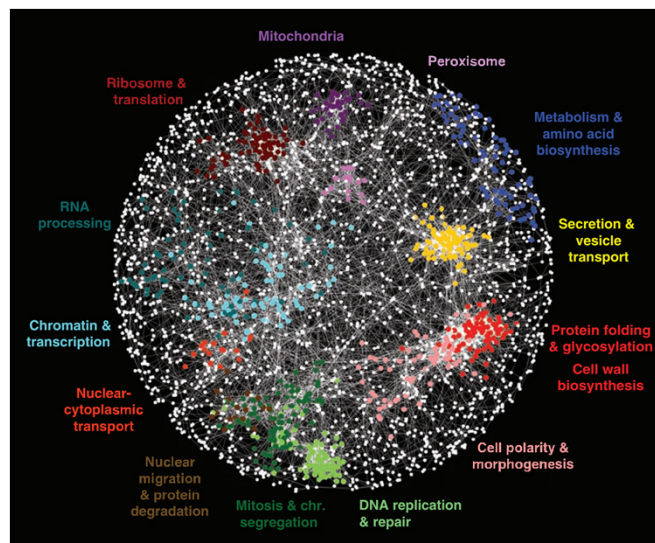
There are still lots of cellular machines left to find

e.g. the “Commander” complex, found in all 3 recent human PPI maps, a 600 kDa protein complex expressed in nearly every human cell type and tissue



Genetic interactions

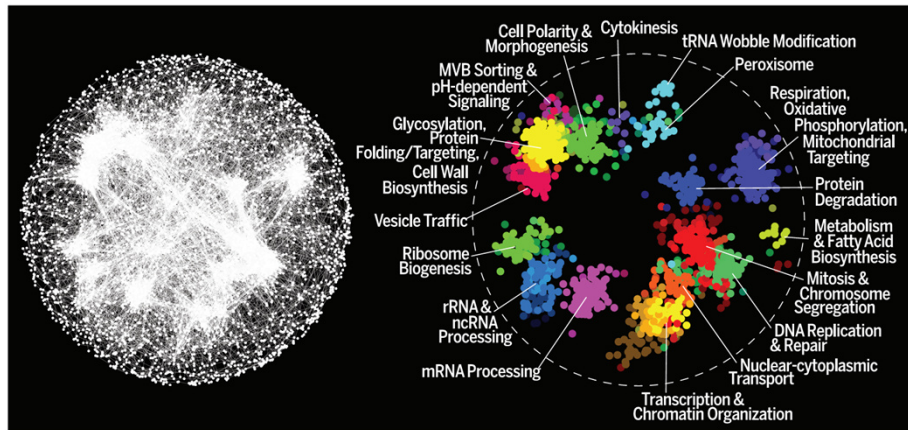
5.4 million gene-gene pairs assayed for synthetic genetic interactions in yeast



Costanzo et al., *Science* 327: 425 (2010)

Genetic interactions, the 2016 version

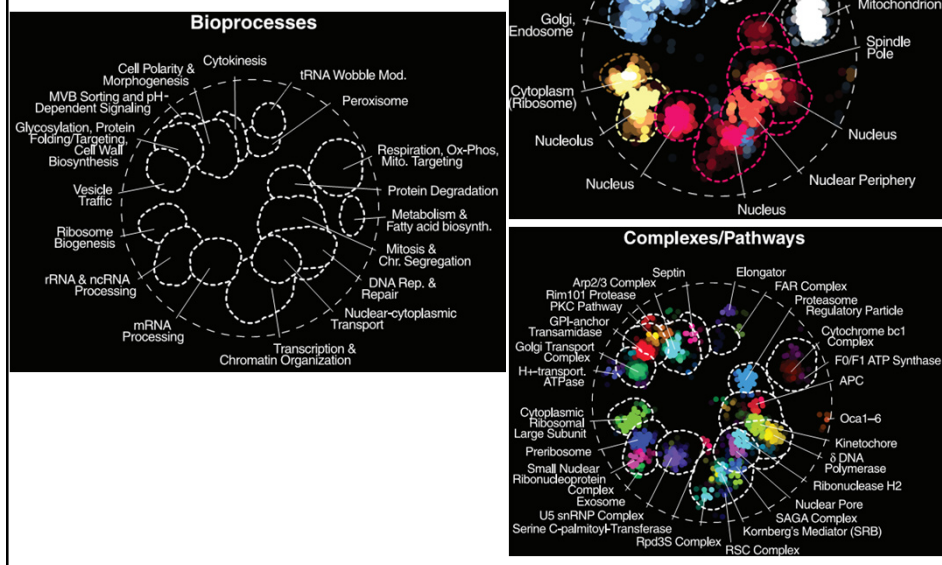
23 million gene-gene pairs assayed for synthetic genetic interactions in yeast, identifying ~550,000 negative and ~350,000 positive genetic interactions

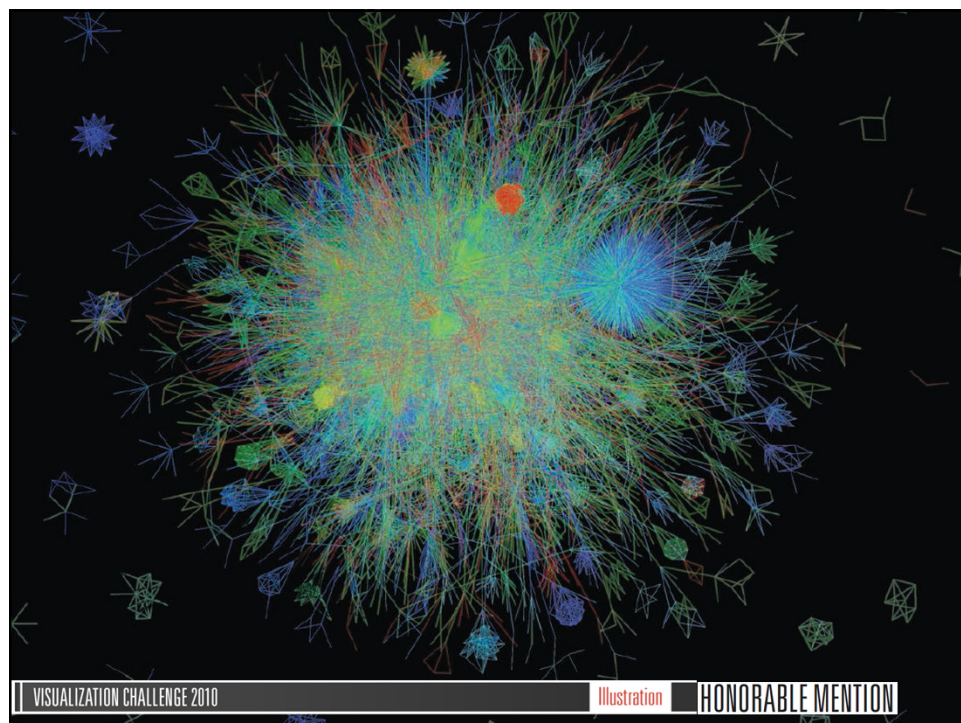
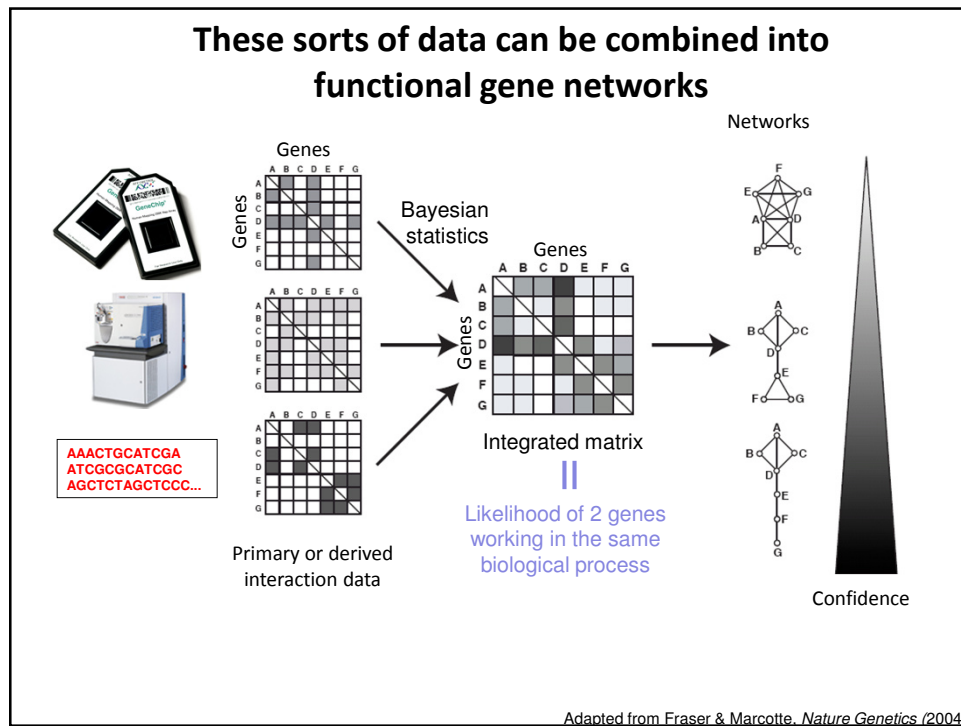


A global network of genetic interaction profile similarities. (Left) Genes with similar genetic interaction profiles are connected in a global network, such that genes exhibiting more similar profiles are located closer to each other, whereas genes with less similar profiles are positioned farther apart. (Right) Spatial

Costanzo *et al.*, *Science* 353: 1381 (2016)

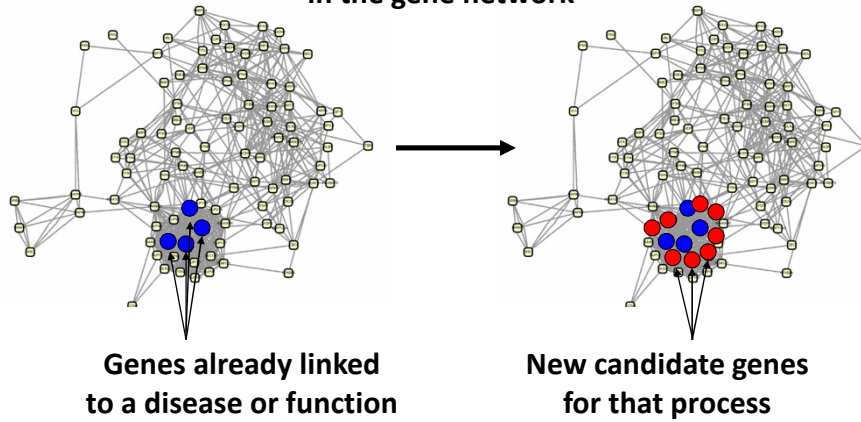
The global genetic interaction profile similarity network reveals a hierarchy of cellular function.



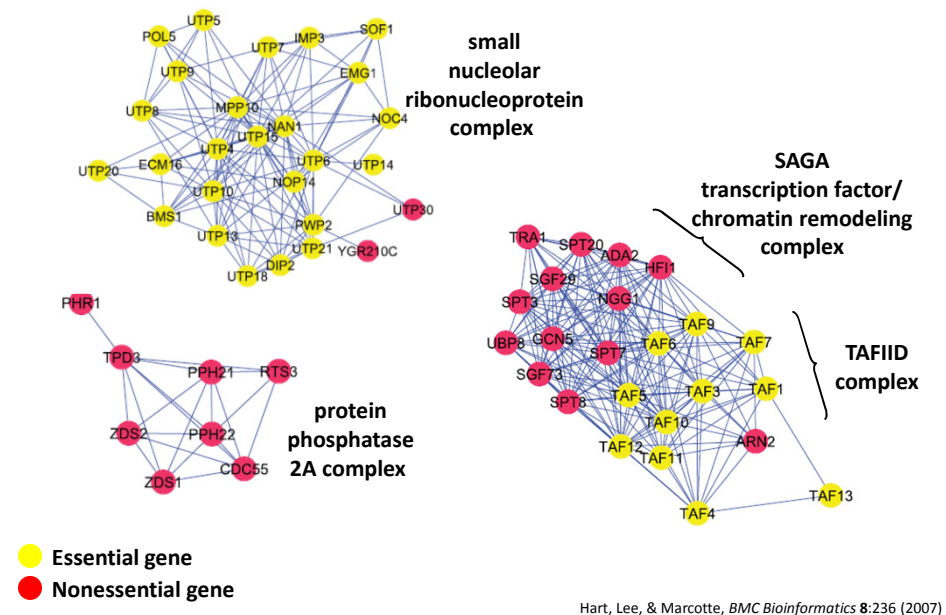


These networks are hypothesis generators.
Given a gene, what other genes does it function with?
What do they do?

**Guilt-by-association
in the gene network**

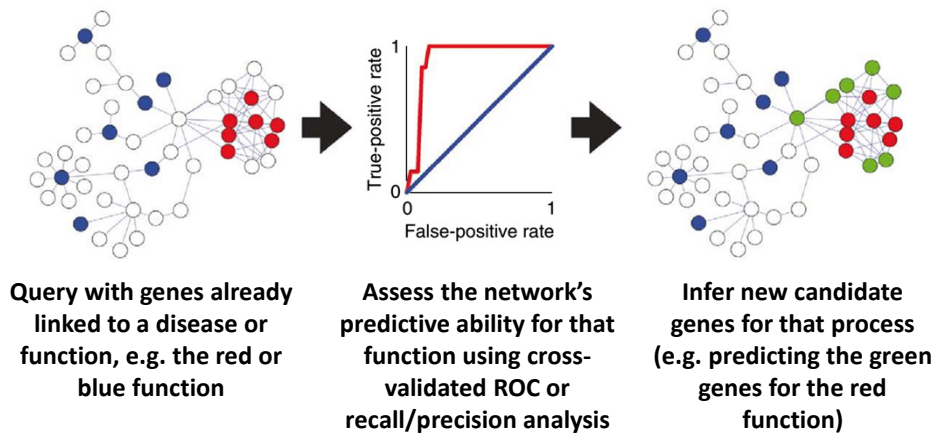


Gene networks frequently reflect functions, pathways, & phenotypes,
e.g., lethality in yeast is linked to the molecular machine, not the gene



We can propagate annotations across the graph to infer new annotations for genes (network “guilt-by-association”, or GBA).

Testing how well this works on hidden, but known, cases let’s us measure how predictive it will be for new cases.

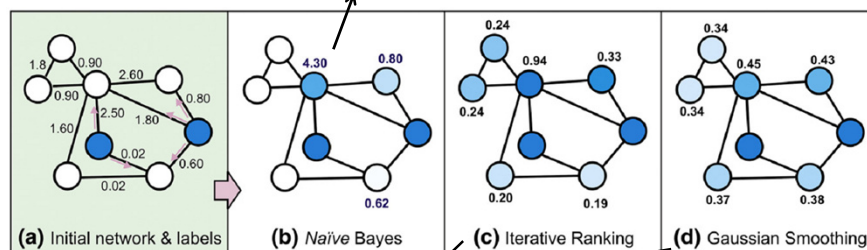


Lee, Ambaru et al. *Nature Biotechnology* 28:149-156 (2010)

Numerous algorithms exist for network GBA

Similar to Google’s personalized PageRank

Naïve Bayes assigns scores to neighboring nodes based on edges



Network diffusion algorithms start with initial annotations and the graph topology, then propagate initial scores across the network, e.g. Gaussian smoothing tries to find scores:

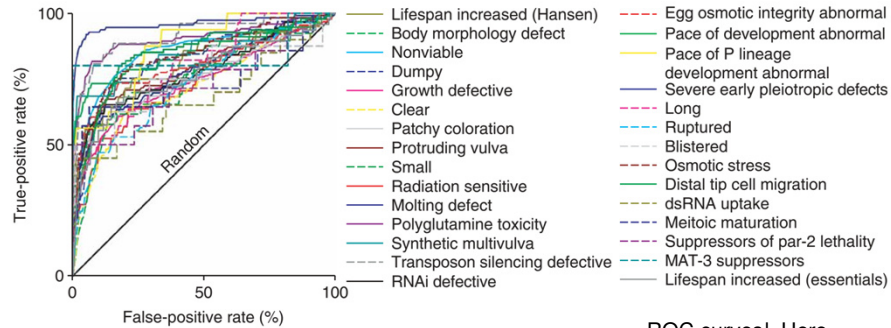
$$f^{final} = \underset{f}{\operatorname{argmin}} \alpha \sum_i (f_i - f_i^0)^2 + (1 - \alpha) \sum_i \sum_j w_{ij} (f_i - f_j)^2$$

minimizing the difference between final and initial scores of a protein

& between a protein’s score and that of each of its neighbors

Reviewed in Wang & Marcotte, *J Proteomics* (2010)

For example, predicting genes linked with worm phenotypes in genome-wide RNAi screens



ROC curves! Here, indicating the likely predictive power of the network for a system of interest, independent of how big the system is.

A poor ROC → no better than random guessing.

Lee, Lehner *et al.*, *Nat Genet*, 40(2):181-8 (2008)

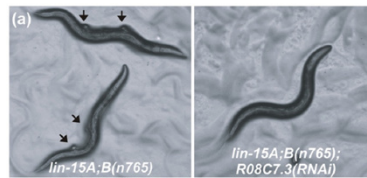
Remarkably, this strategy works quite well

Some examples of network-guided predictions:

In worms:

Genes that can reverse 'tumors' in a nematode model of tumorigenesis

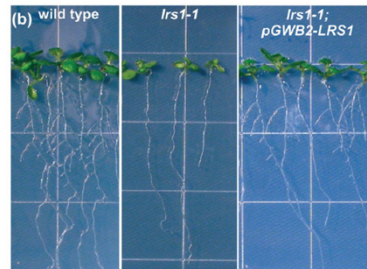
Lee, Lehner *et al.*
Nature Genetics (2008)



In Arabidopsis:

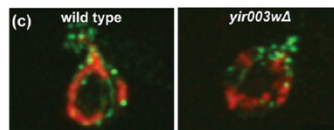
New genes regulating root formation

Lee, Ambaru *et al.*
Nature Biotech (2010)



In yeast: New mitochondrial biogenesis genes

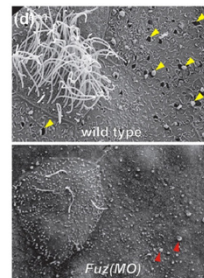
Hess *et al.*, *PLoS Genetics* (2009)



In mice/frogs:

Functions for a birth defect gene

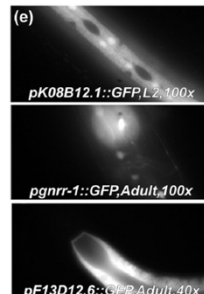
Gray *et al.*, *Nature Cell Biology* (2009)



In worms:

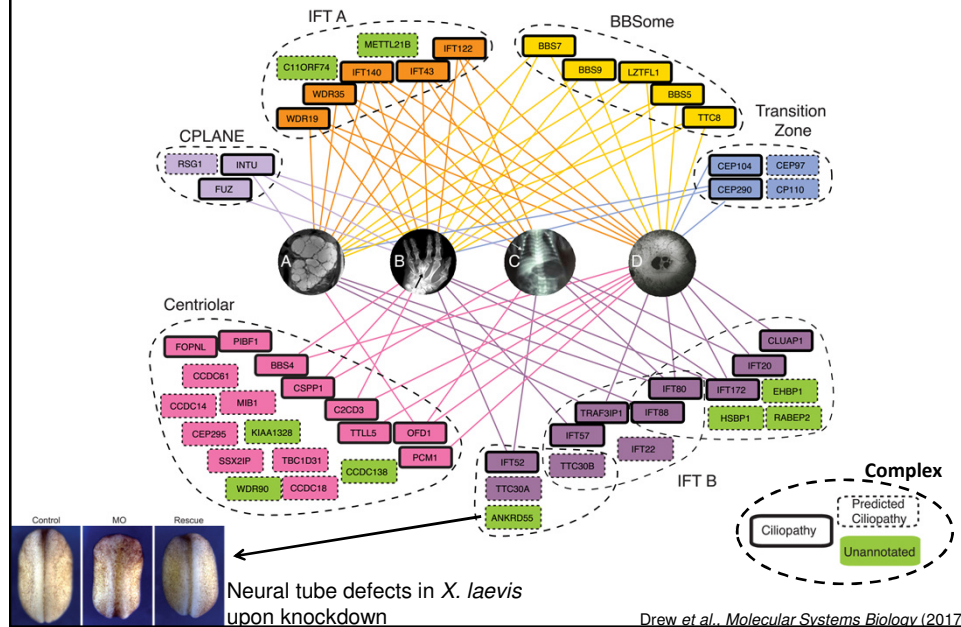
Predicting tissue specific gene expression

Chikina *et al.*, *PLoS Comp Biology* (2009)



Reviewed in Wang & Marcotte, *J Proteomics* (2010)

We use this approach routinely in the lab, e.g. a recent example predicting new ciliopathy genes from protein complexes



**Live demo of
STRING, BioGRID,
GeneMania,
functional networks
and Cytoscape**