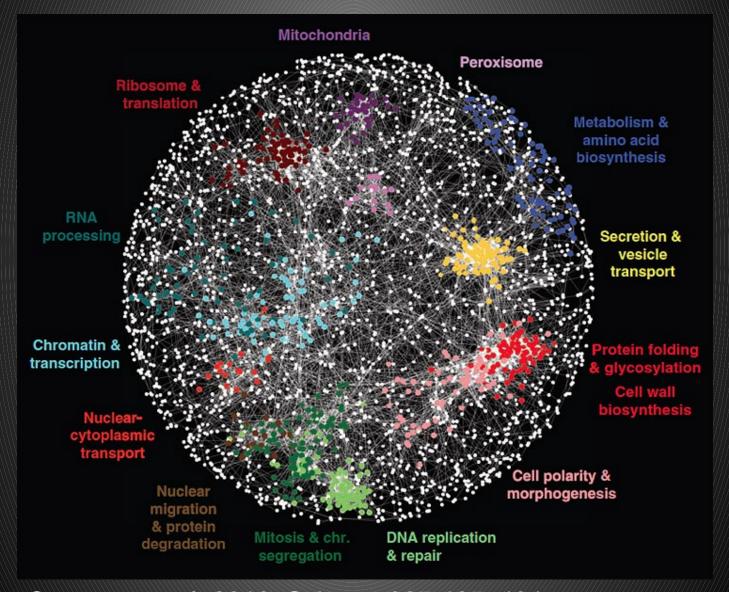
Graph Theory in Computational Biology

Nathan Lazar
B.S. Mathematics - PSU
PhD candidate - OHSU

Overview

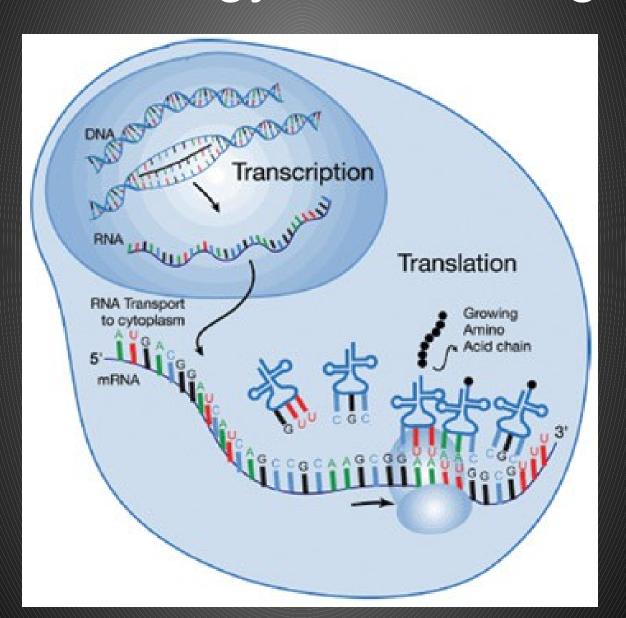
- Gene interaction networks
 - Biology
 - Modeling
 - Inferring
 - Measuring
 - Visualization
- Other networks
- Open questions

Gene interaction networks

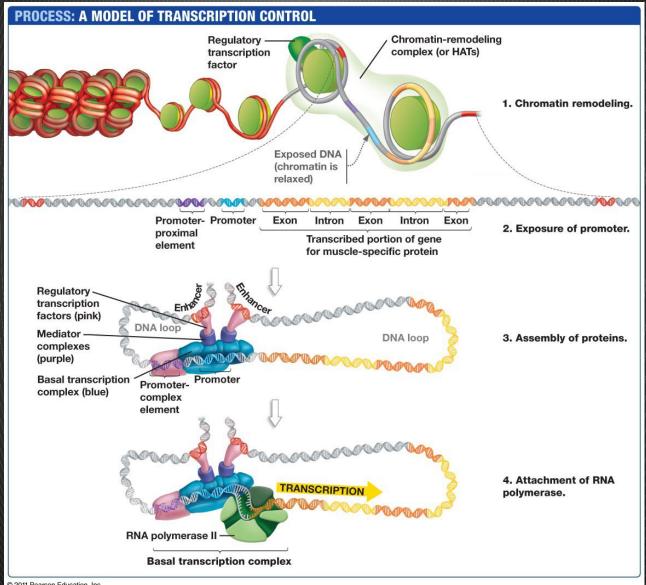


Costanzo et al. 2010. Science 327:425–431.

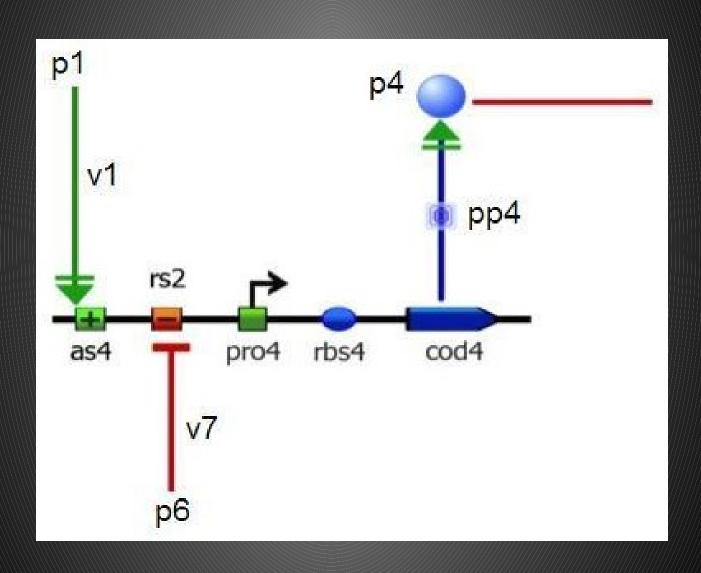
Some biology: central dogma



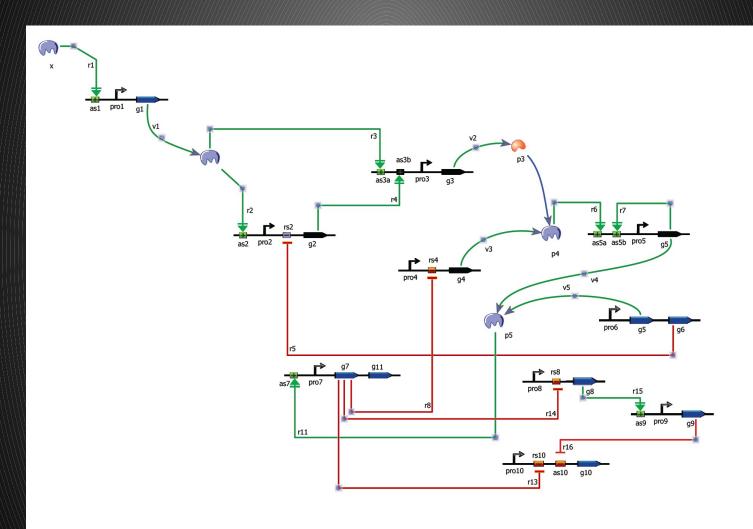
Transcription factors



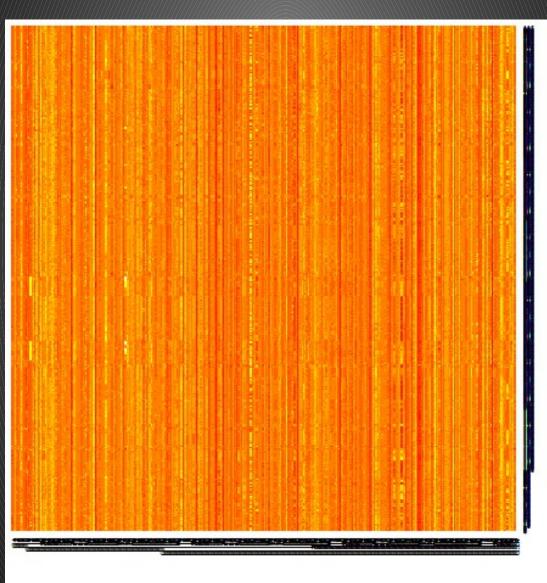
Modeling



Regulation

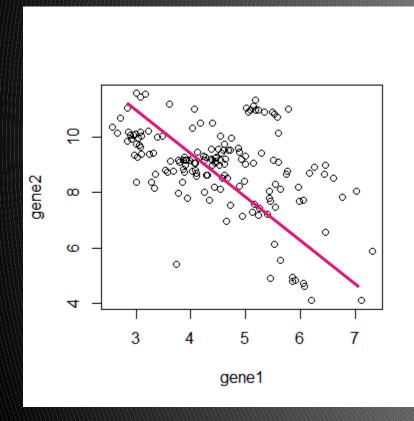


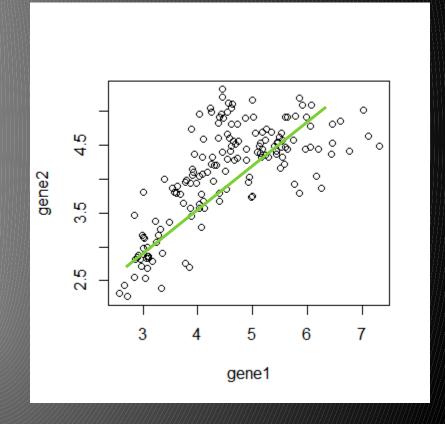
Transcription factor expression

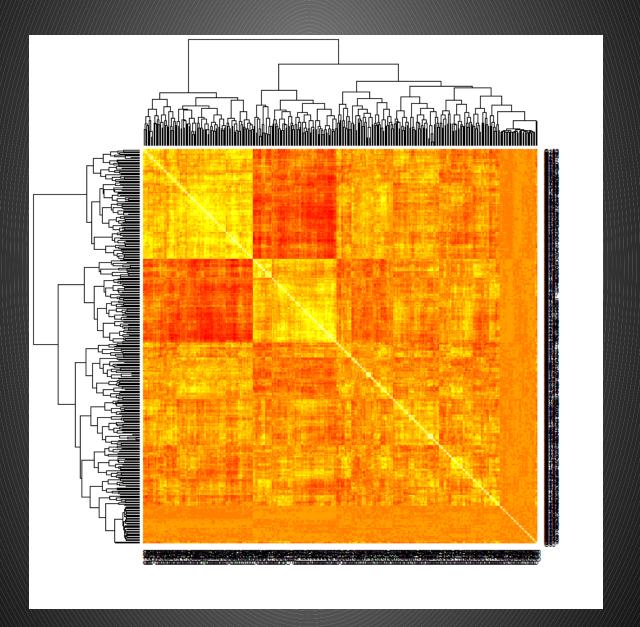


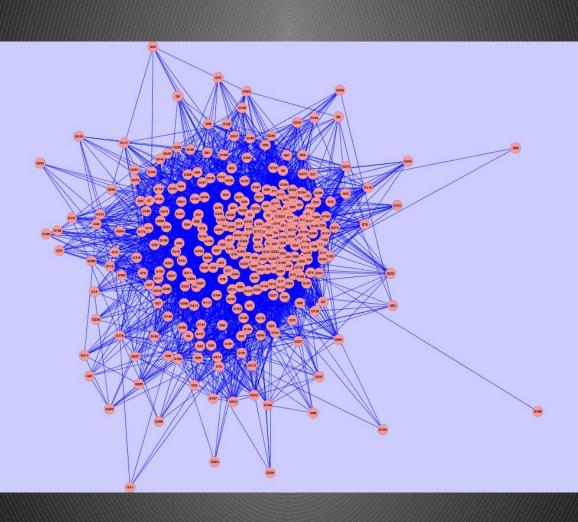
Measurements

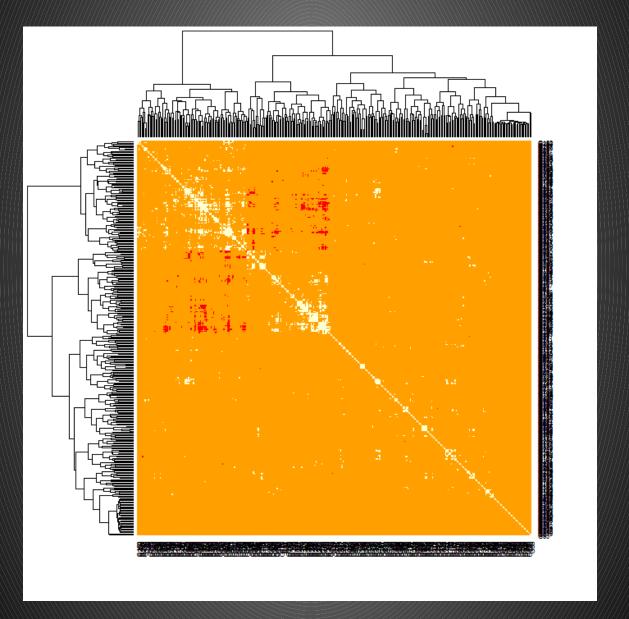
Correlation

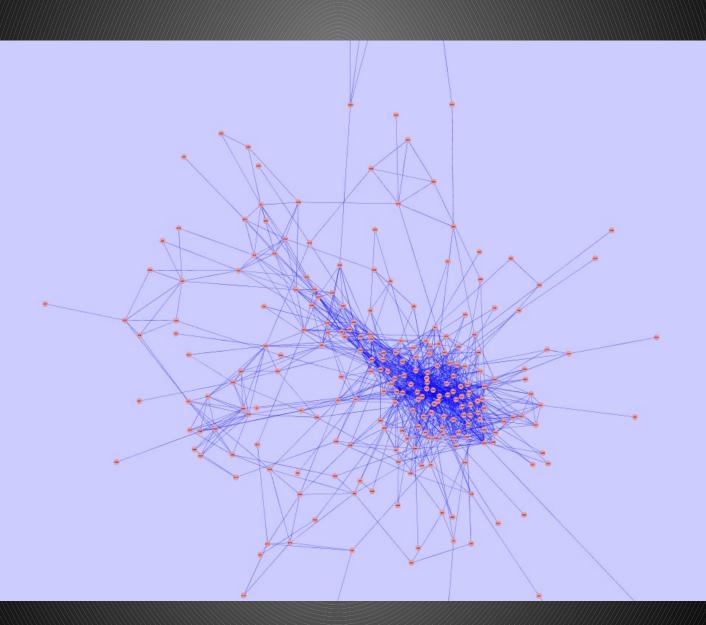








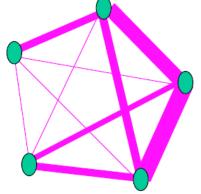




- Problems
 - You only know expression (mRNA or proteomic) at certain times
 - Missing nodes
 - Ton of other stuff going on
 - miRNA
 - Post-translational modifications
 - Spatial effects
 - No directionality

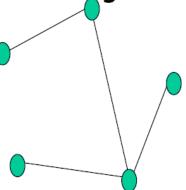
Weighted gene correlation network analysis (WGCNA)

Weighted Network View



- •All genes are connected
- Connection Widths=Connection strenghts

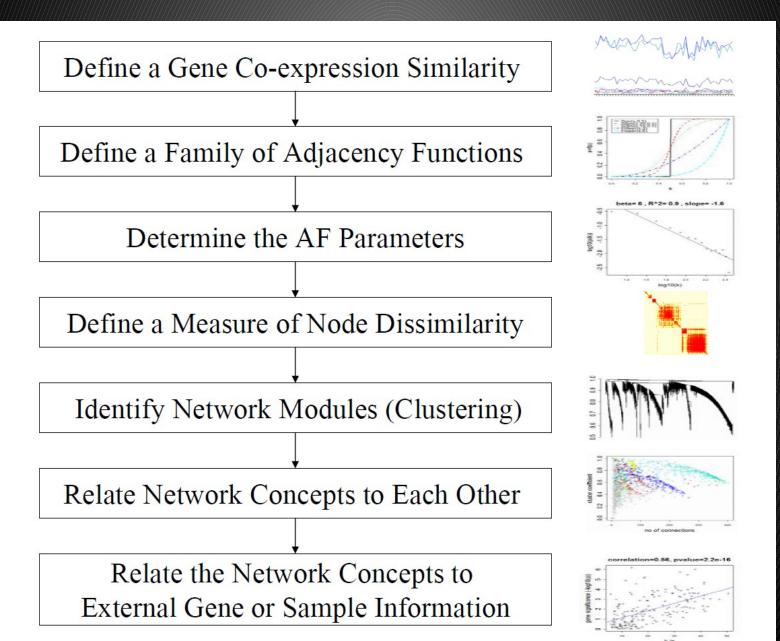
Unweighted View



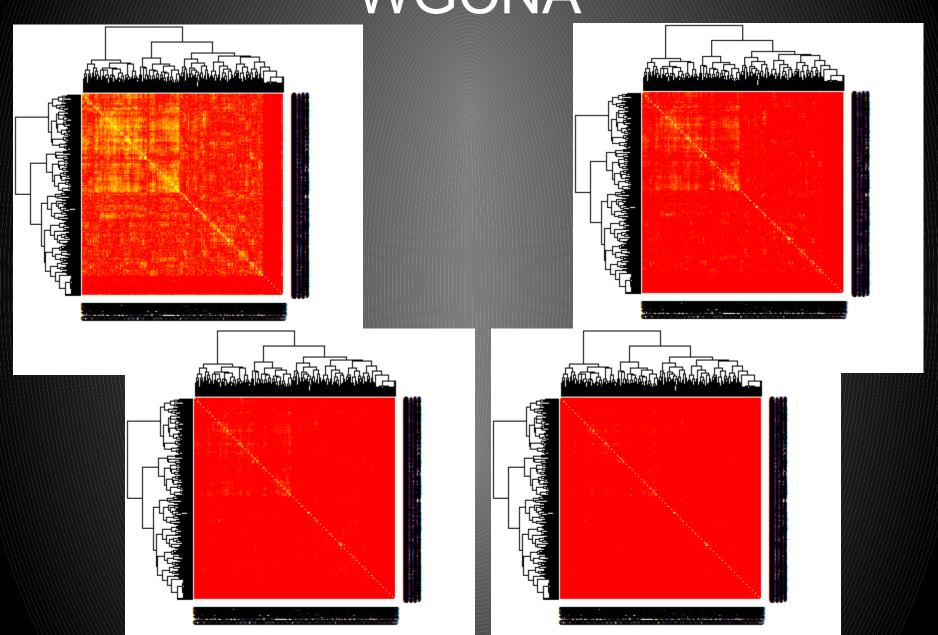
Some genes are connected All connections are equal

Zhang, B., and Horvath, S. (2005). A general framework for weighted gene coexpression network analysis. Stat. Appl. Genet. Mol. Biol. 4, Article17

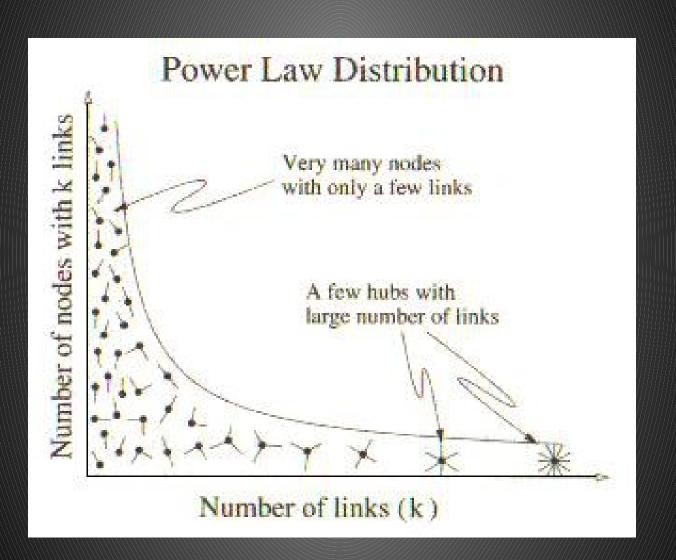
WGCNA



WGCNA



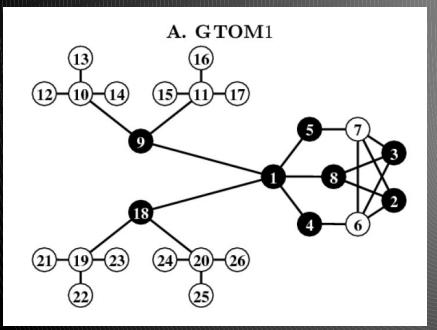
Scale free networks

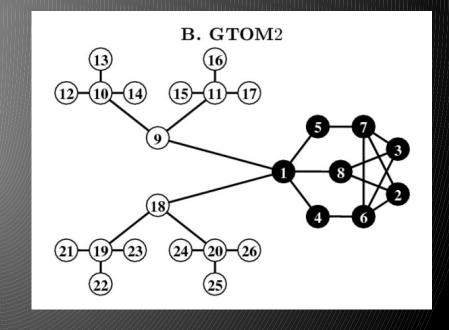


WGCNA node similarity

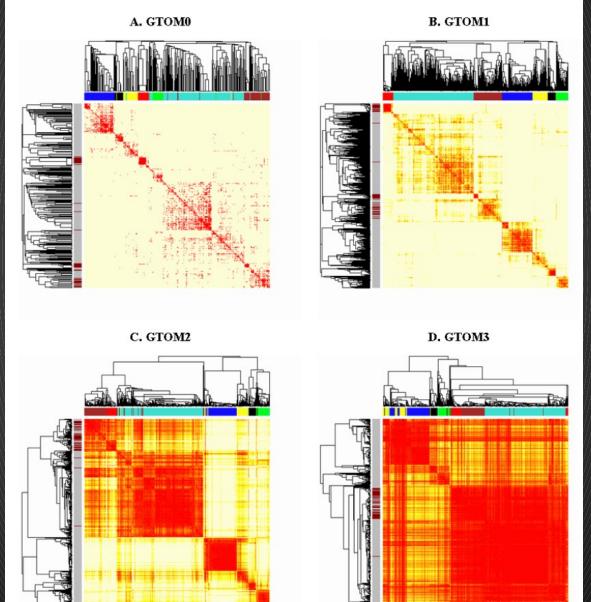
Generalized
topological
overlap
(Yip & Horvath 2007)

$$\begin{split} N_m(i) &:= \big\{ j \neq i \big| \mathrm{dist}(i,j) \leq m \big\} \\ t_{ij}^{[m]} &= \begin{cases} \frac{\big| N_m(i) \cap N_m(j) \big| + a_{ij}}{\min \big\{ \big| N_m(i) \big|, \big| N_m(j) \big| \big\} + 1 - a_{ij}} & \text{if } i \neq j \\ 1 & \text{if } i = j. \end{cases} \end{split}$$

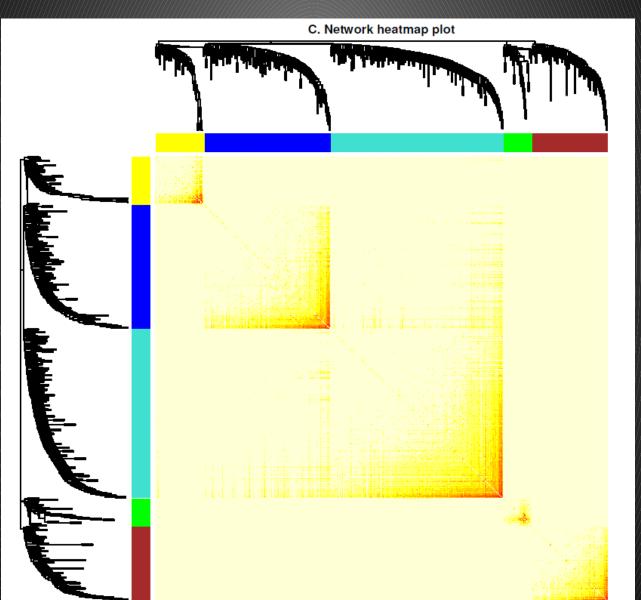




Generalized topological overlap

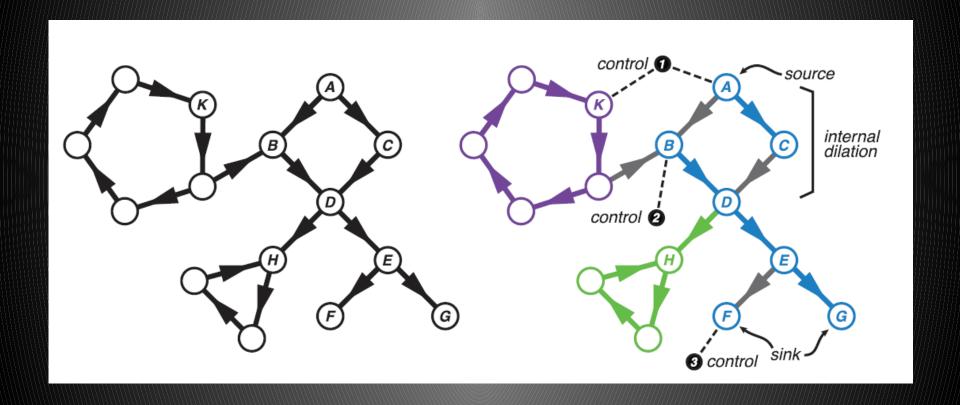


WGCNA

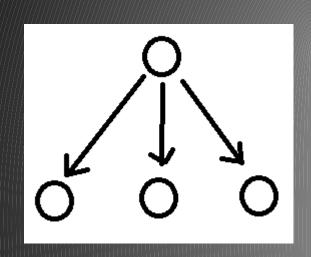


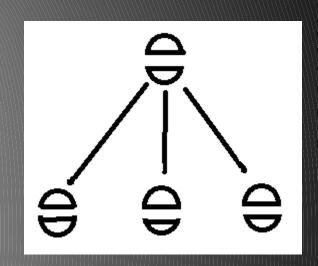
Other graph inference methods

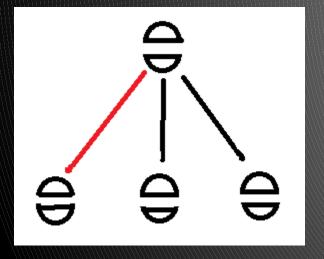
- Information theoretic
 - Use mutual information between nodes instead of correlation
- Regression based (directed graphs)
 - Random forests
 - Regularization using p-norms
- Bayesian inference
- Ensemble methods

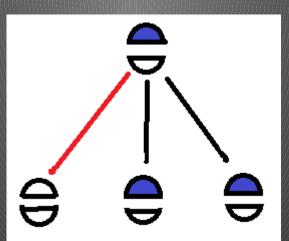


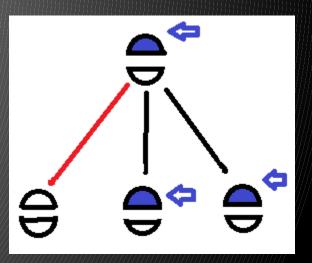
- Find control profiles by solving the 'maximum matching problem'
 - Split nodes into in and out nodes w/o connections between and get rid of directions
 - Find the maximum matching.
 - Largest set of edges w/o sharing a node
 - "In nodes" that don't have an edge from the maximal matching must be controlled.
- Look at what types of nodes need control.

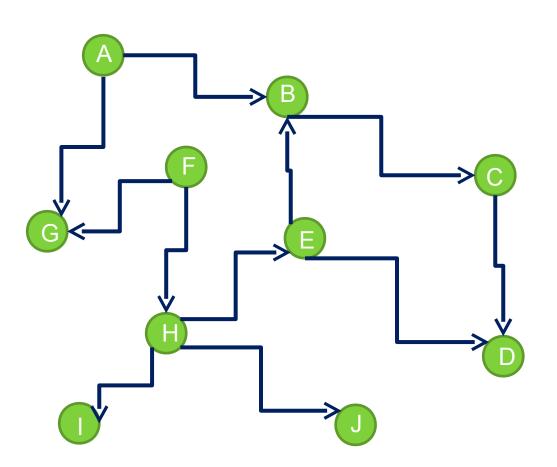


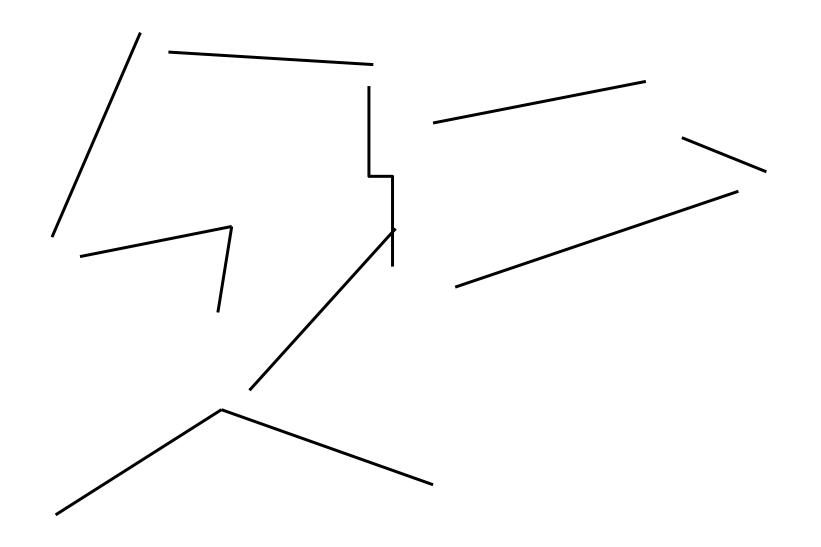


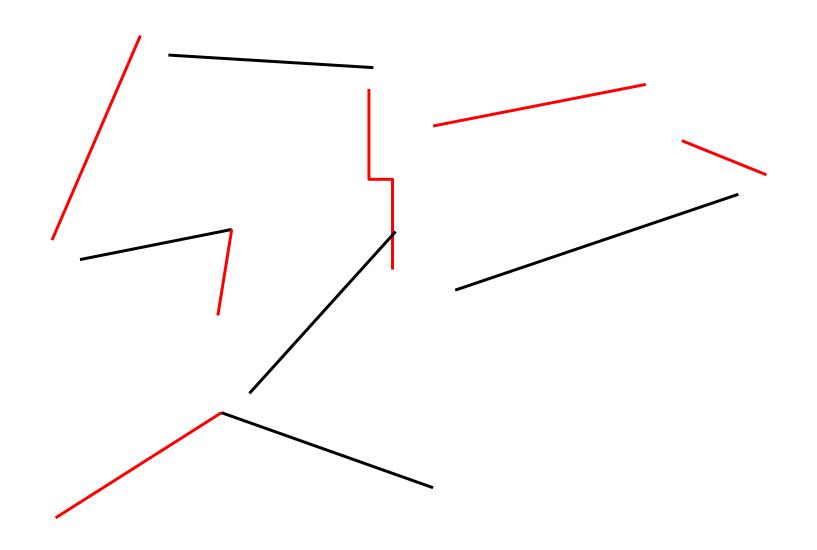


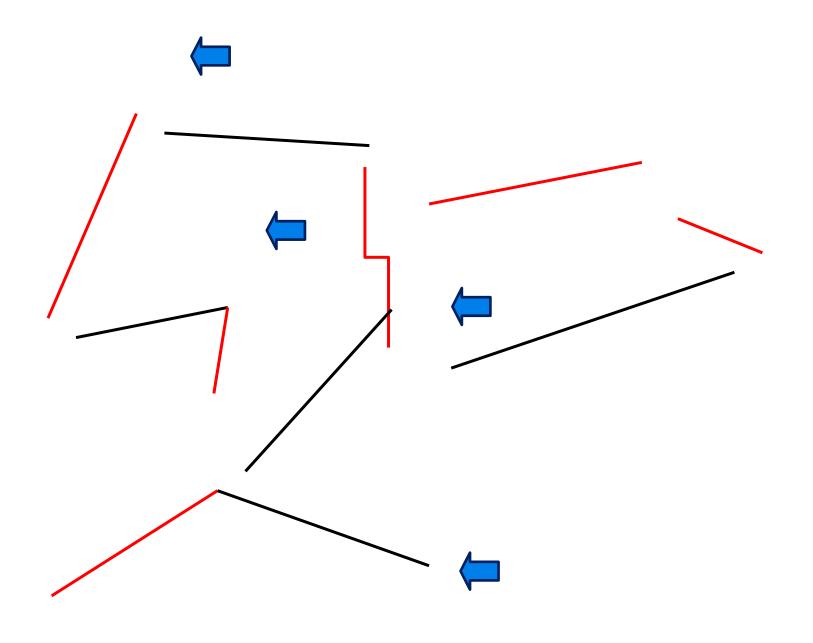


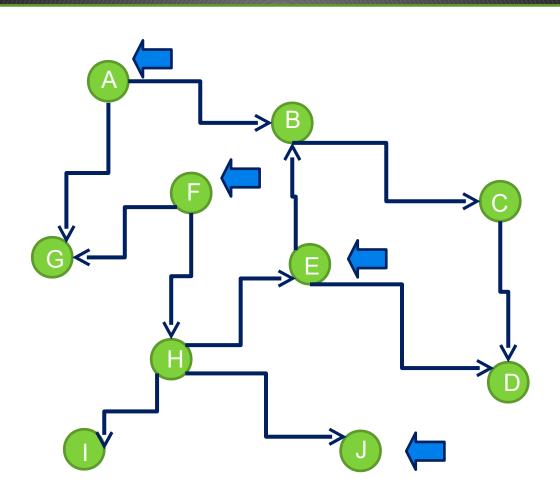


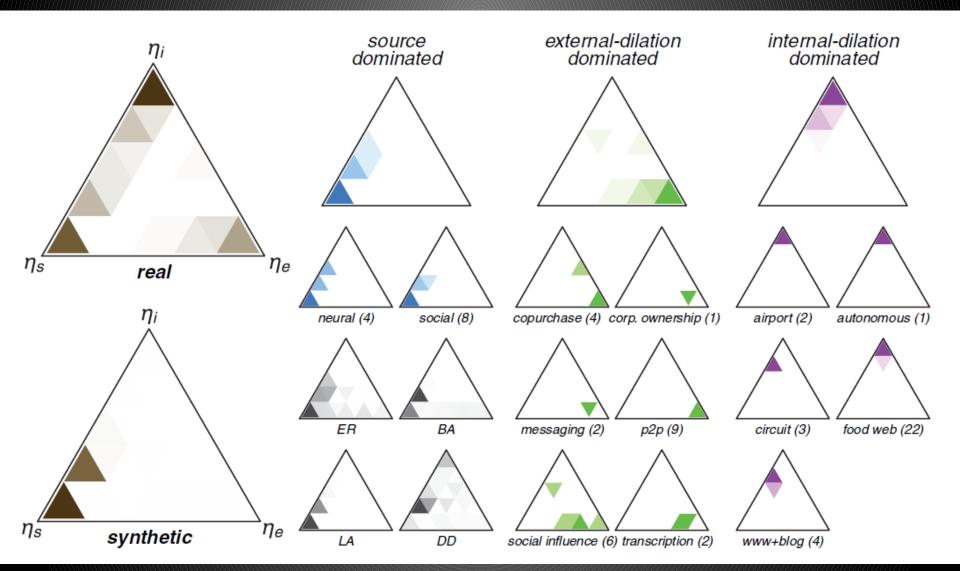




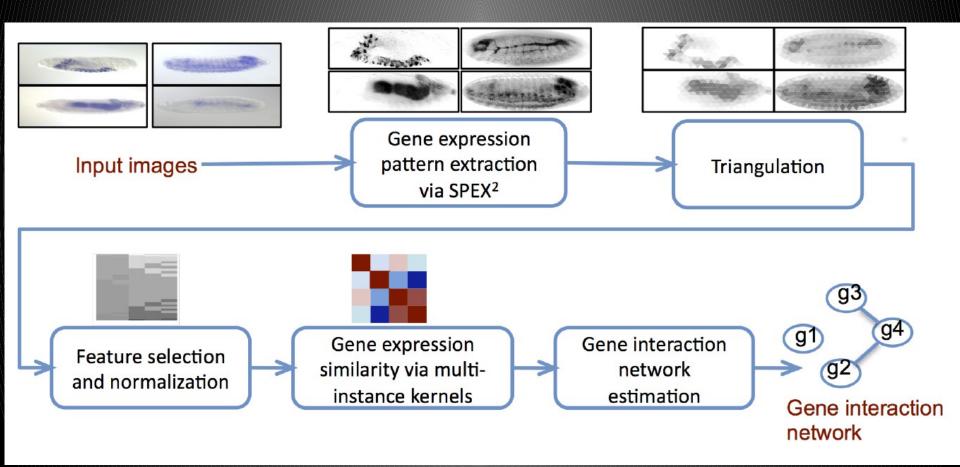








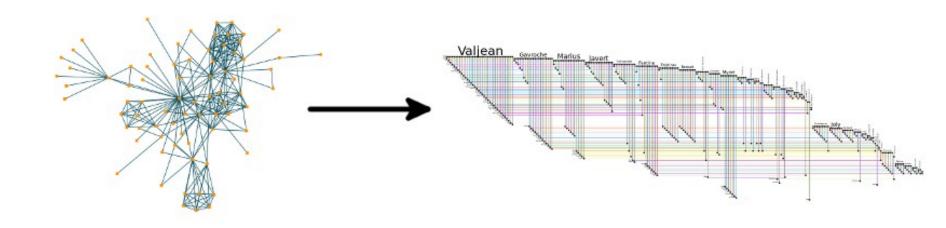
Spatial modeling



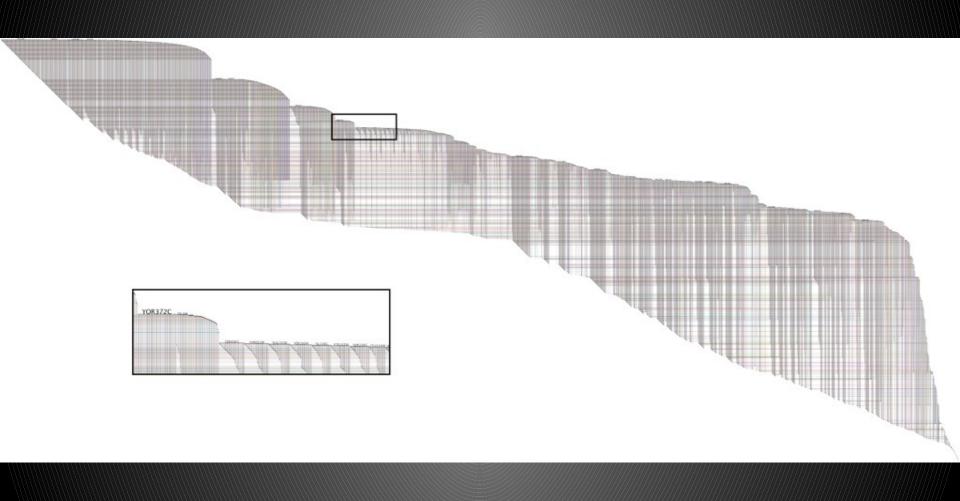
Puniyani, K., and Xing, E.P. (2013). GINI: from ISH images to gene interaction networks. PLoS Comput. Biol. 9, e1003227.

Graph visualization

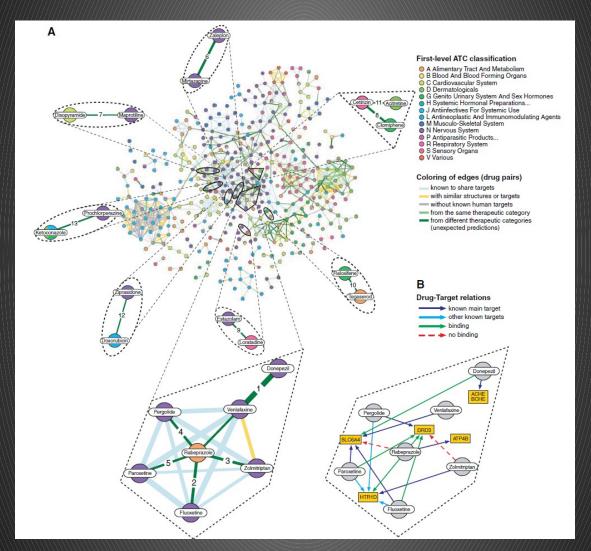
Biofabric (http://www.biofabric.org/)



Biofabric

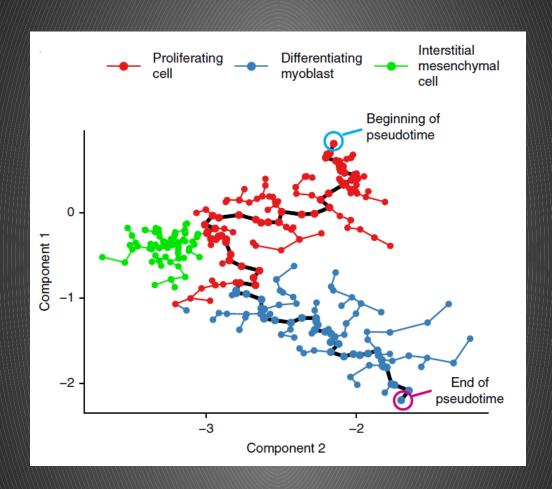


Drug repositioning



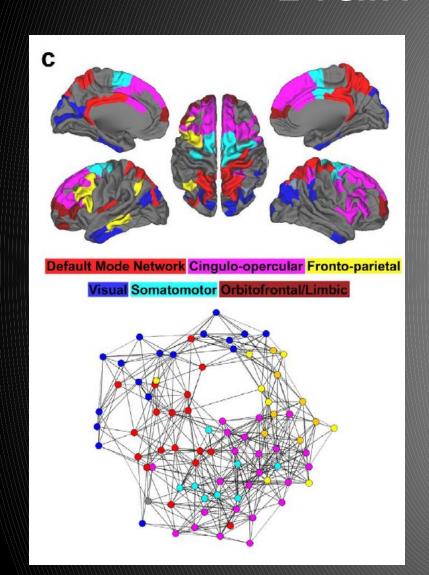
Campillos, M., Kuhn, M., Gavin, A.-C., Jensen, L.J., and Bork, P. (2008). Drug Target Identification Using Side-Effect Similarity. Science 321, 263–266.

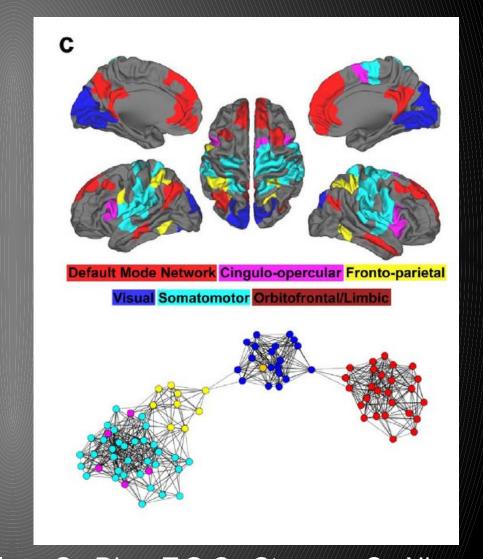
Cell differentiation networks



Trapnell, C., Cacchiarelli, D., Grimsby, J., Pokharel, P., Li, S., Morse, M., Lennon, N.J., Livak, K.J., Mikkelsen, T.S., and Rinn, J.L. (2014). The dynamics and regulators of cell fate decisions are revealed by pseudotemporal ordering of single cells. Nat. Biotechnol. *advance online publication*.

Brain networks





Grayson, D.S., Ray, S., Carpenter, S., Iyer, S., Dias, T.G.C., Stevens, C., Nigg, J.T., and Fair, D.A. (2014). Structural and Functional Rich Club Organization of

Open Questions

- How can we best simulate background models of interaction networks?
- What graph theory measures capture the salient qualities of real graphs?
- How do we model changes of interaction networks over time?
- How do we incorporate spatial aspects?
- Can we use symmetry groups?