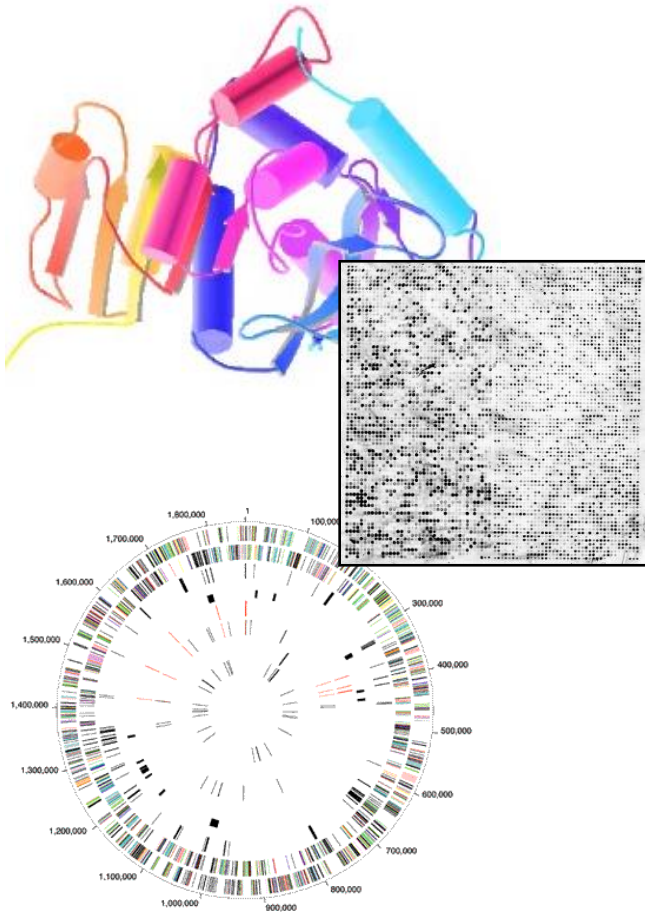


Unsupervised Datamining -- Community Detection



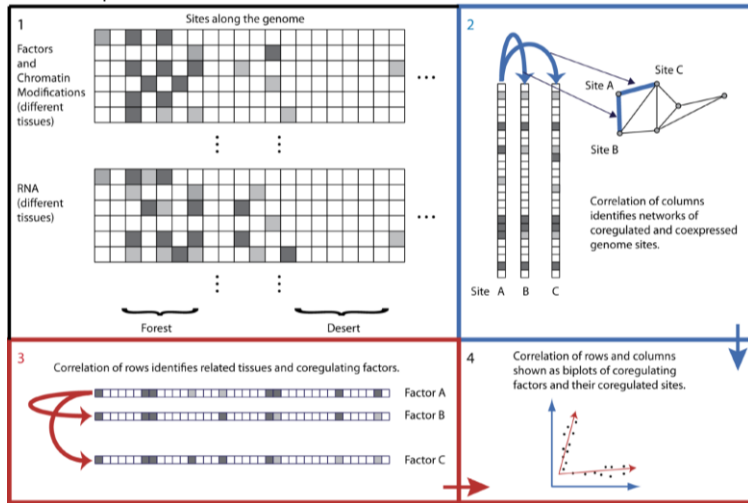
Mark Gerstein, Yale University
gersteinlab.org/courses/452

(last edit in spring '22, pack 22m9b; similar to pack M9b from '21 with additional slide 2 & edits to the TADs section.)

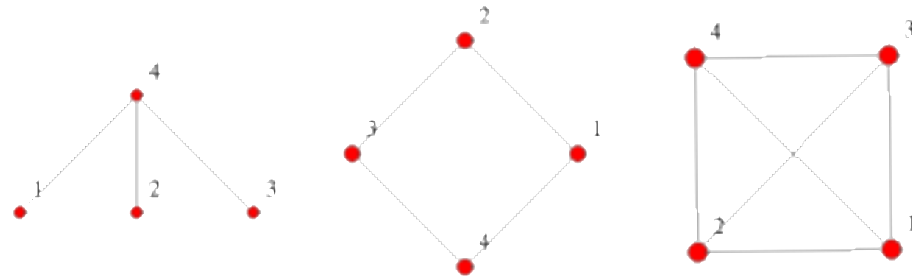
Unsupervised Mining

Graph Analysis &
Community Detection Approaches

Graph Methods & Community Detection



- Turn data into a graph
- Alternate local and global clustering while optimizing for **modularity**
- Can discover the number of clusters given a resolution
 - Cell type detection
- Fast: $O(n \log n)$
- E.g. Louvain, Leiden Community detection



Adjacency Matrix & Graphs
Correlating rows or columns

$$R = AA^T$$

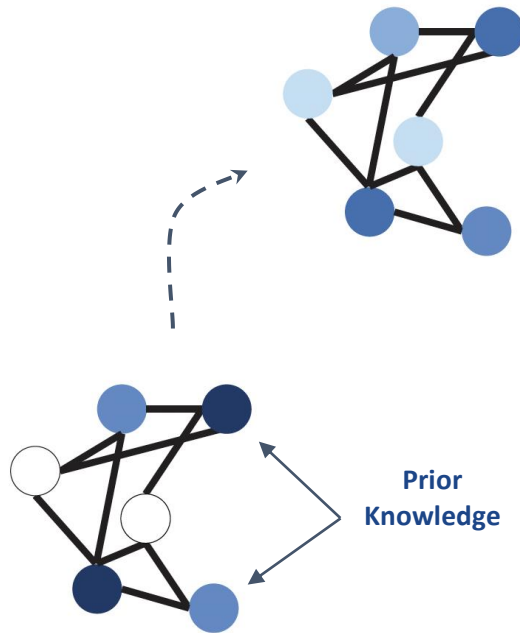
$$C = A^T A$$

$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

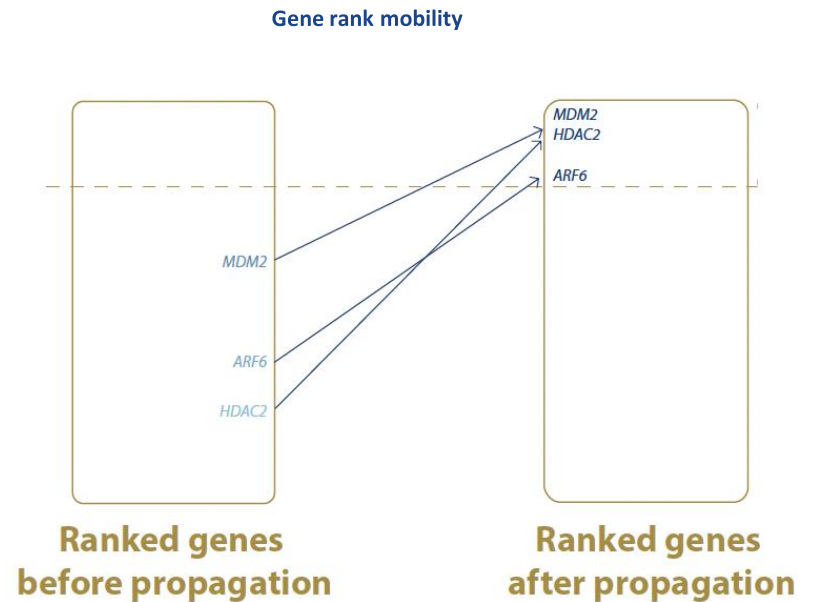
$$\begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

Network propagation-based prioritization



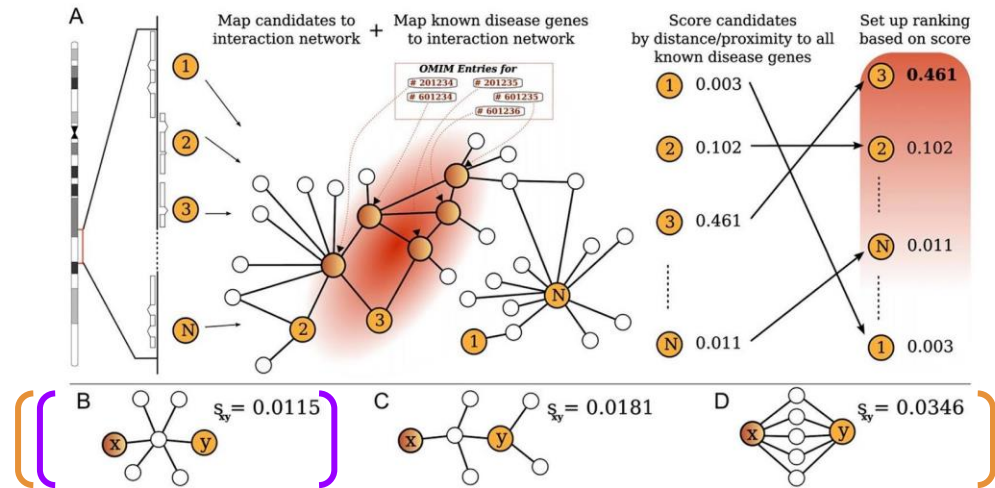
- Changes in gene rank before & after propagation



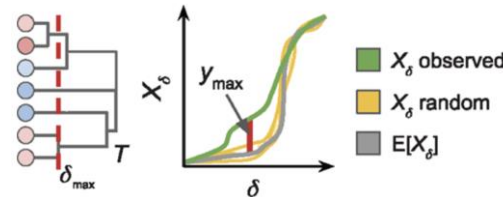
$$S_{(t+1)} = \alpha W' S_{(t)} + (1-\alpha) S_{(0)}$$

Network Propagation in Biomedicine (Label propagation & Diffusion distance)

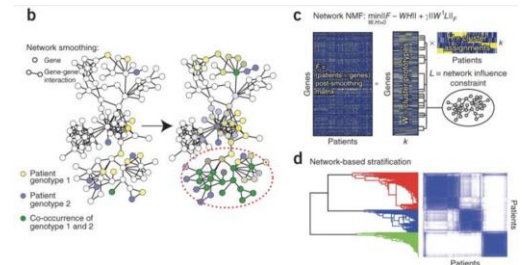
- Starting ~2008
- Limitations in nearest neighbor (B) and shortest distance measures (B-D)
- Leverages local and global network topology
- Mathematically rigorous
- Early methods: function prediction and gene-disease association
- Current methods: gene ranking, **subnetwork detection**, gene-drug and TF-target associations, **patient sample stratification**, etc.



Köhler, Bauer, Horn and Robinson (2008)



Reyna, Leiserson, and Raphael (2018)

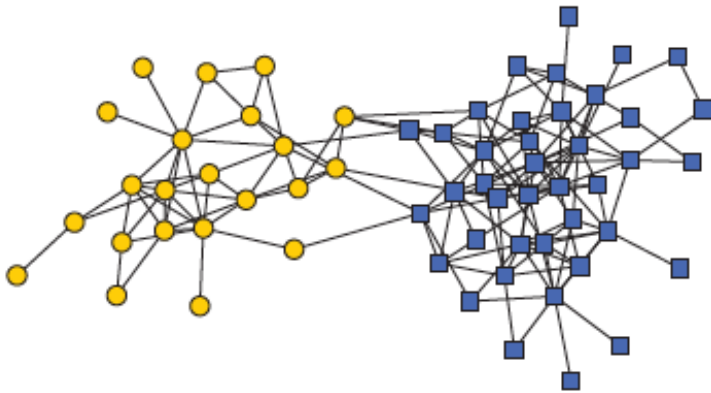


Hofree, Shen, Carter, Gross, and Ideker (2013)

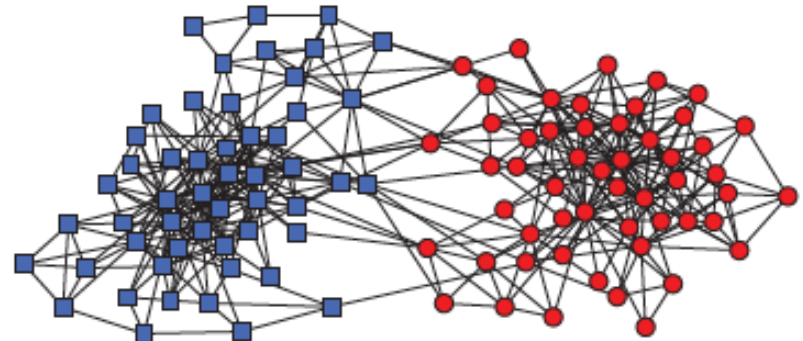
Unsupervised Mining

Community Detection
Application to Hi-C

Network modularity



Dolphin social network



Political books

Newman Phys. Rev. E 2013

adjacency matrix

$$Q = \frac{1}{2m} \sum_{i,j} \left(W_{ij} - \frac{k_i k_j}{2m} \right) \delta_{\sigma_i \sigma_j}$$

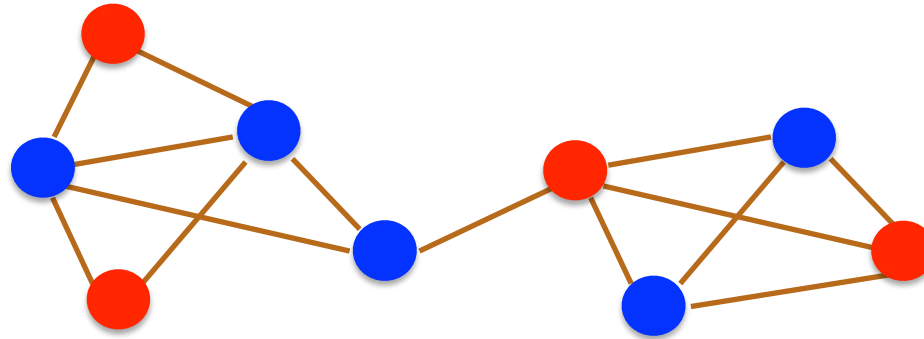
number of edges

degree of node i

whether or not i, j are in the same module

expected number of edges between i and j

Network modularity

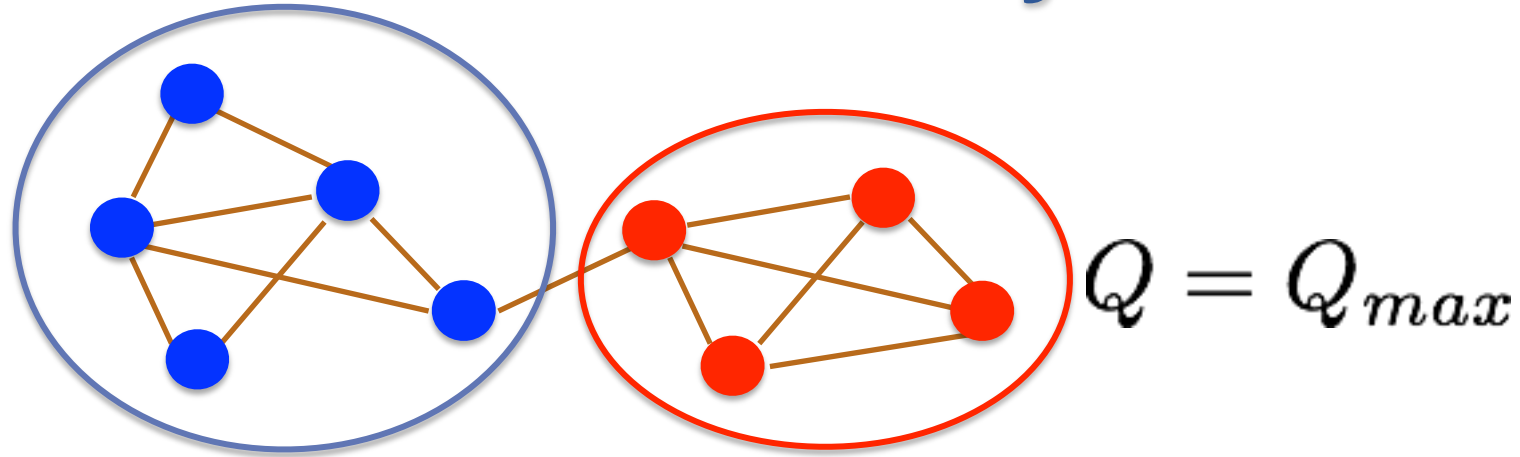


$$Q \approx 0$$

$$Q = \frac{1}{2m} \sum_{i,j} \left(W_{ij} - \frac{k_i k_j}{2m} \right) \delta_{\sigma_i \sigma_j}$$

adjacency matrix W_{ij}
 degree of node i k_i
 number of edges $2m$
 expected number of edges between i and j $\frac{k_i k_j}{2m}$
 whether or not i, j are in the same module $\delta_{\sigma_i \sigma_j}$

Network modularity

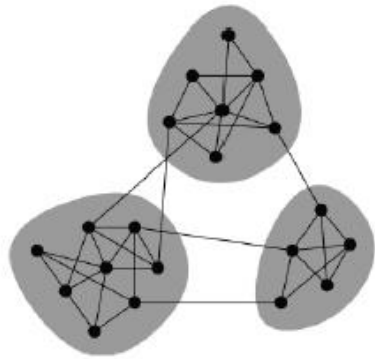


Optimization
problem
for sim.
annealing

$$Q = \frac{1}{2m} \sum_{i,j} \left(W_{ij} - \frac{k_i k_j}{2m} \right) \delta_{\sigma_i \sigma_j}$$

adjacency matrix W_{ij}
 degree of node i k_i
 whether or not i, j are in the same module $\delta_{\sigma_i \sigma_j}$
 number of edges $2m$
 expected number of edges between i and j $\frac{k_i k_j}{2m}$

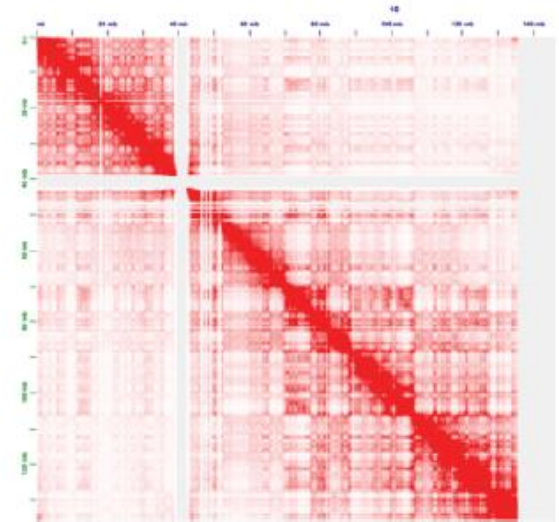
Identifying TADs in multiple resolutions



Modularity maximization

$$Q = \frac{1}{2m} \sum_{i,j} \left(W_{ij} - \frac{k_i k_j}{2m} \right) \delta_{\sigma_i \sigma_j}$$

network	contact map
node	chromosome bin
edge	Hi-C contact
# of connections	coverage
module	domain



schematic adapted from ref. [2]

.....Resuming from “Multi-omics pack”

Identifying TADs in multiple resolutions

