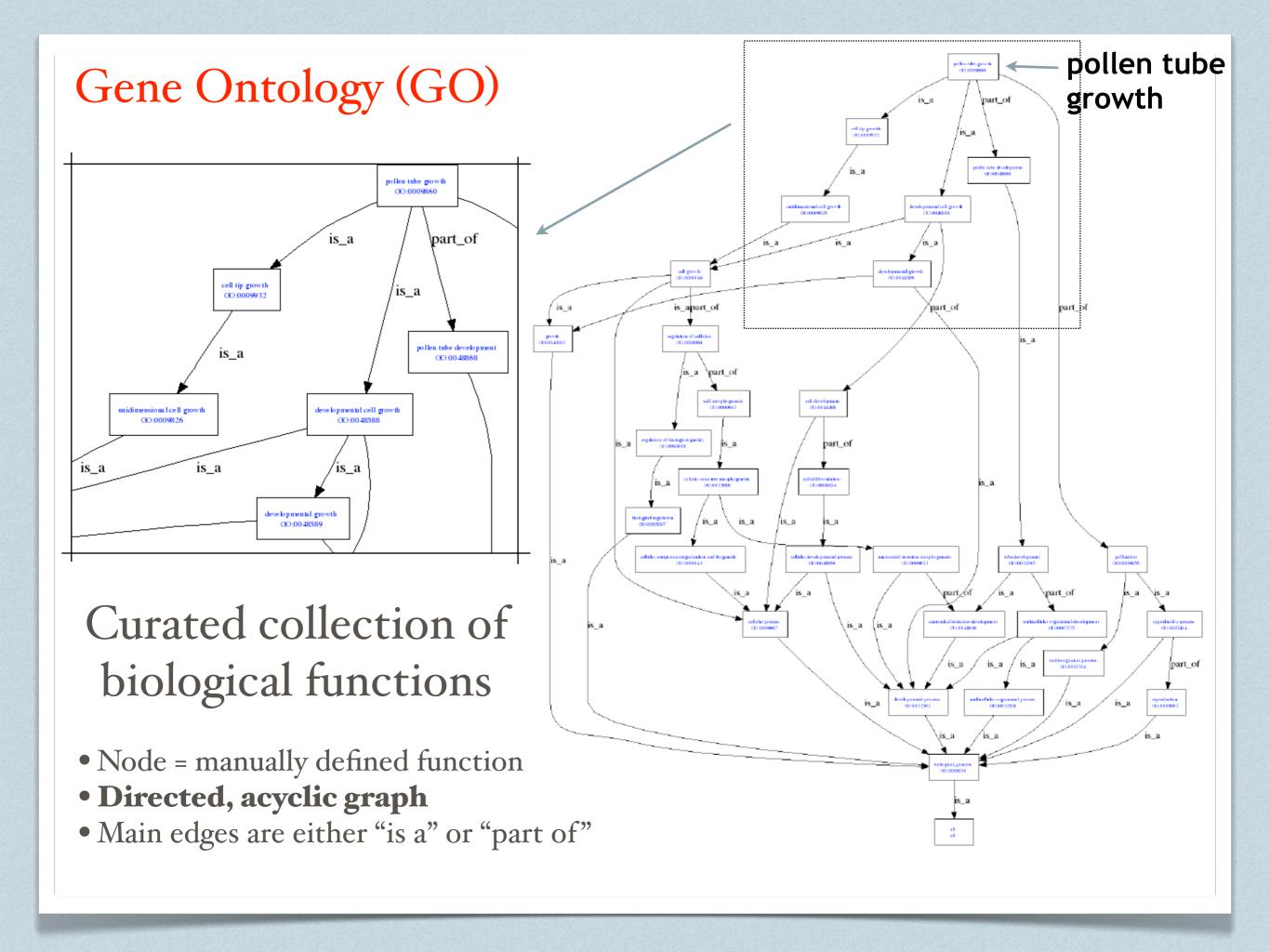
Function Prediction

CMSC 858L

Predicting Protein Function from Networks

- Ultimately, we want to know how various processes in the cell work.
- A first step: figure out which proteins are involved in which biological role.

- What do we mean by a "biological role"?
 - Several different schemes:
 - Gene Ontology (largest, most widely used)
 - MIPS (good collection of known protein complexes)
 - KEGG (manually curated pathways)

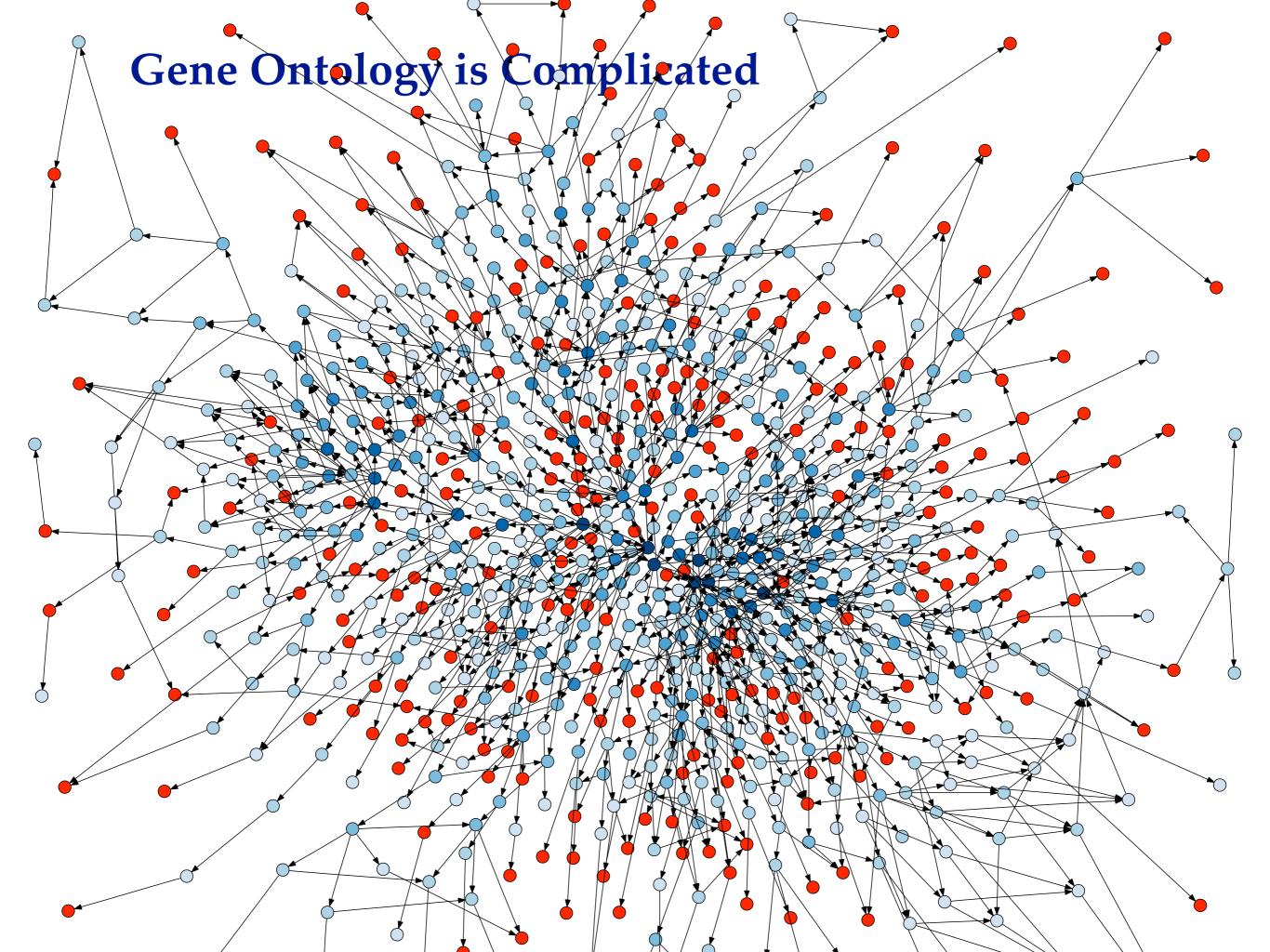


Gene Ontology has 3 Sub-ontologies

- <u>Cellular component</u>: a part of the cell (a location, or organelle, or other structure)
- <u>Biological process</u>: a collection of steps that the cell carries out to achieve some purpose. E.g. cell division.
- Molecular function: a specific mechanism that a protein performs. E.g.
 - a kinase would have molecular function "phosphorylation";
 - a transcription factor would have molecular function "DNA binding"
- Each protein may be *annotated* with several terms from each sub-ontology.

Edge Types

- **is_a**: like a C++ or Java subclass relationship.
 - A is_a B means A is a more specific version of B
 - **–** E.g. "nuclear chromosome" **is_a** "chromosome".
- part_of: A is some part of B
 - A piston is part_of an engine (but a piston is not an specific kind of engine)
- Transitivity:
 - If a protein is annotated with term A, it is implicitly annotated with all the ancestors of A (following every path to the root).
 - GO is explicitly designed so this is always true.



KEGG is a tree of "pathways"

1. Metabolism

1.1 Carbohydrate Metabolism

Glycolysis / Gluconeogenesis

Citrate cycle (TCA cycle)

Pentose phosphate pathway

Pentose and glucuronate interconversions

Fructose and mannose metabolism

Galactose metabolism

Ascorbate and aldarate metabolism

Starch and sucrose metabolism

Aminosugars metabolism

Nucleotide sugars metabolism

Pyruvate metabolism

Glyoxylate and dicarboxylate metabolism

Propanoate metabolism

Butanoate metabolism

C5-Branched dibasic acid metabolism

Inositol metabolism

Inositol phosphate metabolism

1.2 Energy Metabolism

Oxidative phosphorylation

Photosynthesis

Photosynthesis - antenna proteins

Carbon fixation in photosynthetic organisms

Reductive carboxylate cycle (CO2 fixation)

Methane metabolism

Nitrogen metabolism

Sulfur metabolism

1.3 Lipid Metabolism

Fatty acid biosynthesis

Fatty acid elongation in mitochondria

Fatty acid metabolism

Synthesis and degradation of ketone bodies

Biosynthesis of steroids

Bile acid biosynthesis

C21-Steroid hormone metabolism

Androgen and estrogen metabolism

Glycerolipid metabolism

Glycerophospholipid metabolism

Ether lipid metabolism

Sphingolipid metabolism

Arachidonic acid metabolism

Linoleic acid metabolism

alpha-Linolenic acid metabolism

Biosynthesis of unsaturated fatty acids

1.4 Nucleotide Metabolism

Purine metabolism Pyrimidine metabolism

1.5 Amino Acid Metabolism

Clutamate metabolism

1.5 Amino Acid Metabolism

Glutamate metabolism

Alanine and aspartate metabolism

Glycine, serine and threonine metabolism

Methionine metabolism

Cysteine metabolism

Valine, leucine and isoleucine degradation

Valine, leucine and isoleucine biosynthesis

Lysine biosynthesis

Lysine degradation

Arginine and proline metabolism

Histidine metabolism

Tyrosine metabolism

Phenylalanine metabolism

Tryptophan metabolism

Phenylalanine, tyrosine and tryptophan biosynthesis

Urea cycle and metabolism of amino groups

1.6 Metabolism of Other Amino Acids

beta-Alanine metabolism

Taurine and hypotaurine metabolism

Aminophosphonate metabolism

Selenoamino acid metabolism

Cyanoamino acid metabolism

D-Glutamine and D-glutamate metabolism

D-Arginine and D-ornithine metabolism

D-Alanine metabolism

Glutathione metabolism

1.7 Glycan Biosynthesis and Metabolism

N-Glycan biosynthesis

High-mannose type N-glycan biosynthesis

N-Glycan degradation

O-Glycan biosynthesis

Chondroitin sulfate biosynthesis

Heparan sulfate biosynthesis

Keratan sulfate biosynthesis

Glycosaminoglycan degradation

Lipopolysaccharide biosynthesis

Peptidoglycan biosynthesis

Glycosylphosphatidylinositol(GPI)-anchor biosynthesis

Glycosphingolipid biosynthesis - lactoseries

Glycosphingolipid biosynthesis - neo-lactoseries

Glycosphingolipid biosynthesis - globoseries

Glycosphingolipid biosynthesis - ganglioseries

Glycan structures - biosynthesis 1

Glycan structures - biosynthesis 2

Glycan structures - degradation

1.8 Biosynthesis of Polyketides and Nonribosomal Peptides

Type I polyketide structures

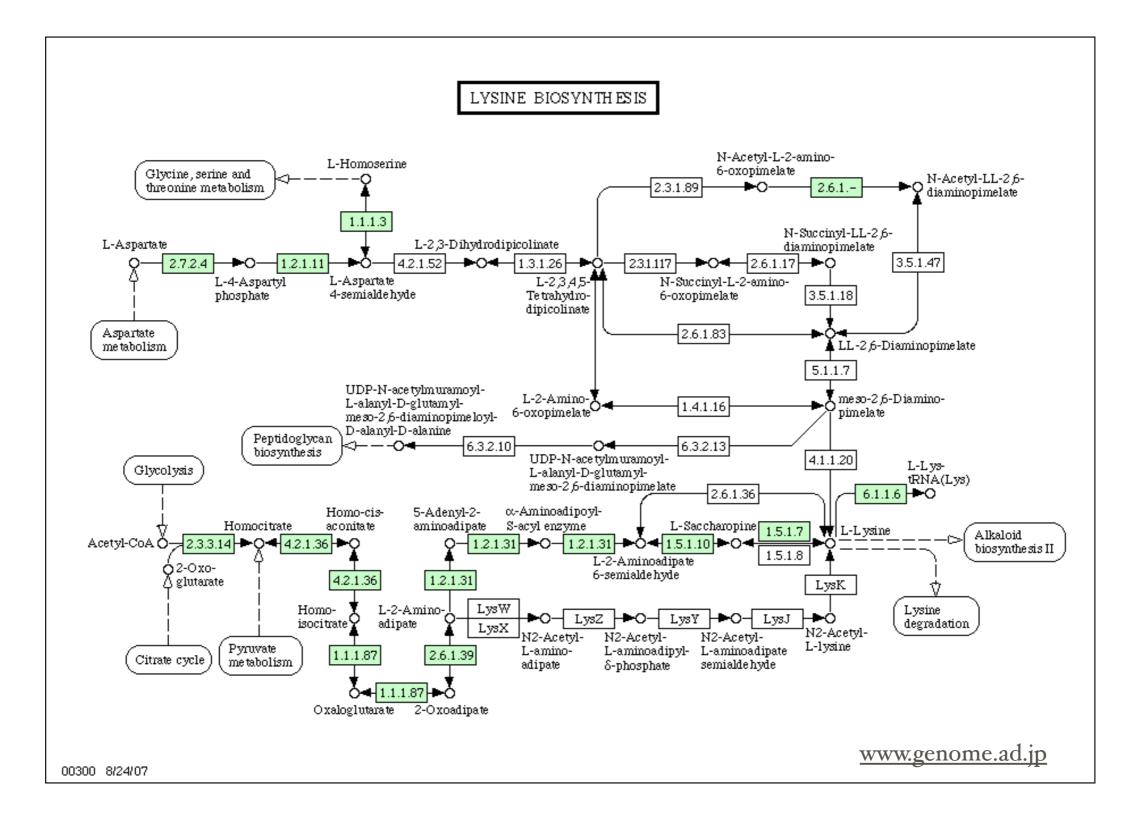
Biosynthesis of 12-, 14- and 16-membered macrolides

Biosynthesis of ansamycins

Biosynthesis of type II polyketide backbone

Biosynthesis of type II polyketide products

KEGG PATHWAY



MIPS has annotation terms organized in trees

- Function Catalog (FunCat):

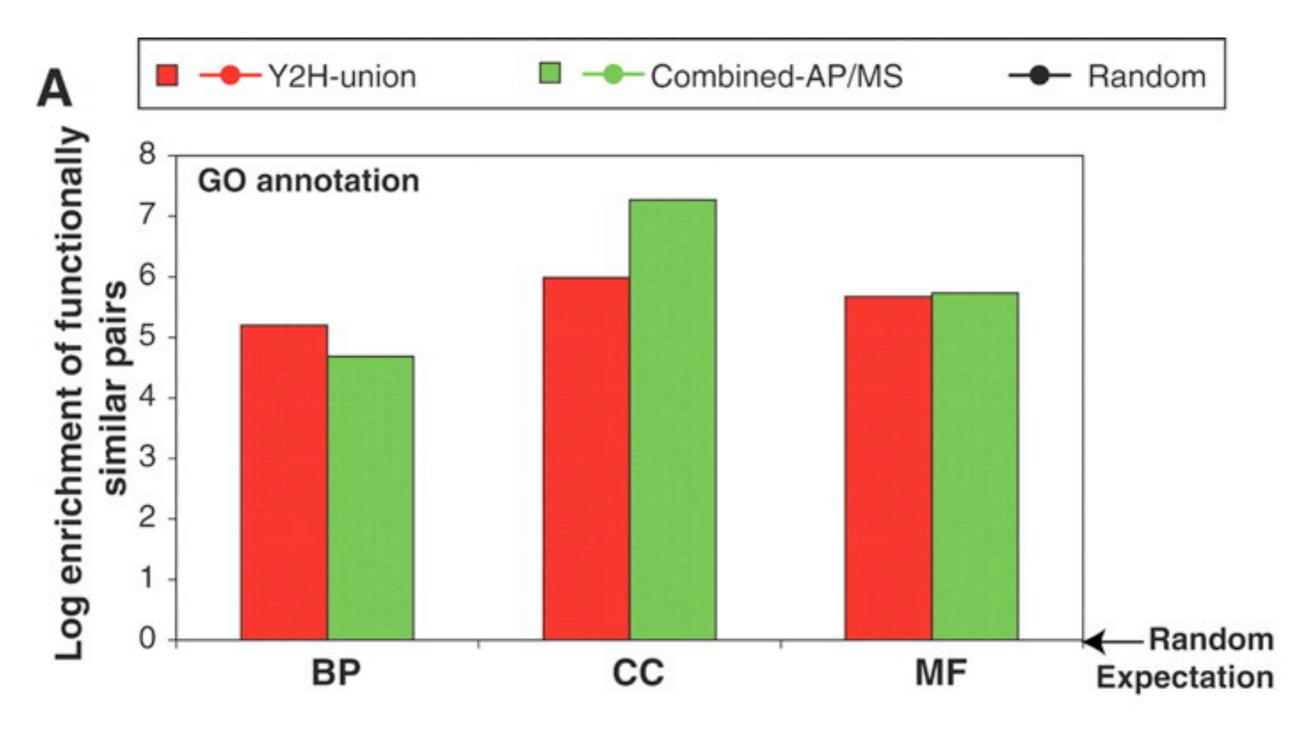
 a collection of functions and
 biological processes
 organized as a tree.
- Manually annotated protein complexes (e.g. at left)
 - also organized as a tree

Complex	Proteins
20 2-oxoglutarate dehydrogenase	3
40 Alpha-agglutinin anchor	2
60 Anaphase promoting complex (APC)	11
70 Anthranilate synthase	2
75 Arginase	1
80 Arginine-specific carbamoylphosphate synthase	2
90 Assembly complexes	7
100 Calcineurin B	3
110 cAMP-dependent protein kinase	4
120 Casein kinase	8
123 Catalase	2
125 Cell cycle checkpoint complexes	2
130 Chaperonine containing T-complex TRiC (TCP RING Complex)	8
132 CTP synthetase	1
133 Cyclin-CDK (Cyclin-dependent kinases) complexes	25
140 Cytoskeleton	73
143 D-arabinose dehydrogenase	1
145 delta3-cis-delta2-trans-enoyl-CoA isomerase	1
150 Endonuclease Scel, mitochondrial	1
160 Exocyst complex	7
170 Fatty acid synthetase, cytoplasmic	2
172 Fatty acid synthetase, mitochondrial	1
177 Gim complexes	5
180 Prenyltransferases	6
190 Glucan synthases	5
200 Glycine decarboxylase	4
210 H+-ATPase, plasma mebrane	4
220 H+-transporting ATPase, vacuolar	15
225 Hexokinase 2	1
230 Histone acetyltransferase complexes	19
240 Histone deacetylase complexes	5

Basic Methods for Predicting Function

- Majority Rule
- Neighborhood enrichment
- Minimum Multiway Cut
- "Functional Flow"

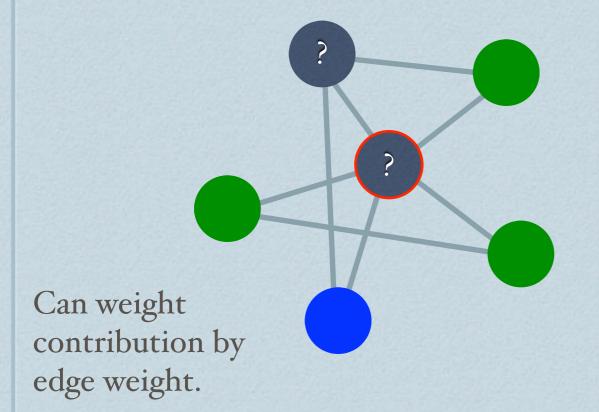
Neighboring Proteins More Likely to Share Function



• (Yu et al., 2008)

Majority Rule

- ❖ Proteins with known function + network topology → function assignment for unknown proteins.
- Guilt by association
- * Majority Rule:





Doesn't take into account connections between neighbors
Or annotations at distance > one

Neighborhood Approaches, e.g.:

Let N(u,r) be all the proteins within distance r to u.

$$f(u, r, a) = |\{u \in N(u, r) : u \text{ has function } a\}|$$
= # of proteins in neighborhood with function a

$$e(u,r,a) = |N(u,r)| \cdot \frac{|\{u \in V : u \text{ has function } a\}|}{|V|}$$

= Expected # of proteins in neighborhood with function *a*

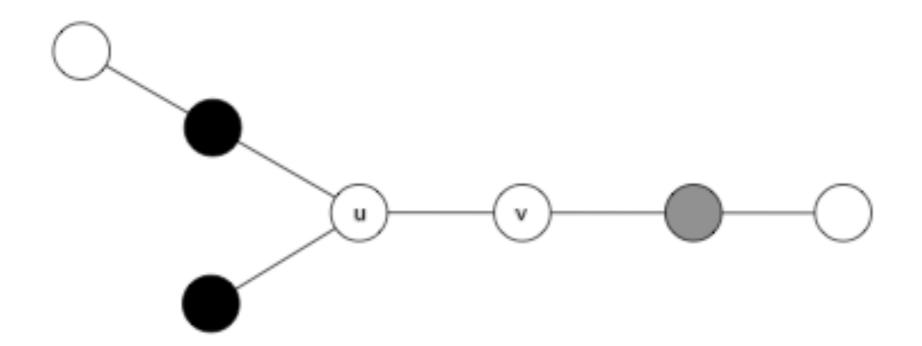
Score
$$(u, r, a) = \frac{(f(u, r, a) - e(u, r, a))^2}{e(u, r, a)}$$

 $\approx \chi_2$ statistic measures how surprising it is to see the observed # of proteins annotated with *a* in the neighborhood

• Protein u is assigned function $\underset{\text{argmax}_a}{\text{argmax}_a}$ Score(u,r,a)

Problems with neighborhood

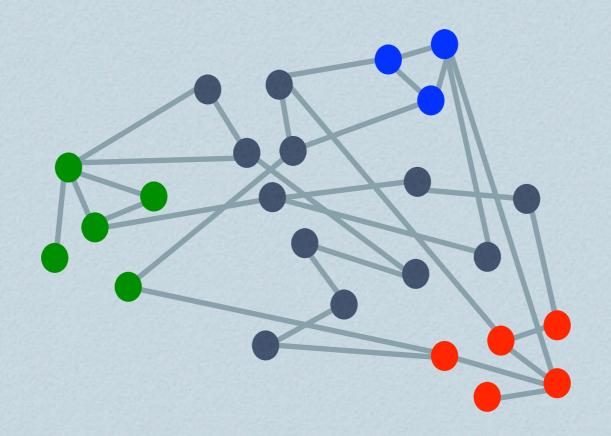
 Neighborhood with radius 2 gives the same scores for black and gray functions to nodes u and v:



(Nabieva, Singh, 2008)

Minimum Multiway *k***-Cut:** Partition the nodes so that each of *k* (sets of) terminal nodes is in a different partition & the number of edges cut is minimized.

- Proposed by Vazquez et al (2003) and Karaoz (2004) for function annotation.
- * One "terminal node set" for each function, containing proteins known to have that function.
- NP-hard: simulated annealing; integer programming



Integer Programming

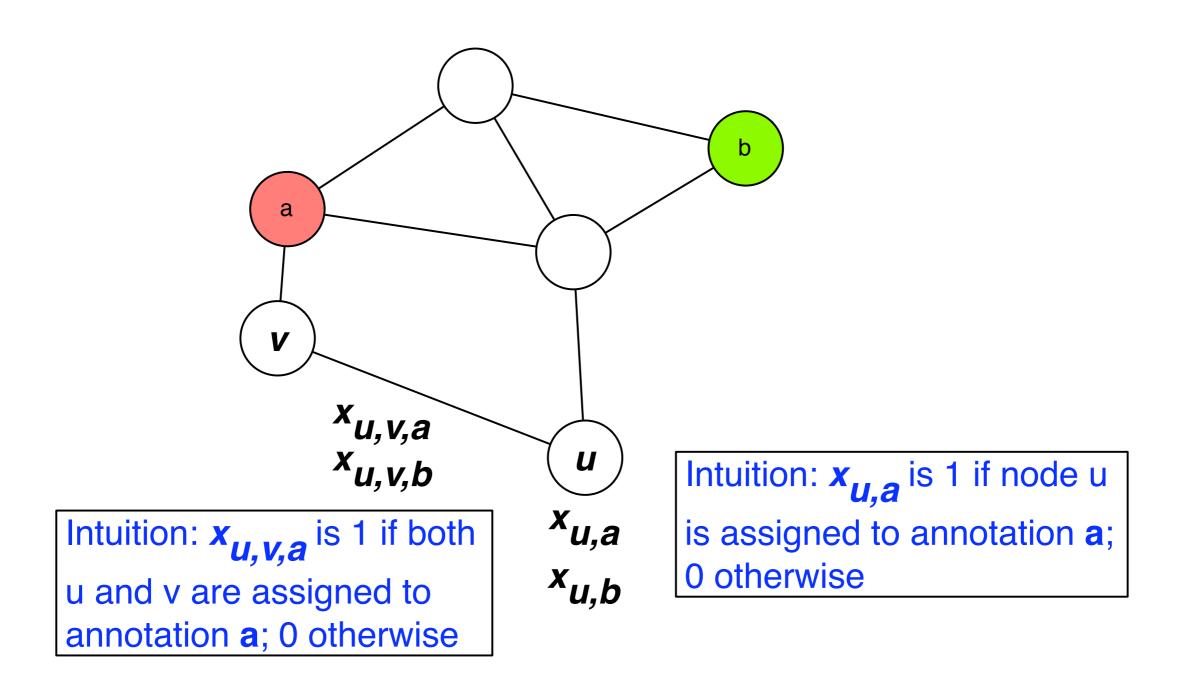
- General optimization framework:
 - Describe system by set of variables

```
IP:= - Minimize a linear function.
- Subject to linear constraints (= or ≥).
- While requiring the variables to be {0,1}.
```

- Computationally hard, but many advanced solver packages:
 - CPLEX, COIN-OR, ABACUS, FortMP, LINGO, ...

Integer Programming (IP) Formulation for Multiway Cut

Introduce 0/1 variables associated with each node and edge:



IP for Min Multiway Cut

maximize
$$\sum_{\{u,v\}\in E,a} x_{u,v,a}$$
 "monochromatic edges" Equivalent to minimizin

Maximize # of Equivalent to minimizing the number of cut edges.

Subject to:

$$x_{u,x} \text{ and } x_{u,v,a} \in \{0,1\}$$

$$\sum_{a} x_{u,a} = 1$$
 Each node gets exactly 1 annotation

$$x_{u,v,a} \le x_{u,a}$$
 Can set $x_{u,v,a}$ to 1 iff both its $x_{u,v,a} \le x_{v,a}$ endpoints are 1

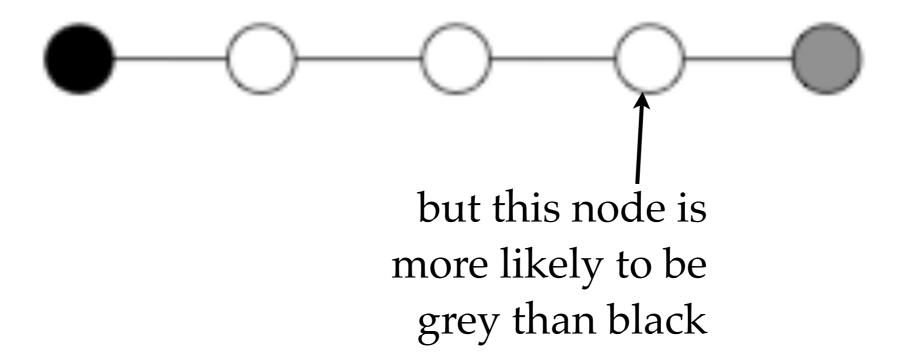
$$x_{u,a} = 1 \text{ if } a \in \text{annot}(u)$$

$$x_{u,a} = 0 \text{ if } a \notin \text{annot}(u) \neq \emptyset$$

Fix variables for nodes with known annotations.

Problem with Simple Cut Approaches

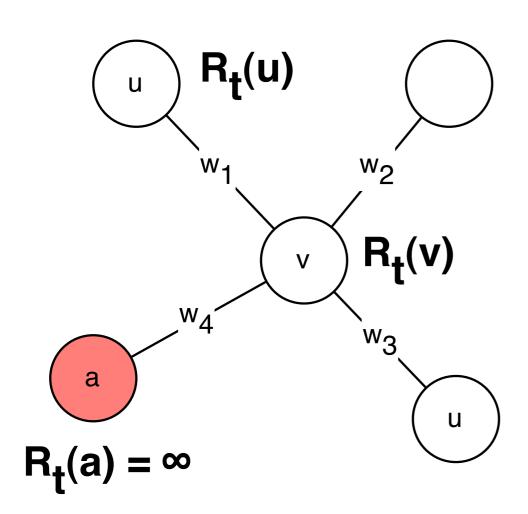
• Every cut is equally likely:



(Nabieva, Singh, 2008)

Functional Flow (Nabieva et al.)

Each node *u* has a "reservoir" at each time step t.



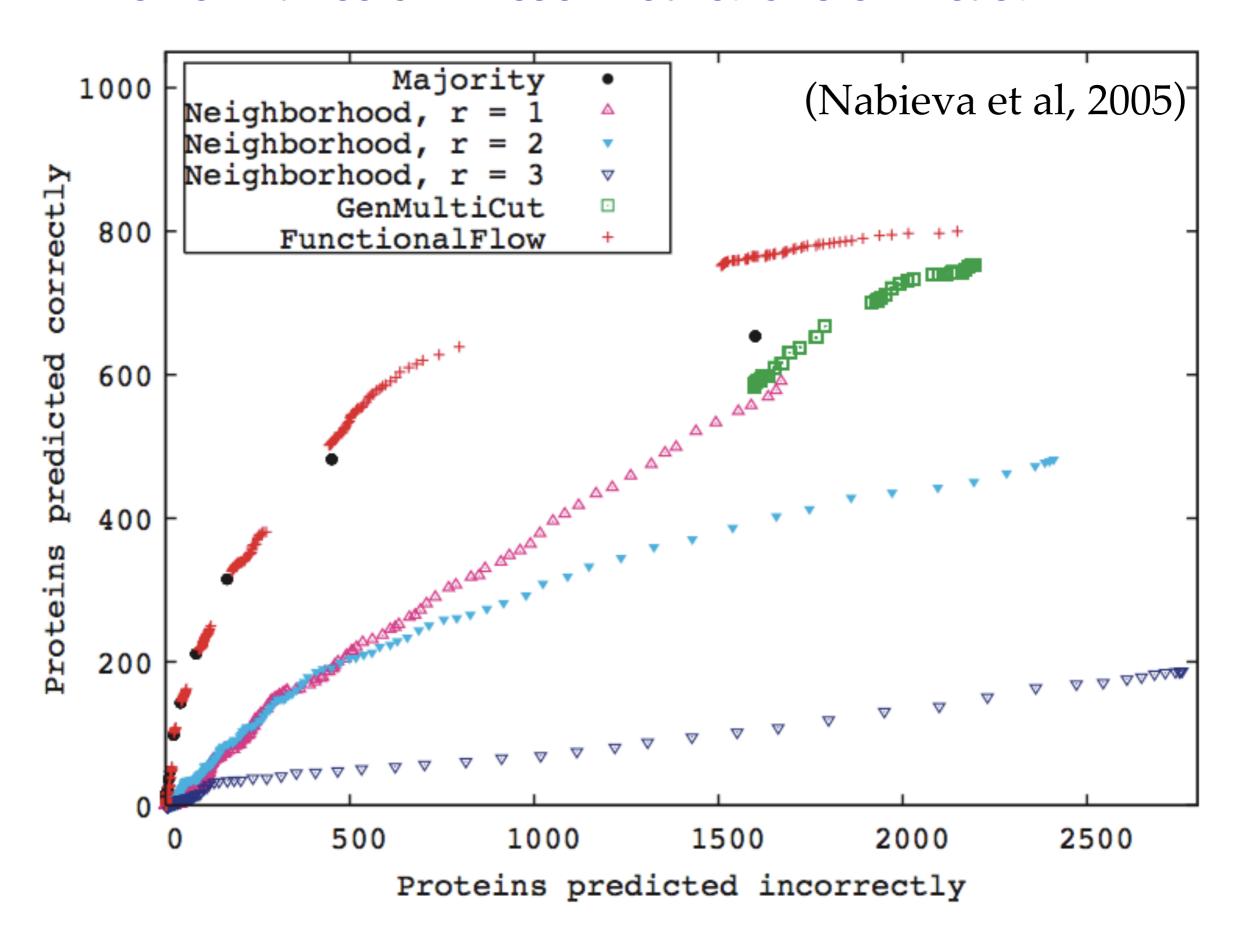
At every time step, water flows "downhill" from the more filled reservoir to the more empty reservoir, up to the capacity of the edge.

If there isn't enough water to fill the downhill pipes, it is distributed proportionally to the capacity of the edge.

Every function f is considered separately.

Score(u,f) is the total water that passed through u when considering f. Predicted function for u is the function with the highest score.

Performance of These Predictions on Yeast



Summary

- Guilt-by-association = proteins near one another in the network are more likely to have the same function.
- Neighborhood 1 does better than larger neighborhoods
 Perhaps because the structure of the neighborhood is not taken into account.
- Integer programming NP-hard, but often practical.
 Can obtain multiple solutions in 2 ways:
 - Random perturbation of weights
 - Solving successive problems with additional constraints.
- "Functional flow" is an embodiment of a general technique: "information" being passed along the network.