

# **Protein Function Prediction In Networks**

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- Determining protein function is one of the most important

problems in proteomics.

-High-throughput data collection is messy, and often has missing data.

-Therefore, many proteomes are missing functional

annotations of their proteins.

- -We present a pattern-matching based algorithm to predict
- values for these missing data

#### **Our Method**

- First we find all interaction pairs in which the query

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protein participates

-Then we look in the neighborhood of the query node to

find other interaction pairs

-We use Bayesian inference analysis on the interaction

pairs collected in the neighborhood to predict values for

the missing function

-We store the prior and conditional probabilities of the

# **Current Methods**

- General idea is to look at attribute values in the

neighborhood of a query protein and use most frequently occurring values to determine attribute values for the query protein.

- Assumes highly assortative networks



observations seen in the neighborhood interaction pairs -Before we store counts of the values seen, we flatten the interaction pairs



# **Experimental Method**

- We use 2-fold cross validation
- Implementation in Java

# **Example Sub-Network**

-Shows a protein-protein interaction sub-network from the earthworm proteome

-Let us assume that the protein labeled 3 is missing its

annotation.

-If we only take into account annotation frequencies, we

mistakenly guess protein 3 is a square

-The weakness of these methods is they don't adapt to

structural features of the graph

# Hypothesis

-Our hypothesis is that looking at the interactions of the query protein will give us a better estimate of its functions
than simply by looking at its neighborhood.
-We explore interaction pair similarities to interpolate missing function.

- We compare with majority (Schwikowski et al.) and

function flow methods (Nabieva et al.)

- Yeast Protein-Protein Interaction. The attribute we interpolate is function (48 possibilities).



0 1 2 3 4 5 6 7 8 9 10 11 Radius

-We treat the graph as the raw data for deducing

similarities and supports for interpolation by performing a

Bayesian inference analysis on interaction pairs



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