

## Metabolic flux prediction via gene expression and metabolomics

#### **Tomer Shlomi**

Department of Computer Science, Technion, Haifa, Israel

October, 2009



## Metabolic stuff we've been doing recently

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- Context-dependent Biomass Flux Balance Analysis (CB-FBA)
  - Predicting flux distributions by accounting for growth-associated demand for biomass production in a context-dependent manner
- RobustKnock: Predicting Metabolic Engineering Knockout Strategies for Chemical Production
  - Improving OptKnock by accounting for alternative pathways
- Predicting Metabolic Gene-Nutrient Interactions (GNIs) in yeast
  - Predicting constraints on nutrient availability in the growth media based on enzyme essentiality data
- Predicting Enzyme Sub-cellular Localization
  - Predicting enzymes' sub-cellular localization based on partial localization data for a subset of the enzymes in the network



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Metabolic Flux Balance Analysis with Context-dependant Biomass.

T. Benyamini, O. Folger, E. Ruppin, T. Shlomi, RECOMB, Systems Biology, 2009 (to appear)

**NOT PUBLISHED YET** 

□ Predicting enzymes' sub-cellular localization based on partial localization data for a subset of the enzymes in the network



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Predicting Metabolic Engineering Knockout Strategies for Chemical Production: Accounting for Competing Pathways.

N. Tepper, T. Shlomi (Submitted)

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Context-dependent Biomass Flux Balance Analysis (CB-FBA)

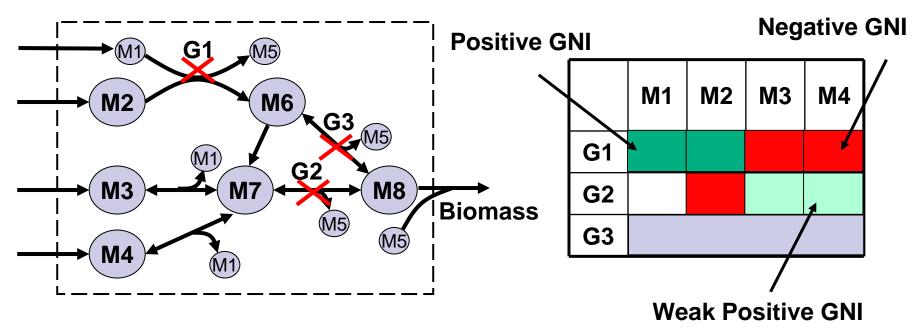
Metabolic Network-based Analysis of Yeast Gene-Nutrient Interactions.

- I. Diamant, Y. Eldar, O. Rokhlenko, E. Ruppin, T. Shlomi. Molecular BioSystems, DOI: 10.1039, 2009
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### **Gene-Nutrient Interactions**

- Under which growth media G1 is essential?
- A GNI represents a constraint on the presence/absence of a nutrient in the growth media under which a gene is essential
- A weak (vs. strong) GNI reflects a non-strict constraint





## Predicting Gene-Nutrient Interactions

- Identified via a bi-level optimization problem
- Transformed into Mixed-Integer Linear Programming (MILP)

```
gene essentiality (wild-type vs. knockout growth-rate)
  (over all growth media)
subject to
              Maximize
                              wild-type growth rate
                 (over fluxes)
               subject to

    uptake of media substrates

    network stoichiometry, thermodynamics and capacity constraints

              Maximize
                              knockout growth rate
                 (over fluxes)
               subject to

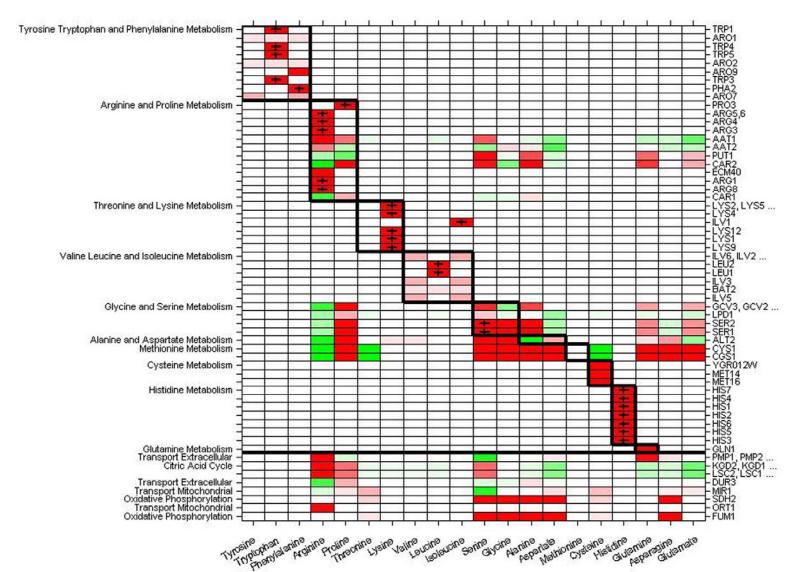
    uptake of media substrates

    network stoichiometry, thermodynamics and capacity constraints

    Inactive knocked-out reactions
```

Maximize

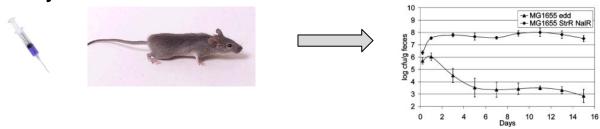
## Gene-Nutrient Interactions in Yeast





# GNI-based 'Reverse Prediction' of Growth Media Composition

- What is the natural growth environment of a pathogen within a host organism?
- Suppose we have in-vivo data on bacterial gene knockout essentiality



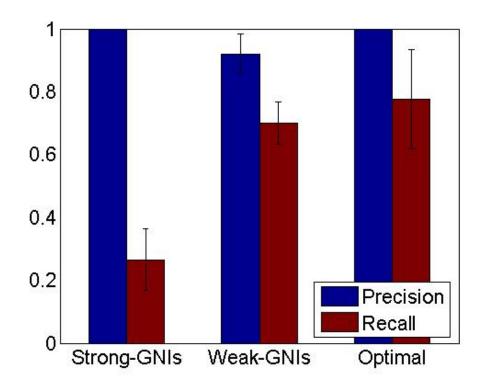
- Can we use the measured pattern of gene essentiality to predict constraints on the in-vivo growth environment of the bacteria?
- Unfortunately, we don't have enough data of this kind. However...





# GNI-based 'Reverse Prediction' of Growth Media Composition

 In simulations, GNI-based analysis provide accurate predictions of growth media composition based on gene essentiality data





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Network-based Prediction of Metabolic Enzymes' Subcellular Localization.

S. Mintz, A. Aharoni, E. Ruppin, T. Shlomi.

Bioinformatics, 25(12): 247-252, 2009 (ISMB'09)

pased on enzyme essentiality data

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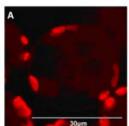
### Derecting Protein Subcellular Localization

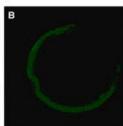
#### **Experimental Methods:**

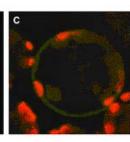
- Green fluorescent protein (GFP) tagging
- Electron microscopy
- Subcellular fractionation + detection

#### Limitations:

- Costly
- Time-consuming







Wormit et.al, Plant Cell, 2006

#### **Computational Methods:**

- Sequence motifs
- Amino acid composition
- Homology
- PPI data

#### **Limitations:**

- Low number of compartments
- Performance varies across different organisms and compartments
- Relatively low availability of PPI networks

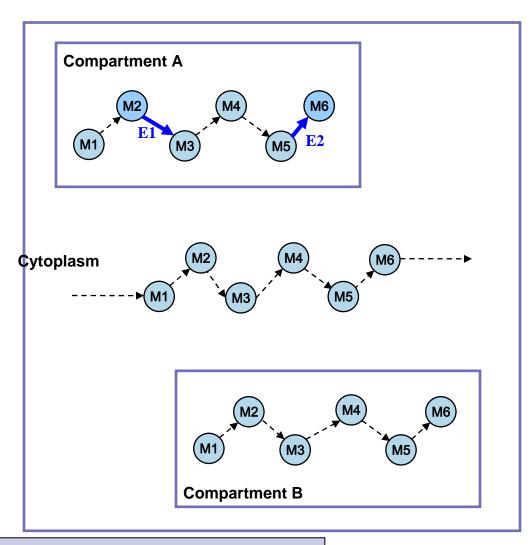


## Research Objective

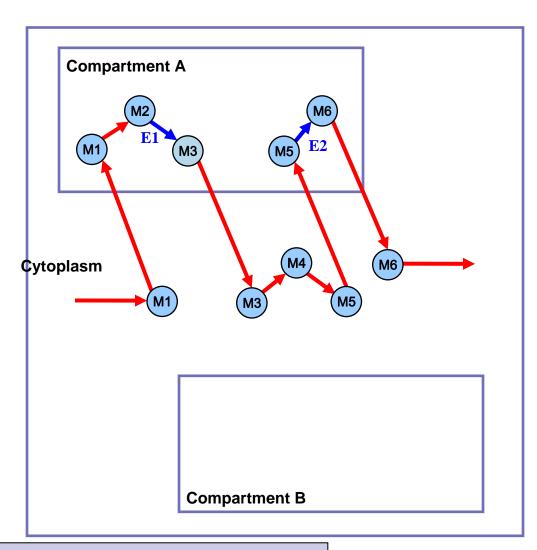
Predict metabolic enzymes' subcellular localization, based on:

- The organism's metabolic network
- Prior knowledge regarding localization of a subset of the enzymes
- Parsimonious assumption of minimal number of crossmembrane metabolite transports between compartments

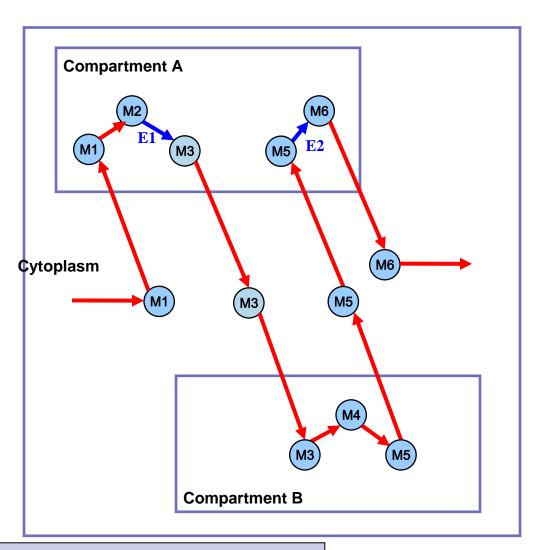










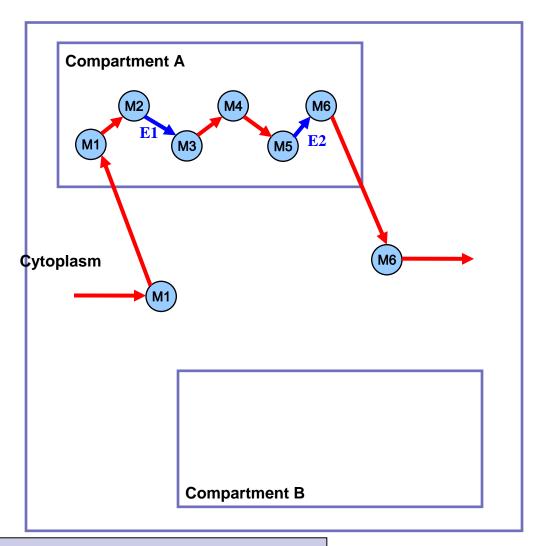


 Transport reactions depend on transporter proteins, imposing energetic cost or requiring the maintenance of a membrane potential



Minimize transport reactions





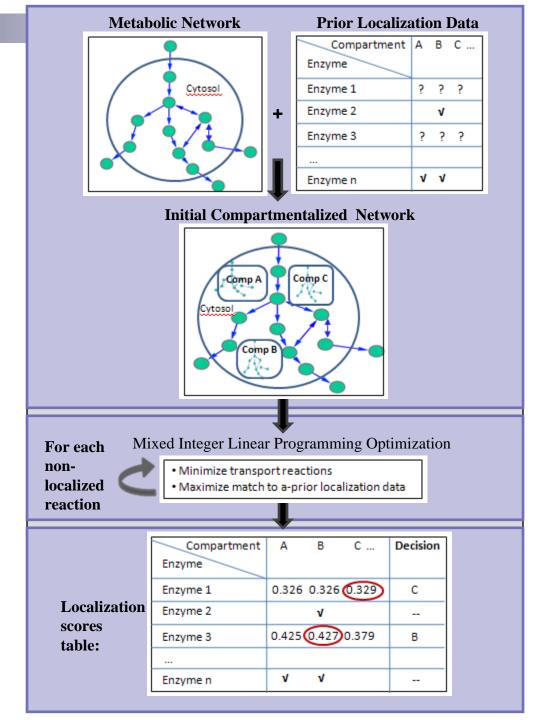
- Match known localization data
- Assume minimal number of metabolite crossmembrane transports

CBM method for predicting localization:

Input:

Optimization process:

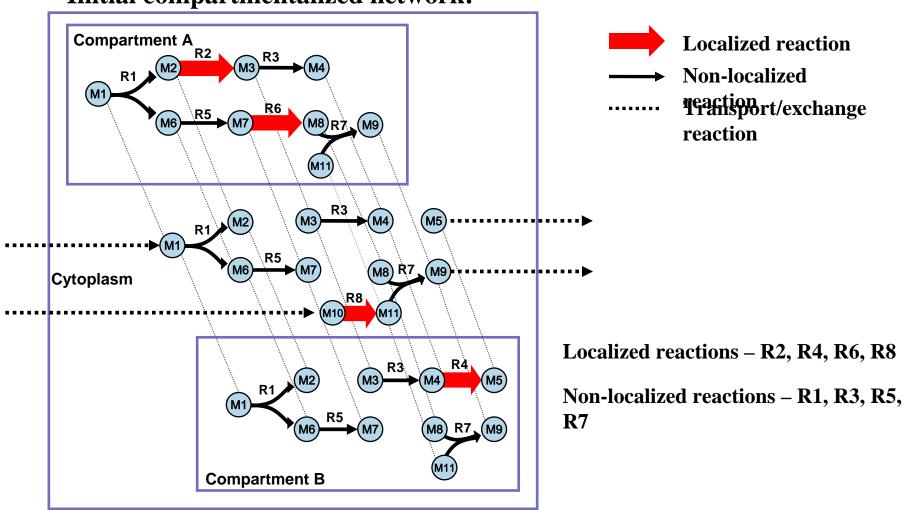
Output (prediction):



Mintz, et al., ISMB & Bioinformatics, 2009

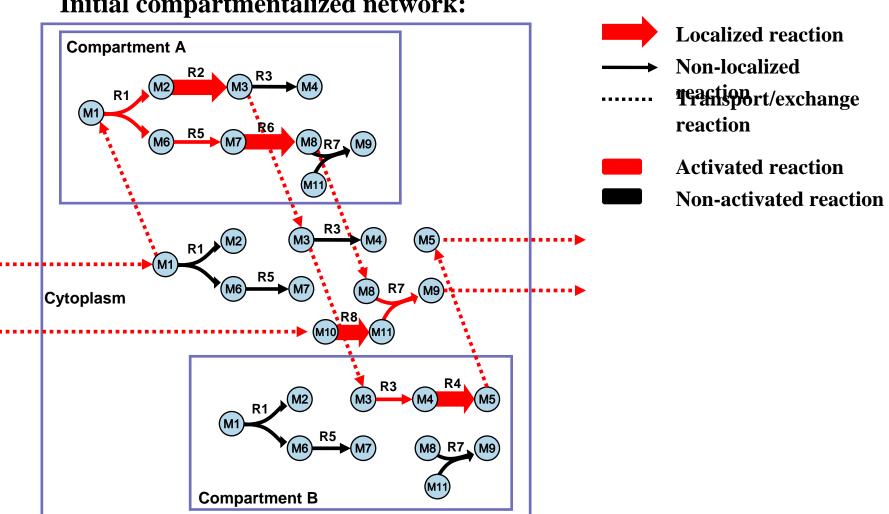
## Example

#### **Initial compartmentalized network:**



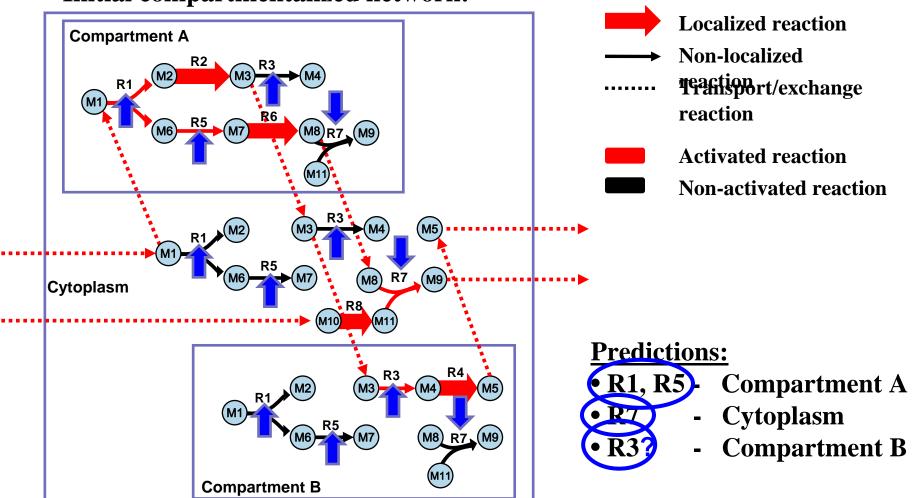
### Example – Flux Distribution

#### **Initial compartmentalized network:**



## Example – Results

#### **Initial compartmentalized network:**



## Validating Predictions via Metabolic Network of *S. cerevisiae*

- Genome-scale, fully compartmentalized metabolic network model of (Duarte et al, 2004)
- 1062 metabolites, 1149 reactions, 7 compartments

- \* To evaluate our method:
  - 1. Remove existing localization data
  - 2. Cross validation test- random localized vs. non-localized sets
  - 3. Apply our method
  - 4. Compute accuracy (compared to experimental data)
    - coverage (portion of predictions with single predicted compartment)



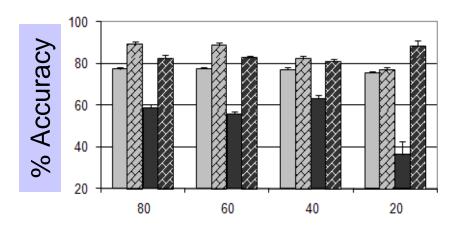
## Comparison to Pathway Enrichment-Based Method

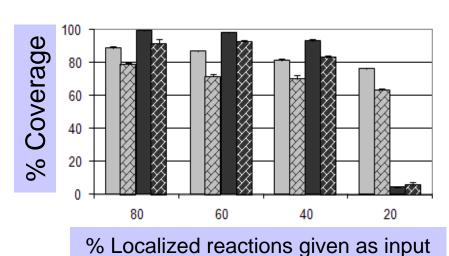
- Localization is determined based on the assignment of enzymes in pre-determined biochemical pathways
- For each pathway compute a set of hyper-geometric *p*-values reflecting the **pathway's enrichment for all compartments**, respectively
- □ Prediction based on compartment yielding the lowest *p*-value in the corresponding pathway



#### Results

#### Accuracy and coverage for various fractions of localized





- Metabolic-network prediction (cytosolic reactions)
- Metabolic-network prediction (non-cytosolic reactions)
- Enrichment-based prediction (cytosolic reactions)
- Enrichment-based prediction (non-cytosolic reactions)
  - Robust accuracy
  - Moderate coverage decline



## Collaborators

My lab
 Naama Tepper
 Edward Vitkin
 Roi Adadi

- Eytan Ruppin's lab (Tel-Aviv)
   Tomer Benyamini
   Ori Folger
   Idit Diamant
- Asaph Aharoni's lab (Weizmann)
   Shira Mintz