An integrative, multi-scale, genome -wide model reveals the phenotypic landscape of Escherichia coli

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Background

Materials and Methods

Results

Discussion

Background

Potential

• The development of an integrative genome-scale model: Holy Grail

discovery of novel properties and emerging behaviors

generating and testing predictable hypotheses

guiding experimentation

accelerating the in-depth understanding of cellular physiology

Early work

• E-cell

a modular software environment for whole-cell simulation that included organelle sub-models

More recently work

genome-scale simulations were performed to study complex phenomena

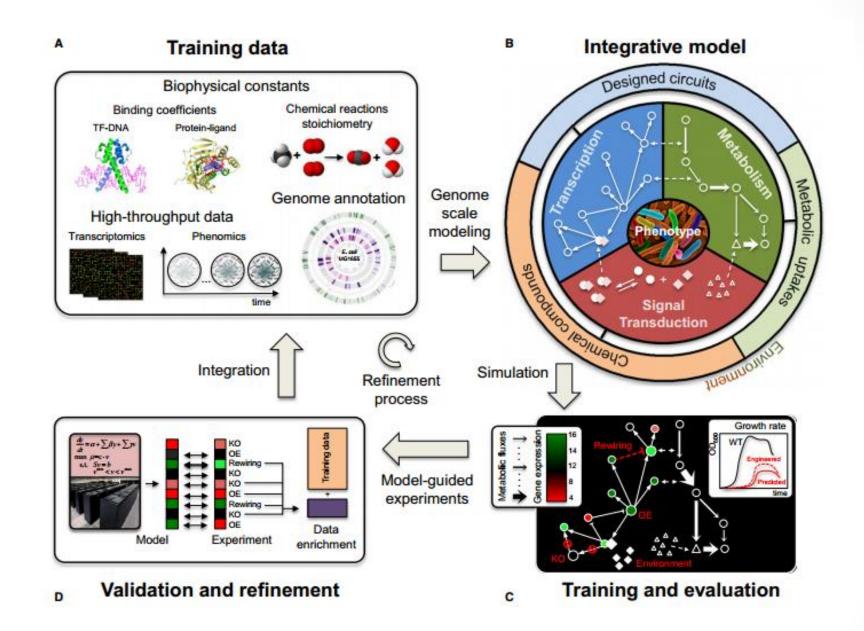
- the emergence of anticipatory behavior during evolution in varying environments
- the noise contributions of an inducible switch
- the effect of stochastic expression to metabolic variability

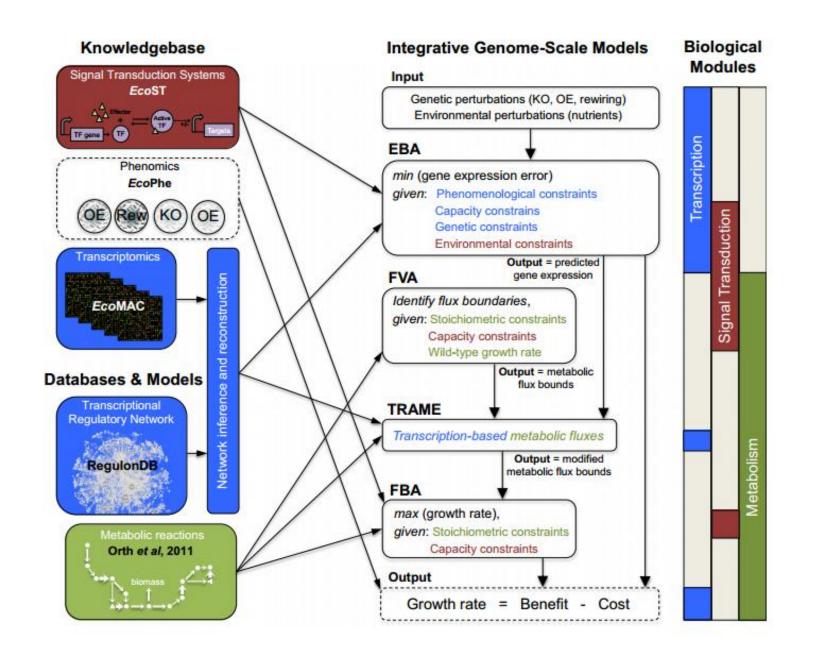
Our aim

To construct a phenomenological model for bacterial organisms that integrates multiple layers of biological organization.

Why E.coli

- The wealth of data and knowledge accumulated over the years
- The easiness to culture and manipulate experimentally
- Its importance in medical and biotechnological applications





Materials and Methods

Data

Cellular sub-models

Model integration

Data

• gene expression

- > from GEO, ASAP database
- constructed a gene expression compendium of 4,189 genes over 2,198 arrays that were collected from 127 scientific articles

signal transduction

- A total of 328 transcription factors (TFs) and 1,357 enzymes were identified by using RegulonDB
- identify 151 instances of signal transduction systems (STSs)

• Phenomics compendium

➤ bacterial growth information for 616 of the arrays in EcoMAC by EcoPhe

Cellular sub-models

Signal transduction model

$$y_{\mathrm{TF}} = y_{\mathrm{TF}}^{\mathrm{wt}} + \Omega (C_{\mathrm{TF}}^{\mathrm{max}} - C_{\mathrm{TF}}^{\mathrm{min}}) \chi_{\mathrm{TF}}^{E} \frac{\Delta n_{E}}{\Delta n_{E}^{\mathrm{max}}}$$

• Transcriptional model and EBA

$$E = \frac{1}{2} \begin{bmatrix} \bar{y}_{\text{TF}} & \bar{\epsilon}_{\text{TF}} \end{bmatrix} \bar{\bar{H}} \begin{bmatrix} \bar{y}_{\text{TF}} \\ \bar{\epsilon}_{\text{TF}} \end{bmatrix} + \bar{f} \begin{bmatrix} \bar{y}_{\text{TF}} \\ \bar{\epsilon}_{\text{TF}} \end{bmatrix}$$

• Metabolic model and Transcription-based Flux Enrichment

$$PV_{\min} \le v \le PV_{\max}$$
, where $P_e = \left(\frac{y_e}{y_e^{\text{wt}}}\right)^n$

Model integration

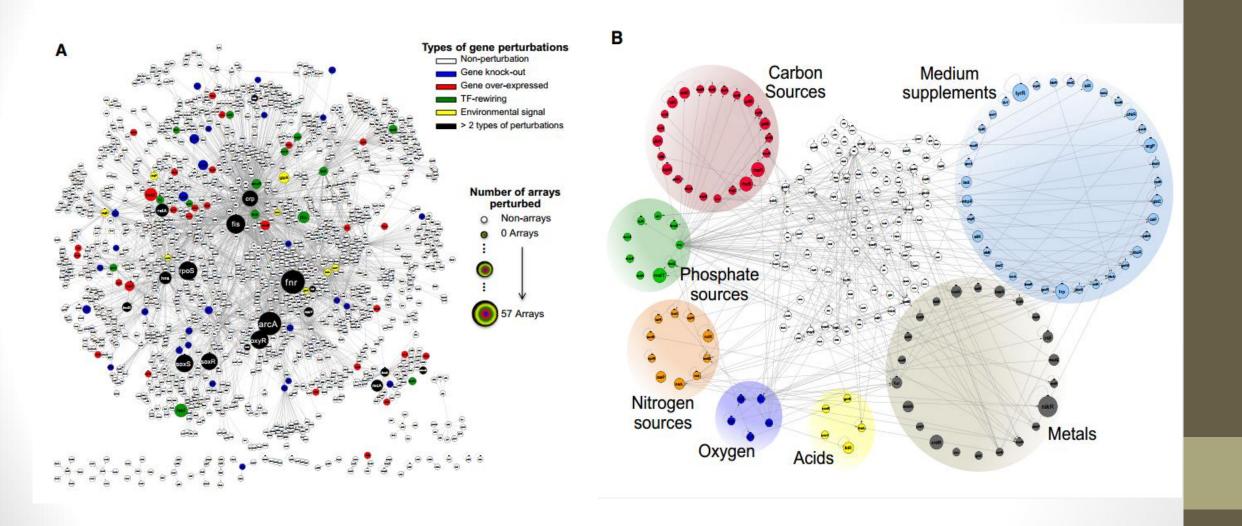
$$c = \frac{1}{N_G} \sum_{g} \left| \frac{\bar{y}_g - y_g^{\text{WT}}}{y_g^{\text{WT}}} \right|$$

$$\bar{\mu} = B - c$$

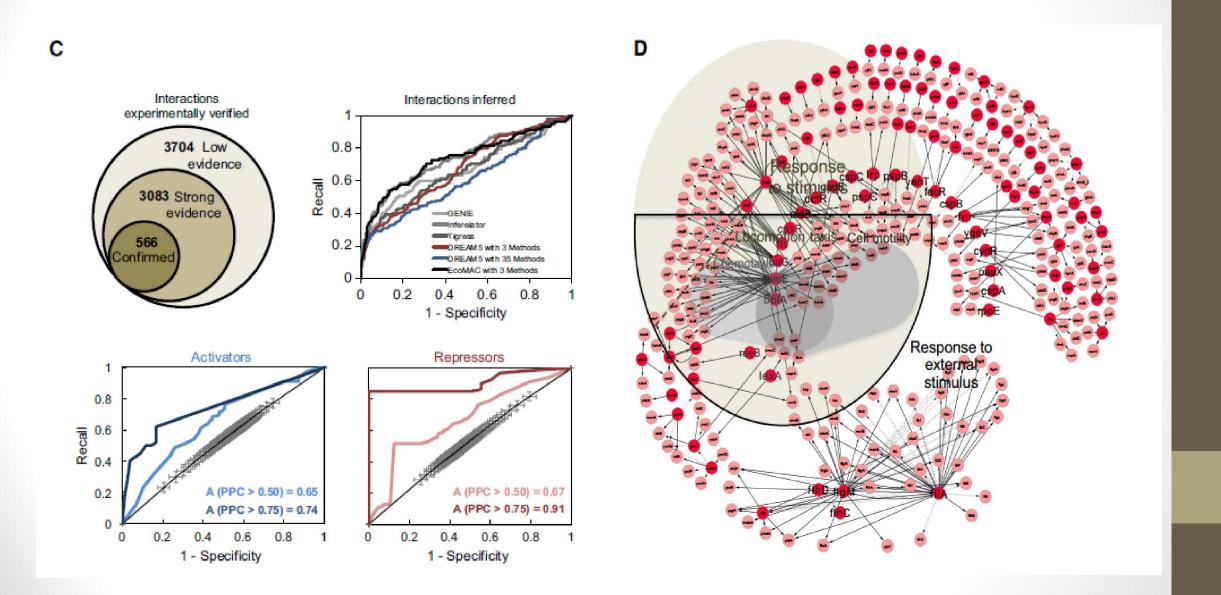
Results

- Genetic and environmental gene expression diversity
- An integrative knowledgebase as a base to regulatory network enrichment
- Expression Balance Analysis
- Phenotypic predictions in an integrated model
- Model enrichment through targeted experimentation

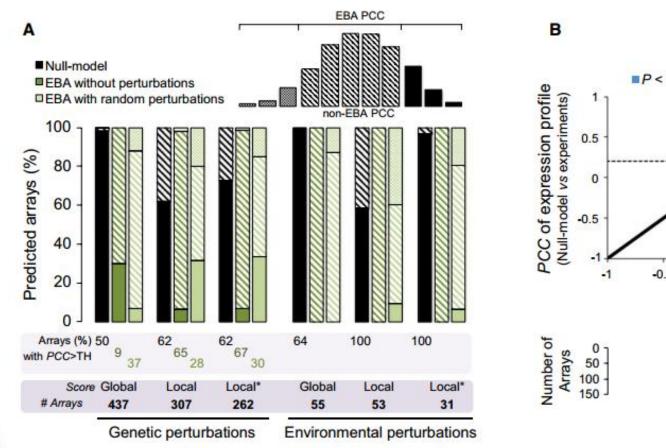
Genetic and environmental gene expression diversity

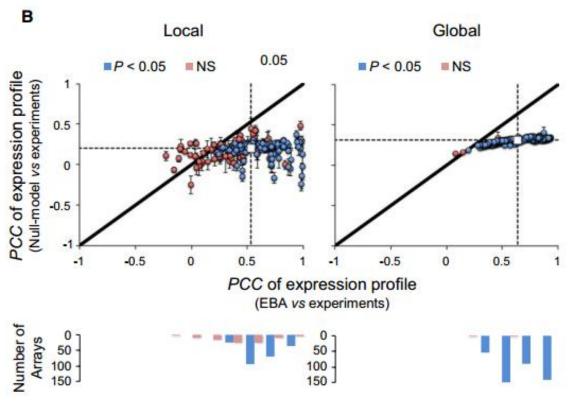


An integrative knowledgebase as a base to regulatory network enrichment

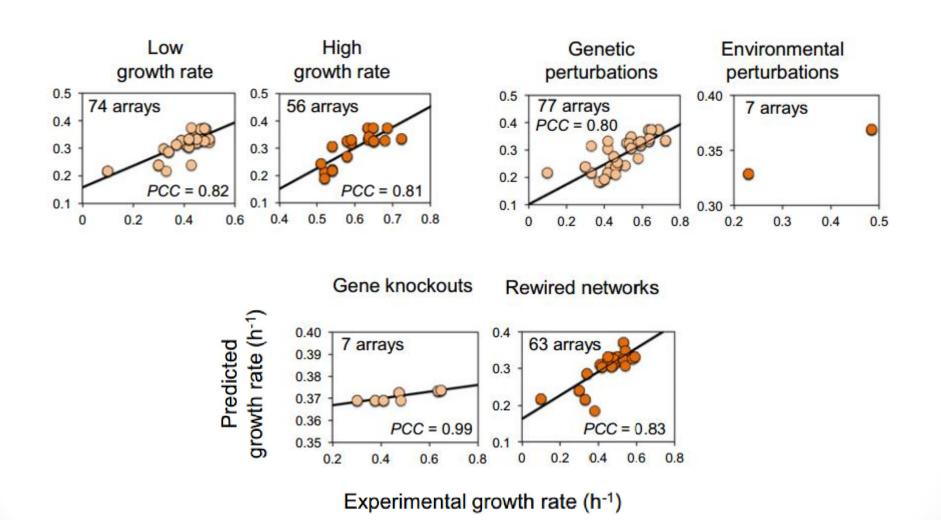


Expression Balance Analysis

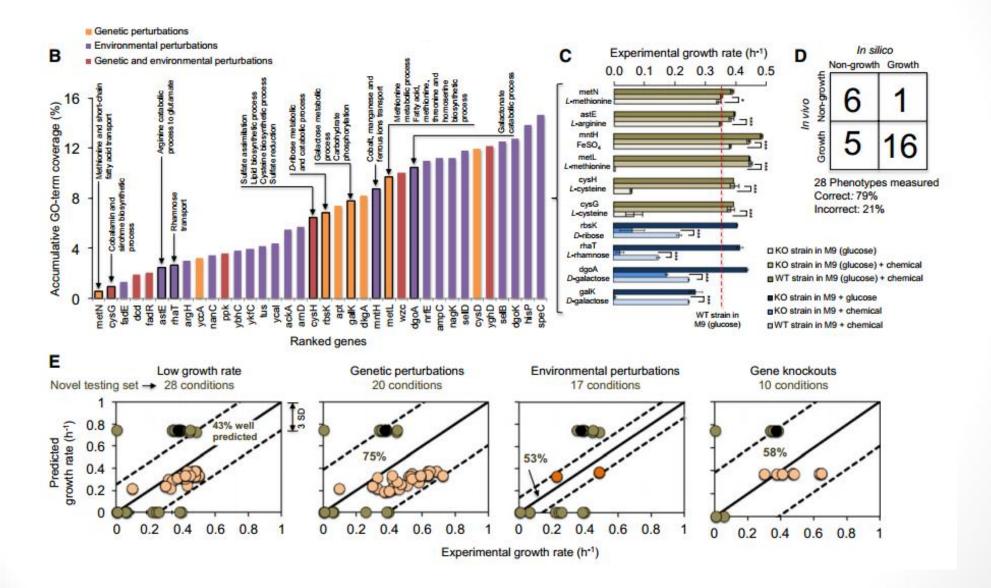




Phenotypic predictions in an integrated model



Model enrichment through targeted experimentation



Discussion

- Advantage
- > the creation of a signal transduction network (EcoST)
- its integration to the transcriptional and metabolic network through constraint modeling
- Disadvantage
- ➤ Coverage
- the severe bias to negative samples in the ground truth

Inspiration

Binding site

Funtion

Thanks for your attention