



Introduction

Hidde de Jong
IBIS
INRIA Grenoble – Rhône-Alpes
Hidde.de-Jong@inria.fr

September 26, 2012

INRIA Grenoble - Rhône-Alpes and IBIS



- IBIS: systems biology group at INRIA/Université Joseph Fourier/CNRS
 - Analysis of bacterial regulatory networks by means of models and experiments
 - Biologists, computer scientists, mathematicians, physicists, ...

<http://ibis.inrialpes.fr>

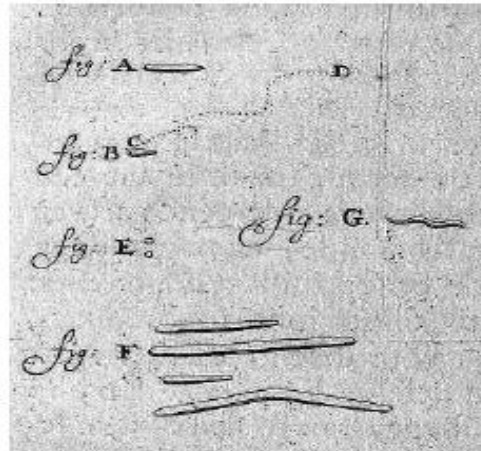


Bacteria

- Bacteria were first observed by Antonie van Leeuwenhoek, using a single-lens microscope of his own design



<http://commons.wikimedia.org/>



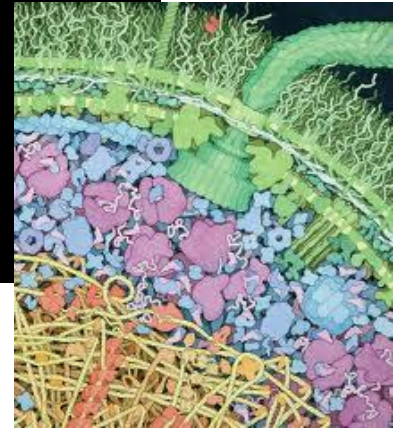
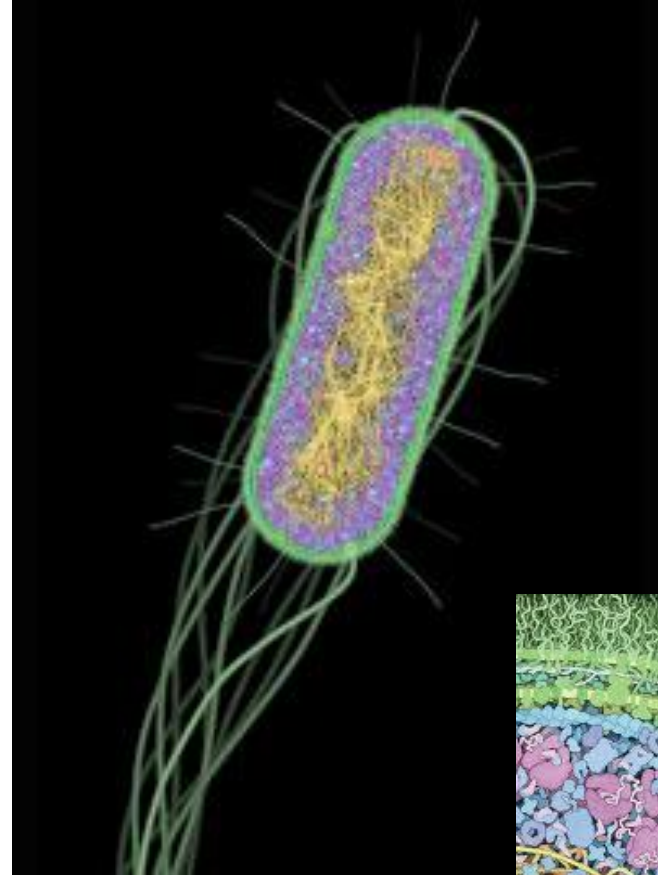
www.euronet.nl/users/wamar/leeuwenhoek.html

van Leeuwenhoek A (1684),
Philosophical Transactions
(1683–1775) 14: 568–574

*"In the morning I used to rub my teeth with salt and rinse my mouth with water and after eating to clean my molars with a toothpick.... I then most always saw, with great wonder, that in the said matter there were many very **little living animalcules**, very prettily a-moving. The biggest sort had a very strong and swift motion, and shot through the water like a pike does through the water; mostly these were of small numbers."*

Bacteria are complex living systems

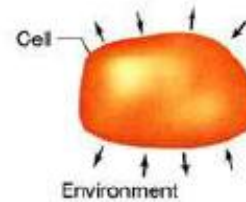
- Bacterial cells are complex biochemical and biophysical machines
 - Wide range of shapes, typically 0.5-5 μm in length
 - 10^6 bacterial cells in 1 ml of fresh water
 - 10 times as much bacterial cells as human cells in human body



Goodsell (2010), *The Machinery of Life*, Springer, 2nd ed.

Bacteria are complex living systems

- Bacterial cells are complex biochemical and biophysical machines
- Bacteria possess characteristics shared by most living systems:
 - Metabolism
 - Growth and reproduction
 - Differentiation
 - Communication
 - Evolution



1. Metabolism

Uptake of chemicals from the environment, their transformation within the cell, and elimination of wastes into the environment. The cell is thus an open system.



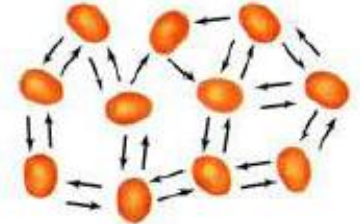
2. Reproduction (growth)

Chemicals from the environment are turned into new cells under the direction of preexisting cells.



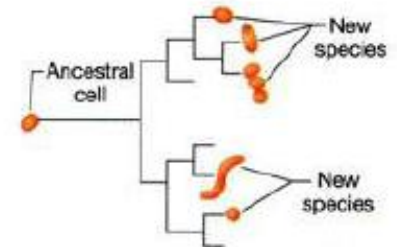
3. Differentiation

Formation of a new cell structure such as a spore, usually as part of a cellular *life cycle*.



4. Communication

Cells *communicate* or *interact* primarily by means of chemicals that are released or taken up.



5. Evolution

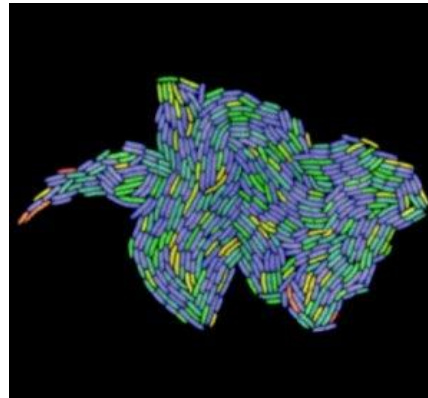
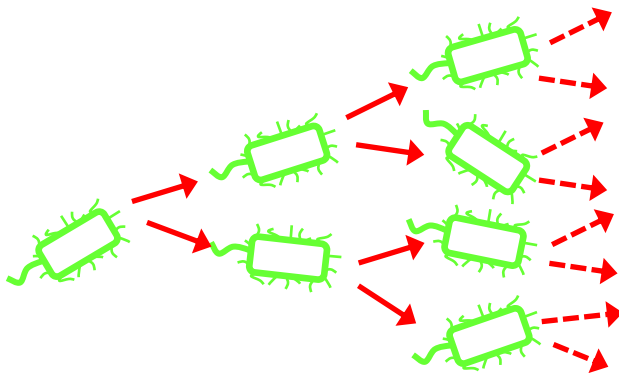
Cells *evolve* to display new biological properties. Phylogenetic trees show the evolutionary relationships between cells.

Madigan *et al.* (2003), *Brock Biology of Microorganisms*, Prentice Hall, 10th ed.

Bacterial growth and metabolism

- Bacteria are geared towards **growth** and **division**

Escherichia coli cells have doubling times up to 20 min



Stewart *et al.* (2005), *PLoS Biol.*, 3(2): e45

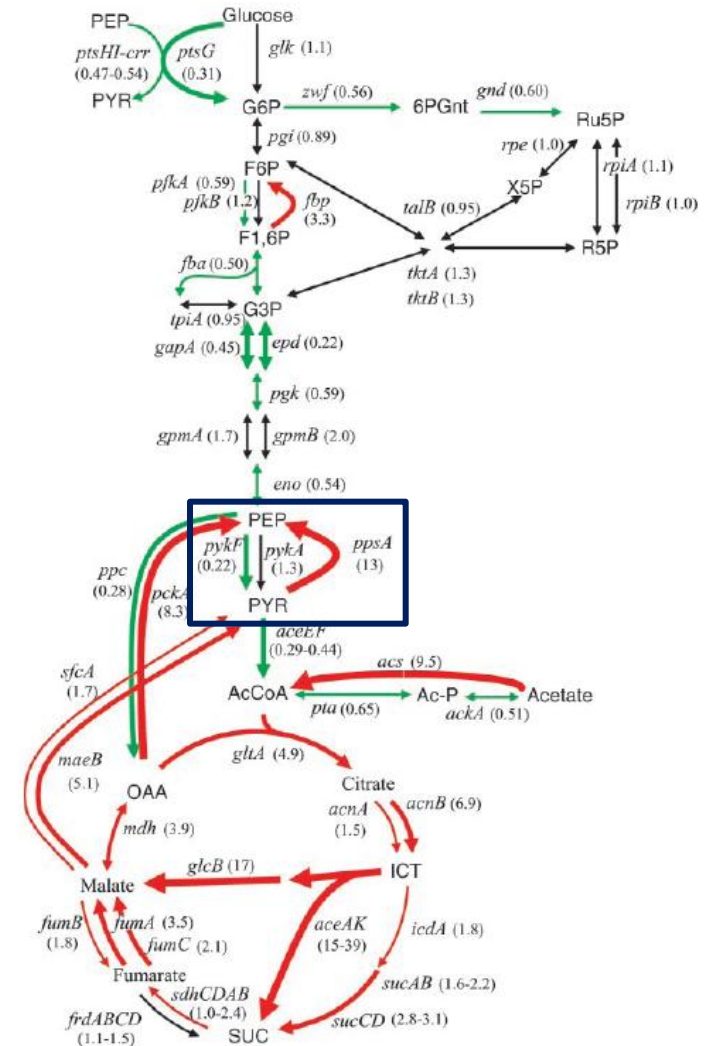
- Metabolism** fuels growth by production of energy and building blocks for macromolecules, using nutriment from environment

ATP, amino acids, nucleotides, ...

Bacterial growth and metabolism

- Central **carbon metabolism** breaks down carbon sources for energy production and macromolecular synthesis

Glucose, acetate, lactose, ...



Oh *et al.* (2002), *J. Biol. Chem.*, 277(15):13175–83

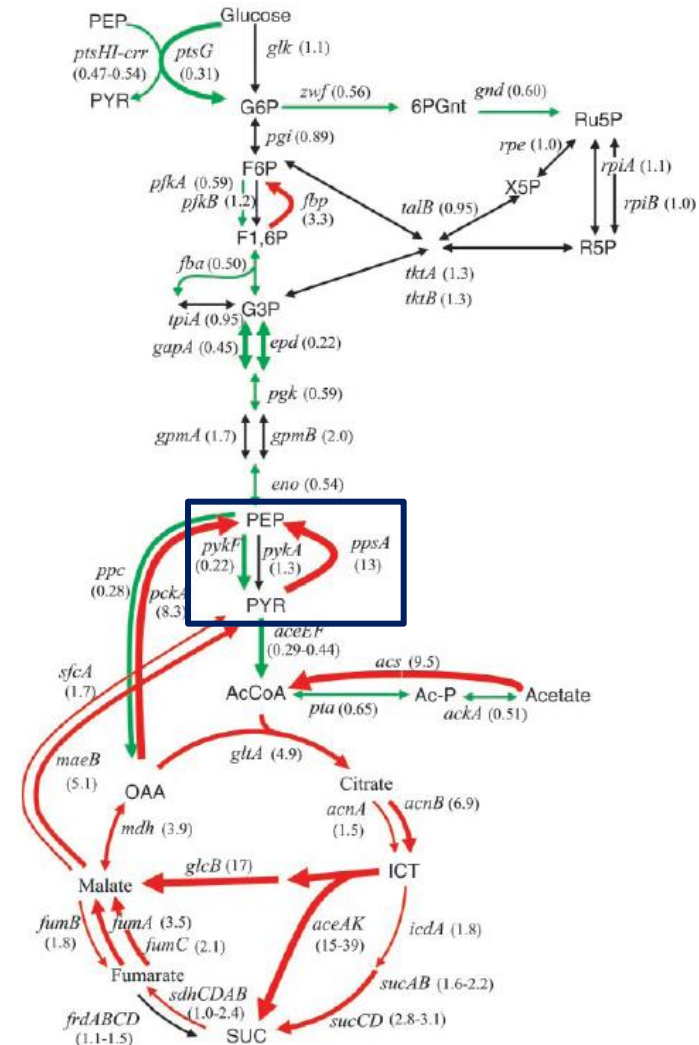
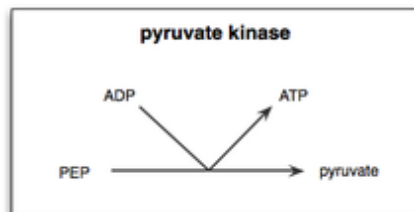
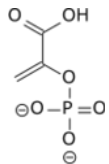
Bacterial growth and metabolism

- Central **carbon metabolism** breaks down carbon sources for energy production and macromolecular synthesis

Glucose, acetate, lactose, ...

- **Enzymes** catalyze individual steps in metabolic network

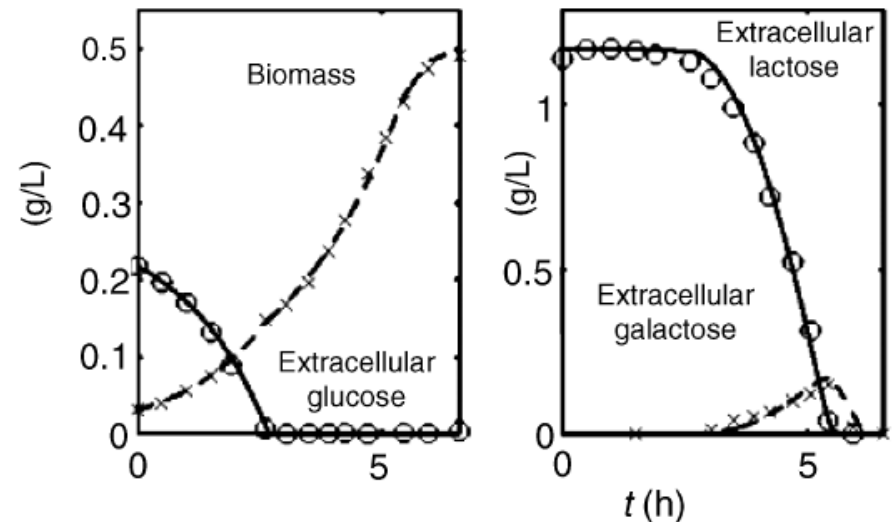
Pyruvate kinase transforms phosphoenolpyruvate (PEP) into pyruvate



Bacterial growth and metabolism

- Bacteria can sequentially use different sugars, in preferential order

Diauxic growth on glucose and lactose



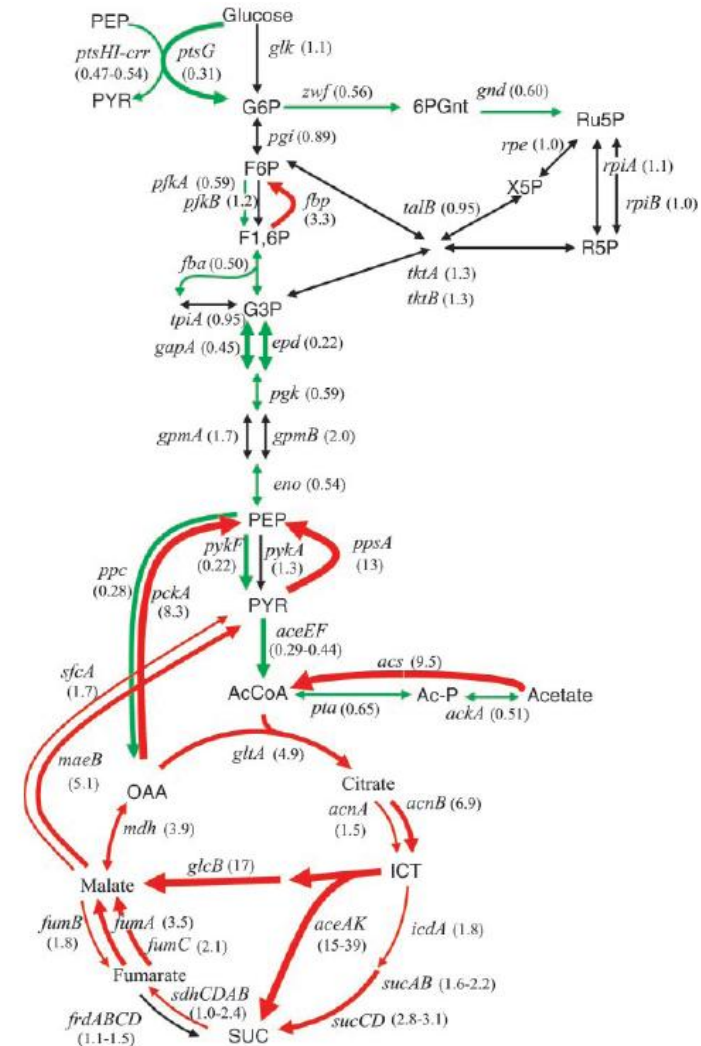
Bettenbrock *et al.* (2006), *J. Biol. Chem.*,
281(5):2578-84

Bacterial growth and metabolism

- Bacteria can sequentially use different carbon sources, in preferential order
- Adaptation of bacteria to growth on different carbon source involves changes in **metabolic fluxes**

Different flux directions in central metabolism of *E. coli* during growth on glucose (**glycolysis**) and acetate (**gluconeogenesis**)

Oh *et al.* (2002), *J. Biol. Chem.*, 277(15):13175–83

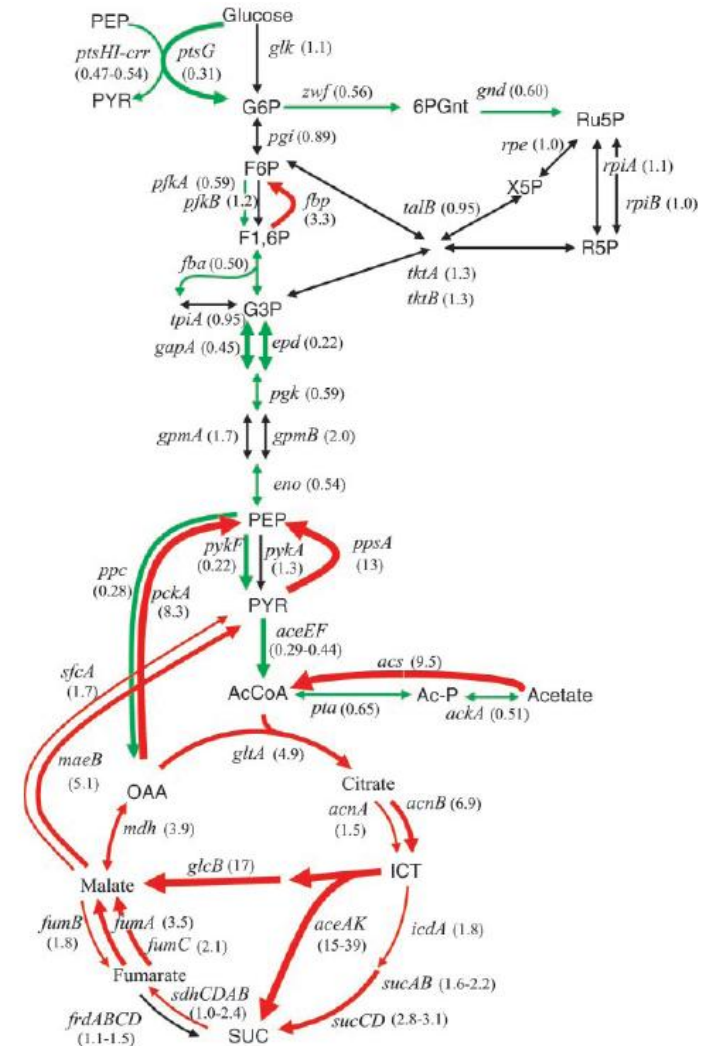


Bacterial growth and metabolism

- Bacteria can sequentially use different carbon sources, in preferential order
- Adaptation of bacteria to growth on different carbon source involves adjustment of **enzyme levels**

Differences in expression of enzymes in central metabolism of *E. coli* during growth on glucose and acetate

Oh *et al.* (2002), *J. Biol. Chem.*, 277(15):13175–83



Bacterial growth and metabolism

- Bacteria can sequentially use different carbon sources, in preferential order
- Adaptation of bacteria to growth on different carbon source involves adjustment of **metabolite levels**

Absolute measurement of metabolite concentrations in *E. coli* cells growing on glucose

Majority of metabolites present at significantly different concentrations in cells growing on acetate rather than glucose

Bennett *et al.* (2009), *Nat. Chem. Biol.*, 5(8):593-9

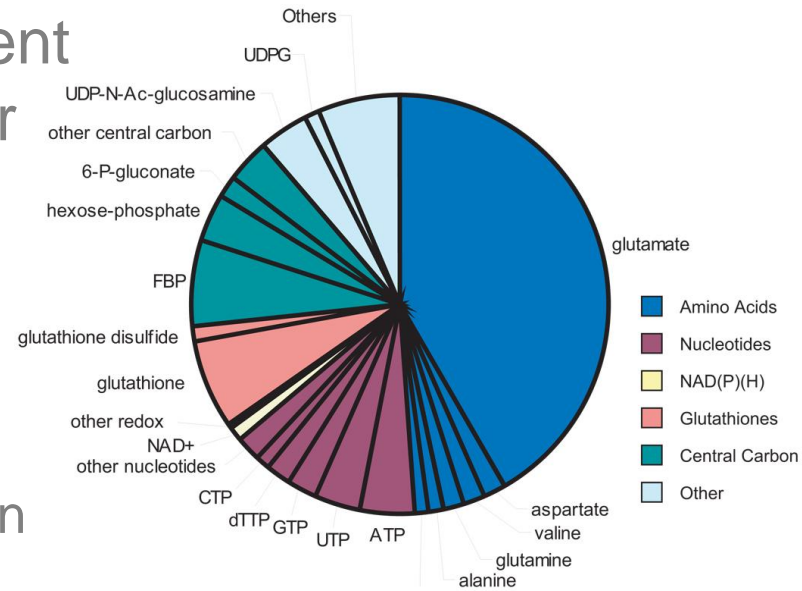


Table 1 Intracellular metabolite concentrations in glucose-fed, exponentially growing *E. coli*

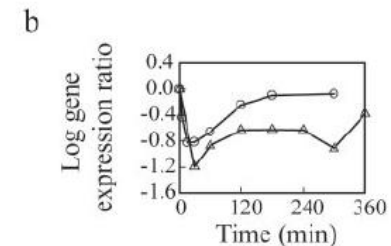
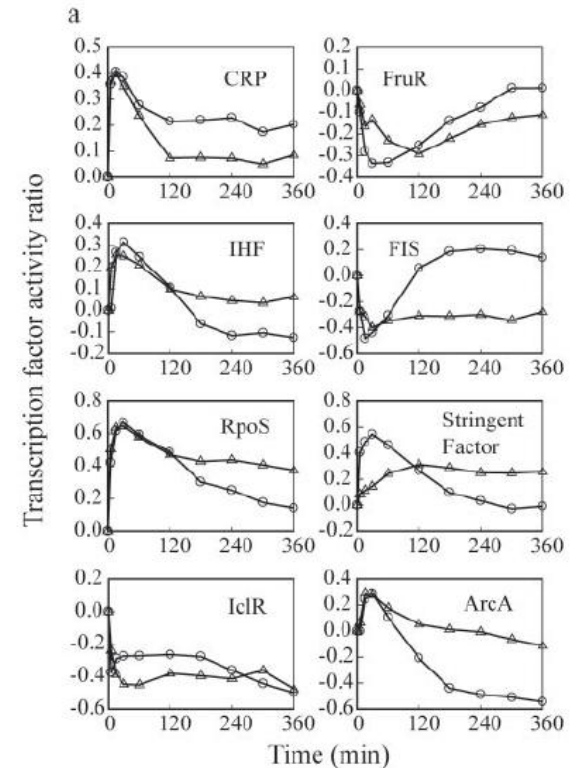
Metabolite	mol l ⁻¹	Metabolite	mol l ⁻¹
Glutamate	9.6×10^{-2}	UDP-glucuronate (51)	5.7×10^{-4}
Glutathione	1.7×10^{-2}	ADP	5.6×10^{-4}
Fructose-1,6-bisphosphate	1.5×10^{-2}	Asparagine (52)	5.1×10^{-4}
ATP	9.6×10^{-3}	α -Ketoglutarate	4.4×10^{-4}
UDP-N-acetylglucosamine (29)	9.2×10^{-3}	Lysine (53)	4.1×10^{-4}
Hexose-P ^a	8.8×10^{-3}	Proline (54)	3.9×10^{-4}
UTP (30)	8.3×10^{-3}	dTDP (55)	3.8×10^{-4}
GTP (31)	4.9×10^{-3}	Dihydroxyacetone phosphate	3.7×10^{-4}
dTTP	4.6×10^{-3}	Homocysteine (56)	3.7×10^{-4}
Aspartate	4.2×10^{-3}	CMP (57)	3.6×10^{-4}
Valine (32)	4.0×10^{-3}	Deoxyribose-5-P (58)	3.0×10^{-4}
Glutamine	3.8×10^{-3}	Isoleucine (59)+leucine (60)	3.0×10^{-4}
6-Phosphogluconate	3.8×10^{-3}	AMP	2.8×10^{-4}

Bacterial growth and metabolism

- Bacteria can sequentially use different carbon sources, in preferential order
- Adaptation of bacteria to growth on different carbon source involves adjustment of **transcription factor levels**

Changes in activity and concentration of transcription factors during glucose-acetate diauxie in *E. coli*

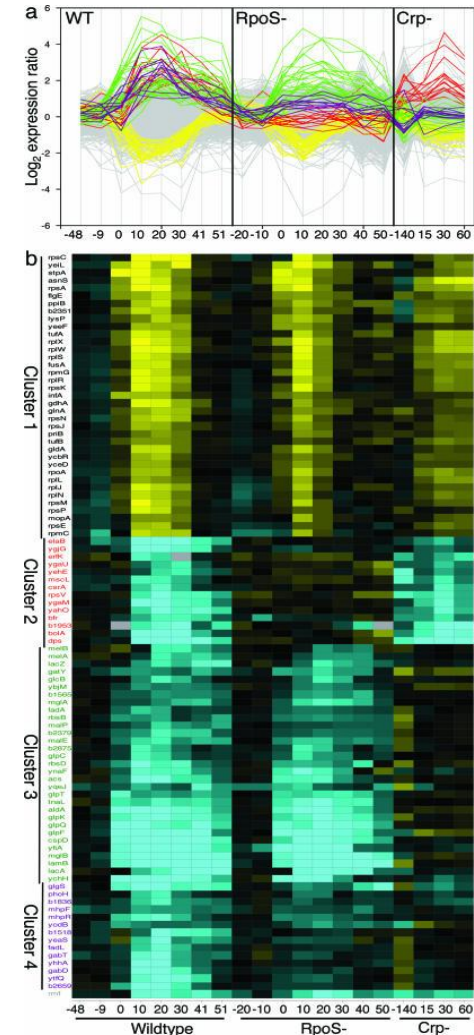
Kao *et al.* (2005), *J. Biol. Chem.*, 280(43):36079–87



Growth adaptation and gene expression

- Genome-wide **reorganization of gene expression** following growth transitions in bacteria

Gene expression during glucose-lactose diauxie in *E. coli*, in wild-type and transcription factor mutants



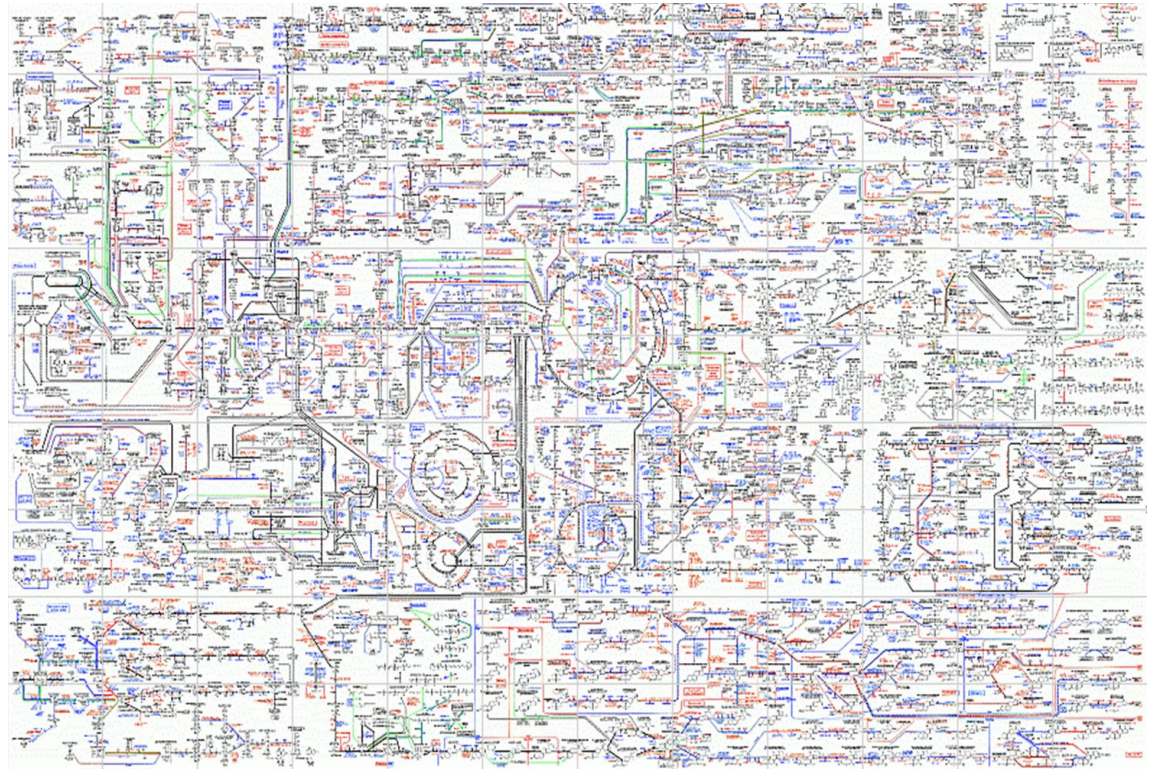
Traxler *et al.* (2006), *Proc. Natl. Acad. Sci. USA*, 103(7):2374–9

General question on cellular adaptation

- Cells are capable of responding to a variety of changes in their environment by adapting their physiology
 - Change in carbon source, starvation, population density, ...
- On the molecular level, these responses involve adjustment of metabolism and gene expression
 - Cellular concentrations of metabolites, enzymes, transcription factors, ...
- **Question:** how does cell coordinate these adaptive responses?

Coordination of adaptive responses

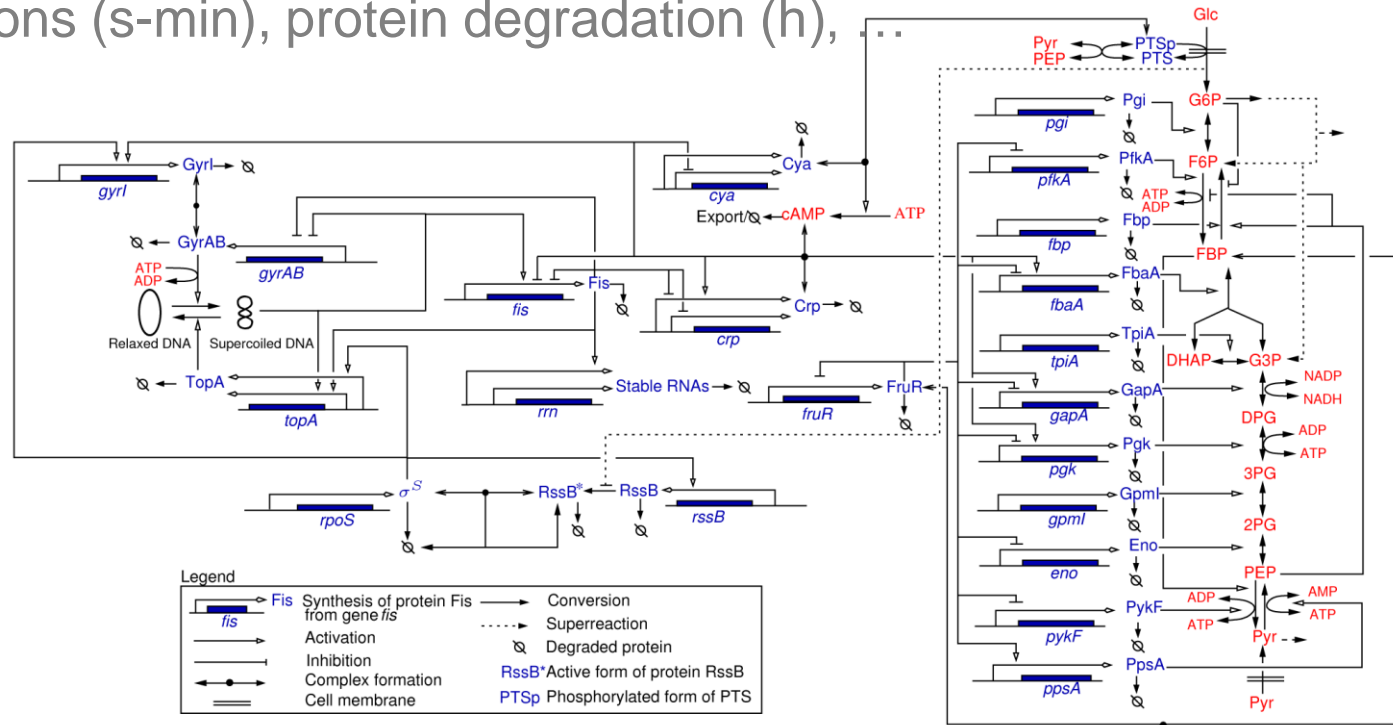
- Coordination involves **regulation** of functioning of **biochemical reaction networks**
 - Most networks are **large and complex**
 - E. coli has ± 200 metabolic pathways, involving ± 900 enzymes and ± 1000 reactions



Karp *et al.* (2007), *Nucleic Acids Res.*,
35(22):7577-90

Coordination of adaptive responses

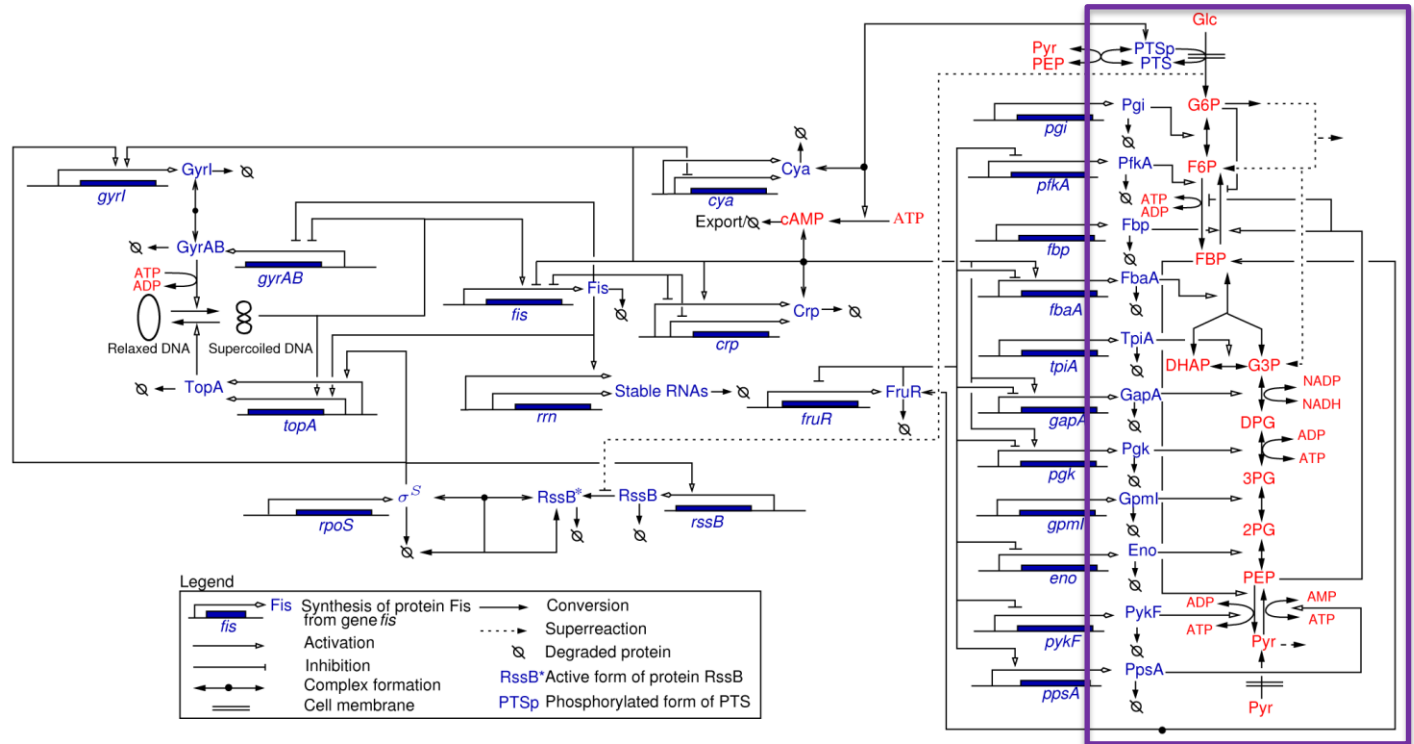
- Coordination involves **regulation** of functioning of **biochemical reaction networks**
 - Most networks involve **variety of biochemical reaction mechanisms**, operating on **different time-scales**: enzymatic reactions (s-min), protein degradation (h), ..



Baldazzi et al. (2010), *PLoS Comput. Biol.*, 6(6):e1000812

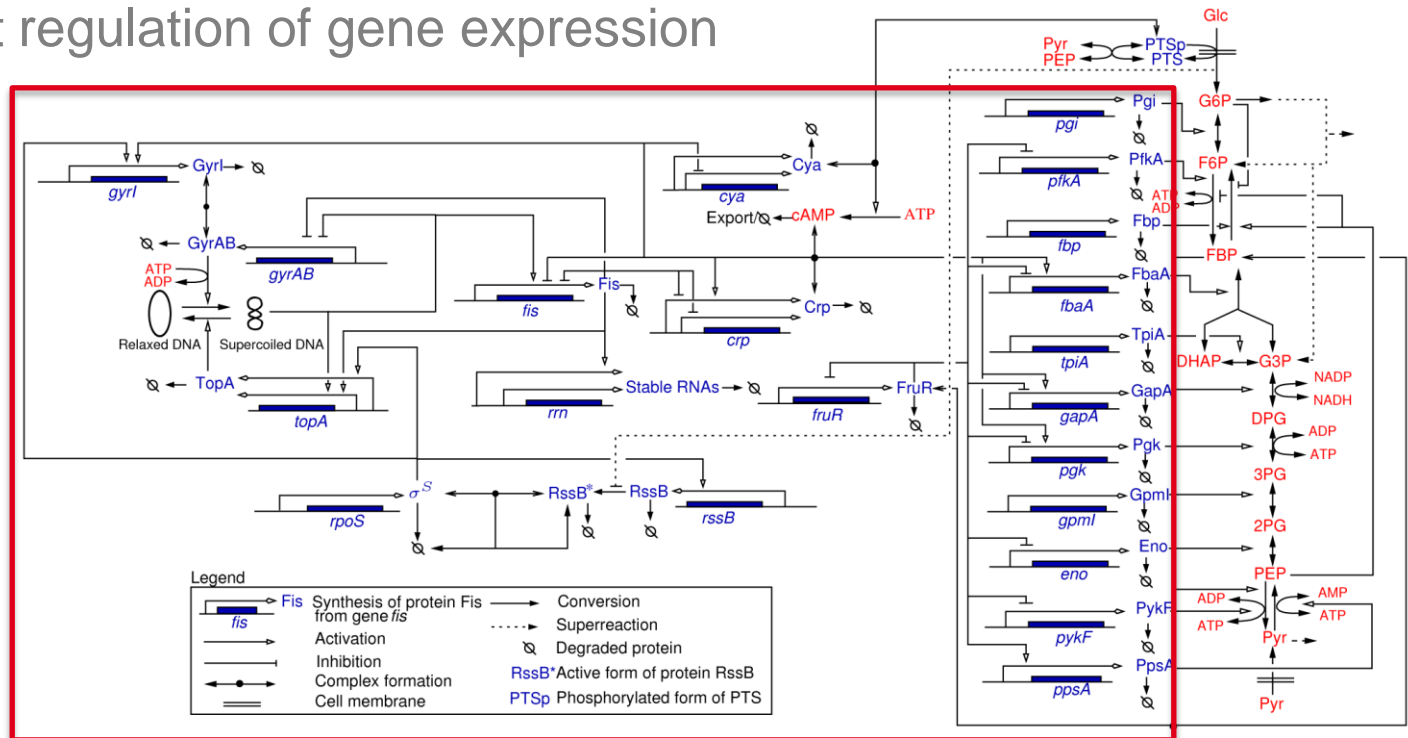
Types of biochemical reaction networks

- Types of networks distinguished by focusing on specific interactions and different time-scales:
 - Metabolic networks:** metabolites and enzymatic reactions



Types of biochemical reaction networks

- Types of networks distinguished by focusing on specific interactions and different time-scales:
 - Metabolic networks:** metabolites and enzymatic reactions
 - Gene regulatory networks:** genes, RNAs, proteins, and direct and indirect regulation of gene expression



Analysis of network functioning: from structure to dynamics

- Wealth of knowledge on network structure in many bacteria
 - Scientific databases and repositories
 - Primary experimental literature
- Comprehension of network functioning requires observed system **dynamics** to be related to network **structure**
- Mathematical modeling and computer simulation indispensable for dynamic analysis of biochemical reaction networks
- Analysis of network functioning has a central place in emerging field of **systems biology**

Alon (2007), *An Introduction to Systems Biology*, Chapman & Hall/CRC Press

Historical note

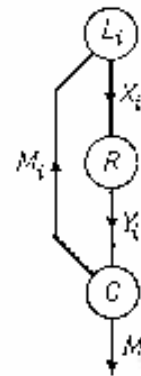
- Systems biology, and more particularly the mathematical modeling and computer simulation of biochemical reaction networks, have a long history

Westerhoff and Palsson, *Nat. Biotechnol.*, 22(10):1249-52

- Simulation of metabolic pathways (glycolysis)

Garfinkel *et al.* (1970), *Ann. Rev. Biochem.*, 39:473-98

- Modeling of gene regulatory networks



Goodwin (1963), *Temporal Organization in Cells*

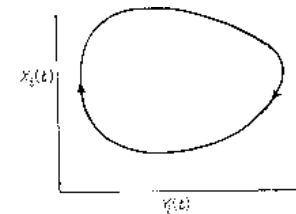
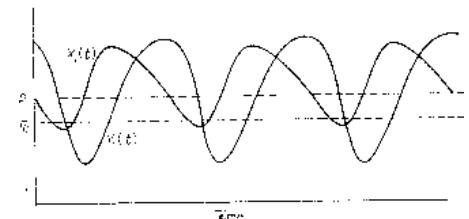


FIGURE 3.



Mathematical modeling of biochemical reaction networks

- Well-established framework for modeling of biochemical reaction networks using **ordinary differential equation (ODE)** models
- General form of ODE models of biochemical reaction networks

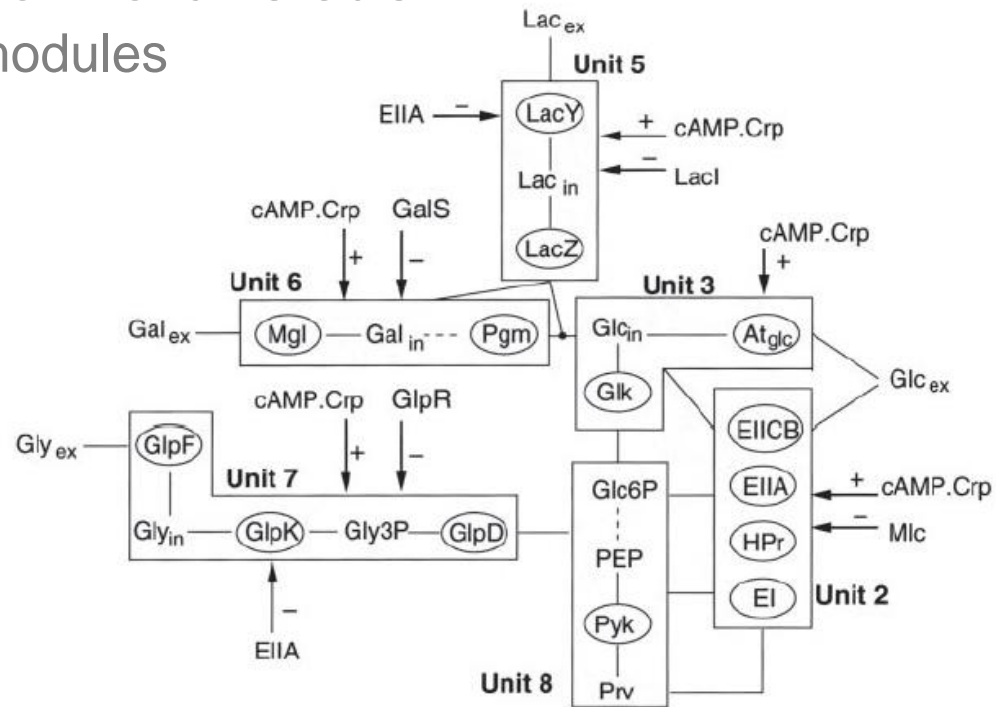
$$\dot{x} = N v(x)$$

- Concentration variables $x \in \mathbb{R}_+^n$
- Reaction rates $v : \mathbb{R}_+^n \rightarrow \mathbb{R}^q$
- Stoichiometry matrix $N \in \mathbb{Z}^{n \times q}$
- Various forms of kinetic rate laws: mass-action, Michaelis-Menten, Hill, Monod-Wyman-Changeux, ...

Heinrich and Schuster (1996), *The Regulation of Cellular Systems*, Chapman & Hall

Example of network modeling

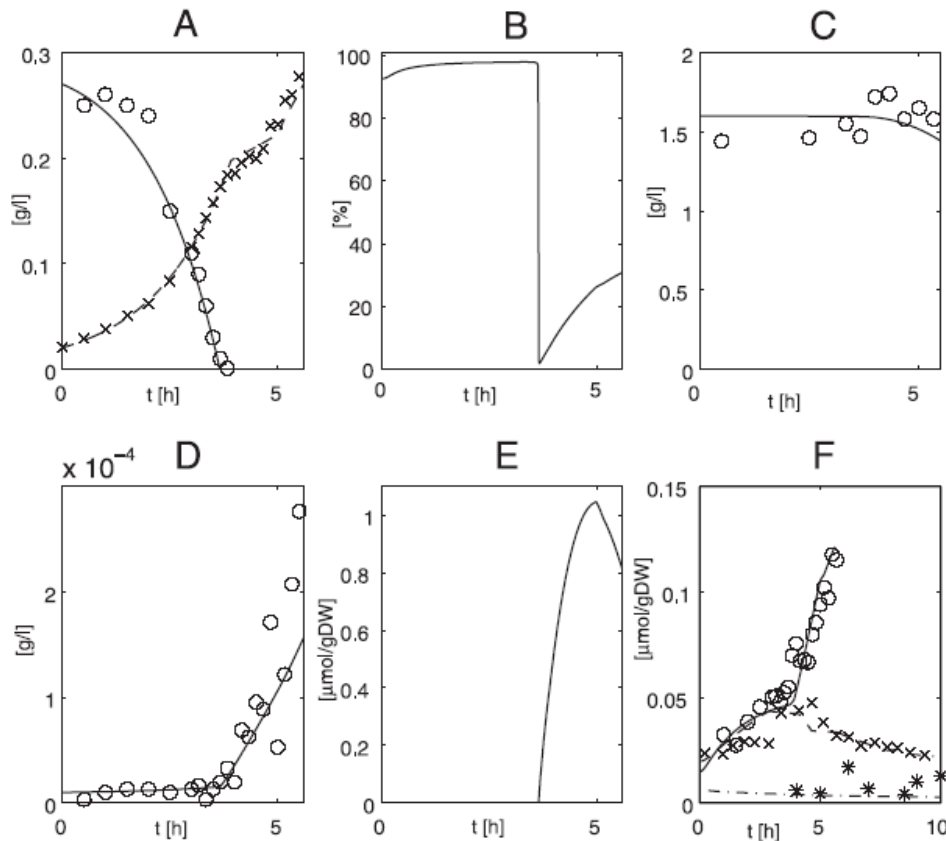
- Model of uptake of carbon sources (glucose, lactose, glycerol, ...) by *E. coli*
 - Several dozens of equations and more than a hundred parameters, many of them unknown or unreliable
 - Mostly metabolic modules



Bettenbrock *et al.* (2005), *J. Biol. Chem.*, 281(5): 2578-2584

Example of network modeling

- Estimation of parameter values from time-series measurements of metabolite concentrations on wild-type and mutant strains



- Model has good predictive capability

Bettenbrock *et al.* (2005), *J. Biol. Chem.*, 281(5): 2578-2584

Issues in mathematical modeling

- Mathematical models are used for explanation, prediction, and control
- Modeler confronted with several **practical problems**
 - Models of actual networks are large systems of nonlinear ODEs
 - Parameter values are generally unknown and difficult to measure directly
 - Reaction mechanisms are often unknown
 - Experimental measurements of variables are scarce, noisy, and indirect
- This raises issues in model reduction and approximation, parameter estimation, network inference, data analysis, ...
- But also: issues in experimental data acquisition

Objective of course "Modeling of biological networks"

- **Course objective** is to master kinetic modelling as applied to metabolic and gene regulatory networks
 - Both the theoretical foundations and concrete applications to diverse systems of biological regulation
 - Applications will rely on the practical use of computer tools for the modelling, analysis and simulation of biological networks

Program and teachers

- Part 1. Systems biology and kinetic modeling (courses 7 h)
 - Reminders on dynamical systems (Hidde de Jong)
 - Introduction to regulatory systems (Hans Geiselmann)
 - Reminders on kinetic modeling and enzymology (Daniel Kahn)
- Part 2. Metabolic network modeling (courses 6 h, and practicals 9 h)
 - Introduction to metabolic networks (Daniel Kahn)
 - Metabolic Control Theory (Daniel Kahn)
 - Practical on the modeling of a metabolic system using COPASI (Daniel Kahn)

Program and teachers

- Part 3. Gene regulatory network modeling (courses 16 h, and practicals 6 h)
 - Introduction to recent techniques for measuring gene expression (Hidde de Jong)
 - Kinetic models of gene expression and dynamics of gene regulatory networks (Hidde de Jong)
 - Identification and inference of gene network models (Eugenio Cinquemani)
 - Practical on the qualitative modeling of bacterial regulatory networks, using GNA (Hidde de Jong)
- Part 4. Towards integrated models of regulatory networks (courses 2 h)
 - MetaGenoReg project (Daniel Kahn and Hidde de Jong)

Evaluation

- Metabolic network modeling:
 - Exercises handed out during course
- Gene regulatory network modeling:
 - Questions on articles handed out during course
 - Or: literature review on specific topic of interest
- Grade is average of grades for two subparts of course
- Articles will be made available via course web site

Merci



www.inrialpes.fr/ibis