



# Metabolic Control Analysis (MCA)

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# INRIA Grenoble - Rhône-Alpes and IBIS



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

<http://team.inria.fr/ibis>



# Overview

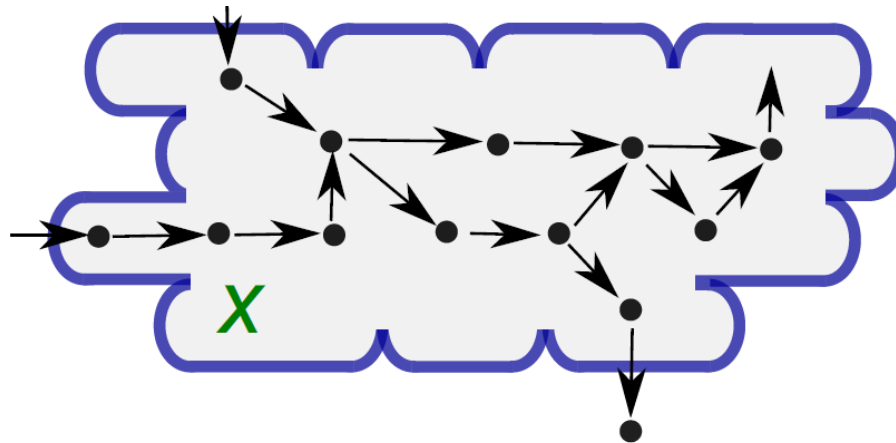
- Part 1. Systems biology and kinetic modeling
  - Introduction
  - Kinetic modeling of cellular reaction networks
- Part 2. Metabolic network modeling
  - Kinetic modeling of metabolism
  - **Metabolic control analysis (MCA)**
  - Flux balance analysis (FBA)
  - Practical on flux balance analysis (COBRA)
- Part 3. Gene regulatory network modeling
- Part 4. Models and data

# Biochemical reaction networks

- ODE model for growth of microbial populations:

$$\dot{x} = N \cdot v(x) - \mu \cdot x,$$

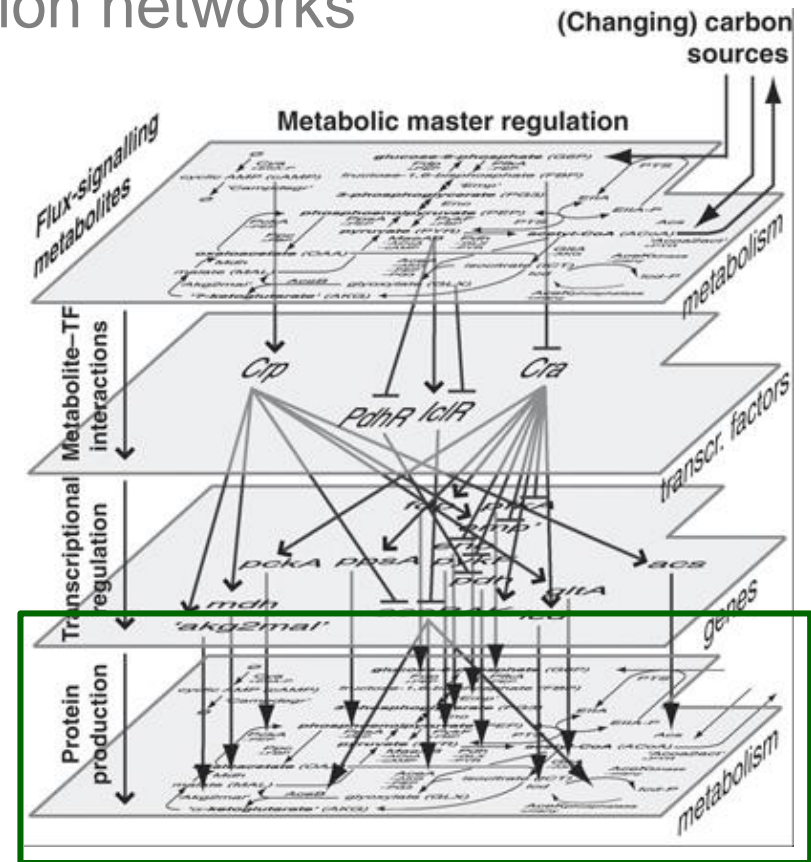
$$\mu = \delta \cdot \sum_i \alpha_i \cdot N_i \cdot v(x).$$



- Reaction rates depend on concentrations  $x$  of substrates, products, effectors

# Metabolic networks

- Focus on **subsystems** that can be studied in isolation due to **modular structure** of reaction networks
  - Time-scale hierarchies
  - Connectivity structure
- **Metabolic networks**
  - Metabolites and enzymatic reactions
  - Short turn-over times of metabolite pools in comparison with enzyme pools



Kotte et al. (2010), *Mol. Syst. Biol.*, 6: 355

# Metabolic networks

- Models describing dynamics of metabolism
  - Effect of growth dilution can often be ignored
  - Variables are metabolites and rates of enzyme-catalyzed reactions
  - Enzyme concentrations constant on time-scale of metabolic dynamics

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

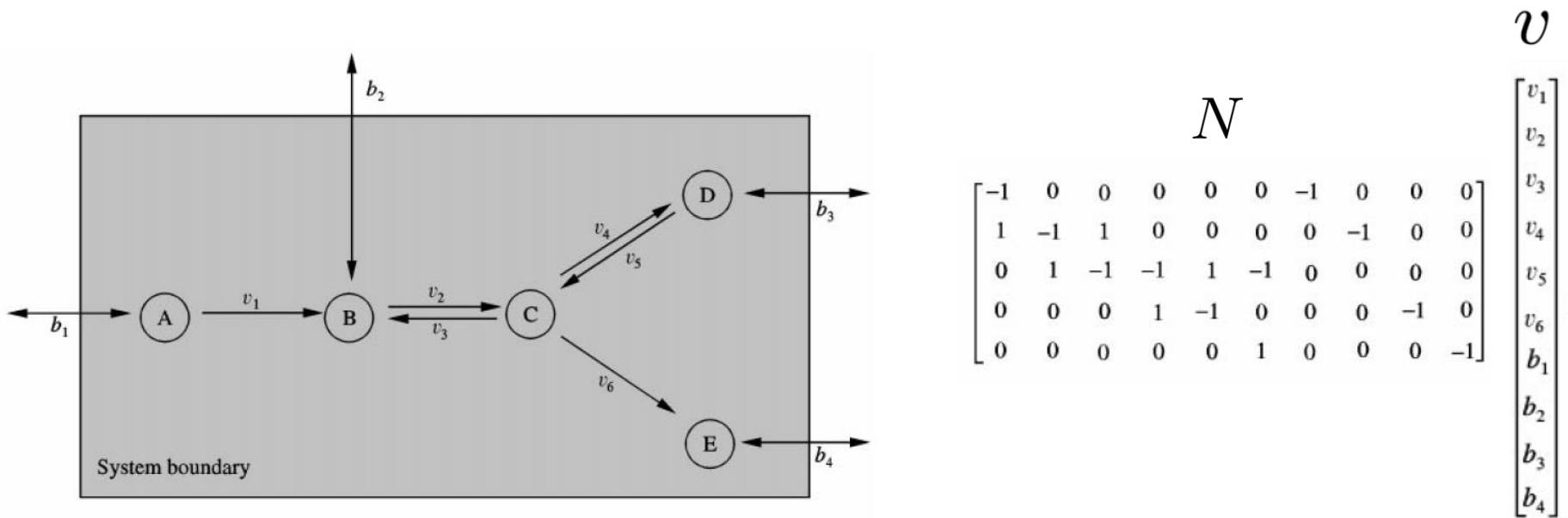
- Explicit introduction of dependency of model dynamics on parameters  $p$  :
  - Enzyme concentrations
  - Half-saturation and catalytic constants
  - Inhibition/activation constants

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

# Stoichiometry matrix

- Stoichiometry matrix  $N$  describes structure of reaction network

Internal reactions and exchange reactions, reversible and irreversible



Schilling *et al.* (2000), *J. Theor. Biol.*, 203(3):229-48

# Stoichiometry matrix

- Stoichiometry matrix may not be full rank
  - Dependencies between rows (variables) due to conservation relations
  - Example:  $[ATP] + [ADP] + [AMP]$  constant
- Reduction of stoichiometry matrix by means of **link matrix**  $L$  :

$$N = L N^0, \quad x = L x^0$$

- Variables in resulting metabolic system are independent

$$\dot{x}^0 = N^0 v(x^0, p)$$

- In what follows, we assume that  $N$  is full rank



# Metabolic networks at steady state

- For many problems of interest, the metabolic system can be considered at **steady state**

$$N v(x, p) = 0$$

- Metabolism relaxes on short time-scale (seconds-minutes) after changes in environment
- Difficult to measure dynamics of metabolic adaptation
- Metabolic rates at steady state: **fluxes**  $v(x^*, p) = v^*(p)$
- Trivial steady state with zero fluxes corresponds to **thermodynamic equilibrium**
- Steady state with non-zero fluxes requires that metabolic system is **open system**  
Non-zero exchange fluxes

# Stability of steady state

- Metabolism concerns almost exclusively **sustainable** processing of chemical inputs into outputs  
Biomass, energy, waste, ...

- Therefore, one expects steady states to be **stable**

Grimbs *et al.* (2007), *Mol. Syst. Biol.*, 3:146

- **Stability criterion** given by sign of (real part of) eigenvalues of **Jacobian matrix**

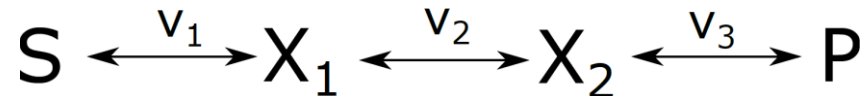
$$J(x^*) = N \left. \frac{\partial v}{\partial x} \right|_{x^*, v^*}$$

System is stable, if real part of **every** eigenvalue is negative

Kaplan and Glass (1995), *Understanding Nonlinear Dynamics*, New York

# Example of simple metabolic pathway

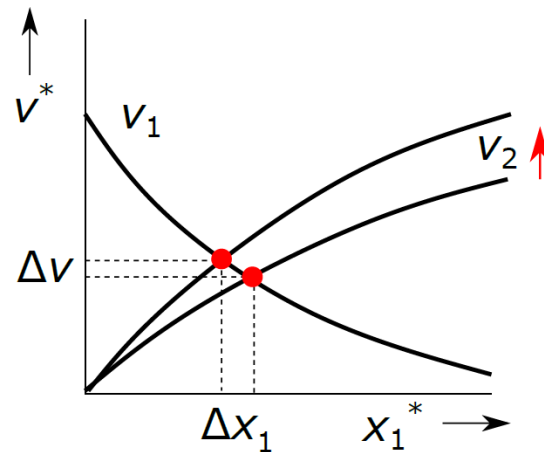
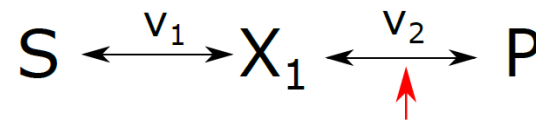
- Pathway of reactions converting substrate to product
  - S and P are supplied/removed (constant concentrations)
  - Reactions are reversible (Michaelis-Menten kinetics)



- **Exercise:** What is the stoichiometry matrix for this system?
- **Exercise:** How do the fluxes relate at steady state?
- **Exercise:** Write out the Jacobian matrix for this system
- **Exercise:** Determine the stability of the system. Hint: use the signs of the partial derivatives and the relation between eigenvalues and trace/determinant

# Metabolic control analysis

- Steady state of system is **sensitive** to (local) changes in enzyme concentrations or kinetic parameters



# Metabolic control analysis

- Steady state of system is **sensitive** to (local) changes in enzyme concentrations or kinetic parameters
- **Metabolic control analysis (MCA)** aims at studying this sensitivity in a systematic and rigorous manner
- MCA applies to arbitrarily complex networks
- Central questions in MCA:
  - How does the system steady state respond to changes in enzyme concentrations or kinetic parameters?
  - How does the system response depend on the network structure?
  - How constrained are sensitivities? Do they show dependencies?

Sauro (2009), Chapter 13 in Jason McDermott et al. (eds.), *Computational Systems Biology*, Humana Press, 269-309

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

Fell (1997), *Understanding the Control of Metabolism*, Portland Press

# Elasticity coefficients

- **Elasticity coefficients** express how the rate of a reaction changes due to a change in the reaction properties
  - Change in substrate, product, enzyme, effector concentrations
  - Change in kinetic parameter

$$\epsilon_{x_i}^{v_j} = \frac{\partial v_j}{\partial x_i} \cdot \frac{x_i}{v_j} = \frac{\partial \ln v_j}{\partial \ln x_i} = \frac{\%v_j}{\%x_i}$$

$$\epsilon_{p_i}^{v_j} = \frac{\partial v_j}{\partial p_i} \cdot \frac{p_i}{v_j} = \frac{\partial \ln v_j}{\partial \ln p_i} = \frac{\%v_j}{\%p_i}$$

- Elasticities are **local** properties of metabolic system
- Elasticities may vary with system state for complex rate laws
- **Exercise: write elasticities with respect to change in enzyme concentration for irreversible Michaelis-Menten rate law**

# Response coefficients

- **Response coefficients** express how steady state of the system changes due to a change in reaction properties
  - **Flux response coefficients**

$$R_{p_i}^{v_j^*} = \left. \frac{\partial v_j}{\partial p_i} \cdot \frac{p_i}{v_j} \right|_{x^*, v^*} = \left. \frac{\partial \ln v_j}{\partial \ln p_i} \right|_{x^*, v^*} = \frac{\%v_j^*}{\%p_i}$$

- **Concentration response coefficients**

$$R_{p_i}^{x_j^*} = \left. \frac{\partial x_j}{\partial p_i} \cdot \frac{p_i}{x_j} \right|_{x^*, v^*} = \left. \frac{\partial \ln x_j}{\partial \ln p_i} \right|_{x^*, v^*} = \frac{\%x_j^*}{\%p_i}$$

- Response coefficients are **global** properties of metabolic system
- Response coefficients generally vary with system state
- **Exercise: give examples of response coefficients for simple pathway**

# Response coefficients

- **Response coefficients** express how steady state of the system changes due to a change in reaction properties

- **Flux response coefficients**

$$R_{p_i}^{v_j^*} = \left. \frac{\partial v_j}{\partial p_i} \cdot \frac{p_i}{v_j} \right|_{x^*, v^*} = \left. \frac{\partial \ln v_j}{\partial \ln p_i} \right|_{x^*, v^*} = \frac{\%v_j^*}{\%p_i}$$

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- Response coefficients are **global** properties of metabolic system
- Response coefficients generally vary with system state
- How can response coefficients be computed? How do they relate to elasticity coefficients?



# Computation of response coefficients

- Differentiation of steady-state equation w.r.t.  $p$  :

$$N \frac{\partial v}{\partial x} \frac{\partial x}{\partial p} \Big|_{x^*, v^*} + N \frac{\partial v}{\partial p} \Big|_{x^*, v^*} = 0$$

# Computation of response coefficients

- Differentiation of steady-state equation w.r.t.  $p$  :

$$N \frac{\partial v}{\partial x} \frac{\partial x}{\partial p} \Big|_{x^*, v^*} + N \frac{\partial v}{\partial p} \Big|_{x^*, v^*} = 0$$

$$\frac{\partial x}{\partial p} \Big|_{x^*, v^*} = - \left( N \cdot \frac{\partial v}{\partial x} \right)^{-1} \cdot N \cdot \frac{\partial v}{\partial p} \Big|_{x^*, v^*}$$

$$\begin{aligned} \text{dg}(x)^{-1} \cdot \frac{\partial x}{\partial p} \cdot \text{dg}(p) \Big|_{x^*, v^*} &= - \text{dg}(x)^{-1} \cdot \left( N \cdot \frac{\partial v}{\partial x} \right)^{-1} \cdot N \cdot \frac{\partial v}{\partial p} \cdot \text{dg}(p) \Big|_{x^*, v^*} \\ &= - \text{dg}(x)^{-1} \cdot \left( N \cdot \text{dg}(v) \cdot \text{dg}(v)^{-1} \cdot \frac{\partial v}{\partial x} \cdot \text{dg}(x) \cdot \text{dg}(x)^{-1} \right)^{-1} \cdot \end{aligned}$$

$$N \cdot \text{dg}(v) \cdot \text{dg}(v)^{-1} \cdot \frac{\partial v}{\partial p} \cdot \text{dg}(p) \Big|_{x^*, v^*}$$

$$R_p^{x^*} = - \left( N \cdot \text{dg}(v) \cdot \epsilon_x^v \right)^{-1} \cdot N \cdot \text{dg}(v) \cdot \epsilon_p^v \Big|_{x^*, v^*}$$

# Control coefficients

- Separation of reaction-specific and systemic contribution to response coefficient

$$\begin{aligned} R_p^{x^*} &= - (N \cdot \text{dg}(v) \cdot \epsilon_x^v)^{-1} \cdot N \cdot \text{dg}(v) \cdot \epsilon_p^v \Big|_{x^*, v^*} \\ &= C_v^{x^*} \cdot \epsilon_p^v \Big|_{x^*, v^*} \end{aligned}$$

with **concentration control coefficients**

$$C_v^{x^*} = - (N \cdot \text{dg}(v) \cdot \epsilon_x^v)^{-1} \cdot N \cdot \text{dg}(v) \Big|_{x^*, v^*}$$

- Concentration control coefficients describe effect of change in rate (by whatever means) on steady-state concentration

$$R_{p_i}^{x_j^*} = C_{v_k}^{x_j^*} \cdot \epsilon_{p_i}^{v_k} \Big|_{x^*, v^*} \qquad C_{v_k}^{x_j^*} = \frac{\%x_j^*}{\%v_k}$$

# Control coefficients

- Similar analysis for flux response coefficients leads to

$$R_p^{v^*} = C_v^{v^*} \cdot \epsilon_p^v \Big|_{x^*, v^*}$$

with **flux control coefficients**

$$C_v^{v^*} = I + \epsilon_x^v \cdot C_v^{x^*} \Big|_{x^*, v^*}$$

- Flux control coefficients describe effect of change in rate (by whatever means) on fluxes at steady state

$$R_{p_i}^{v_j^*} = C_{v_k}^{v_j^*} \cdot \epsilon_{p_i}^{v_k} \Big|_{x^*, v^*} \quad C_{v_k}^{v_j^*} = \frac{\%v_j^*}{\%v_k}$$

- Above analysis provides conceptual framework, but is not very practical for computational purposes

# Summation and connectivity theorems

- Above analysis can be further developed into MCA **summation theorems**:

$$C_v^{x^*} \cdot 1 = 0, \quad \sum_k C_{v_k}^{x_j^*} = 0$$

$$C_v^{v^*} \cdot 1 = 1, \quad \sum_k C_{v_k}^{v_j^*} = 1$$

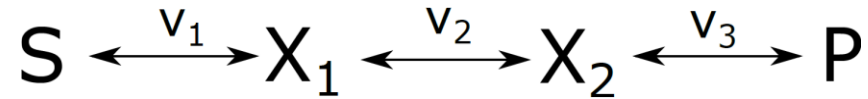
- Flux control is **distributed** over the system
- Idem for MCA **connectivity theorems**:

$$C_v^{x^*} \cdot \epsilon_x^v = -I, \quad \sum_k C_{v_k}^{x_j^*} \cdot \epsilon_{x_i}^{v_k} = -\delta_{ji} \quad \delta_{ji} = \begin{cases} 1, & i = j, \\ 0, & i \neq j. \end{cases}$$

$$C_v^{v^*} \cdot \epsilon_x^v = 0, \quad \sum_k C_{v_k}^{v_j^*} \cdot \epsilon_{x_i}^{v_k} = 0$$

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

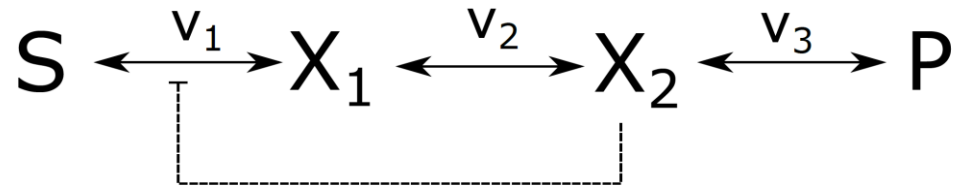
# Example of simple metabolic pathway



- Exercise: write down the flux summation and connectivity theorems for the model of this pathway
- Exercise: find expressions for flux control coefficients in terms of elasticities. What can be learned from these expressions?

# Example of simple pathway with feedback

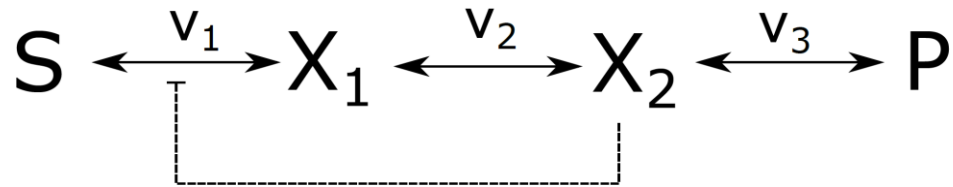
- Pathway with negative feedback on level of enzyme activity



- Question: in case of strong feedback, if we would like to increase production of P, which reaction should we target?

# Example of simple pathway with feedback

- Pathway with negative feedback on level of enzyme activity



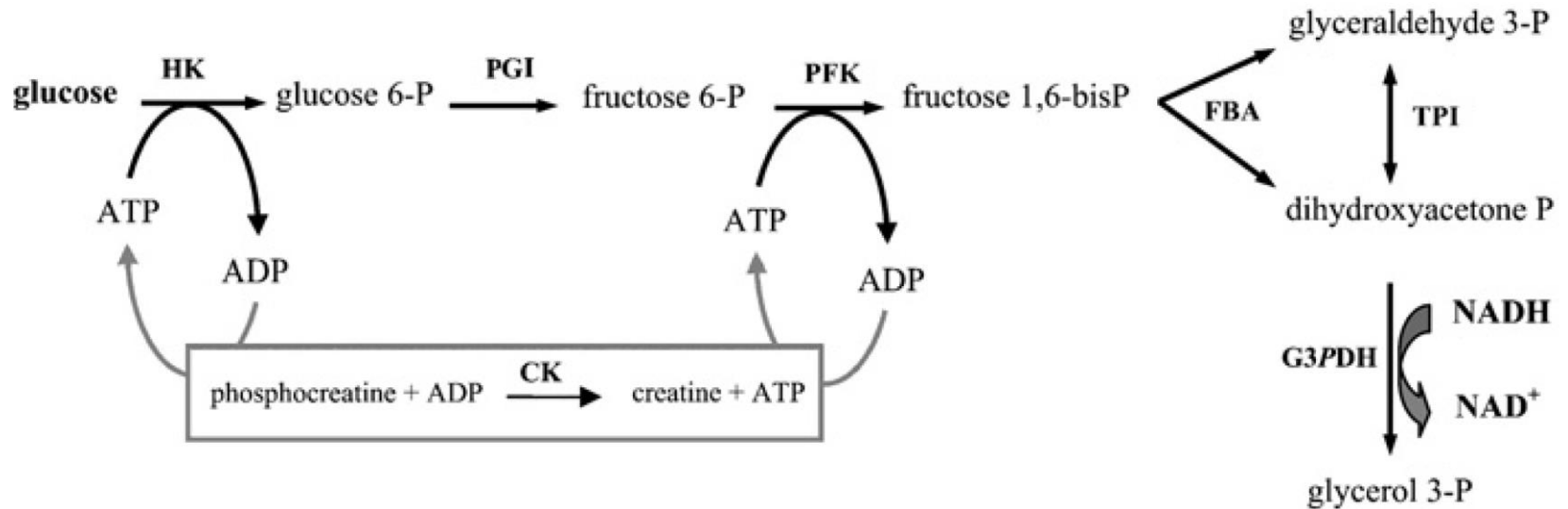
- Question: in case of strong feedback, if we would like to increase production of P, which reaction should we target?
- Answer: reaction 3 (counter-intuitive)

Sauro (2009), Chapter 13 in Jason McDermott et al. (eds.), *Computational Systems Biology*, Humana Press, 269-309



# *In-vitro* reconstruction of glycolysis

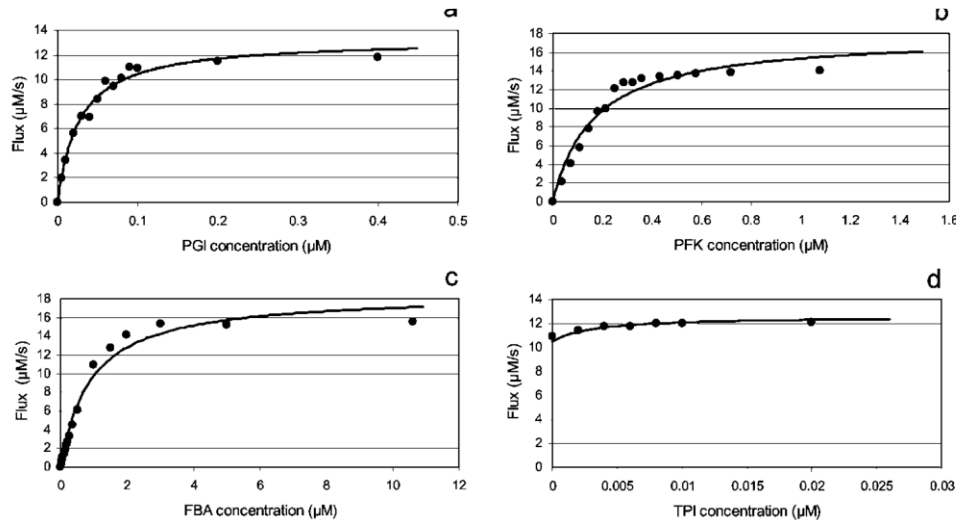
- Upper part of glycolysis pathway has been reconstructed *in vitro* and quantitatively modeled



Fiévet *et al.* (2006), *Biochem. J.*, 396:317–26

# *In-vitro* reconstruction of glycolysis

- Upper part of glycolysis pathway has been reconstructed *in vitro* and quantitatively modeled
- How does flux respond to change in enzyme concentration?

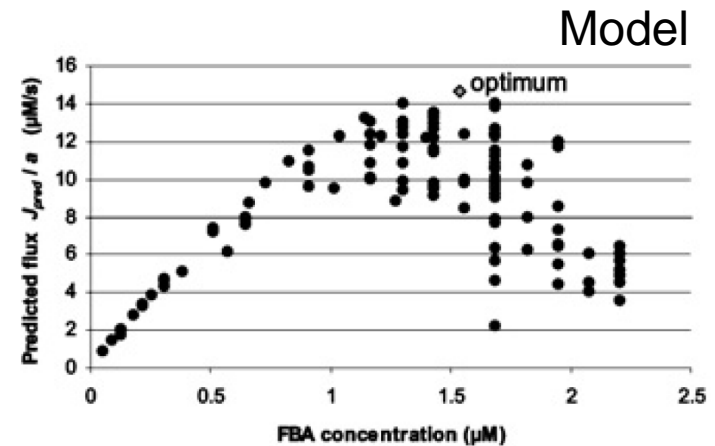
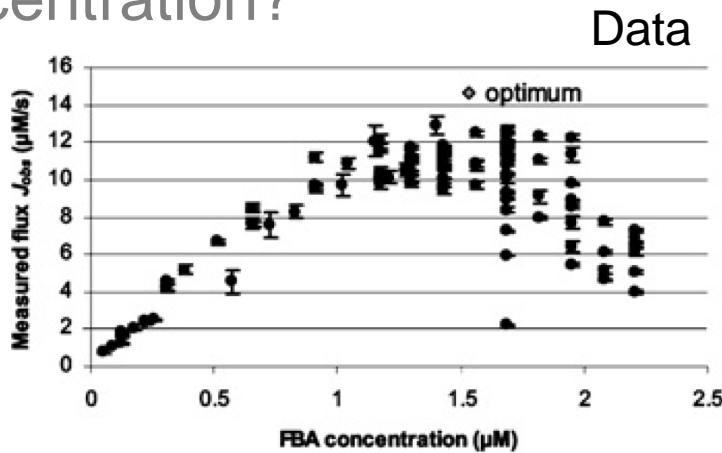


- Positive flux control coefficients for **all** enzymes, as expected from theoretical analysis

Fiévet *et al.* (2006), *Biochem. J.*, 396:317–26

# *In-vitro* reconstruction of glycolysis

- Upper part of glycolysis pathway has been reconstructed *in vitro* and quantitatively modeled
- How does flux respond to change in enzyme concentration?
- How can flux be optimized for given total enzyme concentration?



- Maximum attained for intermediate enzyme concentrations

Fiévet *et al.* (2006), *Biochem. J.*, 396:317–26

# *In-vivo* control of glycolytic flux

- How is flux through glycolysis controlled in bacteria? What is role of ATP demand?
  - ATP produced by glycolysis and consumed by other cellular processes
- Approach: augment intracellular ATP consumption
  - Inducible (uncoupled) ATPase activity

# In-vivo control of glycolytic flux

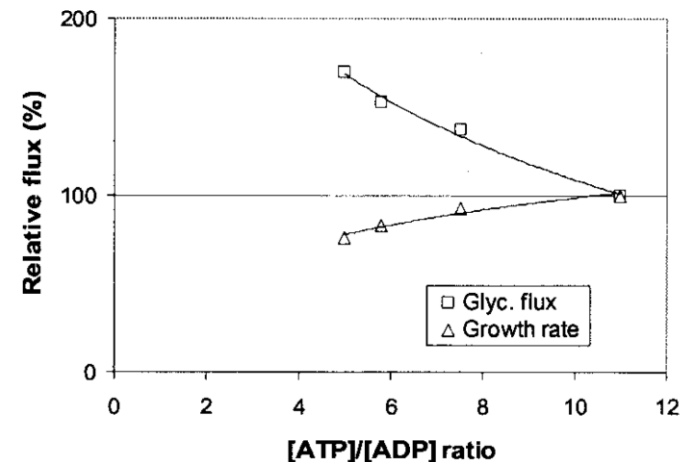
- How is flux through glycolysis controlled in bacteria? What is role of ATP demand?

ATP produced by glycolysis and consumed by other cellular processes

- Approach: augment intracellular ATP consumption
- Metabolic control analysis of simplified system



- ATPase expression decreases growth rate and increases glycolytic fluxes



# In-vivo control of glycolytic flux

- How is flux through glycolysis controlled in bacteria? What is role of ATP demand?

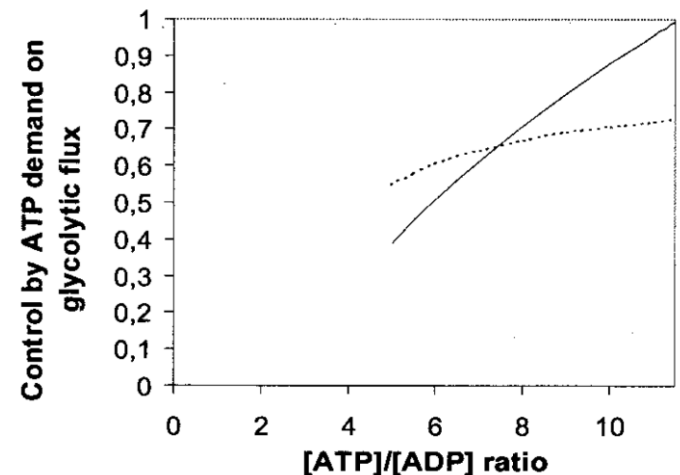
ATP produced by glycolysis and consumed by other cellular processes

- Approach: augment intracellular ATP consumption
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$$C_{e_2}^J = \frac{-\epsilon_p^{e_1}}{\epsilon_p^{e_2} - \epsilon_p^{e_1}}$$

- ATPase expression decreases growth rate and increases glycolytic fluxes
- In wild-type cells at least 75% of glycolytic control exerted by ATP demand



# Conclusions

- Metabolic systems often analyzed at (stable) steady state
- Metabolic flux analysis (MCA) quantifies sensitivity of fluxes and concentrations to changes in parameters and inputs
- Well-established and powerful mathematical framework
- Dedicated computer tools supporting the analysis
- Many applications demonstrating its practical use in systems biology and synthetic biology/metabolic engineering

**Merci !**



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