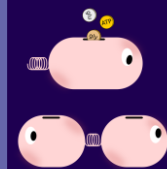


# Economic Principles in Cell Biology

Vienna, July 23-26, 2025



## Optimal Cell Behavior in Time

Dafni Giannari, Hidde de Jong, Diego A. Oyarzún, Steffen Waldherr, Agustín G. Yabo

# Outline of presentation

- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in a metabolic pathway
- Teaser: time-varying resource allocation and cellular growth
- Conclusions and perspectives



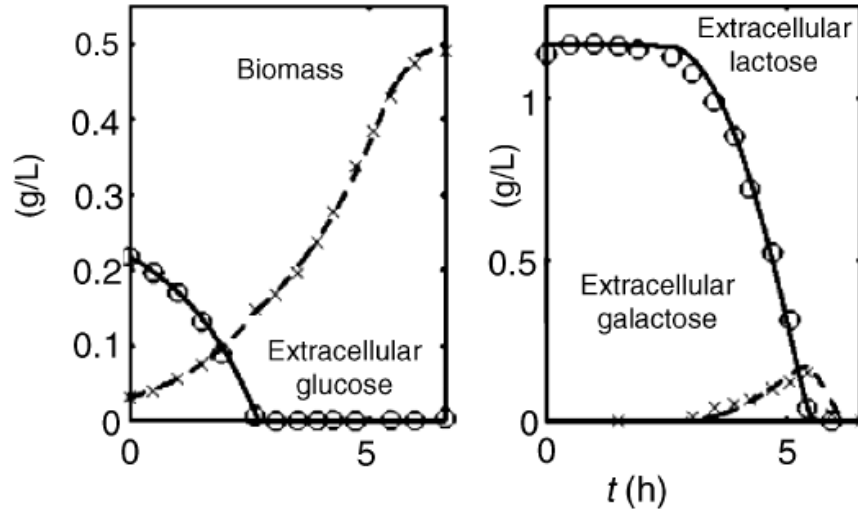
# Outline of presentation

- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in a metabolic pathway
- Teaser: time-varying resource allocation and cellular growth
- Conclusions and perspectives
  
- Many links with previous lectures
  - Optimization of metabolic fluxes
  - The enzyme cost of metabolic fluxes
  - Principles of cell growth
  - Growth balance analysis
  - ...



# Bacterial adaptation

- Bacterial cells need to adapt to dynamically changing environments
  - Example: diauxic growth of *E. coli*

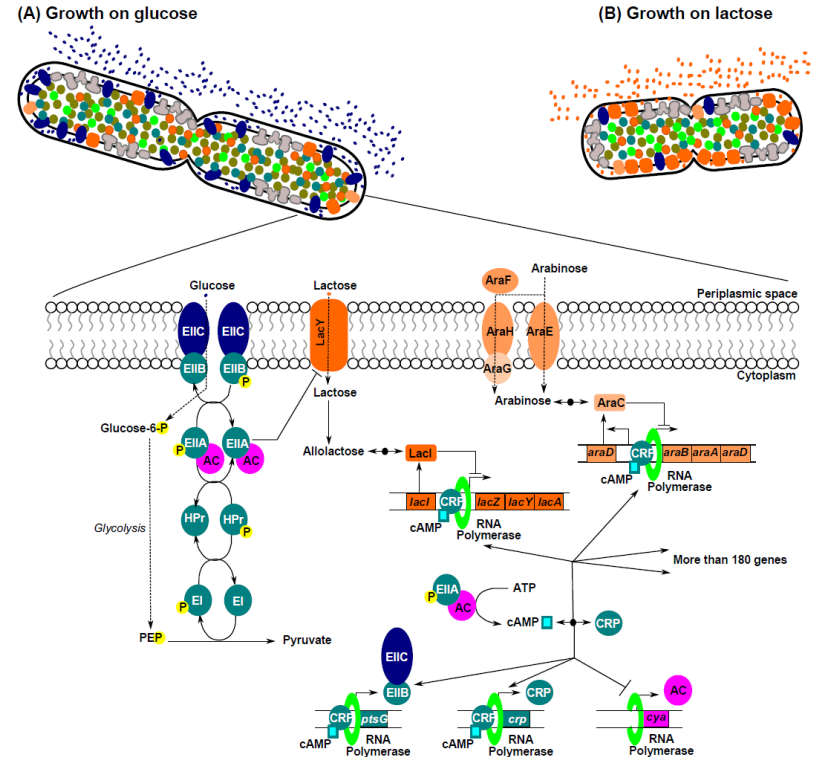


Bettenbrock *et al.* (2006), *J. Biol. Chem.*, 281:2578-84



## Regulation of bacterial adaptation

- Bacterial cells have developed complex sensory and regulatory systems to realize adaptation
- Difficult to obtain quantitative, mechanistic models for these systems
  - Incompletely known mechanisms, unknown parameter values, ...



Kremling *et al.* (2015), *Trends Microbiol.*, 23:99-109



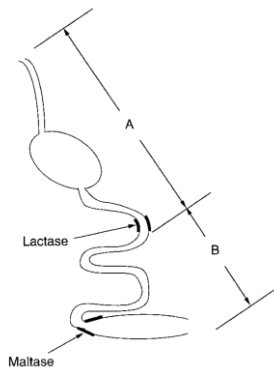
# Dynamic optimization approach

- Alternative: ignore regulatory mechanisms and assume that **cells perform (dynamic) optimization**
  - Limiting resources (proteins, fluxes, ...)
  - ... allocated to cellular processes (reactions)
  - ... so as to maximize some objective (biomass synthesis, adaptation time, ...)
  - ... in a changing environment (nutrients, temperature, light, ...)
  - ... over a time-interval (response time, day/night cycle, ...)
- Dynamic vs static optimization
  - Time-varying allocation of resources
  - Non-trivial features: resource buffers, anticipation of future changes, ...
- **Assumption:** bacteria have evolved to (dynamically) optimize their functioning in competitive, changing environments

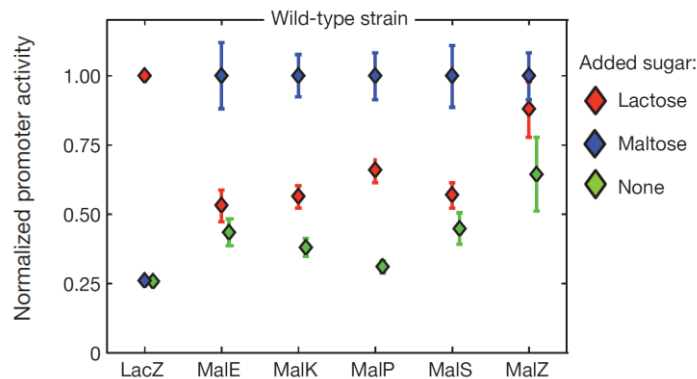


# Evidence for dynamic optimization

- Microorganisms are capable of anticipating changes in their environment
  - Along digestive tract, exposure of *E. coli* to lactose precedes exposure to maltose
  - Expression of maltose genes when lactose is present



Savageau (1998), *Genetics*, 149:1677-91



Mitchell *et al.* (2008), *Nature*, 460:220-4

- But: assumption remains working hypothesis!



# Dynamic optimization and optimal control

- Mathematical formulation of dynamic optimization yields **optimal control problem**:

$$\max_{u \in U} J(x, u, t_0, t_e),$$

such that

$$\frac{dx}{dt} = f(x(t), u(t)), \quad x(t_0) = x_0,$$

$$0 \geq c_1(x(t), u(t)),$$

$$0 \geq c_2(x(t_0), x(t_e)).$$

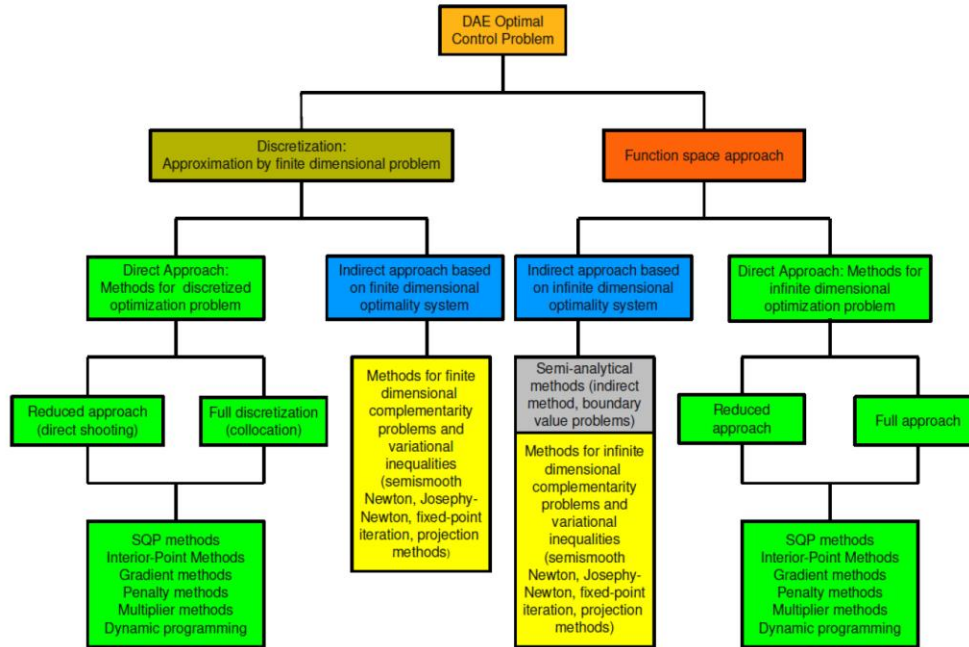
- Specification of optimal control problem:
  - Dynamical system with state  $x$  and dynamics  $f$
  - Time-varying control  $u$ , over time-interval  $[t_0, t_e]$
  - Objective function  $J$ , and path constraints  $c_1$  and time-point constraints  $c_2$





# Dynamic optimization and optimal control

- Rich variety of mathematical techniques exist to **(numerically) solve optimal control problems**

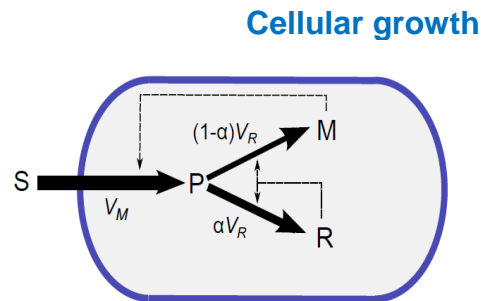
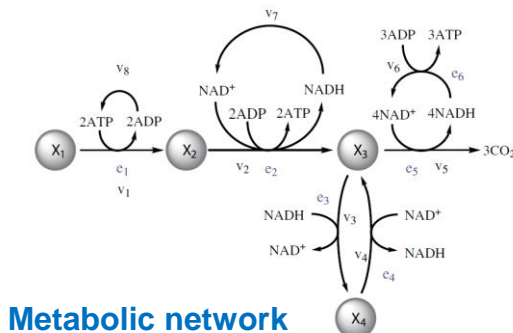
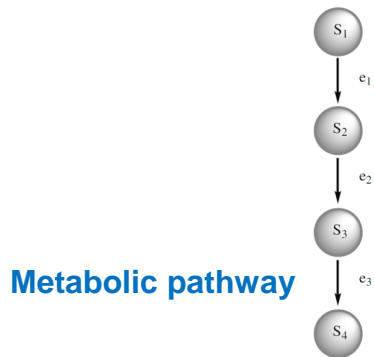


Gerdt (2013), OMPC Summer school Bayreuth, [link](#)



# Examples of dynamic optimization

- Dynamic optimization in two examples, increasingly larger scope:
  - Time-varying expression of enzymes in **metabolic pathways and networks**
  - Teaser: time-varying resource allocation and **cellular growth**



# Outline of presentation

- Dynamic optimization: introduction and motivation
- **Time-varying expression of enzymes in a metabolic pathway**
- Teaser: time-varying resource allocation and cellular growth
- Conclusions and perspectives



# Time-varying expression of enzymes

- **Metabolic pathway:** chain of enzymatic reactions converting substrate into product
- Allocation of enzyme capacity to reactions is resource allocation problem

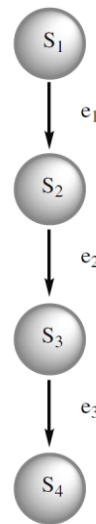
Enzymes are limiting (costly) resource

$S_1$ : substrate

$S_{2,3}$ : intermediate metabolites

$S_4$ : product

$e_1, \dots, e_3$ : enzymes



Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13

Bartl *et al.* (2010), *BioSystems*, 101:67–77

De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1



# Time-varying expression of enzymes

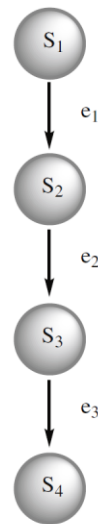
- Model describing dynamics of metabolic pathway

Mass-action kinetics

$$\frac{ds}{dt} = N \cdot v(s(t), e(t)), \quad s(t_0) = [s_{10}, 0, 0, 0]',$$

$$N = \begin{bmatrix} 0 & 0 & 0 \\ 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix},$$

$$v_1(s_i, e_i) = k_i e_i s_i, \quad i = 1, \dots, 3,$$



# Time-varying expression of enzymes

- **Assumption:** pathway has evolved so as to minimize transition time, that is, time to make a (certain amount of) product.
- **Dynamic optimization problem:** given an objective function

$$J(e) = t_e,$$

where  $e(t)$  is a time-dependent function, find

$$e_{opt} = \arg \min_e J(e)$$

under constraints

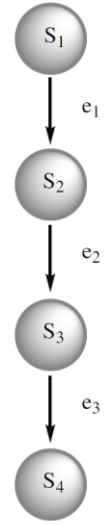
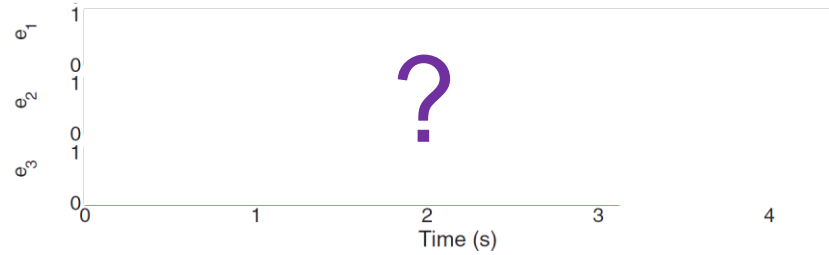
$$e_T \geq e_1 + e_2 + e_3,$$

$$s_4(t_e) = 0.9 \cdot s_{10}.$$



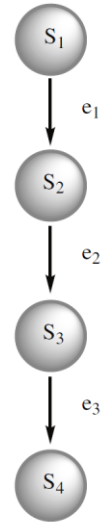
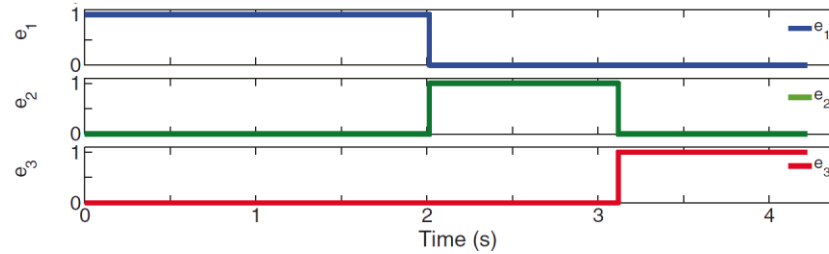
# Time-varying expression of enzymes

- What is the optimal enzyme expression pattern?

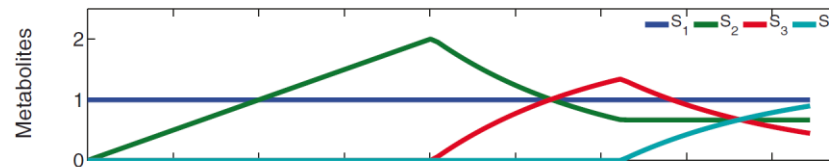


# Time-varying expression of enzymes

- What is the optimal enzyme expression pattern?



- Temporal ordering of expression of enzymes corresponding to ordering of reactions in pathway



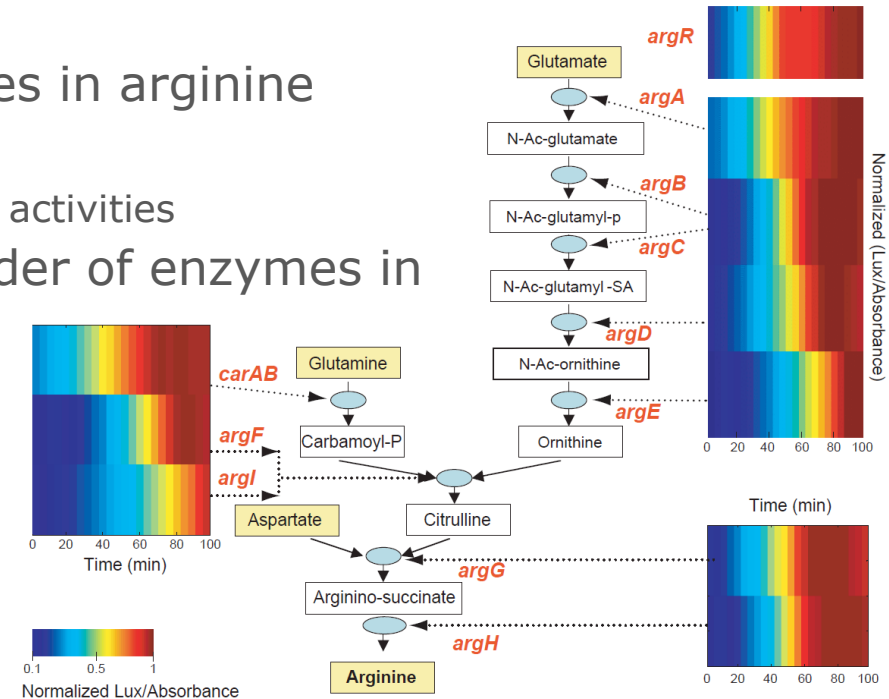
De Hijas-Liste *et al.* (2014),  
*BMC Syst. Biol.*, 8:1





# Time-varying expression of enzymes

- Experimental evidence for temporal expression patterns of enzymes?
- Just-in-time expression of enzymes in arginine metabolism
  - Measurement of (normalized) promoter activities
- Temporal order corresponds to order of enzymes in unbranched pathways



Zaslaver *et al.* (2004), *Nat. Genet.*, 36:486-91

# Time-varying expression of enzymes

- Generalization from pathways to networks
  - Diauxic growth on glucose and ethanol in yeast

$X_1$ : glucose

$X_{2,3}$ : intermediate metabolites

$X_4$ : ethanol

$v_1$ : upper glycolysis

$v_2$ : lower glycolysis

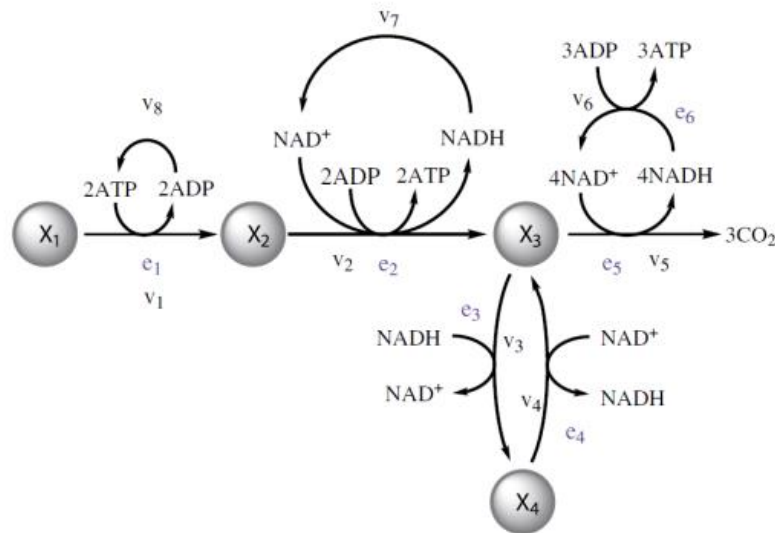
$v_3$ : ethanol production

$v_4$ : ethanol consumption

$v_5$ : TCA cycle

$v_6$ : respiratory chain

$v_{7-8}$ : cofactor recycling

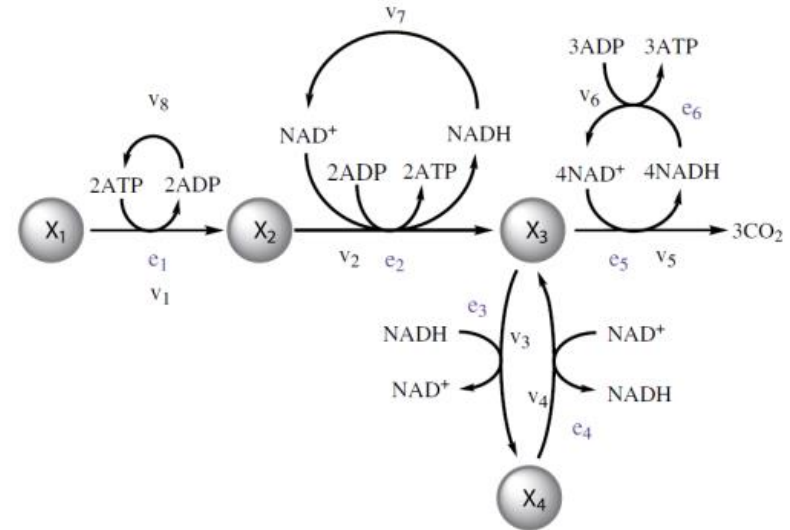


Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13  
De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1



# Time-varying expression of enzymes

- Generalization from pathways to networks
  - Diauxic growth on glucose and ethanol in yeast
  - Mass-action model, constraint on total enzyme concentration
  - Maximization of survival time (quiescent state)

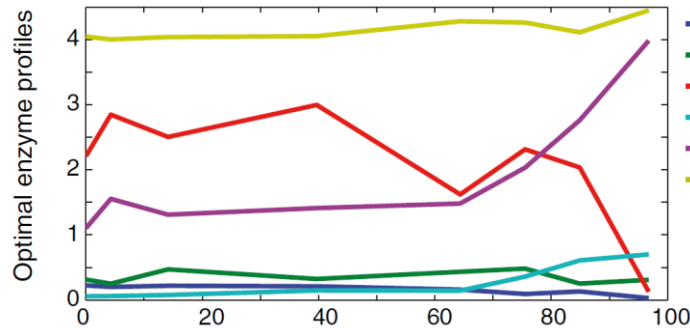


Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13  
De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1

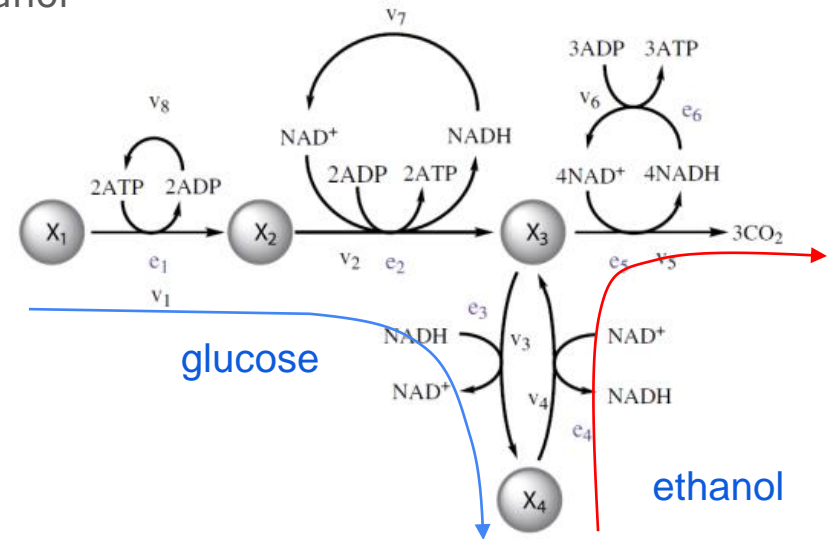


# Time-varying expression of enzymes

- Generalization from pathways to networks
  - Diauxic growth on glucose and ethanol in yeast
  - Mass-action model, constraint on total enzyme concentration
  - Maximization of survival time (quiescent state)
  - Predicted diauxic growth: glucose  $\rightarrow$  ethanol



Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13  
De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1



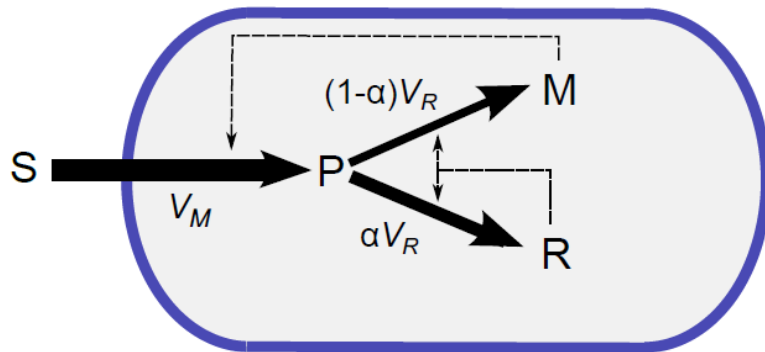
# Outline of presentation

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- **Teaser: time-varying resource allocation and cellular growth**
- Conclusions and perspectives



# Time-varying resource allocation and growth

- Bacterial growth is fundamentally a **resource allocation problem**  
How does the cell distribute available resources over cellular functions?
- Resource allocation can be studied using **self-replicator models** of bacterial growth



**S:** substrate (nutrient)  
**P:** precursor  
**M:** metabolic machinery  
**R:** gene expression machinery  
**Biomass:**  $M + R$

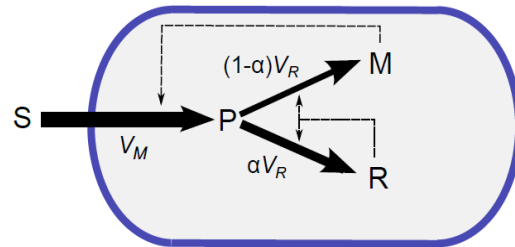
$\alpha$ : resource allocation

Molenaar *et al.* (2009), *Mol. Syst. Biol.*, 5:323  
Scott *et al.* (2014), *Mol. Syst. Biol.*, 10:747  
Giordano *et al.* (2016), *PLoS Comput. Biol.*, 12:e1004802



# Time-varying resource allocation and growth

- Assumption: bacteria have evolved to maximize growth, *i.e.*, total amount of biomass
- Example questions of interest:
  - Which resource allocation strategy (choice of  $\alpha$ ) would optimize biomass in stable environment?
  - Which strategy would optimize biomass after a shift from poor to rich substrate?
  - Which strategy would optimize biomass in an environment with poor substrate with periodic pulses of rich substrate?
- See “**Optimal Cell Behavior in Time**” chapter for more information!



# Outline of presentation

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# Conclusions

- **Pros** of dynamical optimization approach
  - Avoids modeling of (unknown) regulatory mechanisms
  - Allows for specification of constraints on solutions
  - Exploits availability of numerical tools for solving optimal control problems
  - Applies both to explaining observed behavior and designing desired behavior



# Conclusions

- **Pros** of dynamical optimization approach
  - Avoids modeling of (unknown) regulatory mechanisms
  - Allows for specification of constraints on solutions
  - Exploits availability of numerical tools for solving optimal control problems
  - Applies both to explaining observed behavior and designing desired behavior
- **Cons** of dynamical optimization approach
  - Faces problems with numerical solvers: robustness, multiple solutions, ...
  - Requires prior specification of plausible objective function(s): many possibilities...
  - Is based on (questionable) hypothesis that observed behavior has been optimized through natural selection



# Perspectives

- **Multi-objective optimality** (Pareto optimality): system simultaneously optimizes several objectives, leading to trade-offs
- **Inverse optimality**: exploits huge amounts of available data to infer rather than assume objective function(s)
- **Experimental validation** of model predictions



Please give us your feedback about this lecture!

