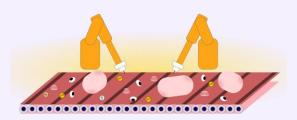


Optimal Cell Behavior in Time

Hidde de Jong, Markus Köbis





Outline of book chapter

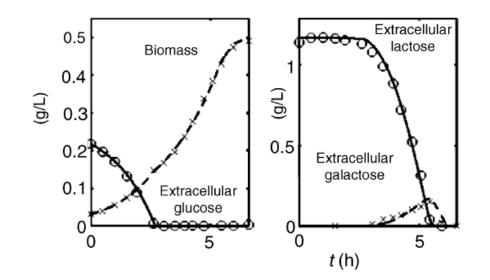
- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in a metabolic pathway
- Time-varying flux distribution in a metabolic network Dynamic flux balance analysis (dFBA)
- Time-varying flux distribution in a metabolic network with enzyme costs Dynamic enzyme-cost flux balance analysis (deFBA)
- Time-varying resource allocation and cellular growth
- Conclusions and perspectives
- Authors: Steffen Waldherr, Markus Köbis, Diego Oyarzun, Hidde de Jong, ...

Outline of presentation

- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in a metabolic pathway
- Time-varying flux distribution in a metabolic network Dynamic flux balance analysis (dFBA)
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- Conclusions and perspectives
- Many links with previous lectures

Bacterial adaptation

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



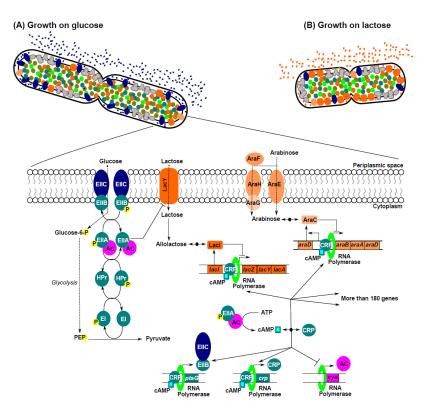
Bettenbrock et al. (2006), J. Biol. Chem., 281:2578-84



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, ...



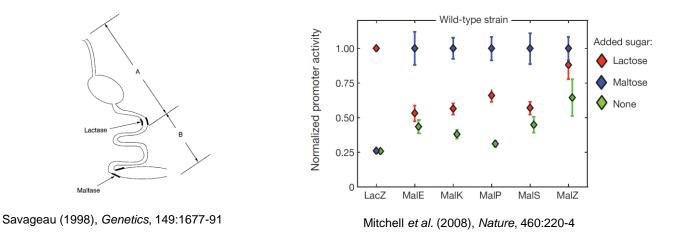
Kremling et al. (2015), Trends Microbiol., 23:99-109

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



But: assumption remains working hypothesis!

Dynamic optimization and optimal control

• Mathematical formulation of dynamic optimization yields **optimal control problem**: $\max_{u \in U} J(x, u, t_0, t_e),$

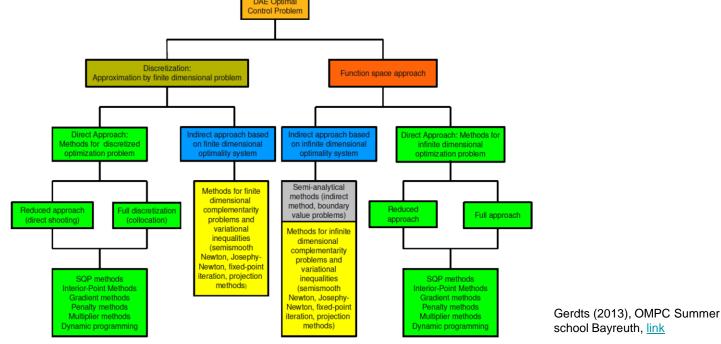
such that
$$\begin{aligned} \frac{dx}{dt} &= f(x(t), u(t)), \quad x(t_0) = x_0, \\ 0 &\geq c_1(x(t), u(t)), \\ 0 &\geq c_2(x(t_0), x(t_e)). \end{aligned}$$

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

Tsiantis and Banga (2020), BMC Bioinform, 21:472

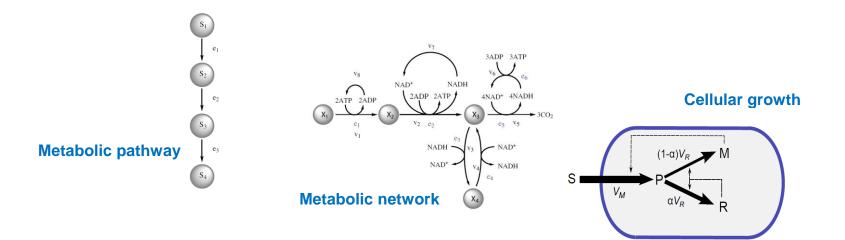
Dynamic optimization and optimal control

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



Examples of dynamic optimization

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



Outline of presentation

- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in metabolic pathways and networks
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- Conclusions and perspectives

- 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 $E_1, ..., E_3$: 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 S_1

 S_2

S₃

 S_4

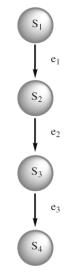
 e_1

 e_2

 e_3

Model describing dynamics of metabolic pathway
 Mass-action kinetics

$$\frac{ds}{dt} = N \cdot v(s(t), e(t)), \quad s(t_0) = [s_{10}, 0, 0, 0]',$$
$$N = \begin{bmatrix} 0 & 0 & 0 \\ 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix},$$
$$v_1(s_i, e_i) = k_i e_i s_i, \quad i = 1, \dots, 3,$$



- **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

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

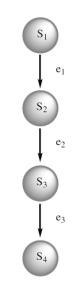
under constraints

$$e_T \ge e_1 + e_2 + e_3$$

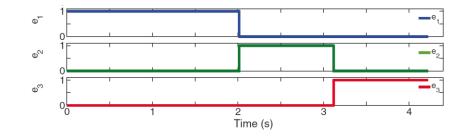
 $s_4(t_e) = 0.9 \cdot s_{10}.$

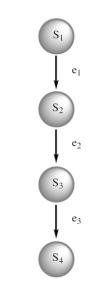
• What is the optimal enzyme expression pattern?



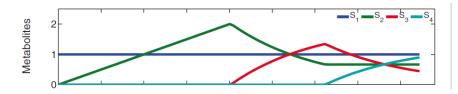


• What is the optimal enzyme expression pattern?



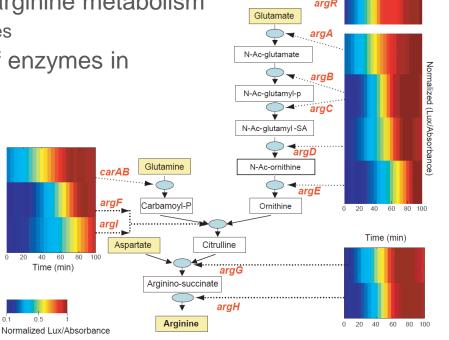


• Temporal ordering of expression of enzymes corresponding to ordering of reactions in pathway



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

- 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



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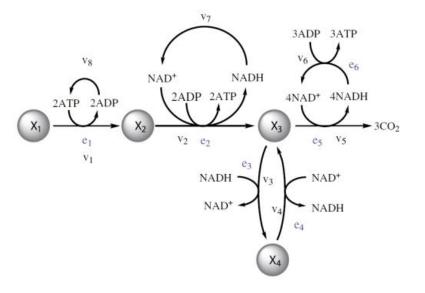
Zaslaver et al. (2004), Nat. Genet., 36:486-91

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

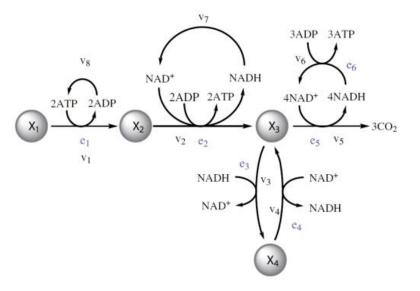
X₁: glucose X_{2,3}: intermediate metabolites X₄: ethanol

 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

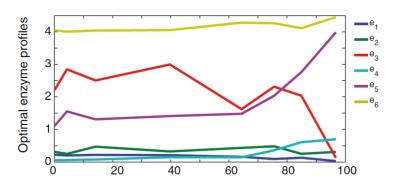


- Generalization from pathways to networks
 - Diauxic growth on glucose and ethanol in yeast
 - Mass-action model, constraint on total enzyme concentration
 - Maximization of survival time (quiescent state)

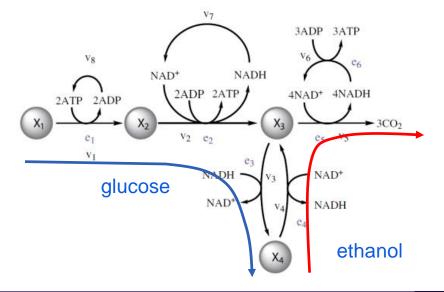


Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13 De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1

- 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)
 - \circ Predicted diauxic growth: glucose \rightarrow ethanol



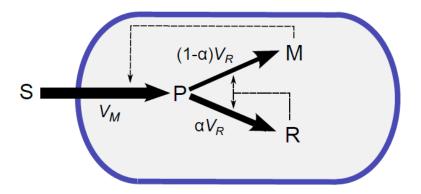
Klipp *et al.* (2002), *Eur. J. Biochem.*, 269:5406–13 De Hijas-Liste *et al.* (2014), *BMC Syst. Biol.*, 8:1



Outline of presentation

- Dynamic optimization: introduction and motivation
- Time-varying expression of enzymes in metabolic pathways and networks
- Time-varying resource allocation and cellular growth
- Conclusions and perspectives

- 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



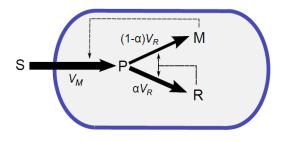
S: substrate (nutrient)
P: precursor
M: metabolic machinery
R: gene expression machinery
Biomass: M + R

 $\boldsymbol{\alpha}$: 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

 Resource allocation can be studied using self-replicator models of bacterial growth

$$\frac{dp}{dt} = v_m(s,m) - v_r(p,r) - \mu p$$
$$\frac{dr}{dt} = \alpha(t) v_r(p,r) - \mu r$$
$$1/\beta = m + r$$
$$\mu = \beta v_r(p,r)$$



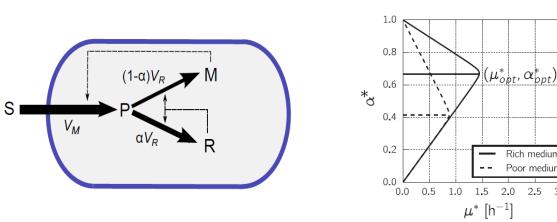
$$v_m(s,m) = k_m m \frac{s}{s+K_m} = k_m \left(1/\beta - r\right) \frac{s}{s+K_m}$$
$$v_r(p,r) = k_r r \frac{p}{p+K_r}$$

- **Assumption:** resource allocation has evolved to maximize growth
- Static optimization problem: given an objective function

$$J(\alpha^*) = \mu^* = \beta v_r(p^*, r^*)$$

find

$$\alpha_{opt}^* = \arg \max_{\alpha^* \in U} J(\alpha^*)$$



3.0

Rich medium Poor medium

2.0 2.5 В

- **Assumption**: resource allocation has evolved to maximize growth
- **Dynamic optimization problem**: given an objective function

$$J(\alpha) = \int_0^\tau \mu(t) dt = \int_0^\tau \beta v_R(p, r) dt,$$

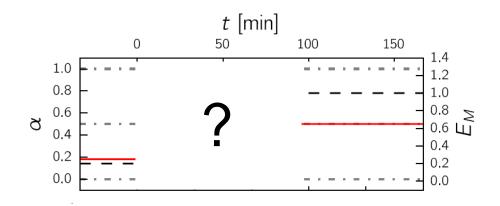
where $\alpha(t)$ is a time-dependent function, find

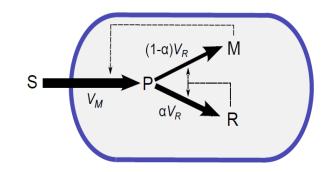
$$\alpha_{opt} = \arg \max_{\alpha \in \mathcal{U}} J(\alpha)$$

Compute analytical or numerical solution using optimal control methods/tools

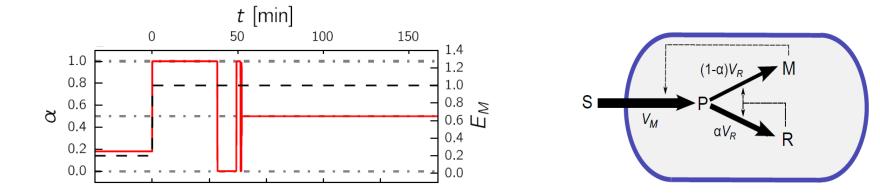
Giordano *et al.* (2016), *PLoS Comput. Biol.*, 12:e1004802 Yabo *et al.* (2022), *SIAM J. Appl. Dyn. Syst.*, 21:137-165

• What is the optimal resource allocation scheme for switch from poor to rich carbon source?

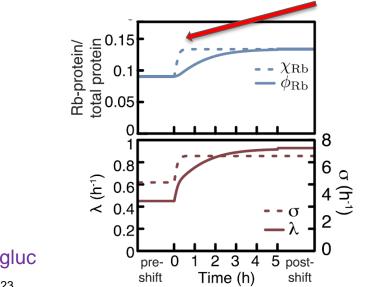




- Optimal resource allocation scheme is **bang-bang singular**
 - Sequence of switches between $\alpha = 1$ (maximal synthesis of gene expression machinery) and $\alpha = 0$ (maximal synthesis of metabolic machinery)
 - \circ α is then set to α^*_{opt} , value leading to maximal growth rate in new medium



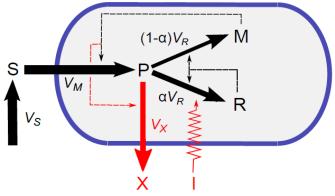
• Experimental evidence for oscillatory patterns in ribosome synthesis? No support from model based on population-level data



Upshift: succ \rightarrow succ + gluc

Erickson et al. (2017), Nature, 551:119-23

- For biotechnological applications, one would like to **change** the natural resource allocation strategies of the cell
- Which external control I would **optimize production of X** over a given time interval?
- Optimal control problem with human-defined instead of naturally-evolved objective
- Optimal solution: first growth, then production



Izard, Gomez Balderas *et al.* (2015), *Mol. Syst. Biol.*, 11:840 Yegorov *et al.* (2019), *J. Math. Biol.*, 78:985-1032

Outline of presentation

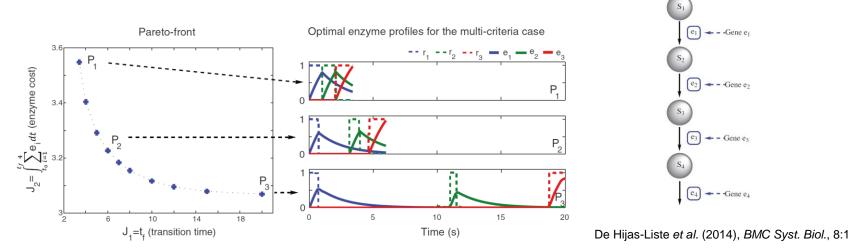
- Dynamic optimization: introduction and motivation
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Conclusions

- **Pros** of dynamical optimization approach
 - Avoids modeling of (unknown) regulatory mechanisms
 - Allows specifications 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



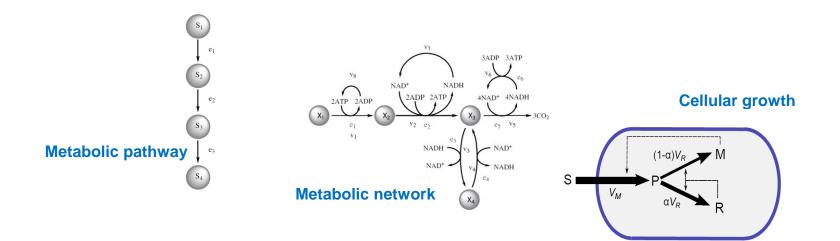
- Inverse optimality: exploits huge amounts of available data to infer rather than assume objective function(s)
 Zhao et al. (2016), Genome Biol., 1 Zhao et al. (2016), Genome Biol., 1
- Experimental validation of model predictions

Zhao *et al.* (2016), *Genome Biol.*, 17:109 Tsiantis *et al.* (2018), *Bioinformatics* 34:2433–40

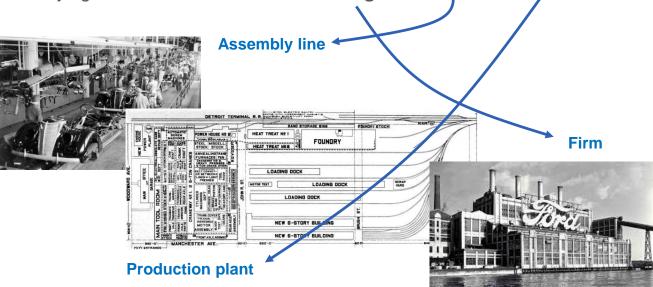
- Dynamic optimization perspective draws comparison between biological and economical processes
- Bacterial cells can be seen as economic agents
 - \circ ... with limited resources
 - ... assigned to productive activities
 - ... so as maximize their profits (or minimize their costs)
 - ... in a changing, competitive market environment
 - ... over a period of time
 - Attractiveness of economic metaphor: provides intuitively plausible language to speak about cellular process

Costs, investment, resource allocation, just-in-time, ...

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



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

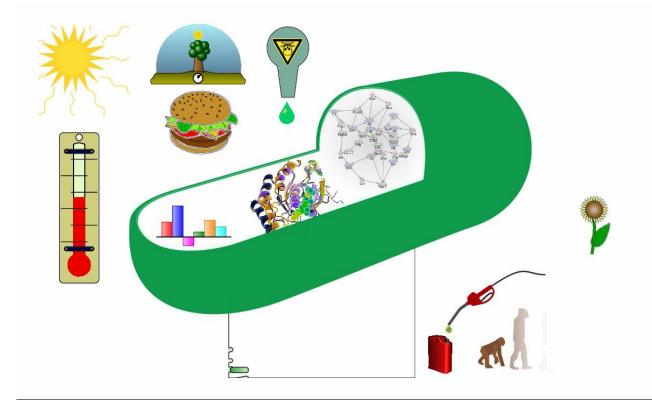


- Economists have developed evolutionary economics
 - Routines (capacities and rules of decision, investment strategies) shape behavior of firm
 - Routines are transmitted within the firm and gradually modified
 - Success of firm determined by ability to adapt routines to changing environment

Nelson and Winter (1982), *An Evolutionary Theory of Economic Change*, Belknap Press of Harvard University Press Gayon (2011), *Biol. Theor.*, 6(4):320–5 https://en.wikipedia.org/wiki/Evolutionary_economics

- Limits to the evolutionary metaphor in economy
 - Firms grow and survive, but do not reproduce
 - Routines are transmitted, but within same evolving firm
 - Economic agents display (bounded) rationality and behavorial learning, rather than blind variation
- This also poses limits on the economic metaphor in biology!

Optimal Cell Behavior in Time – Thank You!



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