

Growth Balance Analysis

Hugo Dourado

Why yet another “balance analysis”?

Growth Balance Analysis (GBA): simplified framework for nonlinear self-replicating cell models at balanced growth¹.

- ▶ **Nonlinear:** includes nonlinear kinetic rate laws.
- ▶ **Self-replicating:** metabolism + protein synthesis and dilution of all components.
- ▶ **Balanced growth:** constant (external and internal) concentrations in time.

A framework, not a model: find common properties to all possible models.

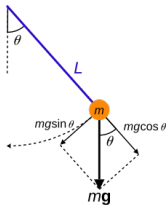
Mathematical simplification: allows analytical study to find fundamental principles.

¹Dourado & Lercher, An analytical theory of balanced cellular growth, *Nature Communications* 2020.

Mathematical simplification: the least number of variables and equations

Not important for linear problems, but critical for nonlinear problems!

Example: Simple pendulum



Angle θ (“generalized coordinate”) completely determines the system state, no need of x, y, z .

Why looking for simplest formulation?

- ▶ Easier numerical calculations.
- ▶ Independent variables are preferable for analytical methods.
- ▶ Deeper understanding of the problem.
- ▶ Most “elegant” solution.



Balanced growth (or steady-state growth)

For a steady-state environment defined by “external” concentrations a:

- ▶ Steady-state growth rate μ (1/h), direct measure of fitness.
- ▶ Steady-state internal concentrations c (g/L) of reactants (substrates, products)

$$c_i = \frac{\text{abundance of "i" (g/cell)}}{\text{volume (L/cell)}} = \text{constant}$$

Mass concentrations (not abundances) better describe cell states: i) constant, ii) reaction kinetics depend on concentrations, iii) relate to cell density (g/L).

Matching units for fluxes v : mass per volume per time ($\text{g L}^{-1} \text{h}^{-1}$).



Main differences to other modeling frameworks

Density constraint: cell density² ρ (g/L) including **all** mass concentrations c (g/L)

$$\rho = c_p + \sum_m c_m$$

where c_p is the total protein concentration, and m are all “non-protein” components³.

Units: to match the units, we normalize \mathbf{N} with the molecular weights \mathbf{w} (g/mol)

$$\mathbf{N} \xrightarrow[\text{columns by } \mathbf{w}]{\text{multiply}} \text{diag}(\mathbf{w}) \mathbf{N} \xrightarrow[\sum(-)=-1, \sum(+)=1]{\text{scale columns, s.t.}} \mathbf{M}_{\text{total}} \xrightarrow[\text{external rows}]{\text{exclude}} \mathbf{M}$$

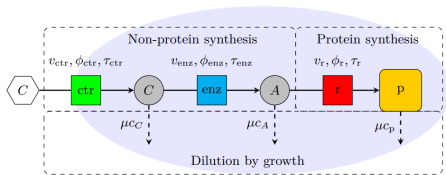
\mathbf{M} entries are mass fractions of reactants into $(-)$ and out $(+)$ each reaction.

² Baldwin et al. *Archives of Microbiology* 1995, Kubitschek et al. *Journal of bacteriology* 1983, Cayley et al. *Journal of Molecular Biology* 1991

³ Dourado et al. *PLOS Comp Bio* 2023.

GBA models: transport, protein synthesis, kinetics, dilution by growth

A) Scheme of a simple GBA model “L3”



B) Mathematical definition (\mathbf{M}, ρ, τ) of the model L3:

$$\mathbf{M} = \begin{bmatrix} \text{ctr} & \text{enz} & \text{r} \\ 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{matrix} (C) \\ (A) \\ (p) \end{matrix}$$

$$\rho = 340 \text{ gL}^{-1}$$

$$\tau = \left(\frac{1}{7} \left(1 + \frac{0.1}{a_C} \right), \frac{1}{7} \left(1 + \frac{23}{c_C} \right), \frac{1}{6} \left(1 + \frac{41}{c_A} \right) \right)$$

C) Mathematical definition of a (static) medium:

$$a_C = 10 \text{ gL}^{-1}$$

D) Cell states ($\mu, \mathbf{v}, \mathbf{c}$) must satisfy:

i) Mass conservation: $\mathbf{M} \mathbf{v} = \mu \mathbf{c}$

$$\underbrace{\begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} v_{\text{ctr}} \\ v_{\text{enz}} \\ v_r \end{bmatrix}}_{\text{flux balance}} = \underbrace{\mu \begin{bmatrix} c_C \\ c_A \\ c_p \end{bmatrix}}_{\text{dilution by growth}}$$

where $\mathbf{s} :=$ column sums of $\mathbf{M} = (1, 0, 0)$

ii) Kinetic rate laws and protein sum: $\mathbf{v} \cdot \tau(\mathbf{a}, \mathbf{c}) = c_p$

$$\frac{v_{\text{ctr}}}{7} \left(1 + \frac{0.1}{10} \right) + \frac{v_{\text{enz}}}{7} \left(1 + \frac{23}{c_C} \right) + \frac{v_r}{6} \left(1 + \frac{41}{c_A} \right) = c_p$$

iii) Density constraint: $\sum \mathbf{c} = \rho$

$$c_C + c_A + c_p = 340$$

The general GBA optimization problem

For some given GBA model $(\mathbf{M}, \boldsymbol{\tau}, \rho)$ and medium concentrations \mathbf{a} :

$$\begin{array}{ll} \text{maximize} & \mu \\ \mathbf{v} \in \mathbb{R}^r, \mathbf{c} \in \mathbb{R}_+^p & \end{array} \quad \text{(Maximize growth rate)}$$

subject to:

$$\mathbf{M} \mathbf{v} = \mu \mathbf{c} \quad \text{(Flux balance)}$$

$$c_p = \mathbf{v} \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c}) \quad \text{(Reaction kinetics and protein sum)}$$

$$\rho = \sum \mathbf{c} \quad \text{(Constant cell density)}$$



No alternative pathways: simplification with \mathbf{c} as independent variables

1) For a square \mathbf{M} , there is an inverse $\mathbf{W} = \mathbf{M}^{-1}$ and

$$\mathbf{M} \mathbf{v} = \mu \mathbf{c} \Rightarrow \mathbf{v} = \mu \mathbf{W} \mathbf{c} \quad .$$

2) Substituting into $c_p = \mathbf{v} \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c})$

$$c_p = \mu (\mathbf{W} \mathbf{c}) \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c}) \quad .$$

3) Solving for μ : we get the objective function $\mu(\mathbf{c}, \mathbf{a})$

$$\mu(\mathbf{a}, \mathbf{c}) = \frac{c_p}{(\mathbf{W} \mathbf{c}) \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c})} \quad .$$

4) The only constraint left:

$$\rho = \sum \mathbf{c} \quad .$$



The GBA problem with no alternative pathways: analytical solution

Reformulated problem: for some given GBA model $(\mathbf{M}, \boldsymbol{\tau}, \rho)$ and medium \mathbf{a}

$$\underset{\mathbf{c} \in \mathbb{R}_+^p}{\text{maximize}} \quad \mu(\mathbf{c}, \mathbf{a}) = \frac{c_p}{(\mathbf{W} \mathbf{c}) \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c})}$$

subject to:

$$\sum \mathbf{c} = \rho \quad .$$

Analytical conditions for optimal states: using Lagrange multipliers, we find

$$\boxed{\mu(\mathbf{W} \mathbf{c}) \cdot \frac{\partial \boldsymbol{\tau}}{\partial c_m} + \mu \boldsymbol{\tau} \cdot (\mathbf{W}_m - \mathbf{W}_p) + 1 = 0 \quad \forall m} \quad (1)$$

We got: # algebraic equations = # variables (solvable).



Cell economics: the costs and benefits of reactants

Substituting $v = \mu W c$ into the solution (1)

$$\frac{c_p}{\mu} \frac{\partial \mu}{\partial c_m} = - \sum_j v_j \frac{\partial \tau_j}{\partial c_m} + \mu \sum_j \tau_j (W_p^j - W_m^j) - 1 = 0 \quad \forall m$$

Economics analogy: costs and benefits (in terms of protein allocation)

(marginal) reactant value = local benefit + global benefit + density cost (= 0 if optimal)

Protein is the underlying “currency”

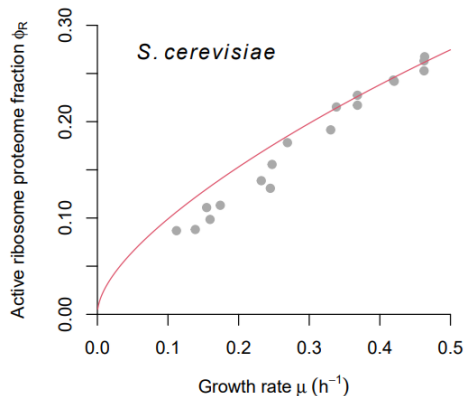
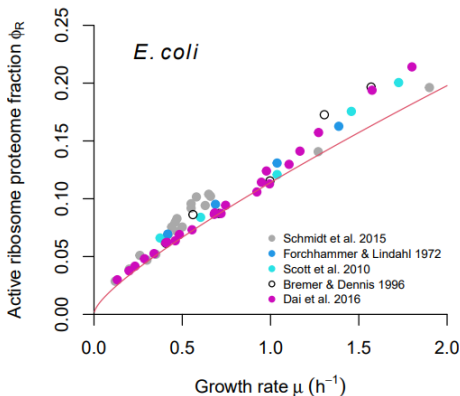
$$- \sum_j v_j \frac{\partial \tau_j}{\partial c_m} = - \sum_j \left(\frac{\partial p_j}{\partial c_m} \right)_{\mathbf{v}=\text{const.}}, \text{ and } \mu \sum_j \tau_j (W_p^j - W_m^j) = - \sum_j \left(\frac{\partial p_j}{\partial c_m} \right)_{\tau, \mu=\text{const.}} \approx 0.03^a$$

^aDourado, Quantitative principles of optimal cellular resource allocation, *PhD Thesis* 2020.



Comparison to data: *E. coli* and yeast ribosome proteome fraction ϕ_R vs. μ

All data available, *in vivo* data close to the predicted optimality⁴ (red lines, no fitting).



⁴ Dourado & Lercher, An analytical theory of balanced cellular growth, *Nature Communications* 2020.

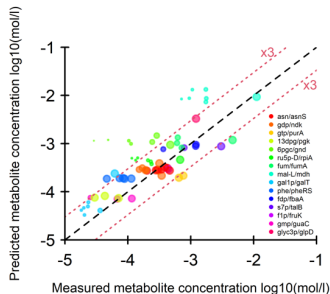


(Approximate) Enzyme-substrate optimality (without k_{cat})

Considering the Michaelis-Menten kinetics and no global benefit:

$$-\sum_j v_j \frac{\partial \tau_j}{\partial c_m} + \mu \sum_j \tau_j (W_p^j - W_m^j) - 1 = 0 \Rightarrow p_j = c_m \left(1 + \frac{c_m}{K_j^m} \right)$$

E. coli enzymes and substrates are close to this optimality⁵



⁵ Dourado et al. On the optimality of the enzyme-substrate relationship in bacteria, *PLOS Biology* 2021.



“Growth Control Analysis”: holistic view of the growing cell

Metabolic Control Analysis (MCA): perturbations on metabolism (open system).

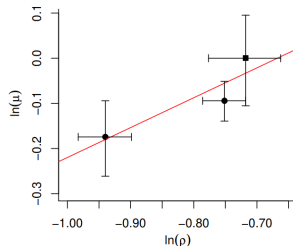
Growth Control Analysis (GCA): perturbations self-replicating system (closed system), all is connected \Rightarrow analytical expressions⁶.

- ▶ Growth Control Coefficients: change in μ by perturbing one concentration c_i .
- ▶ Growth Adaptation Coefficients A : change in optimal μ^* by changing parameters.

E.g.: changing in the density ρ

$$A_\rho = \frac{\rho}{\mu^*} \frac{d\mu^*}{d\rho} = \frac{\rho}{c_p} \left(1 - \mu \sum_j \tau_j W_p^j \right)$$

Comparing to *E. coli* data⁷ \rightarrow



⁶Dourado & Lercher, *Nature Communications* 2020, ⁷ Cayley et. al, *Biophys. J* 2000



The general GBA problem: formulation on \mathbf{q}

Problem simplification on new (adimensional) independent variables:

$$\mathbf{q} := \frac{\mathbf{v}}{\mu \rho}$$

We get the optimality condition⁷ for each reaction j ($s_j := \text{sum of column } M_j$)

$$M_j^p - \mu \tau_j - \mathbf{v} \cdot \frac{\partial \tau}{\partial \mathbf{c}} \mathbf{M}_j + s_j \mathbf{v} \cdot \frac{\partial \tau}{\partial \mathbf{c}} \mathbf{c} / \rho = 0$$

Cell economics: the protein costs and benefits of each reaction

$$\underbrace{\text{production benefit}}_{\text{(protein production)}} + \underbrace{\text{local cost}}_{\text{(protein in } j\text{)}} + \underbrace{\text{local benefit}}_{\text{(local saturation)}} + \underbrace{\text{transport benefit}}_{\text{(global saturation)}} (= 0 \text{ if opt.})$$

⁷ Dourado et al. Mathematical properties of optimal fluxes in cellular reaction networks at balanced growth, *PLOS Comp Biol* 2023.



Grow Control Analysis: Grow Adaptation Coefficient for k_{cat}

We can show from first principles (using the Envelope Theorem)⁸ that:

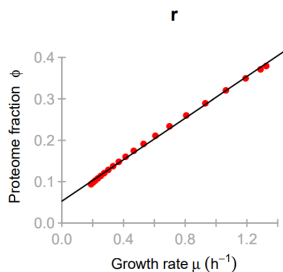
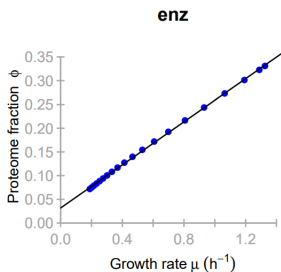
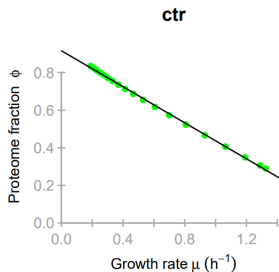
$$A_{k_{\text{cat}}^j} = \frac{k_{\text{cat}}^j}{\mu^*} \frac{d\mu^*}{dk_{\text{cat}}^j} = \phi_j$$

Proportional change in μ^* is exactly the same as proportion of protein allocated to j .

⁸Dourado et al. Mathematical properties of optimal fluxes in cellular reaction networks at balanced growth, *PLOS Comp Biol* 2023.

Simplified mathematical formulation also facilitates numerical solutions

Model L3 on different media (≈ 0.1 s):



Genome-scale GBA models are feasible: 10 reactions ≈ 1 s, 100 reactions ≈ 1 min.



Computational tools for GBA

Numerical implementation in R (including also dynamical simulations):

`https://github.com/HDourado/Growth_Mechanics`

Online tool: Cell growth simulator

`https://cellgrowthsim.com/`



Summary

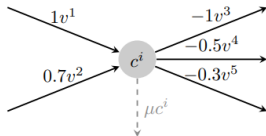
- ▶ **GBA: self-replicating cell models on independent variables, easier to study.**
- ▶ **Analytical conditions for optimal balanced growth (fundamental principles).**
- ▶ **Experimental indications that cells do implement near optimal strategies.**
- ▶ **Proteins emerge as the “currency” in cell economics from first principles.**

(soon chapter in the EPCB book)

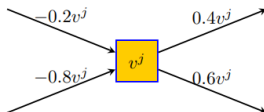


Constraints on GBA

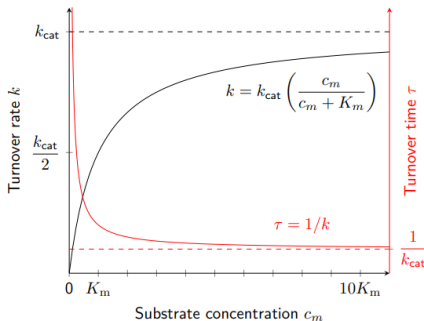
A) Flux balance for each reactant "i"



B) Mass balance within each reaction "j"



C) Kinetics: $v = p \cdot k(\mathbf{c}, \mathbf{x})$ or $p = v \cdot \tau(\mathbf{c}, \mathbf{x})$



D) Density constraints

$$\begin{aligned}
 \text{i)} \quad & \boxed{\phi^1} + \dots + \boxed{} + \boxed{\phi^r} = 1 \\
 \text{ii)} \quad & \textcircled{c^1} + \textcircled{c^2} + \dots + \textcircled{c^p} = \rho
 \end{aligned}$$

Michaelis-Menten kinetics with activation

Based on “Convenience kinetics”⁹, we define the Michaelis-Menten kinetics with activation, corresponding “activation constants” \mathbf{A}

$$\tau_j = \frac{1}{k_{\text{cat}}^j} \prod_m \left(1 + \frac{A_j^m}{c_m} \right) \left(1 + \frac{K_j^m}{c_m} \right) \prod_n \left(1 + \frac{K_j^n}{a^n} \right)$$

⁹ Liebermeister & Klipp, Bringing metabolic networks to life: convenience rate law and thermodynamic constraints, 2006.



Optimal substrate mass concentration = free enzyme mass concentration

The optimal mass concentration balance for minimal ρ :

$$c_m = \frac{p^j K_m^j}{K_m^j + c_m} \quad .$$

But this corresponds exactly to the free enzyme mass concentration

$$p_{\text{free}}^j := p^j - p^j \left(\frac{c_m}{c_m + K_m^j} \right) = \frac{p^j K_m^j}{K_m^j + c_m} \quad .$$

Thus¹⁰,

$$c_m = p_{\text{free}}^j \quad .$$

¹⁰Dourado et al. On the optimality of the enzyme–substrate relationship in bacteria, *PLOS Biology* 2021



Equations for balance growth states: model L3

1) **Original problem: Implicit constraints on μ , involving $v_1, v_2, v_3, c_1, c_2, c_3, a_1$ (6 variables, 5 equations)**

$$v_1 - v_2 = \mu c_1$$

$$v_2 - v_3 = \mu c_2 \quad (\text{mass conservation})$$

$$v_3 = \mu c_3$$

$$\frac{v_1}{7} \left(1 + \frac{1}{a_1}\right) + \frac{v_2}{7} \left(1 + \frac{23}{c_1}\right) + \frac{v_3}{6} \left(1 + \frac{41}{c_2}\right) = c_3 \quad (\text{kinetics and protein sum})$$

$$c_1 + c_2 + c_3 = 340 \quad (\text{constant cell density})$$

2) **GBA: Explicit constraint on $\mu(c_1, c_2, a_1)$ (using $c_3 = 340 - c_1 - c_2$)**

$$\mu(c_1, c_2, a_1) = \frac{340 - c_1 - c_2}{\frac{1}{7} \left(1 + \frac{1}{a_1}\right) + \frac{340 - c_1}{7 \cdot 340} \left(1 + \frac{23}{c_1}\right) + \frac{340 - c_1 - c_2}{6 \cdot 340} \left(1 + \frac{41}{c_2}\right)} \quad (\text{constrained growth rate})$$

3) **Analytical conditions for optimal balanced growth state (system of algebraic equations)**

$$\mu \frac{23 (340 - c_1)}{(c_1)^2} + \mu \left[\frac{1}{7} \left(1 + \frac{23}{c_1}\right) + \frac{1}{6} \left(1 + \frac{41}{c_2}\right) \right] - 1 = 0 \quad (m = 1)$$

$$\mu \frac{40 (340 - c_1 - c_2)}{6 (c_2)^2} + \mu \left[\frac{1}{5} \left(1 + \frac{41}{c_2}\right) \right] - 1 = 0 \quad (m = 2)$$



The general GBA problem: formulation on \mathbf{q}

General formulation on \mathbf{q} in few steps

Substituting $\mathbf{v} = \mu \rho \mathbf{q}$ into $\mathbf{M} \mathbf{v} = \mu \mathbf{c}$

$$\rho \mathbf{M} \mathbf{q} = \mathbf{c} \quad (\text{independent of } \mu).$$

Substituting $\mathbf{c} = \rho \mathbf{M} \mathbf{q}$ into $c_p = \mathbf{v} \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c})$

$$M_r^p q_r = \mu \mathbf{q} \cdot \boldsymbol{\tau}(\rho \mathbf{M} \mathbf{q}, \mathbf{a})$$

Solving for μ :

$$\mu(\mathbf{q}, \mathbf{a}) = \frac{M_r^p q_r}{\mathbf{q} \cdot \boldsymbol{\tau}(\rho \mathbf{M} \mathbf{q}, \mathbf{a})}$$

The density constraint:

$$\rho = \sum \mathbf{c} \Leftrightarrow \boxed{\mathbf{s} \cdot \mathbf{q} = 1}$$

The general GBA problem: analytical “solution”

Reformulated problem: for some given model (\mathbf{M}, τ, ρ) and environment \mathbf{a}

$$\underset{\mathbf{q} \in \mathbb{R}^r}{\text{maximize}} \quad \mu(\mathbf{q}, \mathbf{a}) = \frac{M_r^p q_r}{\mathbf{q} \cdot \tau(\rho \mathbf{M} \mathbf{q}, \mathbf{a})}$$

subject to:

$$\mathbf{s} \cdot \mathbf{q} = 1$$

$$\mathbf{q} \odot \tau(\rho \mathbf{M} \mathbf{q}, \mathbf{a}) \geq \mathbf{0} \quad .$$

Analytical conditions for optimal states: using KKT conditions, we find

$$\boxed{\left(M_j^p - \mu \tau_j - \mu \mathbf{q} \cdot \frac{\partial \tau}{\partial q^j} + s_j \mu \mathbf{q} \cdot \frac{\partial \tau}{\partial \mathbf{q}} \mathbf{q} \right) q_j = 0 \quad \forall j} \quad (2)$$

Using $\mathbf{s} \cdot \mathbf{q} = 1$: # algebraic equations = # variables (solvable)..



Equations for balance growth states: model L3

1) **Original problem: Implicit constraints on μ , involving $v_1, v_2, v_3, c_1, c_2, c_3, a_1$ (6 variables, 5 equations)**

$$v_1 - v_2 = \mu c_1$$

$$v_2 - v_3 = \mu c_2 \quad (\text{mass conservation})$$

$$v_3 = \mu c_3$$

$$\frac{v_1}{7} \left(1 + \frac{1}{a_1}\right) + \frac{v_2}{7} \left(1 + \frac{23}{c_1}\right) + \frac{v_3}{6} \left(1 + \frac{41}{c_2}\right) = c_3 \quad (\text{kinetics and protein sum})$$

$$c_1 + c_2 + c_3 = 340 \quad (\text{constant cell density})$$

2) **GBA: Explicit constraint on $\mu(q_2, q_3, a_1)$ (from the density constraint $q_1 = 1$)**

$$\mu(q_2, q_3, a_1) = \frac{q_3}{\frac{1}{7} \left(1 + \frac{1}{a_1}\right) + \frac{q_2}{7} \left(1 + \frac{23}{340(1 - q_2)}\right) + \frac{q_3}{6} \left(1 + \frac{41}{340(q_2 - q_3)}\right)} \quad (\text{constrained growth rate})$$

3) **Analytical conditions for optimal balanced growth state (system of algebraic equations)**

$$\frac{1}{7} \left(1 + \frac{23}{340(1 - q_2)}\right) + \frac{23q_2}{7 [340(1 - q_2)]^2} - \frac{41q_3}{6 [340(q_2 - q_3)]^2} = 0 \quad (j = 2)$$

$$1 - \mu \frac{1}{6} \left(1 + \frac{41}{340(q_2 - q_3)}\right) - \mu \frac{41q_3}{6 [340(q_2 - q_3)]^2} = 0 \quad (j = 3)$$



The dynamic generalization: fitness optimization

For some given model $(\mathbf{M}, \boldsymbol{\tau}, \rho)$ and **dynamic medium** $\mathbf{a}(t)$:

$$\begin{array}{ll} \text{maximize} & \int_0^T \mu \, dt \quad (\text{Maximize fitness}) \\ \mathbf{v}(t), \mathbf{c}(t) & \end{array}$$

subject to:

$$\mathbf{M} \mathbf{v} = \mu \mathbf{c} + \dot{\mathbf{c}} \quad (\text{Mass conservation})$$

$$c_p = \mathbf{v} \cdot \boldsymbol{\tau}(\mathbf{a}, \mathbf{c}) \quad (\text{Reaction kinetics and protein sum})$$

$$\rho = \sum \mathbf{c} \quad (\text{Constant cell density})$$

Main trick for analytical “solution”: define the “generalized fluxes” \mathbf{q} such that

$$\rho \mathbf{M} \mathbf{q} = \mathbf{c} \quad ,$$

then reformulate the problem on $\dot{\mathbf{q}}, \mathbf{q}, \mathbf{a}$, and solve Euler-Lagrange equations.

