

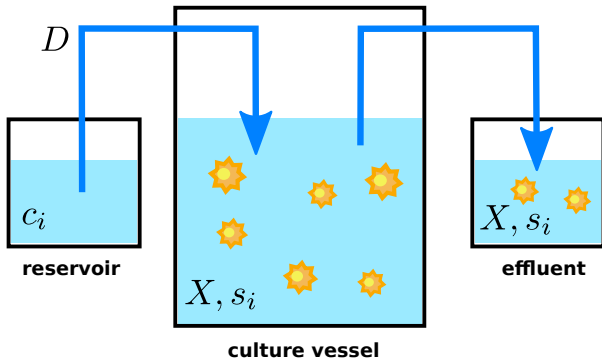
Stationary states of genome scale metabolic networks in continuous cell cultures

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Requirements

- ▶ It must include the internal metabolism
- ▶ It must include the chemostat
- ▶ Be computationally scalable to Genome scale metabolic networks
- ▶ Be flexible
- ▶ Toxicity
- ▶ Heterogeneity

Outline

Homogeneous chemostat

Mathematical framework

Stationary States

From a Toy model to Genome Scale

Heterogeneous chemostat

Maximum Entropy Principle

The Toy model again

Genome Scale Metabolic Network

Conclusions

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Chemostat

$$\frac{dX}{dt} = (\mu - \sigma - D)X \quad (1)$$

$$\mu = \mu(\nu) \quad \sigma = \sigma(s) \quad (2)$$

$$\frac{ds_i}{dt} = -u_i X - (s_i - c_i)D \quad (3)$$

The cell

$$lb_k \leq r_k \leq ub_k$$

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$$\sum_k r_k < K$$

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This is a polytope in very high dimensions

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The cell maximizes biomass production μ
Linear Programming LP

Mathematical framework

$$\frac{dX}{dt} = (\mu - \sigma - D)X$$

$$\mu = \mu(\nu) \quad \sigma = \sigma(s)$$

$$\frac{ds_j}{dt} = -u_j X - (s_j - c_j)D$$

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The cell maximizes biomass production μ LP

$$u_i^*(\xi) \dots \mu(\xi)$$

Equilibrium in metabolite's concentration

$$\frac{ds_i}{dt} = -u_i^* X - (s_i - c_i)D$$

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Stationarity in cell's concentration

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$$0 = (\mu^*(\xi) - \sigma^*(\xi) - D)X^*$$

Stationarity in cell's concentration

$$\frac{dX}{dt} = (\mu - \sigma - D)X$$

$$D = \mu^*(\xi) - \sigma^*(\xi)$$

Stationarity in cell's concentration

$$\frac{dX}{dt} = (\mu - \sigma - D)X$$

$$\frac{X^*}{\xi} = \mu^*(\xi) - \sigma^*(\xi)$$

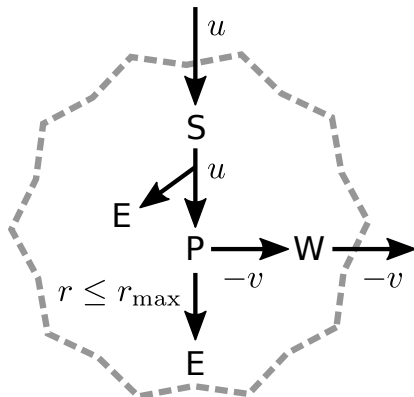
Stationarity equations

$$r_k^* \dots u_i^*(\xi) \dots \mu^*(\xi)$$

$$s_i^*(\xi) = c_i - u_i^*(\xi)\xi$$

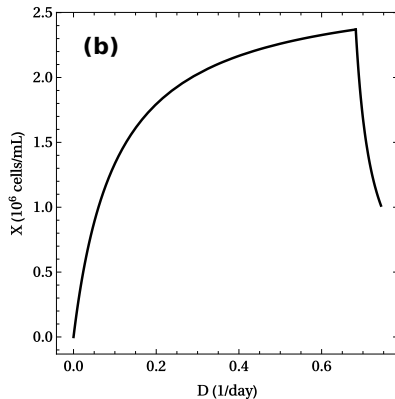
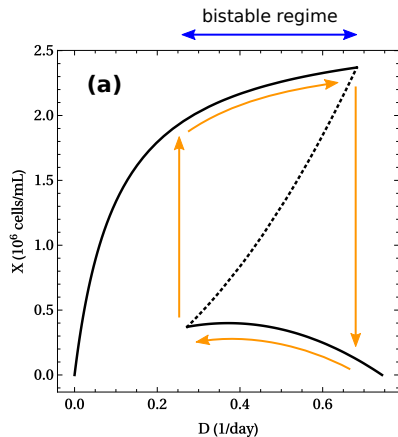
$$\frac{X^*(\xi)}{\xi} = \mu^*(\xi) - \sigma^*(\xi)$$

Small Network

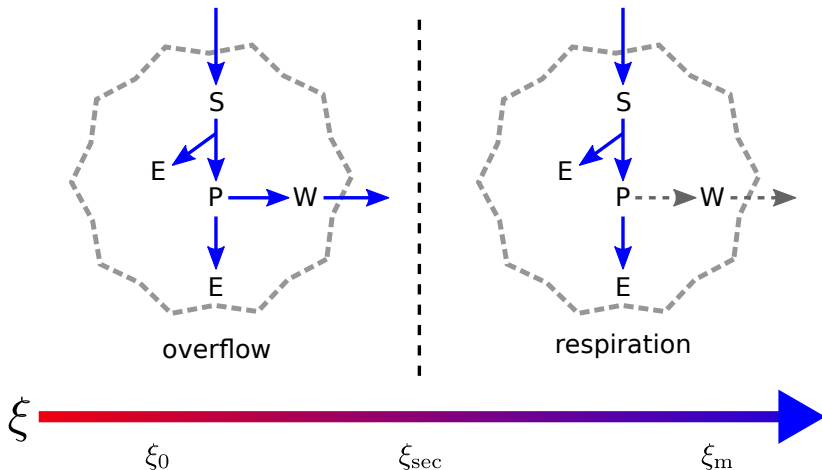


Vazquez et al.. Macromolecular crowding explains overflow metabolism in cells. *Scientific Reports* 6, 31007 (2016)

Toxicity is the key point



General Picture

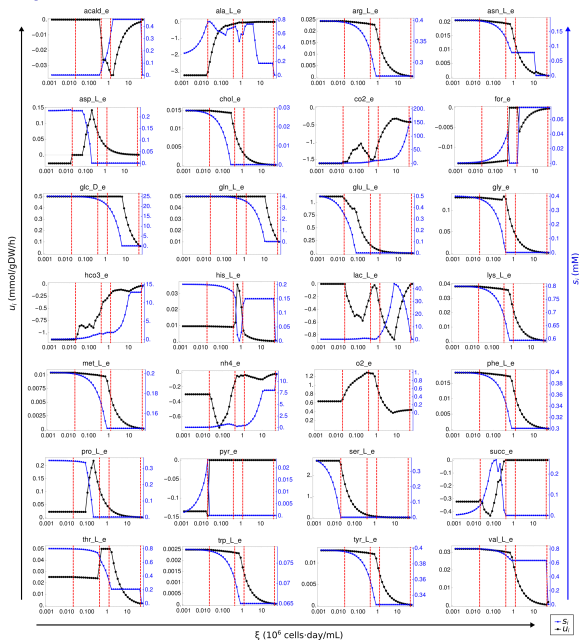


- (a) Overflow. At high enough nutrient uptake the respiratory flux hits the upper bound r_{max} and the remaining nutrients are exported as W . (b) Respiration. The nutrient is completely oxidized with a large energy yield. (c) Threshold values of ξ . ξ_0 delimits the nutrient excess regime ($\xi < \xi_0$) from the competition regime ($\xi > \xi_0$). ξ_{sec} delimits the transition between overflow metabolism ($\xi < \xi_{sec}$) and respiration ($\xi > \xi_{sec}$). Finally, maintenance demand cannot be met beyond $\xi > \xi_m$.

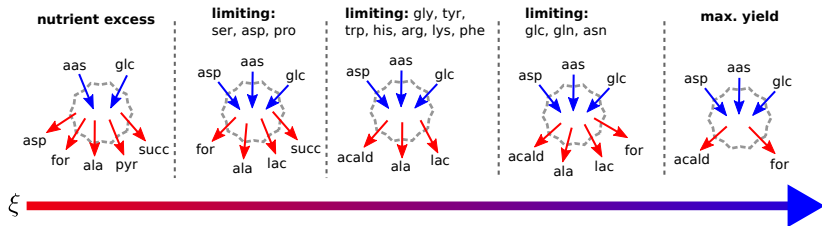
Genome Scale: CHO-K1 line

- ▶ 6663 reactions
- ▶ $V_{glc} = 0.5 \text{ mmol/gDW/h}$
- ▶ $V_i = .1 V_{glc}$

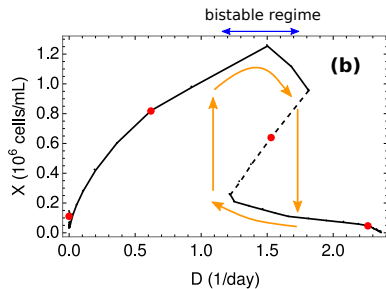
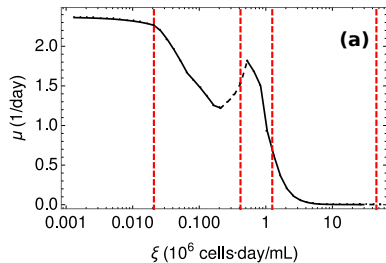
Metabolite uptakes and concentration



General picture of the transitions



Steady state and bifurcation



J. Fernandez-de-Cossio Diaz, K. León and R. M., *Characterizing stationary states of genome scale metabolic networks in continuous culture*, *PLOS Computational Biology*. **13** (11): e1005835 (2017)

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Constraints

$$\sum_k S_{ik} r_k - e_i - y_i \mu + u_i = 0$$

$$lb_k \leq r_k \leq ub_k$$

$$-L_i \leq u_i \leq \min\{V_i, c_i \frac{D}{X}\}$$

$$\sum_i r_i < K$$

We must explore this polytope

Stationarity: Dealing with the heterogeneity

$$\frac{d\vec{X}}{dt} = (\mu - \sigma - D)\vec{X}$$

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$$0 = (\mu(\nu) - \sigma(s) - D)$$

$$\text{Effective Growth rate} = \mu(\nu) - \sigma(s) = \bar{D}$$

Maximum Entropy Principle

If s is fixed, $\mu(\nu) - \sigma(s^) = D$*

Maximum Entropy Principle

$$P_{s^*}(\nu) \sim e^{\beta(\mu(\nu) - \sigma(s^*))}$$

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$$s_i = c_i - \frac{1}{D} \sum_a u_i^a$$

Maximum Entropy Principle

$$P_{s^*}(\nu) \sim e^{\beta(\mu(\nu) - \sigma(s^*))}$$

$$s_i^* = c_i - \frac{X}{D} \int_{\Pi} u_i(\nu) P_{s^*}(\nu) d\nu$$

In short

$$D = \frac{X}{\xi} = \frac{\int_{\Pi} d\nu [\mu(\nu) - \sigma(\mathbf{s}^*)] e^{\beta(\mu(\nu) - \sigma(\mathbf{s}^*))}}{\int_{\Pi} d\nu e^{\beta(\mu(\nu) - \sigma(\mathbf{s}^*))}}$$

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$$s_i^* = c_i - \xi \int_{\Pi} u_i(\nu) P_{s^*}(\nu) d\nu$$

Homogeneous vs Heterogeneous Chemostat

$$D = \frac{X}{\xi} = \langle \mu(\nu) - \sigma(s^*) \rangle_{P_{s^*}}$$

$$s_i^* = c_i - \xi \int_{\Pi} u_i(\nu) P_{s^*}(\nu) d\nu$$

$$\frac{X^*(\xi)}{\xi} = \mu^*(\xi) - \sigma^*(\xi)$$

$$s_i^*(\xi) = c_i - \xi u_i^*(\xi)$$

Summarizing

$$D = \langle \mu(\nu) - \sigma(s^*) \rangle_{P_{s^*}}$$

$$s_i^* = c_i - \frac{X}{D} \int_{\Pi} u_i(\nu) P_{s^*}(\nu) d\nu$$

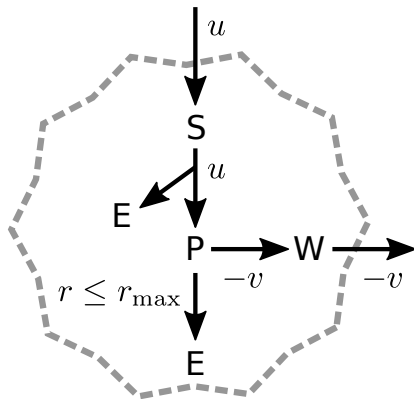
$$\sum_k S_{ik} r_k - e_i - y_i \mu + u_i = 0$$

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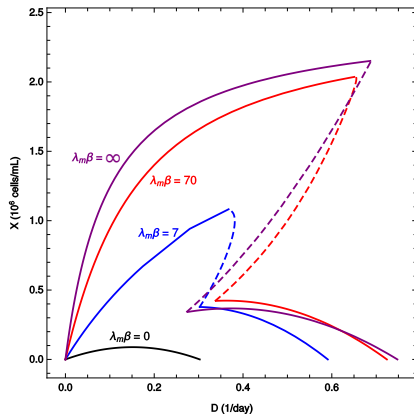
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$$\sum_i r_i < K$$

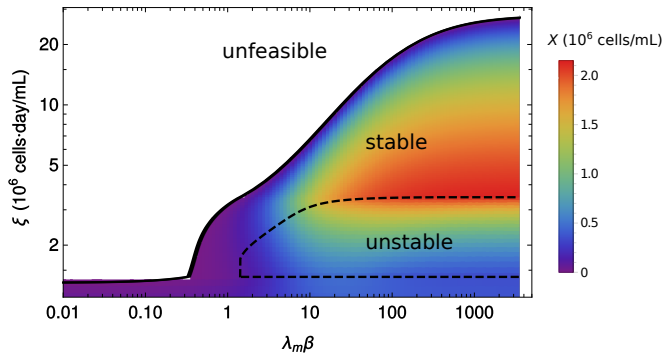
Small Network again



Effect of the heterogeneity



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Exploring the space

Π

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$$-L_i \leq u_i \leq \min\left\{V_i, c_i \frac{\bar{D}}{X}\right\}$$

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$$\prod$$

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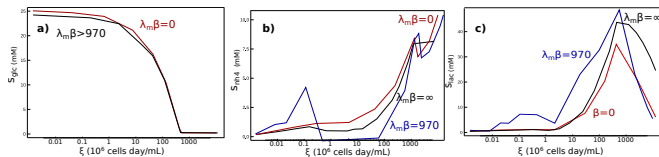
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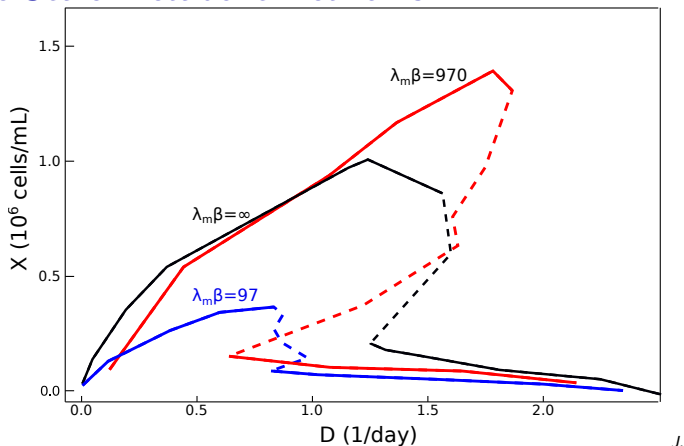
For $\beta = \infty$: **Expectation Propagation** Alfredo Braunstein, Anna Paola Muntoni, Andrea Pagnani, *An analytic approximation of the feasible space of metabolic networks*, *Nat. Comm.* **8**, 14915 (2017)

Here generalized for finite β

Genome Scale Metabolic Networks



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- ▶ We developed a mathematical framework to determine the stationary states in a chemostat
- ▶ The presence of toxic waste:
 - ▶ drives the appearance of many stationary states
 - ▶ makes relevant the history of the system
- ▶ We provided a scheme to estimate the metabolic flux distribution of an heterogeneous culture in a chemostat
- ▶ The presence of heterogeneity in the culture
 - ▶ changes the concentration of metabolites
 - ▶ allows stationary states with a larger number of cells
- ▶ Everything is computationally tractable in Genome Scale metabolic networks

Collaborators and acknowledgments

- ▶ Jorge Fernández de Cossio. Centre for Molecular Immunology-CIM and Physics Faculty, UH. Cuba
- ▶ Kalet León. Centre for Molecular Immunology-CIM. Cuba

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