

Stability analysis of metabolic systems

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Jacobian of a differential system

Let us consider a system of ordinary differential equations (ODEs)

$$d\mathbf{x}/dt = \mathbf{f}(\mathbf{x})$$

We define its Jacobian matrix as the matrix of its partial derivatives

$$\mathfrak{J} := \partial \mathbf{f} / \partial \mathbf{x}$$

which is a square matrix

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System evolution around steady-state

Let us now consider the system around a steady-state \mathbf{X}

$$d\mathbf{x}/dt(\mathbf{X}) = \mathbf{f}(\mathbf{X}) = \mathbf{0}$$

In the vicinity of \mathbf{X} we may use the first order approximation

$$d\mathbf{x}/dt \sim \mathfrak{J} \cdot [\mathbf{x} - \mathbf{X}]$$

which integrates into

$$\mathbf{x} - \mathbf{X} = \exp(\mathfrak{J}t) \cdot [\mathbf{x}(0) - \mathbf{X}]$$

using the matrix exponential

$$\exp(\mathfrak{J}t) := \sum_{k=0}^{\infty} \frac{1}{k!} \mathfrak{J}^k t^k$$

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Stability conditions around steady-state

Consider the eigenvalues λ_i of the Jacobian matrix

The steady-state is unstable if

$$\exists i, \operatorname{Re}(\lambda_i) > 0$$

The steady-state is exponentially stable if

$$\forall i, \operatorname{Re}(\lambda_i) < 0$$

with relaxation times $\tau_i = 1 / |\operatorname{Re}(\lambda_i)|$

and frequencies $\omega_i = \frac{|\operatorname{Im}(\lambda_i)|}{2\pi}$

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Bifurcations

Consider the eigenvalues $\lambda_i(\mathbf{p})$ of the Jacobian matrix when parameters vary

A saddle-node bifurcation corresponds to a zero-crossing of one real eigenvalue λ_i

A Hopf bifurcation corresponds to a zero-crossing of the real parts $\operatorname{Re}(\lambda_j)$ of one pair of conjugated eigenvalues

$$\operatorname{Re}(\lambda_j) \pm 2i\pi\omega_j$$

There are several other more complex bifurcation types

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From the evolution equation

$$d\mathbf{x}/dt = \mathbf{N} \cdot \mathbf{v}(\mathbf{x}, \mathbf{p})$$

we derive the Jacobian

$$\mathfrak{J} = \mathbf{N} \cdot \partial \mathbf{v} / \partial \mathbf{x}$$

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However this Jacobian is singular if \mathbf{N} has not maximal rank.
It is then useful to reduce the system to independent variables:

$$d\mathbf{x}^0/dt = \mathbf{N}^0 \cdot \mathbf{v}(\mathbf{x}, \mathbf{p})$$

with
$$\mathbf{N} = \mathbf{L} \cdot \mathbf{N}^0$$

$$\partial \mathbf{x} / \partial \mathbf{x}^0 = \mathbf{L}$$

and we derive the Jacobian

$$\mathfrak{J} = \mathbf{N}^0 \cdot \partial \mathbf{v} / \partial \mathbf{x} \cdot \mathbf{L}$$

that must be definite negative for the system to be stable

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What makes a metabolic system stable?

Structural kinetic modeling of metabolic networks

Ralf Steuer^{***}, Thilo Gross^{**}, Joachim Selbig^{**}, and Bernd Blasius^{*}

Steuer *et al.* (2006), *PNAS* 103:11868-11873

Different notations:

x_i normalized by X_i (dimensionless)

v_j normalized by J_j

$$\mu_j := v_j / J_j$$

$$\Lambda_{ij} := N_{ij} J_j / X_i$$

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What makes a metabolic system stable?

so that the system evolution follows:

$$d\mathbf{x} / dt = \mathbf{A} \cdot \boldsymbol{\mu}(\mathbf{x})$$

$$\mathfrak{J} = \mathbf{A} \cdot \frac{\partial \boldsymbol{\mu}}{\partial \mathbf{x}}$$

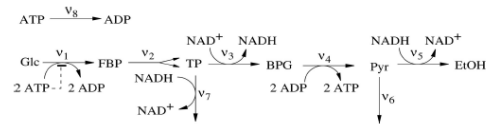
where

$$\boldsymbol{\theta} := \frac{\partial \boldsymbol{\mu}}{\partial \mathbf{x}}$$

is the matrix of normalized elasticities (usually noted $\boldsymbol{\epsilon}$)

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Example: simplified yeast glycolysis

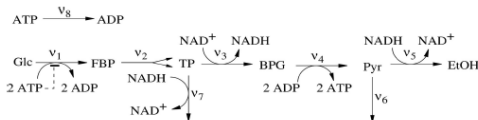


with an inhibition parameter ξ for PFK by ATP:

$$\theta_{ATP}^{\mu_1} = 1 - \xi$$

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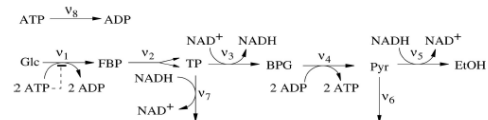
Stoichiometry matrix



	ν_1	ν_2	ν_3	ν_4	ν_5	ν_6	ν_7	ν_8
FBP	+1	-1	0	0	0	0	0	0
TP	0	+2	-1	0	0	0	-1	0
BPG	0	0	+1	-1	0	0	0	0
Pyr/ACA	0	0	0	+1	-1	-1	0	0
ATP	-2	0	0	+2	0	0	0	-1
NADH	0	0	+1	0	-1	0	-1	0
NAD+	0	0	-1	0	+1	0	+1	0
ADP	+2	0	0	-2	0	0	0	+1

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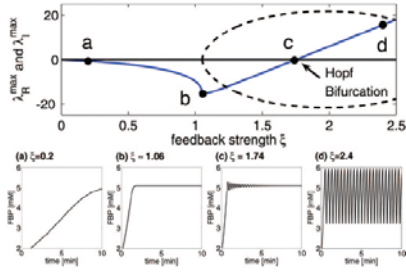
Normalized elasticity matrix



	FBP	TP	BPG	Pyr/ACA	ATP	NADH	NAD+	ADP
ν_1	0	0	0	0	$\theta_{ATP}^{\mu_1}$	0	0	0
ν_2	$\theta_{FBP}^{\mu_2}$	0	0	0	0	0	0	0
ν_3	0	$\theta_{TP}^{\mu_3}$	0	0	0	0	$\theta_{NAD}^{\mu_3}$	0
ν_4	0	0	$\theta_{BPG}^{\mu_4}$	0	0	0	0	$\theta_{ADP}^{\mu_4}$
ν_5	0	0	0	$\theta_{Pyr}^{\mu_5}$	0	$\theta_{NADH}^{\mu_5}$	0	0
ν_6	0	0	0	$\theta_{Pyr}^{\mu_6}$	0	0	0	0
ν_7	0	$\theta_{TP}^{\mu_7}$	0	0	0	$\theta_{NADH}^{\mu_7}$	0	0
ν_8	0	0	0	0	$\theta_{ATP}^{\mu_8}$	0	0	0

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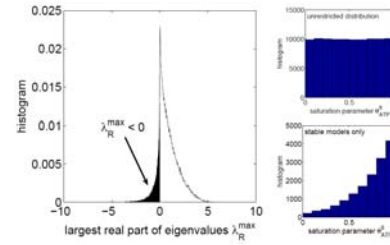
Effect of ATP feedback on stability



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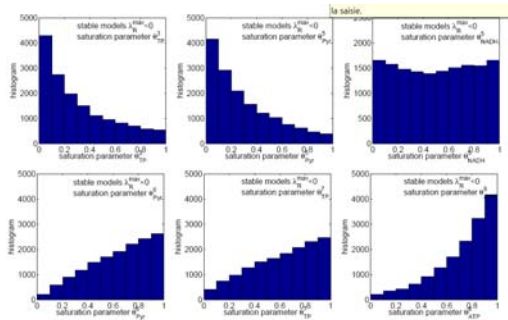
Stabilization and saturation

Random sampling of parameters:
destabilization by saturation of ATP consumption



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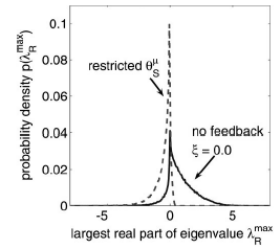
Stabilization and saturation



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Stabilization and saturation

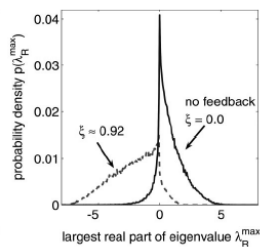
Random sampling of parameters
Constraining $\theta_{ATP}^{\mu} = \theta_{Pyr}^{\mu} = \theta_{TP}^{\mu} = 0.9$ far from saturation



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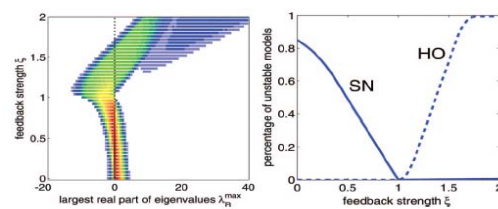
Stabilization by feedback

Random sampling of parameters



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Destabilization by feedback



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