Robustness/plasticity of biochemical reaction systems and network topology

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Mochizuki A, Fiedler B. JTB (2015), Okada T., Mochizuki A. PRL (2016) Okada T., Mochizuki A. PRE (2017) Okada T., Tsai JC, Mochizuki A arXiv:1711.00250



KEGG database

Enzyme knockdown experiment



Reaction network of central carbon metabolism

Reaction rate functions are not known precisely.

Can we determine system's behaviors from network structure alone??? (without knowledge of reaction rate functions)

OUTLINE



Chemical Reaction systems Dynamics





A. Qualitative response is determined from network structure alone.

$$A \equiv \left(\begin{array}{c} \sum_{i} S_{mi} W_{i}(k_{i}, \bar{x}) = 0\\ \sum_{i} S_{mi} W_{i}(k_{i} + \delta k, \bar{x} + \delta \bar{x}) = 0 \end{array} \right)$$

$$A \equiv \left(\begin{array}{c} J \\ J \\ J_{im} = \frac{\partial W_{i}}{\partial x_{m}} = \begin{cases} + (m \in \text{substrate})\\ 0 (\text{otherwise}) \end{cases} \right) \left(\begin{array}{c} \frac{\partial \bar{x}_{m}}{\partial k_{j}} = -(A^{-1})_{mj} \right) \\ \end{array} \right)$$

$$H = \left(\begin{array}{c} \frac{\partial W_{i}}{\partial x_{m}} = \left(\begin{array}{c} + (m \in \text{substrate})\\ 0 (\text{otherwise}) \end{array} \right) \\ \end{array} \right) \left(\begin{array}{c} \frac{\partial \bar{x}_{m}}{\partial k_{j}} = -(A^{-1})_{mj} \right) \\ \end{array} \right)$$

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OUTLINE



限局則 (the law of localization)

Okada T. & Mochizuki A. (2016) Phys. Rev. Lett. 117, 048101.



BS corresponds to a square submatrix



 $\chi(\Gamma) \equiv |\mathcal{M}| - |\mathcal{R}| + \# \text{cycle in } \Gamma \stackrel{???}{=} 0$







$$\chi(\Gamma) \equiv \#$$
metabolite $- \#$ reac. $+ \#$ cycle $\stackrel{???}{=} 0$



$$\chi(\Gamma) \equiv \#$$
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$$\chi(\Gamma) \equiv \#$$
metabolite $- \#$ reac. $+ \#$ cycle $\stackrel{???}{=} 0$



$$\chi(\Gamma) \equiv \#$$
metabolite - $\#$ reac. + $\#$ cycle $\stackrel{???}{=} 0$



note: There are 17 buffering structures in total.





A hierarchy appears from a nest of buffering structures









A hierarchy appears from a nest of buffering structures

G3P 7

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Plasticity and Bifurcation Phenomena

Ertugrul M. Ozbudak, et al, Nature, 2004



The concentration of the protein (necessary to utilize the nutrient) change discontinuously.

Bifurcation theory

$$\frac{dx}{dt} = f(x) = -k + x^2$$

$$\int_{0}^{\bar{x}} \int_{k}^{\bar{x}} k J = \frac{\partial f}{\partial x}|_{x=x^*} = 2x^* = \pm 2\sqrt{k}$$

At bifurcation point (k=0), the Jacobian J becomes 0.

In a multivariate case, bif point ⇔ det J =0

$$\begin{aligned} \dot{x} &= f(x, y) \\ \dot{y} &= g(x, y) \end{aligned} \mathbf{J} = \begin{pmatrix} \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \\ \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \end{pmatrix} \Big|_{\vec{x} = \vec{x}^*} \\ \frac{\partial f}{\partial x} & \frac{\partial f}{\partial y} \end{aligned}$$

$$\det \mathbf{J} = \frac{\partial f}{\partial x} \frac{\partial g}{\partial y} - \frac{\partial f}{\partial y} \frac{\partial g}{\partial x}$$

Bifurcation analysis based on the matrix A

At bif. point, $\det J = 0$

$$\mathbf{A} = \left(\frac{\partial \boldsymbol{r}}{\partial \boldsymbol{x}}\Big|_{\boldsymbol{x}=\boldsymbol{x}^*} \mid \ker S \right) \qquad \qquad \frac{d\mathbf{x}}{dt} := S\mathbf{r}(x) \\ \mathbf{J} := S\frac{\partial \mathbf{r}}{\partial \mathbf{x}}|_{\boldsymbol{x}=\boldsymbol{x}^*}$$



$$\mathbf{A} = \begin{pmatrix} \mathbf{A}_{\Gamma} & \ast \\ \mathbf{A}_{\bar{\Gamma}} & \mathbf{A}_{\bar{\Gamma}} \end{pmatrix} \qquad \det \mathbf{A} = \det \mathbf{A}_{\Gamma} \times \det \mathbf{A}_{\bar{\Gamma}} = 0$$

$$\operatorname{decomposition}$$



 $\overline{\Gamma}$

Γ

Bifurcation analysis of reaction systems based on network structures

Okada T., Tsai JC, Mochizuki A. arXiv:1711.00250





Summary

Mochizuki A, Fiedler B. JTB (2015), Okada T., Mochizuki A. PRL (2016) Okada T., Mochizuki A. PRE (2017)

- Responses and network topology
 - Responses are determined from network topology
 - Buffering structures explain response patterns

- Bifurcation and network topology
 - Det J = Det A

Okada T., Tsai JC, Mochizuki A arXiv:1711.00250