

BRIEF SYNOPSIS:

BIOCHEMICAL NETWORKS

- information processing
- decision & control
- effector function

λ PHAGE

- 2 states; long-term stability
- multiple interactions among proteins & binding sites

DIFFERENTIAL EQUATION FORMULATION:

$$\frac{d\vec{x}}{dt} = F(\vec{x}, \vec{u}) = A^{(1)}\vec{x} + A^{(2)}(\vec{x} \odot \vec{x}) + B\vec{u} + C(\vec{u} \otimes \vec{x}) + D(\vec{u} \odot \vec{u})$$

generally don't write

For steady states:

$$\frac{d\vec{x}}{dt} = 0 \Rightarrow F(\vec{x}, \vec{u}) = 0$$

Solve using NEWTON'S METHOD:

$$\frac{dF(\vec{x}, \vec{u})}{d\vec{x}} = \nabla_{\vec{x}} F \equiv J_F(\vec{x})$$

$$F(\vec{x}'; \vec{u}) = F(\vec{x}^0; \vec{u}) + \underbrace{\frac{dF}{d\vec{x}}(\vec{x}^0; \vec{u})}_{J_F(\vec{x}^0)} (\vec{x}' - \vec{x}^0) + H.O.T.$$

$$\text{Set } F(\vec{x}'; \vec{u}) = 0 \Rightarrow J_F(\vec{x}^0) (\vec{x}' - \vec{x}^0) = -F(\vec{x}^0; \vec{u})$$

solve for \vec{x}'

$$\text{iterate: } J_F(\vec{x}^i) (\vec{x}^{i+1} - \vec{x}^i) = -F(\vec{x}^i; \vec{u})$$

⋮

RUN SIMULATIONS:

- Forward Euler - explicit
- Backward Euler - implicit (iterative)

PARAMETER SENSITIVITIES:

$$\min_{\Delta A} f(\Delta A) = \min_{\Delta A} \sum_{i=1}^{\# \text{ of experiments}} [C^T \vec{x}^i(\Delta A) - C^T \vec{x}^i_{\text{measured}}]^2$$

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OPTIMIZE:

• GRADIENT DESCENT

$$\Delta A' = \Delta A^0 + \beta \nabla_{\Delta A} f$$

pick stepsize β s.t. $\min f(\Delta A^0 + \beta \nabla_{\Delta A} f)$

This method does not work well

• ADJOINT OPTIMIZATION

Set gradient to zero

$$\frac{\partial}{\partial(\Delta A)} [C^T \vec{x}^L(\Delta A) - C^T \vec{x}^L_{\text{measured}}]^2$$

$$= 2 [C^T \vec{x}^L(\Delta A) - C^T \vec{x}^L_{\text{measured}}] \left[\frac{\partial}{\partial(\Delta A)} C^T \vec{x}^L(\Delta A) \right]$$

straightforward difficult

$$\frac{\partial}{\partial(\Delta A)} C^T \vec{x}^L(\Delta A) = \frac{1}{2} C^T \frac{\partial \vec{x}^L(\Delta A)}{\partial(\Delta A)}$$

→ sensitivities
n = # of concentrations
p = # of parameters

$$\frac{dF}{d(\Delta A)} = 0 \Rightarrow \frac{\partial F}{\partial(\Delta A)} + \frac{\partial F}{\partial \vec{x}} \frac{\partial \vec{x}}{\partial(\Delta A)} = 0 \Rightarrow \frac{\partial F}{\partial \vec{x}} \frac{\partial \vec{x}}{\partial(\Delta A)} = - \frac{\partial F}{\partial(\Delta A)}$$

⋮
solve for this term & plug-in
HARDER TO SOLVE

Rather than solve the previous problem, solve its DUAL:

$$\frac{\partial}{\partial(\Delta A)} C^T \vec{x}^L(\Delta A) = \vec{z}^T \left(\frac{\partial F}{\partial(\Delta A)} \right)$$

where \vec{z} is obtained from:

$$n \left(\frac{\partial F}{\partial \vec{x}} \right)^T \vec{z} = - \vec{c}$$

EASIER TO SOLVE

SHOW THAT THE DUAL IS EQUIVALENT:

$$C^T \frac{\partial \vec{x}^L(\Delta A)}{\partial(\Delta A)} = C^T \left(- \frac{\partial F}{\partial \vec{x}} \right)^{-1} \left(\frac{\partial F}{\partial(\Delta A)} \right)$$

FROM * $\Rightarrow \frac{\partial}{\partial(\Delta A)} C^T \vec{x}^L(\Delta A) = \vec{z}^T \frac{\partial F}{\partial(\Delta A)}$

$$\vec{z}^T = C^T \left(- \frac{\partial F}{\partial \vec{x}} \right)^{-1}$$

$$\Rightarrow \vec{z}^T \left(\frac{\partial F}{\partial \vec{x}} \right) = - C^T$$

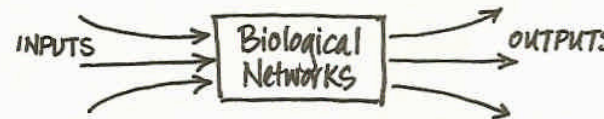
$$\Rightarrow \left(\frac{\partial F}{\partial \vec{x}} \right)^T \vec{z} = - C \quad \square$$

* Finding steady States

* Running Simulations

- Model Construction

- parameter estimation
- can also be used in design mode
- parameter sensitivities



- absence of glucose
- presence of galactose

- synthesize multiple enzymes for galactose utilization
- turns on galactose transporters

PROBLEMS IN BIOLOGY:

- Difficulty in maintaining constant concentration
 - cell volume changes
 - small counts make uniformity difficult
- Highly variable environment
 - Temperature
 - Humidity
- Certain functions persist across all stages: development, cell cycle, tissue types
- Evolve

