

A design principle in biochemical reaction networks based on realization theory

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Outline

- Some special features of the differential equations of biochemical reaction networks
- Why are there so many reactant species?
- Possible connection with a nonlinear realization problem
- A conjecture and a problem formulation
- A hack for the solution

Some features of the equations

- Biochemical reaction networks seem to always contain *terms with special forms*
- Many reactant species, but *“function” appears to be described by a small subset*
- e.g. Circadian Oscillation in Drosophila:

$$\begin{aligned}\dot{x}_4 &= v_{3p} \frac{x_3}{k_{3p} + x_3} - v_{4p} \frac{x_4}{k_{4p} + x_4} - k_3 x_4 x_8 + k_4 x_9 \\ &\quad - v_{dp} \frac{x_4}{K_{dp} + x_4} - k_d x_4 \\ \dot{x}_5 &= K_{iT}^n \frac{v_{st}}{K_{iT}^n + x_{10}^n} - v_{mt} \frac{x_5}{k_{mt} + x_5} - k_d x_5\end{aligned}$$

- Each special form \iff Specific biochemical reaction mechanism

A model reduction question

Given a biochemical reaction network, say

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ \vdots \\ x_{30} \end{bmatrix} = f(x_1, \dots, x_{30}) \quad (1)$$

If function is described by a small number of states,

say x_1, x_{12}

does there exist a reduced model

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_{12} \end{bmatrix} = g(x_1, x_{12}) \quad (2)$$

such that *behavior of (x_1, x_{12}) in (1) \iff behavior of (x_1, x_{12}) in (2)*

A Similar Question in Reverse

Given a low order system

$$\dot{z} = g(z),$$

z : *small number of variables,*

g : *possibly complicated function,*

can this system be imbedded in

$$\dot{X} = F(X),$$

X : *possibly large number of variables,*

F *has special form?*

$$\{z_1, \dots, z_r\} \subset \{X_1, \dots, X_n\}$$

embedded: Every trajectory of z is part of a trajectory of X

A nonlinear realization problem

Implementation with *Linear* Elements: Carleman Linearization

Example: $\dot{z} = g(z) = z^2$

Define: $X_1 := z, X_2 := z^2, \dots, X_n := z^n, \dots$

$$\frac{d}{dt} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ \vdots \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & \dots \\ 0 & 0 & 2 & 0 & 0 & \dots \\ 0 & 0 & 0 & 3 & 0 & \dots \\ 0 & 0 & 0 & 0 & 4 & \dots \\ \vdots & & & & & \ddots \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ \vdots \end{bmatrix}, \quad \text{i.e.} \quad \dot{\mathcal{X}} = \mathcal{A}\mathcal{X}$$

The original system is imbedded in this linear system

Easily generalizable to vector z and analytic g using

$$X_{(i_1, \dots, i_n)} := z^{i_1} \dots z^{i_n}$$

Implementation with *Quadratic* Elements

Consider: $\dot{z} = g(z)$,

Can embed in a larger system

$$\dot{\mathcal{X}} = \mathcal{Q}(\mathcal{X})$$

where \mathcal{Q} is **Quadratic**

g polynomial $\Rightarrow \mathcal{X}$ finite
 \mathcal{X}, \mathcal{Q} non-unique

(Starkl & del Re, ACC, CDC '03)

Note:

- Can approximate any system with \mathcal{Q} a neural network
- Many other possibilities for the building blocks of \mathcal{Q}

Implementation with *Quadratic* Elements (cont.)

Example: $\dot{z} = z^4$

Define $x_i := z^i$

$$\dot{x}_1 = x_4$$

$$\dot{x}_4 = 4z^3 \dot{z} = 4z^3 z^4 = 4x_3 x_4$$

$$\dot{x}_3 = 3z^2 \dot{z} = 3x_2 x_4 = 3x_3^2$$

$$\dot{x}_2 = 2z \dot{z} = 2x_1 x_4$$

Choice of Z and RHS are non-unique!

Procedure terminates in finite number of steps

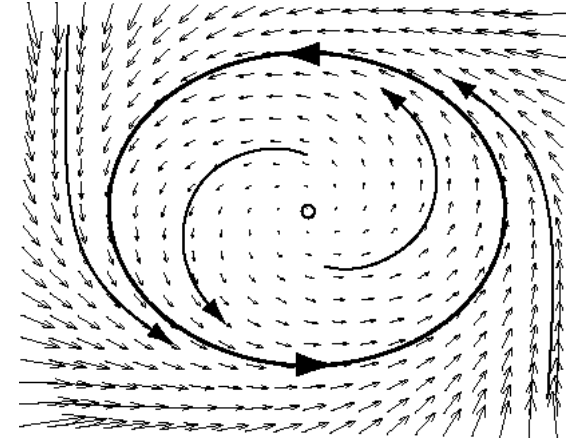
Realization (Implementation) Issues

Goal: Realize given dynamics $\dot{z} = g(z)$
with a given set of motifs (building blocks)

- Choice of new intermediate variables
 - Monomials ($x_i := z^i$)
 - Orthogonal Polynomials ($x_i := J_i(z)$)
 - ...
- Available building blocks, e.g.
 - Quadratic terms ($k_{ij}x_ix_j$, restrictions on reaction rates k_{ij})
 - Hill terms $\frac{x_i}{K+x_i}$
- Original dynamics are an invariant manifold of $\dot{z} = g(z)$
 - for implementation: must be a *stable* invariant manifold

A Design Principle

- Design dynamics for prescribed function,
e.g. as p increases, **oscillator** \longrightarrow **bistable switch**



- Design vector field f for dynamics $\dot{z} = f(z, p)$
 - Can generally be done with 2 or 3 states.
 - **but f is not realizable with given motifs**
- Apply an expansion procedure to imbed dynamics in a larger system

$$\dot{z} = g(z, p)$$

- g constructed from available motifs

A Conjecture

Why so many intermediate reactant species?

One answer:

To realize function using building blocks available with biochemistry

In other words:

Biological function may be easily described by $\dot{z} = g(z)$, w/ small z

But g may contain terms not realizable with biochemical reactions

The Reverse Embedding Problem in General

Given a large system

$$\dot{x} = f(x),$$

with specified building blocks (allowable reaction terms).

Did it arise from the “expansion” of a smaller system

$$\dot{z} = g(z),$$

where z a subset of x ?

Possibly much more interesting than forward embedding problems
which are hard to employ

A Hack for the Embedding Problem

- Start with original large system

$$\dot{x} = f(x)$$

- Simulate to produce trajectories “representative” of typical bio-function
- Identify a subset of states (relabel them z) that describe bio-function
- Use simulation data to fit a model

$$\dot{z} = g(z),$$

where g is parametrized, e.g. polynomial of some order.

Reformulation as Linear Regression

For polynomial g , fitting can be made into linear regression. Define

$$\begin{aligned} z^{(1)} &= z \\ z^{(2)} &= z \otimes z \\ z^{(3)} &= z \otimes z^{(2)} \\ &\vdots \end{aligned} \tag{3}$$

Then

$$g(z) = G Z,$$

$$Z := [1 \ z' \ z^{(2)'} \ \dots \ z^{(m)'}]'$$

Coefficients of polynomial g are entries of matrix G

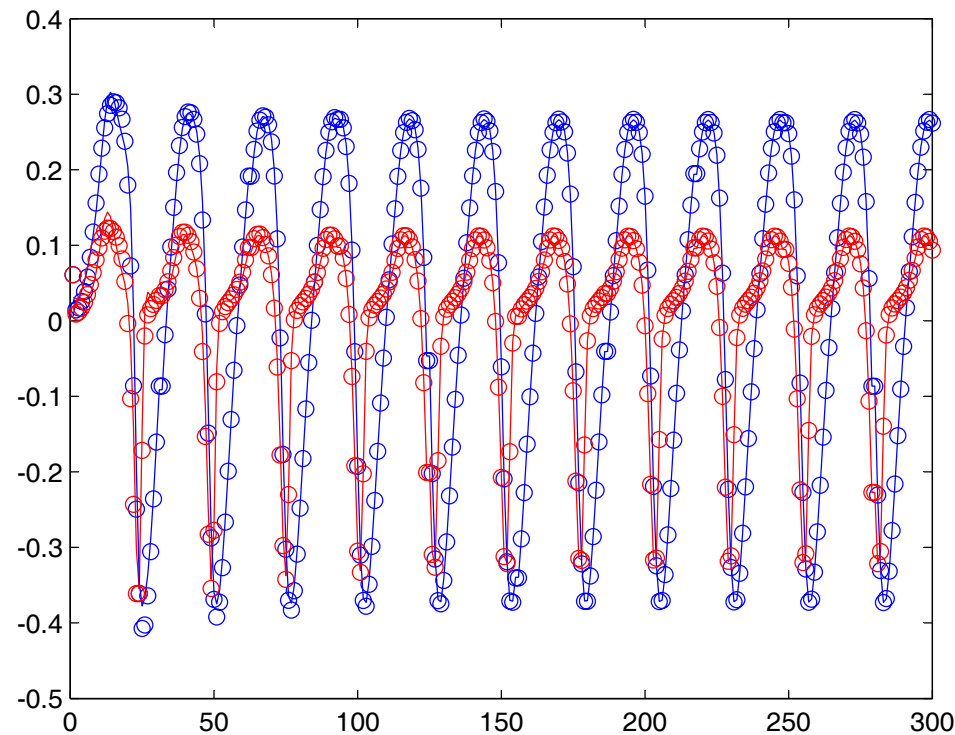
Fitting trajectories to $\dot{z} = GZ \Leftrightarrow$ Linear regression for entries of G

Example: Circadian Oscillation in Drosophila

Original model has 10 species

Function is described by 2: PER & TIM

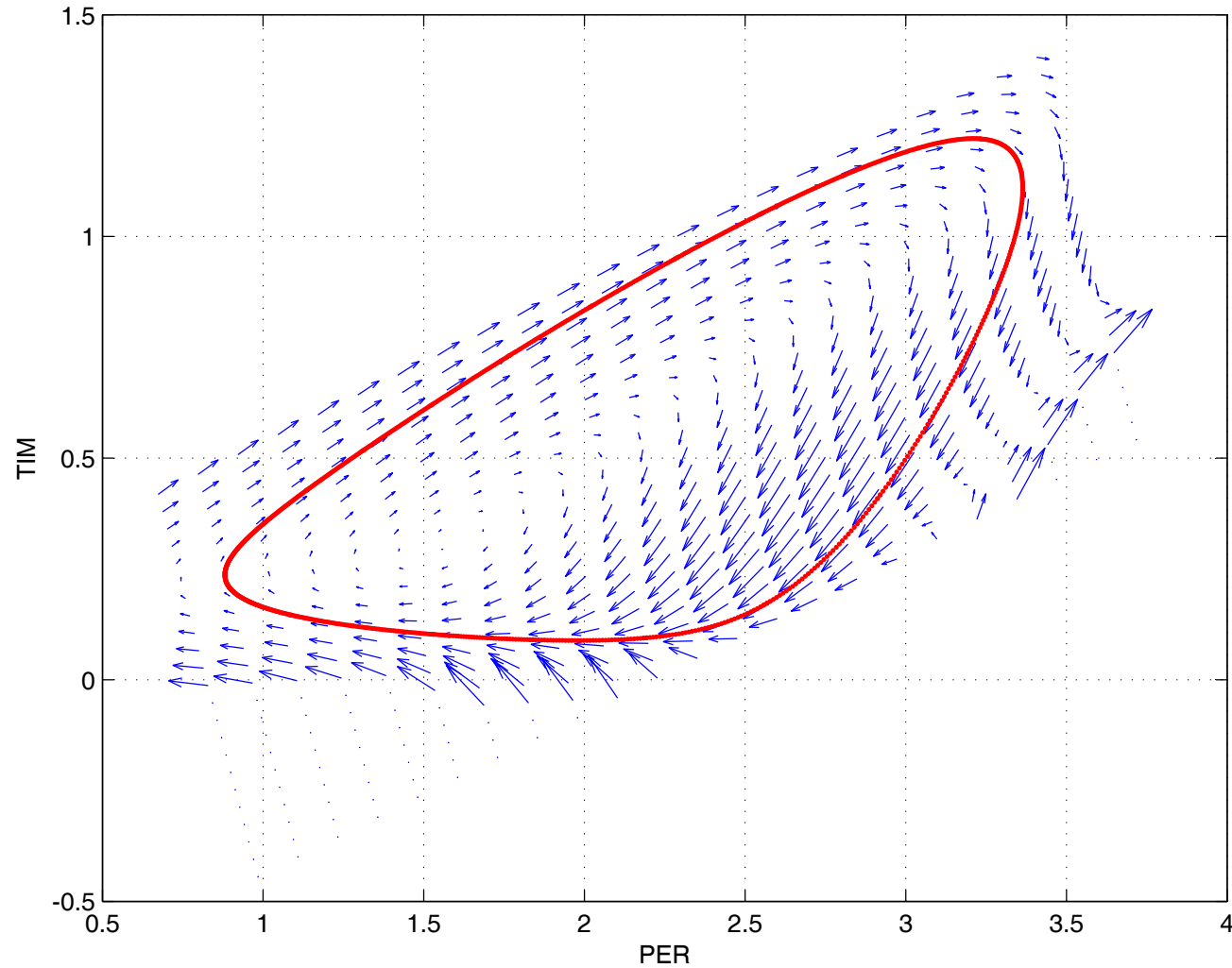
Reduced model in example shown: 2 states, g of degree 5



Original PER/TIM (solid)

Reduced PER/TIM (dots)

Example: Circadian Oscillation in Drosophila (Cont.)



Vector field of the reduced model

Limit cycle trajectory of original system (Bamieh & Giarre, ACC '07)

Comments

- Hack is easily generalizable to include parameters and/or inputs

$$\dot{z} = g(z, p, u)$$

- Difficult to compare reduced and original model
- Need better quantification of reduction error or
comparison of behavior of original and reduced
- The original forward and reverse mathematical problems are open