

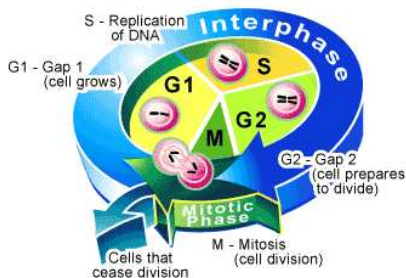
## Inverse bifurcation analysis for reverse engineering of gene systems

James Lu

Johann Radon Institute for Computational and Applied Mathematics  
(RICAM) Linz, Austria

*Joint work with Heinz W. Engl (Inverse Problems Group, RICAM)  
Peter Schuster (Theoretical Biochemistry Group, University of Vienna)*

# Motivation



Schematic of phases of cell cycle

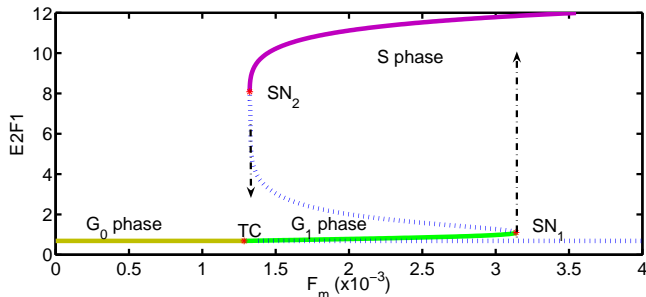
- Concentrations of regulatory species  $x$  typically modelled by **parameterized** ODE systems:

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

- Mathematical correspondence for changes in cycle phase: **bifurcation** of solution

# Bifurcation analysis in molecular biology

- Bifurcation diagram for genetic "switch" regulating mammalian cell cycle



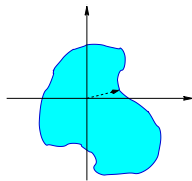
- Biologically important geometric characterization of bifurcation diagram:
  - $x$ -location of  $SN_2$ : whether switch of state is permanent
  - length of "nose" for saddle-node: whether transition of state is delayed

# Inverse bifurcation analysis

- In **forward problems** one looks at predictions of given model
- In **inverse problems** one looks for causes of observed or desired effects
  - typically ill-posed: the solution may not *exist*, may not be *unique*, or does not *depend continuously* on data
  - $\Rightarrow$  mathematical theory of regularization
- **Forward** bifurcation analysis:
  - map **parameters**  $\rightarrow$  **bifurcation diagrams**
  - e.g.: *is the genetic switch irreversible for given parameter values?*
- **Inverse** bifurcation analysis:
  - map **geometry of bifurcation diagrams**  $\rightarrow$  **parameters**
  - e.g.: *what are the mechanisms for the nose of saddle-node to increase under given signals?*

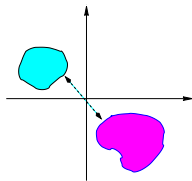
# Geometric properties of bifurcation diagrams

- Geometric properties: **location**, **size**, and **spatial relationship**
- Parametric distance to bifurcation



quantify: robustness; ability of 'tuning' to environment

- Parametric distance between qualitative regions

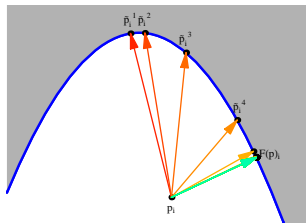


quantify: evolvability to perform different functions

# Robustness to simultaneous parameter variation

- Need to compute distance to bifurcation under **simultaneous** parameter variation
- Iterative method for computing (locally minimum) distance to bifurcation manifold (I. Dobson 1993, M. Mönnigmann and W. Marquardt 2002):
  - based on performing a series of one-param continuation
  - geometric convergence if conditions on magnitude of principal curvatures are satisfied

# Iteratively computing (locally) closest point

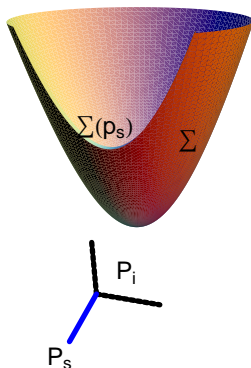


ALGORITHM:  $\text{LOCMINDIST}(x, (p_i, p_s), v, \varepsilon)$

- Set initial parameter:  $p^0 \leftarrow p, x^0 \leftarrow x$
- FOR  $j = 1, \dots, j_{max}$ 
  - 1 From  $p$  and  $x$ , continue along parameter ray  $\{(p_i + rv, p_s) : r \in \mathbb{R}_+\}$  until bifurcation point  $p^b$  detected
  - 2 Compute normal vector at  $p^b$ :  $v \leftarrow N_i(p^b)$
  - 3 Update: parameter  $p^j \leftarrow p^b$ , ODE solution  $x^j \leftarrow x(p^b)$
  - 4 Terminate if  $\|p^j - p^{j-1}\| / \|p^0\| < \varepsilon$

# Mathematical description

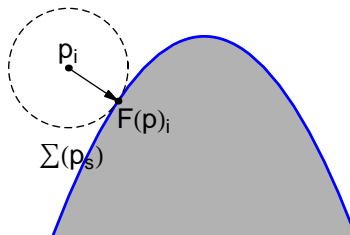
- Consider splitting of parameter space  $P: p = (p_i, p_s) \in P_i \times P_s$ ,
  - 1  $P_i$ : space of *control inputs*
  - 2  $P_s$ : space of *system parameters*
- $\Sigma$ : bifurcation manifold in total parameter space  $P$
- Define  $\Sigma(p_s) \equiv \Sigma \cap \{p_s\}$  the intersection of  $\Sigma$  with the  $p_s$ -plane:



# Forward operator and functional

- Consider the orthogonal projection operator  $F : P \rightarrow P$ ,

$$\begin{aligned} F(p) &\equiv (F(p)_i, F(p)_s) \\ &= (\mathbb{P}_{\Sigma(p_s)} p_i, p_s). \end{aligned}$$



- Functional:  $l_2$ -distance  $J(p) = \|F(p) - p\|_{l_2}$
- Adjoint analysis: allows for efficient implementation by gradient-based solution methods

# Sensitivity analysis by adjoint method

- Denote linearization of forward operator and functional as  $F'(p)$  and  $\langle \cdot, l \rangle$
- Define *adjoint operator*  $F'^*(p)$  (as in general Hilbert space setting) by:

$$\langle F'(p)\delta p, \delta \tilde{p} \rangle = \langle \delta p, F'^*(p)\delta \tilde{p} \rangle, \quad \forall \delta \tilde{p}, \delta \tilde{p} \in \mathbb{R}^m.$$

- By defining the *adjoint solution*  $\psi \equiv F'^*(p)l$ , obtain equivalence relation:

$$\langle \delta p, \psi \rangle = \langle F'(p)\delta p, l \rangle$$

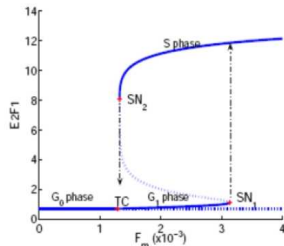
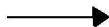
- For codim-1 bifurcations of equilibrium, components of adjoint solutions given by:

$$\begin{aligned} \text{Saddle node} & : \mathbf{w}^T \mathbf{f}_p \\ \text{Hopf} & : \operatorname{Re}\{\mathbf{w}^T \mathbf{f}_{xp} \mathbf{v} - (\mathbf{w}^T \mathbf{f}_{xx} \mathbf{v}) \mathbf{f}_x^{-1} \mathbf{f}_p\} \end{aligned}$$

# Software implementation for inverse bifurcation

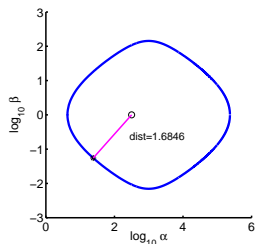
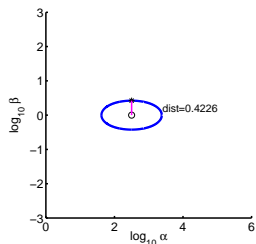
- Packages involved
  - **MathSBML** (B. E. Shapiro): reads biological models in **Systems Biology Mark-up Language (SBML)** format into Mathematica
  - **Mathematica Symbolic Toolbox for Matlab** (B. E. Barrowes): allows communication between Matlab and Mathematica
  - **CL\_MATCONT** (A. Dhooge, W. Govaerts, Y. Kuznetsov, B. Sautois, etc.): performs bifurcation analysis
  - **Matlab Optimization Toolbox** (MathWorks): solves inverse bifurcation using Sequential Quadratic Programming (SQP)

```
<model name = "oscillating_MAPK">
  <XMLListofCompartments>
    <compartment name = "uVol" volume = "1"/>
  </listOfCompartments>
  <listOfSpecies>
    <specie name = "MKKK" boundaryCondition = "false" initialAmount = "90"
    <specie name = "MKKK_P" boundaryCondition = "false" initialAmount = "10"
  </listOfSpecies>
  <listOfParameters>
    <parameter name = "V1" value = "2.5"/>
    <parameter name = "k1" value = "9"/>
    <parameter name = "ki" value = "10"/>
  </listOfParameters>
  <listOfReactions>
    <reaction name = "J0" reversible = "false">
      <listOfReactants>
        <specieReference specie = "MKKK" stoichiometry = "1"/>
      </listOfReactants>
      <listOfProducts>
        <specieReference specie = "MKKK_P" stoichiometry = "1"/>
      </listOfProducts>
      <kineticLaw formula = "V1*MKKK/((1+(MAPK_PP/k1)^n)*(K1+MKKK)"/>
      <listOfParameters>
        <parameter name = "n" value = "1"/>
      </listOfParameters>
    </reaction>
  </listOfReactions>
</model>
```





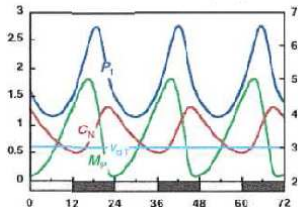
# Applications: symmetric repressilator (Müller *et. al.*)



$$(\delta, h) : (10^{-3}, 1.5) \rightarrow (10^{-4}, 2)$$

# Application: circadian rhythm

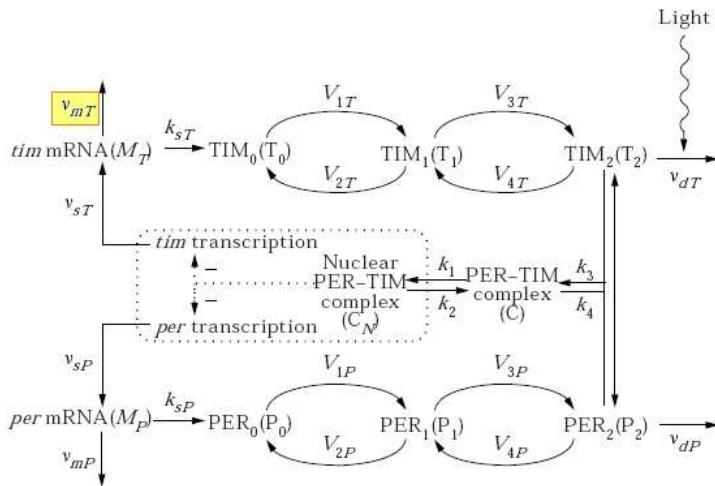
- Sustained, autonomous biochemical oscillations with  $\approx 24$  hour period



- Observed property: **period** of circadian rhythm **insensitive to temperature variations**
- Homeostasis: under environmental perturbation,
  - 1 **oscillatory behavior** is conserved
  - 2 **period of oscillation** remains constant

# Test case: Leloup/Goldbeter model (1999)

- Model: 10 variables, 38 parameters



# Limit cycle continuation in CL\_MATCONT

- Numerical continuation of limit cycles: reformulation as boundary value problem
- Time transformation  $t \rightarrow t/T$ . Find on unit circle  $\mathbb{S}$ ,  $(x(t), T) \in C^1(\mathbb{S}) \times \mathbb{R}$  satisfying

$$\begin{aligned}\dot{x} - Tf(x, p) &= 0 \\ \langle x, \tilde{f} \rangle_{L_2(\mathbb{S})} &= 0\end{aligned}$$

- At limit point for cycles (**LPC**), existence of null space for the adjoint operator:  $(\phi(t), \sigma) \in C^1(\mathbb{S}) \times \mathbb{R}$ ,

$$\begin{aligned}\dot{\phi} + Tf_x(x, p)^t \phi + \sigma f(x, p) &= 0 \\ \langle \phi, f(x, p) \rangle_{L_2(\mathbb{S})} &= 0\end{aligned}$$

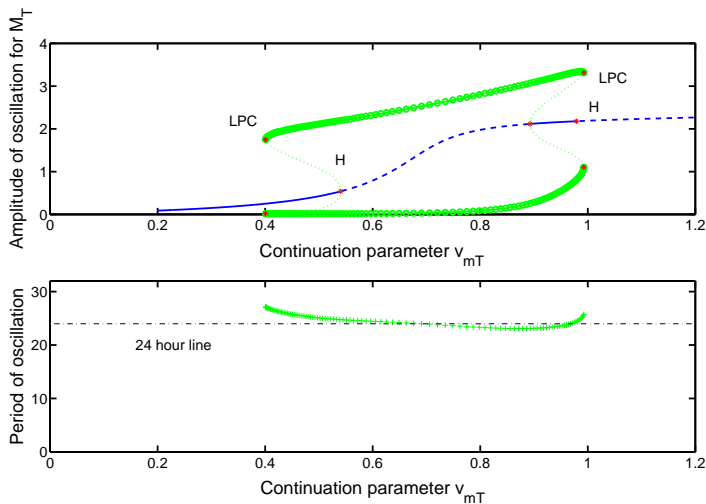
- Sensitivity of distance to **LPC** is (up to scalar factor):

$$\langle \phi, Tf_p(x, p) \rangle_{L_2(\mathbb{S})}$$

# Numerical analysis

- Adjoint analysis can be carried out in either discrete or continuous settings
  - convergence of discrete adjoint to continuous counterpart
- Collocation methods for optimal control:
  - **J. A. Pietz, M. Heinkenschloss** (2003, Rice Univ. thesis): convergence of discrete adjoint approximation in optimal control of ODEs using Legendre pseudospectral collocation
  - **S. Kameswaran, L. T. Biegler** (2004): convergence of optimal control using collocation at Radau points
- CL\_MATCONT: Lagrange interpolation at evenly-spaced points, collocation at Legendre points
- If integral constraint  $\langle \cdot, f \rangle_{L_2(\mathbb{S})} = 0$  computed using Gauss-Legendre quadrature, discrete adjoint coefficients divided by G-L weights provide approximation to continuous adjoint approximation at Legendre points
- Further work: prove convergence rates, numerical tests

# Bifurcation diagram

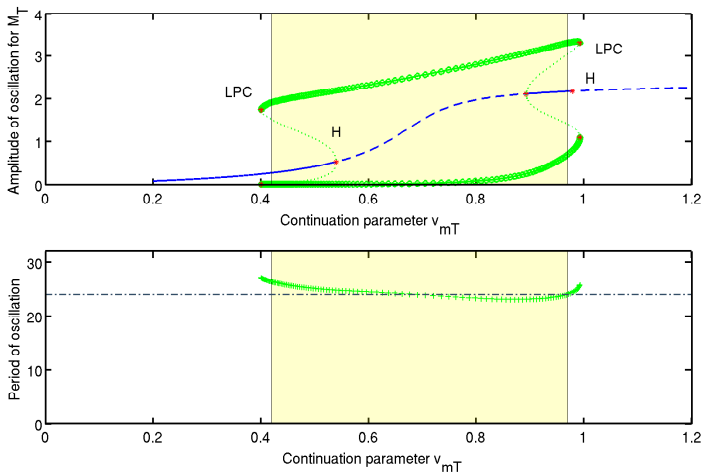


Stable **limit cycles** within range of parameter  $v_{mT}$

# Inferring robustness of oscillation period

- Intrinsic period of circadian rhythm robust to temperature variations
- Inverse problem: what is the underlying homeostatic mechanism?
- First test problem: **identify parameters** resulting in constancy of period under  $\approx 2$ -fold change in rate parameter  $v_{mT}$

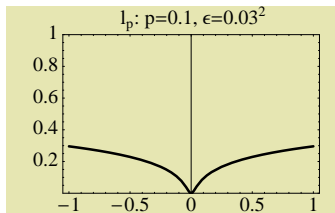
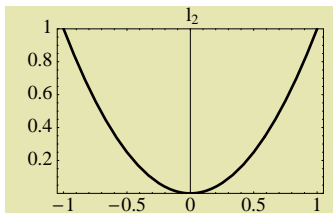
# Test case: constancy of period



Goal: eliminate variation of period within a **window of values** for  $v_{mT}$

# Objective function for identifying period constancy

- Right functional for measuring deviation of period: **total variation**?
- Large number (38) of parameters in the model
- Goal: identify **sparse** subset of parameters **impacting** the robustness of period
- Use  $l_p$ -functional,  $p \leq 1$ , for identifying sparse parameter set

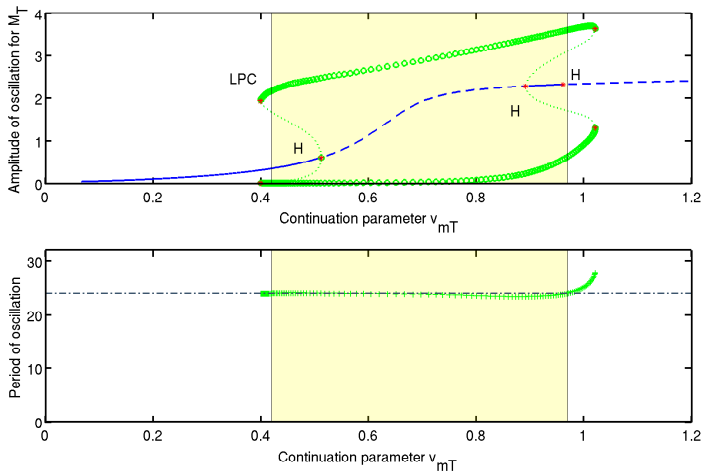


$$l_2(x) = \sum_i x_i^2, \quad \text{and} \quad l_p(x) = \sum_i (x_i^2 + \epsilon)^{p/2}$$

- Minimizing objective function:

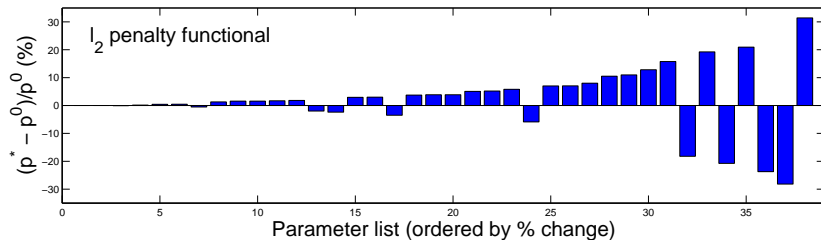
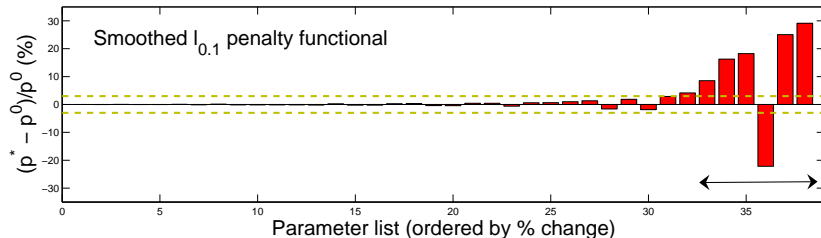
$$\begin{aligned} \min_p J(p) &= TV(\text{period} - 24hr) + \alpha \|p - p^0\|_{l_p} \\ \text{s.t.} \quad &LPC_{\text{left}} \leq p_{\text{left}}, LPC_{\text{right}} \geq p_{\text{right}} \end{aligned}$$

# Test case: result



Bifurcation diagram of system for computed system

# Effectiveness of $l_p$ functional for sparsity





# Conclusions

- Inverse bifurcation analysis: methodology for studying how qualitative and quantitative behavior arises
- Sparsity useful for identifying crucial parameters in high-dimensional systems
- Future directions:
  - **upscaling**: regularization methods, summarizing bifurcation diagrams in high dimensions
  - infer aspects of **network topology** from system dynamics