

Inferring Switched Nonlinear Dynamical Systems

推断切换型非线性动力系统

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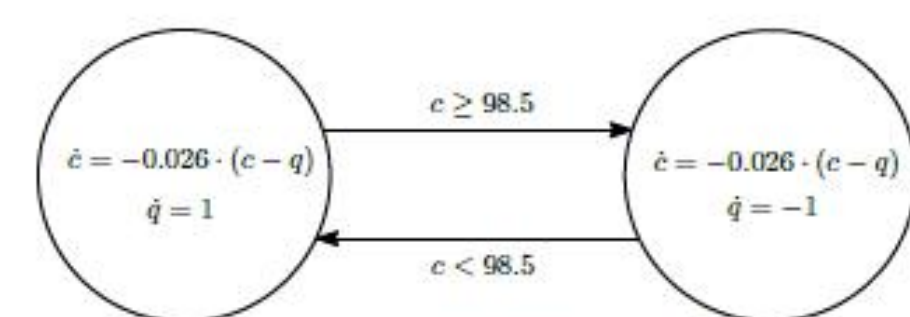
Formal Aspects of Computing

System Identification

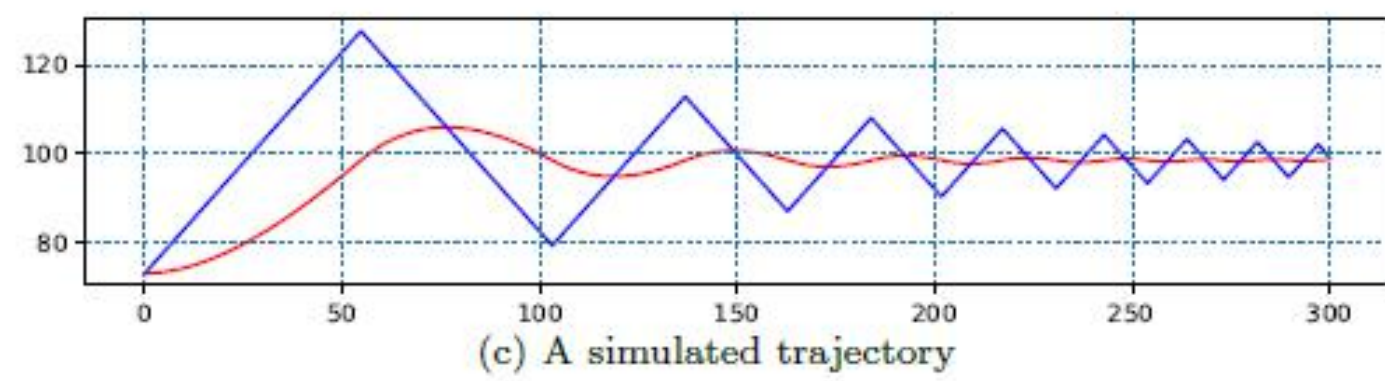
- Purpose:
Learning models for dynamical and hybrid systems from trajectories



(a) The Isolette system



(b) The corresponding SNDS modelled as a hybrid automaton



(c) A simulated trajectory

SNDS

Switched Nonlinear Dynamical Systems

$$\dot{x} = \begin{cases} f_1(x), & \text{if } x \in G_1(x) \\ f_2(x), & \text{if } x \in G_2(x) \\ \vdots \\ f_N(x), & \text{if } x \in G_N(x) \end{cases}$$

where f_i are polynomials and

$\forall i, j \in \{1, 2, \dots, N\}, i \neq j \rightarrow G_i \cap G_j = \emptyset$

trajectory: $x(t_0), x(t_1), x(t_2), \dots$ with $t_{i+1} - t_i = h$

Main Contributions

- A heuristic method to segment trajectories of a switched nonlinear dynamical system
- An inference procedure based on the segmented trajectories with a pruning search.
- An inference procedure by extending the identification methods of piecewise affine models.

Inferring Dynamics

- linear multistep method to estimate the derivatives on discrete points

$$f(x(t_n)) \approx \frac{1}{h} \left(\frac{137}{60} x(t_n) - \frac{300}{60} x(t_{n-1}) + \frac{300}{60} x(t_{n-2}) - \frac{200}{60} x(t_{n-3}) + \frac{75}{60} x(t_{n-4}) - \frac{12}{60} x(t_{n-5}) \right)$$

$$f(x(t_n)) \approx \frac{1}{h} \left(-\frac{137}{60} x(t_n) + \frac{300}{60} x(t_{n+1}) - \frac{300}{60} x(t_{n+2}) + \frac{200}{60} x(t_{n+3}) - \frac{75}{60} x(t_{n+4}) + \frac{12}{60} x(t_{n+5}) \right)$$

- linear regression to infer the coefficients of the dynamic polynomials

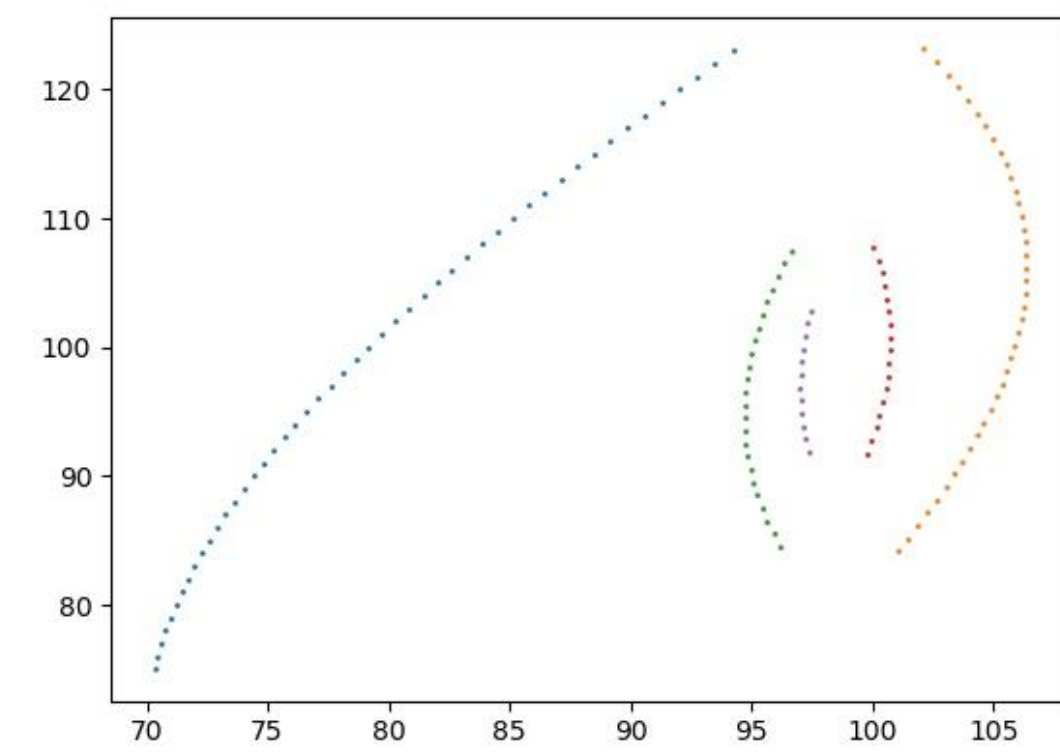
$$f(x_1, x_2) = \begin{bmatrix} a_{0,0} + a_{1,0}x_1 + a_{0,1}x_2 + a_{2,0}x_1^2 + a_{1,1}x_1x_2 + a_{0,2}x_2^2 + \dots \\ b_{0,0} + b_{1,0}x_1 + b_{0,1}x_2 + b_{2,0}x_1^2 + b_{1,1}x_1x_2 + b_{0,2}x_2^2 + \dots \end{bmatrix}^T$$

$$\begin{bmatrix} 1 & x_1(t_5) & x_2(t_5) & x_1(t_5)^2 & x_1(t_5)x_2(t_5) & x_2(t_5)^2 & \dots \\ 1 & x_1(t_6) & x_2(t_6) & x_1(t_6)^2 & x_1(t_6)x_2(t_6) & x_2(t_6)^2 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 1 & x_1(t_{N-1}) & x_2(t_{N-1}) & x_1(t_{N-1})^2 & x_1(t_{N-1})x_2(t_{N-1}) & x_2(t_{N-1})^2 & \dots \\ 1 & x_1(t_N) & x_2(t_N) & x_1(t_N)^2 & x_1(t_N)x_2(t_N) & x_2(t_N)^2 & \dots \end{bmatrix} \cdot \begin{bmatrix} a_{0,0} & b_{0,0} \\ a_{1,0} & b_{1,0} \\ a_{0,1} & b_{0,1} \\ a_{2,0} & b_{2,0} \\ a_{1,1} & b_{1,1} \\ a_{0,2} & b_{0,2} \\ \vdots & \vdots \end{bmatrix} = \begin{bmatrix} f(x_1(t_5), x_2(t_5)) \\ f(x_1(t_6), x_2(t_6)) \\ \vdots \\ f(x_1(t_{N-1}), x_2(t_{N-1})) \\ f(x_1(t_N), x_2(t_N)) \end{bmatrix}$$

$$X \cdot \theta = y$$

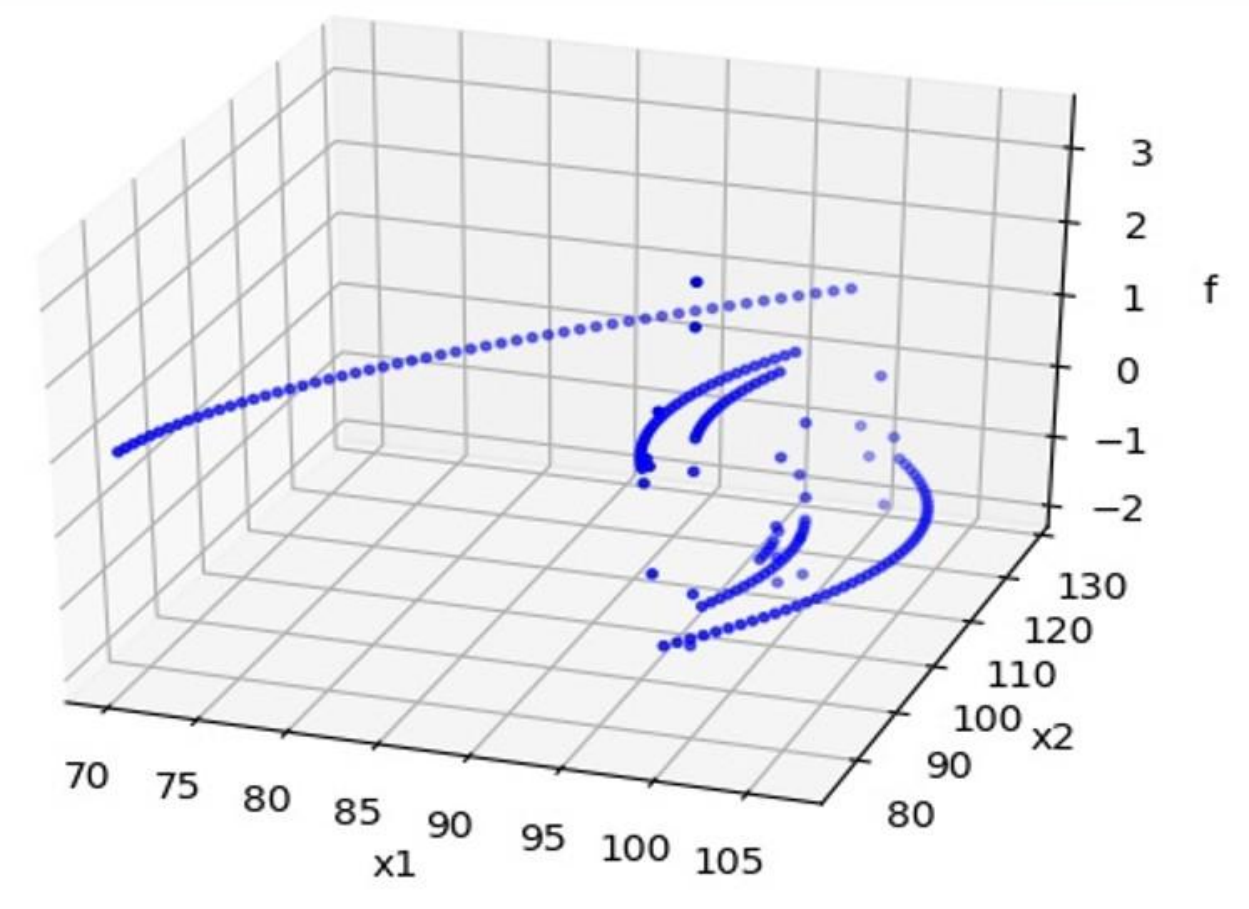
Segmenting and Searching

Segmenting the trajectory by comparing the relative difference between the two estimates of derivatives.



Pruning searching to find the best clustering and then infer dynamics.

Identification of piecewise affine models



Without prior segmentation, the estimated derivatives are captured by a piecewise affine model except the change points. The dynamics can be inferred by identification of this pwa model.

Experiments

- The Isolette example

Origin system

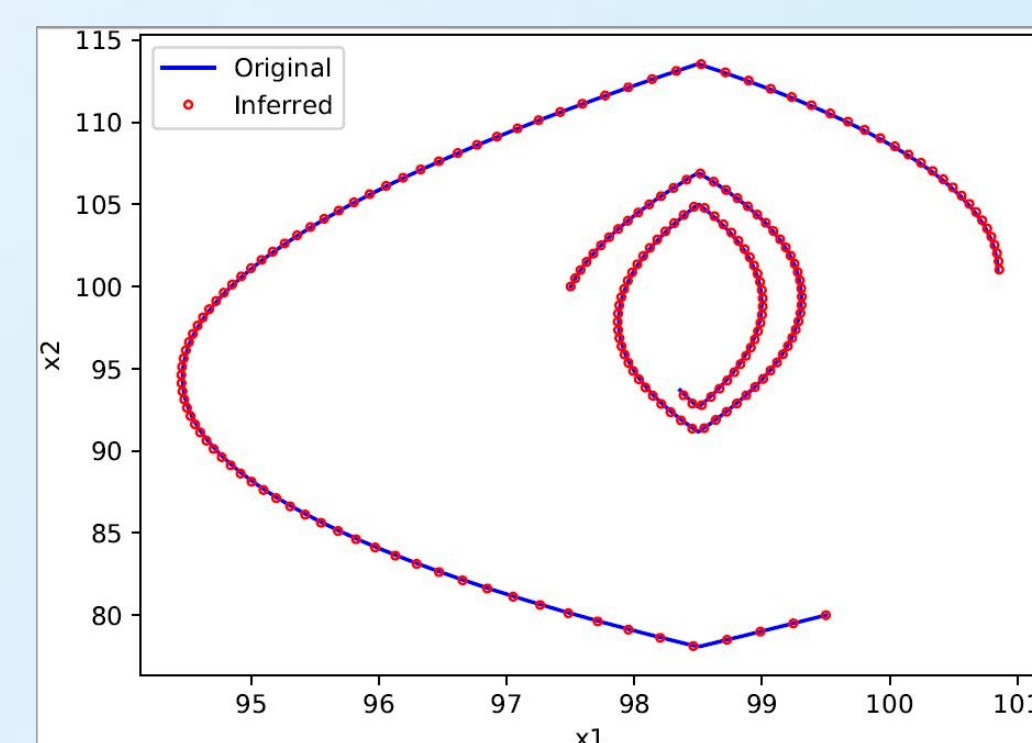
$$\begin{bmatrix} -0.026 & 0 \\ 0.026 & 0 \\ 0 & 1 \end{bmatrix} \quad \begin{bmatrix} -0.026 & 0 \\ 0.026 & 0 \\ 0 & -1 \end{bmatrix}$$

$x_1 = 98.5$

Inferred system

$$\begin{bmatrix} -2.60000 \times 10^{-2} & -7.93683 \times 10^{-14} \\ 2.60000 \times 10^{-2} & -2.58099 \times 10^{-14} \\ -6.58505 \times 10^{-13} & 1.00000 \end{bmatrix} \quad \begin{bmatrix} -2.60056 \times 10^{-2} & -1.08319 \times 10^{-4} \\ 2.60007 \times 10^{-2} & 1.42407 \times 10^{-5} \\ 4.86417 \times 10^{-4} & -9.90636 \times 10^{-1} \end{bmatrix}$$

$x_1 - 0.00077x_2 = 98.42313$



- Other examples

