

# Manifold Construction and Parameterization for Nonlinear Manifold-Based Model Reduction

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# Outline

- **Background**

- Introduction to MOR and maniMOR
- Manifold construction and parameterization

- **Manifold construction using integral curves**

- DC manifold and the normalized integral curve equation
- Ideal and almost-ideal manifold
- Algorithm

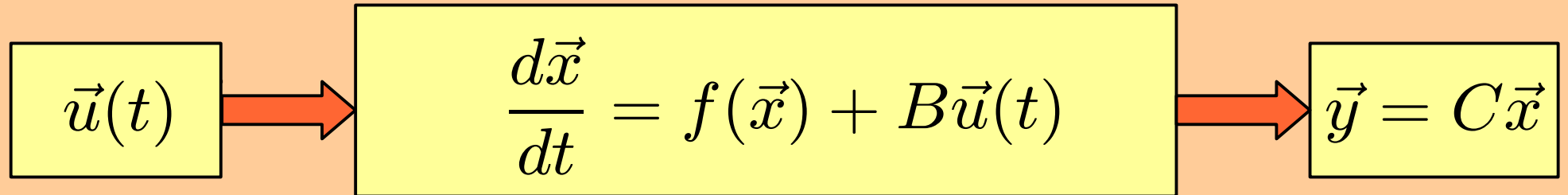
- **Experimental results**

- **Conclusion**

# Background

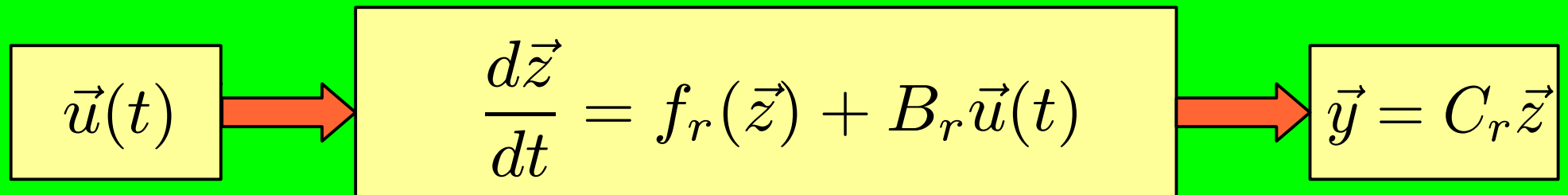
# Model Order Reduction

Original system (size  $n$ )



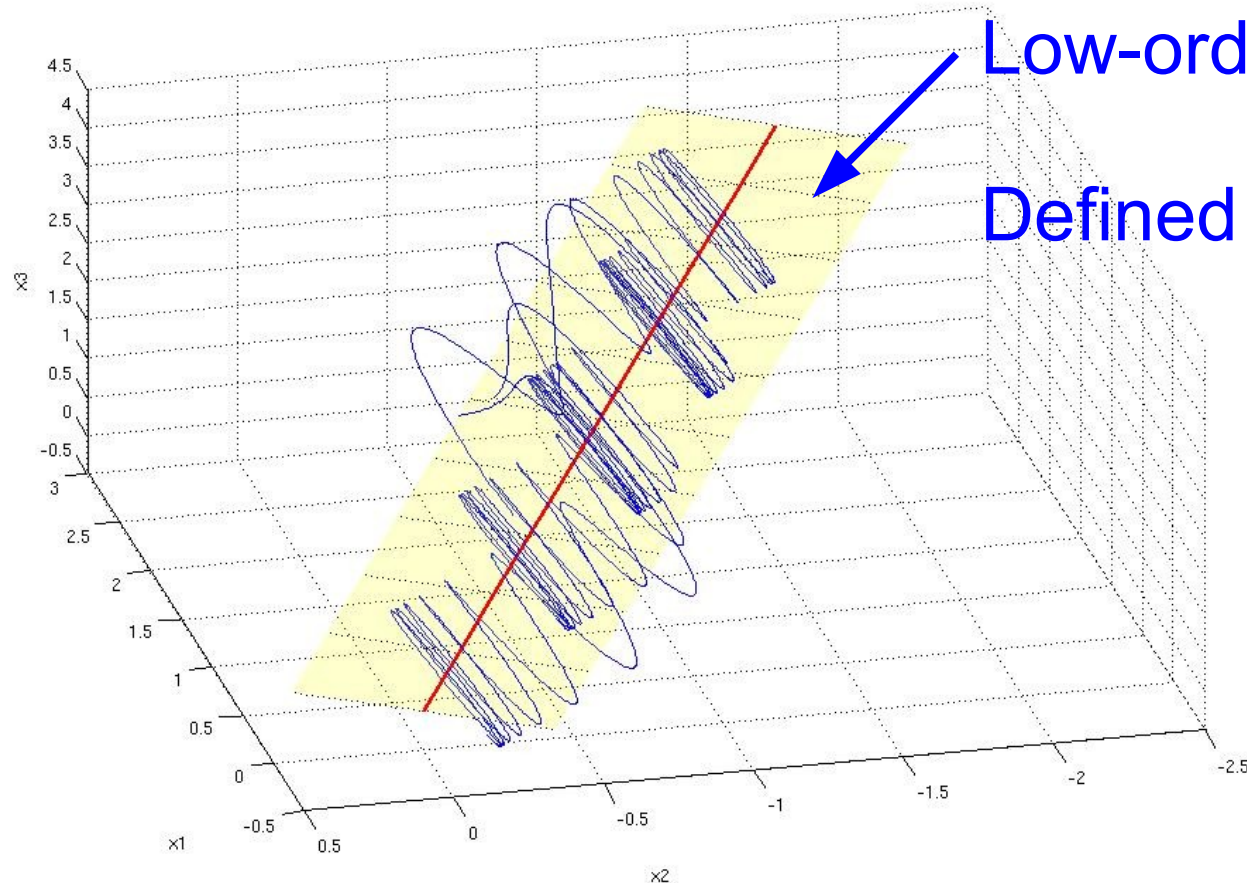
$$v : \vec{x} \mapsto \vec{z}, \quad \vec{x} \in \mathbb{R}^n, \vec{z} \in \mathbb{R}^q, \quad q \ll n$$

Reduced system (size  $q$ )



# Low-order Linear Subspace

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -10 & 1 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} u(t)$$

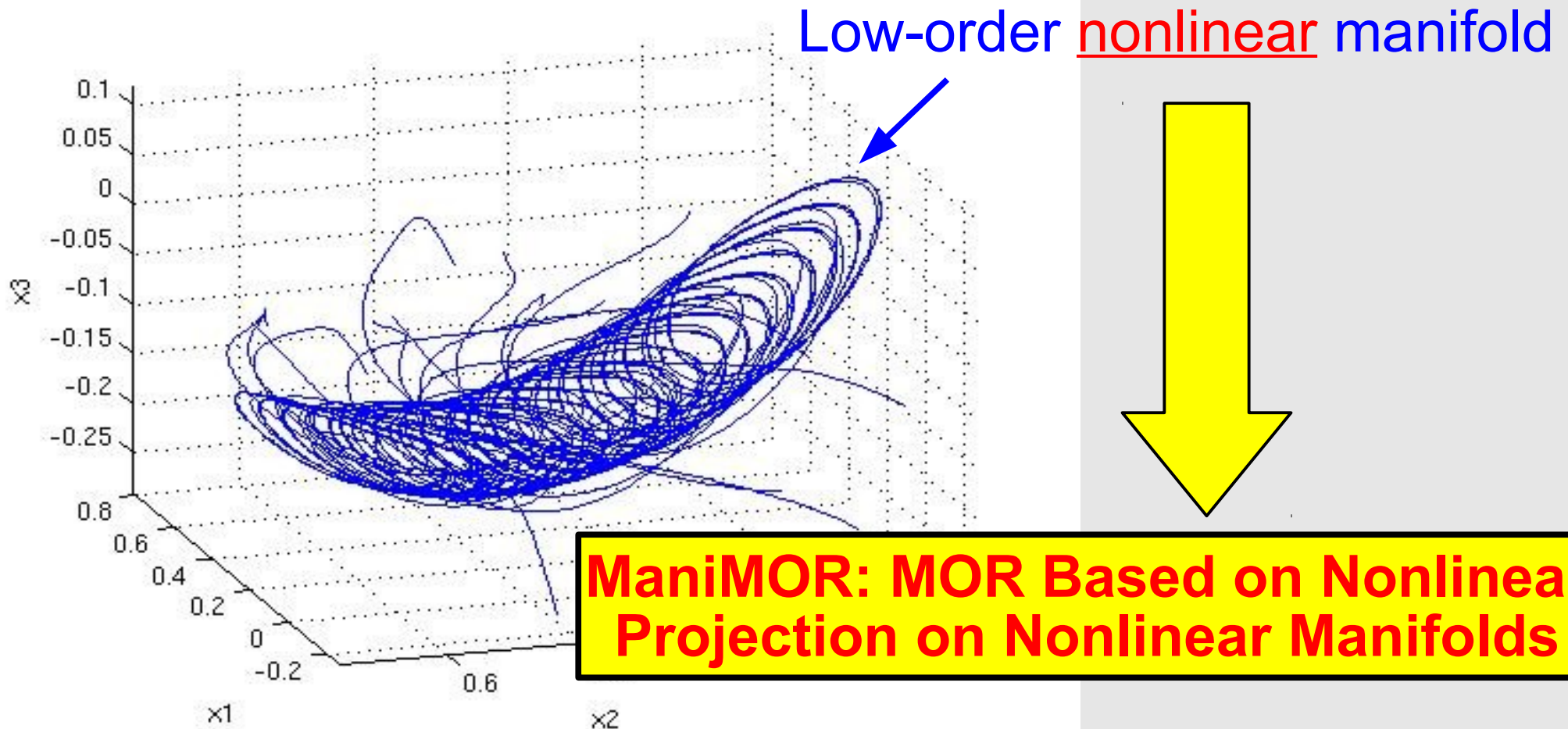


Low-order linear subspace

Defined by  $x = Vz$

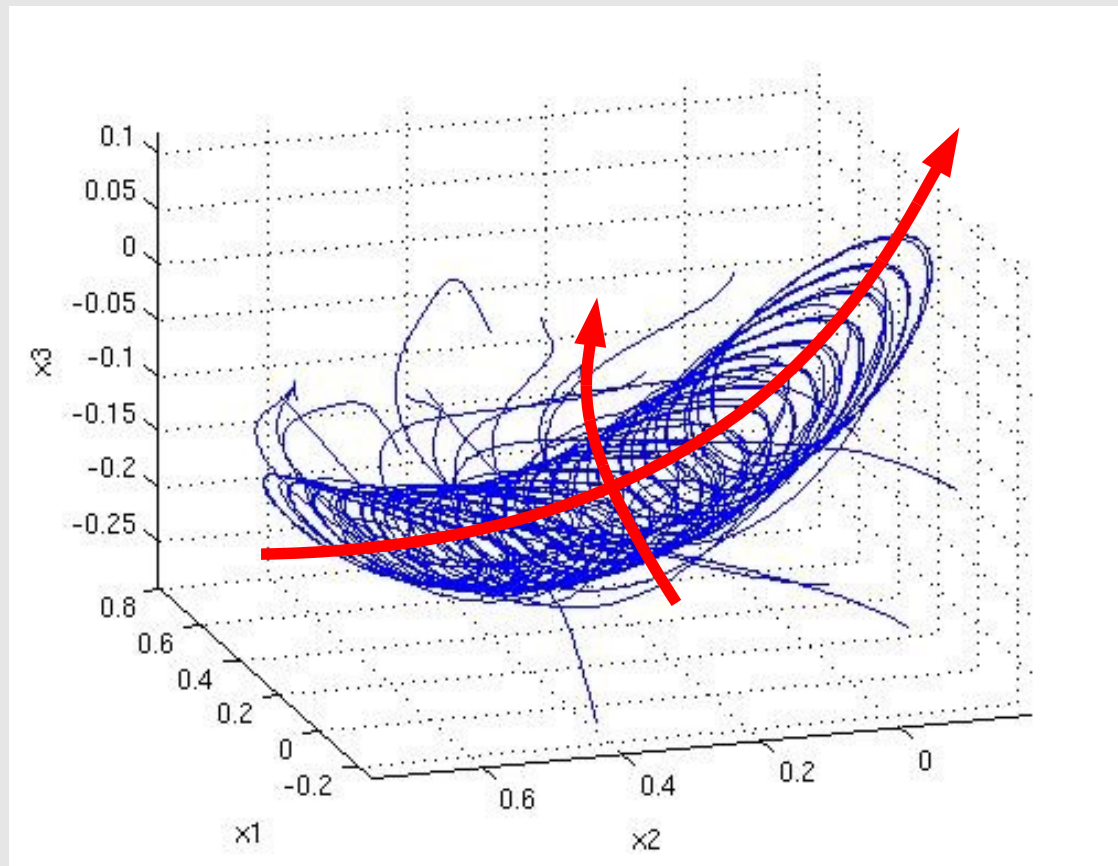
# Low-order Nonlinear Manifold

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -10 & 1 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ x_1^2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} u(t)$$

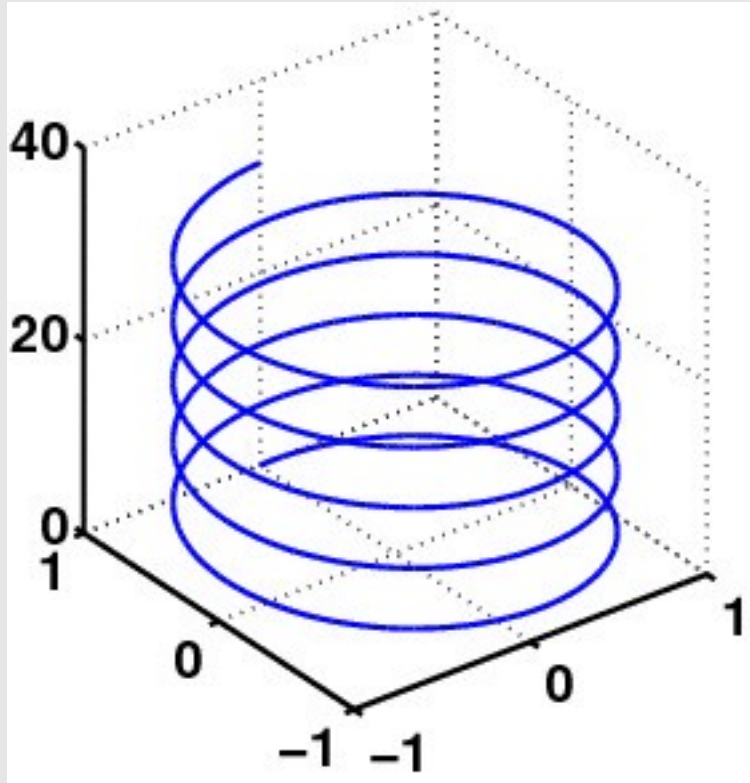


# Key Steps in ManiMOR

- **“Find” the nonlinear manifold**
  - Capture important dynamics
- **“Parameterize” the manifold**
  - Build up the coordinate system



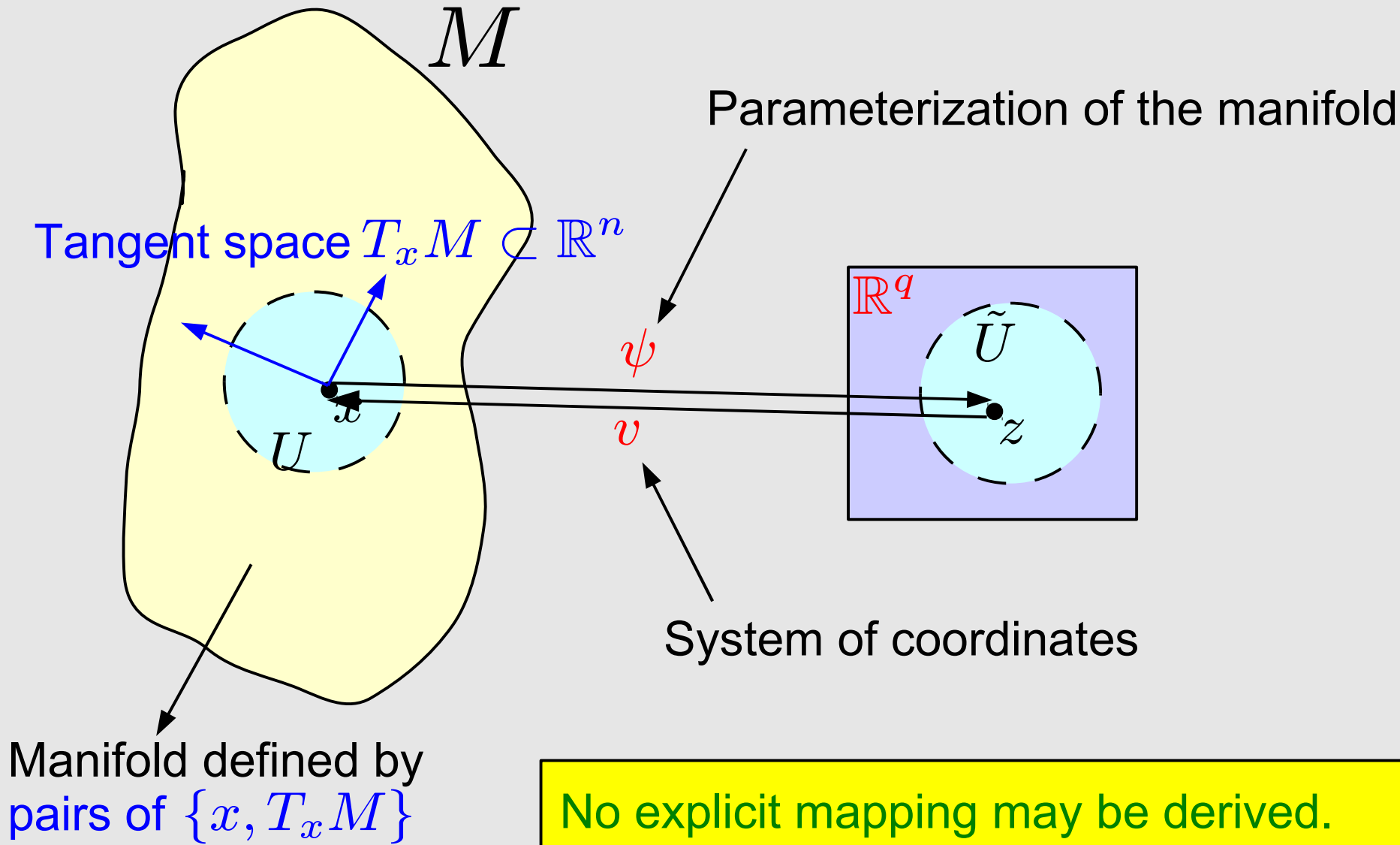
# Manifold and Its Parameterization



$$\begin{cases} x = \cos(t) \\ y = \sin(t) \\ z = t \end{cases}$$



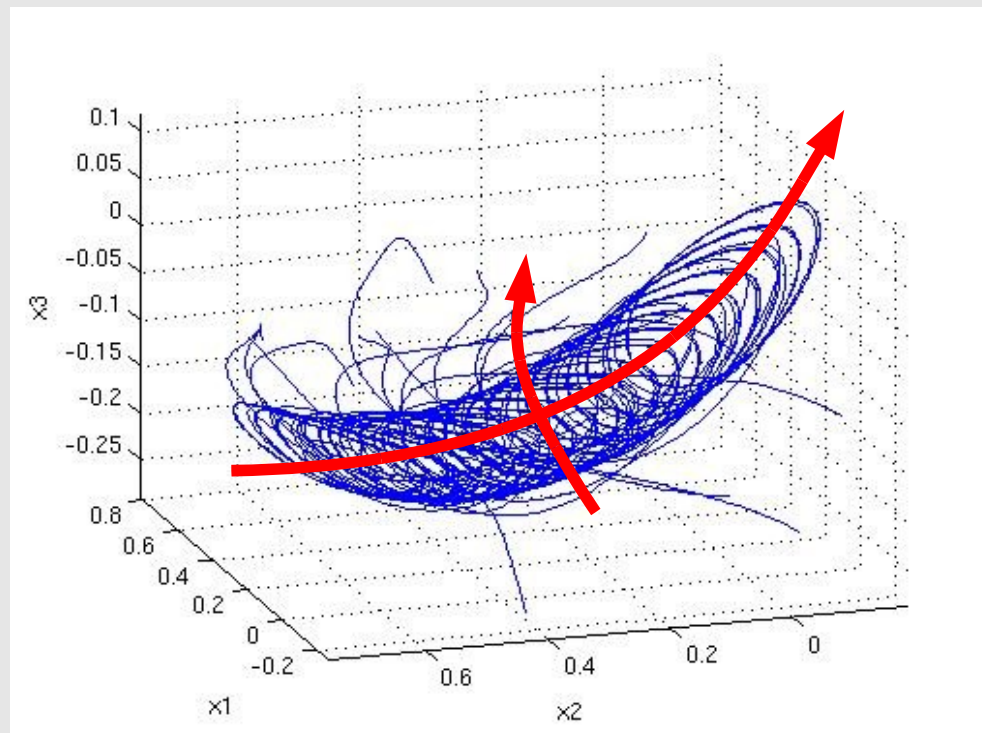
# Manifold and Its Parameterization



No explicit mapping may be derived.  
Instead, use **piecewise linear** approximation.

# Manifold and Its Parameterization

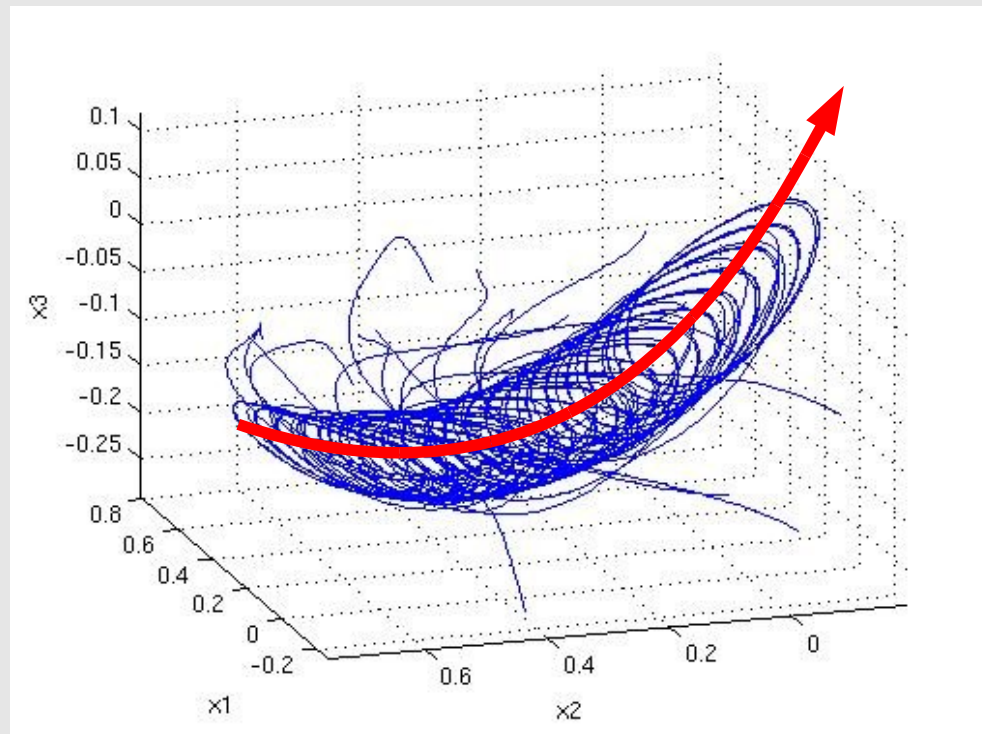
1. Identify the manifold that capture important dynamics
2. Compute and store pairs of  $\{x, T_x M\} / \{z, T_z M\}$



# DC Manifold

DC operating points constitute a DC manifold.

$$\frac{d\vec{x}}{dt} = f(\vec{x}) + B\vec{u}(t) = 0$$



How to compute and parameterize the DC manifold?

# DC Manifold

$$f(\vec{x}) + B\vec{u}(t) = 0$$

## A straight-forward solution:

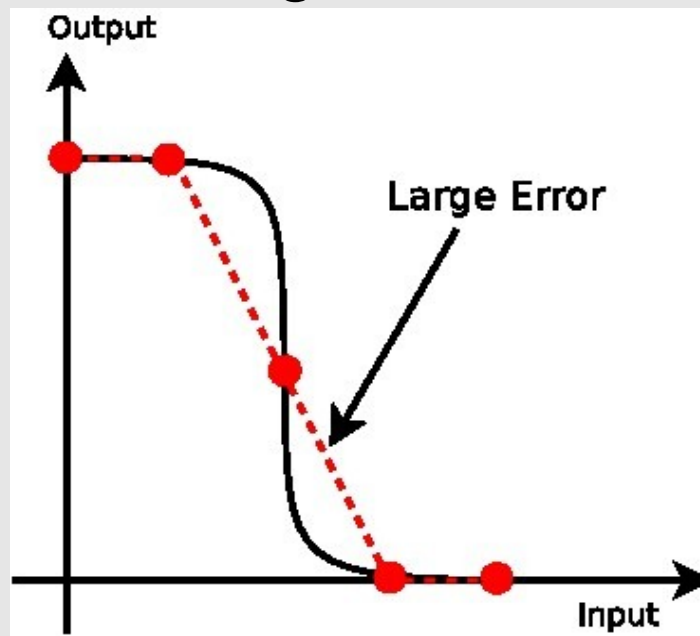
Computation: Perform DC sweep analysis

Parameterization: Define  $z$  coordinates using values of  $u$

## Problems:

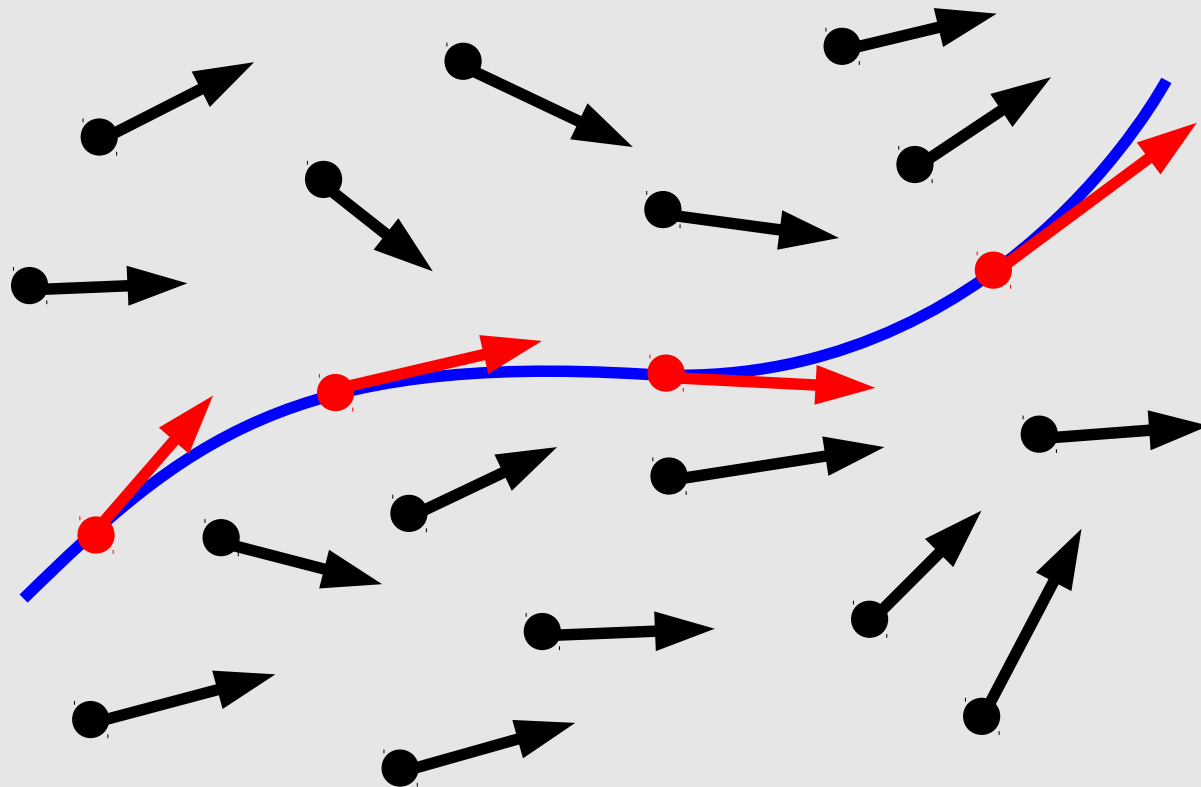
Hard to choose step size in DC sweep analysis

Not generalizable to higher dimensions



# Introduction to Integral Curve

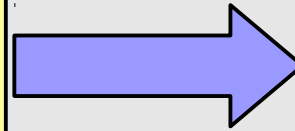
Given a vector field  $v(x)$  ,  
its **integral curve** is the curve  $\gamma \equiv x(t)$  such that  $\frac{dx}{dt} = v(x)$



# DC Manifold as an Integral Curve

Need to derive the relationship between  $dx$  and  $du$

$$f(\vec{x}) + B\vec{u}(t) = 0$$



$$\frac{\partial f}{\partial x} \frac{dx}{du} + B = 0$$

$$\frac{dx}{du} = -[G(x)]^{-1} B$$

The first  
Krylov basis.

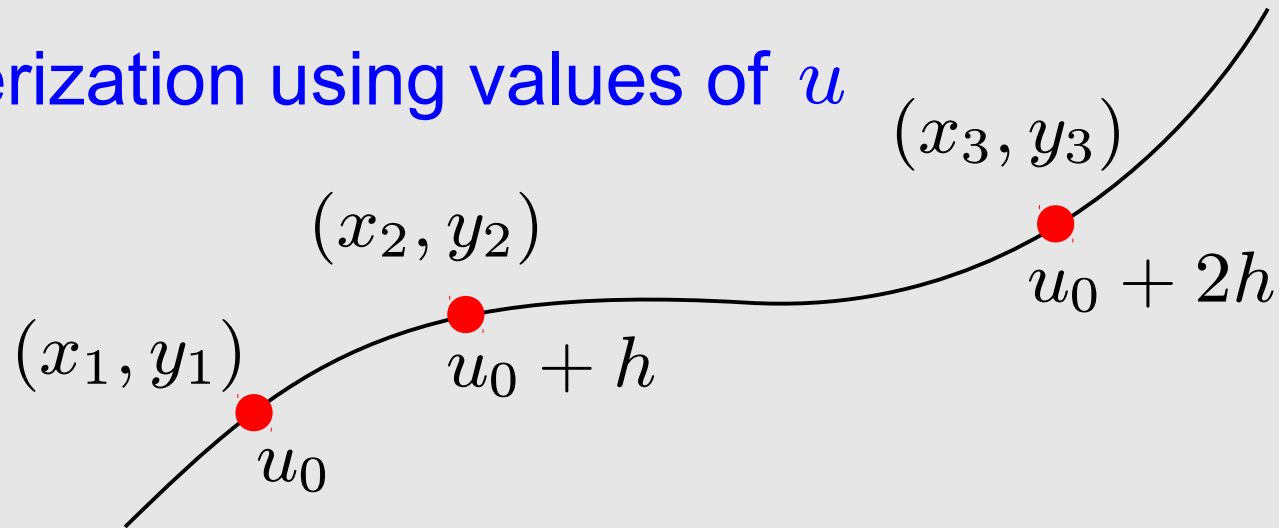
Initial condition:  $x(u = 0) = x_{DC} |_{u=0}$

Solutions are DC operating points.

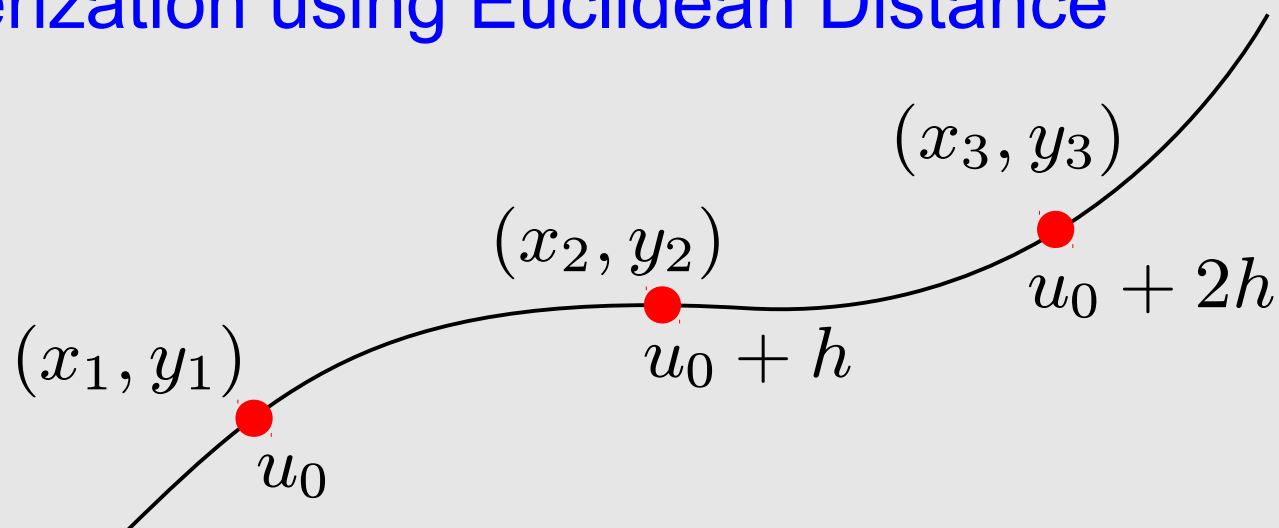
Any numerical integration / transient analysis code can be applied.

# Parameterization using Euclidean Distance

Parameterization using values of  $u$

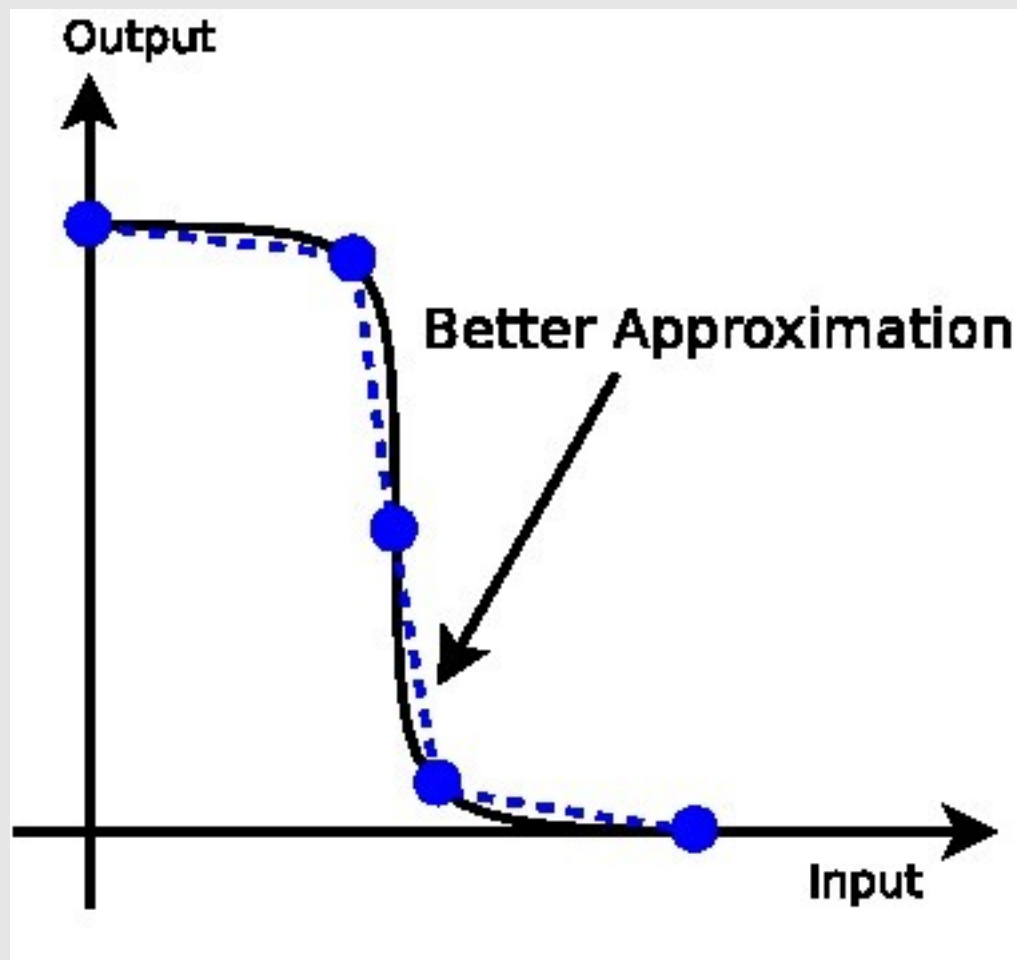
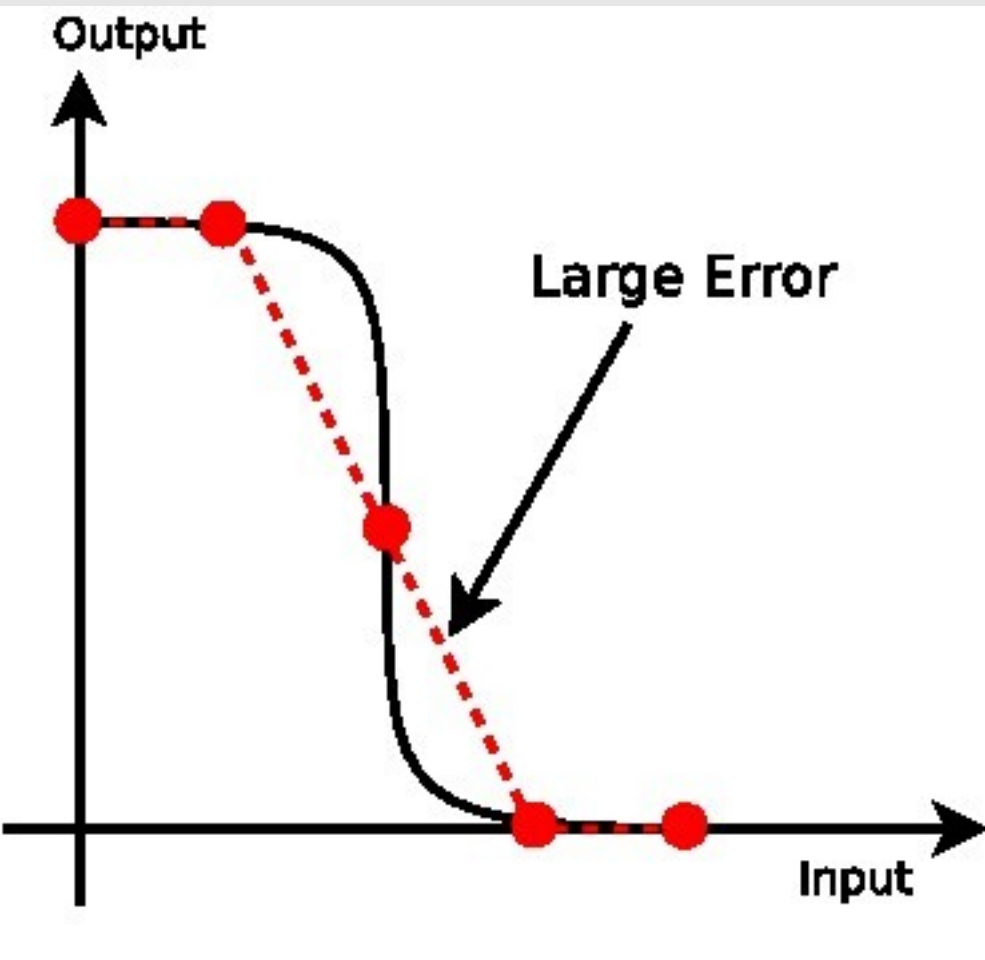


Parameterization using Euclidean Distance



Sample points equally spaced on the DC manifold

# Parameterization using Euclidean Distance





# Normalized Integral Curve Equation

Local Euclidean distance is

$$\|dx\|_2 = |du|$$

$$\frac{dx}{du} = -[G(x)]^{-1} B$$

$$\left\| \frac{dx}{du} \right\|_2 = \|[G(x)]^{-1} B\|_2 = 1$$

Generally not satisfied

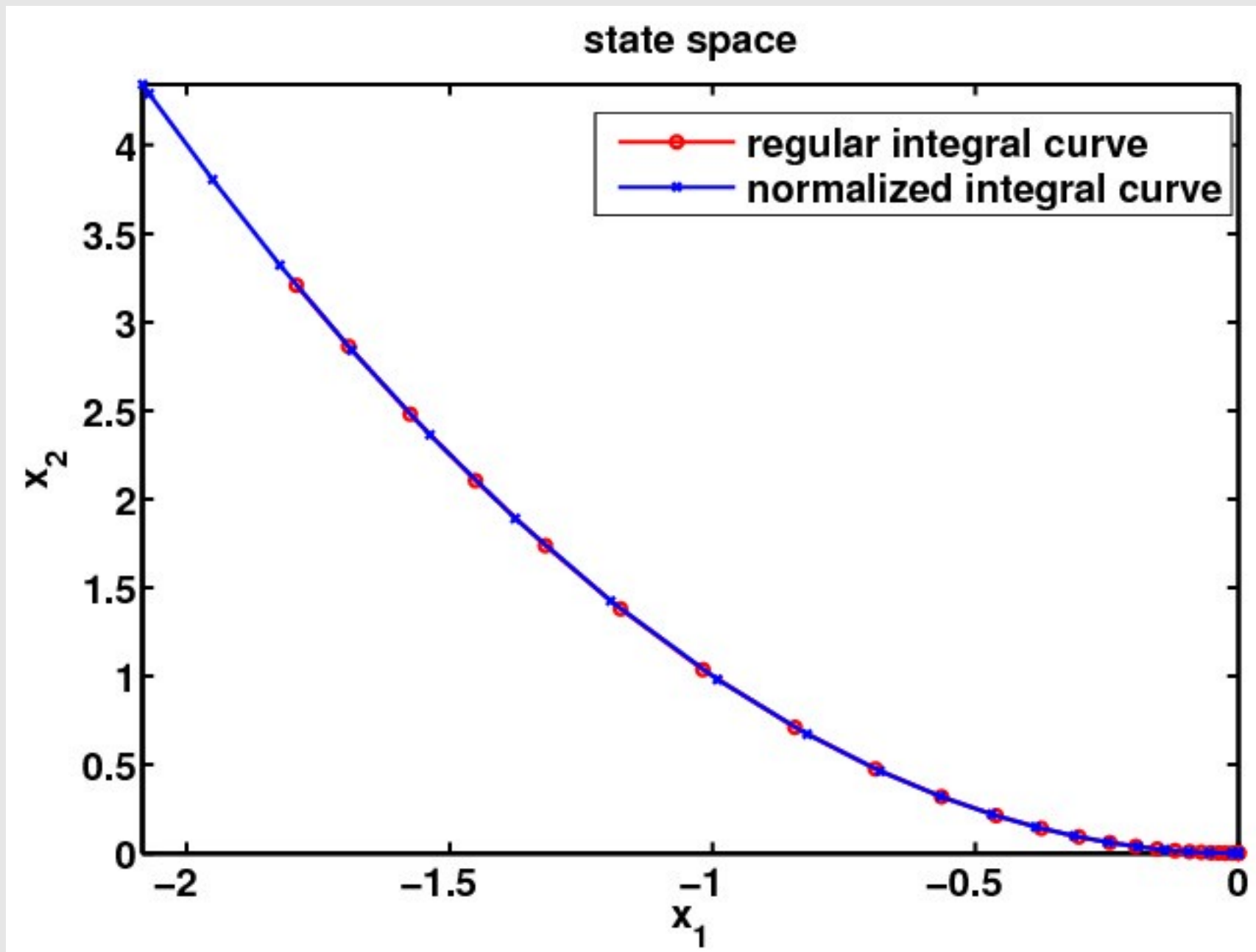
Normalize RHS

$$\frac{dx}{du} = \frac{[G(x)]^{-1} B}{\|[G(x)]^{-1} B\|_2}$$

Normalized Integral Curve Equation

Does it define the same integral curve?

# Validation



# Normalized Integral Curve Equation

## Theorem:

Suppose  $t = \sigma(\tau)$ ;  $x(t)$  and  $\hat{x}(\tau)$  satisfy

$$\frac{d}{dt}x(t) = g(x(t)) \quad \text{and} \quad \frac{d}{d\tau}\hat{x}(\tau) = \sigma'(\tau)g(\hat{x}(\tau)), \quad \text{respectively.}$$

Then  $x(t)$  and  $\hat{x}(\tau)$  span the same state space.

## Sketch of proof:

Since  $t = \sigma(\tau)$ , we have  $dt = \sigma'(\tau)d\tau$ .

Define  $\hat{x}(\tau) \equiv x(t) = \hat{x}(\sigma(t))$ , then

$$\frac{d}{d\tau}\hat{x}(\tau) = \frac{d\hat{x}(\tau)}{dt} \frac{dt}{d\tau} = \sigma'(\tau)g(x(t)) = \sigma'(\tau)g(\hat{x}(\tau))$$

# Normalized Integral Curve Equation

$$\frac{dx}{du} = -[G(x)]^{-1} B$$

$$\frac{dx}{du} = \frac{[G(x)]^{-1} B}{\| [G(x)]^{-1} B \|_2}$$

Solution:  $x(u)$

Solution:  $\hat{x}(\hat{u})$

Define  $u = \sigma(\hat{u}) = \int_0^{\hat{u}} \frac{1}{\| [G(\hat{x}(\mu))]^{-1} B \|_2} d\mu$

From the theorem,

$x(u)$  and  $\hat{x}(\hat{u})$  define the same integral curve.

# Normalized Integral Curve Equation

$$\frac{dx}{du} = \frac{[G(x)]^{-1} B}{\| [G(x)]^{-1} B \|_2}$$

← The first normalized Krylov basis.

Directly available from Krylov subspace methods.

Generalizable to higher dimensions.

# Ideal Nonlinear Manifold

$$\frac{\partial x}{\partial z_1} = v_1(x), \quad \frac{\partial x}{\partial z_2} = v_2(x), \quad \dots, \quad \frac{\partial x}{\partial z_q} = v_q(x).$$

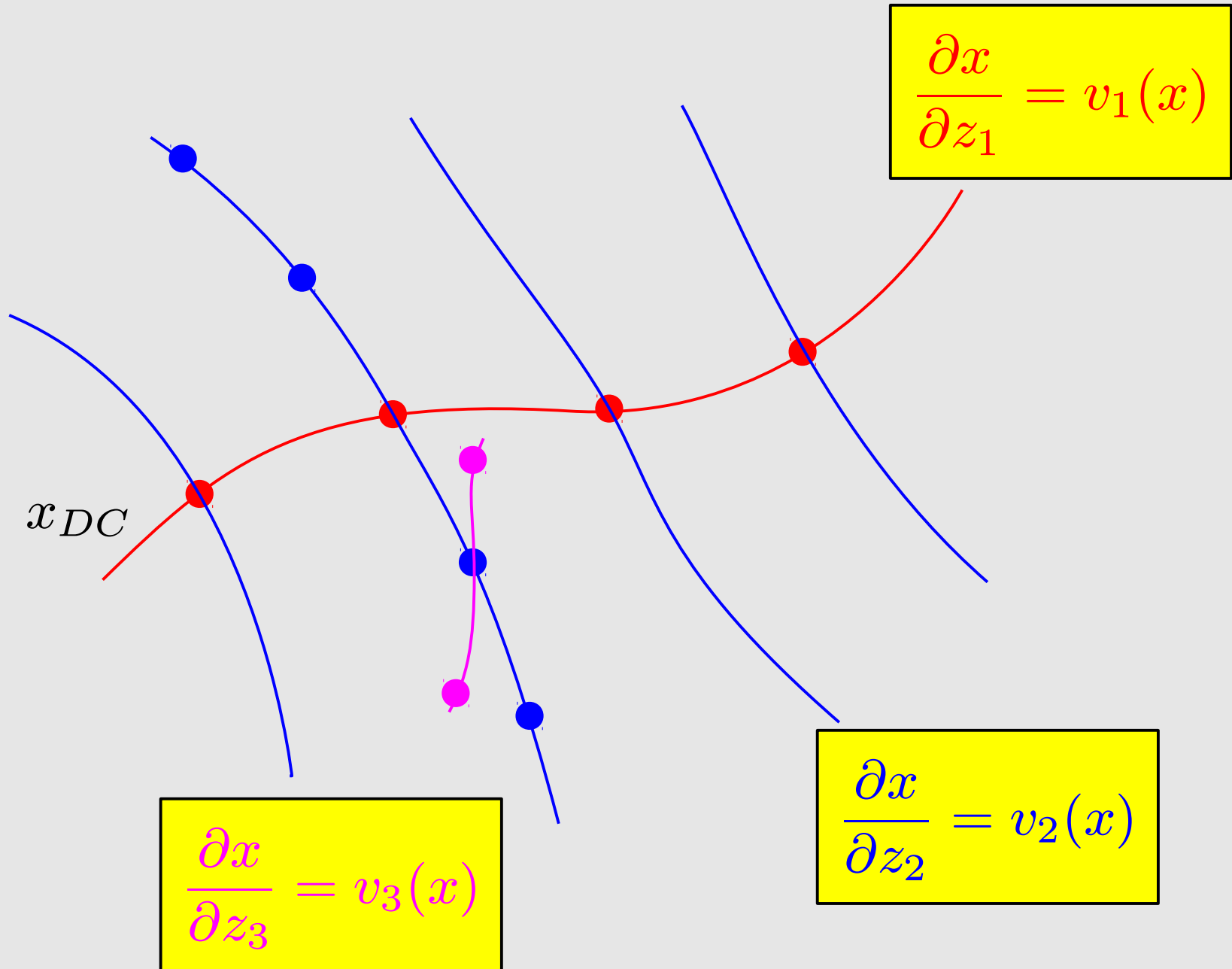
$V(x) = [v_1(x), \dots, v_q(x)]$  is the projection matrix for the reduced linearized system (at  $x$ ).

For example, Arnoldi algorithm generates a basis for

$$\mathcal{K}_q([G(x)]^{-1}, B) = \{[G(x)]^{-1}B, [G(x)]^{-2}B, \dots, [G(x)]^{-q}B\}$$

However, this set of PDEs is over-determined.

# Almost-Ideal Manifold Construction



# Almost-Ideal Manifold Construction

## Algorithm 1 Manifold Construction by Finding Integral Curves

- 1: Given the region to be parameterized  $(z_{i,min}, z_{i,max}), i \in [1, q]$ ;
- 2: Let  $x_0(0, \dots, 0) = x_{DC}$ , where  $x_{DC}$  is the DC solution when  $u = 0$ ;
- 3:  $X \leftarrow \{x_0\}, Z \leftarrow (0, \dots, 0)$ ;
- 4: **for**  $i = 1$  to  $q$  **do**
- 5:     **for all**  $x \in X$  **do**
- 6:         Integrate the integral curve equation
$$\frac{\partial x}{\partial z_i} = v_i(x)$$

with initial condition  $x$ ;
- 7:          $X \leftarrow \{x(z)\}, Z \leftarrow z$ ;
- 8:     **end for**
- 9: **end for**
- 10: Output  $X$  as the set of points on the manifold;
- 11: Output  $Z$  as the parameterization of the manifold for each point  $x \in X$ .



# Experimental Results

# A Hand-Calculable Example

$$\frac{d}{dt}x_1 = -x_1 + x_2 - u(t)$$

$$\frac{d}{dt}x_2 = x_1^2 - x_2$$

$$f(x) = \begin{bmatrix} -x_1 + x_2 \\ x_1^2 - x_2 \end{bmatrix}, \quad B = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

$$G(x) = \begin{bmatrix} -1 & 1 \\ 2x_1 & -1 \end{bmatrix}, \quad [G(x)]^{-1} = \frac{1}{2x_1 - 1} \begin{bmatrix} 1 & 1 \\ 2x_1 & 1 \end{bmatrix}$$

# DC and AC Manifold

$$[G(x)]^{-1} = \frac{1}{2x_1 - 1} \begin{bmatrix} 1 & 1 \\ 2x_1 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} -1 \\ 0 \end{bmatrix}$$

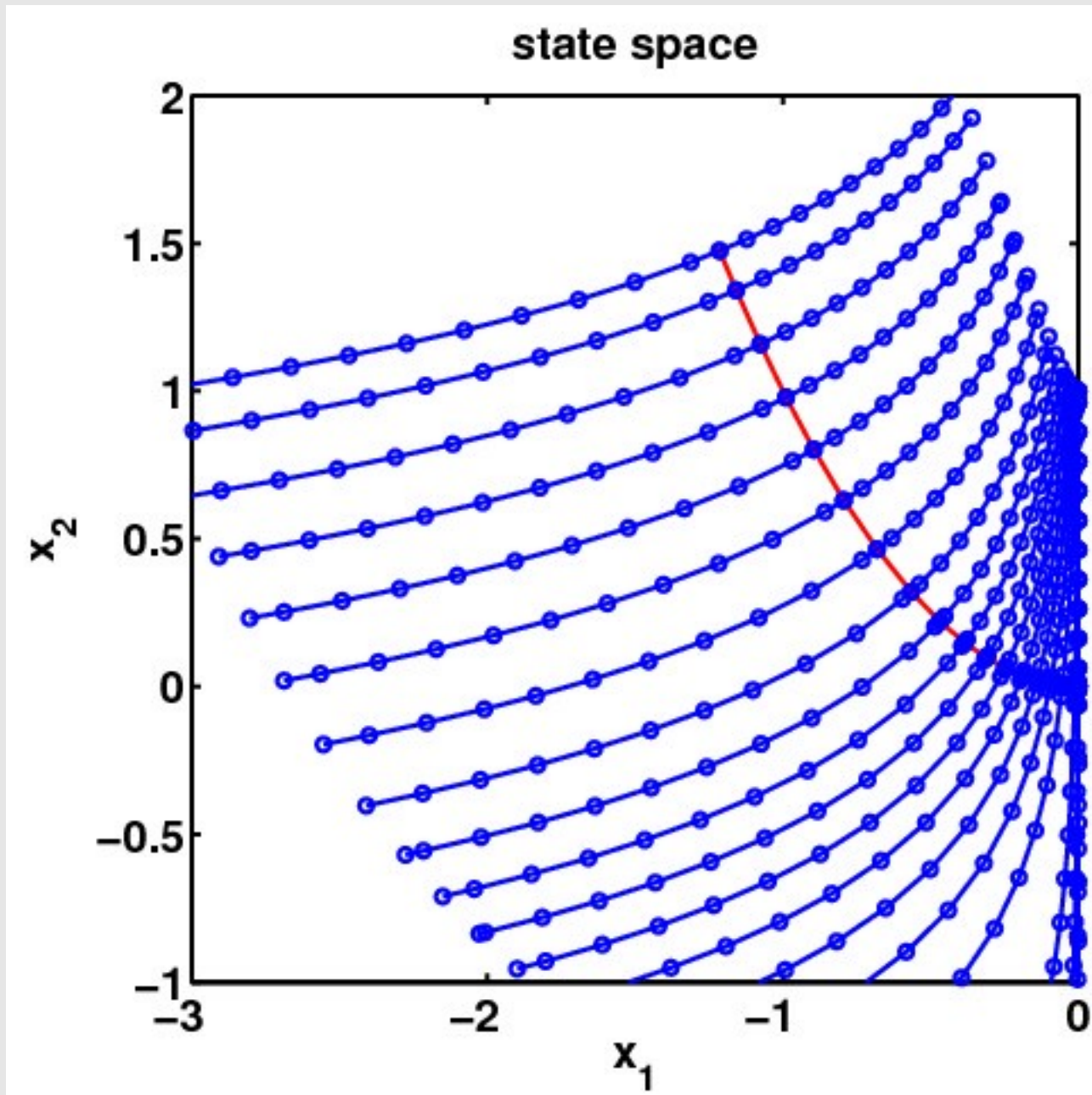
$$w_1(x) = [G(x)]^{-1} B = \frac{1}{2x_1 - 1} \begin{bmatrix} 1 \\ 2x_1 \end{bmatrix}$$

$$w_2(x) = [G(x)]^{-2} B = \frac{1}{(2x_1 - 1)^2} \begin{bmatrix} -1 - 2x_1 \\ -4x_1 \end{bmatrix}$$

DC manifold:  $\frac{\partial x}{\partial z_1} = v_1(x) = \frac{w_1(x)}{\|w_1(x)\|_2}$

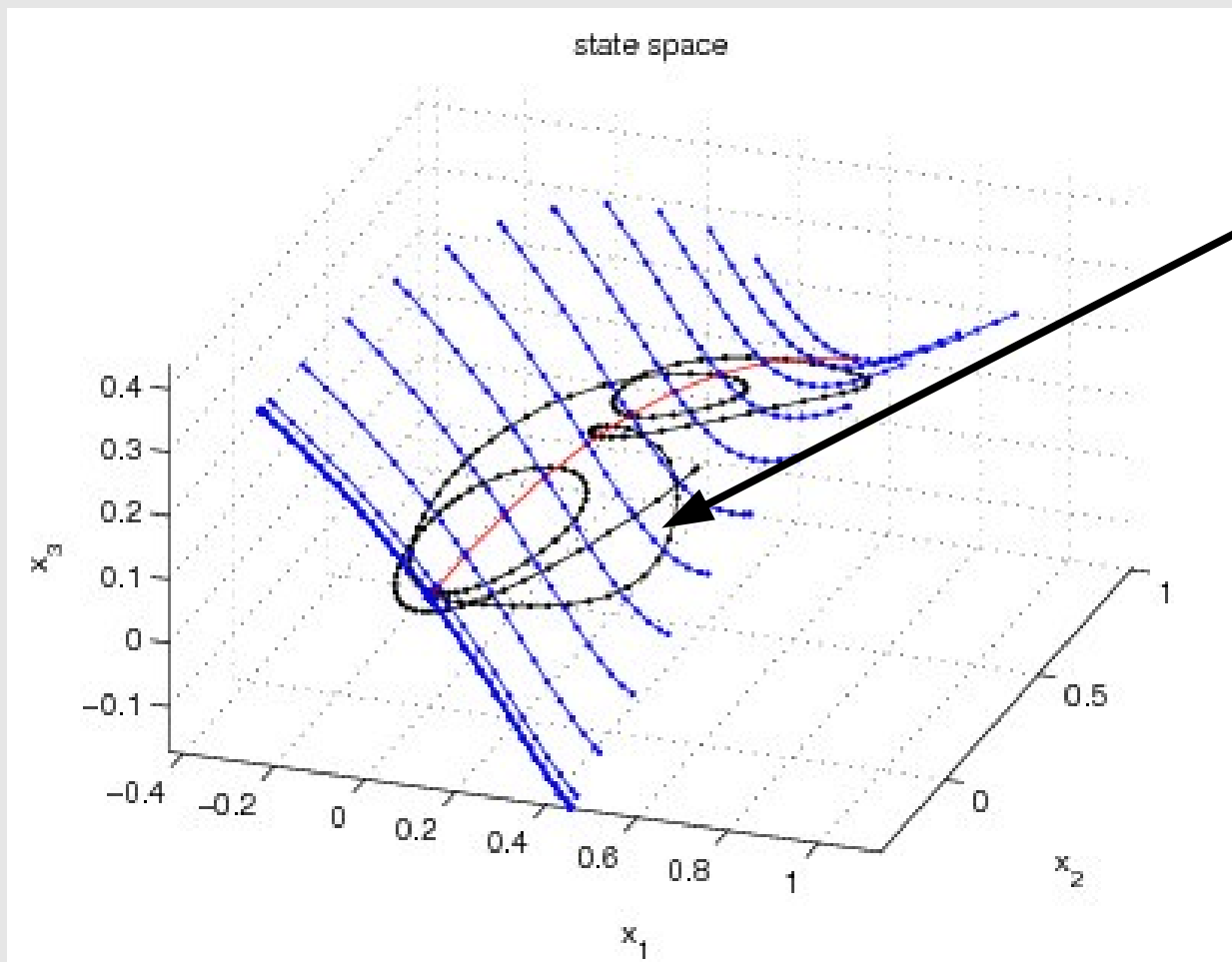
AC manifold:  $\frac{\partial x}{\partial z_1} = v_2(x) = \frac{w_2 - \langle w_2, v_1 \rangle v_1}{\|w_2 - \langle w_2, v_1 \rangle v_1\|_2}$

# DC and AC Manifold



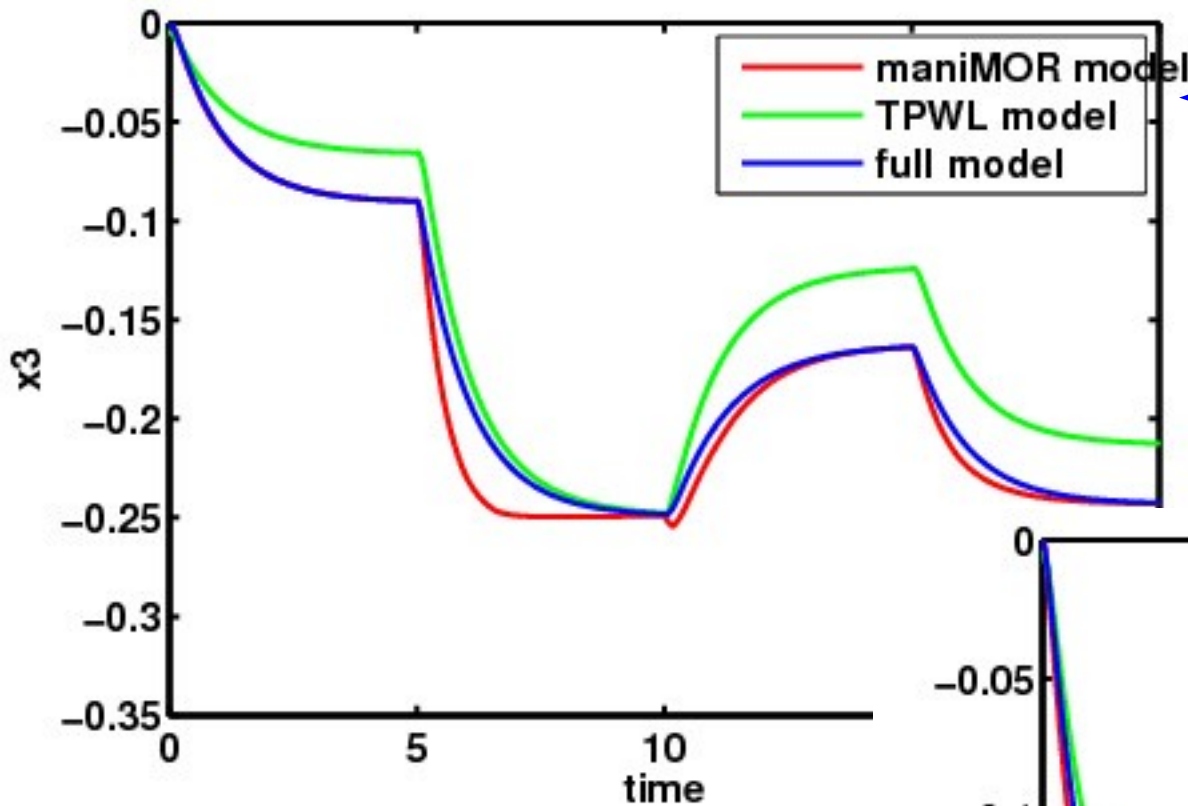
# Application to MOR

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -10 & 1 & 1 \\ 1 & -1 & 0 \\ 1 & 0 & -1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ x_1^2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} u(t)$$



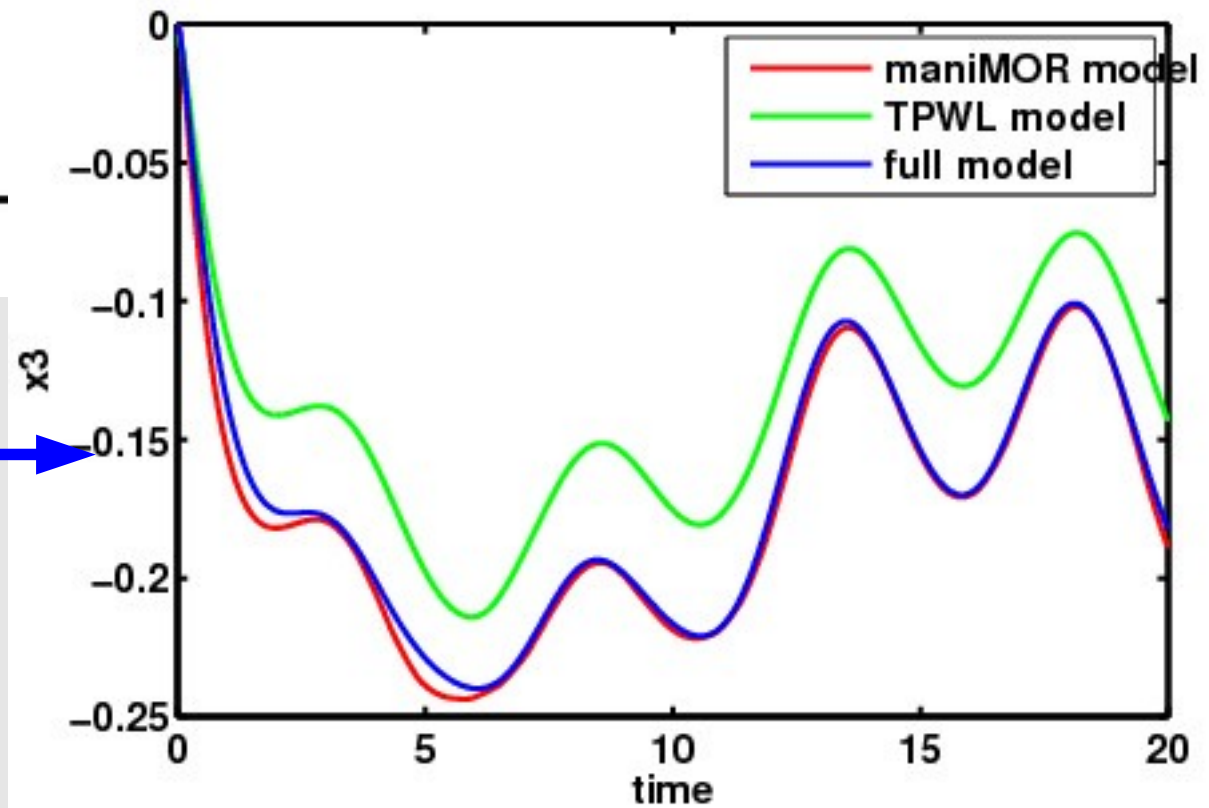
Trajectory of  
the full system  
stays close  
to the manifold

# Simulation of the Reduced Order Model



Response to a step input

Response to a sinusoidal input



# Conclusion

- **Presented a manifold construction and parameterization procedure**
  - Based on computing integral curves
  - Preserves local distance
  - Captures important system responses
    - Such as DC and AC responses
- **Application to manifold-based MOR**
  - Validated against several examples