

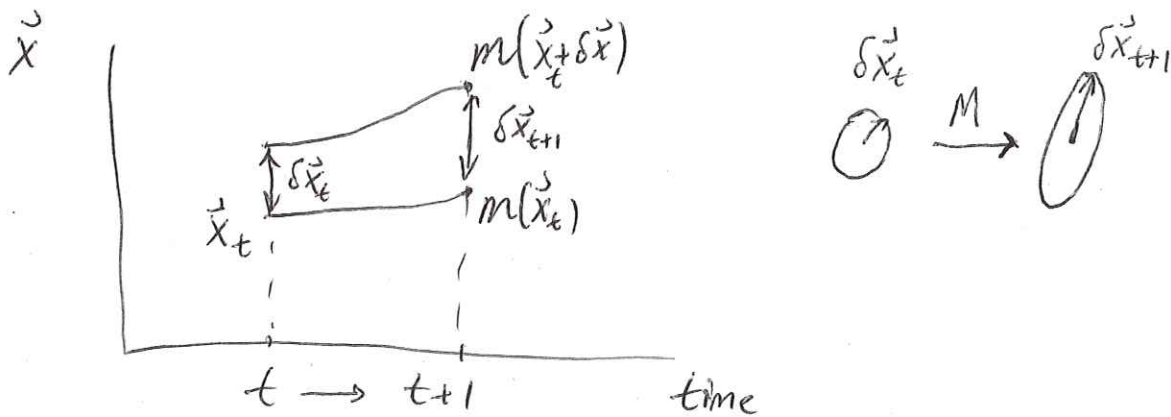
## Tangent Linear Model

A dynamic model (dynamical system) has its own error growth mechanism and predictability. By studying a linearized version of the system, we can characterize how errors evolve over time.

Let the nonlinear dynamical system be denoted as

$$\vec{x}_{t+1} = m(\vec{x}_t) = \vec{x}_t + \Delta t f(\vec{x}_t) \quad (\text{Euler forward}).$$

when there is a small deviation  $\delta \vec{x}_t$  at time  $t$  the model trajectory from  $\vec{x}_t + \delta \vec{x}_t$  is different from that starting from  $\vec{x}_t$ :



The difference of state at time  $t+1$  can be approximated

$$m(\vec{x}_t + \delta \vec{x}_t) = m(\vec{x}_t) + \left. \frac{\partial m}{\partial \vec{x}} \right|_{\vec{x}_t} \delta \vec{x}_t + \dots$$

$$\delta \vec{x}_{t+1} = m(\vec{x}_t + \delta \vec{x}_t) - m(\vec{x}_t) \cong \left. \frac{\partial m}{\partial \vec{x}} \right|_{\vec{x}_t} \delta \vec{x}_t, \quad M \equiv \left. \frac{\partial m}{\partial \vec{x}} \right|_{\vec{x}_t} = \nabla_{\vec{x}} m(\vec{x}_t)$$

-  $M$  propagates a perturbation forward in time.

$M$  is called "Tangent Linear" model (TLM)

it is also the Jacobian matrix  $M_{ij} = \frac{\partial m_i}{\partial x_j}$

Example 1: find the TLM for L63:

$$\frac{d}{dt} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} -\sigma x_1 + \sigma x_2 \\ \rho x_1 - x_2 - x_1 x_3 \\ x_1 x_2 - \beta x_3 \end{pmatrix} \quad \frac{d\vec{x}}{dt} = f(\vec{x})$$

in discrete form using Euler forward:  $\vec{x}_{t+1} = \vec{x}_t + \Delta t \cdot f(\vec{x}_t)$   
 $= m(\vec{x}_t)$

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}_{t+1} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}_t + \Delta t \begin{pmatrix} -\sigma x_1 + \sigma x_2 \\ \rho x_1 - x_2 - x_1 x_3 \\ x_1 x_2 - \beta x_3 \end{pmatrix}_t = \begin{pmatrix} m_1 \\ m_2 \\ m_3 \end{pmatrix}$$

Linearize:  $\frac{\partial m_i}{\partial x_j}$

$$M = \begin{pmatrix} \frac{\partial m_1}{\partial x_1} & \frac{\partial m_1}{\partial x_2} & \frac{\partial m_1}{\partial x_3} \\ \frac{\partial m_2}{\partial x_1} & \frac{\partial m_2}{\partial x_2} & \frac{\partial m_2}{\partial x_3} \\ \frac{\partial m_3}{\partial x_1} & \frac{\partial m_3}{\partial x_2} & \frac{\partial m_3}{\partial x_3} \end{pmatrix} = \begin{pmatrix} 1 + \Delta t(-\sigma) & \Delta t\sigma & 0 \\ \Delta t(\rho - x_3) & 1 - \Delta t & -\Delta t x_1 \\ \Delta t x_2 & \Delta t x_1 & 1 - \Delta t\beta \end{pmatrix}$$

$$= I + \Delta t \begin{pmatrix} -\sigma & \sigma & 0 \\ \rho - x_3 & -1 & -x_1 \\ x_2 & x_1 & -\beta \end{pmatrix}_t = I + \Delta t F$$

$$\delta \vec{x}_{t+1} = M \delta \vec{x}_t = \delta \vec{x}_t + \underbrace{\Delta t \cdot F \cdot \delta \vec{x}_t}_{\text{change in perturbation due to model forcing}}$$

change in perturbation  
due to model forcing

## Example 2: TLM for L96

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Discrete form using Euler forward:

$$x_i^{t+1} = m(x_i^t) = x_i^t + \Delta t \left[ (x_{i+1}^t - x_{i-2}^t) x_{i-1}^t - x_i^t + F \right]$$

$$\vec{\delta x}_{t+1} = M \vec{\delta x}_t = \underbrace{(I + \Delta t F^*)}_{\text{TLM}} \vec{\delta x}_t$$

$$M \vec{\delta x}_t = \vec{\delta x}_t + \Delta t$$

$\begin{bmatrix} \ddots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \dots & -1 & x_{i-3} & 0 & 0 & \dots \\ \dots & x_i - x_{i-3} & -1 & x_{i-2} & 0 & \dots \\ \dots & -x_{i-1} & x_{i+1} - x_{i-2} & -1 & x_{i-1} & \dots \\ \dots & 0 & -x_i & x_{i+2} - x_{i-1} & -1 & \dots \\ \dots & 0 & 0 & -x_{i+1} & x_{i+3} - x_i & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \ddots \end{bmatrix}_t$	$\begin{bmatrix} \vdots \\ \delta x_{i-2} \\ \delta x_{i-1} \\ \delta x_i \\ \delta x_{i+1} \\ \delta x_{i+2} \\ \vdots \end{bmatrix}_t$
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ADM
TLM

- Nonlinear L96 code:

for  $i=1, 2, \dots, n$

$$x_i^{t+1} = x_i^t + \Delta t \left[ (x_{i+1}^t - x_{i-2}^t) x_{i-1}^t - x_i^t + F \right]$$

- TLM L96 code:

for  $i=1, 2, \dots, n$

$$\delta x_i^{t+1} = (1 - \Delta t) \delta x_i^t + \Delta t \left[ -x_{i-1}^t \delta x_{i-2}^t + (x_{i+1}^t - x_{i-2}^t) \delta x_{i-1}^t + x_{i-1}^t \delta x_{i+1}^t \right]$$

- ADM L96 code:

for  $i=1, 2, \dots, n$

$$\delta x_i^{t*} = (1 - \Delta t) \delta x_i^{t+1} + \Delta t \left[ x_{i-2}^{t+1} \delta x_{i-1}^{t+1} + (x_{i+2}^{t+1} - x_{i-1}^{t+1}) \delta x_{i+1}^{t+1} - x_{i+1}^{t+1} \delta x_{i+2}^{t+1} \right]$$