

# Variational Data Assimilation (4D - Var)

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- Smoothing for large, deterministic systems.
- Commonly used in numerical weather prediction.

$\mathbf{x}_{k+1} = M_k(\mathbf{x}_k)$       Deterministic Dynamics

$\mathbf{x}_b, \mathbf{B}$       Prior for Initial Condition (Background)

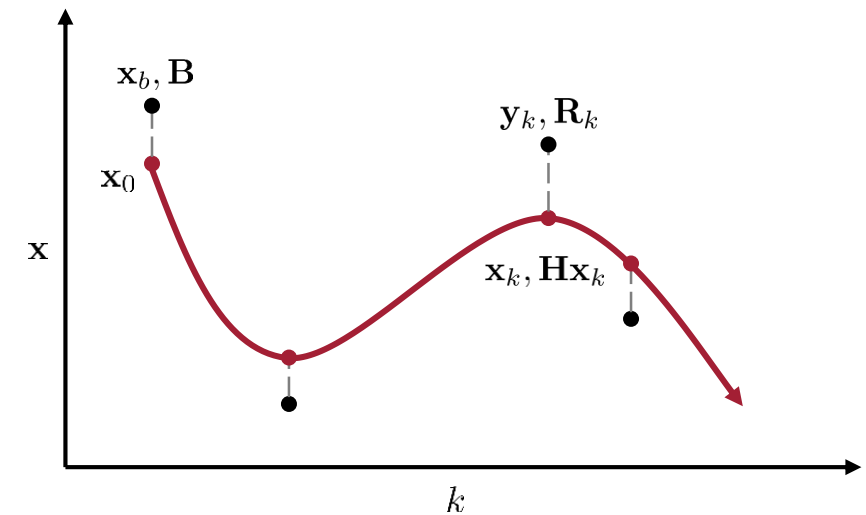
$\mathbf{y}_k, \mathbf{R}_k$       Noisy Observations at times  $k$

- Key idea: the best initial condition creates a trajectory that minimizes

$$J(\mathbf{x}_0) = \underbrace{\|\mathbf{x}_0 - \mathbf{x}_b\|}_{\text{Distance from Prior}} \underbrace{\mathbf{B}^{-1}}_{\text{Prior Precision}} + \sum_k \underbrace{\|\mathbf{y}_k - \mathbf{H}\mathbf{x}_k\|}_{\text{Distance from Data}} \underbrace{\mathbf{R}_k^{-1}}_{\text{Data Precision}}$$



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# Computational Considerations

## How do we do this?

- $J$  depends on all of the  $\mathbf{x}_k$ , but the optimization requires  $dJ/d\mathbf{x}_0$
- Adjoint Model (Linearization):

$$\begin{cases} \mathbf{x}_{j+1} = M_j(\mathbf{x}_j) & \text{Nonlinear Dynamics} \\ \mathbf{M}_j = M'_j(\mathbf{x}_j) & \text{Jacobian Matrix} \end{cases}$$

Forward Pass: Dynamics



$$\mathbf{x}_k = [M_k \circ M_{k-1} \circ \dots \circ M_2 \circ M_1](\mathbf{x}_0)$$

$$\frac{d}{d\mathbf{x}_0} = \mathbf{M}_1^T \cdot \mathbf{M}_2^T \cdot \dots \cdot \mathbf{M}_{k-1}^T \cdot \mathbf{M}_k^T \cdot \frac{d}{d\mathbf{x}_k}$$



Backward/Adjoint Pass: Gradients

## Algorithm:

Initialize  $\mathbf{x}_0$ .

Loop until  $\|dJ/d\mathbf{x}_0\|$  is small:

- Forward pass: calculate  $J(\mathbf{x}_0)$  and  $\{dJ/d\mathbf{x}_k\}$
- Adjoint pass: calculate  $dJ/d\mathbf{x}_0$  from  $\{dJ/d\mathbf{x}_k\}$
- Update  $\mathbf{x}_0$  based on  $dJ/d\mathbf{x}_0$

## Why do we do this?

- Large, nonlinear, deterministic systems.
- Minimization can be “sloppy” – approximate gradients, reduced models

# Example Application

## Infer the trajectory of an elastic rod

- System of constrained nonlinear ODEs
- Noisy observations of marked points
- See write-up for details on adjoints, covariance localization

### **In one sentence:**

Variational Data Assimilation (4D-Var) smooths large deterministic systems using gradient-based minimization and adjoint models.

