

**Tutorial 10:
Observation Impact &
Observation Sensitivity**

Impact/Sensitivity Functions

We will use as an example, the time-averaged transport along a line of constant latitude, ϕ :

$$I = \frac{1}{(t_2 - t_1)} \int_{t_1}^{t_2} \int_{\lambda_1}^{\lambda_2} \int_{z_1}^{z_2} v(\lambda, \phi, z) dz d\lambda dt$$

$$\equiv \frac{1}{(k_2 - k_1)} \sum_{k=k_1}^{k_2} \mathbf{h}_k^T \mathbf{x}_k$$

where $[t_1, t_2] \equiv [k_1 \Delta t, k_2 \Delta t]$ and $\mathbf{x}_k \equiv \mathbf{x}(k \Delta t)$

$$\mathbf{x}_k = \begin{bmatrix} \mathbf{T}_k \\ \mathbf{S}_k \\ \boldsymbol{\varsigma}_k \\ \mathbf{u}_k \\ \mathbf{v}_k \end{bmatrix}$$

$$\mathbf{h}_k = \begin{bmatrix} \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \\ \mathbf{0} \\ \Delta z \Delta \lambda \end{bmatrix}$$

$$\frac{\partial I}{\partial \mathbf{x}_j} = \frac{1}{(k_2 - k_1)} \mathbf{h}_j$$

Impact/Sensitivity Functions

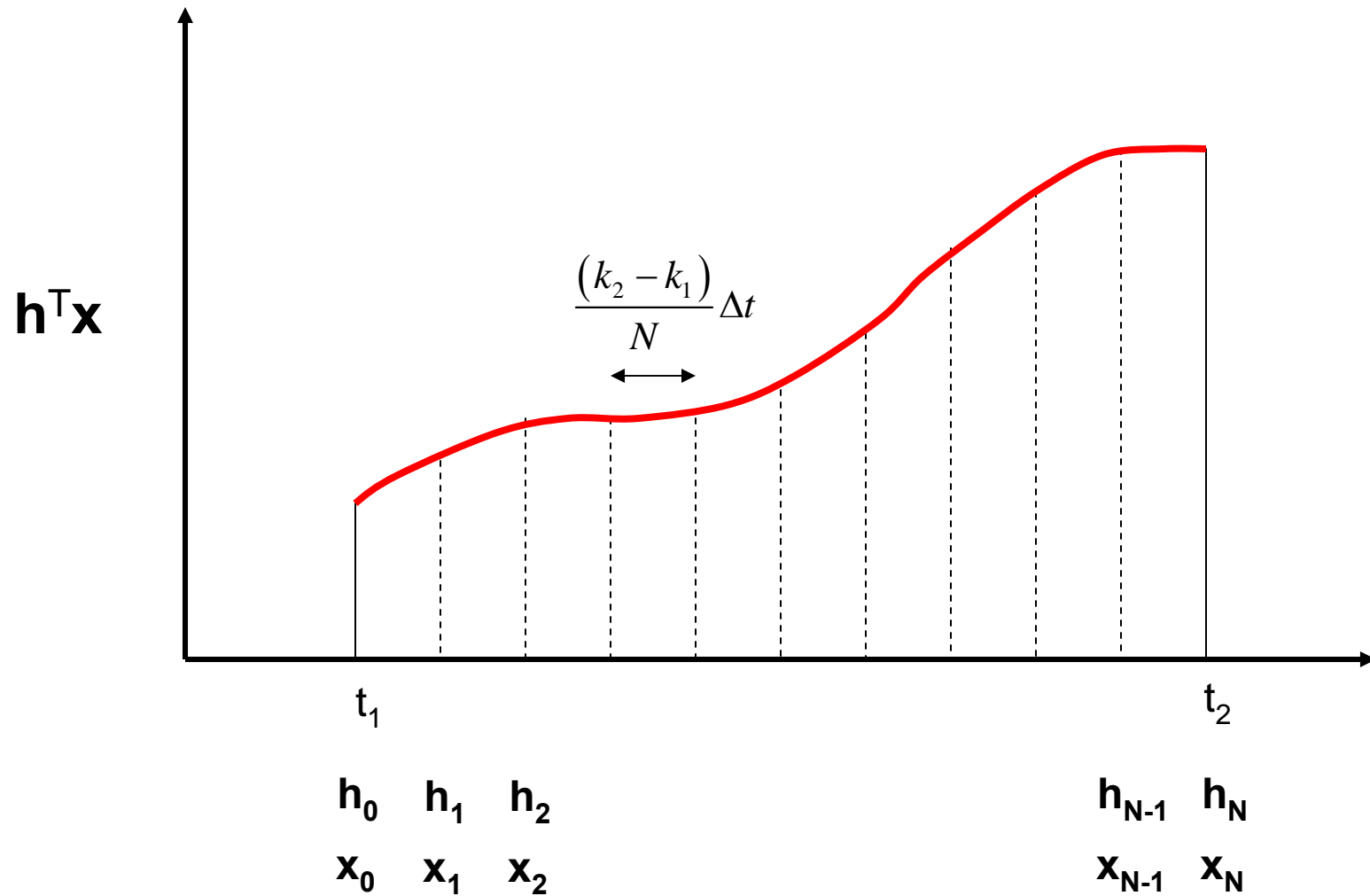
Recall that observation impact and observation sensitivity calculations involve forcing ADROMS with $\partial I/\partial \mathbf{x}$.

$$I = \frac{1}{(k_2 - k_1)} \sum_{k=k_1}^{k_2} \mathbf{h}_k^T \mathbf{x}_k$$

$$\frac{\partial I}{\partial \mathbf{x}_j} = \frac{1}{(k_2 - k_1)} \mathbf{h}_j$$

The Trapezoidal Rule

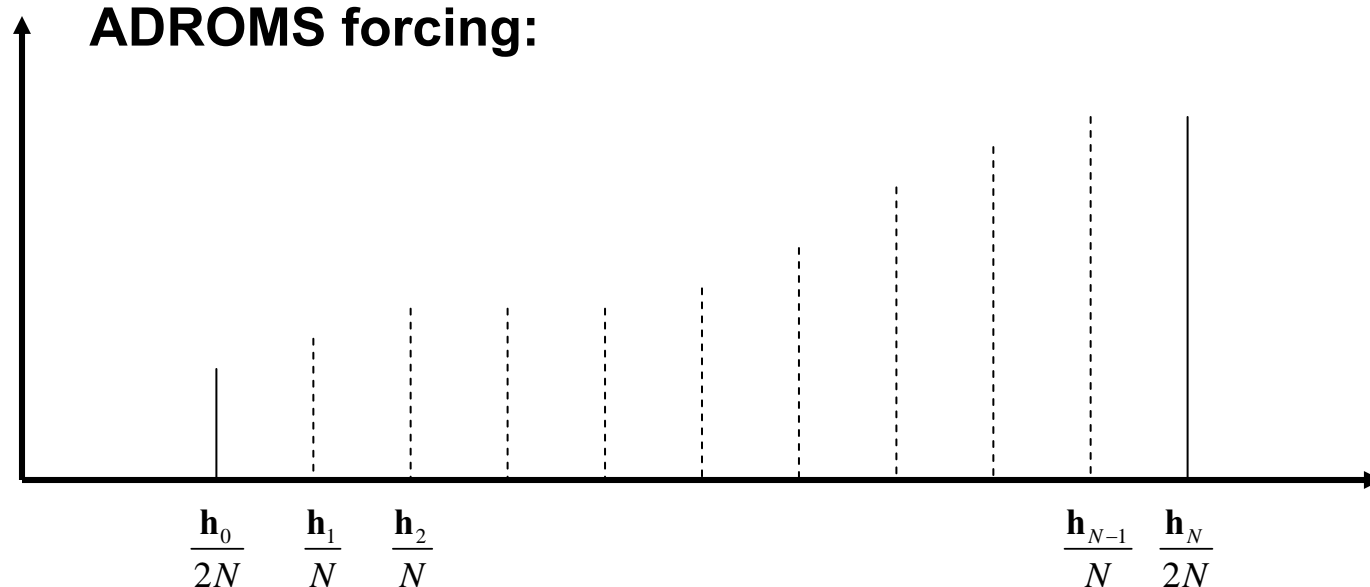
Saving x every timestep is costly, so instead use an approx.



The Trapezoidal Rule

$$I = \frac{1}{N} \left(\frac{1}{2} (\mathbf{h}_0^T \mathbf{x}_0 + \mathbf{h}_N^T \mathbf{x}_N) + \sum_{i=1}^{N-1} \mathbf{h}_i^T \mathbf{x}_i \right)$$

$$\frac{\partial I}{\partial \mathbf{x}_0} = \frac{1}{2N} \mathbf{h}_0; \quad \frac{\partial I}{\partial \mathbf{x}_N} = \frac{1}{2N} \mathbf{h}_N; \quad \frac{\partial I}{\partial \mathbf{x}_i} = \frac{1}{N} \mathbf{h}_i;$$



define AD_IMPULSE

Some Common General Cases

(i) Linear I :

$$I = \sum_{k=k_1}^{k_2} \mathbf{h}_k^T \mathbf{x}_k; \quad \partial I / \partial \mathbf{x}_k = \mathbf{h}_k$$

(i) Quadratic I :

$$I = \sum_{k=k_1}^{k_2} (\mathbf{x}_k \pm \mathbf{g})^T \mathbf{L} (\mathbf{x}_k \pm \mathbf{g}); \quad \partial I / \partial \mathbf{x}_k = (\mathbf{L} + \mathbf{L}^T) (\mathbf{x}_k \pm \mathbf{g})$$

where \mathbf{g} is an arbitrary vector that is not a function of \mathbf{x} .