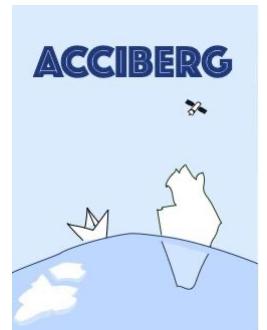
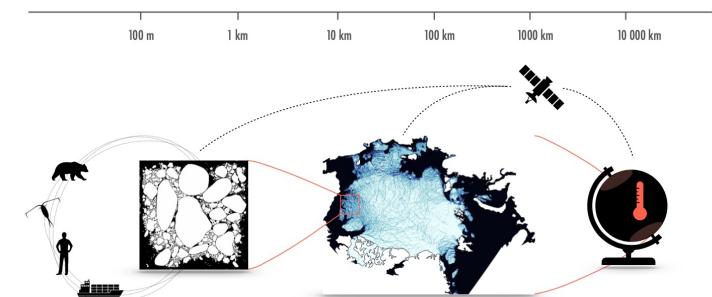


Assimilating observations of deformation to improve short-term ensemble forecasts of sea ice features

Yue (Michael) Ying

NERSC



*SASIP, *ACCIBERG

Pierre Rampal, Einar Olason, Anton Korosov, Laurent Bertino

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Motivation

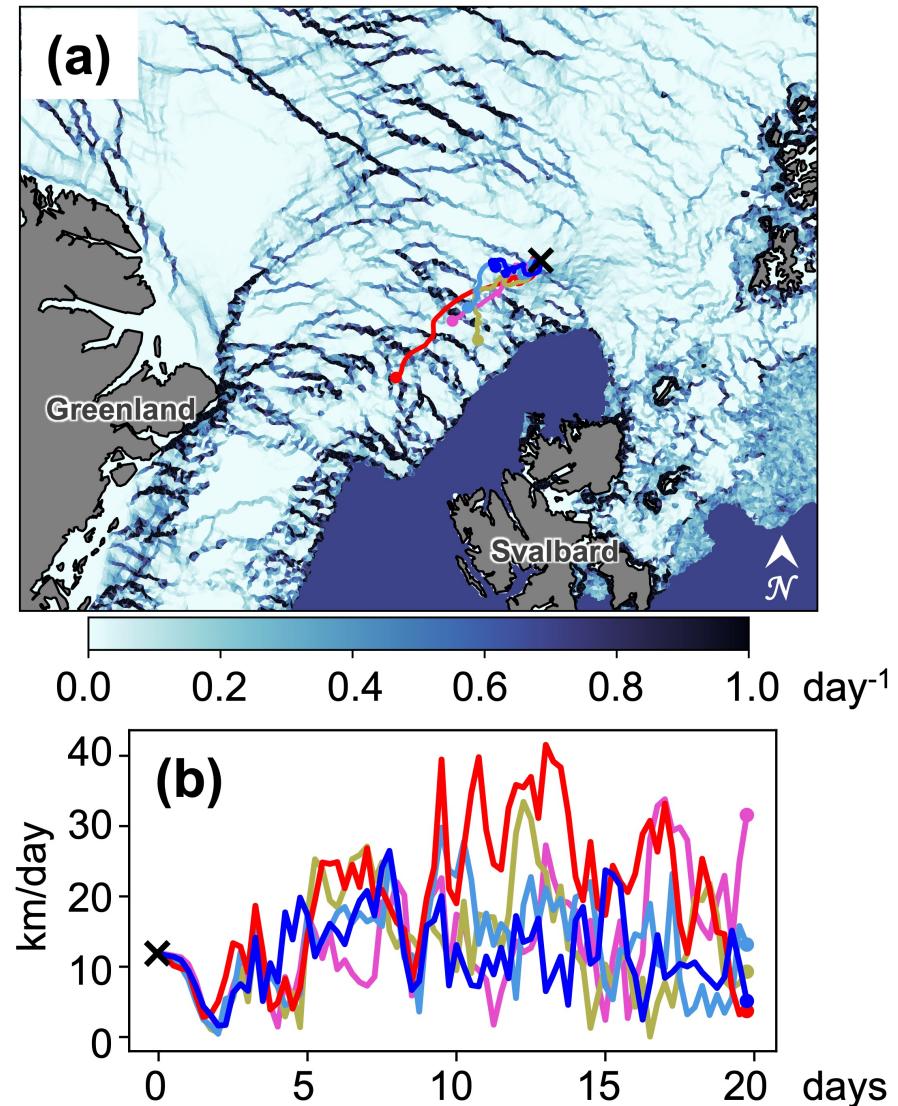
Deformation is important features for short-term sea ice forecasts: both large-scale **wind forcing** and small-scale ice **rheology** influences.

Challenging **multifractal** and **nonlinear** problems!

Part 1: some visual example of sea ice deformation

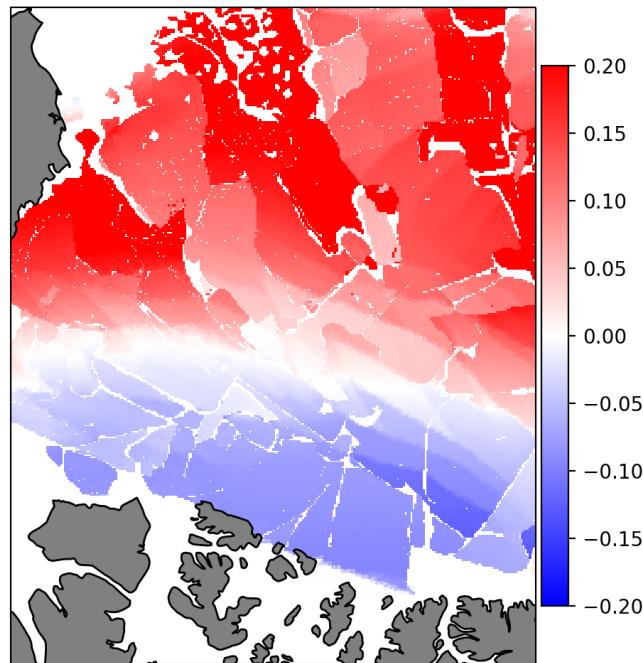
Part 2: a 1-D illustration of the assimilation problem

Part 3: Bayesian formulation for the multiscale filter

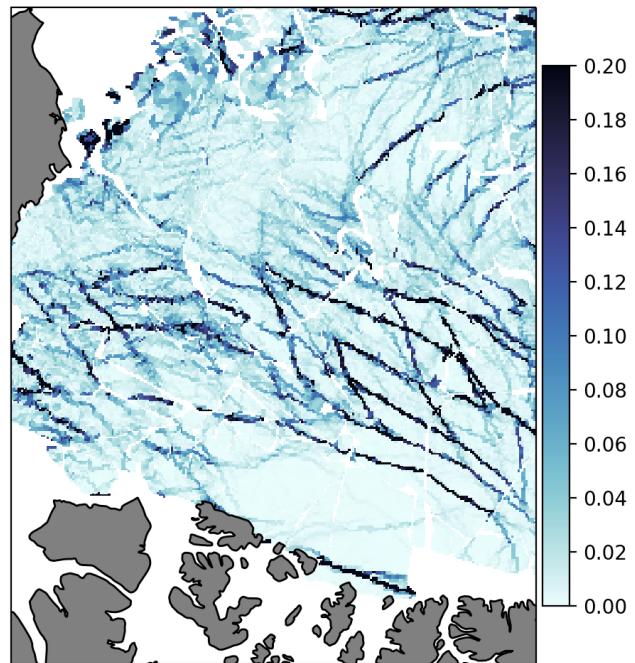


Part 1: Sea ice drift and deformation

ice drift u (m/s)



ice shear deformation (1/day)



RGPS SAR imagery
(Synthetic Aperture Radar)

Tracking ice image pairs for
coherent features: ice drift

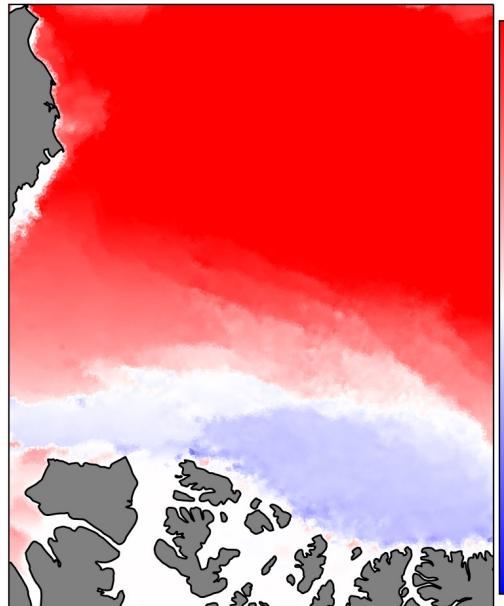
$$(u, v) = \left(\frac{x(t + \delta t) - x(t)}{\delta t}, \frac{y(t + \delta t) - y(t)}{\delta t} \right)$$

$$\dot{\varepsilon}_{\text{shear}} = \sqrt{(u_x - v_y)^2 + (u_y + v_x)^2},$$

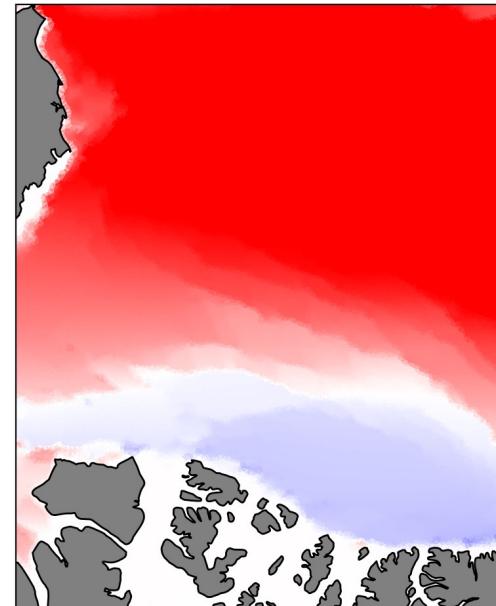
as $\delta t \rightarrow 0$, (u, v) is the
instantaneous velocity in model
restart files

Sea ice drift definition

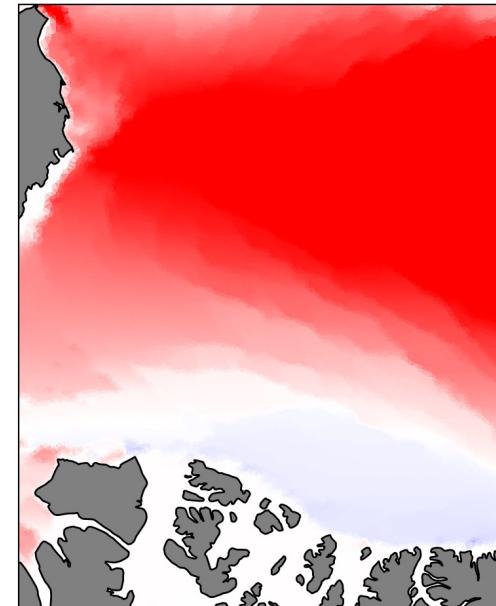
instantaneous u



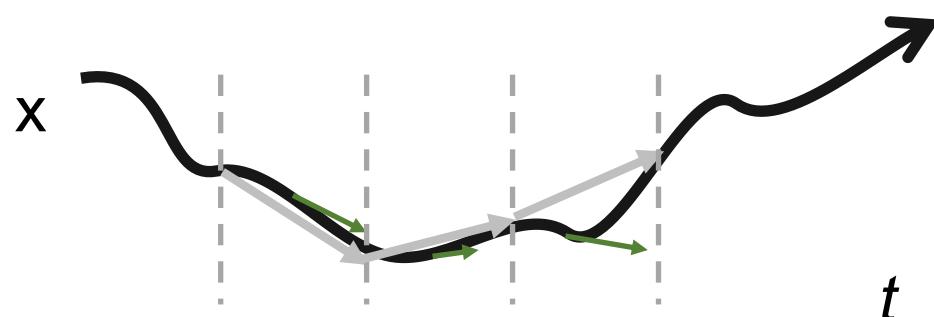
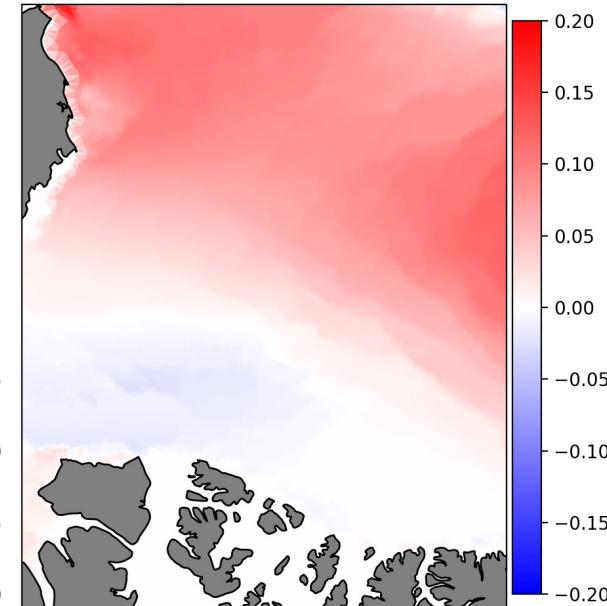
u from 12-h trajectory



u from 24-h trajectory



u from 72-h trajectory

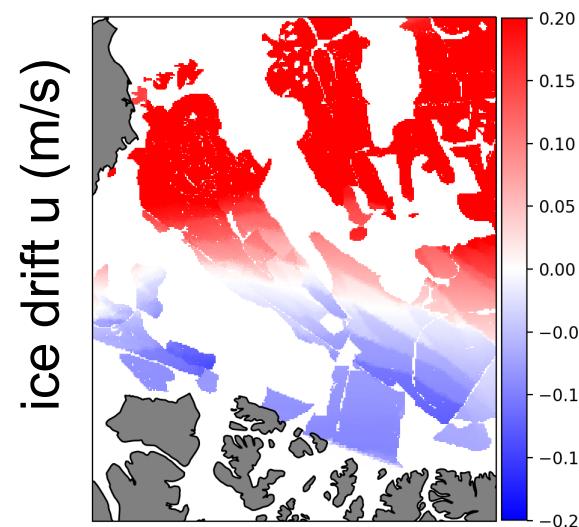


we need:

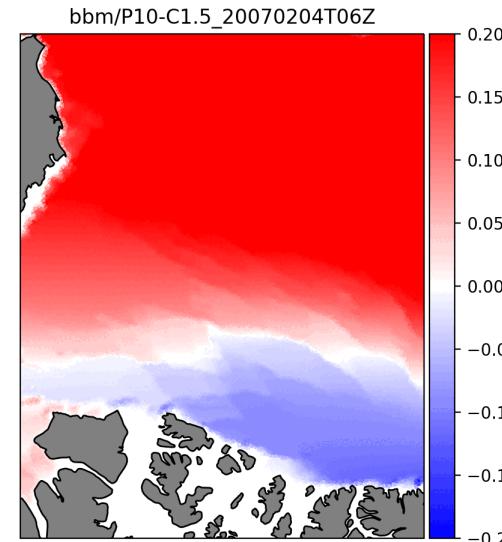
- enough temporal resolution
- asynchronous assimilation

neXtSIM ensemble simulation of sea ice deformation

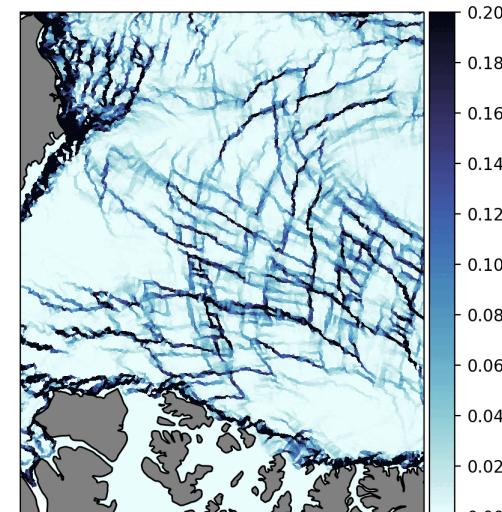
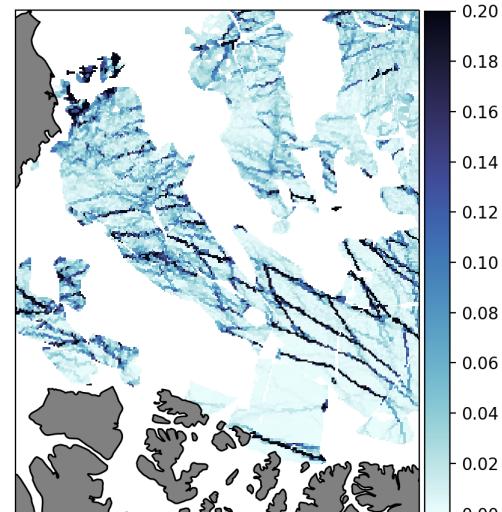
observed (RGPS)



model simulated (neXtSIM)



ice shear deformation
(1/day)



ensemble using different rheology parameters (P, C)

but using the same atmospheric forcing

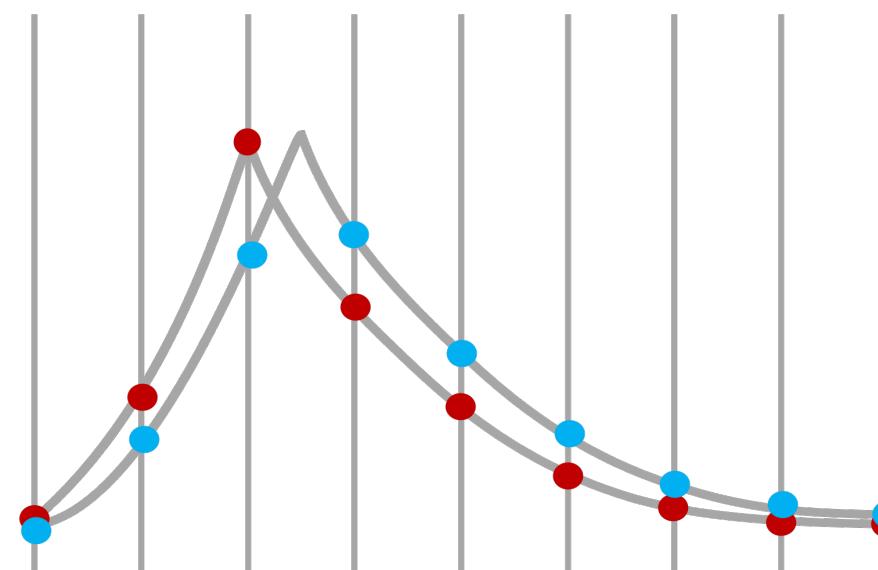
Challenges in assimilating small-scale features

- **irregular triangular mesh:** members have different mesh positions

Interpolation to a common analysis grid:

at higher resolution, it's costly

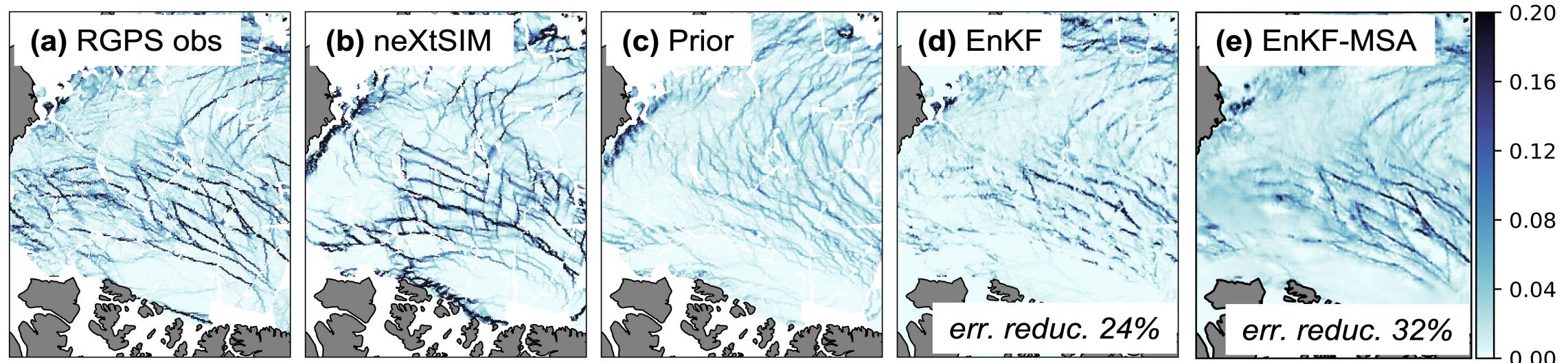
at the same resolution, interpolation is diffusive



Challenges in assimilating small-scale features

- **irregular triangular mesh:** members have different mesh positions
- **deformation is fractal:** hierarchy of features with position errors

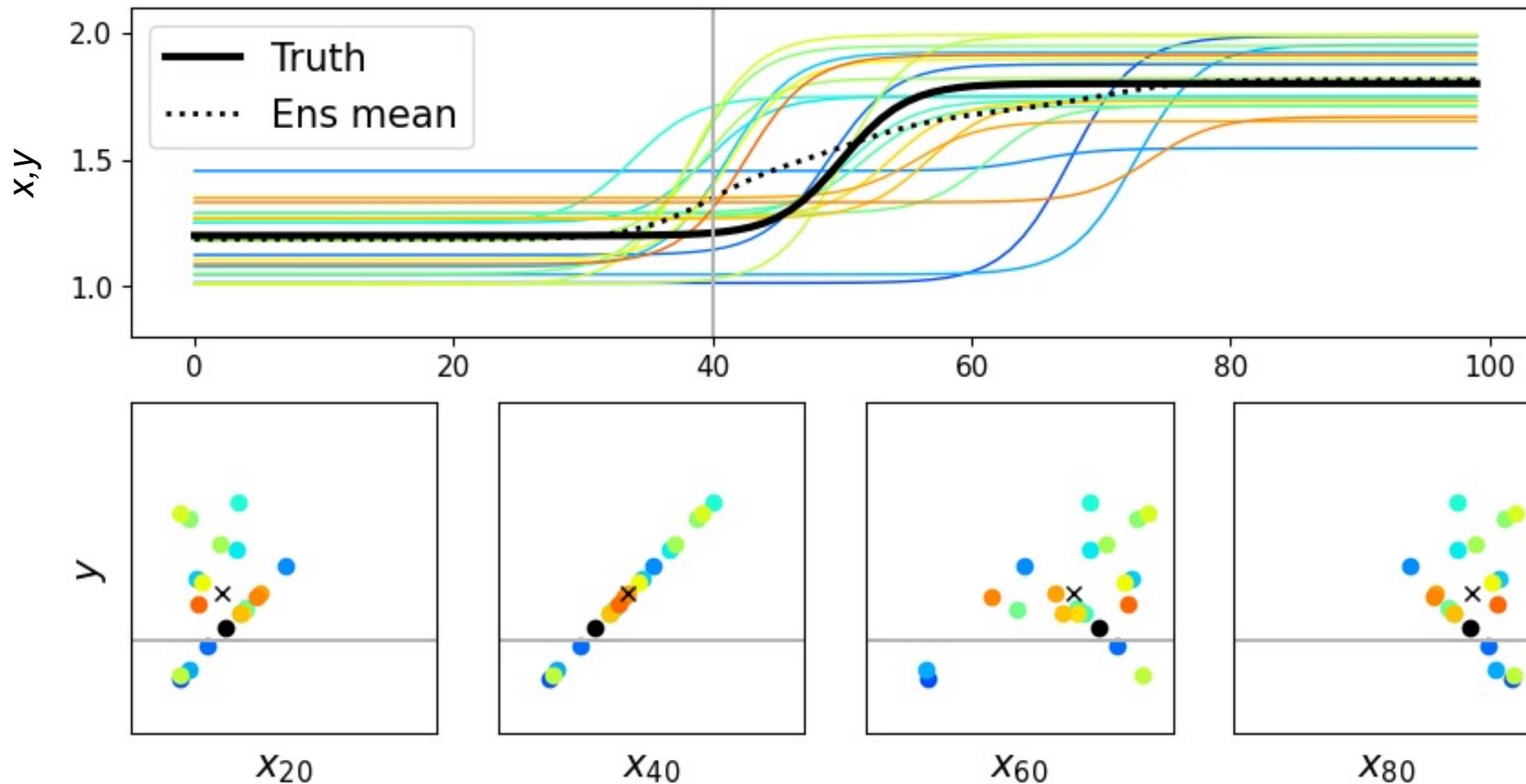
Using multiscale alignment method, we can better assimilate deformation:



How to update the model states (drift, concentration, damage...)?

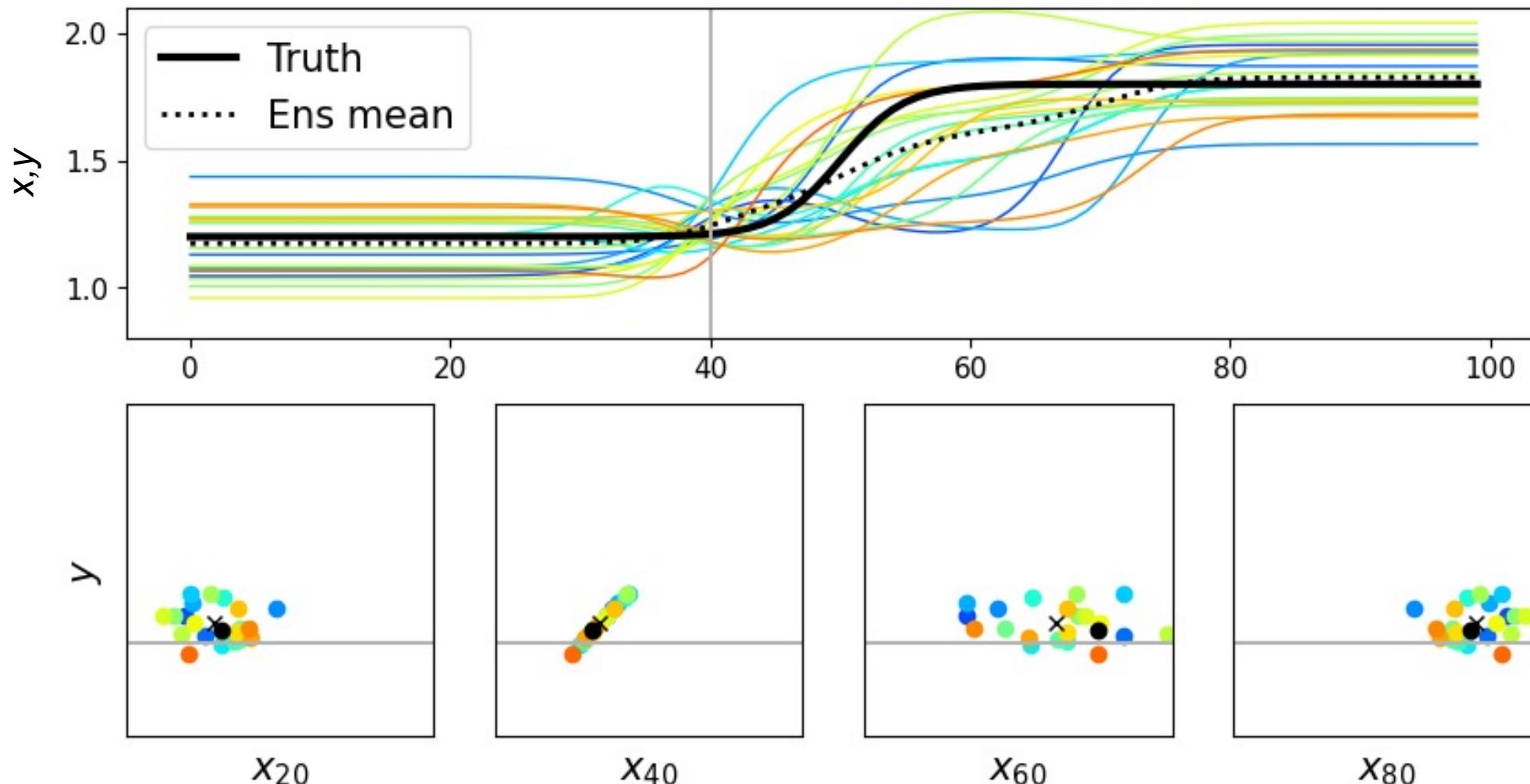
Part 2: 1-D illustration of the problem

Prior

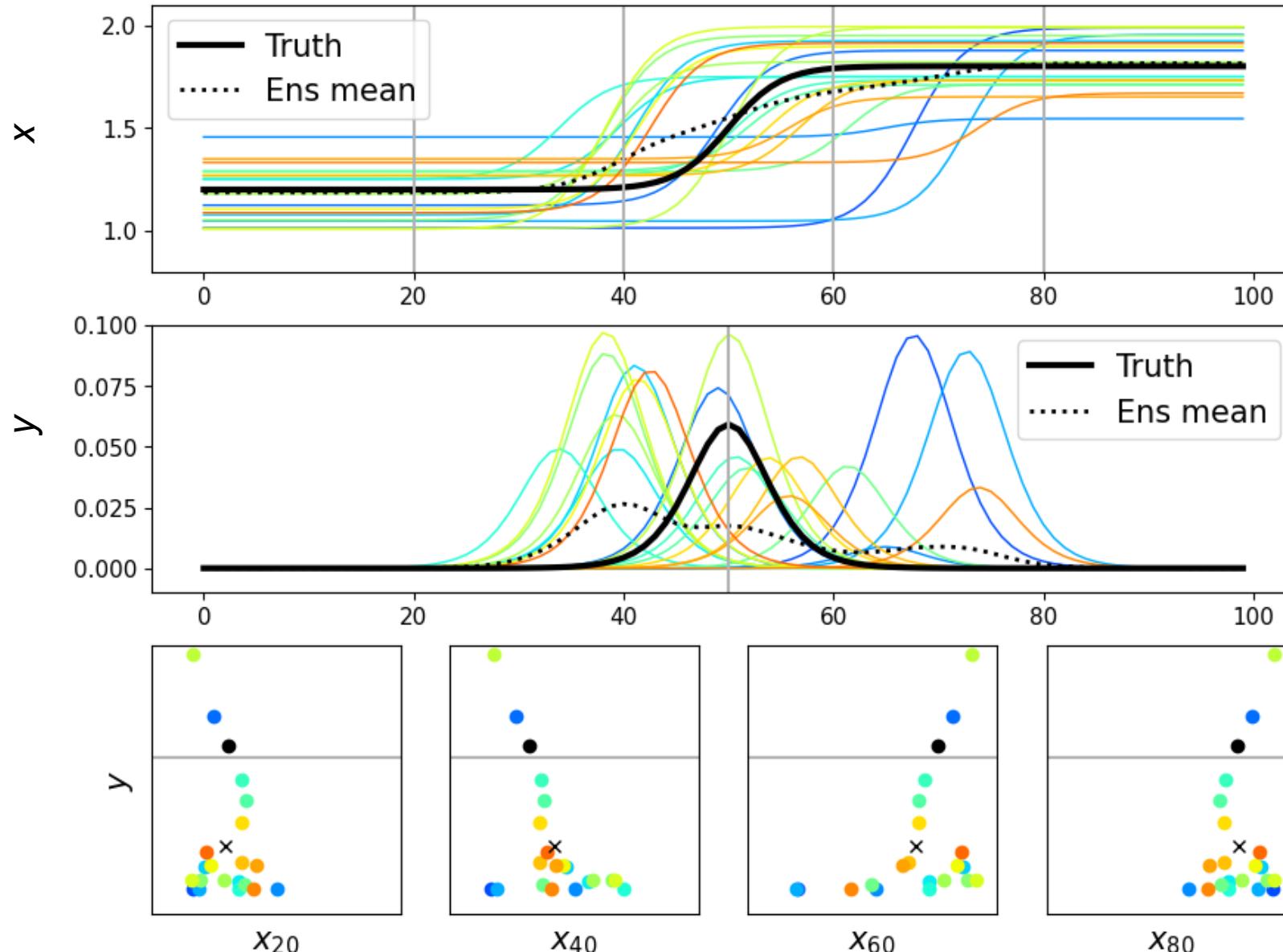


Part 2: 1-D illustration of the problem

Posterior, assimilating drift to update drift

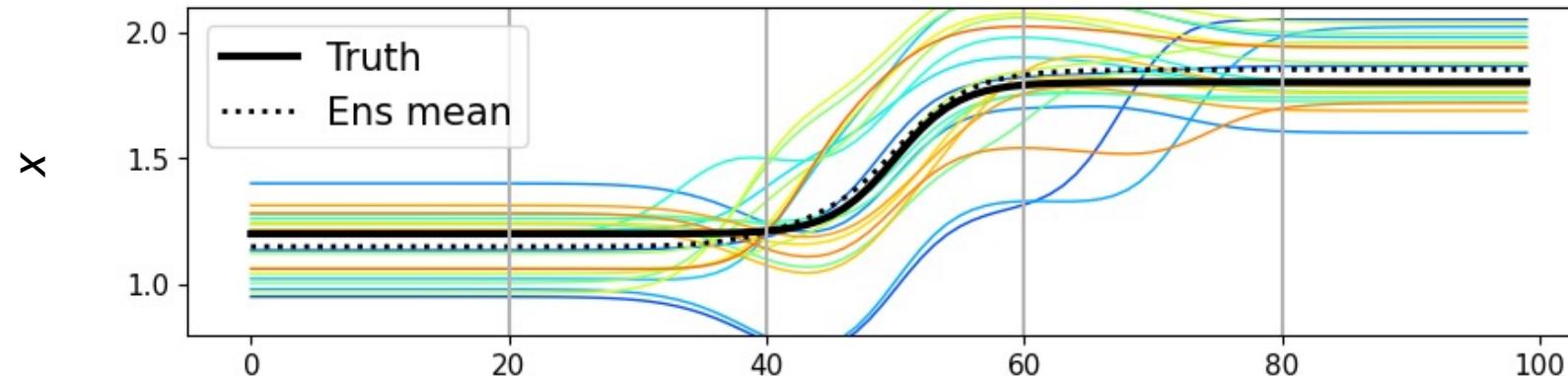


Advantage of assimilating deformation

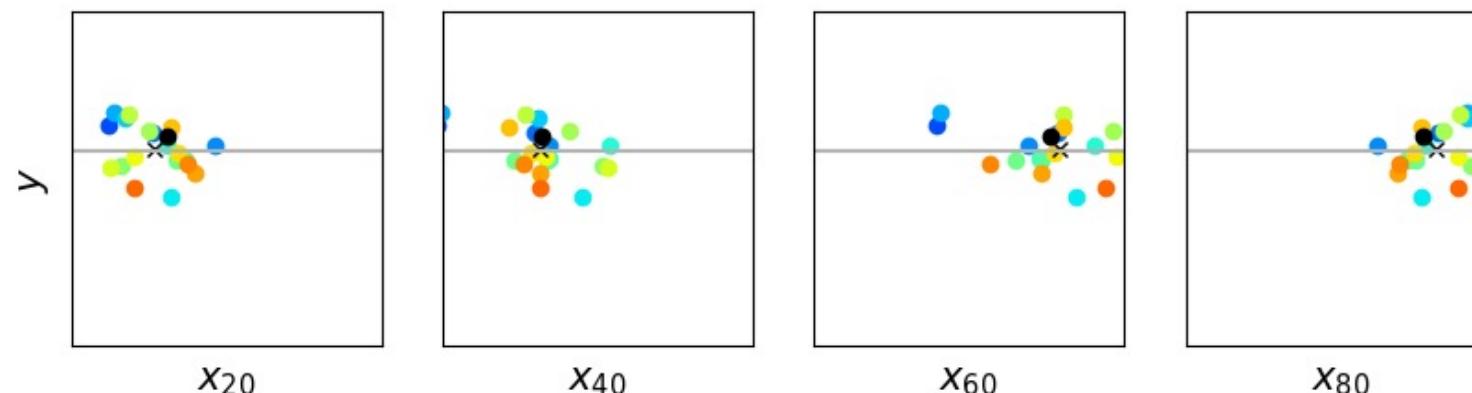
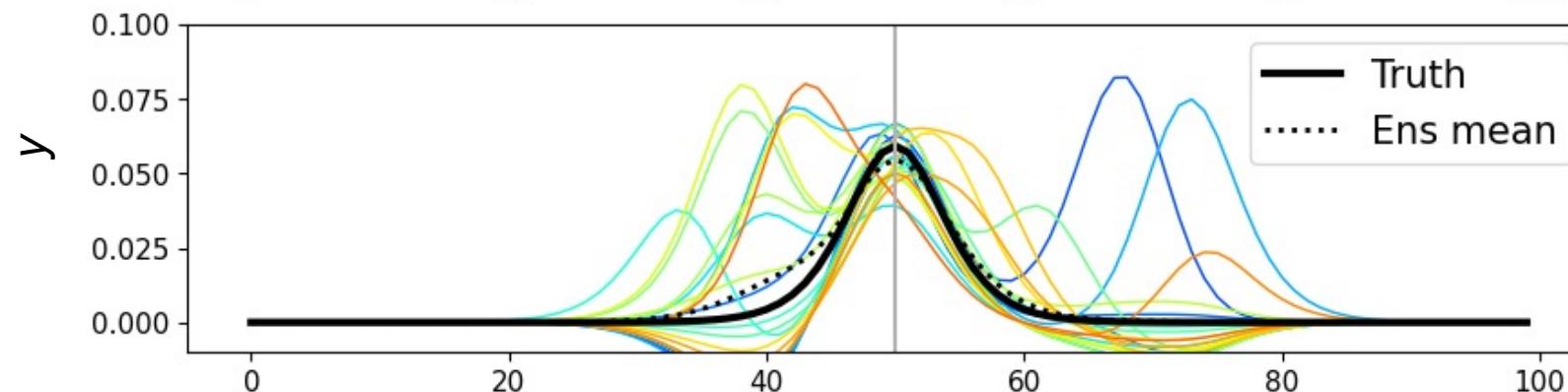


x : drift u
 y : deform, grad(u)

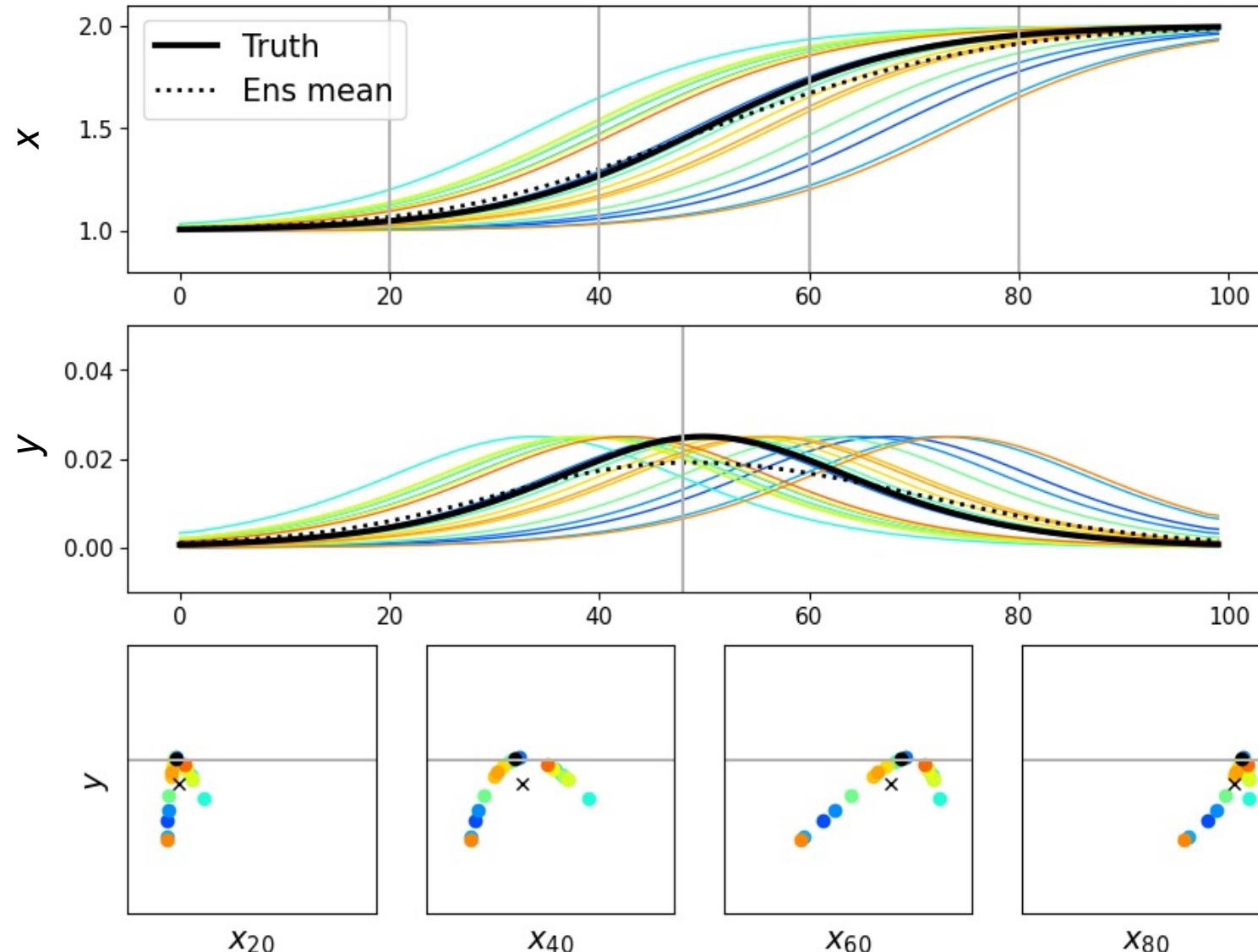
Advantage of assimilating deformation



x: drift u
y: deform, $\text{grad}(u)$



Effect of nonlinear position errors



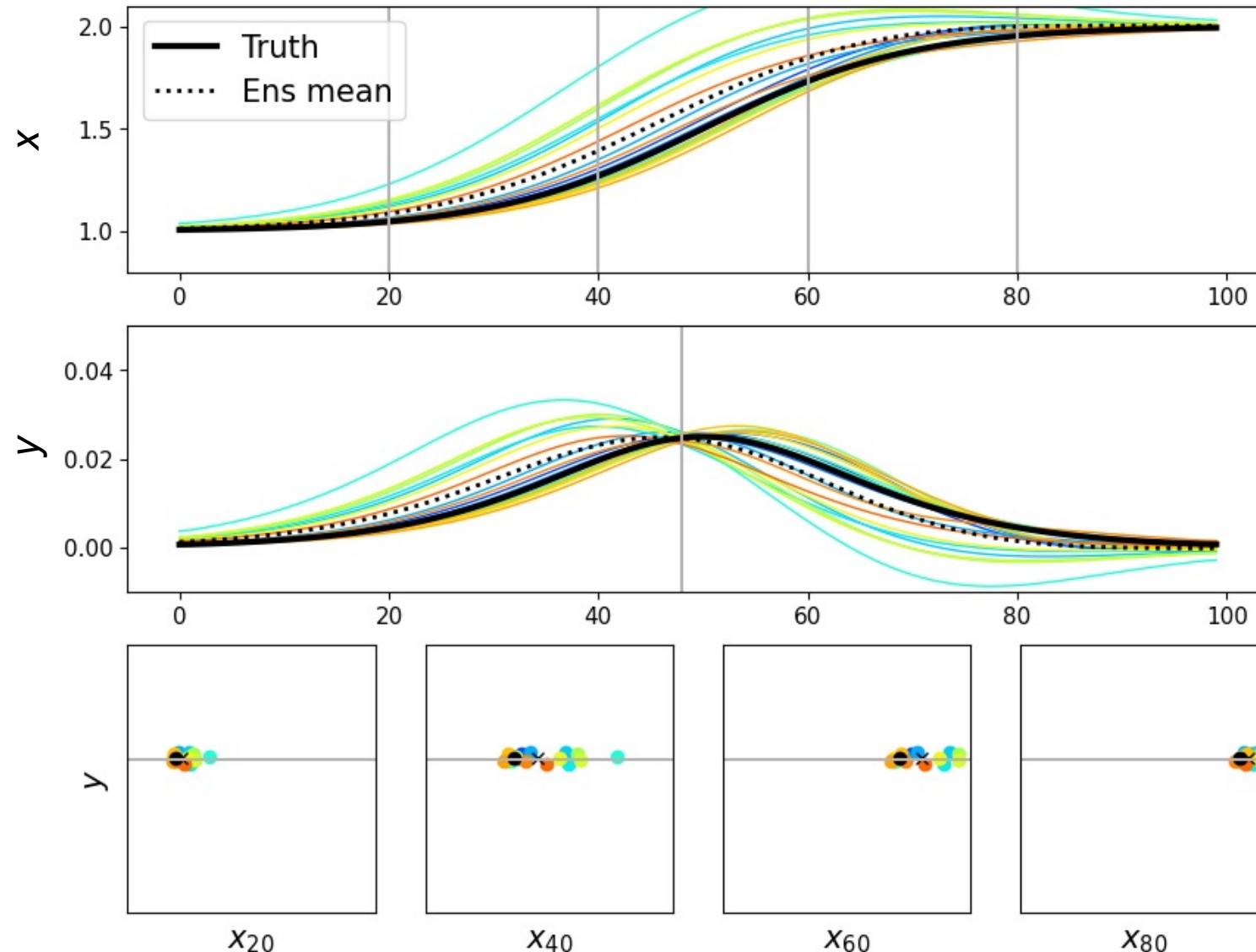
x : drift u

y : deform, $\text{grad}(u)$

only position errors

nonlinearity is shown

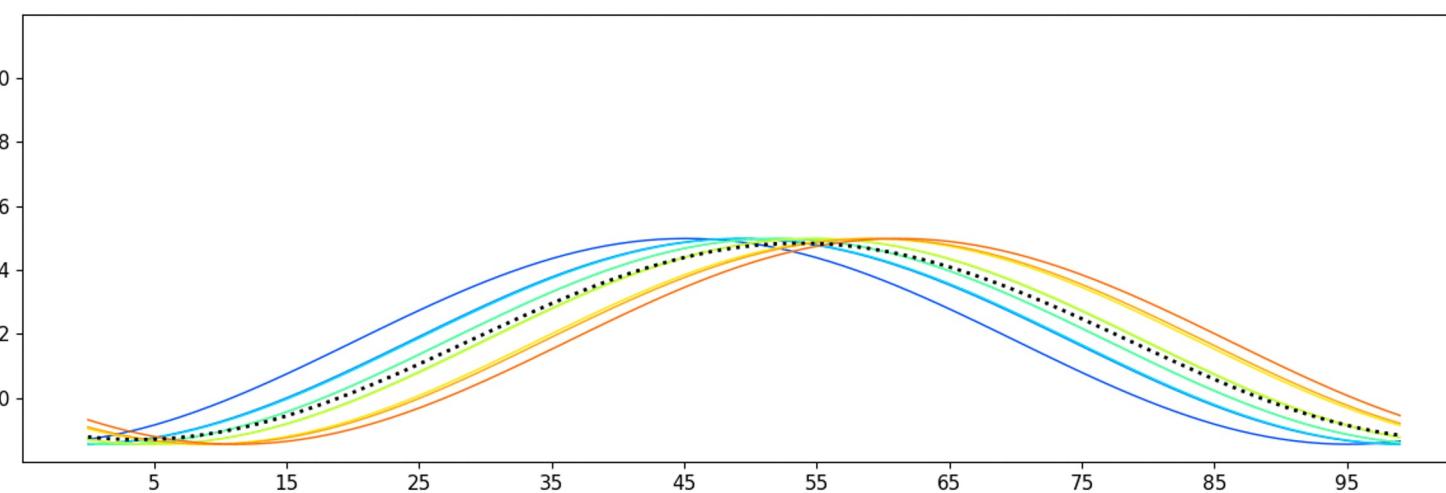
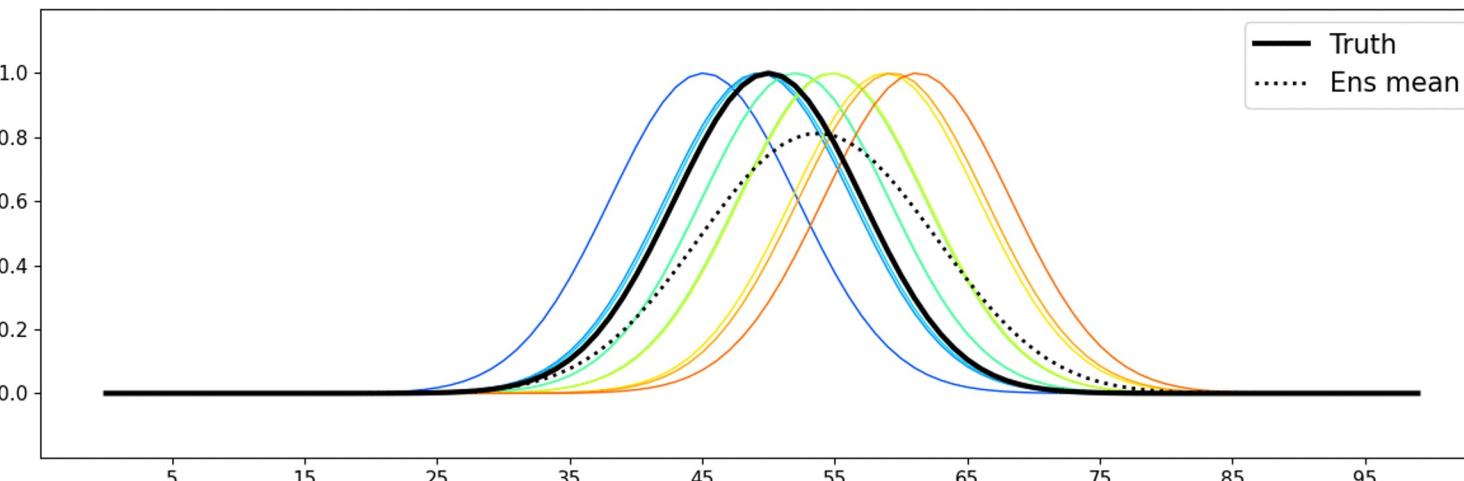
Effect of nonlinear position errors



x : drift u
 y : deform, grad(u)

assimilation corrects
some position errors,
but introduced bias in
drift speed

Remedy: multiscale alignment



Run filter at coarse resolution

Find displacement vectors
and warp the fields

(optional) run filter again for
higher resolution ...

*This is also useful for irregular
mesh problems, to preserve
sharp gradients!*

Part 3: Bayesian framework for multiscale filtering

The underlying idea for multiscale alignment:

Decompose state into sum of scale component $s = 1, 2 \dots N_s$

$$x^b = x_1^b + x_2^b + \dots + x_{N_s}^b$$

At each scale, the error is further separated into phase and amplitude:

$$x_s^b(r) = x_s^t(r + q_s) + x'_s$$

e.g., Fourier transform $x(r) \rightleftharpoons \hat{x}(k)e^{ikr}$

The new cost function

$$p(x, q | y) \propto p(y | x, q) p(x | q) p(q)$$

$$J(x, q) = \|y - h[x(q)]\|_R^2 + \|x(q) - x^b(q)\|_{B(q)}^2 + \ln[B(q)] + L(q)$$

two-step solution:

1. filter update $J(x)|_{q=0} = \|y - h[x]\|_R^2 + \|x - x^b\|_B^2 \rightarrow x^a$

2. alignment $J(q)|_{x^a} = \|x^a - x^b(q)\|^2 + L(q) \rightarrow q$

Relation between large- and small-scale components

Ying 2019 assumed that small-scale displacement inherits the large-scale ones

$$x_s^b(q_1 + q_2 + \dots + q_{s-1})$$

Ying et al. 2023: large- and small-scale position errors can be incoherent, making this a bad assumption

Instead, maybe using covariance to update $q_s^b \rightarrow q_s^a$?

Summary and ongoing work

Sea ice deformation has fractal features, which are important information for models to maintain prediction skill

Traditional EnKF has limitation when assimilating sea ice observation

- Assimilating deformation in addition to drift
- Treatment of irregular lagrangian mesh
- Nonlinear position errors tackled by alignment techniques
- Rethinking the large- and small-scale relationship