

A comparison of the LETKF and the equivalent weights particle filter on the barotropic vorticity model

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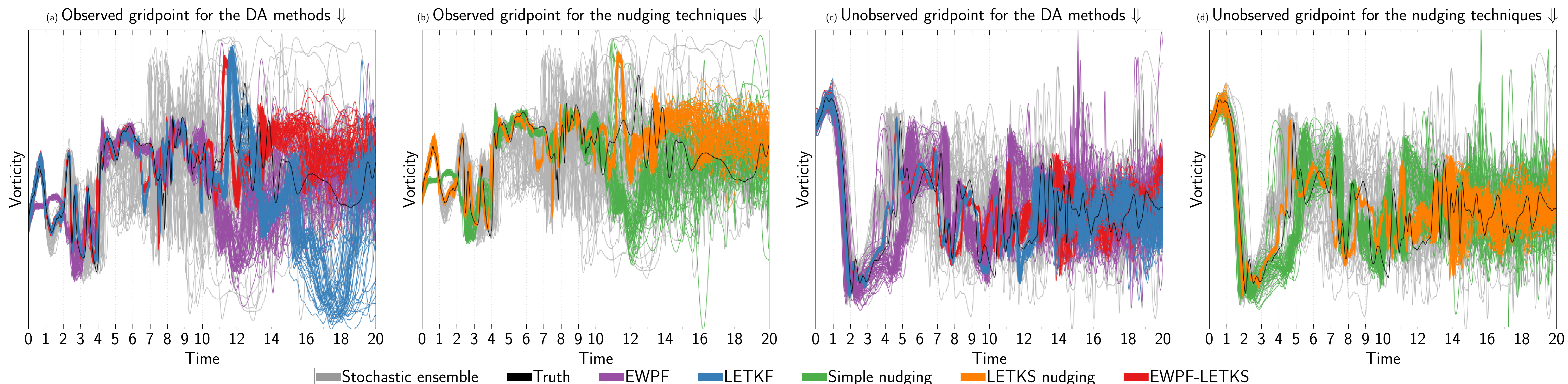


Figure 1: Trajectories of 2 different points in the domain when using the different assimilation methods with observing network 3 for a single experiment

We consider the *barotropic vorticity* model:

$$\frac{\partial q}{\partial t} + u \frac{\partial q}{\partial x} + v \frac{\partial q}{\partial y} = 0$$

Example true model state ↓

Example observations ↓

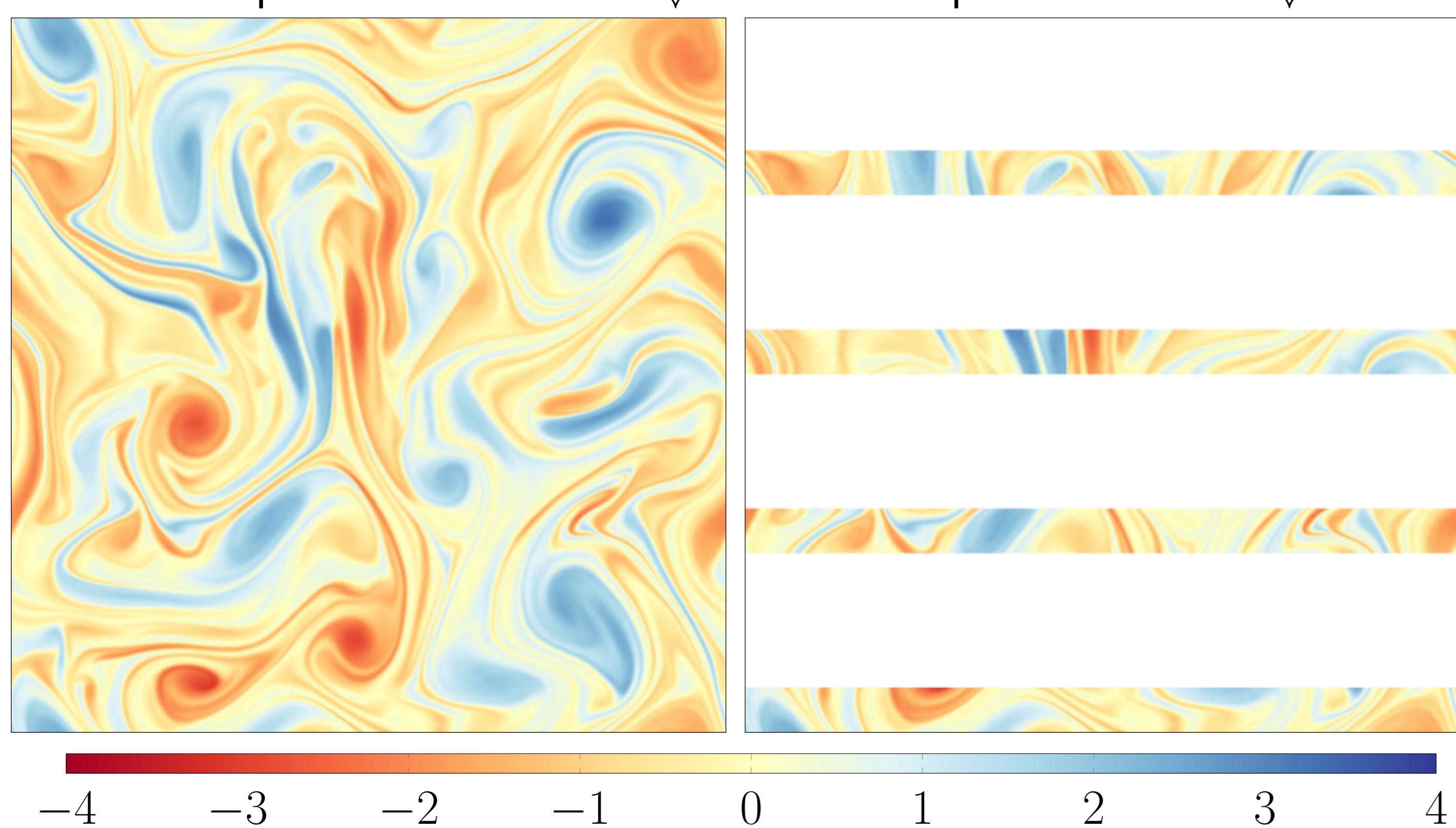


Figure 2. Plots of vorticity q for the true state and the resulting observations at the 6th analysis time

$N_x = 2^{18} = 262,144$. $N_y = 65,336$. $N_e = 48$. All the experiments used the EMPIRE data assimilation system.

The EWPF is a 2-stage filter:

1. For all timesteps before an observation, relax using g s.t.

$$x_i^{k+1} = f(x_i^k) + g(x_i^k, y, z) + \beta_i^k$$
2. At analysis time, α_i chosen to make weights *equivalent*, s.t.

$$x_i^n = f(x_i^{n-1}) + \beta_i^n + \alpha_i Q H^T (H Q H^T + R)^{-1} (y^n - H(f(x_i^{n-1})))$$

For the basic EWPF, $g(x_i^k, y, z) = \rho Q H^T R^{-1} (y - H x_i^k)$

- Q 1: what *should* the relaxation term look like?
 Q 2: why are the trajectories for ensembles that use nudging and particle filters using the same relaxations so similar?

Answers (partially): **P.A. Browne** (2016). A comparison of the equivalent weights particle filter and the local ensemble transform Kalman filter in application to the barotropic vorticity equation. *Tellus A*. In press

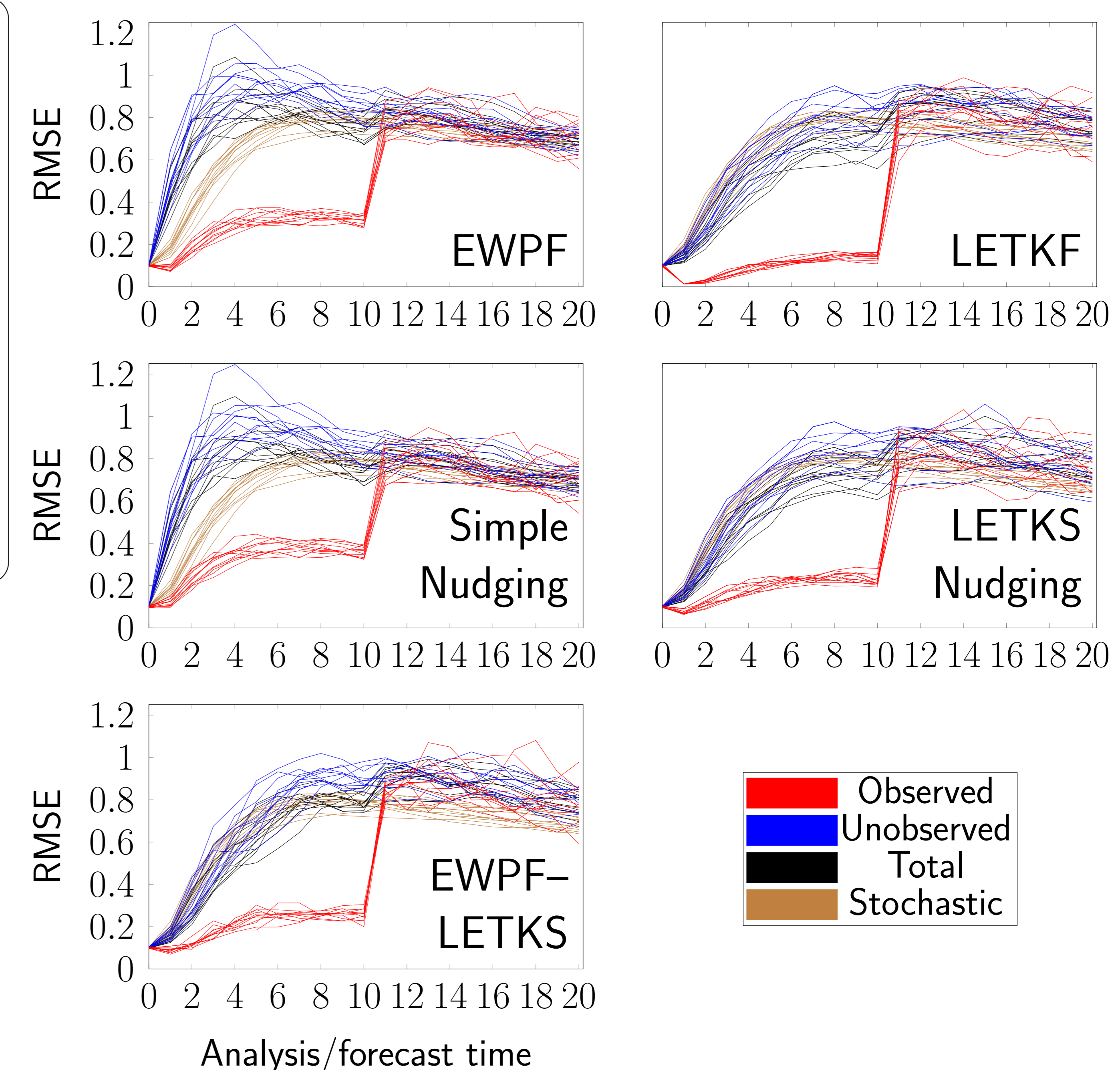


Figure 3. Root mean squared errors sampled over different parts of the domain