

# Localisation and Balance in an Ensemble Kalman Filter

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# Outline

- Illustrate that covariance localisation can produce imbalance.
- Present new localisation formulations.
- Show results from identical-twin experiments in a global shallow-water model, comparing localisation formulations.
- Conclusions.

# Covariance Localisation – Why?

$$\mathbf{X}^a = \mathbf{X}^b + \mathbf{P}_e^b \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^b \mathbf{H}^T + \mathbf{R})^{-1} (\mathbf{Y} - \mathbf{H} \mathbf{X}^b)$$

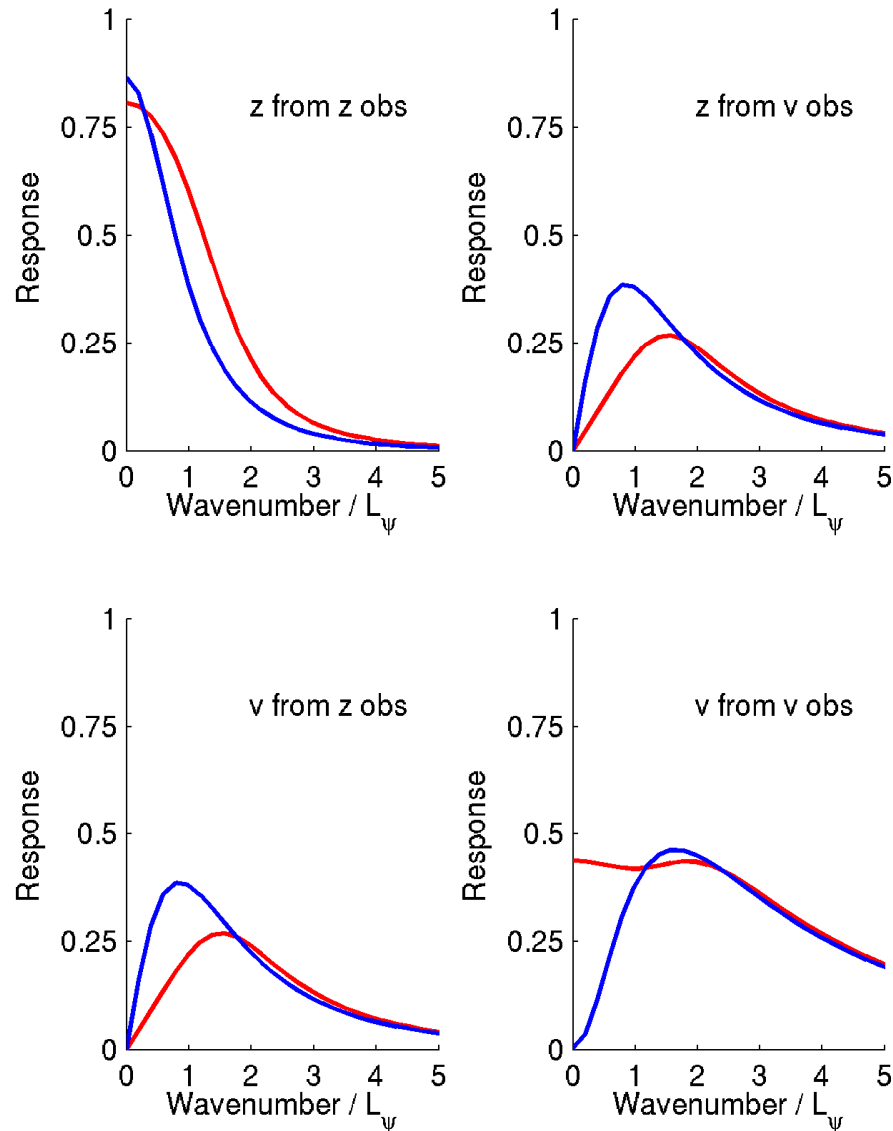
$$\mathbf{X}^a = \mathbf{X}^b + [(\mathbf{C} \mathbf{H}^T) \circ (\mathbf{P}_e^b \mathbf{H}^T)] [(\mathbf{H} \mathbf{C} \mathbf{H}^T) \circ (\mathbf{H} \mathbf{P}_e^b \mathbf{H}^T) + \mathbf{R}]^{-1} (\mathbf{Y} - \mathbf{H} \mathbf{X}^b)$$

- $\mathbf{P}_e^b$  is estimated by a Monte Carlo process, so is subject to sampling error.
- Localisation
  - Removes spurious long-range correlations
  - Corrects erroneous response of analysis mean and variance to distant observations.
  - $\text{CoP}_e^b$  is still a covariance (Gaspari and Cohn, 1999)
- $\text{rank}(\mathbf{P}_e^b) = \mathcal{O}(100)$  is too small, gives poor fit of analysis to observations
  - Localisation increases rank of  $\mathbf{P}_e^b$
- Eliminates analysis discontinuities between sub-domains.

# Disadvantages of Localisation

- Weakens balances encoded in  $\mathbf{P}_e^b$ .
  - Exact balances correspond to the null space of  $\mathbf{P}_e^b$ . (Lorenc 2003)
  - Advantage of EnKF advantage is better covariances - localisation reduces this.
  - Canadian experience confirms balance problems. (Houtkamer et al 2005)
- Makes analysis suboptimal.
  - Changes the gain matrix.
- *ad hoc* procedure which requires tuning.

# Localisation and Balance



- One-dimensional cyclic domain.
- Very dense observations.
- Bivariate  $(z, v)$  analysis using OI covariance model with exact geostrophic balance, nondivergent.
- Plotted response is from diagonals of Fourier transformed gain matrix. (c.f. Daley 1991)
- **Blue:** global analysis, standard OI covariance model.
- **Red:** localised covariance model (support =  $6L$  ).

# Improved Localisation

Write the velocity components in terms of streamfunction  $\psi$  and velocity potential  $\chi$ :

$$\begin{aligned}u &= -\frac{\cos \phi}{a} \frac{\partial \psi}{\partial \phi} + \frac{1}{a} \frac{\partial \chi}{\partial \lambda} \\v &= \frac{1}{a} \frac{\partial \psi}{\partial \lambda} + \frac{\cos \phi}{a} \frac{\partial \chi}{\partial \phi}\end{aligned}\tag{1}$$

The two-point covariance  $\langle \psi_1, \psi_2 \rangle$  is

$$F_{\psi,\psi} = F_{\psi,\psi}(\lambda_1, \phi_1, \lambda_2, \phi_2) = \langle \psi(\lambda_1, \phi_1), \psi(\lambda_2, \phi_2) \rangle = \langle \psi_1, \psi_2 \rangle\tag{2}$$

and similarly for  $F_{\psi,\chi}$ ,  $F_{\chi,\psi}$  and  $F_{\chi,\chi}$ . Then the two-point covariance  $F_{u,u} = \langle u_1, u_2 \rangle$  becomes

$$\begin{aligned}F_{u,u} &= \frac{\cos \phi_1 \cos \phi_2}{a^2} \frac{\partial}{\partial \phi_1} \frac{\partial}{\partial \phi_2} F_{\psi,\psi} - \frac{\cos \phi_1}{a^2} \frac{\partial}{\partial \phi_1} \frac{\partial}{\partial \lambda_2} F_{\psi,\chi} \\&\quad - \frac{\cos \phi_2}{a^2} \frac{\partial}{\partial \lambda_1} \frac{\partial}{\partial \phi_2} F_{\chi,\psi} + \frac{1}{a^2} \frac{\partial}{\partial \lambda_1} \frac{\partial}{\partial \lambda_2} F_{\chi,\chi}\end{aligned}\tag{3}$$

# Improved Localisation (cont'd)

Let  $C$  be a localisation function, and apply the localisation by replacing  $F_{\psi,\psi} \leftarrow CF_{\psi,\psi}$  in (3) and similarly for  $F_{\psi,\chi}$ ,  $F_{\chi,\psi}$  and  $F_{\chi,\chi}$ . Then the *localised* two-point covariance  $\langle u, u \rangle_L$  becomes

$$\begin{aligned} \langle u_1, u_2 \rangle_L &= C_{o,o} F_{u,u} \\ &+ C_{x,o} F_{\chi,u} - C_{y,o} F_{\psi,u} + C_{o,x} F_{u,\chi} - C_{o,y} F_{u,\psi} \\ &+ C_{x,x} F_{\chi,\chi} - C_{x,y} F_{\chi,\psi} - C_{y,x} F_{\psi,\chi} + C_{y,y} F_{\psi,\psi} \end{aligned} \quad (4)$$

The first term on the RHS is the usual localisation. The  $F$ 's can all be calculated from the ensemble in the usual way. (The EnKF will need to read in  $\psi$  and  $\chi$  in addition to  $u$  and  $v$ .)

# Improved Localisation (cont'd)

The  $C$ 's are spatial derivatives of the localiser:

$$\begin{aligned}C_{o,o} &= C = C(r(\lambda_1, \phi_1, \lambda_2, \phi_2)) \\C_{x,o} &= \frac{1}{a} \frac{\partial C}{\partial \lambda_1} \\C_{y,o} &= \frac{\cos \phi_1}{a} \frac{\partial C}{\partial \phi_1} \\C_{o,x} &= \frac{1}{a} \frac{\partial C}{\partial \lambda_2} \\C_{o,y} &= \frac{\cos \phi_2}{a} \frac{\partial C}{\partial \phi_2} \\C_{x,x} &= \frac{1}{a^2} \frac{\partial^2 C}{\partial \lambda_1 \partial \lambda_2} \\C_{x,y} &= \frac{\cos \phi_2}{a^2} \frac{\partial^2 C}{\partial \lambda_1 \partial \phi_2} \\C_{y,x} &= \frac{\cos \phi_1}{a^2} \frac{\partial^2 C}{\partial \phi_1 \partial \lambda_2} \\C_{y,y} &= \frac{\cos \phi_1 \phi_2}{a^2} \frac{\partial^2 C}{\partial \phi_1 \partial \phi_2}\end{aligned} \tag{5}$$

# Improved Localisation (cont'd)

Similarly, the localised versions of the other covariances are

$$\begin{aligned}\langle u_1, v_2 \rangle_L &= C_{o,o}F_{u,v} \\ &+ C_{x,o}F_{\chi,v} - C_{y,o}F_{\psi,v} + C_{o,x}F_{u,\psi} + C_{o,y}F_{u,\chi} \\ &- C_{x,x}F_{\psi,\psi} + C_{x,y}F_{\chi,\chi} - C_{y,x}F_{\psi,\psi} + C_{y,y}F_{\chi,\psi}\end{aligned}\quad (6)$$

$$\begin{aligned}\langle v_1, u_2 \rangle_L &= C_{o,o}F_{v,u} \\ &+ C_{x,o}F_{\psi,u} + C_{y,o}F_{\chi,u} + C_{o,x}F_{v,\chi} - C_{o,y}F_{v,\psi} \\ &+ C_{x,x}F_{\psi,\chi} - C_{x,y}F_{\psi,\psi} + C_{y,x}F_{\chi,\chi} - C_{y,y}F_{\chi,\psi}\end{aligned}\quad (7)$$

$$\begin{aligned}\langle v_1, v_2 \rangle_L &= C_{o,o}F_{v,v} \\ &+ C_{x,o}F_{\psi,v} + C_{y,o}F_{\chi,v} + C_{o,x}F_{v,\psi} + C_{o,y}F_{v,\chi} \\ &+ C_{x,x}F_{\psi,\psi} + C_{x,y}F_{\psi,\chi} + C_{y,x}F_{\chi,\psi} + C_{y,y}F_{\chi,\chi}\end{aligned}\quad (8)$$

$$\langle u_1, z_2 \rangle_L = C_{o,o}F_{u,z} + C_{x,o}F_{\chi,z} - C_{y,o}F_{\psi,z}\quad (9)$$

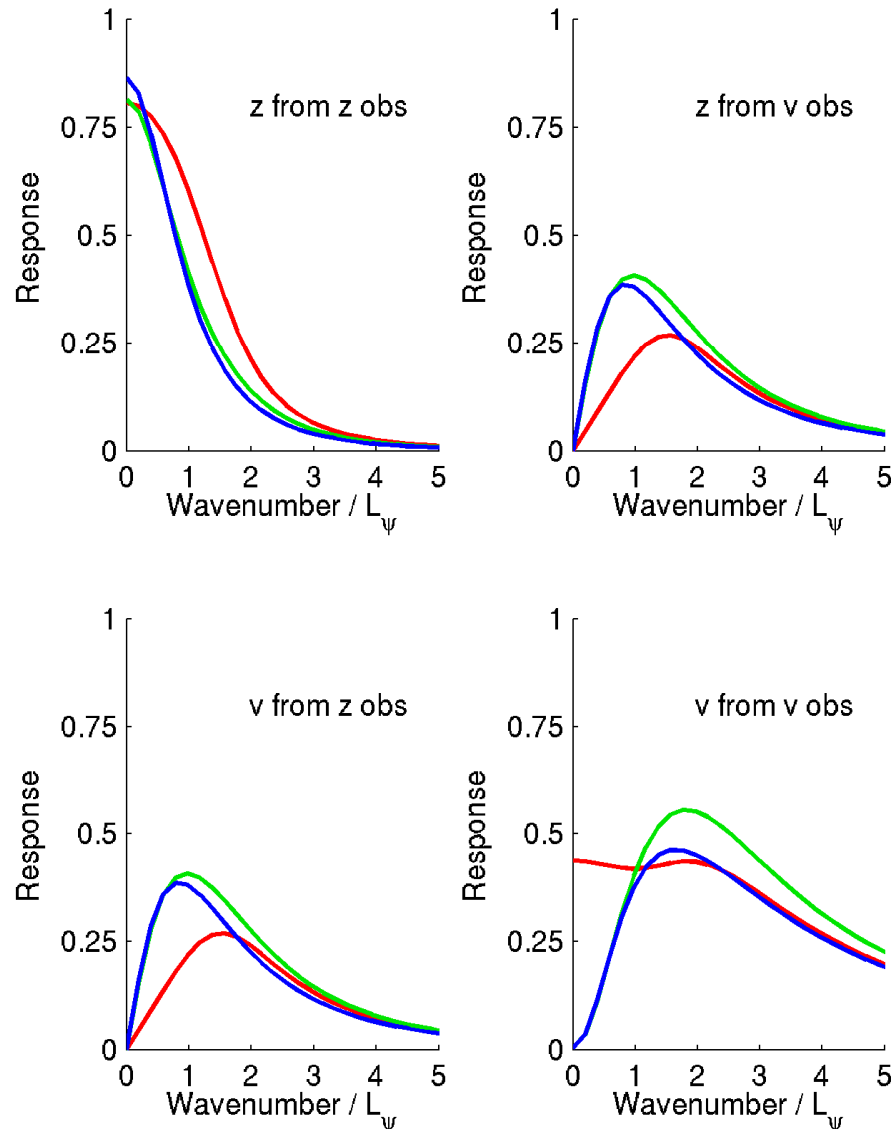
$$\langle v_1, z_2 \rangle_L = C_{o,o}F_{v,z} + C_{x,o}F_{\psi,z} + C_{y,o}F_{\chi,z}\quad (10)$$

$$\langle z_1, u_2 \rangle_L = C_{o,o}F_{z,u} + C_{o,x}F_{z,\chi} - C_{o,y}F_{z,\psi}\quad (11)$$

$$\langle z_1, v_2 \rangle_L = C_{o,o}F_{z,v} + C_{o,x}F_{z,\psi} + C_{o,y}F_{\psi,z}\quad (12)$$

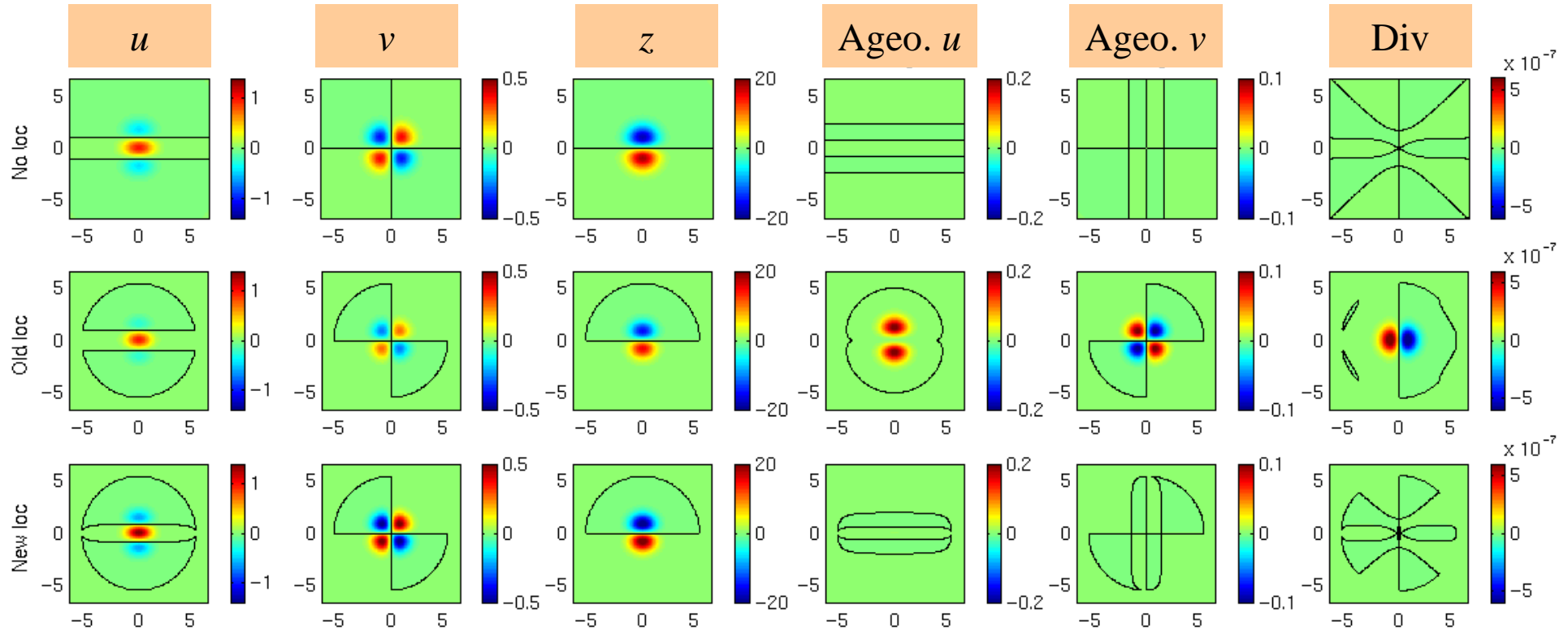
$$\langle z_1, z_2 \rangle_L = C_{o,o}F_{z,z}\quad (13)$$

# Analysis response using new localisation



- **Blue:** standard OI covariance model.
  - **Red:** localised covariance model.
  - **Green:** covariances localised using new method.
- 
- **Analysis response is geostrophically balanced and nondivergent.**

# Improved Localisation Example



- Analysis of single  $u$  observation, using idealised nondivergent and geostrophically balanced covariances.
- **Top:** No localisation. **Middle:** Standard localisation. **Bottom:** New localisation.

# Localise $\chi$ cross-covariances to zero

- Experience shows that  $z$ - $\chi$  and  $\psi$ - $\chi$  correlation often nearly zero (except in boundary layer).
- It is therefore sensitive to sampling error.
- Can repeat derivation with the  $\chi$  cross-covariances “localised” to zero.
- Slightly more complex expressions, EnKF now needs to read in rotational and divergent wind components.

# A cheap hybrid

- About half of the additional terms in the new formulation can be included in the observation-space to model-space covariances by changing model-space from  $(z, u, v)$  to  $(z, \psi, \chi)$ .

$$\mathbf{X}^a = \mathbf{X}^b + [(\mathbf{C}\mathbf{H}^T) \circ (\mathbf{P}_e^b \mathbf{H}^T)] [(\mathbf{H}\mathbf{C}\mathbf{H}^T) \circ (\mathbf{H}\mathbf{P}_e^b \mathbf{H}^T) + \mathbf{R}]^{-1} (\mathbf{Y} - \mathbf{H}\mathbf{X}^b)$$

$$\mathbf{P}_e^b \mathbf{H}^T = \frac{1}{N-1} \begin{bmatrix} \mathbf{X}^b - \bar{\mathbf{X}}^b \\ \mathcal{H}(\mathbf{X}^b) - \bar{\mathcal{H}}(\bar{\mathbf{X}}^b) \end{bmatrix}^T$$

$$\mathbf{H}\mathbf{P}_e^b \mathbf{H}^T = \frac{1}{N-1} \begin{bmatrix} \mathcal{H}(\mathbf{X}^b) - \bar{\mathcal{H}}(\bar{\mathbf{X}}^b) \\ \mathcal{H}(\mathbf{X}^b) - \bar{\mathcal{H}}(\bar{\mathbf{X}}^b) \end{bmatrix}^T$$

model space deviations  
 $(z, \psi, \chi)$

obs space deviations  $(z, u, v)$

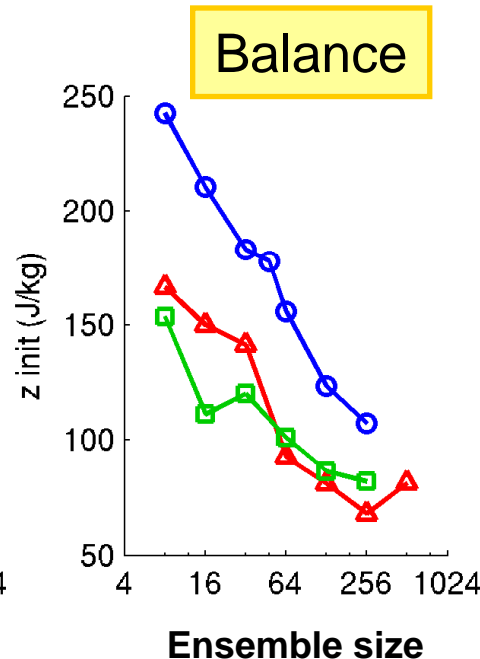
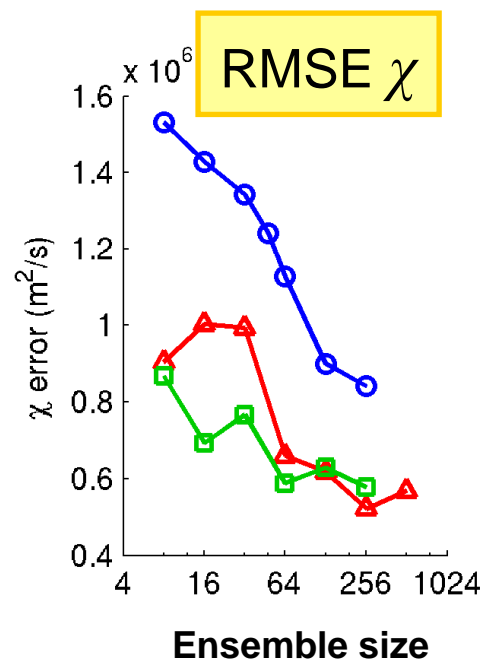
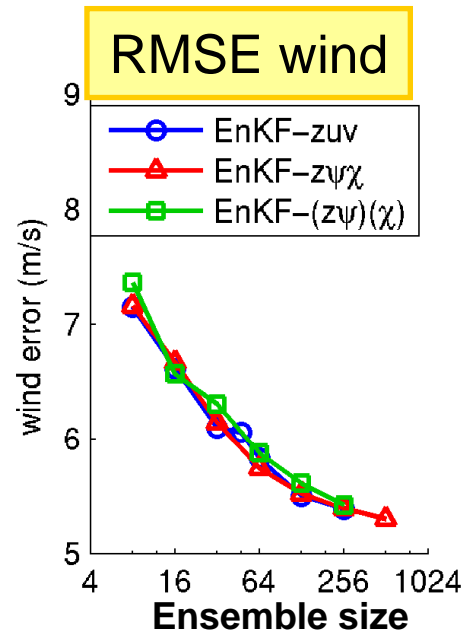
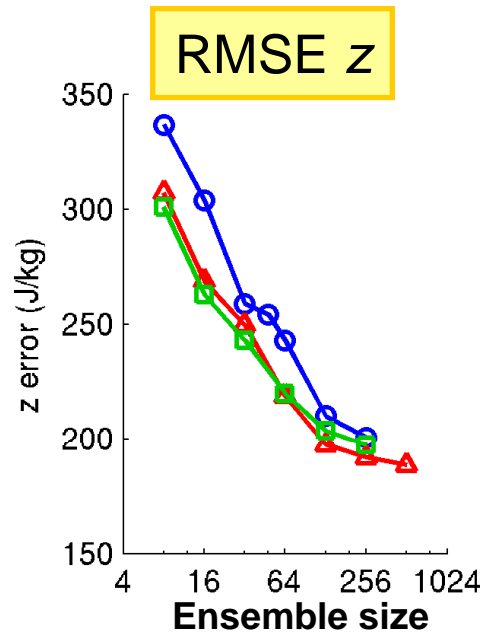
# Experimental Design

- Global spectral shallow water of Bourke (1972) at T31.
  - Accurate, non-dissipative, includes variable transforms  $(u, v)$  to/from  $(\psi, \chi)$ .
  - Simplest model to have main atmospheric balances.
  - Error-doubling time  $\sim 5$  days.
- Identical twin with 60-day truth run
  - Start from ERA-40 analysis for 500 hPa 18 Jan 1962.
  - Initial ensemble from a climate forecast beginning one year before.
  - Analyse last 20 days of assimilation cycle.
- Obs network based on 50% of global radiosondes, observe  $(z, u, v)$  12-hourly.
  - Height error = 100 J/kg, wind error = 5 m/s.
- Standard perturbed-obs EnKF with covariance inflation, various localisations, various ensemble sizes.
  - Tune localisation length and covariance inflation to minimise errors.
  - Verify spread using rank histograms.

# Four localisations

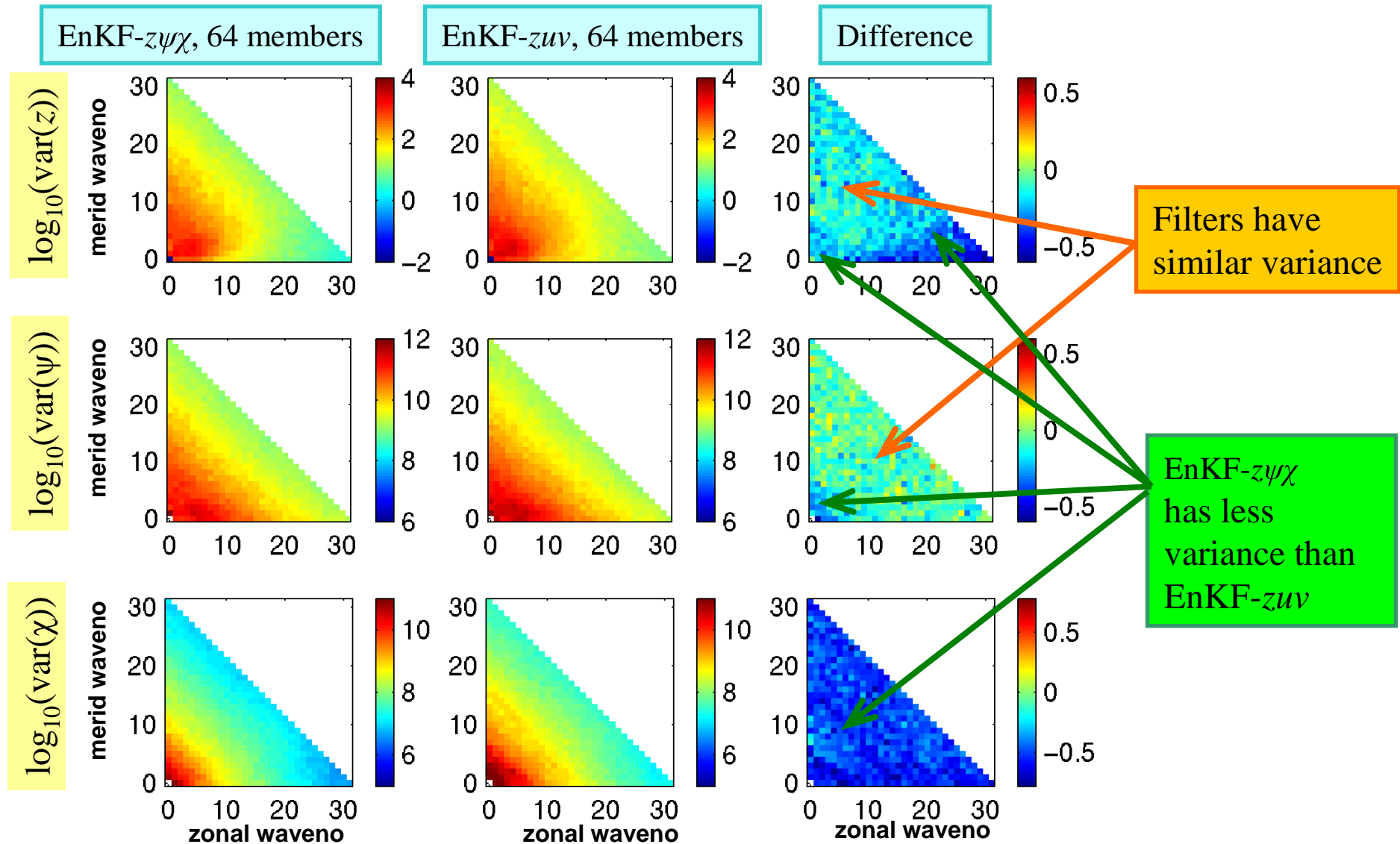
- EnKF- $zuv$  : standard localisation.
- EnKF- $z\psi\chi$  : new localisation in  $(z, \psi, \chi)$  space.
- EnKF-hybrid : partial implementation.
- EnKF- $(z\psi)(\chi)$  : zeroes out  $z - \chi$  and  $\psi - \chi$  correlations.

# Performance



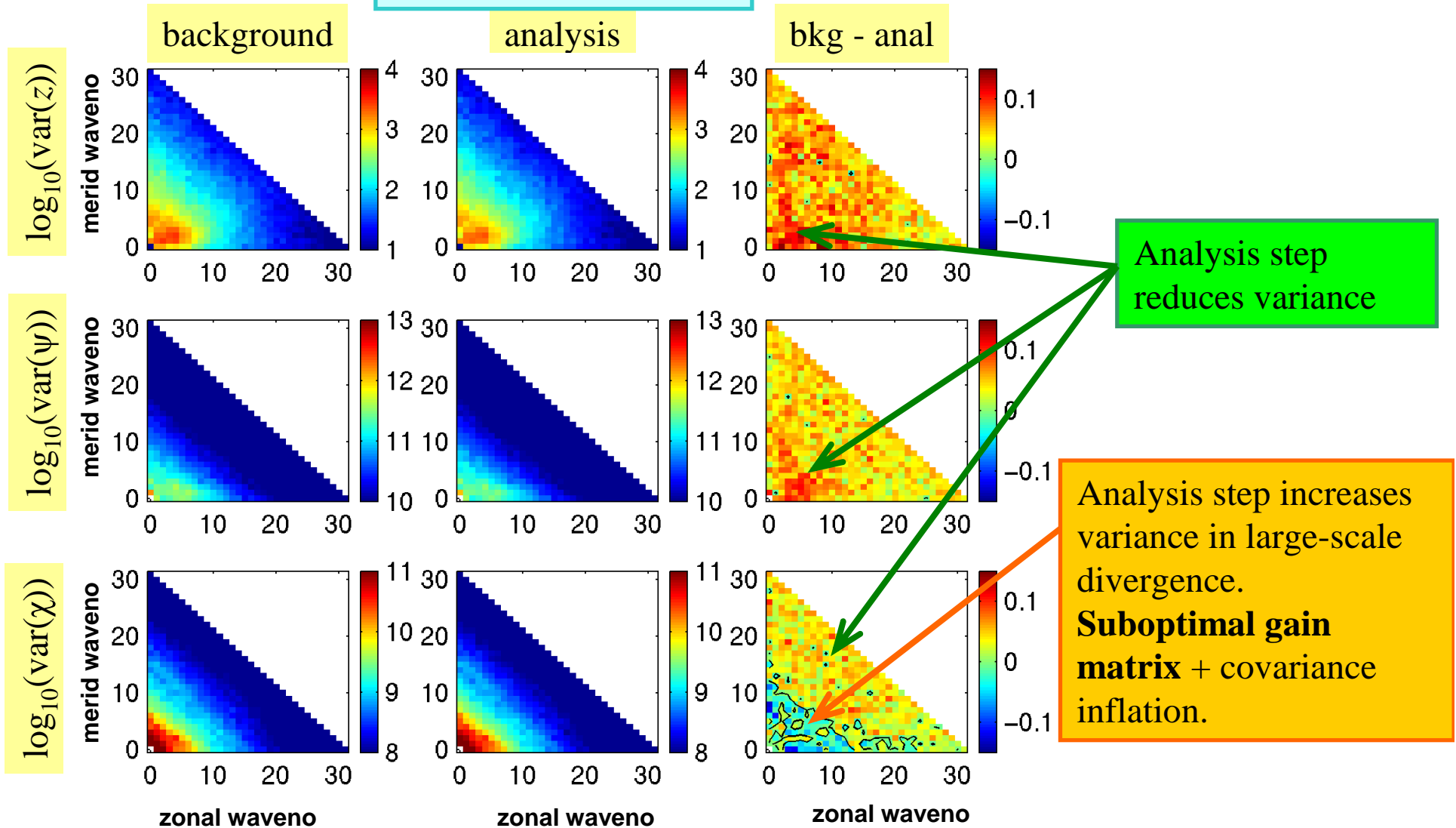
- Scores improve with bigger ensemble.
  - Better estimation of errors.
- New localisations better than old for height.
  - Equivalent to 50-100% bigger ensemble.
  - Similar for wind.
  - Much better for  $\chi$ .
- New localisations better balanced.
- EnKF-(z $\psi$ )( $\chi$ ) best for small ensembles, EnKF-z $\psi\chi$  best for large.
- Hybrid is intermediate (not shown).

# Background variance in spectral space

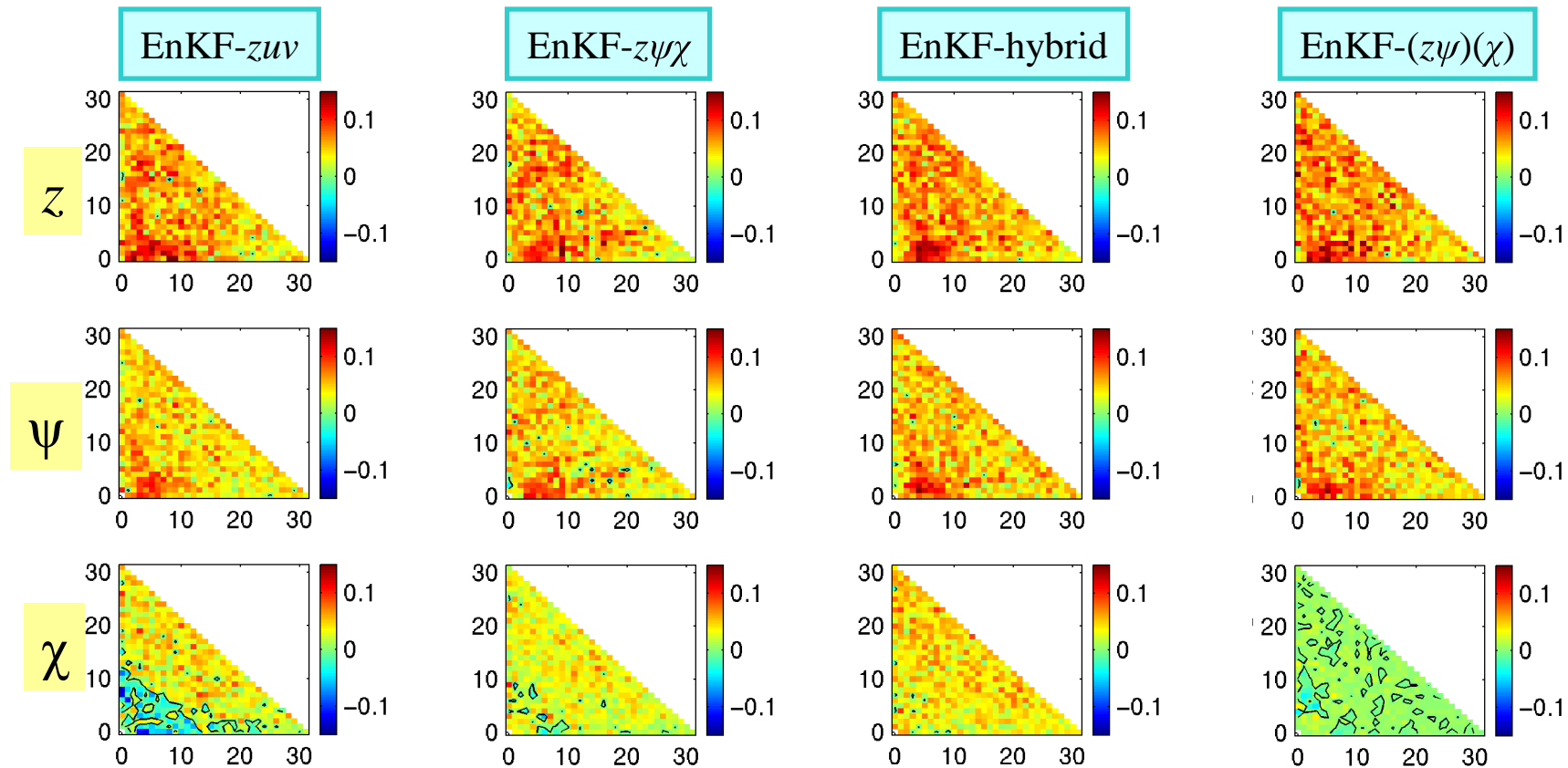


# Variance change, background - analysis

EnKF-*zuv*, 64 members

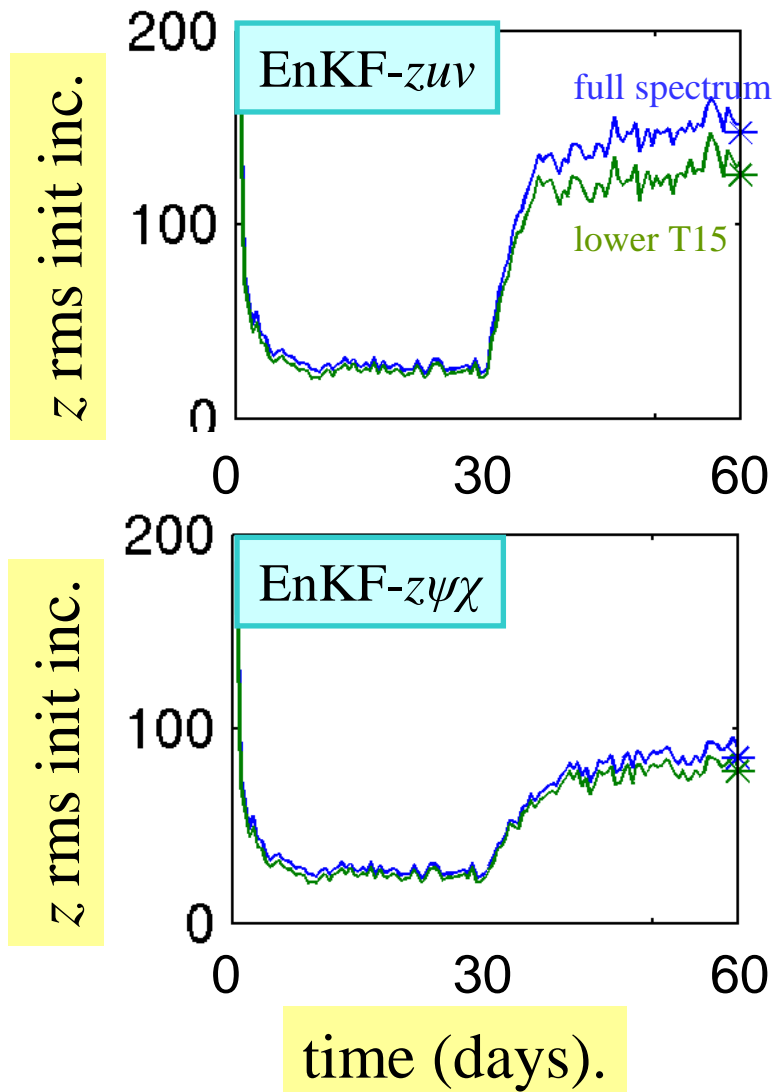


# Variance change, background - analysis



- New localisations give
  - Reduced evidence of suboptimal analysis.
  - Somewhat less analysis impact on  $\chi$  variance.
- This diagnostic includes impact of covariance inflation.

# Balance



- Diagnostic is rms(analysis – initialised analysis).
- Switched normal-mode initialisation off in filter at day 30.
- Both filters asymptote to new level of imbalance.
- EnKF- $zuv$  has faster imbalance growth, to higher level, with more small-scale imbalance.
- Positive feedback: The analysis will project more of the observations onto the modes with the most variance.
- Similar results with other balance diagnostics (e.g. rms tendencies).

# Conclusions

- New localisation is better (produce more accurate and better balanced analyses).
- Removing  $\chi$  ensemble cross-covariances a further improvement for small ensembles.
- Hybrid realises some of the gain of full scheme.
- Disadvantages of new localisations:
  - Marginally inferior when analysis cycle includes initialisation.
  - Due to additional sampling error (more ensemble covariances calculated)?
  - Extra cost (small except in highly sequential schemes).
  - Increases response to wind observations (can be compensated for).
- Positive feedback on unbalanced modes in EnKF.
- Tested in perturbed-observation-single-EnKF, but applicable to other flavours (except ETKF).