

Predicting forecast uncertainty with singular vectors and ensemble-derived analysis error covariances

Mark Buehner

Meteorological Service of Canada
Dorval, Quebec

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Introduction

- several applications require accurate prediction of forecast uncertainty or its origins: ensemble prediction, observation targeting, data assimilation, data QC
- singular vectors (SVs) identify the most unstable subspace of the (linearized) forecast model according to specified norms at initial and optimization times
- current applications (ECMWF, NRL) use SVs with total energy (TE) norms for ensemble prediction and observation targeting
- if initial-time norm is inverse analysis error covariances, SVs optimally describe forecast uncertainty at optimization-time (if model error negligible, TL error growth)
- EnKF provides estimate of flow-dependent analysis error covariances via Monte Carlo simulation to model effect of all error sources in forecast-analysis cycle
- goal: evaluate use of these flow-dependent or (more easily obtained) stationary analysis error cov from ensemble approaches to define initial-time norm for SVs

Outline

- SV theory and calculation
- approaches for estimating analysis error cov with:
 1. EnKF (flow-dependent)
 2. perturbed 3d-var (stationary)
- impact of covariance norms on structure of SVs (vs TE)
- impact of norms on distribution of sensitive regions (targeting application)
- in context of ensemble prediction (fraction of explained NH fcterr), compare:
 1. SVs
 2. EnKF members
 3. random perturbations
- conclusions and future directions

SV Theory and Calculation

- given an analysis, find perturbation \mathbf{x}_0 that maximizes ratio:

$$\frac{\mathbf{x}_t^T \mathbf{W}_t \mathbf{x}_t}{\mathbf{x}_0^T \mathbf{W}_0 \mathbf{x}_0}, \text{ where } \mathbf{x}_t = \mathbf{M} \mathbf{x}_0$$

- reformulate in terms of control variable γ , where $\mathbf{x}_0 = \mathbf{W}_0^{-1/2} \gamma$

- results in eigenvalue problem: $\mathbf{W}_0^{-T/2} \mathbf{M}^T \mathbf{W}_t \mathbf{M} \mathbf{W}_0^{-1/2} \gamma_k = \sigma_k^2 \gamma_k$

- \mathbf{W}_t : final-time norm typically TE, restricted to geographical region of interest
- \mathbf{W}_0 : initial-time norm commonly TE (TESV), but...
- when $\mathbf{W}_0 = \mathbf{P}_a^{-1}$, the final-time SVs are the leading eigenvectors of $\mathbf{M} \mathbf{P}_a \mathbf{M}^T$ (predictability error \equiv forecast error for perfect model)

Experimental Setup

- like 4d-var, transfer data between 3d-var software (Lanczos solver and norms) and GEM (TLM and ADJ):

3d-var	GEM
$\times \mathbf{W}_0^{-1/2}$	
\longrightarrow	run tlm
$\times \mathbf{W}_t$	\longleftarrow
\longrightarrow	run adj
$\times \mathbf{W}_0^{-T/2}$	\longleftarrow

- TLM and ADJ at 120x60x28L with simplified linear physics package
- final time norm is TE north of 30°N, 48 hr optimization time interval
- SVs computed each day, 1-14 Dec, 2003
- different initial time (global) norms used :
 - total energy (as commonly used at several NWP centres)
 - inverse analysis error covariances (stationary and flow-dependent)

Approaches for Sampling Analysis Error

Kalman Filter equations: $\mathbf{P}_a = (\mathbf{I} - \mathbf{K}\mathbf{H})\mathbf{P}_f$, $\mathbf{P}_f = \mathbf{M}\mathbf{P}_a\mathbf{M}^T + \mathbf{Q}$

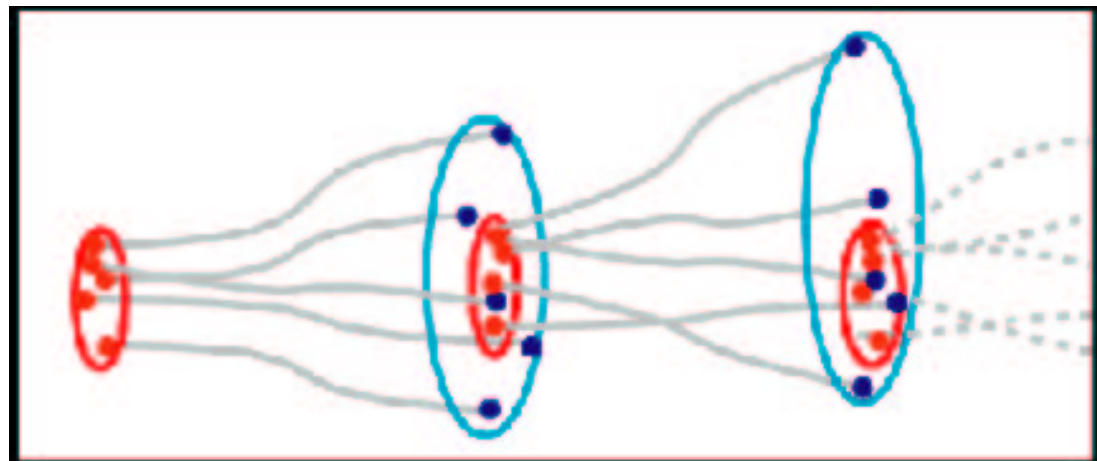
- not feasible for realistic model, only valid for linear model, little knowledge of \mathbf{Q}
- approx with Monte Carlo simulation approaches (perturb uncertain quantities):

EnKF: different ensemble of 128 error samples every 6h

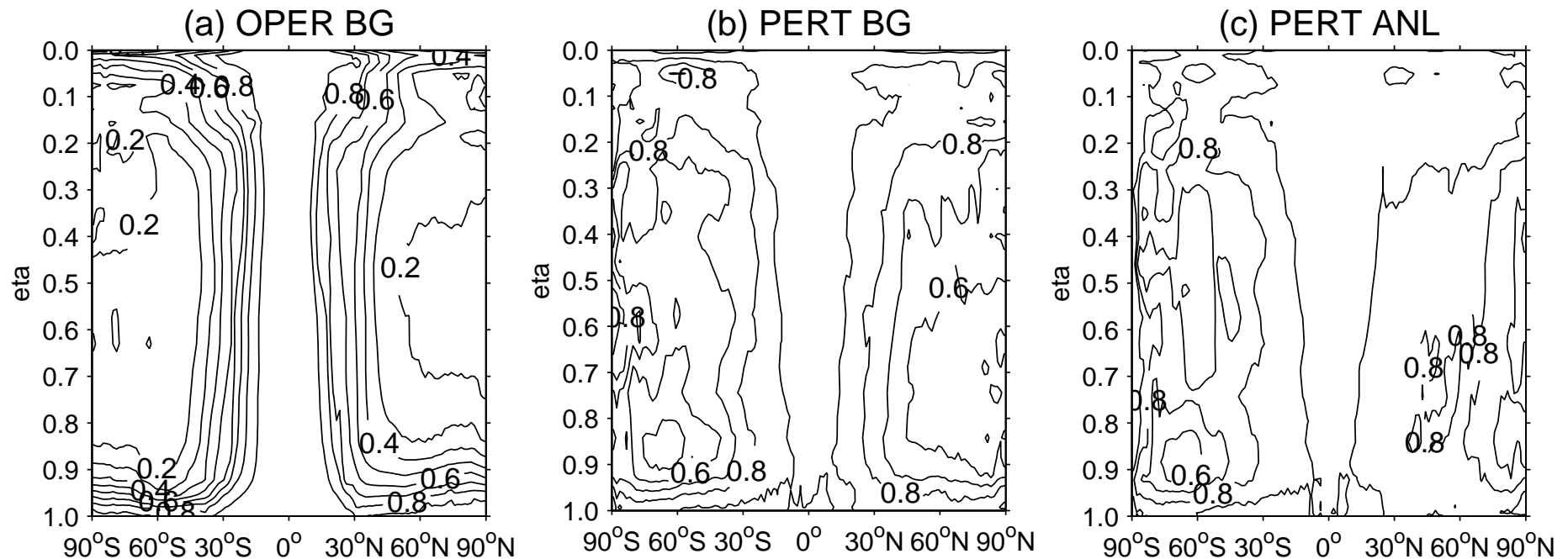
- analysis uses perturbed observations and flow-dependent ensemble covariances
- model error: random perturbations with similar cov as \mathbf{B} in oper. 3d-var
- for SV norm: deviations of analyses from ensemble mean: flow-dependent \mathbf{P}_a

Perturbed 3d-var: single ensemble of 138 error samples from 4 week period

- analysis uses perturbed observations and 3d-var (ECMWF; EPS at CMC)
- model error: random perturbations from adaptively scaled \mathbf{B}
- for SV norm: differences between analyses from perturbed and unperturbed exp'ts: stationary \mathbf{P}_a



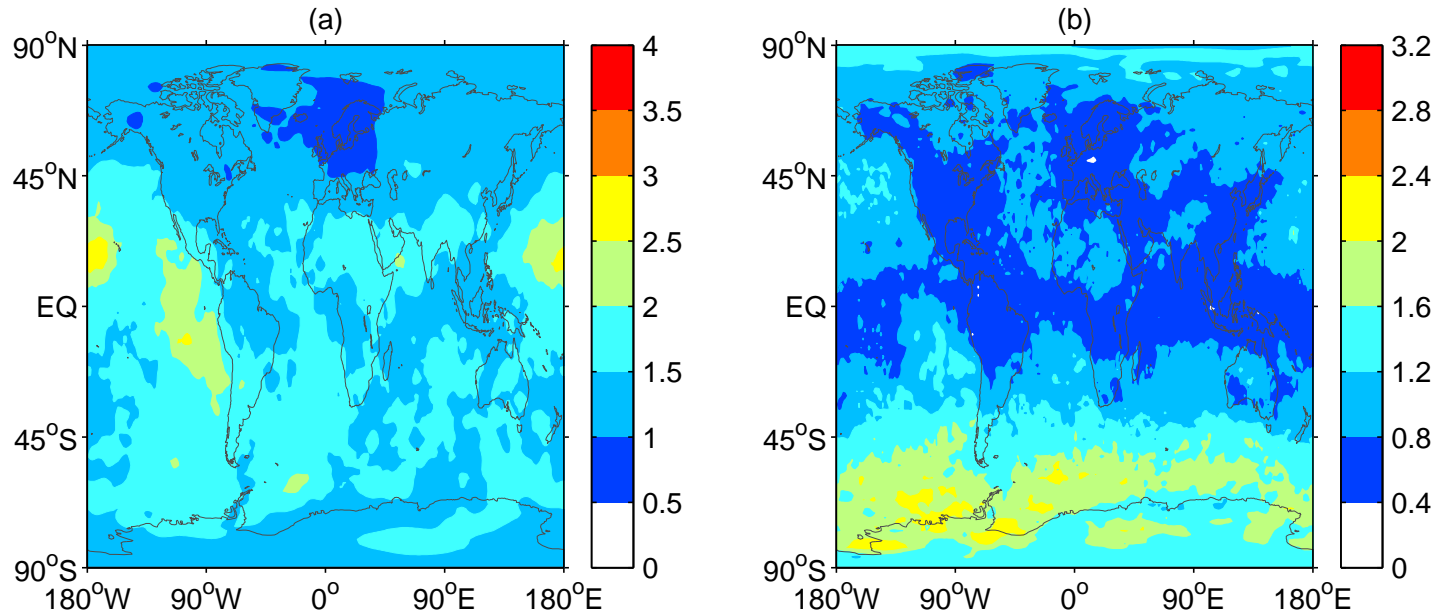
Geostrophic Balance in Perturbed 3d-var Covariances



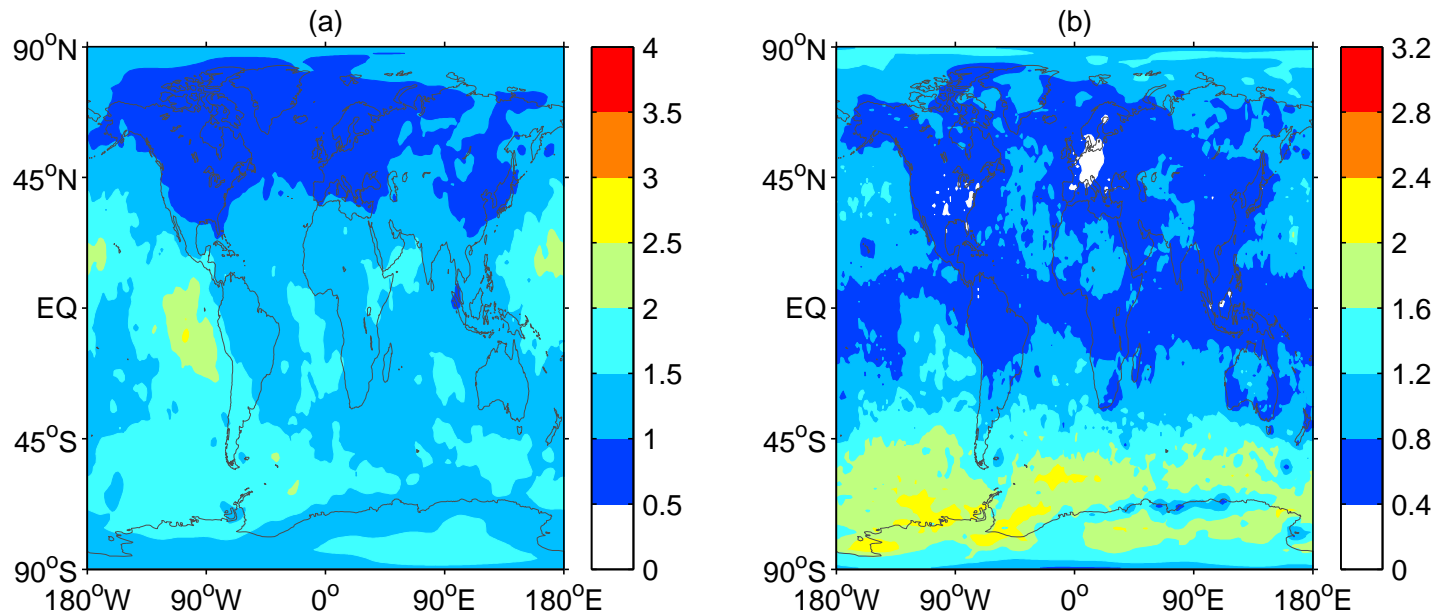
- ratio of unbalanced T to total T variance
- in operational BG covariances T almost completely balanced with winds
- less mass/wind balance in perturbed 3d-var (and EnKF) BG covariances
- analysis error even less balanced than background error

Std Dev of (a) Ψ at 250hPa and (b) T at 500hPa

BG error cov from perturbed 3d-var (138 error samples from 4 week period)

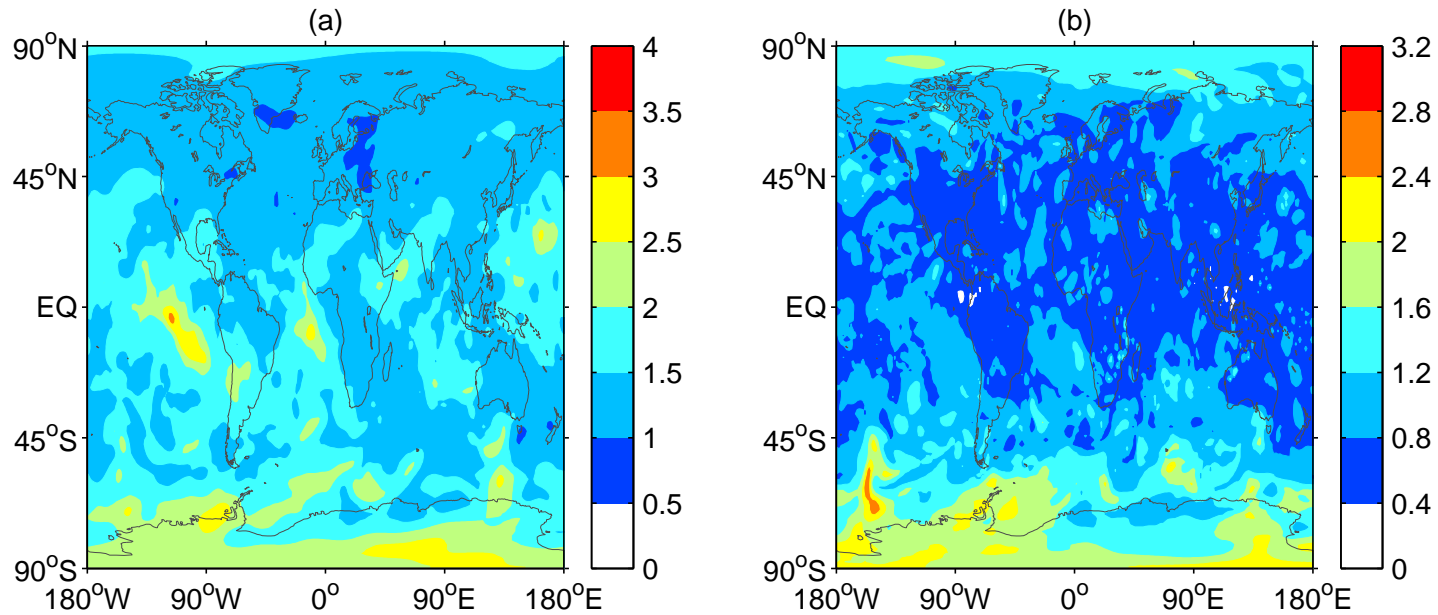


Analysis error cov from perturbed 3d-var (138 error samples from 4 week period)

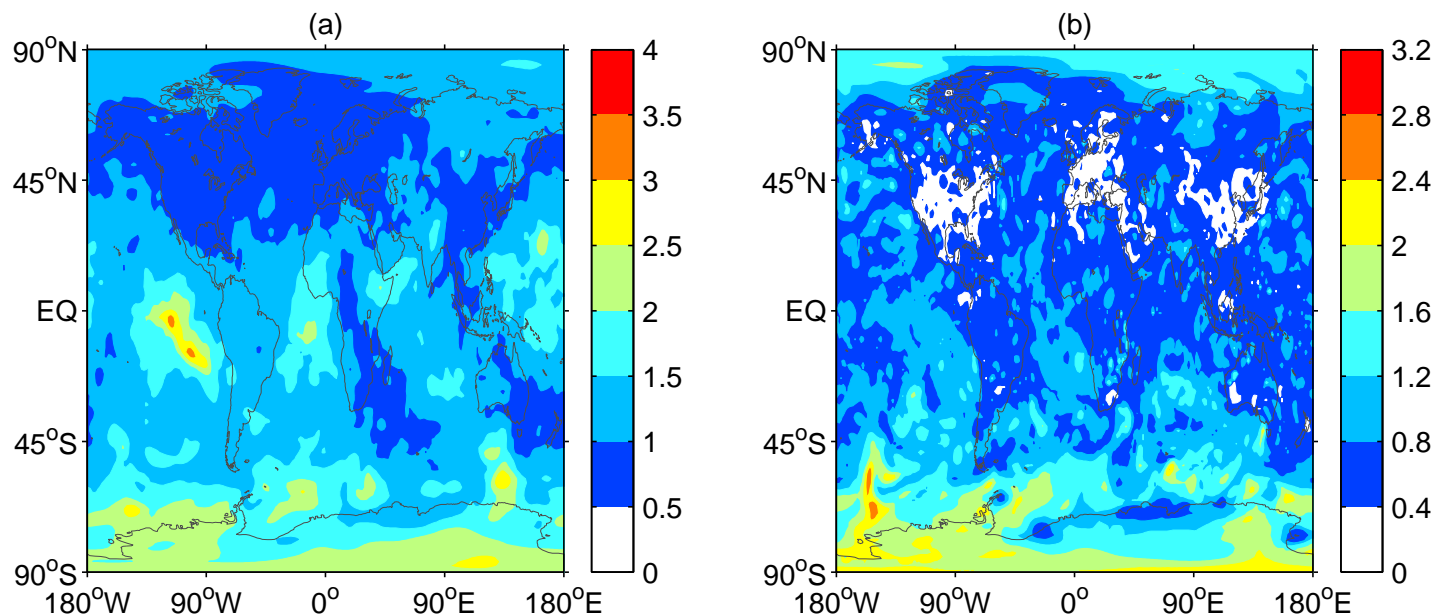


Std Dev of (a) Ψ at 250hPa and (b) T at 500hPa (2)

BG error cov from EnKF (128 error samples valid 2003121212)



Analysis error cov from EnKF (128 error samples valid 2003121212)

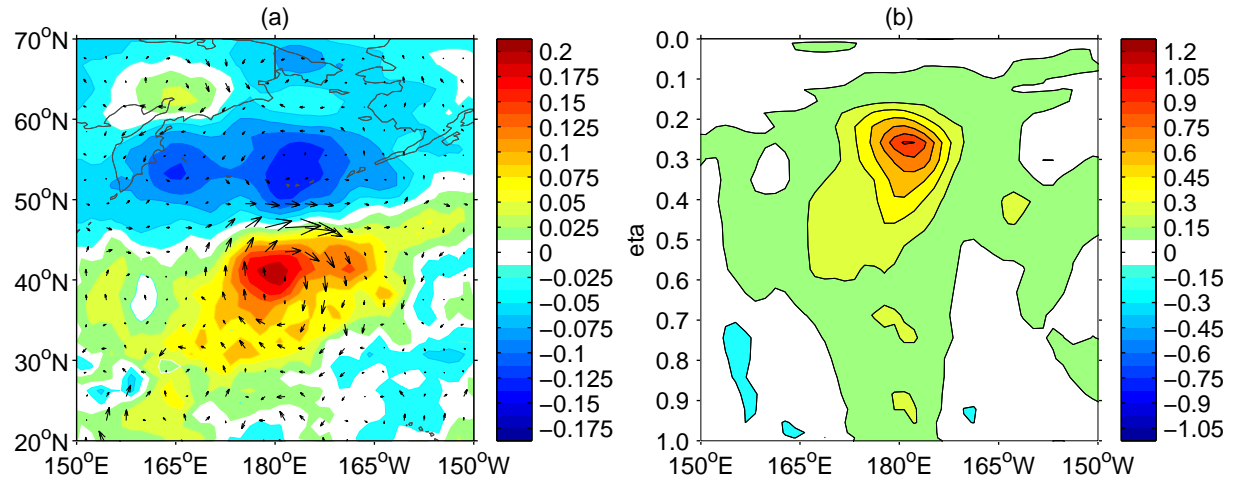


Estimation of Error Correlations

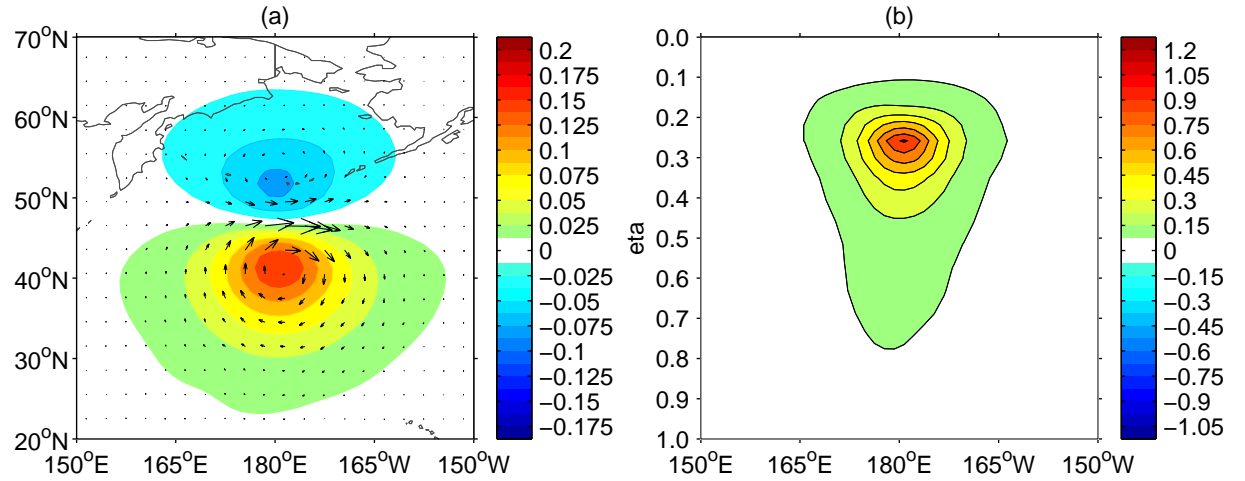
- cannot accurately estimate full correlation matrix (rank $O(10^6)$) from small number of samples ($O(10^2)$)
- therefore must impose constraints on structure of correlations (extra information)
- operational BG error correlations assume homogeneity/isotropy
- alternatively, we can spatially localize horizontal and vertical correlations (currently done in EnKF analysis and in experimental 3d-var)
- demonstrate by plotting BG error covariances of U @ 250 hPa with wind and GZ (next slide)

Estimation of Error Correlations (2)

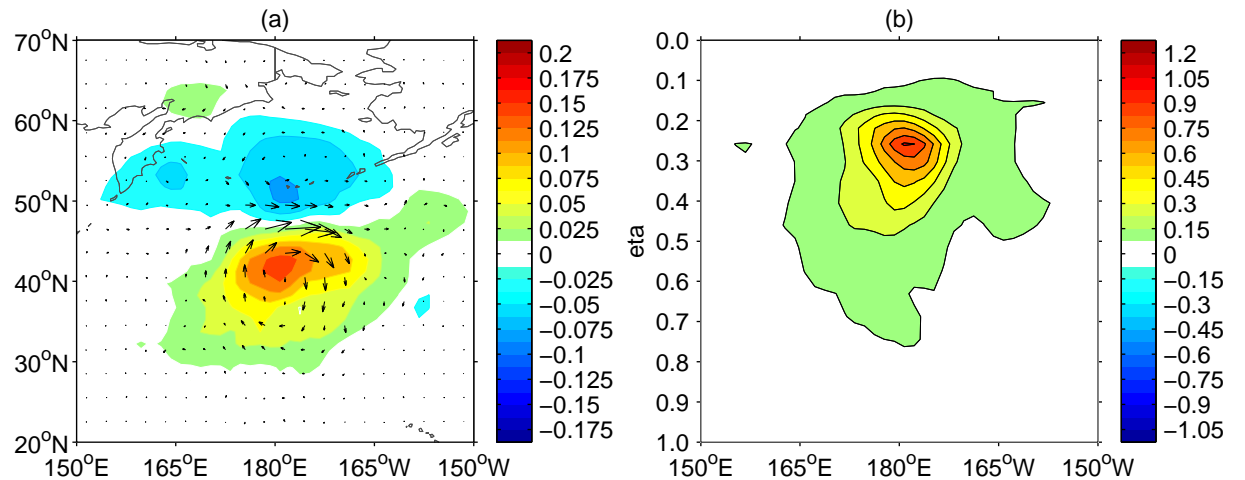
Original sample estimate of correlations:



Homogeneous/isotropic correlations:



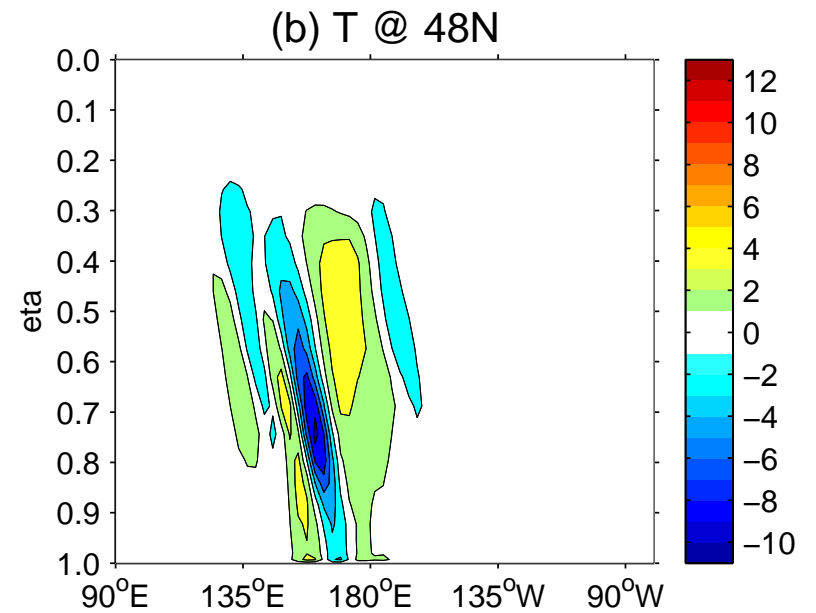
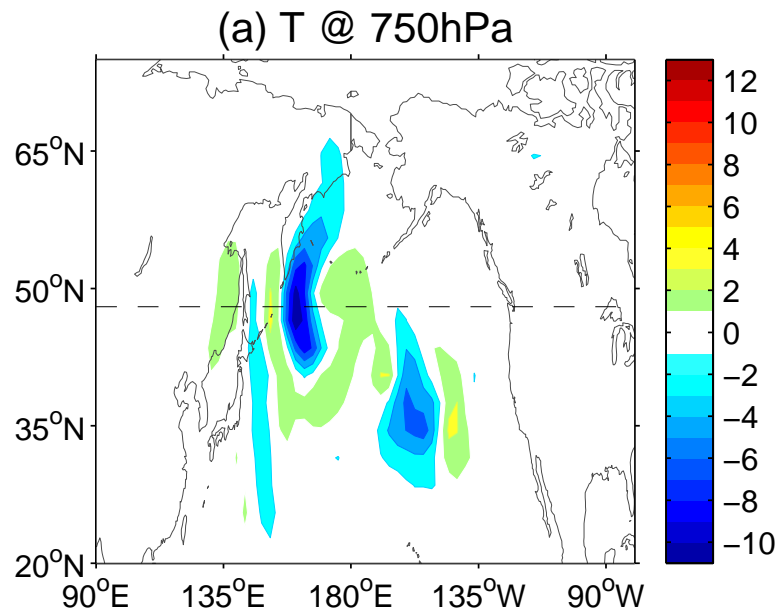
Spatially localized correlations:



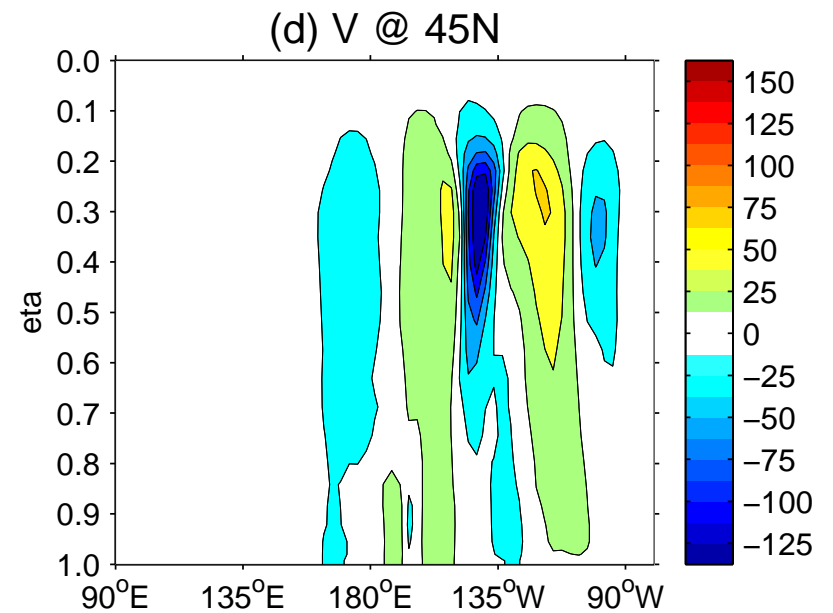
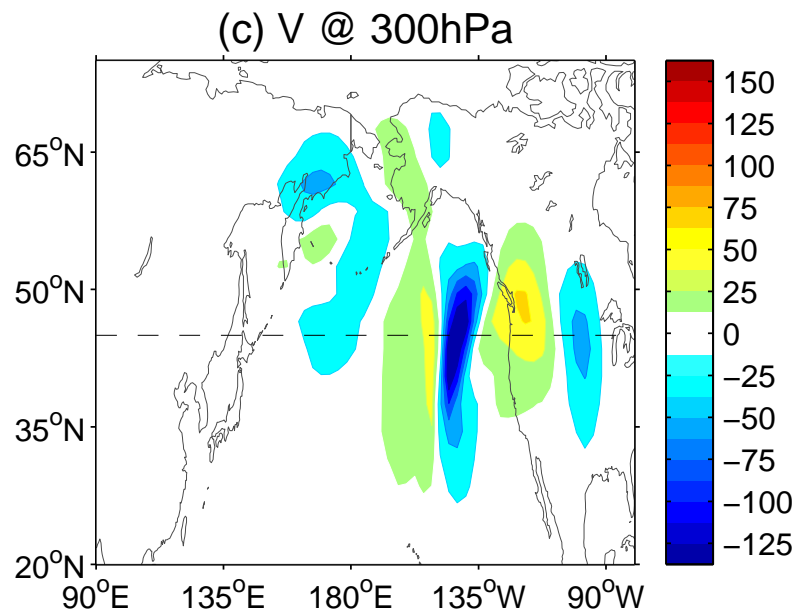
Impact of Initial-time Norm on SV structure

Leading total energy SV on 2003121600

initial
time:



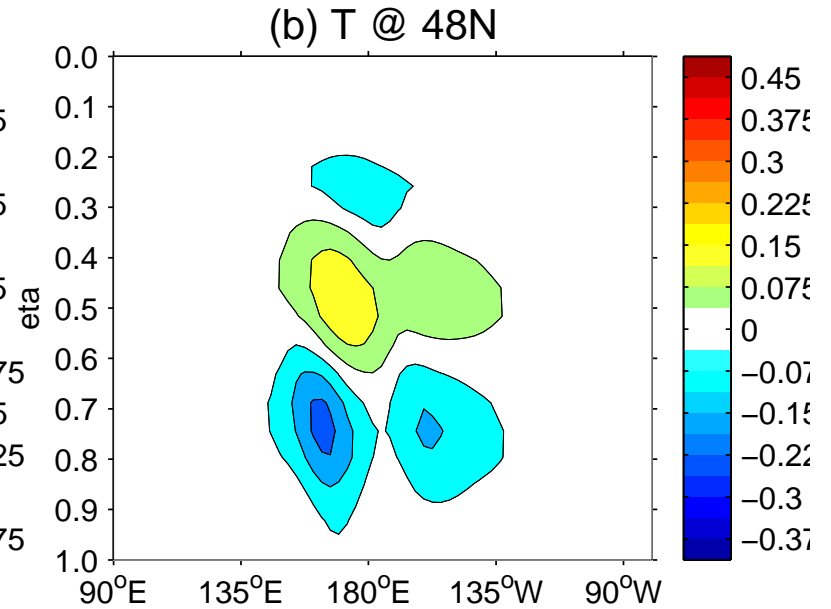
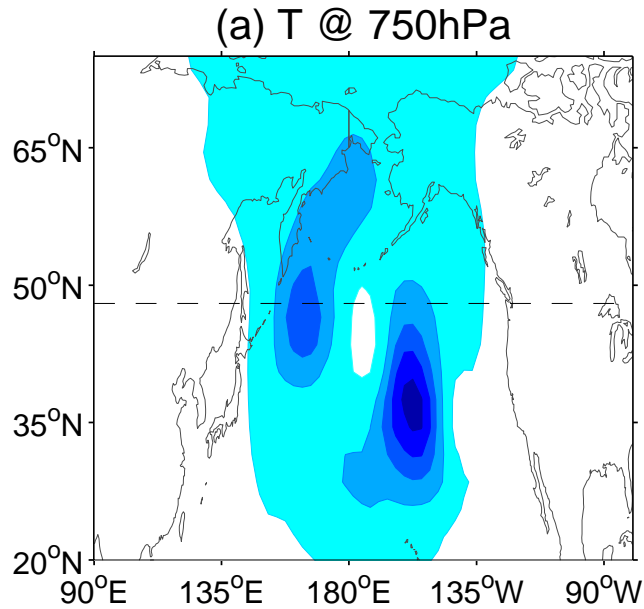
final
time:



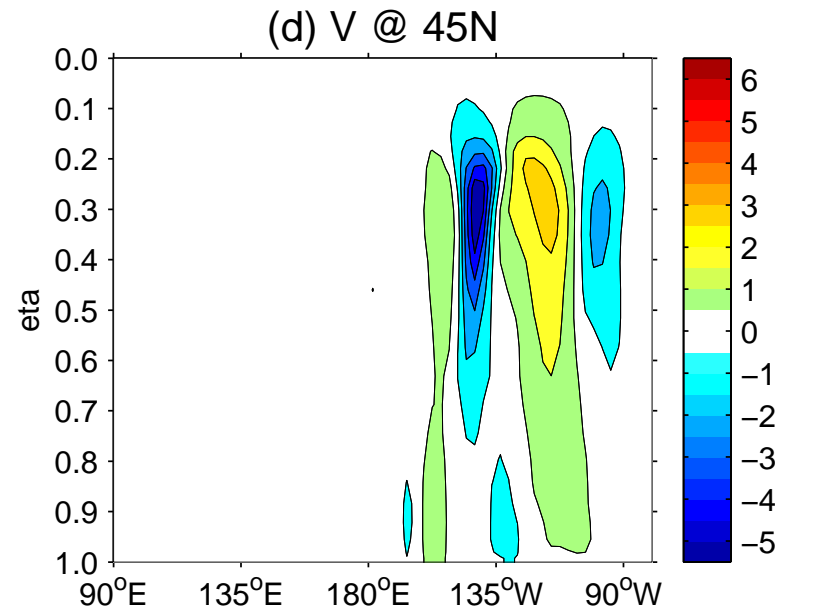
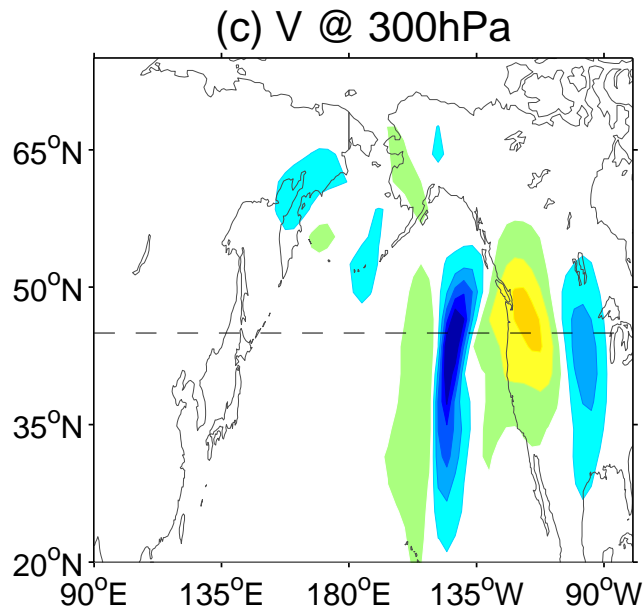
Impact of Initial-time Norm on SV structure (2)

Leading analysis error covariance SV on 2003121600
(perturbed 3d-var, homogeneous/isotropic correlations)

initial
time:



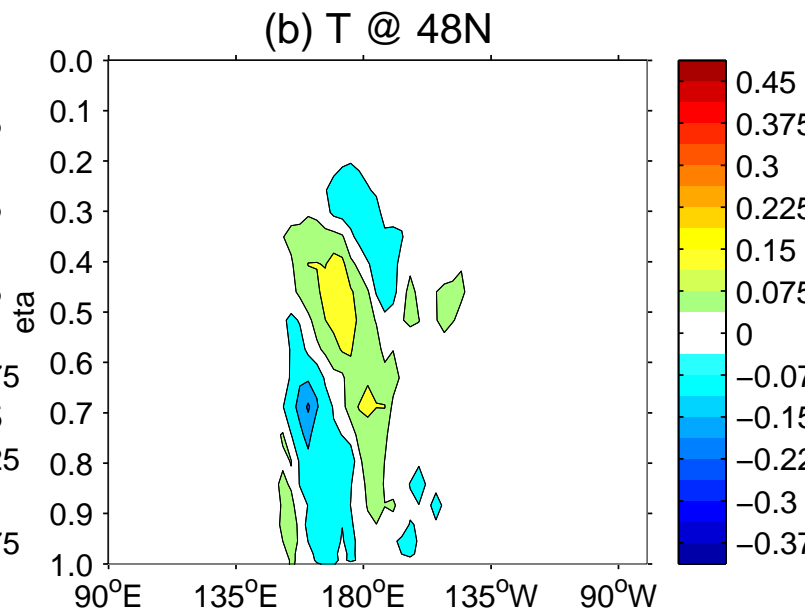
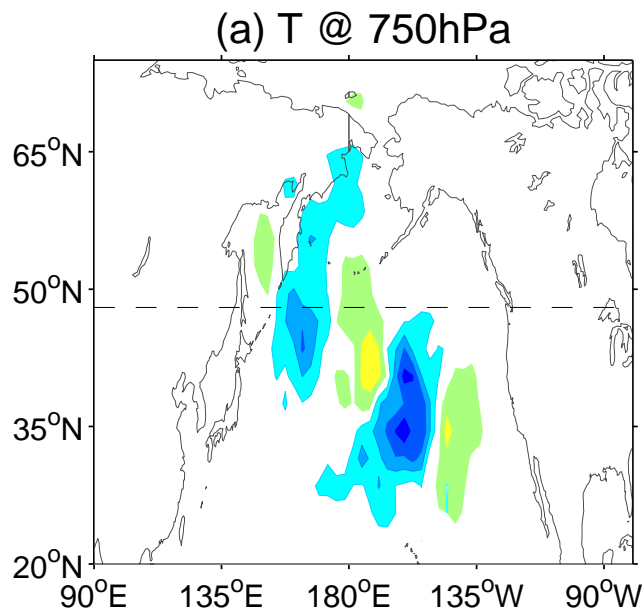
final
time:



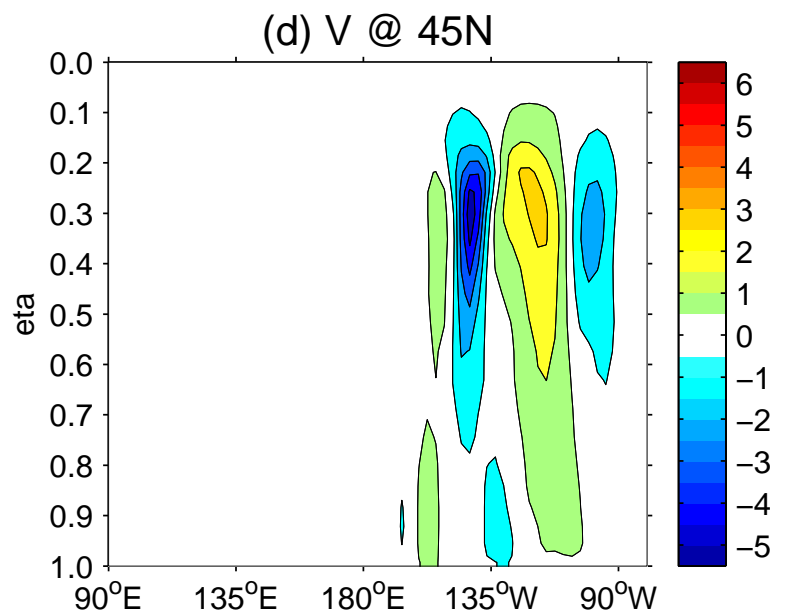
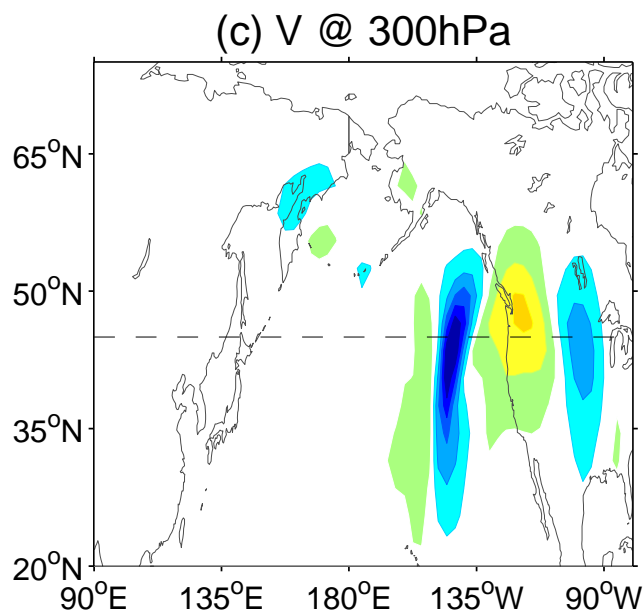
Impact of Initial-time Norm on SV structure (3)

Leading analysis error covariance SV on 2003121600
(perturbed 3d-var, localized correlations)

initial
time:

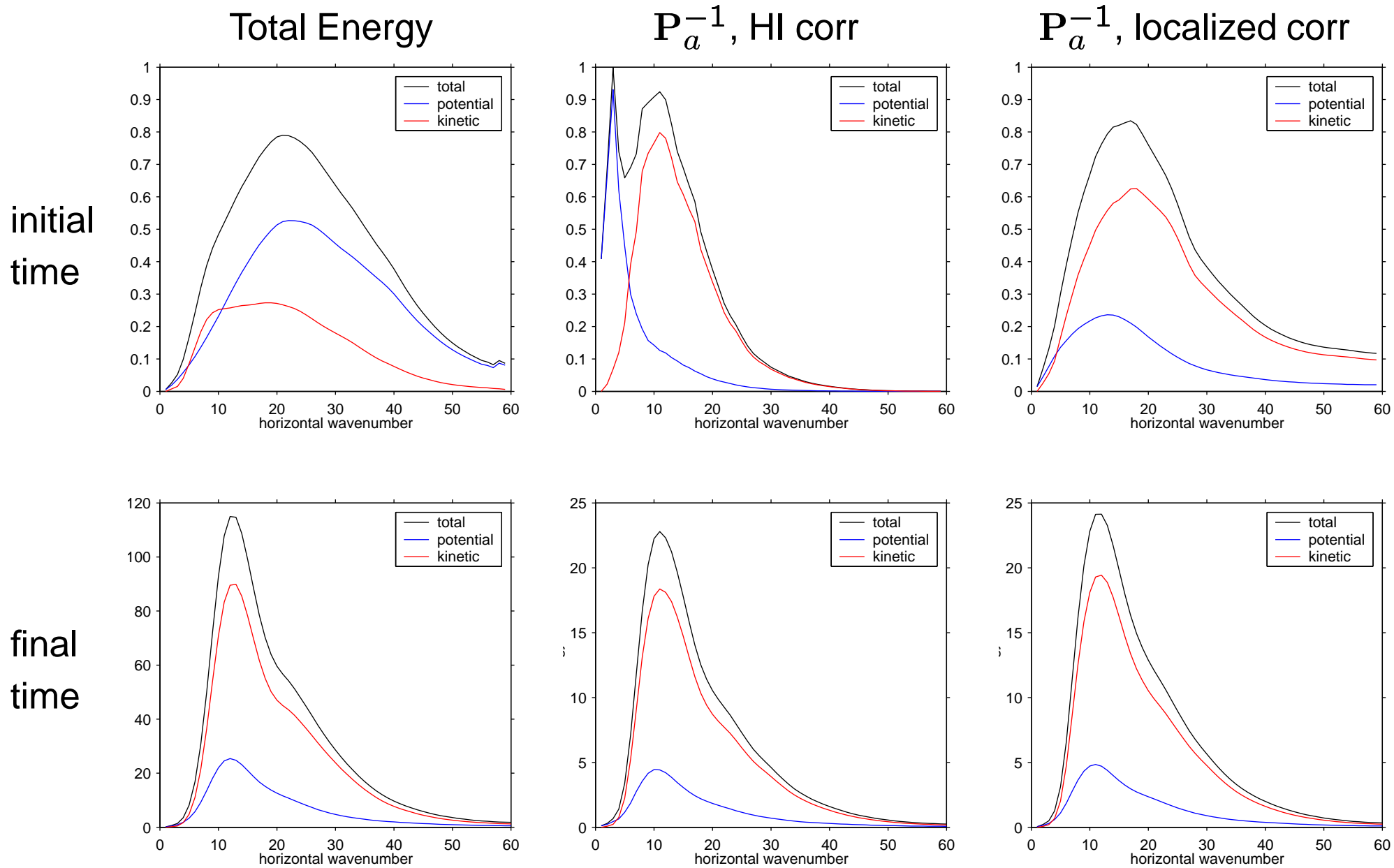


final
time:



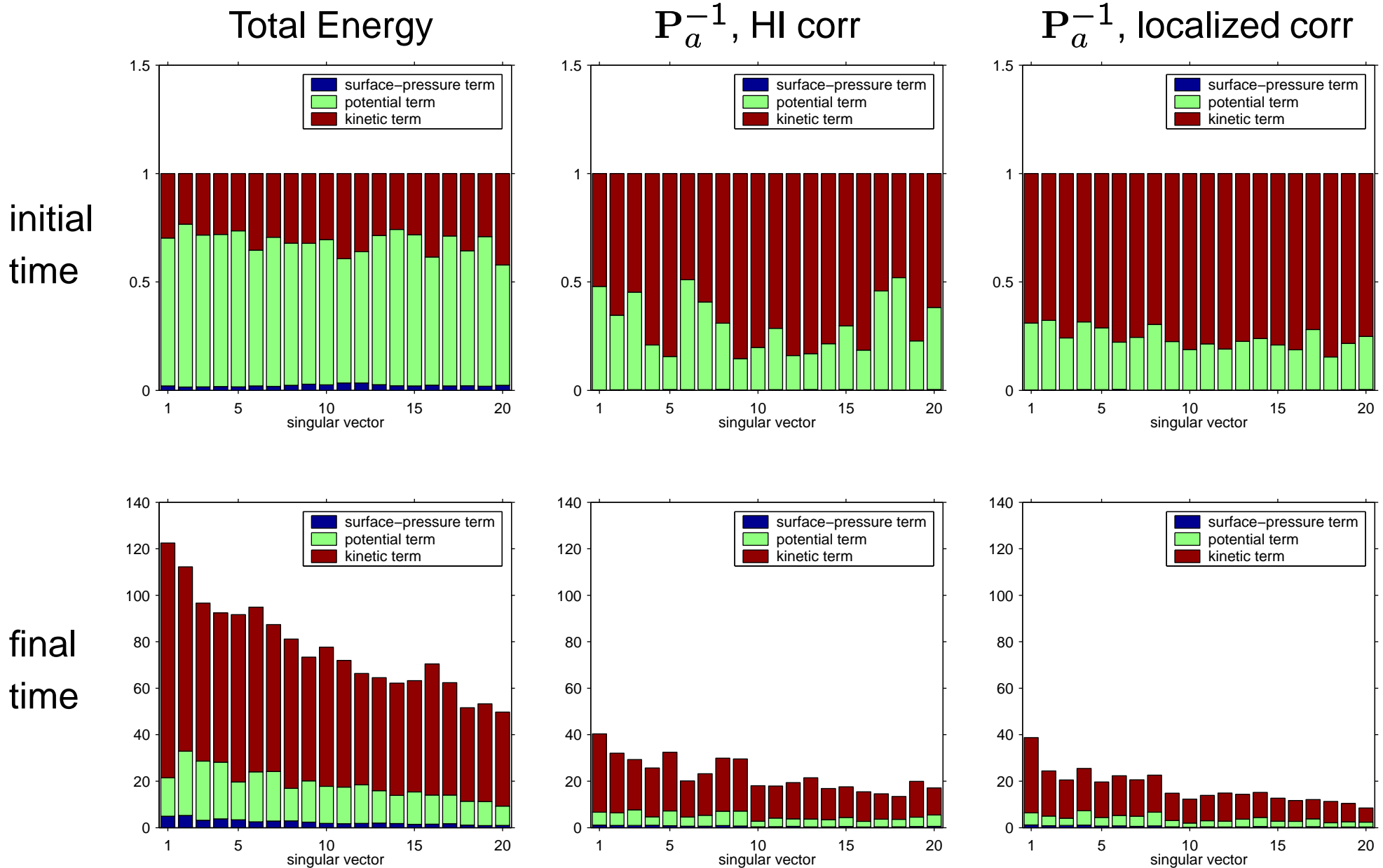
Impact of Initial-time Norm on SV structure (4)

Vertically integrated power spectra for 20 SVs on 2003121600 with initial-time norm:



Impact of Initial-time Norm on SV structure (5)

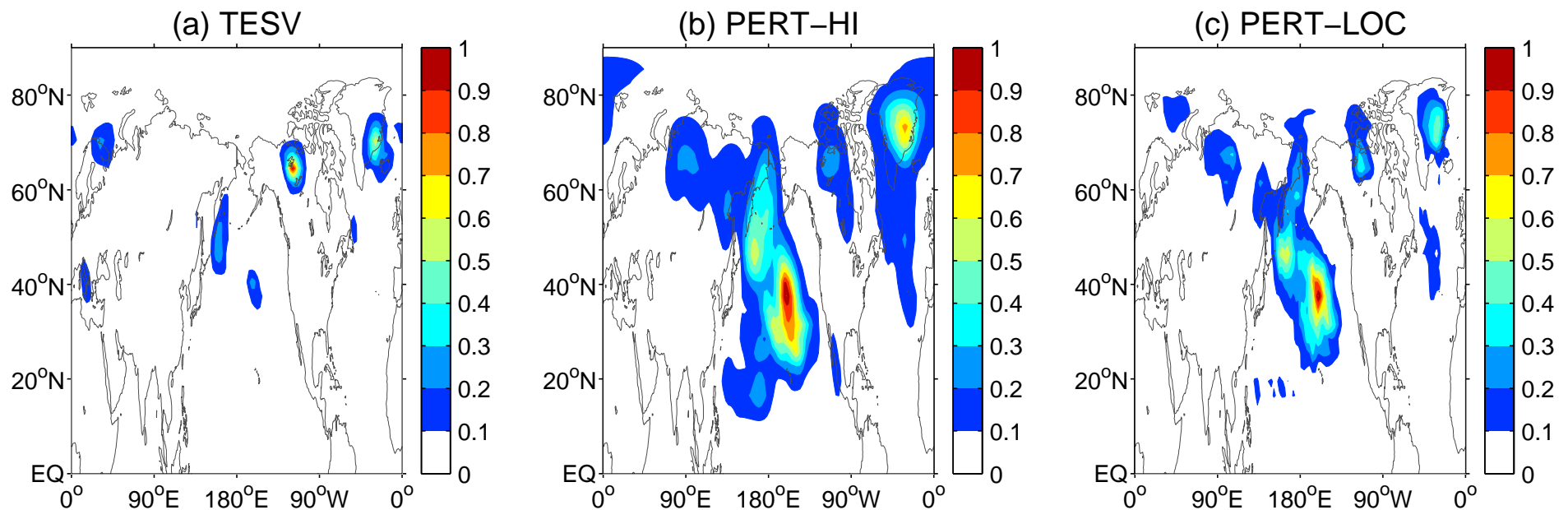
Partition of energy for 20 SVs on 2003121600 with initial-time norm:



Impact of Initial-time Norm on Observation Targeting

Initial-time vertically integrated total energy, growth factor weighted average of 20 SVs with initial-time norms on 2003121600:

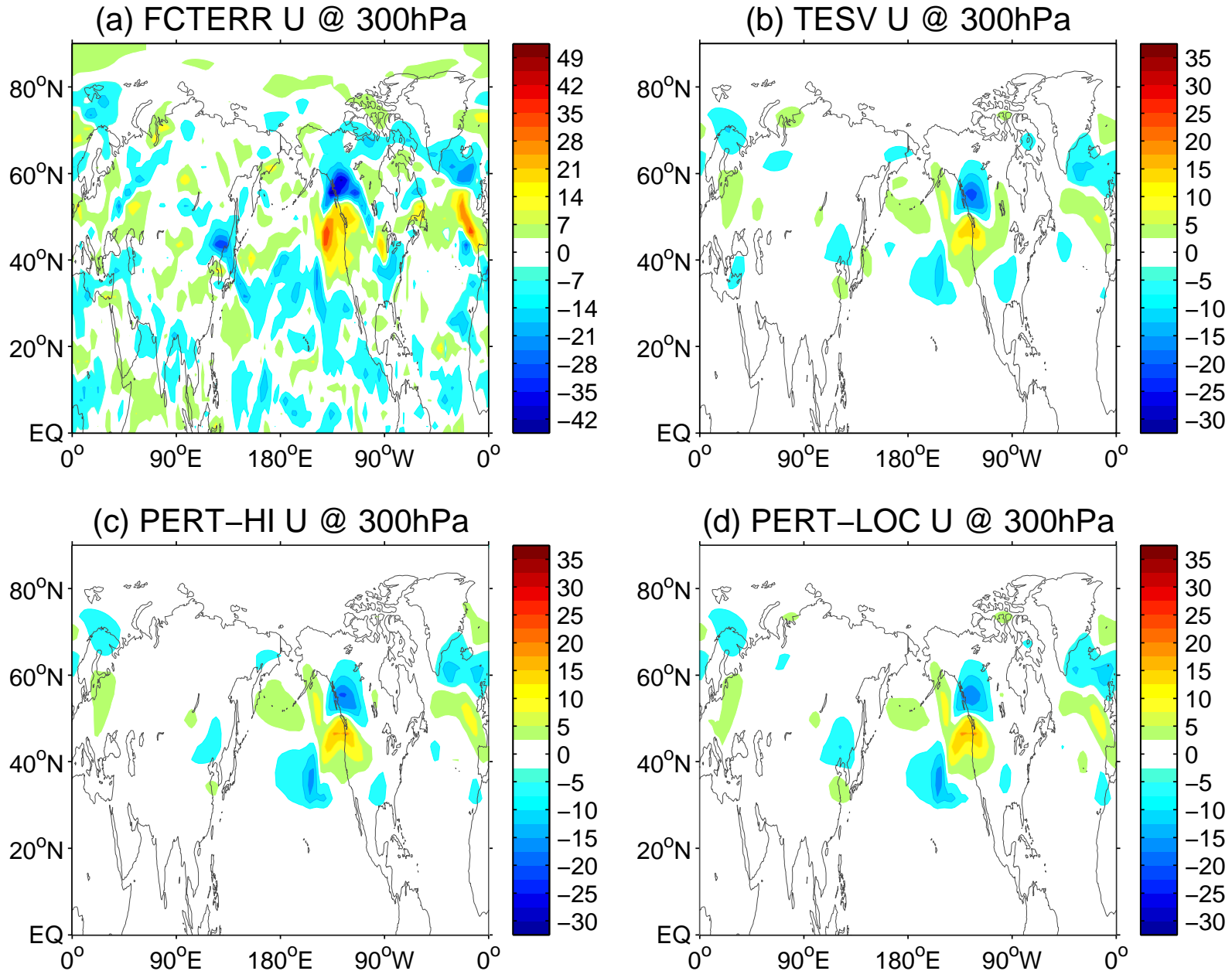
- (a) total energy
- (b) analysis error cov (perturbed 3d-var, homogeneous/isotropic correlations)
- (c) analysis error cov (perturbed 3d-var, localized correlations)



Type of map used to select locations for adaptive observation deployments (e.g. dropsondes)

Impact of Initial-time Norm on Observation Targeting (2)

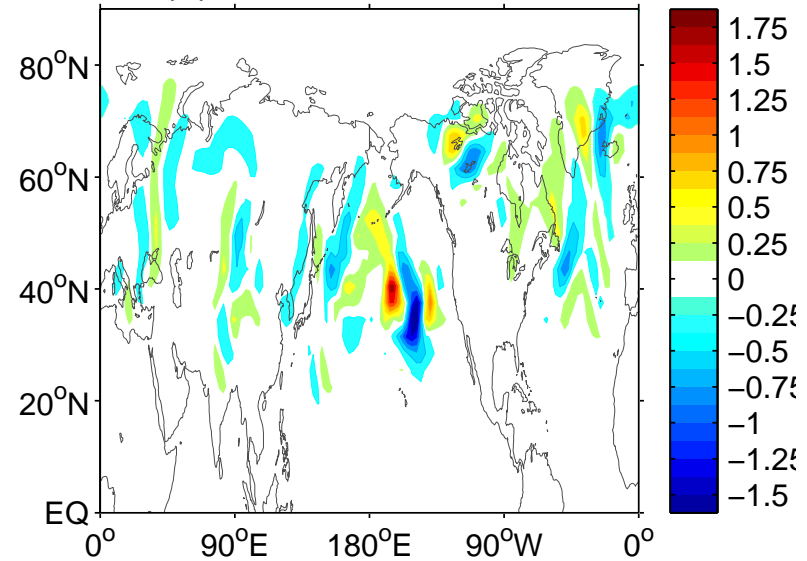
NH FctErr and its projection onto 20 SVs at optimization time (2003121800)



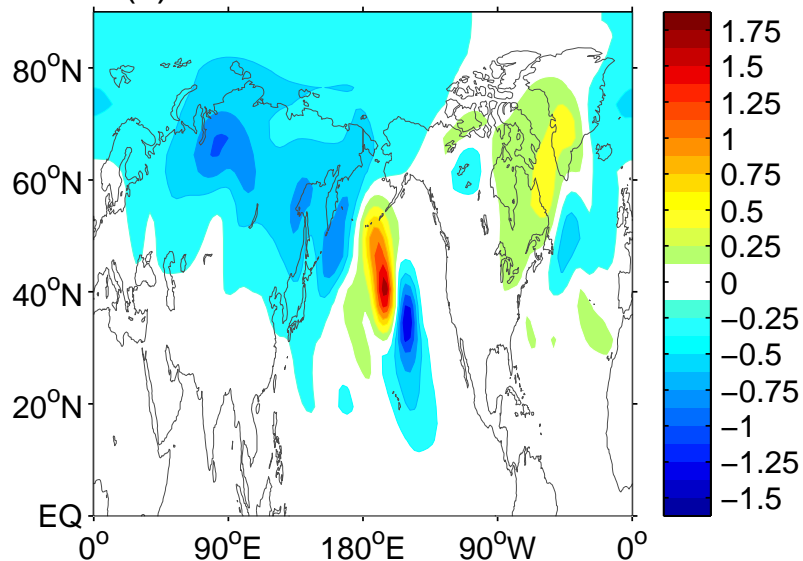
Impact of Initial-time Norm on Observation Targeting (3)

Pseudo-Inverse of FctErr projection onto 20 SVs - like KAE
(linear combination of initial-time SVs on 2003121600)

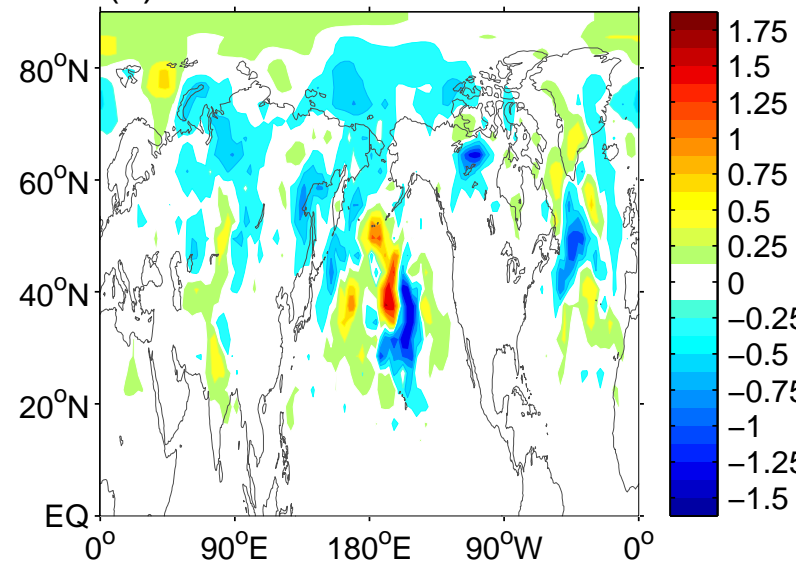
(b) TESV T @ 750hPa



(c) PERT-HI T @ 750hPa



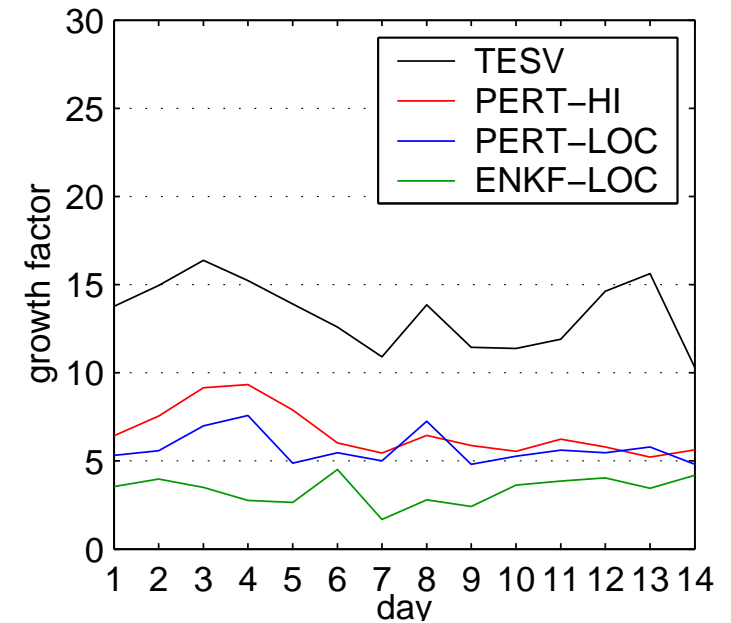
(d) PERT-LOC T @ 750hPa



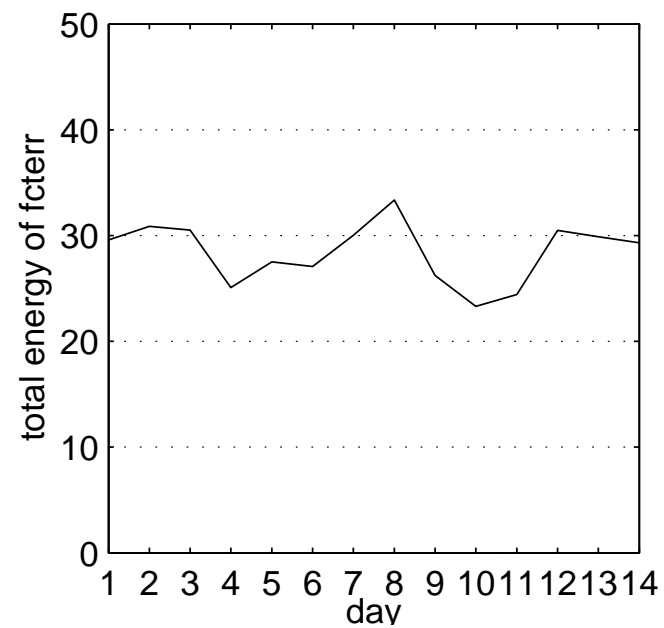
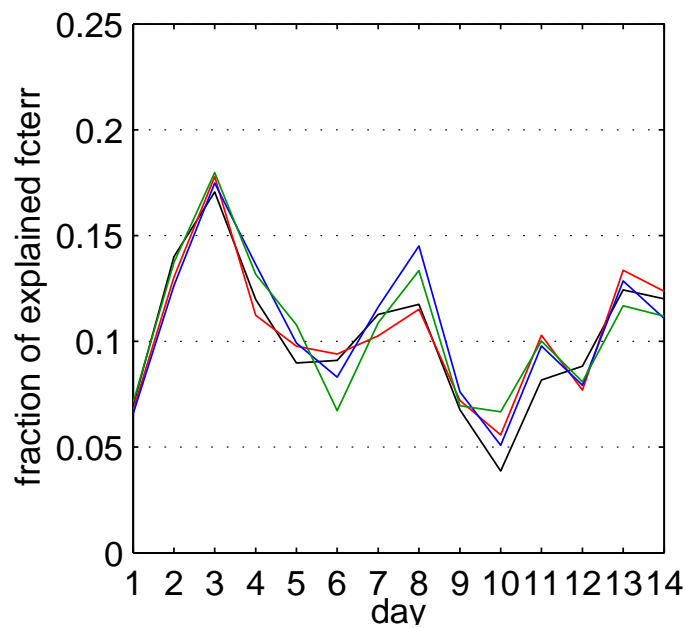
Impact of Initial-time Norm on Predicting FctErr

- total energy growth factor of leading SV ($\sqrt{TE_{48}/TE_0}$)

- TESV: initial SV norm is total energy
- PERT-HI: initial SV norm from perturbed 3d-var analysis error cov with homogeneous/isotropic corr
- PERT-LOC: initial SV norm from perturbed 3d-var analysis error cov with localized corr
- ENKF-LOC: initial SV norm from flow-dependent EnKF analysis error cov with localized cor



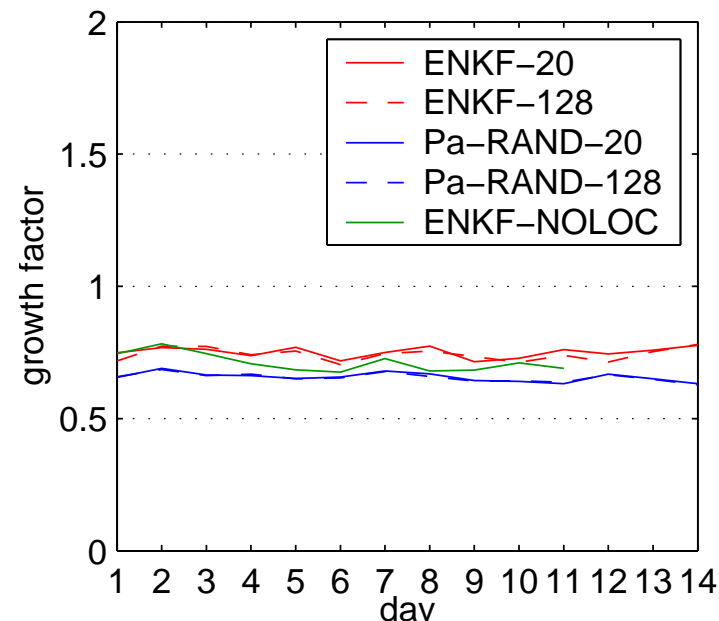
- fraction of 48-hour NH forecast error energy explained by 20 SVs and total fcterr



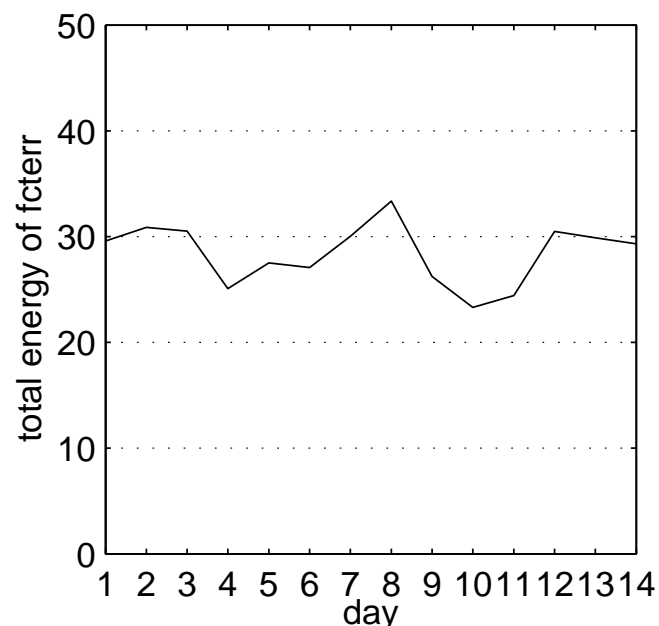
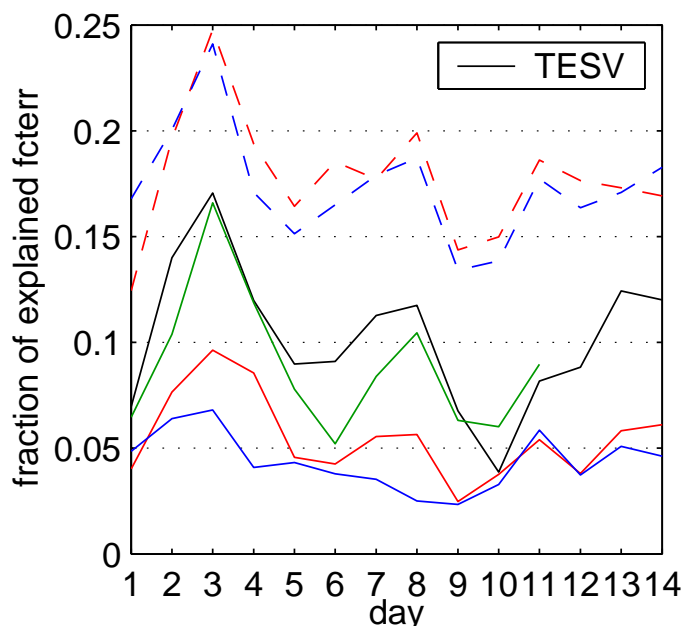
EnKF Members vs Random Pert. for Predicting FctErr

- total energy growth factor of one ensemble member ($\sqrt{TE_{48}/TE_0}$)

- ENKF-20,-128: 48h TLM forecasts initialized with 20 or 128 members from analysis ensembles
- Pa-RAND-20,-128: 48h TLM forecasts initialized with 20 or 128 random perturbations drawn from Gaussian distribution with perturbed 3d-var analysis error cov (homogeneous/isotropic corr)
- ENKF-NOLOC: 20 SVs from non-localized EnKF P_a (orthogonal linear combinations of EnKF members)



- fraction of 48-hour NH forecast error energy explained and total fcterr



Conclusions

1. **large impact** on spatial and multi-variate structure of **initial-time SVs** and location of most sensitive regions from using analysis error covariance norms, and
2. for the case examined, SVs using stationary \mathbf{P}_a^{-1} norm better predict sensitive region than TESV (i.e. more consistent with pseudo-inverse of \mathbf{fcterr})
3. using different SV norms results in **small impact on final-time SVs** and on ability to explain 48-hour forecast error
4. EnKF analysis **ensemble members and random perturbations** from stationary analysis error distribution explain **similar fraction of 48-hour forecast error**
5. 20 ensemble members (or random perturbations) explain about half the amount of forecast error as 20 SVs, but 128 members explain about 50%-100% more

Future Directions

- apply standard ensemble prediction diagnostics to evaluate SVs, EnKF and random perturbations (diagnostics that account for spread)
- perform longer ensemble predictions with nonlinear model
- explore sensitivity to ensemble size for SVs, EnKF and random perturbations (determine relative cost of each approach for equivalent quality)
- explore application of SVs to observation targeting (A. Zadra) and application of analysis error covariance norms to Key Error Analysis (S. Laroche)

The End