1. Preconditioning with Hessian Eigenvectors

2. Singular Vectors: Diagnostic Tool and Covariance Propagation

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Outline of Talk

Background:

- Lanczos-type eigenvalue solvers compute leading eigenvalues/vectors of a large matrix L requiring only: w = Lv, for arbitrary v
- ARPACK is a freely available package of this type

1. Preconditioning with HEVs:

- 3d/4d-var cost function
- estimation of the Hessian
- results from preconditioning 3d/4d-var

2. Singular Vectors:

- formulation
- identifying numerical instabilities
- structure of TESVs, BGSVs, HSVs
- covariance propagation with HSVs
- RRKF and time-averaged SVs

Cost Function for 3d/4d-var

• 3d/4d-var produces analysis increment by minimizing:

$$J = J_{bg} + J_o$$

= $\Delta \mathbf{x}^T \mathbf{B}^{-1} \Delta \mathbf{x} + (\mathbf{H} \Delta \mathbf{x} - \mathbf{y}')^T \mathbf{R}^{-1} (\mathbf{H} \Delta \mathbf{x} - \mathbf{y}')$
where $\mathbf{y}' = \mathbf{y} - H(\mathbf{x}^{bg})$, and
 $\Delta \mathbf{x} = \mathbf{x} - \mathbf{x}^{bg} = [\Delta \psi, \Delta \chi, \Delta T, \Delta P_s, \Delta \ln(q)]$

- Note: ${\bf H}$ also involves integration of the TLM of GEM for 4d-var
- B spreads information from observations spatially according to covariance structure
- current B: stationary, correlations horizontally homog./isotropic, variances zonally invariant
- between-variable covariances governed by balance operators (geostrophic and Ekman balances)
- variances, univariate correlations, and balance operators estimated from ensemble of lagged forecast differences (NMC method)

Preconditioning with HEVs

• to improve efficiency J is rewritten in terms of the control vector ξ , where $\Delta \mathbf{x} = \mathbf{B}^{1/2} \xi$:



- J possibly still poorly conditioned due to J_o
- solution: use (approximate) Hessian to define norm for minimizer
- Hessian (and inverse analysis error cov) given by:

$$\mathbf{J}_{\xi\xi} = \mathbf{A}_{\xi}^{-1} = \mathbf{I} + \mathbf{B}^{T/2} \mathbf{H}^T \mathbf{R}^{-1} \mathbf{H} \mathbf{B}^{1/2}$$

- obtain truncated eigen-decomp of second term (Lanczos algorithm): extracts dominant effect of observations on Hessian (and A)
- for 3d-var: **B**, **H** and **R** nearly constant, therefore eigenvectors can be re-used (differences in **H** between 0/12UTC and 6/18UTC)
- for 4d-var: dependence of TLM/ADJ on background trajectory imposes more time-dependence

Preconditioning 3d/4d-var

 3d-var: eigenvectors computed from outside period at 6:00UTC and held fixed Impact of Preconditioning with Hessian Eigenmodes and 1e–2 Precision



 preliminary result with 4d-var, eigenvectors calculated from first time only:

date	NHEV	#iterations	User Time
2001020712	0	131	5943
2001020800	0	139	6565
2001020712	10	92	4096
2001020800	10	94	4400

Estimation of A: ψ , 250 hPa

SQRT of Correction to BG Variances from Radiosonde data only:



Approximate Analysis Error Std.Dev. (from 300 eigenvectors):



Singular Vectors

Background:

- compute perturbations that will optimally grow (linear dynamics) over specified period (t) w.r.t. specified norms
- currently used at ECMWF to define subspace for RRKF and to generate initial perturbations for EPS
- useful for identifying numerical instabilities in the model

Formulation:

• maximize 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 eigenvector problem: $\mathbf{W}_0^{-T/2} \mathbf{M}^T \mathbf{W}_t \mathbf{M} \mathbf{W}_0^{-1/2} \gamma_k = \lambda_k^2 \gamma_k$
- final time norm typically TE, at initial time: TE (TESV), A^{-1} (HSV), or B^{-1} (BGSV)
- HSV: only by using approximation to A^{-1} can $W_0^{-1/2}$ be calculated (at ECMWF full Hessian used, resulting in a more expensive general-ized eigenvalue problem)

Experimental Setup

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



- experiments at low resolution (64x32x28L) with simplified vert diffusion (S. Laroche) and no humidity
- final time norm is TE north of 30°N
- different initial time (global) norms used: TESV, HSV, BGSV
- A⁻¹ estimated using 300 HEVs calculated from 3d-var cost function and full set of obs
- computational cost per SV: approximately 3 times cost of TLM+ADJ integration

Results: Computational Modes in TESVs

- unlike HSV and BGSV, no balance or smoothness constraints imposed by \mathbf{W}_{0}
- quickly highlights numerical instabilities related to model numerics:
 - with no vertical diffusion: spurious growth at sfc
 - then, modes at pole dominate
 - after applying smoothness constraint or sponge at pole, modes at the top appear
 - after applying sponge at the top or zeroing a few levels at t=0h, localized modes near the jet appear (similar to noise seen by M. Roch?)
- also grow in nonlinear model (lo-res with simplified physics)
- more realistic modes with sponge near pole and zeroing U,V,T down to 200 hPa at the initial time
- further experiments required to determine:
 - if modes grow in model at full-res with full physics and if so, what is their importance generally and for 4d-var
 - best way to eliminate computational modes (at ECMWF also zero values above level 6 at initial time and use enhanced horizontal diffusion in TL/AD and possibly smaller timestep)

No Vertical Diffusion: TESV structure (T)



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Pole Problem: TESV structure (T)



Top Problem (mode 23): TESV structure (T)



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Near the jet (mode 21): TESV structure (T)



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Meteorological Modes Using 3 Norms

At initial time:

- structures tilted vertically against the shear
- TESV generally sharper hor and vert structure and higher growth rates than BGSV or HSV
- TESV: KE concentrated around 800 hPa, BGSV and HSV: around 650 hPa
- TESV: temperature relatively more dominant at initial time than for BGSV and HSV
- leading TESV and BGSV in similar location, HSV over oceans (high anl error)

At optimization time (48h):

- tilt removed (especially for T)
- leading TESV and BGSV very similar
- all norms give similar structure, KE concentrated around 300 hPa
- nonlinearly evolved SVs very similar structure (no scaling)
- amplification factors lower than other studies (case dependent)

Results: TESV structure (T)



15

7

6

5

4

3

2

1

0

-1 -2

-3

-4 -5

-6

Results: TESV structure (V)



Results: BGSV structure (T)



Results: BGSV structure (V)







Results: HSV structure (T)



Results: HSV structure (V)



7

6

5

4

3

2

1

0

-1

-2

-3

-4 -5

-6

35

30

25

20

15 10

5

0

-5

-10

-15

-20 -25

-30

Results: Amplification Factors for Different

Norms/Optimization Times



Time-Averaged SVs

- limitation of SVs: dependent on chosen optimization time, may be significantly sub-optimal at other times (nonmodal growth)
- instead, maximize the average growth over multiple optimization times, e.g.: $\frac{\mathbf{x}_1^T \mathbf{W}_1 \mathbf{x}_1 + \mathbf{x}_2^T \mathbf{W}_2 \mathbf{x}_2}{\mathbf{x}_2^T \mathbf{W}_0 \mathbf{x}_0}$, where $\mathbf{x}_1 = \mathbf{M}_1 \mathbf{x}_0$, and $\mathbf{x}_2 = \mathbf{M}_2 \mathbf{x}_1$
- leads to eigenvalue problem:

$$\mathbf{W}_{0}^{-T/2}\mathbf{M}_{1}^{T}\left(\mathbf{W}_{1}+\mathbf{M}_{2}^{T}\mathbf{W}_{2}\mathbf{M}_{2}\right)\mathbf{M}_{1}\mathbf{W}_{0}^{-1/2}\gamma_{k}=\lambda_{k}^{2}\gamma_{k}$$

- almost same computational cost
- different weighting of term at each time may be desirable

Growth of TESVs: 48h vs. 24h+48h



Covariance Propagation with HSVs

- initial time HSVs evolve into (scaled) leading eigenvectors of propagated covariances: $\mathbf{MAM}^T \approx \sum \mathbf{x}_t \mathbf{x}_t^T$ (KF: $\mathbf{P}^f = \mathbf{MAM}^T + \mathbf{Q}$)
- suggests use as basis functions in RRKF
- compare with covariances propagated with Monte-Carlo approach:

$$\mathbf{MAM}^T \approx \overline{\epsilon_a \epsilon_a^T}$$
, where $\epsilon_a = \mathbf{MA}^{1/2} \epsilon$, and $\epsilon \sim N(0, \mathbf{I})$

• with 25 HSVs, variance appears to be underestimated (especially eastern NA, EUR) vs. 150 random perturbations (similar cost)



Standard Deviation of Propagated A (V)

From 25 leading HSVs evolved 24 hours:





From 150 random perturbations evolved 24 hours:





Flow-Dependent B in 3d-var

- use 25 leading HSVs partially evolved by 6 hours
- covariance matrix composed of two components:

$$\mathbf{B}_{sv} = \alpha_1 \mathbf{B}_{hi} + \alpha_2 \sum \mathbf{x}_{6h} \mathbf{x}_{6h}^T$$

- similar to existing approaches (ENKF: Hamill and Snyder, HSV: Fisher, Bred mode: UKMet)
- 1-obs exp't in 3d-var shows spatial covariance structure of B_{sv} (V at 250 hPa, 40°N, 140°W):



Only SVs ($\alpha_1 = 0, \alpha_2 = 1$)

Flow-Dependent B in 3d-var

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RRKF using HSVs

- lack of projection of evolved SVs with coincident initial time SVs (Gelaro et. al. 1998) prevents cycling of covariances with HSVs
- possible solution: use "balanced" combination of finite-time (or timeaveraged) SVs and evolved SVs (Farrell and Ioannou 2001)
- "balancing" defines a single subspace that contains both the subspace in which the model is most sensitive and the optimal response subspace \rightarrow should improve overlap
- time-averaged SVs similar to Stochastic Optimals (B. Farrell) defined for <u>autonomous</u> systems as leading eigenvectors of:

$$\tilde{\mathbf{Q}} = \sum_{n=0}^{\infty} \left(\mathbf{M}^n \right)^T \left(\mathbf{M}^n \right) \Delta t$$

• FI (2001) showed effectiveness of using balanced truncation of ${\bf Q}$ and:

$$\tilde{\mathbf{P}} = \sum_{n=0}^{\infty} \left(\mathbf{M}^n \right) \left(\mathbf{M}^n \right)^T \Delta t$$

to define the subspace for propagating covariances in a RRKF

- FI (1999) showed good results when approximating $\mathbf{\tilde{Q}}$ and $\mathbf{\tilde{P}}$ with a single well-chosen term in sum

Conclusions

Results:

- preconditioning with HEVs reduces required number of iterations in 3d/4d-var by 35%-50% (reuse eigenvectors from 6 or 18UTC)
- SVs useful for identifying unwanted numerical instabilities in GEM
- new type of SV (time-averaged) allows optimizing growth over a range of lead times
- SVs may provide effective way for cycling covariances in 3d/4d-var

Future work:

- more testing of preconditioning in 3d/4d-var (determine effectiveness at earlier cut-off times)
- investigate approaches to eliminate numerical instabilities
- evaluate impact of increasing resolution and including more physics on SVs (moist processes, subgrid scale orography)
- compare SVs (growth rates) from specific case with ECMWF
- more fully compare covariance propagation using HSVs and Monte Carlo approach
- evaluate impact of balancing SVs and evolved SVs for covariance propagation in a cycling context (compare with ENKF)