

First steps towards the application of the Ensemble Transform Kalman Filter technique at the Hungarian Meteorological Service

Gergely Bölöni, Petra Csomós

Hungarian Meteorological Service

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First steps towards the application of the Ensemble Transform Kalman Filter technique at the Hungarian Meteorological Service

- Data assimilation
- Theoretical aspects of ETKF
- Realization of ETKF at HMS



Data assimilation

Aim: to obtain the “best” estimate of the present state of the atmosphere



Data assimilation

- Aim:** to obtain the “best” estimate of the present state of the atmosphere
- From:** observations and previous forecast



Data assimilation

Aim: to obtain the “best” estimate of the present state of the atmosphere

From: observations and previous forecast

How?

- ① Optimal interpolation
- ② Variational methods
- ③ Kalman filter



Kalman filter

Analysis:

$$x_a = x_f + P_f H^\top (H P_f H^\top + P_o)^{-1} (y - \mathcal{H}(x_f))$$

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Denotation

- P_f error covariance matrix of forecast
- P_a error covariance matrix of analysis
- P_o error covariance matrix of observations
- P_M error covariance matrix of model error
and linearization



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P_f	error covariance matrix of forecast
P_a	error covariance matrix of analysis
P_o	error covariance matrix of observations
P_M	error covariance matrix of model error and linearization
M	linearized model ($i \rightarrow i+1$)
\mathcal{H}	observation operator
H	linearized of \mathcal{H}
$(\cdot)^\top$	transpose
i	time level



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Idea: *Kalman (1960):*

$$P_f^{i+1} = M P_a^i M^\top + P_M^i$$



Kalman filter

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However: not applicable in weather forecast

- $P_f \in \mathbb{R}^{n \times n}$, $n \approx 10^7$
- $M \in \mathbb{R}^{n \times n} \Rightarrow n$ integrations
- need for tangent linear and adjoint models



Kalman filter

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⇒ Extended KF

⇒ Ensemble KF

⇒ Reduced Rank KF

⇒ Ensemble Transform KF



Ensemble Transform Kalman Filter

Idea: error statistics from an ensemble: $x_{f,j}$ ($j = 1, \dots, k$)



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Definition

Dispersions: $z_j := x_j - \bar{x}$ ($j = 1, \dots, k$)

Their “vector”: $Z := \frac{1}{\sqrt{k-1}}(z_1, z_2, \dots, z_k)$

Error cov. matrix: $P \approx \frac{1}{k-1} \sum_{j=1}^k (x_j - \bar{x})(x_j - \bar{x})^\top = ZZ^\top$

(Each for f and a as well.)



Ensemble Transform Kalman Filter

The main point: connection between Z_a and Z_f (described by T):

$$Z_a = Z_f T$$



Ensemble Transform Kalman Filter

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Question: transformation matrix $T = ?$



Assumptions

- transformation: $Z_a = Z_f T$
 - square-root: $P_a = Z_a Z_a^\top$
 - from BLUE: $P_a = (I - KH)P_f$
- $$K = P_f H^\top (H P_f H^\top + P_o)^{-1} \text{ (Kalman gain)}$$



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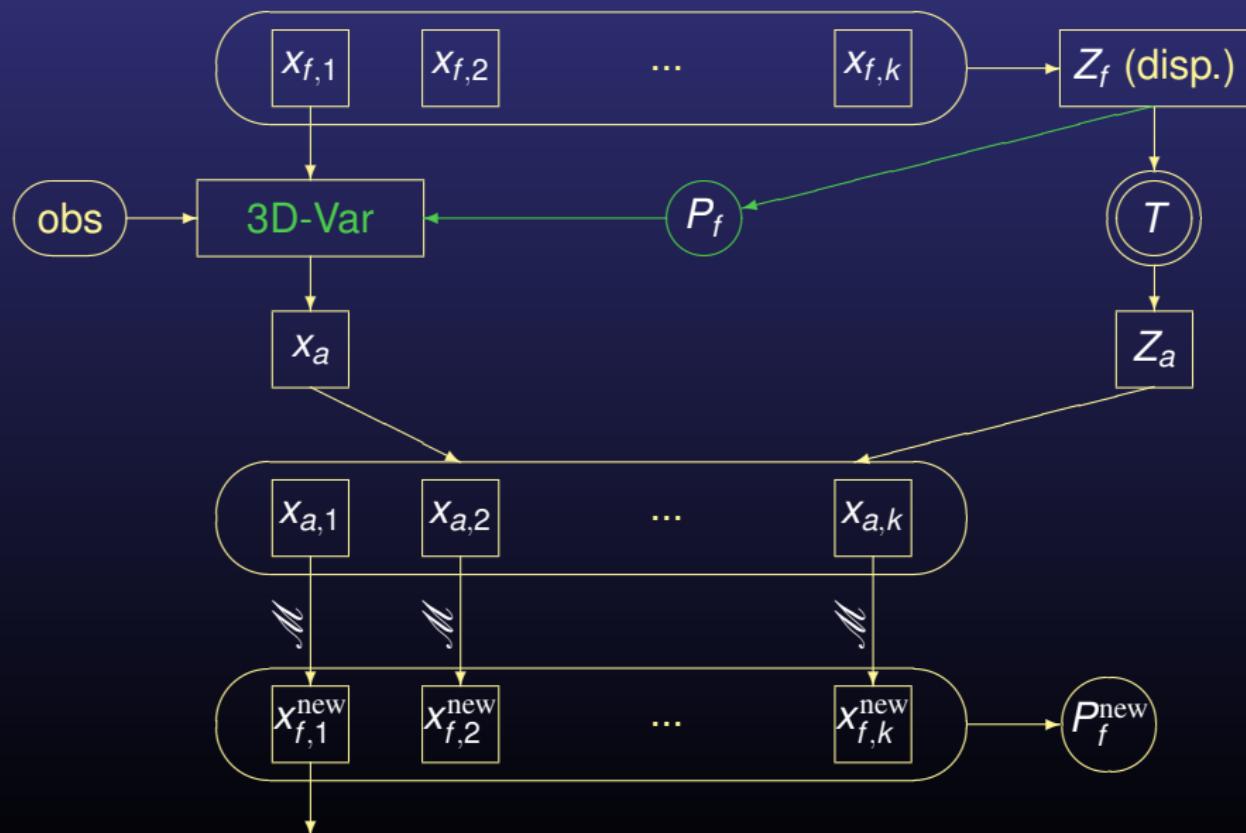
Bishop et al. (2001) \Rightarrow $T = C(\Gamma + I)^{-1/2}$ with

$$Z_f^\top H^\top P_o^{-1} H Z_f = C \Gamma C^\top \in \mathbb{R}^{k \times k}$$

(eigenvectors, eigenvalues)



Ensemble Transform Kalman Filter



Realization of ETKF

- ① Build matrix $Z_f^\top H^\top P_o^{-1} H Z_f$
- ② Solve the eigenvalue problem $Z_f^\top H^\top P_o^{-1} H Z_f = C \Gamma C^\top$
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 $(j = 1, \dots, k)$
- ⑥ Generate new ensemble members with $x_{f,j}^{\text{new}} = \mathcal{M}(x_{a,j})$
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Matrix $Z_f^\top H^\top P_o^{-1} H Z_f = V^\top V$ **with** $V = \underbrace{P_0^{-1/2}}_{\sigma_{o,j}} \underbrace{H Z_f}_{?}$



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$$Hz_{f,j} = H(x_{f,j} - \bar{x})$$



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$$\Rightarrow HZ_{f,j} = \text{fg_depar}_{\bar{x}_f} - \text{fg_depar}_{x_{f,j}} \quad j = 1, \dots, k$$



Realization of ETKF

Hence:

$$V(i,j) = (P_0^{-1/2} H Z_f)(i,j) =$$

$$= \frac{1}{\sqrt{k-1}} \frac{1}{\text{obs_error}(i)} \left[\text{fg_depar}_{\bar{x}_f}(i) - \text{fg_depar}_{x_{f,j}}(i) \right]$$

$$\Rightarrow Z_f^\top H^\top P_o^{-1} H Z_f = V^\top V = \dots$$



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Discussion

- ETKF: time-dependent P_f , ensemble system
- Compare with EnKF: less MIN but computation of T
- HMS: development is ready but no cycling
- Next step: cycling, diagnostics
- Open problems: sampling noise, coupling



