

Basic Equations for the Ensemble Kalman and Square-root Filters

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List of symbols

ψ_i^f	Forecast state (ensemble member i)
ψ^t	Truth state
ψ_i^a	Analysis state (ensemble member i)
y_i	Observation set (associated with ensemble member i)
$y_i^{(m)}$	Model observation set (associated with ensemble member i)
\mathbf{P}^f	Forecast error covariance matrix
\mathbf{P}_e^f	Ensemble estimate of forecast error covariance matrix
\mathbf{P}^a	Analysis error covariance matrix
\mathbf{P}_e^a	Ensemble estimate of analysis error covariance matrix
\mathbf{R}_e	Observation error covariance matrix
h	Non-linear observation operator
\mathbf{H}	Tangent linear model of observation operator
\mathbf{K}	Kalman gain matrix
\mathbf{A}^f	Forecast state vector matrix (forecasts are columns)
\mathbf{A}'^f	Forecast state vector perturbation matrix
\mathbf{A}^a	Analysis state vector matrix (analyses are columns)
\mathbf{A}'^a	Analysis state vector perturbation matrix
\mathbf{D}	Observation vector matrix (observation sets are columns)
\mathbf{Y}	Observation vector perturbation matrix
$\mathbf{I}_{N \times N}$	$N \times N$ identity matrix
$\mathbf{1}_{N \times N}$	$N \times N$ matrix of $1 / N$ values
\mathbf{S}	Model observation perturbations calculated from forecast perturbations
\mathbf{C}	$(N - 1) \times$ model ob + ob covariance matrices

Let there be n elements in a state vector, N ensemble members and p observations.

Definition of error covariance matrices

Forecast error covariance matrices

$$\mathbf{P}^f = \overline{(\psi_i^f - \psi^f)(\psi_i^f - \psi^f)^T}$$

$$\mathbf{P}_e^f = \overline{(\psi_i^f - \overline{\psi^f})(\psi_i^f - \overline{\psi^f})^T}$$

Analysis error covariance matrices

$$\mathbf{P}^a = \overline{(\psi_i^a - \psi^f)(\psi_i^a - \psi^f)^T}$$

$$\mathbf{P}_e^a = \overline{(\psi_i^a - \overline{\psi^a})(\psi_i^a - \overline{\psi^a})^T}$$

Observation error covariance matrix

$$\mathbf{R}_e = \overline{(\mathbf{y}_i - \overline{\mathbf{y}_i})(\mathbf{y}_i - \overline{\mathbf{y}_i})^T}$$

The observation operator

$$\mathbf{y}_i^{(m)} = \mathbf{h}(\overline{\psi_i^f}) + \mathbf{H}(\psi_i^f - \overline{\psi_i^f})$$

The KF analysis update step for each member and for the mean

Analysis step for ensemble member i

$$\psi_i^a = \psi_i^f + \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1} (\mathbf{y}_i - \mathbf{h}(\overline{\psi_i^f}) - \mathbf{H}(\psi_i^f - \overline{\psi_i^f}))$$

$$\overline{\psi_i^a} = \overline{\psi_i^f} + \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1} (\overline{\mathbf{y}_i} - \mathbf{h}(\overline{\psi_i^f}))$$

Deviation of each member from the mean

$$\psi_i^a - \overline{\psi_i^a} = \psi_i^f - \overline{\psi_i^f} + \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1} (\mathbf{y}_i - \overline{\mathbf{y}_i} - \mathbf{H}(\psi_i^f - \overline{\psi_i^f}))$$

$$= (\mathbf{I} - \mathbf{K} \mathbf{H})(\psi_i^f - \overline{\psi_i^f}) + \mathbf{K}(\mathbf{y}_i - \overline{\mathbf{y}_i})$$

$$\mathbf{K} = \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1}$$

The ensemble analysis error covariance

Assume no correlation between forecast and observation errors.

$$\begin{aligned} \mathbf{P}_e^a &= \langle [(\mathbf{I} - \mathbf{K} \mathbf{H})(\psi_i^f - \overline{\psi_i^f}) + \mathbf{K}(\mathbf{y}_i - \overline{\mathbf{y}_i})][(\mathbf{I} - \mathbf{K} \mathbf{H})(\psi_i^f - \overline{\psi_i^f}) + \mathbf{K}(\mathbf{y}_i - \overline{\mathbf{y}_i})]^T \rangle \\ &= \langle [(\mathbf{I} - \mathbf{K} \mathbf{H})(\psi_i^f - \overline{\psi_i^f})][(\mathbf{I} - \mathbf{K} \mathbf{H})(\psi_i^f - \overline{\psi_i^f})]^T \rangle + \langle [\mathbf{K}(\mathbf{y}_i - \overline{\mathbf{y}_i})][\mathbf{K}(\mathbf{y}_i - \overline{\mathbf{y}_i})]^T \rangle \\ &= (\mathbf{I} - \mathbf{K} \mathbf{H}) \mathbf{P}_e^f (\mathbf{I} - \mathbf{K} \mathbf{H})^T + \mathbf{K} \mathbf{R}_e \mathbf{K}^T \\ &= \mathbf{P}_e^f - \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T - \mathbf{K} \mathbf{H} \mathbf{P}_e^f + \mathbf{K} \mathbf{H} \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T + \mathbf{K} \mathbf{R}_e \mathbf{K}^T \\ &= \mathbf{P}_e^f + \mathbf{K} (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e) \mathbf{K}^T - \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T - \mathbf{K} \mathbf{H} \mathbf{P}_e^f \\ &= \mathbf{P}_e^f + \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1} (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e) \mathbf{K}^T - \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T - \mathbf{K} \mathbf{H} \mathbf{P}_e^f \end{aligned}$$

$$\begin{aligned}
&= \mathbf{P}_e^f - \mathbf{KHP}_e^f \\
&= (\mathbf{I} - \mathbf{KH})\mathbf{P}_e^f
\end{aligned}$$

Matrix representation of the ensemble

The forecast states ψ_i^f may be assembled into columns of the $n \times N$ matrix \mathbf{A}^f .

$$\mathbf{A}^f = (\psi_1^f \psi_2^f \dots \psi_N^f)$$

The analysis states ψ_i^a may be assembled into columns of the $n \times N$ matrix \mathbf{A}^a .

$$\mathbf{A}^a = (\psi_1^a \psi_2^a \dots \psi_N^a)$$

The perturbed observations y_i may be assembled into the columns of the $p \times N$ matrix \mathbf{D} .

$$\mathbf{D} = (y_1 y_2 \dots y_N)$$

The ensemble means, $\bar{\mathbf{A}}^f$, $\bar{\mathbf{A}}^a$ and $\bar{\mathbf{D}}$ are matrices that comprise the respective ensemble mean state repeated in each column.

$$\begin{aligned}
\bar{\mathbf{A}}^f &= \mathbf{A}^f \mathbf{1}_{N \times N} & \bar{\mathbf{A}}^a &= \mathbf{A}^a \mathbf{1}_{N \times N} & \bar{\mathbf{D}} &= \mathbf{D} \mathbf{1}_{N \times N} \\
\mathbf{1}_{N \times N} &= \begin{pmatrix} 1/N & 1/N & \dots & 1/N \\ 1/N & 1/N & \dots & 1/N \\ \dots & \dots & \dots & \dots \\ 1/N & 1/N & \dots & 1/N \end{pmatrix}
\end{aligned}$$

Matrix representation of the ensemble perturbations from the mean

$$\mathbf{A}'^f = \mathbf{A}^f - \bar{\mathbf{A}}^f = \mathbf{A}^f (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N})$$

$$\mathbf{A}'^a = \mathbf{A}^a - \bar{\mathbf{A}}^a = \mathbf{A}^a (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N})$$

$$\Upsilon = \mathbf{D} - \bar{\mathbf{D}} = \mathbf{D} (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N})$$

$$\text{Note: } (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N})^2 = (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N})$$

The error covariance matrices from the matrix representation of ensemble perturbations

$$\mathbf{P}_e^f = \frac{1}{N-1} \mathbf{A}'^f \mathbf{A}'^{fT} \quad \mathbf{P}_e^a = \frac{1}{N-1} \mathbf{A}'^a \mathbf{A}'^{aT} \quad \mathbf{R}_e = \frac{1}{N-1} \Upsilon \Upsilon^T$$

The analysis equation in terms of the matrix representation of the ensemble

$$\mathbf{A}^a = \mathbf{A}^f + \mathbf{K} (\mathbf{D} - \mathbf{H} \mathbf{A}^f)$$

$$\mathbf{K} = \mathbf{P}_e^f \mathbf{H}^T (\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e)^{-1}$$

$$= \mathbf{A}^f \mathbf{A}'^{fT} \mathbf{H}^T (\mathbf{H} \mathbf{A}^f \mathbf{A}'^{fT} \mathbf{H}^T + \Upsilon \Upsilon^T)^{-1}$$

$$= \mathbf{A}^f (\mathbf{H} \mathbf{A}^f)^T ((\mathbf{H} \mathbf{A}^f) (\mathbf{H} \mathbf{A}^f)^T + \Upsilon \Upsilon^T)^{-1}$$

These equations do not depend upon explicit knowledge of the forecast error covariance matrix. It is implied by the ensemble of forecast states. Warning: there is a possible rank deficiency if $N < p$.

The ensemble mean analysis equation in terms of matrix representation of the ensemble

$$\begin{aligned}\bar{\mathbf{A}}^a &= \mathbf{A}^a \mathbf{1}_{N \times N} = \mathbf{A}^f \mathbf{1}_{N \times N} + \mathbf{K} (\mathbf{D} - \mathbf{H} \mathbf{A}^f) \mathbf{1}_{N \times N} \\ &= \bar{\mathbf{A}}^f + \mathbf{K} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f)\end{aligned}$$

The analysis error covariance in terms of matrix representations of the ensemble

$$\begin{aligned}\mathbf{P}_e^a &= \frac{1}{N-1} \mathbf{A}^a \mathbf{A}^{aT} = \frac{1}{N-1} \mathbf{A}^a (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N}) (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N}) \mathbf{A}^{aT} \\ &= \frac{1}{N-1} \mathbf{A}^a (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N}) \mathbf{A}^{aT} \\ &= \frac{1}{N-1} [\mathbf{A}^f + \mathbf{K} (\mathbf{D} - \mathbf{H} \mathbf{A}^f)] (\mathbf{I}_{N \times N} - \mathbf{1}_{N \times N}) [\mathbf{A}^f + \mathbf{K} (\mathbf{D} - \mathbf{H} \mathbf{A}^f)]^T \\ &= \mathbf{P}_e^f - \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T - \mathbf{K} \mathbf{H} \mathbf{P}_e^f + \mathbf{K} \mathbf{R}_e \mathbf{K}^T + \mathbf{K} \mathbf{H} \mathbf{P}_e^f \mathbf{H}^T \mathbf{K}^T \\ &= (\mathbf{I} - \mathbf{K} \mathbf{H}) \mathbf{P}_e^f \quad \text{using analysis beforehand.}\end{aligned}$$

Factorization in terms of S and C matrices

The ensemble mean analysis can be written in terms of matrices **S** and **C**.

$$\bar{\mathbf{A}}^a = \bar{\mathbf{A}}^f + \mathbf{A}^f \mathbf{S}^T \mathbf{C}^{-1} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f)$$

$$\text{where} \quad \mathbf{S} = \mathbf{H} \mathbf{A}^f \quad \mathbf{C} = \mathbf{S} \mathbf{S}^T + (N-1) \mathbf{R}_e$$

S is the matrix containing model observation perturbations (calculated from the matrix of forecast perturbations), and **C** is $(N-1)$ times the sum of model observation covariances, $\mathbf{S} \mathbf{S}^T$, and observation error covariances, \mathbf{R}_e .

The above result can be checked by substitution

$$\begin{aligned}\bar{\mathbf{A}}^a &= \bar{\mathbf{A}}^f + \mathbf{A}^f \mathbf{A}^{fT} \mathbf{H}^T [\mathbf{H} \mathbf{A}^f \mathbf{A}^{fT} \mathbf{H}^T + (N-1) \mathbf{R}_e]^{-1} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f) \\ &= \bar{\mathbf{A}}^f + (N-1) \mathbf{P}_e^f \mathbf{H}^T [\mathbf{H} (N-1) \mathbf{P}_e^f \mathbf{H}^T + (N-1) \mathbf{R}_e]^{-1} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f) \\ &= \bar{\mathbf{A}}^f + \mathbf{P}_e^f \mathbf{H}^T [\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e]^{-1} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f) \\ &= \bar{\mathbf{A}}^f + \mathbf{K} (\bar{\mathbf{D}} - \mathbf{H} \bar{\mathbf{A}}^f) \quad \text{expression is confirmed.}\end{aligned}$$

\mathbf{P}_e^a can also be written in terms of matrices **S** and **C**.

$$\mathbf{P}_e^a = \frac{1}{N-1} \mathbf{A}^f (\mathbf{I} - \mathbf{S}^T \mathbf{C}^{-1} \mathbf{S}) \mathbf{A}^{fT}$$

This result can also be checked by substitution

$$\begin{aligned}\mathbf{P}_e^a &= \frac{1}{N-1} \mathbf{A}^f \mathbf{A}^{fT} - \frac{1}{N-1} \mathbf{A}^f \mathbf{A}^{fT} \mathbf{H}^T [\mathbf{H} \mathbf{A}^f \mathbf{A}^{fT} \mathbf{H}^T + (N-1) \mathbf{R}_e]^{-1} \mathbf{H} \mathbf{A}^f \mathbf{A}^{fT} \\ &= \mathbf{P}_e^f - \mathbf{P}_e^f \mathbf{H}^T [\mathbf{H} (N-1) \mathbf{P}_e^f \mathbf{H}^T + (N-1) \mathbf{R}_e]^{-1} (N-1) \mathbf{H} \mathbf{P}_e^f \\ &= \mathbf{P}_e^f - \mathbf{P}_e^f \mathbf{H}^T [\mathbf{H} \mathbf{P}_e^f \mathbf{H}^T + \mathbf{R}_e]^{-1} \mathbf{H} \mathbf{P}_e^f\end{aligned}$$

$$\begin{aligned}
&= \mathbf{P}_e^f - \mathbf{KHP}_e^f \\
&= (\mathbf{I} - \mathbf{KH})\mathbf{P}_e^f \quad \text{expression is confirmed.}
\end{aligned}$$

The expressions involving \mathbf{S} and \mathbf{C} are useful because the matrices \mathbf{S} and \mathbf{C} are calculable for a given ensemble. Both exist in observation space.

A 'square root' analysis scheme

It is desirable to calculate the square-root of \mathbf{P}_e^a . The square-root matrix may be regarded as a set of ensemble perturbations that have covariance \mathbf{P}_e^a . Given that \mathbf{C} is available, write it in terms of its eigenvalue decomposition

$$\mathbf{C} = \mathbf{Z}\mathbf{\Lambda}\mathbf{Z}^T$$

where $\mathbf{\Lambda}$ is the (diagonal) matrix of eigenvalues and \mathbf{Z} is the orthonormal matrix of eigenvectors. We assume that all eigenvalues of \mathbf{C} are non-zero, which will allow calculation of its inverse. \mathbf{P}_e^a is then written

$$\begin{aligned}
\mathbf{P}_e^a &= \frac{1}{N-1} \mathbf{A}^{af} (\mathbf{I} - \mathbf{S}^T \mathbf{Z} \mathbf{\Lambda}^{-1} \mathbf{Z}^T \mathbf{S}) \mathbf{A}^{afT} \\
&= \frac{1}{N-1} \mathbf{A}^{af} (\mathbf{I} - \mathbf{X}^T \mathbf{X}) \mathbf{A}^{afT} \\
&\quad \text{where} \quad \mathbf{X} = \mathbf{\Lambda}^{-1/2} \mathbf{Z}^T \mathbf{S}
\end{aligned}$$

Matrix \mathbf{X} may be further written in terms of a singular value decomposition in the following way

$$\mathbf{X} = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^T$$

where $\mathbf{\Sigma}$ is the matrix of singular values, \mathbf{U} is the orthonormal matrix of left singular vectors and \mathbf{V} is the orthonormal matrix of right singular vectors. \mathbf{P}_e^a is then written

$$\begin{aligned}
\mathbf{P}_e^a &= \frac{1}{N-1} \mathbf{A}^{af} (\mathbf{I} - \mathbf{V} \mathbf{\Sigma}^T \mathbf{U}^T \mathbf{U} \mathbf{\Sigma} \mathbf{V}^T) \mathbf{A}^{afT} \\
&= \frac{1}{N-1} \mathbf{A}^{af} (\mathbf{I} - \mathbf{V} \mathbf{\Sigma}^T \mathbf{\Sigma} \mathbf{V}^T) \mathbf{A}^{afT} \\
&= \frac{1}{N-1} \mathbf{A}^{af} \mathbf{V} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma}) \mathbf{V}^T \mathbf{A}^{afT}
\end{aligned}$$

$\mathbf{\Sigma}^T \mathbf{\Sigma}$ is a diagonal matrix of size $N \times N$. The square-root of the combined diagonal matrix $\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma}$ is therefore trivial to compute.

$$\begin{aligned}
\mathbf{P}_e^a &= \frac{1}{N-1} \mathbf{A}^{af} \mathbf{A}^{afT} = \frac{1}{N-1} \mathbf{A}^{af} \mathbf{V} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma})^{1/2} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma})^{1/2} \mathbf{V}^T \mathbf{A}^{afT} \\
\therefore \quad \mathbf{A}^{af} \mathbf{A}^{afT} &= [\mathbf{A}^{af} \mathbf{V} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma})^{1/2}] [\mathbf{A}^{af} \mathbf{V} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma})^{1/2}]^T \quad \text{and} \\
\mathbf{A}^{af} &= \mathbf{A}^{af} \mathbf{V} (\mathbf{I} - \mathbf{\Sigma}^T \mathbf{\Sigma})^{1/2}
\end{aligned}$$