GAME-DA Documentation

GAME-DA Development Team

Chapter 1

Optimum interpolation

Let $M \ge 1$ be the number of degrees freedom of the model, $N \ge 1$ the number of observations, $\mathbf{h} \in \mathbb{R}^N$ the the observations as reconstructed from the background state, $\mathbf{y} \in \mathbb{R}^N$ the actual observations, $\overrightarrow{H} \in \mathbb{R}^{N \times M}$ the Jacobian of the observations operator, $\overrightarrow{H} \in \mathbb{R}^{M \times M}$ the background error covariance matrix and $\overrightarrow{R} \in \mathbb{R}^{N \times N}$ the model error covariance matrix. It is

$$\left(\overleftrightarrow{H} \stackrel{\longleftrightarrow}{B}\right)_{i,j} = \sum_{k=1}^{M} \stackrel{\longleftrightarrow}{H}_{i,k} \stackrel{\longleftrightarrow}{B}_{k,j}. \tag{1.1}$$

If \overrightarrow{B} is diagonal, this reduces to

$$\left(\overrightarrow{H} \overrightarrow{B}\right)_{i,j} = \sum_{k=1}^{M} \overrightarrow{H}_{i,k} \overrightarrow{B}_{k,j} \delta_{k,j} = \sum_{k=1}^{M} \overrightarrow{H}_{i,k} \overrightarrow{B}_{j,j} \delta_{k,j} = \overrightarrow{H}_{i,j} \overrightarrow{B}_{j,j}.$$
(1.2)

For $\overrightarrow{H} \overrightarrow{B} \overrightarrow{H}^T$, one obtains

$$\left(\overrightarrow{H} \stackrel{\longleftrightarrow}{B} \stackrel{\longleftrightarrow}{H}^T\right)_{i,j} = \left[\left(\overrightarrow{H} \stackrel{\longleftrightarrow}{B}\right) \stackrel{\longleftrightarrow}{H}^T\right]_{i,j} = \sum_{k=1}^{M} \left(\overrightarrow{H} \stackrel{\longleftrightarrow}{B}\right)_{i,k} \stackrel{\longleftrightarrow}{H}_{k,j} = \sum_{k=1}^{M} \left(\overrightarrow{H} \stackrel{\longleftrightarrow}{B}\right)_{i,k} \stackrel{\longleftrightarrow}{H}_{j,k} = \sum_{k=1}^{M} \left(\sum_{l=1}^{M} \stackrel{\longleftrightarrow}{H}_{i,l} \stackrel{\longleftrightarrow}{B}_{l,k}\right) \stackrel{\longleftrightarrow}{H}_{j,k} = \sum_{k,l=1}^{M} \stackrel{\longleftrightarrow}{H}_{i,l} \stackrel{\longleftrightarrow}{B}_{l,k} \stackrel{\longleftrightarrow}{H}_{j,k}.$$
 (1.3)

For the final result, we obtain

$$x_{i} = x_{B,i} + \sum_{j=1}^{N} \left[\overrightarrow{B} \overrightarrow{H}^{T} \left(\overrightarrow{H} \overrightarrow{B} \overrightarrow{H}^{T} + \overrightarrow{R} \right)^{-1} \right]_{i,j} (y_{j} - h_{j}) = x_{B,i} + \sum_{j=1}^{N} \left[\sum_{k=1}^{N} \left(\overrightarrow{B} \overrightarrow{H}^{T} \right)_{i,k} \left(\overrightarrow{H} \overrightarrow{B} \overrightarrow{H}^{T} + \overrightarrow{R} \right)_{k,j}^{-1} \right] (y_{j} - h_{j})$$

$$= x_{B,i} + \sum_{j=1}^{N} \left[\sum_{k=1}^{N} \left(\overrightarrow{H} \overrightarrow{B} \overrightarrow{H}^{T} + \overrightarrow{R} \right)_{k,j}^{-1} \left(\overrightarrow{B} \overrightarrow{H}^{T} \right)_{i,k} \right] (y_{j} - h_{j}) = x_{B,i} + \sum_{j=1}^{N} \left[\sum_{k=1}^{N} \left(\overrightarrow{H} \overrightarrow{B} \overrightarrow{H}^{T} + \overrightarrow{R} \right)_{k,j}^{-1} \left(\sum_{l=1}^{M} \overrightarrow{B} \overrightarrow{H}^{T} \right)_{l,k} \right] (y_{j} - h_{j})$$

$$\Rightarrow x_{i} = x_{B,i} + \sum_{j=1}^{N} \left[\sum_{k=1}^{N} \left(\overleftrightarrow{H} \stackrel{\longleftrightarrow}{B} \stackrel{\longleftrightarrow}{H}^{T} + \stackrel{\longleftrightarrow}{R} \right)_{k,j}^{-1} \left(\sum_{l=1}^{M} \stackrel{\longleftrightarrow}{B}_{i,l} \stackrel{\longleftrightarrow}{H}_{k,l} \right) \right] (y_{j} - h_{j}). \tag{1.4}$$

1.1 Inclusion of moisture

So far, moisture is taken into account in a separate assimilation process for efficiency.

Chapter 2

3D-Var

2.1 Technicalities

The observations used come from a time window $\left[-\frac{T}{2},\frac{T}{2}\right]$ around the analysis time and are all taken to be valid at the analysis time. T=3 h is a typical value..

Chapter 3

4D-Var

3.1 Technicalities

The observations used come from a time window $\left[-\frac{T}{2},\frac{T}{2}\right]$ around the analysis time and are taken to be valid at individual time steps $n\Delta t$. T=6 h and $\Delta t=15$ min are typical values.