

Inverse Theory Cheat Sheet

Linear Inverse Problems

Definition of the problem:

$$\mathbf{A}\mathbf{m} = \mathbf{d}$$

Where \mathbf{A} is the forward problem, \mathbf{m} is the model, and \mathbf{d} is the data vector. \mathbf{A} and \mathbf{d} are known, and \mathbf{m} is not.

Rescale using covariance matrix:

$$\begin{aligned}\hat{\mathbf{A}} &= \mathbf{C}^{1/2}\mathbf{A} \\ \hat{\mathbf{d}} &= \mathbf{C}^{1/2}\mathbf{d}\end{aligned}$$

We want:

$$\hat{\boldsymbol{\varepsilon}} \equiv \hat{\mathbf{d}} - \hat{\mathbf{A}}\mathbf{m} \rightarrow \mathbf{0}$$

and

$$\nu\mathbf{L}\mathbf{m} \rightarrow \mathbf{0}$$

Find \mathbf{m} that minimizes

$$\hat{\boldsymbol{\varepsilon}}^T\hat{\boldsymbol{\varepsilon}} + \nu^2(\mathbf{L}\mathbf{m})^T(\mathbf{L}\mathbf{m})$$

This requirement is satisfied by finding \mathbf{m} using the following equation:

$$\begin{bmatrix} \hat{\mathbf{A}} \\ \nu\mathbf{L} \end{bmatrix} \mathbf{m} = \begin{bmatrix} \hat{\mathbf{d}} \\ \mathbf{0} \end{bmatrix}$$

The solution for \mathbf{m} is:

$$\mathbf{m} = (\hat{\mathbf{A}}^T\hat{\mathbf{A}} + \nu^2\mathbf{L}^T\mathbf{L})^{-1}(\hat{\mathbf{A}}\hat{\mathbf{d}} + \nu\mathbf{L}^T\mathbf{0})$$

(Weakly) Non-linear Inverse Problems

In this case we make an initial guess \mathbf{m}_o :

$$\mathbf{m}_o \Rightarrow \mathbf{m} = \mathbf{m}_o + \delta\mathbf{m}$$

The forward problem is defined by

$$\begin{aligned}d_i &= F_i(\mathbf{m}) \\ d_i &= F_i(\mathbf{m} + \delta\mathbf{m})\end{aligned}$$

F_i is expanded using the first term of a Taylor series expansion:

$$d_i = F_i(\mathbf{m}_o) + \frac{\partial F_i}{\partial m_j} \delta m_j$$

$$\delta d \equiv \mathbf{d} - \mathbf{d}^o = \mathbf{A}^o \delta \mathbf{m}$$

where \mathbf{d} is the real data, and \mathbf{d}^o is the data predicted by \mathbf{m}_o .

Re-scaling:

$$\begin{aligned}\hat{\mathbf{A}}^o &= \mathbf{C}^{-1/2}\mathbf{A}^o \\ \delta\hat{\mathbf{d}} &= \mathbf{C}^{-1/2}\delta\mathbf{d}\end{aligned}$$

Scaled version:

$$\hat{\mathbf{A}}^o \delta \mathbf{m} = \delta \hat{\mathbf{d}}$$

We want:

$$\hat{\boldsymbol{\varepsilon}} \equiv \delta\hat{\mathbf{d}} - \hat{\mathbf{A}}^o\delta\mathbf{m} \rightarrow \mathbf{0}$$

and

$$\nu\mathbf{L}\mathbf{m} \rightarrow \mathbf{0}$$

What is \mathbf{m} ?

$$\begin{aligned}\nu\mathbf{L}(\mathbf{m}_o + \delta\mathbf{m}) &\rightarrow \mathbf{0} \\ \nu\mathbf{L}\delta\mathbf{m} &= -\nu\mathbf{L}\mathbf{m}_o\end{aligned}$$

$$\begin{bmatrix} \hat{\mathbf{A}}^o \\ \nu\mathbf{L} \end{bmatrix} \delta \mathbf{m} = \begin{bmatrix} \delta \hat{\mathbf{d}} \\ -\nu\mathbf{L}\mathbf{m}_o \end{bmatrix}$$

The solution for $\delta\mathbf{m}$ is :

$$\delta\mathbf{m} = (\hat{\mathbf{A}}^{oT}\hat{\mathbf{A}}^o + \nu^2\mathbf{L}^T\mathbf{L})^{-1}(\hat{\mathbf{A}}^{oT}\delta\hat{\mathbf{d}} - \nu^2\mathbf{L}^T\mathbf{L}\mathbf{m}_o)$$

$$\mathbf{m}^{new} = \mathbf{m}_o + \delta\mathbf{m} \quad (1)$$

Now start again, substituting \mathbf{m}^{new} for \mathbf{m}_o . Iterate until $\delta\mathbf{m}$ is sufficiently small (depends on the problem).