

# Optimal filtering

## 1 Fitting data

We want to fit equispaced data points imposing Tikhonov regularization on the fitting function. We do this by minimizing a functional  $\Omega$ :

$$\Omega \equiv \sum_i (f_i - g_i)^2 + \Lambda \cdot \sum_i (f_i - f_{i-1})^2 = \text{minimum.} \quad (1)$$

The solution of Eq. 1 can be found by setting all components of the gradient vector to 0:

$$\frac{1}{2} \frac{\partial \Omega}{\partial f_k} = (f_k - g_k) + \Lambda \cdot (2f_k - f_{k-1} - f_{k+1}) = 0. \quad (2)$$

We can re-write this system of linear equations in the form:

$$\sum_i A_{ij} f_i = g_j, \quad (3)$$

where:

$$A_{ij} = \begin{pmatrix} 1 + \Lambda & -\Lambda & 0 & \dots & 0 \\ -\Lambda & 1 + 2\Lambda & -\Lambda & & \\ 0 & -\Lambda & 1 + 2\Lambda & -\Lambda & \\ \dots & & & & \\ & & & -\Lambda & 1 + 2\Lambda & -\Lambda \\ & & & & -\Lambda & 1 + \Lambda \end{pmatrix} \quad (4)$$

Matrix  $A_{ij}$  is tri-diagonal and, therefore, one can quickly solve Eq. 3 and find  $f_i$ .

You may try to program this and fit different data: sine curve, Gaussian, noise with different values of  $\Lambda$ .

## 2 Interpolation

Interpolation here means that we want to find  $f_l$  on an equispaced grid *which is denser than the grid of the original data*.

The Eq. 1 has to be re-written as following:

$$\Omega \equiv \sum_{i=0}^{N_{\text{data}}-1} (f_{i \cdot m} - g_i)^2 + \Lambda \cdot \sum_{l=0}^{N_{\text{fit}}-1} (f_l - f_{l-1})^2 = \text{minimum,} \quad (5)$$

where  $m = N_{\text{fit}}/N_{\text{data}}$ .

1. Follow the derivation for the case of fitting and find the equation of  $f_l$  in case of interpolation.
2. Write a program to find  $f_l$  for a given data,  $\Lambda$  and  $m$
3. Construct data that samples a step-function and try to interpolate it with cubic spline and with the optimal filtering technique.