In the most general sense, we seek to estimate a smoothed signal,  $u_{\text{smooth}}$ , from observed data,  $u_{\text{obs}}$ , by incorporating prior knowledge of the signals statistical properties. In this paper, we describe the smoothed and observed signal as

$$u_{\text{smooth}} = u_{\text{obs}} + \epsilon, \quad \epsilon \sim \mathcal{N}(0, \mathbf{C}_{\text{obs}})$$
 (1)

$$u_{\text{smooth}} = u_{\text{prior}}, \quad u_{\text{prior}} \sim \mathcal{N}(0, \mathbf{C}_{\text{prior}}).$$
 (2)

We find  $u_{\rm smooth}$  that simulataneously satisfies eq. (1) and eq. (2) in a least-squares sense by minimization the objection function

$$F(u_{\text{smooth}}) = ||u_{\text{smooth}} - u_{\text{obs}}||_{\mathbf{C}_{\text{obs}}}^2 + ||u_{\text{smooth}}||_{\mathbf{C}_{\text{prior}}}^2.$$
(3)

The solution that minimizes eq. 3 is simply

$$u_{\text{smooth}} = (\mathbf{C}_{\text{obs}}^{-1} + \mathbf{C}_{\text{prior}}^{-1})^{-1} \mathbf{C}_{\text{obs}}^{-1} u_{\text{obs}}.$$
 (4)

The challenge is in choosing an appropriate  $C_{prior}$ . Below we discuss selection of  $C_{prior}$  is one-dimensional problems, which then naturally leads to an extension to selection of  $C_{prior}$  for higher dimensions.

## One-dimensional smoothing

When the signal is only assumed to covary in time,  $u_{\text{prior}}$  is commonly treated as Brownian motion (eg. Me Segall, McQuire, Murray, etc.). We treat a  $u_{\text{prior}}$  as Brownian motion by assuming that its velocity is white noise with constant variance  $\lambda^2$ . That is to say

$$\mathbf{D}_1 u_{\text{prior}} = q, \quad q \sim \mathcal{N}(0, \lambda^2) \tag{5}$$

where  $\mathbf{D}_N$  denotes an N'th order differentiation matrix. We then model  $u_{\text{prior}}$  as Brownian motion by setting  $\mathbf{C}_{\text{prior}}$  to be

$$\mathbf{C}_{\text{prior}} = \lambda^2 (\mathbf{D}_1^T \mathbf{D}_1)^{-1}. \tag{6}$$

It is also common to treate  $u_{\text{prior}}$  as integrated Brownian motion. In such case, the appropriate choice of  $\mathbf{C}_{\text{prior}}$  would just use  $\mathbf{D}_2$  rather than  $\mathbf{D}_1$ . There is still a need to select  $\lambda$ , which described how rapidly we expect the smoothed signal to vary. There are numerous methods for selecting  $\lambda$ . For example, one could use maximum likelihood methods, a trade-off curve, or simply vary  $\sigma$  until the smoothed signal looks appropriate when compared to the observations. We note that smoothing is fundamentally a low-pass filter and so it is perhaps most useful to choose  $\sigma$  based on its cut-off attenuation frequencies. We then consider the solution for  $u_{\text{smooth}}$  in the frequency domain.

For the purpose analytical tractability, we assume that  $u_{\rm smooth}$  and  $\epsilon$  are stationary stochastic processes and furthermore

$$\mathbf{C}_{\text{obs}} = \sigma^2 \mathbf{I} \tag{7}$$

and

$$\mathbf{C}_{\text{prior}} = \lambda^2 (\mathbf{D}_N^T \mathbf{D}_N)^{-1}. \tag{8}$$

The solution for  $u_{\text{smooth}}$  in the frequency domain is

$$\hat{u}_{\text{smooth}}(\omega) = \frac{\left(\frac{1}{\sigma}\right)^2}{\left(\frac{1}{\sigma}\right)^2 + \left(\frac{(2\pi\omega)^N}{\lambda}\right)^2} \hat{u}_{\text{obs}}(\omega). \tag{9}$$

We make the change of variables

$$\lambda = (2\pi\omega_c)^N \sigma \tag{10}$$

where  $\omega_c$  is now our free parameter. This simplifies eq. 9 to

$$\hat{u}_{\text{smooth}}(\omega) = \frac{1}{1 + \left(\frac{\omega}{\omega_c}\right)^{2N}} \hat{u}_{\text{obs}}(\omega). \tag{11}$$

We can recognize eq. 11 as a N'th order Butterworth filter, and  $\omega_c$  is the cut-off frequency. Thus selecting the hyperparameter,  $\lambda$ , can be translated through eq. 10 into a more tangible question of identifying an upper bound on the frequency content of the underlying signal. In the limit as  $N \to \infty$  eq. 11 becomes an ideal low-pass filter which removes all frequencies above  $\omega_c$  and leaves lower frequencies unaltered. Of course, an ideal low-pass filter is often undesirable in practice because it will tend to produce ringing artifacts in the smoothed solutions. When modeling  $u_{\text{prior}}$  as a Brownian motion or integrated Brownian motion, where N=1 and N=2 respectively, the transfer function is tapered across  $\omega_c$ , which ameliorates ringing in the smoothed solution.