

Let  $r_{sj}$  be the residual for the  $j$ th observation on the  $s$ th subject. Make a list of all within-subject normalized cross-products  $r_{sj}r_{sj'}/\sigma^2$  whether or not  $j = j'$ , and a corresponding list of absolute time lags  $|t_{sj} - t_{sj'}|$ . Let  $y_i$  be the  $i$ th cross-product on the list, and let  $x_i$  be the  $i$ th time lag. Our generalized AR-1 model is

$$E(y_i) = \rho^{x_i} + \epsilon_i$$

where  $\epsilon_i$  is some error with mean zero and finite variance and  $0 < \rho < 1$ . We rewrite this as

$$E(y_i) = \exp(-\theta x_i)$$

where  $\theta = -\log \rho$  and  $0 < \theta < \infty$ .

So we will try to minimize the loss function

$$Q(\theta) = \frac{1}{2} \sum_{i=1}^n (y_i - f(y_i))^2$$

where  $f(y_i) = \exp(-\theta x_i)$ . Then

$$\frac{dQ}{d\theta} = - \sum_{i=1}^n (y_i - f(y_i)) \frac{df(y_i)}{d\theta}$$

We will use Fisher scoring: first argue that

$$E\left(\frac{d^2Q}{d\theta^2}\right) = \sum_{i=1}^n \left(\frac{df(y_i)}{d\theta}\right)^2$$

Then to find the place where  $\frac{dQ}{d\theta} = 0$ , iterate

$$\hat{\theta}_{new} = \hat{\theta}_{old} - \left(E\left(\frac{d^2Q}{d\theta^2}\right)\right)^{-1} \frac{dQ}{d\theta}$$

Here

$$\frac{df(y_i)}{d\theta} = -x_i \exp(-\theta x_i)$$

so

$$\frac{dQ}{d\theta} = \sum_{i=1}^n (y_i - \exp(-\theta x_i)) (\exp(-\theta x_i)) x_i$$