

## Successive 1-Dimensional Maximization

To maximize  $L(\alpha, \beta|x)$  over both  $\alpha$  and  $\beta$ :

1. Fix  $\alpha$  maximize  $L(\alpha, \beta|x)$  over  $\beta$
2. Let  $\hat{\beta}(\alpha)$  be the value of  $\beta$  that maxs  $L(\alpha, \beta|x)$  for fixed  $\alpha$
3. Profiled likelihood function for  $\alpha$ ,  $H(\alpha|x) = L(\alpha, \hat{\beta}(\alpha)|x)$  depends on  $\alpha$
4. The MLE of  $\beta$  is  $\hat{\beta}(\hat{\alpha}_H)$  where  $\hat{\alpha}_H$  maxs  $H(\alpha|x)$

## Invariance Property of MLE

**Thm:** If  $\hat{\theta}$  is the MLE of  $\theta$ , then for any function  $\tau(\theta)$ , the MLE of  $\tau(\theta)$  is  $\tau(\hat{\theta})$

If mapping  $\theta \rightarrow \tau(\theta)$  is one-to-one, letting  $\eta = \tau(\theta)$ , the MLE of  $\eta$  is the same since:

$$L^*(\eta|x) = \prod_{i=1}^n f(x_i|\tau^{-1}(\eta)) = L(\tau^{-1}(\eta)|x) \text{ and}$$

$$\sup_{\eta} L^*(\eta|x) = \sup_{\eta} L(\tau^{-1}(\eta)|x) = \sup_{\theta} L(\theta|x)$$