

## Confidence Intervals for Nonlinear Regression Parameters

1. The matrix  $\mathbf{F}(\boldsymbol{\theta}) = \mathbf{F}$  plays the role in nonlinear least squares the  $\mathbf{X}$  matrix plays in linear least squares.
2. If  $\boldsymbol{\varepsilon} \sim N(\mathbf{0}, \mathbf{I}\sigma^2)$ , then  $\hat{\boldsymbol{\theta}}$  is *approximately* normally distributed with mean  $\boldsymbol{\theta}$  and  $\text{Var}(\hat{\boldsymbol{\theta}}) = (\mathbf{F}'\mathbf{F})^{-1}\sigma^2$ . (Gallant 1987)
3. The residual sum of squares  $S(\hat{\boldsymbol{\theta}})$  when divided by  $\sigma^2$ , has *approximately* a chi-squared distribution with  $n - p$  degrees of freedom.
4. In practice,  $\mathbf{F}(\boldsymbol{\theta})$  is computed as  $\mathbf{F}(\hat{\boldsymbol{\theta}}) = \hat{\mathbf{F}}$  and  $\sigma^2$  is estimated with  $s^2 = S(\hat{\boldsymbol{\theta}})/(n - p)$ , so that the estimated variance-covariance matrix for  $\hat{\boldsymbol{\theta}}$  is  $s^2(\hat{\boldsymbol{\theta}}) = (\hat{\mathbf{F}}'\hat{\mathbf{F}})^{-1}s^2$
5. Confidence Intervals for  $\boldsymbol{\theta}$ . Let  $C = \mathbf{K}'\boldsymbol{\theta}$  be any linear function of interest (point estimate is  $\hat{C} = \mathbf{K}'\hat{\boldsymbol{\theta}}$ ).

100(1 -  $\alpha$ )% CI estimate of  $C$  is:  $\hat{C} \pm t_{\alpha/2}(n - p)s(\hat{C})$

where  $s(\hat{C}) = \{\mathbf{K}'[s^2(\hat{\boldsymbol{\theta}})]\mathbf{K}\}^{1/2}$

6. Likelihood Ratio Theory: Approximate joint 100(1 -  $\alpha$ )% confidence region for  $\hat{\boldsymbol{\theta}}$ :  
Set of  $\hat{\boldsymbol{\theta}}$  s.t.

$$S(\boldsymbol{\theta}) \leq S(\hat{\boldsymbol{\theta}}) \left[ 1 + \frac{p}{n - p} F_{\alpha}(p, n - p) \right]$$