

# MAXIMUM LIKELIHOOD ESTIMATION

## 1. General Linear Model Estimation

$$y = X\beta + \varepsilon, \varepsilon \sim N(0, \sigma^2 V) \Rightarrow y \sim N(X\beta, \sigma^2 V)$$

$$\Rightarrow L(\beta, \sigma^2 | y, X, V) = \ln \phi(y | X, V, \beta, \sigma^2) \quad (V \text{ known})$$

$$= -\frac{n}{2} \ln(2\pi) - \frac{1}{2} \ln(\sigma^2) - \frac{1}{2} \ln |V| - \frac{1}{2\sigma^2} (y - X\beta)' V^{-1} (y - X\beta)$$

$$\Rightarrow \hat{\beta} = (X' V^{-1} X)^{-1} X' V^{-1} y$$

$$\hat{\sigma}^2 = \frac{1}{n} (y - X\hat{\beta})' V^{-1} (y - X\hat{\beta})$$

## 2. SAR-Model Estimation

$$y = X\beta + u, u = \rho W u + \varepsilon, \varepsilon \sim N(0, \sigma^2 I_n)$$

$$\Rightarrow u = (I_n - \rho W)^{-1} \varepsilon = B_\rho^{-1} \varepsilon \Rightarrow y \sim N(X\beta, \sigma^2 (B_\rho' B_\rho)^{-1})$$

$$\Rightarrow \hat{\beta}_\rho = (X' B_\rho' B_\rho X)^{-1} X' B_\rho' B_\rho y$$

$$\hat{\sigma}_\rho^2 = \frac{1}{n} (y - X\hat{\beta}_\rho)' B_\rho' B_\rho (y - X\hat{\beta}_\rho)$$

Solve for  $\rho$  by substitution into the **Concentrated Likelihood**:

$$\max_\rho L(\rho | y, X) = -\frac{n}{2} (\ln(2\pi) + 1) + \ln |B_\rho| - \frac{n}{2} \ln(\hat{\sigma}_\rho^2)$$

## 2. SL-Model Estimation

$$y = \lambda W y + X \beta + \varepsilon, \quad \varepsilon \sim N(0, \sigma^2 I_n)$$

$$\Rightarrow B_\lambda y = (I_n - \lambda W) y = X \beta + \varepsilon$$

$$\Rightarrow y = B_\lambda^{-1} X \beta + B_\lambda^{-1} \varepsilon \Rightarrow y \sim N\left(B_\lambda^{-1} X \beta, \sigma^2 (B_\lambda' B_\lambda)^{-1}\right)$$

$$\Rightarrow \hat{\beta}_\lambda = (X' X)^{-1} X' B_\lambda y$$

$$\hat{\sigma}_\lambda^2 = \frac{1}{n} (B_\lambda y - X \hat{\beta}_\lambda)' (B_\lambda y - X \hat{\beta}_\lambda)$$

Solve for  $\lambda$  by substitution into **Concentrated Likelihood**:

$$\max_\lambda L(\lambda | y, X) = -\frac{n}{2} (\ln(2\pi) + 1) + \ln |B_\lambda| - \frac{n}{2} \ln(\hat{\sigma}_\lambda^2)$$