

# I. Concluding Topics in Hypothesis Testing

## A. Testing Nonlinear Restrictions

### 1. Format (Wald Strategy):

$$H_0: f(\mathbf{\beta}) = q \Rightarrow z = \frac{f(\mathbf{b}) - q}{s.e.(f(\mathbf{b}))} \sim^A N(0,1) ; \text{ n.b., usually use } t_{n-k}$$

### 2. Issues:

a) Calculating  $f(\mathbf{b})$  given estimates  $\mathbf{b}$  is no problem

b) But how do we calculate  $V(f(\mathbf{b}))$  for  $f(\mathbf{b})$  nonlinear?

### 3. Taylor-Series Approximation (Linearization) $\Rightarrow$ "Delta Method"

a)  $f(\mathbf{b}) \approx f(\mathbf{\beta}) + \nabla_{\mathbf{\beta}} f(\mathbf{\beta}) \cdot (\mathbf{b} - \mathbf{\beta})$  ; recall  $\nabla_{\mathbf{\beta}} f(\mathbf{\beta}) \equiv \frac{\partial f(\mathbf{\beta})}{\partial \mathbf{\beta}}$  is vector

b)  $V\{\nabla_{\mathbf{\beta}} f(\mathbf{\beta}) \cdot (\mathbf{b} - \mathbf{\beta})\} = "[\nabla_{\mathbf{\beta}} f(\mathbf{\beta})]^2 V(\mathbf{b})" = [\nabla_{\mathbf{\beta}} f(\mathbf{\beta})]' V(\mathbf{b}) [\nabla_{\mathbf{\beta}} f(\mathbf{\beta})]$

### 4. So our test statistic:

$$t = \left\{ \begin{array}{l} \frac{f(\mathbf{b}) - q}{\sqrt{[\nabla_{\mathbf{\beta}} f(\mathbf{\beta})]' [V(\mathbf{b})] [\nabla_{\mathbf{\beta}} f(\mathbf{\beta})]}} = \frac{f(\mathbf{b}) - q}{\sqrt{\left[ \frac{\partial f(\mathbf{\beta})}{\partial \mathbf{\beta}} \right]' [V(\mathbf{b})] \left[ \frac{\partial f(\mathbf{\beta})}{\partial \mathbf{\beta}} \right]}} = \\ \frac{f(\mathbf{b}) - q}{\sqrt{\left[ \frac{\partial f(\mathbf{\beta})}{\partial \beta_0} \quad \frac{\partial f(\mathbf{\beta})}{\partial \beta_1} \quad \dots \quad \frac{\partial f(\mathbf{\beta})}{\partial \beta_k} \right] [V(\mathbf{b})] \left[ \frac{\partial f(\mathbf{\beta})}{\partial \beta_0} \quad \frac{\partial f(\mathbf{\beta})}{\partial \beta_1} \quad \dots \quad \frac{\partial f(\mathbf{\beta})}{\partial \beta_k} \right]'}} \end{array} \right\} \sim^A N(0,1) ; \text{ (often use } t_{n-k})$$

### 5. Joint Nonlinear Hypotheses:

$$H_0: f(\mathbf{\beta}) = \mathbf{q} \Rightarrow z = [f(\mathbf{b}) - \mathbf{q}]' [V(f(\mathbf{b}))]^{-1} [f(\mathbf{b}) - \mathbf{q}] \sim^A \chi^2_J ; J = \# \text{ restricts/hypothesis}$$

### 6. Example/Illustration! (See handwritten notes, next page)

$$y_t = \beta_0 + \beta_1 y_{t-1} + \beta_2 X_t + \varepsilon_t$$

• let  $\beta_0 = \varepsilon_t = 0$ ,  
 if  $X_t = 0$  &  $X_{t+i} = 1 \quad \forall i > 0$

$$\Rightarrow y_{t+1} = \beta_2 X_{t+1}$$

$$y_{t+2} = \beta_1 \beta_2 X_{t+1} + \beta_2 X_{t+2}$$

$$y_{t+3} = \beta_1 (\beta_1 \beta_2 X_{t+1} + \beta_2 X_{t+2}) + \beta_2 X_{t+3}$$

$$y_{t+4} = \beta_1^3 \beta_2 X_{t+1} + \beta_1^2 \beta_2 X_{t+2} + \beta_1 \beta_2 X_{t+3} + \beta_2 X_{t+4}$$

$$y_{t+\infty} = \sum_{i=0}^{\infty} \beta_1^i \beta_2 X_{t+i}$$

$$= \frac{1}{1-\beta_1} \cdot \beta_2 X_{t+\infty} \quad \text{for } 0 < \beta_1 < 1$$

LR effect of  $X_t = \frac{\beta_2}{1-\beta_1} \equiv \text{LRE}$

$$V(\hat{\beta}^2 \cdot (1-\hat{\beta}_1)^{-1}) \approx \left[ \frac{\partial \text{LRE}}{\partial \beta} \right]' V(\hat{\beta}) \left[ \frac{\partial \text{LRE}}{\partial \beta} \right]$$

$$\approx \begin{bmatrix} \hat{\beta}_2 \\ (1-\hat{\beta}_1)^2 \\ 1 \\ 1-\hat{\beta}_1 \end{bmatrix}' V(\hat{\beta}) \begin{bmatrix} \hat{\beta}_2 \\ (1-\hat{\beta}_1)^2 \\ 1 \\ 1-\hat{\beta}_1 \end{bmatrix}$$