

# **MGE-201: Test basados en la verosimilitud**

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## Test basados en la verosimilitud

### Problema:

Deseamos probar hipótesis (no lineales) de la forma:

$$H_0 : \mathbf{g}(\boldsymbol{\theta}) = \mathbf{0}, \quad \text{versus} \quad H_1 : \mathbf{g}(\boldsymbol{\theta}) \neq \mathbf{0},$$

donde  $\mathbf{g} : \mathbb{R}^p \rightarrow \mathbb{R}^q$ , tal que  $\mathbf{G}(\boldsymbol{\theta}) = \partial \mathbf{g}(\boldsymbol{\theta}) / \partial \boldsymbol{\theta}^\top$  es una matriz  $q \times p$  con rango  $q$ . En otras palabras, bajo  $H_0$  deseamos resolver el problema restringido:

$$\max_{\boldsymbol{\theta} \in \Theta} \ell_n(\boldsymbol{\theta}), \quad \text{sujeto a: } \mathbf{g}(\boldsymbol{\theta}) = \mathbf{0}.$$

## Test basados en la verosimilitud

*Observación:*

Hipótesis del tipo

$$H_0 : \mathbf{g}(\boldsymbol{\theta}) = \mathbf{0}, \quad \text{vs.} \quad H_1 : \mathbf{g}(\boldsymbol{\theta}) \neq \mathbf{0},$$

se dicen en **forma implícita**.

Mientras que,

$$H_0 : \boldsymbol{\theta} = \mathbf{h}(\boldsymbol{\gamma}), \quad \text{vs.} \quad H_1 : \boldsymbol{\theta} \neq \mathbf{h}(\boldsymbol{\gamma}),$$

con espacio paramétrico nulo,

$$\Theta_0 = \{\boldsymbol{\theta} : \boldsymbol{\theta} = \mathbf{h}(\boldsymbol{\gamma}), \boldsymbol{\gamma} \in \Gamma\}, \quad \mathbf{h}(\boldsymbol{\gamma}) = (h_1(\boldsymbol{\gamma}), \dots, h_p(\boldsymbol{\gamma}))^\top,$$

está en **forma explícita**. Finalmente, la hipótesis nula definida como:

$$H_0 : \{\boldsymbol{\theta} : \exists \mathbf{a} \in \mathcal{A} \subset \mathbb{R}^k, \mathbf{g}(\boldsymbol{\theta}, \mathbf{a}) = \mathbf{0}\},$$

está escrita en **forma mixta**.<sup>1</sup>

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<sup>1</sup>Gourieroux y Monfort (1989). Econometric Theory 5, 63-82.

## Test basados en la verosimilitud

### Resultado 1 (test de Wald):<sup>2</sup>

El **test de Wald** para probar  $H_0 : \mathbf{g}(\boldsymbol{\theta}) = \mathbf{0}$ , es definido por la región crítica,

$$\{W_n \geq \chi^2_{1-\alpha}(q)\},$$

donde

$$W_n = n\mathbf{g}^\top(\widehat{\boldsymbol{\theta}}_n)[\mathbf{G}(\widehat{\boldsymbol{\theta}}_n)\mathcal{F}^{-1}(\widehat{\boldsymbol{\theta}}_n)\mathbf{G}^\top(\widehat{\boldsymbol{\theta}}_n)]^{-1}\mathbf{g}(\widehat{\boldsymbol{\theta}}_n),$$

y bajo  $H_0$  es asintóticamente de tamaño  $\alpha$ .

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<sup>2</sup>Wald (1943). Transactions of the American Mathematical Society **54**, 426-482.

## Test basados en la verosimilitud

*Demostración:*

Sabemos que

$$\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \xrightarrow{D} N_p(\mathbf{0}, \mathcal{F}^{-1}(\boldsymbol{\theta}_0)).$$

De ahí que

$$\sqrt{n}(\mathbf{g}(\hat{\boldsymbol{\theta}}_n) - \mathbf{g}(\boldsymbol{\theta}_0)) \xrightarrow{D} N_q(\mathbf{0}, \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)).$$

Bajo  $H_0$ , tenemos  $\mathbf{g}(\boldsymbol{\theta}_0) = \mathbf{0}$ . Luego,

$$\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \xrightarrow{D} N_q(\mathbf{0}, \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)). \quad (\text{W.1})$$

Como  $\mathbf{G}(\boldsymbol{\theta}_0)$  es de rango fila completo, sigue que

$$\sqrt{n}[\mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)]^{-1/2}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \xrightarrow{D} N_q(\mathbf{0}, \mathbf{I}).$$

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Note que  $\widehat{\boldsymbol{\theta}}_n$  es un estimador consistente de  $\boldsymbol{\theta}_0$ . Haciendo,

$$\mathbf{Z}_n = \sqrt{n}[\mathbf{G}(\widehat{\boldsymbol{\theta}}_n)\mathcal{F}^{-1}(\widehat{\boldsymbol{\theta}}_n)\mathbf{G}^\top(\widehat{\boldsymbol{\theta}}_n)]^{-1/2}\mathbf{g}(\widehat{\boldsymbol{\theta}}_n) \xrightarrow{\text{D}} \mathbf{N}_q(\mathbf{0}, \mathbf{I}).$$

De ahí que, bajo  $H_0$

$$\begin{aligned} W_n &= n\mathbf{g}^\top(\widehat{\boldsymbol{\theta}}_n)[\mathbf{G}(\widehat{\boldsymbol{\theta}}_n)\mathcal{F}^{-1}(\widehat{\boldsymbol{\theta}}_n)\mathbf{G}^\top(\widehat{\boldsymbol{\theta}}_n)]^{-1}\mathbf{g}(\widehat{\boldsymbol{\theta}}_n) \\ &= \mathbf{Z}_n^\top \mathbf{Z}_n \xrightarrow{\text{D}} \chi^2(q). \end{aligned}$$

## Test basados en la verosimilitud

*Observación:*

Para hipótesis de la forma  $H_0 : \theta_1 = \theta_1^0$ , donde  $\theta = (\theta_1^\top, \theta_2^\top)^\top$ . En este caso el estadístico de Wald asume la forma:

$$W_n = n(\hat{\theta}_{1n} - \theta_1)^\top \mathbf{K}_{11}^{-1}(\hat{\theta}_n)(\hat{\theta}_{1n} - \theta_1),$$

con

$$\mathbf{K}(\theta) = \begin{pmatrix} \mathbf{K}_{11}(\theta) & \mathbf{K}_{12}(\theta) \\ \mathbf{K}_{21}(\theta) & \mathbf{K}_{22}(\theta) \end{pmatrix} = \mathcal{F}^{-1}(\theta).$$

Note que

$$\mathbf{K}_{11}(\theta) = (\mathcal{F}_{11}(\theta) - \mathcal{F}_{12}(\theta)\mathcal{F}_{22}^{-1}(\theta)\mathcal{F}_{21}(\theta))^{-1},$$

con

$$\mathcal{F}(\theta) = \begin{pmatrix} \mathcal{F}_{11}(\theta) & \mathcal{F}_{12}(\theta) \\ \mathcal{F}_{21}(\theta) & \mathcal{F}_{22}(\theta) \end{pmatrix}.$$

## Test basados en la verosimilitud

Sea  $\tilde{\theta}_n$  el MLE de  $\theta$  sujeto a la restricción  $g(\theta) = \mathbf{0}$ . La función Langrangiana asociada con el problema restringido es

$$\ell_n(\theta) - \mathbf{g}^\top(\theta)\lambda,$$

y las condiciones de primer orden son

$$\begin{aligned}\frac{\partial \ell_n(\tilde{\theta}_n)}{\partial \theta} - \mathbf{G}^\top(\tilde{\theta}_n)\tilde{\lambda}_n &= \mathbf{0}, \\ \mathbf{g}(\tilde{\theta}_n) &= \mathbf{0},\end{aligned}$$

donde  $\tilde{\lambda}_n$  es un vector de multiplicadores de Lagrange.

## Test basados en la verosimilitud

### Resultado 2 (test score o de multiplicadores de Lagrange):<sup>3</sup>

El estadístico score o de multiplicadores de Lagrange para probar la hipótesis  $H_0 : \mathbf{g}(\boldsymbol{\theta}_0) = \mathbf{0}$ , es dado por

$$\begin{aligned} R_n &= \frac{1}{n} \left( \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \right)^\top \mathcal{F}^{-1}(\tilde{\boldsymbol{\theta}}_n) \left( \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \right) \\ &= \frac{1}{n} \tilde{\boldsymbol{\lambda}}_n^\top \mathbf{G}(\tilde{\boldsymbol{\theta}}_n) \mathcal{F}^{-1}(\tilde{\boldsymbol{\theta}}_n) \mathbf{G}^\top(\tilde{\boldsymbol{\theta}}_n) \tilde{\boldsymbol{\lambda}}_n, \end{aligned}$$

y bajo  $H_0$  tiene distribución asintótica chi-cuadrado con  $q$  grados de libertad.

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<sup>3</sup>Rao (1948). Proceedings of the Cambridge Philosophical Society 44, 50-57.

## Test basados en la verosimilitud

*Demostración:*

Considere una expansión de Taylor en torno de  $\theta_0$ ,<sup>4</sup> como

$$\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \stackrel{\text{a}}{=} \mathbf{g}(\boldsymbol{\theta}_0) + \frac{\partial \mathbf{g}(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}^\top} (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0),$$

y análogamente para  $\mathbf{g}(\tilde{\boldsymbol{\theta}}_n)$ . De ahí que

$$\begin{aligned}\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) &\stackrel{\text{a}}{=} \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0), \\ \sqrt{n}\mathbf{g}(\tilde{\boldsymbol{\theta}}_n) &\stackrel{\text{a}}{=} \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0),\end{aligned}$$

tomando diferencias, obtenemos

$$\begin{aligned}\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) - \sqrt{n}\mathbf{g}(\tilde{\boldsymbol{\theta}}_n) &\stackrel{\text{a}}{=} \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) - \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &= \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n),\end{aligned}$$

como  $\tilde{\boldsymbol{\theta}}_n$  es el MLE restringido, tenemos  $\mathbf{g}(\tilde{\boldsymbol{\theta}}_n) = \mathbf{0}$ . De este modo,

$$\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \stackrel{\text{a}}{=} \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n). \tag{R.1}$$

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<sup>4</sup>  $\mathbf{X} \stackrel{\text{a}}{=} \mathbf{Y}$  indica que  $\mathbf{X} - \mathbf{Y} = o_{\mathbb{P}}(1)$

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Por otro lado,

$$\frac{1}{\sqrt{n}} \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \stackrel{a}{=} \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} - \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n} (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0),$$

y

$$\mathbf{0} = \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\hat{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \stackrel{a}{=} \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} - \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n} (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0). \quad (\text{R.2})$$

Tomando diferencias, obtenemos

$$\begin{aligned} \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} &\stackrel{a}{=} \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} - \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} \\ &\quad - \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n} (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) + \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n} (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &\stackrel{a}{=} \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n} (\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n). \end{aligned}$$

Por tanto,

$$\sqrt{n} (\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) \stackrel{a}{=} \mathcal{F}^{-1}(\boldsymbol{\theta}_0) \frac{1}{\sqrt{n}} \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}}. \quad (\text{R.3})$$

## Test basados en la verosimilitud

Desde Ecuación (R.1), obtenemos

$$\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \stackrel{a}{=} \mathbf{G}(\boldsymbol{\theta}_0)\sqrt{n}(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) \stackrel{a}{=} \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\frac{1}{\sqrt{n}}\frac{\partial\ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial\boldsymbol{\theta}}.$$

Por la condición de primer orden, podemos escribir

$$\frac{\partial\ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial\boldsymbol{\theta}} = \mathbf{G}^\top(\tilde{\boldsymbol{\theta}}_n)\tilde{\boldsymbol{\lambda}}_n. \quad (\text{R.4})$$

De este modo,

$$\begin{aligned}\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) &\stackrel{a}{=} \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\tilde{\boldsymbol{\theta}}_n)\frac{\tilde{\boldsymbol{\lambda}}_n}{\sqrt{n}} \\ &\stackrel{a}{=} \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)\frac{\tilde{\boldsymbol{\lambda}}_n}{\sqrt{n}}.\end{aligned}$$

Es decir,

$$\frac{\tilde{\boldsymbol{\lambda}}_n}{\sqrt{n}} \stackrel{a}{=} [\mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)]^{-1}\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n).$$

## Test basados en la verosimilitud

Por (W.1), sigue que

$$\frac{\tilde{\lambda}_n}{\sqrt{n}} \xrightarrow{D} N_q(\mathbf{0}, [\mathbf{G}(\boldsymbol{\theta}_0) \mathcal{F}^{-1}(\boldsymbol{\theta}_0) \mathbf{G}^\top(\boldsymbol{\theta}_0)]^{-1}).$$

De ahí que la forma cuadrática

$$R_n = \frac{\tilde{\lambda}_n^\top}{\sqrt{n}} \mathbf{G}(\tilde{\boldsymbol{\theta}}_n) \mathcal{F}^{-1}(\tilde{\boldsymbol{\theta}}_n) \mathbf{G}^\top(\tilde{\boldsymbol{\theta}}_n) \frac{\tilde{\lambda}_n}{\sqrt{n}} \xrightarrow{D} \chi^2(q).$$

Nuevamente por la condición en (R.4), podemos escribir:

$$R_n = \frac{1}{n} \left( \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \right)^\top \mathcal{F}^{-1}(\tilde{\boldsymbol{\theta}}_n) \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}}.$$

## Test basados en la verosimilitud

### Resultado 3 (test de razón de verosimilitudes):<sup>5</sup>

El **test de razón de verosimilitudes** es definido por el estadístico

$$LR_n = 2(\ell_n(\hat{\theta}_n) - \ell_n(\tilde{\theta}_n)),$$

y bajo  $H_0$  tiene región crítica asintótica de tamaño  $\alpha$ , dada por

$$\{LR_n \geq \chi^2_{1-\alpha}(q)\}.$$

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<sup>5</sup>Wilks (1938). The Annals of Mathematical Statistics **9**, 60-62.

## Test basados en la verosimilitud

*Demostración:*

Considere las expansiones de Taylor de  $\ell_n(\hat{\boldsymbol{\theta}}_n)$  y  $\ell_n(\tilde{\boldsymbol{\theta}}_n)$  en torno de  $\boldsymbol{\theta}_0$ . Bajo  $H_0 : \mathbf{g}(\boldsymbol{\theta}_0) = \mathbf{0}$ , estas expresiones son:

$$\ell_n(\hat{\boldsymbol{\theta}}_n) \stackrel{a}{=} \ell_n(\boldsymbol{\theta}_0) + \left( \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} \right)^\top (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) - \frac{n}{2} (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0) (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)$$

$$\ell_n(\tilde{\boldsymbol{\theta}}_n) \stackrel{a}{=} \ell_n(\boldsymbol{\theta}_0) + \left( \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} \right)^\top (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) - \frac{n}{2} (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0) (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0).$$

Tomando diferencias, obtenemos

$$\begin{aligned} \ell_n(\hat{\boldsymbol{\theta}}_n) - \ell_n(\tilde{\boldsymbol{\theta}}_n) &\stackrel{a}{=} \ell_n(\boldsymbol{\theta}_0) - \ell_n(\boldsymbol{\theta}_0) + \left( \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} \right)^\top (\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) \\ &\quad - \frac{n}{2} (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0) (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) + \frac{n}{2} (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0) (\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \end{aligned}$$

## Test basados en la verosimilitud

Es decir,

$$\begin{aligned} LR_n &= 2(\ell_n(\hat{\boldsymbol{\theta}}_n) - \ell_n(\tilde{\boldsymbol{\theta}}_n)) \\ &\stackrel{a}{=} 2\left(\frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}}\right)^\top (\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) - n(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &\quad + n(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \end{aligned} \tag{L.1}$$

Usando (R.3), tenemos

$$\frac{1}{\sqrt{n}} \frac{\partial \ell_n(\boldsymbol{\theta}_0)}{\partial \boldsymbol{\theta}} \stackrel{a}{=} \mathcal{F}(\boldsymbol{\theta}_0) \sqrt{n}(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0).$$

Substituyendo en (L.1), sigue que

$$\begin{aligned} LR_n &\stackrel{a}{=} 2n(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) - (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &\quad + n(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\tilde{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &\stackrel{a}{=} 2n(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n) - (\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \\ &\quad + n(\tilde{\boldsymbol{\theta}}_n - \hat{\boldsymbol{\theta}}_n + \hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\tilde{\boldsymbol{\theta}}_n - \hat{\boldsymbol{\theta}}_n + \hat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0). \end{aligned}$$

## Test basados en la verosimilitud

Después de simple álgebra, obtenemos

$$LR_n \stackrel{a}{=} n(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n)^\top \mathcal{F}(\boldsymbol{\theta}_0)(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n)$$

*Observación:*

Por Ecuación (R.3), sigue que

$$LR_n \stackrel{a}{=} \frac{1}{n} \left( \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} \right)^\top \mathcal{F}^{-1}(\boldsymbol{\theta}_0) \frac{\partial \ell_n(\tilde{\boldsymbol{\theta}}_n)}{\partial \boldsymbol{\theta}} = R_n.$$

Es decir,  $LR_n$  es asintóticamente equivalente a  $R_n$ . De ahí que, bajo  $H_0$  tienen la misma distribución asintótica chi-cuadrado.

## Test basados en la verosimilitud

### Definición 1 (test gradiente):<sup>6</sup>

El estadístico gradiente  $T_n$ , para probar la hipótesis nula  $H_0 : \boldsymbol{\theta} = \boldsymbol{\theta}_0$  versus  $H_1 : \boldsymbol{\theta} \neq \boldsymbol{\theta}_0$  es dado por

$$T_n = \mathbf{U}_n^\top(\boldsymbol{\theta}_0)(\widehat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0),$$

y es asintóticamente chi-cuadrado con  $p$  grados de libertad.

### Demostración:

Sigue de notar que

$$\frac{1}{\sqrt{n}} \mathbf{U}_n(\boldsymbol{\theta}_0) \xrightarrow{D} \mathcal{N}_p(\mathbf{0}, \mathcal{F}(\boldsymbol{\theta}_0)), \quad \sqrt{n}(\widehat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \xrightarrow{D} \mathcal{N}_p(\mathbf{0}, \mathcal{F}^{-1}(\boldsymbol{\theta}_0)),$$

de ahí que

$$\left\{ \frac{1}{\sqrt{n}} \mathcal{F}^{-1/2}(\boldsymbol{\theta}_0) \mathbf{U}_n(\boldsymbol{\theta}_0) \right\}^\top \sqrt{n} \mathcal{F}^{1/2}(\boldsymbol{\theta}_0) (\widehat{\boldsymbol{\theta}}_n - \boldsymbol{\theta}_0) \xrightarrow{D} \chi^2(p).$$

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<sup>6</sup>Terrell (2002). Computing Sciences and Statistics **34**, 206-215.

## Test basados en la verosimilitud

### Resultado 4 (test de forma bilineal):<sup>7</sup>

El estadístico de forma bilineal  $BF_n$ , para probar la hipótesis  $H_0 : \mathbf{g}(\boldsymbol{\theta}_0) = \mathbf{0}$ , adopta la forma

$$BF_n = \tilde{\lambda}_n^\top \mathbf{g}(\hat{\boldsymbol{\theta}}_n),$$

y bajo  $H_0$ ,  $BF_n$  tiene una distribución asintótica chi-cuadrado con  $q$  grados de libertad.

*Demostración:*

Sabemos que

$$\frac{\tilde{\lambda}_n}{\sqrt{n}} \xrightarrow{D} N_q(\mathbf{0}, [\mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)]^{-1}),$$

$$\sqrt{n}\mathbf{g}(\hat{\boldsymbol{\theta}}_n) \xrightarrow{D} N_q(\mathbf{0}, \mathbf{G}(\boldsymbol{\theta}_0)\mathcal{F}^{-1}(\boldsymbol{\theta}_0)\mathbf{G}^\top(\boldsymbol{\theta}_0)),$$

y el resultado sigue.

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<sup>7</sup>Crudu y Osorio (2020). Economics Letters 187, 108885.

## Test basados en la verosimilitud

*Observación:*

Podemos escribir el estadístico  $BF_n$  de forma equivalente como

$$BF_n = \mathbf{U}_n^\top (\tilde{\boldsymbol{\lambda}}_n) \mathbf{G}^+ \mathbf{g}(\hat{\boldsymbol{\theta}}_n),$$

donde  $\mathbf{G}^+ = \mathbf{G}^\top (\mathbf{G}\mathbf{G}^\top)^{-1}$  denota la inversa Moore-Penrose de  $\mathbf{G} = \mathbf{G}(\tilde{\boldsymbol{\theta}}_n)$ . Por (R.1), tenemos que

$$BF_n \stackrel{\text{a}}{=} \mathbf{U}_n^\top (\tilde{\boldsymbol{\theta}}_n) \mathbf{G}^+ \mathbf{G}(\hat{\boldsymbol{\theta}}_n - \tilde{\boldsymbol{\theta}}_n).$$

Además,

$$BF_n \stackrel{\text{a}}{=} \frac{1}{n} \tilde{\boldsymbol{\lambda}}_n^\top \mathbf{G}(\tilde{\boldsymbol{\theta}}_n) \mathcal{F}^{-1}(\tilde{\boldsymbol{\theta}}_n) \mathbf{G}^\top (\tilde{\boldsymbol{\theta}}_n) \tilde{\boldsymbol{\lambda}}_n = R_n.$$

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