

STAT461-561: Delta Method

AD

January 2008

- Assume first that $\theta \in \mathbb{R}$ and that you have

$$\theta_n \xrightarrow{P} \theta$$

and

$$\sqrt{n}(\theta_n - \theta) \Rightarrow \mathcal{N}(0, \sigma^2(\theta))$$

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- Then we have

$$\sqrt{n}(g(\theta_n) - g(\theta)) \Rightarrow \mathcal{N}(0, g'(\theta)^2 \sigma^2(\theta))$$

- We use a Taylor's expansion

$$g(\theta_n) = g(\theta) + g'(\bar{\theta}_n)(\theta_n - \theta)$$

where $\bar{\theta}_n$ is on the line between θ and θ_n .

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- So Slutsky again yields

$$\sqrt{n}(g(\theta_n) - g(\theta)) \Rightarrow \mathcal{N}(0, g'(\theta)^2 \sigma^2)$$

Multivariate delta method

- Assume that $\theta \in \mathbb{R}^d$ and that you have

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Multivariate delta method

- Assume that $\theta \in \mathbb{R}^d$ and that you have

$$\theta_n \xrightarrow{P} \theta$$

and

$$\sqrt{n}(\theta_n - \theta) \Rightarrow \mathcal{N}(0, \Sigma(\theta))$$

- Now if we have $g : \mathbb{R}^d \rightarrow \mathbb{R}^m$ then

$$\sqrt{n}(g(\theta_n) - g(\theta)) \Rightarrow \mathcal{N}\left(0, \nabla g(\theta) \Sigma(\theta) (\nabla g(\theta))^T\right)$$

where $\nabla g := \left(\frac{\partial g}{\partial \theta_1}, \dots, \frac{\partial g}{\partial \theta_d}\right)^T$ is a $d \times m$ matrix.

- We use a multivariate Taylor's expansion

$$g(\theta_n) = g(\theta) + \nabla g(\bar{\theta}_n)(\theta_n - \theta)$$

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- Now we have $\nabla g(\bar{\theta}_n) \rightarrow \nabla g(\theta)$ and

$$\begin{aligned} \text{cov}(g(\theta_n) - g(\theta)) &= \text{cov}(\nabla g(\theta)(\theta_n - \theta)) \\ &= \nabla g(\theta) \text{cov}(\theta_n - \theta) (\nabla g(\theta))^T \end{aligned}$$

and the result follows.