

Lecture 15 : Delta Method

$$\bar{X} \sim AN(\mu, \sigma^2/N)$$

CLT: X_n iid \neq , $E X_n = \mu$, $\text{Var } X_n = \sigma^2 < \infty$

$$\text{then } \sqrt{N} \left(\frac{\bar{X} - \mu}{\sigma} \right) \xrightarrow{d} N(0, 1)$$

Theorem: First-Order Delta Method

If Y_N is a seq. of RVs where

$$\sqrt{N} (Y_n - \theta) \xrightarrow{d} N(0, \Psi^2)$$

$$\uparrow \Psi^2(\theta)$$

then if g is a differentiable function and $g'(\theta) \neq 0$ then

$$\sqrt{N} (g(Y_n) - g(\theta)) \xrightarrow{d} N(0, [g'(\theta)]^2 \Psi^2)$$

Another way:

$$Y_N \sim AN(\theta, \Psi^2/N)$$

then

$$g(Y_N) \sim AN(g(\theta), [g'(\theta)]^2 \Psi^2/N)$$

The most obv. example is when $\frac{Y}{N} = \bar{X}_N$

then by CLT: $\bar{X}_N \sim AN(\mu, \sigma^2/N)$

and so by FO Δ -method:

$$\bar{X}_N \sim AN(g(\mu), [g'(\mu)]^2 \sigma^2/N)$$

Ex. CLT: $\sqrt{N}(\bar{X} - \mu) \xrightarrow{d} N(0, \sigma^2)$

if $g(x) = \log(x)$

then $g'(x) = 1/x$

and so (as long as $g'(\mu) \neq 0$)

then

$$\sqrt{N}(g(\bar{X}) - g(\mu)) = \sqrt{N}(\log(\bar{X}) - \log(\mu))$$

→

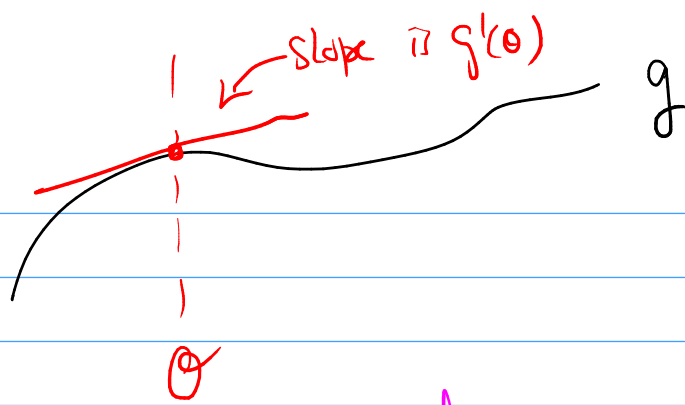
$$\xrightarrow{d} N(0, \frac{1}{\mu^2} \sigma^2)$$

i.e. → $g(\bar{X}) = \log(\bar{X}) \sim AN(\log(\mu), \frac{\sigma^2}{N\mu^2})$

pf of Δ -method

Consider a Taylor Approx. of g about θ

$$g(x) = g(\theta) + g'(\theta)(x - \theta)$$



Idea: $E[cX+d] = cEX + d$
and $\text{Var}(cX+d) = c^2 \text{Var}X$

if near θ $g(x) \approx g(\theta) + g'(\theta)(x-\theta)$

then near θ $E[g(x)] \approx g(\theta) + g'(\theta) \underbrace{[EX - \theta]}_0$
 $= g(\theta)$

and near θ $\text{Var}(g(x)) \approx [g'(\theta)]^2 \text{Var}X$

more formally: $g(x) \approx g(\theta) + g'(\theta)(x-\theta)$

then $g(x) - g(\theta) \approx g'(\theta)(x-\theta)$

so $\sqrt{N}(g(x) - g(\theta)) \approx g'(\theta)[\sqrt{N}(x-\theta)]$

Plug in Y_N for x [recall $\sqrt{N}(Y_N - \theta) \xrightarrow{d} N(0, \psi^2)$]

then

$$\sqrt{N}(g(Y_N) - g(\theta)) \approx g'(\theta) \underbrace{[\sqrt{N}(Y_N - \theta)]}_{\xrightarrow{d} N(0, \psi^2)}$$

$$\xrightarrow{d} N(0, [g'(\theta)]^2 \psi^2).$$

Ex. Let $X_n \stackrel{iid}{\sim} \text{Pois}(\lambda)$ then CLT says
$$\sqrt{N}(\bar{X} - \lambda) \xrightarrow{d} N(0, \lambda)$$

consider $g(x) = 1/x$ then $g'(x) = -1/x^2$
and as long as $\lambda > 0$ $g'(x) \neq 0$

and so $[g'(x)]^2 = (-1/x^2)^2 = 1/x^4$

and the Δ -method says

$$\sqrt{N}(g(\bar{X}) - g(\lambda)) \xrightarrow{d} N(0, [g'(\lambda)]^2 \lambda)$$

i.e.

$$\sqrt{N}\left(\frac{1}{\bar{X}} - \frac{1}{\lambda}\right) \xrightarrow{d} N(0, 1/\lambda^3)$$

Ex. Variance-Stabilizing Transformation

Generically $Y_N \sim AN(\theta, \psi^2(\theta)/N)$

↑ variance depends on θ

Q: is there some trans g so that

$$g(Y_N) \sim AN(g(\theta), \text{doesn't depend on } \theta)$$

Soln: use Δ -method.

$$g(Y_N) \sim AN(g(\theta), [g'(\theta)]^2 \psi^2(\theta)/N)$$

could choose g to make

$$[g'(\theta)]^2 \psi^2(\theta)/N \text{ independent of } \theta$$

So our condition to satisfy is

$$[g'(\theta)]^2 \psi^2(\theta) = c$$

Ex. $X_n \stackrel{\text{iid}}{\sim} \text{Pois}(\lambda)$

$$\psi^2(x) = \lambda$$

CLT: $\bar{X}_N \sim AN(\lambda, \lambda/N)$

GPE: $[g'(x)]^2 \psi^2(x) = c$

$$\Rightarrow [g'(x)]^2 \lambda = c$$

$$\Rightarrow \left(\frac{dg}{dx}\right)^2 \lambda = c$$

$$\Rightarrow \frac{dg}{dx} = \sqrt{c/\lambda}$$

$$dg = \frac{\sqrt{c}}{\sqrt{x}} dx$$

$$\Rightarrow g = \int dg = \int \frac{\sqrt{c}}{\sqrt{x}} dx$$

$$\Rightarrow g \propto \sqrt{x}$$

Claim: $g(x) = \sqrt{x}$ (var. stab. trans for $\text{Pois}(\lambda)$)

$$\bar{X} \sim \text{AN}(\lambda, \lambda/N) \quad g'(x) = \frac{1}{2\sqrt{x}}$$

$$\sqrt{X} \sim \text{AN}(\sqrt{\lambda}, \underbrace{[g'(x)]^2 \lambda/N}_{\left(\frac{1}{2\sqrt{x}}\right)^2 \lambda/N})$$

$$\left(\frac{1}{2\sqrt{x}}\right)^2 \lambda/N = \frac{1}{4N}$$

Theorem: Second Order Δ -method

Assume $\sqrt{N}(\bar{Y}_N - \theta) \xrightarrow{d} N(0, \psi^2(\theta))$

and $g'(\theta) = 0$ but g is twice diff'able

then

$$N(g(\bar{Y}_N) - g(\theta)) \xrightarrow{d} \frac{\psi^2 g''(\theta)}{2} \chi^2(1)$$

$\nwarrow g''(\theta) \neq 0$

Ex. $X_n \stackrel{\text{iid}}{\sim} \text{Bern}(p)$ and let

$$g(t) = t \log(t/p) - (1-t) \log\left(\frac{1-t}{1-p}\right)$$

\uparrow dist. metric b/w $\text{Bern}(p)$ and $\text{Bern}(t)$

KL divergence

Q: What can I say about $g(\bar{X})$?

CLT: $\sqrt{N}(\bar{X} - p) \xrightarrow{d} N(0, \underbrace{p(1-p)}_{\psi^2(p)})$

notice that

$$g'(t) = \log(t/1-t) - \log(p/1-p)$$

and so $g'(p) = 0$

Can't use FO Δ -method. Can use SO Δ -method.

$$g''(t) = \frac{1}{t} + \frac{1}{1-t} = \frac{1}{t(1-t)}$$

$$g''(p) = \frac{1}{p(1-p)} \neq 0 \quad (0 < p < 1)$$

So by the SO Δ -method,

$$N(g(\bar{X}) - g(p)) \xrightarrow{d} \frac{\psi^2(p) g''(p)}{2} \chi^2(1)$$

In our case

$$\begin{aligned} N(g(\bar{X}) - g(p)) &\xrightarrow{d} \frac{p(1-p) \frac{1}{p(1-p)}}{2} \chi^2(1) \\ &= \frac{1}{2} \chi^2(1) \end{aligned}$$

pf. of SO Δ -method

Consider a SO Taylor expansion of $g(x)$

$$g(x) \approx g(\theta) + g'(\theta)(x-\theta) + \frac{g''(\theta)}{2}(x-\theta)^2$$

plug in Y_N into this

$$[\sqrt{N}(Y_N - \theta) \xrightarrow{d} N(0, \psi^2)]$$

$$g(Y_N) \approx g(\theta) + \cancel{g'(\theta)}(Y_N - \theta) + \frac{g''(\theta)}{2}(Y_N - \theta)^2$$

$$g'(\theta) = 0$$

$$\approx g(\theta) + \frac{g''(\theta)}{2}(Y_N - \theta)^2$$

$$\Rightarrow g(Y_N) - g(\theta) \approx \frac{g''(\theta)}{2}(Y_N - \theta)^2$$

$$\Rightarrow N(g(Y_N) - g(\theta)) \approx \frac{g''(\theta)}{2} [\underbrace{\sqrt{N}(Y_N - \theta)}_{\xrightarrow{d} N(0, \psi^2)}]^2$$

$$\xrightarrow{d} N(0, \psi^2) = \psi N(0, 1)$$

$$\xrightarrow{d} \frac{g''(\theta)}{2} [\psi N(0, 1)]^2$$

$$\xrightarrow{d} \frac{\psi^2 g''(\theta)}{2} \chi^2(1)$$
