

- Δ - method

가정

$$\sqrt{n}(X_n - \theta) \xrightarrow{D} N(0, \sigma^2)$$

$g(x)$ is differentiable at θ , $g'(\theta) \neq 0$.

$$\sqrt{n}(g(X_n) - g(\theta)) \xrightarrow{D} N(0, \sigma^2(g'(\theta))^2)$$

proof) Taylor expansion

$g(X_n)$ 을 θ 에서 테일러 전개

$$g(X_n) = g(\theta) + g'(\theta)(X_n - \theta) + \frac{1}{2}g''(\theta)(X_n - \theta)^2 + \cdots$$

remainder $O_p(|X_n - \theta|^3) \Rightarrow 0$

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

$$g(X_n) - g(\theta) = g'(\theta)(X_n - \theta)$$

$$\sqrt{n}(g(X_n) - g(\theta)) = \sqrt{n}(g'(\theta)(X_n - \theta))$$

$$\therefore g'(\theta) \cdot \sqrt{n}(X_n - \theta) \xrightarrow{D} N(0, \sigma^2(g'(\theta))^2)$$

- MGF Technique

목표 : $X_n \xrightarrow{D} X$

If

$$M_{X_n}(t) : \text{mgf of } X_n, \quad M(t) : \text{mgf of } X$$

,

$$\lim_{n \rightarrow \infty} M_{X_n}(t) = M(t), \quad |t| \leq h.$$

\Rightarrow 즉, mgf가 같으면 해당 분포로 분포 수렴.

- 중심극한정리 Central Limit Theorem

: 모든 분포를 “정규분포”로 수렴시킬수 있다. (n 개의 “합”의 분포, $\sum X_i$)

Definition

$$X_1, \dots, X_n \sim (\mu, \sigma^2)$$

$$Y_n = \frac{\sum X_i - n\mu}{\sqrt{n}\sigma} \xrightarrow{D} N(0, 1)$$

$$\text{기초통계} : \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}}$$

proof)

1. mgf technique, Taylor expansion

$$X_1, \dots, X_n \sim (\mu, \sigma^2), E(\sum X_i) = n\mu, \text{Var}(\sum X_i) = n\sigma^2$$

$$Y_n = \frac{\sum X_i - n\mu}{\sqrt{n}\sigma} = \frac{\sqrt{n}(\bar{X}_n - \mu)}{\sigma}$$

$$\text{mgf of } (X - \mu), M(t) = E(e^{tx})$$

$$m(t) = E(e^{t(x-\mu)}) = e^{-\mu t} M(t), \quad |t| \leq h$$

$$m(0) = 1$$

$$m'(0) = E((x - \mu)e^{t(x-\mu)}|_{t=0}) = E(x - \mu) = 0$$

$$m''(0) = E((x - \mu)^2 e^{t(x-\mu)}|_{t=0}) = E((x - \mu)^2) = \sigma^2$$

Taylor expansion

$$\begin{aligned} m(t) &= m(0) + m'(0)t + \frac{m''(\xi)t^2}{2}, \quad 0 < \xi < t \\ &= 1 + \frac{m''(\xi)t^2}{2} \\ &= 1 + \frac{\sigma^2 t^2}{2} + \frac{(m''(\xi) - \sigma^2)t^2}{2} \end{aligned}$$

Consider $M(t; n)$

$$\begin{aligned}
M(t; n) &= E \left[\exp \left(t \frac{\sum X_i - n\mu}{\sigma\sqrt{n}} \right) \right] \\
&= E \left[\exp \left(t \frac{X_1 - \mu}{\sigma\sqrt{n}} \right) \cdot \exp \left(t \frac{X_2 - \mu}{\sigma\sqrt{n}} \right) \cdots \exp \left(t \frac{X_n - \mu}{\sigma\sqrt{n}} \right) \right] \\
&= E \left[\exp \left(t \frac{X_1 - \mu}{\sigma\sqrt{n}} \right) \right] \cdots E \left[\exp \left(t \frac{X_n - \mu}{\sigma\sqrt{n}} \right) \right] \\
&= \left\{ E \left[\exp \left(t \frac{X - \mu}{\sigma\sqrt{n}} \right) \right] \right\}^n \\
&= \left[m \left(\frac{t}{\sigma\sqrt{n}} \right) \right]^n, \quad -h < \frac{t}{\sigma\sqrt{n}} < h \\
&= \left\{ 1 + \frac{t^2}{2n} + \frac{[m''(\xi) - \sigma^2]t^2}{2n\sigma^2} \right\}^n, \quad 0 < \xi < \frac{t}{\sigma\sqrt{n}}, \quad -h\sigma\sqrt{n} < t < h\sigma\sqrt{n} \\
\lim_{n \rightarrow \infty} M(t; n) &= \lim_{n \rightarrow \infty} \left\{ 1 + \frac{t^2}{2n} + \frac{[m''(\xi) - \sigma^2]t^2}{2n\sigma^2} \right\}^n \\
&= \lim_{n \rightarrow \infty} \left[1 + \frac{t^2/2}{n} \right]^n \\
&= e^{\frac{t^2}{2}}, \quad \text{mgf of } N(0, 1)
\end{aligned}$$

5.2.16

$$\lim_{n \rightarrow \infty} \psi(n) = 0,$$

$$\lim_{n \rightarrow \infty} \left[1 + \frac{b}{n} + \frac{\psi(n)}{n} \right]^{cn} = \lim_{n \rightarrow \infty} \left(1 + \frac{b}{n} \right)^{cn} = e^{bc}$$

2. 로피탈의 정리를 이용한 증명

L'Hopital's Theorem

극한값이 부정형 $(0/0, \infty/\infty)$ 일 경우 미분을 이용하여 계산할 수 있도록 하는 방법
함수 $f(x), g(x)$ 가 미분가능하고 $f(a) = 0, g(a) = 0$ 이고, $\lim_{x \rightarrow a} g(x) \neq 0$ 일때,

$$\lim_{x \rightarrow a} \frac{f(x)}{g(x)} = \lim_{x \rightarrow a} \frac{f'(x)}{g'(x)}$$

Law of Large Numbers

(Strong)

$$\bar{X}_n \rightarrow \mu \text{ as } n \rightarrow \infty \text{ with prob 1.}$$

(Weak)

$$\text{For any } C > 0, P(|\bar{X}_n - \mu| > c) \rightarrow 0 \text{ as } n \rightarrow \infty$$

★ $\bar{X}_n - \mu \rightarrow 0$ with prob 1, but what dose the distribution of \bar{X}_n look like?

CLT

$$\sqrt{n} \frac{\bar{X}_n - \mu}{\sigma} \rightarrow N(0, 1) \text{ in distribution.}$$

Equivalently

$$\frac{\sum_{j=1}^n X_j - n\mu}{\sqrt{n}\sigma} \rightarrow N(0, 1) \text{ in distribution.}$$

proof) (assume MGF $M(t)$ of X_j exists)

Can assume $\mu = 0, \sigma = 1$, since consider $\frac{1}{\sqrt{n}} \sum_{j=1}^n \frac{X_j - \mu}{\sigma}$. Let $S_n = \sum_{j=1}^n X_j$, show MGF of $\frac{S_n}{\sqrt{n}}$ goes to $N(0, 1)$ MGF.

$$\begin{aligned} E\left(e^{tS_n/\sqrt{n}}\right) &= E\left(e^{tX_1/\sqrt{n}}\right) \cdots E\left(e^{tX_n/\sqrt{n}}\right) \\ &= \left(M\left(\frac{t}{\sqrt{n}}\right)\right)^n \end{aligned}$$

Take logs

$$\begin{aligned} \lim_{n \rightarrow \infty} n \log M\left(\frac{t}{\sqrt{n}}\right) &= \lim_{n \rightarrow \infty} \frac{\log M\left(\frac{t}{\sqrt{n}}\right)}{1/n}, \quad y = \frac{1}{\sqrt{n}} \\ &= \lim_{y \rightarrow 0} \frac{\log M(yt)}{y^2} \\ &= \lim_{y \rightarrow 0} \frac{tM'(yt)}{2yM(yt)}, \quad \lim_{y \rightarrow 0} M(yt) = 1 \\ &= \frac{t}{2} \lim_{y \rightarrow 0} \frac{M'(yt)}{y} \\ &= \frac{t^2}{2} \lim_{y \rightarrow 0} \frac{M''(yt)}{1} \\ &= \frac{t^2}{2} \end{aligned}$$

Which is the log of $e^{\frac{t^2}{2}}$ ($N(0, 1)$ MGF)

$$M(t) = E(e^{tx})$$

$$M(0) = 1$$

$$M'(0) = E(te^{tx})|_{t=0} = 0$$

$$M''(0) = E(e^{tx} + t^2 e^{tx})|_{t=0} = 1$$