

10.5

$$\begin{aligned}\text{Cov}[F_n(u), F_n(v)] &= \text{Cov}\left[\frac{1}{n} \sum_{i=1}^n I(X_i \leq u), \frac{1}{n} \sum_{j=1}^n I(Y_j \leq v)\right] \\ &= \frac{1}{n^2} \sum_{i,j} \text{Cov}[I(X_i \leq u), I(X_j \leq v)] \\ &\quad \quad \quad (i \neq j \text{ だけ}) \\ &= \frac{1}{n^2} \sum_{i=1}^n \text{Cov}[I(X_i \leq u), I(X_i \leq v)] \\ &= \frac{1}{n} \text{Cov}[I(X_1 \leq u), I(X_1 \leq v)] \\ &= \frac{1}{n} P(X_1 \leq u, X_1 \leq v) - \frac{1}{n} P(X_1 \leq u) P(X_1 \leq v) \\ &= \frac{1}{n} (F(\min\{u, v\}) - F(u)F(v))\end{aligned}$$

10.9.

$$E[\log(1-\hat{F}(t))] - \log(1-F(t)) = \text{bias}$$

利用 δ -method ... $g(x) = \log(1-x)$

$$g(x) \approx g(\mu) + g'(\mu)(x-\mu) + \frac{1}{2!} g''(\mu)(x-\mu)^2 \quad \begin{matrix} (x \rightarrow \hat{F}(t)) \\ (\mu \rightarrow F(t)) \end{matrix}$$
$$\begin{aligned} g(\hat{F}(t)) &\approx g(F(t)) + g'(F(t))(\hat{F}(t) - F(t)) \\ &\quad + \frac{1}{2!} g''(F(t))(\hat{F}(t) - F(t))^2 \end{aligned}$$

$$g(\hat{F}(t)) - g(F(t)) \approx g'(F(t))(\hat{F}(t) - F(t)) + \frac{1}{2} g''(F(t))(\hat{F}(t) - F(t))^2$$

$$\therefore \text{bias} \approx \frac{1}{2} g''(F(t)) V[\hat{F}(t)]$$

$$\left(g'(x) = \frac{-1}{1-x} \quad g''(x) = \frac{-1}{(1-x)^2} \right)$$

$$\text{bias} \approx \frac{1}{2} \frac{-1}{(1-F(t))^2} \cdot \frac{1}{n} \cdot F(t)(1-F(t))$$

$$= \frac{-F(t)}{2n(1-F(t))} \quad (F(t) \rightarrow 1 \text{ 時, bias 會很大})$$

10.11.

$$\frac{d}{dt} f(t) = f(t) = \beta t^{\beta-1} e^{-\alpha t^{\beta}}$$

$$\frac{f(t)}{1-f(t)} = \frac{\beta t^{\beta-1} e^{-\alpha t^{\beta}}}{e^{-\alpha t^{\beta}}} = \beta t^{\beta-1}$$

10.13.

Weibull 分布 $\beta > 1$

$$h(t) = \alpha \beta t^{\beta-1} \quad (\because 10.11) \quad \int (t^{\beta})$$

10. K.

Weibull 分布 $0 < \beta < 1$

$$h(t) = \alpha \beta t^{\beta-1} \quad (\text{=: '10.11}) \quad \rightarrow \quad (t \uparrow)$$

11.13.

$$\textcircled{1} X_1, X_2, \dots, X_n \sim N(\mu, 1) \quad (\mu \neq 0) \quad (\mu = 0.3, n = 25)$$

$$T = \sum_{j=1}^n I(X_j > 0) \sim \text{Bin}(n, \Phi(\mu))$$

$$\left(\because \Pr(X_j > 0) = \Pr(X_j - \mu > -\mu) = 1 - \Phi(-\mu) = \Phi(\mu) \right)$$

$$n \text{ 很大时 } T \approx N(n\Phi(\mu), n\Phi(\mu)(1-\Phi(\mu)))$$

$$\frac{T - n\Phi(\mu)}{\sqrt{n\Phi(\mu)(1-\Phi(\mu))}} \sim N(0, 1)$$

$$\text{棄卻域 } C = \{T \mid T > c\}$$

$$\mu = 0 \text{ 時, } \Pr(T \in C \mid \mu = 0) = 0.05$$

$$\therefore \frac{T - n\Phi(0)}{\sqrt{n\Phi(0)(1-\Phi(0))}} = \frac{T - \frac{n}{2}}{\sqrt{\frac{n}{4}}} \geq 1.645$$

$$\therefore T \geq \underbrace{\frac{n}{2} + \frac{\sqrt{n}}{2} \cdot 1.645}_c$$

z

||

$$\beta(\mu) = \Pr(T > c | \mu) = \Pr\left(\frac{T - n\Phi(\mu)}{\sqrt{n\Phi(\mu)(1-\Phi(\mu))}} > \frac{c - n\Phi(\mu)}{\sqrt{n\Phi(\mu)(1-\Phi(\mu))}} \mid \mu\right)$$

$$\Pr\left(z > \frac{\frac{n}{2} + \frac{\sqrt{n}}{2} \cdot 1.645 - n\Phi(\mu)}{\sqrt{n\Phi(\mu)(1-\Phi(\mu))}} \mid z \sim N(0,1)\right)$$

$$= 1 - \Phi\left(\frac{\frac{n}{2} + \frac{\sqrt{n}}{2} \cdot 1.645 - n\Phi(\mu)}{\sqrt{n\Phi(\mu)(1-\Phi(\mu))}}\right)$$

$$\mu = 0.3 \quad n = 25$$

$$\beta(0.3) = 1 - \Phi\left(\frac{12.5 + 25 \cdot 1.645 - 25 \cdot \Phi(0.3)}{\sqrt{25 \Phi(0.3)(1-\Phi(0.3))}}\right)$$

$$\Phi(0.3) = 0.6179$$

$$= 1 - \Phi(0.4795)$$

$$= 1 - 0.6844 = \underline{\underline{0.3156}} \quad (\text{sign-test})$$

② Normal Theory...

$$\sqrt{n}(\bar{X} - \mu) \sim N(0, 1)$$

$$\mu = 0 \text{ 時} \dots \sqrt{n} \bar{X} \geq 1.645$$

$$\therefore \bar{X} \geq \frac{1.645}{\sqrt{n}} \Rightarrow \text{棄卻}$$

$$\beta(\mu) = P_r\left(\bar{X} \geq \frac{1.645}{\sqrt{n}} \mid \mu\right)$$

$$= P_r\left(\sqrt{n}(\bar{X} - \mu) \geq 1.645 - \sqrt{n}\mu \mid \mu\right)$$

$$= 1 - \Phi(1.645 - \sqrt{n}\mu)$$

$$n = 25, \mu = 0.3$$

$$\beta(0.3) = 1 - \Phi(1.645 - 1.5)$$

$$= 1 - \Phi(0.145) \doteq 0.44$$

(檢力較大)

11.15.

某種分布

$$X_1, X_2, \dots, X_n \sim D_x(\mu_x, \sigma^2) \quad (\sigma^2=100)$$

$$Y_1, Y_2, \dots, Y_n \sim D_y(\mu_y, \sigma^2)$$

根據中央極限定理, $\frac{\sqrt{n}(\bar{X}-\mu_x)}{\sigma} \xrightarrow{d} N(0,1)$

$$\frac{\sqrt{n}(\bar{Y}-\mu_y)}{\sigma} \xrightarrow{d} N(0,1)$$

$$\therefore \frac{\sqrt{n}(\bar{X}-\bar{Y}-\mu_x+\mu_y)}{\sqrt{2} \cdot \sigma} \sim N(0,1) \quad (n \text{ 很大})$$

$$\therefore -1.96 \leq \frac{\sqrt{n}(\bar{X}-\bar{Y}-\mu_x+\mu_y)}{10\sqrt{2}} \leq 1.96$$

$$\frac{-1.96 \cdot 10\sqrt{2}}{\sqrt{n}} \leq (\bar{X}-\bar{Y}) - (\mu_x - \mu_y) \leq \frac{1.96 \cdot 10\sqrt{2}}{\sqrt{n}}$$

$$\bar{X}-\bar{Y} \pm \frac{1.96 \cdot 10\sqrt{2}}{\sqrt{n}} \quad (\text{信賴區間})$$

$$\text{寬度} = \frac{392 \cdot 10\sqrt{2}}{\sqrt{n}} \leq 2$$

$$\therefore n = (1.96 \cdot 10\sqrt{2})^2$$

$$\therefore n \geq 769$$

11.19.

$$X_1 \sim \dots \sim X_{25} \sim N(\mu_X, \sigma^2)$$

$$Y_1 \sim \dots \sim Y_{25} \sim N(\mu_Y, \sigma^2)$$

$$(1) \bar{Y} - \bar{X} \sim N(\mu_Y - \mu_X, \frac{2\sigma^2}{25}) = N(\mu_Y - \mu_X, 2)$$

$$SE(\bar{Y} - \bar{X}) = \sqrt{2}$$

$$(2) \mu = \mu_Y - \mu_X \quad H_0: \mu = 0 \quad \text{vs} \quad H_1: \mu > 0$$

↓ Neyman-Pearson's lemma

$$C: \{ (X_1, \dots, X_{25}, Y_1, \dots, Y_{25}) \mid \bar{Y} - \bar{X} > c \}$$

$$\mu = 0 \quad \Pr(\bar{Y} - \bar{X} > c \mid \mu = 0)$$

$$= \Pr\left(\frac{\bar{Y} - \bar{X}}{\sqrt{2}} > \frac{c}{\sqrt{2}} \mid \mu = 0\right) = 1 - \Phi\left(\frac{c}{\sqrt{2}}\right) = 0.05$$

$$c = \sqrt{2} \cdot \Phi^{-1}(0.95) = \sqrt{2} \cdot 1.645$$

$$\therefore C = \{ (X_1, \dots, X_{25}, Y_1, \dots, Y_{25}) \mid \bar{Y} - \bar{X} > \sqrt{2} \cdot 1.645 \}$$

(3) $\mu_T - \mu_X = 1$ 時

$$\begin{aligned}
 & P_r(\bar{Y} - \bar{X} > \sqrt{2} \cdot 1.645 \mid \mu=1) \\
 &= P_r\left(\frac{\bar{Y} - \bar{X} - 1}{\sqrt{2}} > \frac{\sqrt{2} \cdot 1.645 - 1}{\sqrt{2}} \mid \mu=1\right) \\
 &= 1 - \Phi\left(1.645 - \frac{1}{\sqrt{2}}\right) = 1 - \Phi(0.938) = 1 - 0.8264 \\
 &= 0.174
 \end{aligned}$$

(4) YES

(5) $C = \{(X_1, \dots, X_{25}, Y_1, \dots, Y_{25}) \mid |\bar{X} - \bar{Y}| \geq \sqrt{2} \cdot 1.96\}$

$$\begin{aligned}
 & P_r(\bar{X} - \bar{Y} \geq \sqrt{2} \cdot 1.96 \mid \mu=1) \\
 &+ P_r(\bar{X} - \bar{Y} \leq -\sqrt{2} \cdot 1.96 \mid \mu=1) \\
 &= P_r\left(\frac{\bar{X} - \bar{Y} - 1}{\sqrt{2}} \geq 1.96 - \frac{1}{\sqrt{2}} \mid \mu=1\right) \\
 &+ P_r\left(\frac{\bar{X} - \bar{Y} - 1}{\sqrt{2}} \leq -1.96 - \frac{1}{\sqrt{2}} \mid \mu=1\right) \\
 &= 1 - \Phi\left(1.96 - \frac{1}{\sqrt{2}}\right) + \Phi\left(-1.96 - \frac{1}{\sqrt{2}}\right) \\
 &= 1 - \Phi(1.25) + \Phi(-2.67) \\
 &= 1 - (0.8944) + (0.0038) \approx 0.11.
 \end{aligned}$$

11.26

$$\begin{cases} X_1, X_2, \dots, X_n \sim N(0,1) \\ Y_1, Y_2, \dots, Y_n \sim N(1,1) \end{cases}$$

$$(a) T = \frac{n(n+1)}{2} + \sum_{i=1}^n \sum_{j=1}^n I(X_i > Y_j) \quad (\text{rank sum})$$

$$\begin{aligned} E[T] &= \frac{n(n+1)}{2} + E\left[\sum_{i=1}^n \sum_{j=1}^n I(X_i > Y_j)\right] \\ &= \frac{1}{2}n(n+1) + n^2 \cdot P(X_1 > Y_1) \end{aligned}$$

$$\begin{aligned} P(X_1 > Y_1) &= P(Y_1 - X_1 < 0) \quad Y_1 - X_1 \sim N(1, 2) \\ &= P\left(\frac{Y_1 - X_1 - 1}{\sqrt{2}} \leq \frac{-1}{\sqrt{2}}\right) \\ &= \Phi\left(\frac{-1}{\sqrt{2}}\right) = 1 - \Phi\left(\frac{1}{\sqrt{2}}\right) \end{aligned}$$

$$\therefore E[T] = \frac{n(n+1)}{2} + n^2 \cdot \left(1 - \Phi\left(\frac{1}{\sqrt{2}}\right)\right)$$

$$(b) V\left[\frac{n(n+1)}{2} + \sum_{i=1}^n \sum_{j=1}^n I(X_i > Y_j)\right]$$

$$= V\left[\underbrace{\sum_{i=1}^n \sum_{j=1}^n I(X_i > Y_j)}_{U_i}\right]$$

$$= V\left[\sum_{i=1}^n U_i\right] = nV[U] + n(n-1)\text{cov}[U_1, U_2]$$

$$(1) V[U_i] = V\left[\sum_{j=1}^n I(X_i > Y_j)\right]$$

$$= nV[I(X_i > Y_1)] + n(n-1)\text{cov}[I(X_i > Y_1), I(X_i > Y_2)]$$

bernoulli: $(1 - \Phi(\frac{1}{\sqrt{2}}))$

$$\therefore (1 - \Phi(\frac{1}{\sqrt{2}})) \cdot \Phi(\frac{1}{\sqrt{2}})$$

$$(2) \text{cov}[U_1, U_2] = \sum_{i=1}^n \sum_{j=1}^n \text{cov}[I(X_i > Y_i), I(X_2 > Y_j)]$$

($i \neq j \Rightarrow$ 独立)

$$= \sum_{j=1}^n \text{cov}[I(X_1 > Y_j), I(X_2 > Y_j)]$$

$$= n \text{cov}[I(X_1 > Y_1), I(X_2 > Y_1)]$$

$$\therefore (1) + (2) \dots n^2 V[I(X_1 > Y_1)] + n^2(n-1) \text{cov}[I(X_1 > Y_1), I(X_1 > Y_2)]$$

$$+ n^2(n-1) \text{cov}[I(X_1 > Y_1), I(X_2 > Y_1)]$$

$$\textcircled{\pm} \text{cov}[I(X_1 > Y_1), I(X_1 > Y_2)]$$

$$= \underbrace{P(X_1 > Y_1, Y_2)} - \underbrace{P(X_1 > Y_1)} \underbrace{P(X_1 > Y_2)}$$

↓
計算有點困難

$$1 - \Phi\left(\frac{1}{\sqrt{2}}\right) \quad 1 - \Phi\left(\frac{1}{\sqrt{2}}\right)$$

$$\text{cov}[I(X_1 > Y_1), I(X_2 > Y_1)]$$

$$= \underbrace{P(X_1, X_2 > Y_1)} - \underbrace{P(X_1 > Y_1)} \underbrace{P(X_2 > Y_1)}$$

↓
計算有點困難

$$(1 - \Phi\left(\frac{1}{\sqrt{2}}\right)) \cdot (1 - \Phi\left(\frac{1}{\sqrt{2}}\right))$$

11.31.

$$\hat{\pi} = \frac{1}{mn} \sum_{i=1}^n \sum_{j=1}^m Z_{ij}$$

$$E[\hat{\pi}] = E[Z_{11}] = P(X_1 < Y_1) = \frac{1}{2}$$

$$V[\hat{\pi}] = \frac{1}{(mn)^2} V\left[\sum_{i=1}^n \sum_{j=1}^m Z_{ij}\right]$$

$$W_i \stackrel{\text{def}}{=} \sum_{j=1}^m Z_{ij}$$

$$= \frac{1}{(mn)^2} V\left[\sum_{i=1}^n W_i\right]$$

$$= \left(\frac{1}{mn}\right)^2 \left(nV[W] + n(n-1) \text{cov}[W_1, W_2] \right)$$

$$\textcircled{1} V[W] = V\left[\sum_{j=1}^m Z_{1j}\right]$$

$$= mV[Z_{11}] + m(m-1) \text{cov}[Z_{11}, Z_{12}]$$

$$= \frac{m}{4} + m(m-1) \left\{ \underbrace{E[Z_{11}Z_{12}]}_{\frac{1}{3}} - \underbrace{E[Z_{11}]}_{\frac{1}{2}} \underbrace{E[Z_{12}]}_{\frac{1}{2}} \right\}$$

$$Z_{11} \sim \text{bernoulli}\left(\frac{1}{2}\right)$$

$$P(X_1 < Y_1, X_1 < Y_2)$$

$$= \frac{1}{3} \quad (\because \text{排列} = 3!)$$

$$X_1 < Y_1 < Y_2$$

$$X_1 < Y_2 < Y_1$$

$$= \frac{m}{4} + \frac{m(m-1)}{12}$$

$$\therefore \frac{2}{3!} = \frac{1}{3}$$

$$\text{COV}[W_1, W_2] = \sum_{k=1}^m \sum_{l=1}^m \text{COV}[Z_{1,k}, Z_{2,l}]$$

($k \neq l$, ~~not~~)

$$= \sum_{k=1}^m \text{COV}[Z_{1,k}, Z_{2,k}] = m \text{COV}[Z_{1,1}, Z_{2,1}]$$

$$= m \left(\underbrace{\text{Pr}(X_1 < Y_1, X_2 < Y_1)}_{\frac{1}{3}} - \underbrace{\text{Pr}(X_1 < Y_1)}_{\frac{1}{2}} \underbrace{\text{Pr}(X_2 < Y_1)}_{\frac{1}{2}} \right)$$

$$= \frac{m}{12}$$

$$\therefore V[\hat{\rho}] = \left(\frac{1}{mn} \right)^2 \left(n \cdot \left(\frac{m}{4} + \frac{m(m+1)}{12} \right) + \frac{m \cdot n \cdot (n+1)}{12} \right)$$

$$= \frac{1}{(mn)^2} \left(\frac{mn}{4} + \frac{mn(m+1)}{12} + \frac{mn(n+1)}{12} \right)$$

$$= \frac{1}{(mn)^2} \left(\frac{mn}{12} (m+n+2) + \frac{mn}{4} \right)$$

$$= \frac{1}{(mn)^2} \left(\frac{mn}{12} (m+n+1) \right)$$

$$= \frac{1}{12mn} (m+n+1)$$

11.33.

(1) one-sided test

• $H_0: G_X = G_Y$ vs $H_1: G_X > G_Y$ ~~AND CASE...~~

$$C = \left\{ (X_1, \dots, X_n, Y_1, \dots, Y_m) \mid \frac{S_X^2}{S_Y^2} \geq \chi^2_{(1-\alpha)} \right\}$$

χ^2 F-d, m, 1 ~~AND~~ cdf

• $H_0: G_X = G_Y$ vs $H_1: G_X < G_Y$

$$C = \left\{ (X_1, \dots, X_n, Y_1, \dots, Y_m) \mid \frac{S_X^2}{S_Y^2} \leq \chi^2_{(\alpha)} \right\}$$

two-sided test

• $H_0: G_X = G_Y$ vs $H_1: G_X \neq G_Y$

$$C = \left\{ (X_1, \dots, X_n, Y_1, \dots, Y_m) \mid \frac{S_X^2}{S_Y^2} \geq \chi^2_{(1-\frac{\alpha}{2})} \text{ or } \frac{S_X^2}{S_Y^2} \leq \chi^2_{(\frac{\alpha}{2})} \right\}$$

$$(2) \quad \frac{\sigma^2}{\sigma_X^2}, \frac{S_X^2}{S^2} \sim F_{n-1, m-1}$$

$$\therefore P\left(\chi^2\left(\frac{\alpha}{2}\right) \leq \frac{\sigma^2}{\sigma_X^2} \cdot \frac{S_X^2}{S^2} \leq \chi^2\left(1-\frac{\alpha}{2}\right)\right) = 1-\alpha$$

$$\therefore P\left(\frac{S_X^2}{S^2} \cdot \frac{1}{\chi^2\left(1-\frac{\alpha}{2}\right)} \leq \frac{\sigma^2}{\sigma_X^2} \leq \frac{S_X^2}{S^2} \cdot \frac{1}{\chi^2\left(\frac{\alpha}{2}\right)}\right)$$

$$\therefore \left[\frac{S_X^2}{S^2} \cdot \frac{1}{\chi^2\left(1-\frac{\alpha}{2}\right)}, \frac{S_X^2}{S^2} \cdot \frac{1}{\chi^2\left(\frac{\alpha}{2}\right)} \right]$$

$$(3) \quad 0.13 \sim 2.16$$

11.32

$$X - Y \sim N(\mu_X - \mu_Y, \sigma_X^2 + \sigma_Y^2)$$

$$P(X - Y < 0) = P\left(\frac{(X - Y) - (\mu_X - \mu_Y)}{\sqrt{\sigma_X^2 + \sigma_Y^2}} < \frac{-(\mu_X - \mu_Y)}{\sqrt{\sigma_X^2 + \sigma_Y^2}}\right)$$

$$= P\left(Z < \frac{\mu_Y - \mu_X}{\sqrt{\sigma_X^2 + \sigma_Y^2}}\right) = \Phi\left(\frac{\mu_Y - \mu_X}{\sqrt{\sigma_X^2 + \sigma_Y^2}}\right)$$

No.

Date