

1.

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$$

$$(a) \quad \sum_{i=1}^n y_i = a \left(\sum_{i=1}^n x_i \right) + nb$$

$$\bar{y} = a\bar{x} + b$$

(b)

$$S_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

$$S_y^2 = \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

$$\frac{S_x^2}{S_y^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\sum_{i=1}^n (y_i - \bar{y})^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\sum_{i=1}^n (ax_i + b - a\bar{x} - b)^2} = \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\sum_{i=1}^n a^2 (x_i - \bar{x})^2} = \frac{1}{a^2}$$

$$= \frac{\sum_{i=1}^n x_i^2 - n\bar{x}^2}{a^2 \left(\sum_{i=1}^n x_i^2 - n\bar{x}^2 \right)} = \frac{1}{a^2} = \frac{S_x^2}{S_y^2}$$

$$\bar{y}^2 = a^2 \bar{x}^2 + 2ab\bar{x} + b^2$$

$$y_i^2 = a^2 x_i^2 + 2abx_i + b^2$$

$$a^2 \left(\sum_{i=1}^n x_i^2 \right) + 2ab \sum_{i=1}^n x_i + nb^2$$

$$= \frac{a^2 \sum_{i=1}^n (x_i - \bar{x})^2}{a^2} = \frac{1}{a^2}$$

$$a^2 \left(\sum_{i=1}^n x_i^2 \right) + 2abn\bar{x} + nb^2 - a^2 n\bar{x}^2 - 2abn\bar{x} - b^2 n$$

2.

$$\text{Cov}(\bar{x}, S^2) = E((\bar{x} - E(\bar{x})) (S^2 - E(S^2)))$$

$$E(\bar{x}) = E\left(\frac{x_1 + x_2 + \dots + x_n}{n}\right) = E(x_1) = E(X) \quad \text{因为 } x_1 \text{ 与 } x_2 \text{ 同分布}$$

$$E(S^2) = E\left(\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2\right) = \frac{1}{n-1} E\left(\sum_{i=1}^n x_i^2 - 2\bar{x} \sum_{i=1}^n x_i + n\bar{x}^2\right)$$

$$= \frac{1}{n-1} E\left(\sum_{i=1}^n x_i^2 - n\bar{x}^2\right) = \frac{1}{n-1} (n \text{Var}(X) + nE(X)^2 - \text{Var}(X) - nE(X)^2) = \text{Var}(X)$$

$$E(X^2) = \text{Var}(X) + (E(X))^2$$

$$E(n\bar{x}^2) = n E(\bar{x}^2) = n (\text{Var}(\bar{x}) + E(\bar{x})^2) = n \left(\frac{\text{Var}(X)}{n} + E(X)^2 \right)$$

$$\therefore \text{Cov}(\bar{x}, S^2) = E((\bar{x} - E(X)) (S^2 - \text{Var}(X)))$$

$$= E(\bar{x} S^2) - \text{Var}(X) E(\bar{x}) - E(X) E(S^2) + E(X) \text{Var}(X)$$

$$= E(\bar{x} S^2) - E(S^2) E(\bar{x}) \quad \text{设 } E(\bar{x}) = E(X) = E(X) = \mu$$

$$= E((\bar{x} - \mu) S^2)$$



$$(\bar{x} - \mu)^2 = (\bar{x} - \mu) \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2 = \frac{\bar{x} - \mu}{n-1} \sum_{i=1}^n (x_i^2 - 2x_i\bar{x} + \bar{x}^2)$$

$$= \frac{\bar{x} - \mu}{n-1} \left(\sum_{i=1}^n x_i^2 - 2n\bar{x} + n\bar{x}^2 \right) = \frac{\bar{x} - \mu}{n-1} \left(\sum_{i=1}^n x_i^2 - n\bar{x}^2 \right)$$

$$= \frac{\bar{x} - \mu}{n-1} \left(\sum_{i=1}^n x_i^2 - 2\mu \sum_{i=1}^n x_i + n\mu^2 - n\bar{x}^2 + 2\mu \sum_{i=1}^n x_i - n\mu^2 \right)$$

$$= \frac{\bar{x} - \mu}{n-1} \left[\sum_{i=1}^n (x_i - \mu)^2 - (n\bar{x}^2 - 2n\mu\bar{x} + \mu^2 n) \right]$$

$$= \frac{\bar{x} - \mu}{n-1} \left[\sum_{i=1}^n (x_i - \mu)^2 - n(\bar{x} - \mu)^2 \right]$$

$$E[(\bar{x} - \mu)^2] = E \left[\frac{\bar{x} - \mu}{n-1} \sum_{i=1}^n (x_i - \mu)^2 - \frac{n(\bar{x} - \mu)}{n-1} (\bar{x} - \mu)^2 \right]$$

$$= \frac{1}{n-1} E \left[(\bar{x} - \mu) \sum_{i=1}^n (x_i - \mu)^2 - n(\bar{x} - \mu)^3 \right]$$

$$= \frac{1}{n-1} \left[\underbrace{E[(\bar{x} - \mu) \sum_{i=1}^n (x_i - \mu)^2]}_{\text{①}} - \underbrace{E[n(\bar{x} - \mu)^3]}_{\text{②}} \right]$$

$$\text{②} : E[n(\bar{x} - \mu)^3] = n E[(\bar{x} - \mu)^3]$$

$$E[(\bar{x} - \mu)^3] = E \left[\left(\frac{\sum_{i=1}^n x_i}{n} - \mu \right)^3 \right] = \frac{1}{n^3} E \left[\left(\sum_{i=1}^n x_i - n\mu \right)^3 \right]$$

$$= \frac{1}{n^3} E \left[(x_1 - \mu + x_2 - \mu + x_3 - \mu + \dots + x_n - \mu)^3 \right]$$

$$\text{assume 只有两项} \quad E(x_1 - \mu + x_2 - \mu)^3 = E[(x_1 - \mu)^3 + 3(x_1 - \mu)^2(x_2 - \mu) + 3(x_1 - \mu)(x_2 - \mu)^2 + (x_2 - \mu)^3]$$

$$= E(x_1 - \mu)^3 + 3E \underbrace{(x_1 - \mu)^2(x_2 - \mu)}_{x_1 - \mu \text{ 与 } x_2 - \mu \text{ 独立}} + 3E \underbrace{(x_1 - \mu)(x_2 - \mu)^2}_{x_1 - \mu \text{ 与 } x_2 - \mu \text{ 独立}} + E(x_2 - \mu)^3$$

$$\therefore = E(x_1 - \mu)^3 + 3E(x_1 - \mu)^2 E(x_2 - \mu) + 3E(x_1 - \mu) E(x_2 - \mu)^2 + E(x_2 - \mu)^3$$

$$E(x_2 - \mu) = E(x_2) - \mu = \mu - \mu = 0 = E(x_1 - \mu)$$

$$\therefore \text{两项的情况是} \quad E(x_1 - \mu)^3 + E(x_2 - \mu)^3$$

$$\therefore \text{推广到 } n \text{ 项}$$

$$\therefore \frac{1}{n^3} E(x_1 - \mu + x_2 - \mu + x_3 - \mu + \dots + x_n - \mu)^3 = \frac{1}{n^3} \sum_{i=1}^n E(x_i - \mu)^3$$

$$\therefore n E[(\bar{x} - \mu)^3] = \frac{n}{n^3} \sum_{i=1}^n \sigma_3 = \frac{\sigma_3}{n}$$



$$\textcircled{1}: E \left((\bar{x} - \mu) \sum_{i=1}^n (X_i - \mu)^2 \right) = E \left((\bar{x} - \mu) [(X_1 - \mu)^2 + (X_2 - \mu)^2 + \dots + (X_n - \mu)^2] \right)$$

$$E(\bar{x} - \mu)(X_1 - \mu)^2 = E \left(\frac{\sum_{i=1}^n X_i}{n} - \frac{n\mu}{n} \right) (X_1 - \mu)^2 =$$

$$\frac{1}{n} E(X_1 - \mu + X_2 - \mu + X_3 - \mu + \dots + X_n - \mu)(X_1 - \mu)^2$$

$$= \frac{1}{n} E \left((X_1 - \mu)(X_1 - \mu)^2 + (X_2 - \mu)(X_1 - \mu)^2 + \dots + (X_n - \mu)(X_1 - \mu)^2 \right)$$

$$E(X_n - \mu)(X_1 - \mu)^2 \quad \text{当 } n \neq 1 \text{ 时 } X_n - \mu \text{ 与 } (X_1 - \mu)^2 \text{ 独立}$$

$$\therefore = E(X_n - \mu) E(X_1 - \mu)^2$$

$$= \underbrace{(E(X_n) - \mu)}_{=0} E(X_1 - \mu)^2$$

$$= 0$$

$$\therefore E(\bar{x} - \mu)(X_1 - \mu)^2 = \frac{1}{n} E(X_1 - \mu)^3$$

$$\therefore E \left((\bar{x} - \mu) \sum_{i=1}^n (X_i - \mu)^2 \right) = \frac{1}{n} \sum_{i=1}^n E(X_i - \mu)^3 = \frac{1}{n} \cdot n \cdot V_3 = V_3$$

$$\therefore \text{校正} = \frac{1}{n-1} V_3 - \frac{V_3}{n(n-1)} = \frac{nV_3 - V_3}{n(n-1)} = \frac{V_3}{n}$$



$$3. \quad \bar{X}_1 \sim N(\mu, \frac{\sigma^2}{n}) \quad \bar{X}_2 \sim N(\mu, \frac{\sigma^2}{n})$$

$$Z = \bar{X}_1 - \bar{X}_2 \sim N(0, \frac{2\sigma^2}{n})$$

$$P(|\bar{X}_1 - \bar{X}_2| > \sigma) \leq 0.01$$

$$P(Z > \sigma) + P(Z < -\sigma) \leq 0.01$$

$$2P(Z < -\sigma) \leq 0.01$$

$$P(Z < -\sigma) \leq \frac{0.01}{2}$$

$$P\left(\frac{Z-0}{\sqrt{\frac{2\sigma^2}{n}}} < \frac{-\sigma}{\sqrt{\frac{2\sigma^2}{n}}}\right) \leq \frac{0.01}{2}$$

$$\Phi\left(\frac{\sqrt{n}}{\sqrt{2}}\right) \leq 0.005$$

$$1 - \Phi\left(\frac{\sqrt{n}}{\sqrt{2}}\right) \leq 0.005$$

$$\Phi\left(\frac{\sqrt{n}}{\sqrt{2}}\right) \geq 0.995$$

$$\frac{\sqrt{n}}{\sqrt{2}} \geq 1.575$$

$$n \geq 13.261$$

$\therefore n$ 至少为 14

X 为总体

4.

$$E(X) = \frac{1}{\theta} = \theta \quad \text{Var}(X) = \frac{1}{\theta^2} = \theta^2$$

$$\therefore \bar{X} \sim N(\theta, \frac{\theta^2}{40})$$

5.
(a)

$$g_i = F(X_{(i)}) \quad g_1 = F(X_{(1)}) \quad g_2 = F(X_{(2)}) \cdots g_n = F(X_{(n)})$$

$\because F(X)$ 为分布函数 \therefore 其单调非减

$$\because X_1 \leq X_2 \leq \cdots \leq X_n \quad \therefore F(X_{(1)}) \leq F(X_{(2)}) \leq \cdots F(X_{(n)})$$

$$\therefore g_1 \leq g_2 \leq \cdots \leq g_n$$

\therefore 分布函数有界

$$P(g_i = x) = P(F(X_{(i)}) = x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\therefore g_i \sim U(0, 1)$$

$\therefore g_i$ 是来自 $U(0, 1)$ 总体的次序统计量



(b)

$\therefore \eta_i$ 是来自均匀分布 $U(0,1)$ 总体的次序统计量

$$\therefore \eta_i \sim \text{Be}(i, n-i+1)$$

$$\therefore E(\eta_i) = \frac{i}{n+1} \quad \text{Var}(\eta_i) = \frac{i(n-i+1)}{(n+1)^2(n+2)} \quad i=1, 2, \dots, n$$

(3)

$$\text{Cov}(\eta_i, \eta_j) = E[(\eta_i - E(\eta_i))(\eta_j - E(\eta_j))] = E(\eta_i \eta_j) - E(\eta_i)E(\eta_j)$$

$$E(\eta_i)E(\eta_j) = \frac{i}{n+1} \cdot \frac{j}{n+1} = \frac{ij}{(n+1)^2}$$

$$P_{ij}(y, z) = \frac{n!}{(i-1)!(j-i-1)!(n-j)!} [F(y)]^{i-1} p(y) [F(z) - F(y)]^{j-i-1} [1-F(z)]^{n-j} \quad 0 < y < z < 1$$

$$p(y) = p_i(y) = \begin{cases} 1 & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$p(z) = p_j(z) = \begin{cases} 1 & 0 < z < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$F(y) = \begin{cases} y & 0 < y < 1 \\ 0 & y \leq 0 \\ 1 & y \geq 1 \end{cases}$$

$$F(z) = \begin{cases} z & 0 < z < 1 \\ 0 & z \leq 0 \\ 1 & z \geq 1 \end{cases}$$

$$\therefore P_{ij}(y, z) = \frac{n!}{(i-1)!(j-i-1)!(n-j)!} y^{i-1} (z-y)^{j-i-1} (1-z)^{n-j} \quad 0 < y < z < 1$$



$$E(\eta_i \eta_j) = \int_0^1 z dz \int_0^z y \frac{n!}{(i-1)!(j-i-1)!(n-j)!} y^{i-1} (z-y)^{j-i-1} (1-z)^{n-j} dy$$

$$= \frac{n!}{(i-1)!(j-i-1)!(n-j)!} \int_0^1 z dz \int_0^z y^i (z-y)^{j-i-1} (1-z)^{n-j} dy$$

$$= \frac{n!}{(i-1)!(j-i-1)!(n-j)!} \int_0^1 z^{j-i} (1-z)^{n-j} dz \int_0^z y^i (1-\frac{y}{z})^{j-i-1} dy$$

$$= \frac{n!}{(i-1)!(j-i-1)!(n-j)!} \int_0^1 z^{j-i} (1-z)^{n-j} dz \int_0^1 z^i x^i (1-x)^{j-i-1} z dx$$

$$= \frac{n!}{(i-1)!(j-i-1)!(n-j)!} \int_0^1 z^{j+1} (1-z)^{n-j} dz \int_0^1 x^i (1-x)^{j-i-1} dx$$

$$= \frac{ij+1}{(n+1)(n+2)} \quad \text{Cov}(\eta_i, \eta_j) = \frac{a_1(1-a_2)}{(n+2)}$$

$$\begin{pmatrix} \text{Var}(\eta_i) & \text{Cov}(\eta_i, \eta_j) \\ \text{Cov}(\eta_j, \eta_i) & \text{Var}(\eta_j) \end{pmatrix} = \begin{pmatrix} \frac{a_1(1-a_1)}{(n+2)} & \frac{a_1(1-a_2)}{(n+2)} \\ \frac{a_1(1-a_2)}{(n+2)} & \frac{a_2(1-a_2)}{(n+2)} \end{pmatrix}$$



6. $m_{0.5} \sim N \left(X_{0.5}, \frac{0.5(1-0.5)}{n \cdot p(0.5)} \right)$

$$p(0.5)^2 = \left(6 \times \frac{1}{2} \times \frac{1}{2} \right)^2 = \frac{9}{4}$$

$\therefore m_{0.5} \sim N \left(0.5, \frac{1}{9n} \right)$ 渐近分布



概统 hw-16

$$1. \quad x_1, x_2, \dots, x_n \sim N(\mu, 1)$$

$$\therefore \bar{x} \sim N(\mu, \frac{1}{n})$$

$$P(|\bar{x}| < c) = P(-c < \bar{x} < c) = P(\bar{x} < c) - P(\bar{x} < -c)$$

$$= P\left(\frac{\bar{x} - \mu}{\sqrt{\frac{1}{n}}} < \frac{c - \mu}{\sqrt{\frac{1}{n}}}\right) - P\left(\frac{\bar{x} - \mu}{\sqrt{\frac{1}{n}}} < \frac{-c - \mu}{\sqrt{\frac{1}{n}}}\right)$$

$$= \Phi\left(\frac{c - \mu}{\sqrt{\frac{1}{n}}}\right) - \Phi\left(\frac{-c - \mu}{\sqrt{\frac{1}{n}}}\right) \leq \alpha$$

$$\Phi(\sqrt{n}(c - \mu)) - \Phi(\sqrt{n}(-c - \mu)) \leq \alpha$$

$$\varphi(\sqrt{n}(c - \mu)) \cdot -\sqrt{n} - \varphi(\sqrt{n}(-c - \mu)) \cdot -\sqrt{n} \leq \alpha$$

$$-\sqrt{n}(\varphi(\sqrt{n}(c - \mu)) - \varphi(\sqrt{n}(-c - \mu))) \leq \alpha$$

$$\varphi(\sqrt{n}(-c - \mu)) \leq \varphi(\sqrt{n}(c - \mu))$$

$$\therefore -\sqrt{n}(\varphi(\sqrt{n}(c - \mu)) - \varphi(\sqrt{n}(-c - \mu))) \leq 0$$

其最大值为 $-c - \mu + c - \mu = 0$ 时 $\mu = 0$

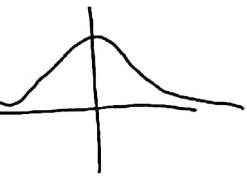
$$\Phi(\sqrt{n}c) - \Phi(\sqrt{n}c) \leq \alpha$$

$$2\Phi(\sqrt{n}c) - 1 \leq \alpha$$

$$\Phi(\sqrt{n}c) \leq \frac{\alpha + 1}{2}$$

$$\therefore 0 \leq \sqrt{n}c \leq U_{\frac{\alpha+1}{2}}$$

$$0 \leq c \leq \frac{U_{\frac{\alpha+1}{2}}}{\sqrt{n}}$$



2. $X \sim F(n, m)$

$$p_X(x) = \frac{\Gamma(\frac{m+n}{2}) (\frac{n}{m})^{\frac{n}{2}}}{\Gamma(\frac{n}{2}) \Gamma(\frac{m}{2})} \cdot x^{\frac{n}{2}-1} (1 + \frac{n}{m}x)^{-\frac{(m+n)}{2}} \quad x \geq 0$$

$$z = \frac{\frac{n}{m}x}{1 + \frac{n}{m}x}$$

$$(1 + \frac{n}{m}x)z = \frac{n}{m}x$$

$$x = \frac{mz}{n-nz}$$

$$\frac{d(\frac{m}{n-nz})}{dz} = \frac{mn}{(n-nz)^2}$$

$$p_z(z) = \frac{\Gamma(\frac{m+n}{2}) (\frac{n}{m})^{\frac{n}{2}}}{\Gamma(\frac{n}{2}) \Gamma(\frac{m}{2})} \cdot \left(\frac{mz}{n-nz}\right)^{\frac{n}{2}-1} \left(1 + \frac{n}{m} \frac{mz}{n-nz}\right)^{-\frac{m+n}{2}} \cdot \frac{mn}{n(1-z)^2} \quad 0 < z < 1$$

$$= \frac{\Gamma(\frac{m+n}{2}) (\frac{n}{m})^{\frac{n}{2}}}{\Gamma(\frac{n}{2}) \Gamma(\frac{m}{2})} \left(\frac{m}{n}\right)^{\frac{n}{2}-1} \left(\frac{z}{1-z}\right)^{\frac{n}{2}-1} \left(\frac{1}{1-z}\right)^{-\frac{(m+n)}{2}} \cdot \frac{1}{(1-z)^2}$$

$$= \frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{n}{2}) \Gamma(\frac{m}{2})} \left(\frac{1}{1-z}\right)^{\frac{n}{2}-1} (z)^{\frac{n}{2}-1} \left(\frac{1}{1-z}\right)^{-\frac{(m+n)}{2}} \frac{1}{(1-z)^2}$$

$$= \frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{n}{2}) \Gamma(\frac{m}{2})} (z)^{\frac{n}{2}-1} (1-z)^{\frac{m}{2}-1} \quad 0 < z < 1$$

$\therefore z \sim \text{Be}(\frac{n}{2}, \frac{m}{2})$ 参数为 F 分布的分子分母自由度的 $\frac{1}{2}$



hw-16

3.

$$S_w^2 = \frac{(n-1)S_x^2 + (m-1)S_y^2}{n+m-2}$$

$$S_x^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

$$S_y^2 = \frac{1}{m-1} \sum_{i=1}^m (Y_i - \bar{Y})^2$$

$$= \frac{\sum_{i=1}^n (X_i - \bar{X})^2 + \sum_{i=1}^m (Y_i - \bar{Y})^2}{n+m-2}$$

$$= \frac{\sum_{i=1}^n X_i^2 - n\bar{X}^2 + \sum_{i=1}^m Y_i^2 - m\bar{Y}^2}{n+m-2}$$

$$\bar{X} \sim (\mu_1, \frac{\sigma^2}{n}), \quad \bar{Y} \sim (\mu_2, \frac{\sigma^2}{m})$$

$$c\bar{X} \sim (c\mu_1, \frac{c^2\sigma^2}{n}), \quad d\bar{Y} \sim (d\mu_2, \frac{d^2\sigma^2}{m})$$

$$c\bar{X} + d\bar{Y} \sim (c\mu_1 + d\mu_2, \frac{c^2\sigma^2}{n} + \frac{d^2\sigma^2}{m})$$

$$\frac{c\bar{X} + d\bar{Y} - (c\mu_1 + d\mu_2)}{\sqrt{\frac{c^2\sigma^2}{n} + \frac{d^2\sigma^2}{m}}} \sim N(0, 1)$$

$$\frac{(n-1)S_x^2}{\sigma^2} \sim \chi^2(n-1), \quad \frac{(m-1)S_y^2}{\sigma^2} \sim \chi^2(m-1)$$

$$\frac{(n-1)S_x^2}{\sigma^2} \text{ 与 } \frac{(m-1)S_y^2}{\sigma^2} \text{ 独立}$$

$$\therefore \frac{(n-1)S_x^2 + (m-1)S_y^2}{\sigma^2} \sim \chi^2(n+m-2) = \frac{(n+m-2)S_w^2}{\sigma^2}$$

$$\therefore S_w^2 \text{ 与 } \bar{X} + \bar{Y} \text{ 独立} \quad \frac{(c\bar{X} + d\bar{Y} - (c\mu_1 + d\mu_2))\sigma}{\sqrt{\frac{c^2\sigma^2}{n} + \frac{d^2\sigma^2}{m}} S_w} \sim t(n+m-2)$$

$$\therefore \frac{c(\bar{X} - \mu_1) + d(\bar{Y} - \mu_2)}{S_w \sqrt{\frac{c^2}{n} + \frac{d^2}{m}}} \sim t(n+m-2)$$

