

第十一周第二次作业.

1. P52 4. $F(x)$ 对称 $\Leftrightarrow F(x) = 1 - F(-x + 0)$

Prove: 对称 \Leftrightarrow ch.f. 为实. 偶函数

$$\text{Proof: "} \Rightarrow \text{" } f(t) = \int_{-\infty}^{+\infty} e^{itx} dF(x) = \int_0^{+\infty} (e^{itx} + e^{-itx}) dF(x) \\ = \int_0^{+\infty} (e^{i|t|x} + e^{-i|t|x}) dF(x) \text{ 为实. 偶函数}$$

" \Leftarrow " 对于实. 偶函数 $f(t): \mathbb{R} \rightarrow \mathbb{R}$. $f(t) = f(-t)$

$$dF(x) = dx \cdot \frac{1}{2\pi} \int_{-\infty}^{+\infty} e^{-itx} f(t) dt \\ = dx \cdot \frac{1}{2\pi} \int_0^{+\infty} (e^{itx} + e^{-itx}) f(t) dt$$

$$\Rightarrow dF(x) = -dF(-x)$$

$\Rightarrow F(x)$ 对称.

11. $\xi_1, \xi_2, \dots, \xi_n$ iid. $\sim U(1, 0)$, 求 $\bar{\xi} = \frac{1}{n} \sum_{k=1}^n \xi_k$ 分布

设 $\xi_1, \xi_2, \dots, \xi_n$ 的 ch.f. 为 f_1, f_2, \dots, f_n , $\bar{\xi}$ 的 ch.f. 为 \bar{f}

$$\text{则 } \bar{f} = \frac{1}{n} \sum_{k=1}^n f_k = f_1 = f_2 = \dots = f_n$$

$\Rightarrow \bar{\xi} \sim U(1, 0)$ 与 $\xi_1, \xi_2, \dots, \xi_n$ 相同.

15. Prove: y.v. $\xi_1, \xi_2, \dots, \xi_n$ 独立 \Leftrightarrow 联合 ch.f. 为边缘 ch.f. 之积.

$$\text{"} \Rightarrow \text{" } dF(x_1, x_2, \dots, x_n) = dF(x_1) dF(x_2) \dots dF(x_n)$$

$$f(t) = \int_{\mathbb{R}^n} e^{it_1 x_1 + it_2 x_2 + \dots + it_n x_n} dF(x_1, x_2, \dots, x_n)$$

$$f_1(t_1) = \int_{-\infty}^{+\infty} e^{it_1 x_1} dF(x_1)$$

$$f_2(t_2) = \int_{-\infty}^{+\infty} e^{it_2 x_2} dF(x_2)$$

...

$$f_n(t_n) = \int_{-\infty}^{+\infty} e^{it_n x_n} dF(x_n)$$

$$\begin{aligned} \text{乘积: } \prod_{k=1}^n f_k(t_k) &= \int_{\mathbb{R}^n} e^{it_1 x_1 + it_2 x_2 + \dots + it_n x_n} dF_1(x_1) dF_2(x_2) \dots dF_n(x_n) \\ &= \int_{\mathbb{R}^n} e^{it_1 x_1 + it_2 x_2 + \dots + it_n x_n} dF(x_1, x_2, \dots, x_n) \end{aligned}$$

$$“\Leftarrow” f(t) = f_1(t_1) f_2(t_2) \dots f_n(t_n) \quad (已证)$$

$$\begin{aligned} |\mathbb{R}| \quad dF(x) &= dx \cdot \left(\frac{1}{2\pi}\right)^n \cdot \int_{\mathbb{R}^n} e^{-it_1 x_1 - it_2 x_2 - \dots - it_n x_n} df(t_1, t_2, \dots, t_n) \\ &= dx \cdot \left(\frac{1}{2\pi}\right)^n \cdot \int_{\mathbb{R}^n} e^{-it_1 x_1 - it_2 x_2 - \dots - it_n x_n} df_1(t_1) df_2(t_2) \dots df_n(t_n) \\ &= dx \cdot \prod_{k=1}^n \frac{1}{2\pi} \int_{-\infty}^{+\infty} e^{-it_k x_k} df_k(t_k) \\ &= \prod_{k=1}^n dF_k(x_k) \end{aligned}$$

Prob 5. ξ, η iid $\sim N(0,1)$

$$Z = \begin{cases} |\eta| & \xi > 0 \\ -|\eta| & \xi < 0 \end{cases}$$

则 Z 服从 $N(0,1)$ 分布, 但协方差阵:

$$B = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix} \Rightarrow |\Sigma| > 1. \quad \text{不是元正态分布}$$

6. (ξ, η) 服从元正态分布

$$\xi + \eta \text{ 与 } \xi - \eta \text{ 独立} \Leftrightarrow f(\xi^2 - \eta^2) = f(\xi + \eta) f(\xi - \eta)$$

$$\Leftrightarrow \iint_{\mathbb{R}^2} e^{it(x^2-y^2)} p(x,y) dx dy = \iint_{\mathbb{R}^2} e^{it(x+y)} p(x,y) dx dy \cdot \iint_{\mathbb{R}^2} e^{it(x-y)} p(x,y) dx dy$$

$$(\text{LHS} = \iint_{\mathbb{R}^2} e^{it(x^2-y^2)} p(x,y) dx dy \cdot \iint_{\mathbb{R}^2} p(x,y) dx dy)$$

$$\Leftrightarrow \iint_{\mathbb{R}^2} [e^{it(x^2-y^2)} - e^{it(x+y)} e^{it(x-y)}] p(x,y) dx dy = 0$$

$$\Leftrightarrow \iint_{\mathbb{R}^2} [e^{it(x^2-y^2)} - e^{-t^2(x^2-y^2)}] p(x,y) dx dy = 0$$

分别为高斯矩量得: 上式 $\Leftrightarrow G_1^2 = G_2^2 \Leftrightarrow b_{11} = b_{22}$

8. $\xi_1, \xi_2, \dots, \xi_n \text{ iid. } \sim N(0, 1), \eta_1 = \sum_{k=1}^n \xi_k, \eta_2 = \sum_{k=1}^m \xi_k \quad (m < n)$

求 η_1 和 η_2 的联合分布

$$D(\eta_1) = m, \quad D(\eta_2) = n$$

$$D(\eta_1, \eta_2) = m$$

$$\text{协方差阵 } B = \begin{pmatrix} m & m \\ m & n \end{pmatrix}$$

$$\text{联合分布 } (\eta_1, \eta_2) \sim N(\vec{0}, B)$$

2. P202 1. X and Y independent r.v.'s, $X \sim N(\mu_1, \sigma^2), Y \sim N(\mu_2, \sigma^2)$

Prove: $X+Y$ and $X-Y$ independent

$$\text{Proof: } U = X+Y, V = X-Y \Rightarrow X = \frac{U+V}{2}, Y = \frac{U-V}{2}$$

$$\Rightarrow \text{cov}(U, V) = \text{cov}(X+Y, X-Y) = D(X) - \text{cov}(X, Y) + \text{cov}(Y, X) - D(Y)$$

$$= D(X) - D(Y) = \sigma^2 - \sigma^2 = 0$$

Since U and V are jointly normal and $\text{cov}(U, V) = 0$

Then U and V are independent

2. Suppose x_1, x_2, \dots, x_n jointly normal, $\text{cov}(x_i, x_j) = 0$ ($i \neq j$)

Then x_1, x_2, \dots, x_n are independent

Proof: $X = (x_1, x_2, \dots, x_n)^T$ random vector whose cov matrix is Σ

if $\Sigma = \text{diag}(\sigma_1^2, \sigma_2^2, \dots, \sigma_n^2)$

Then x_1, x_2, \dots, x_n uncorrelated.

Combined by X follows normal distribution,

Thus x_1, x_2, \dots, x_n independent

2. z_1, z_2, \dots, z_n iid $\sim N(0, \sigma^2)$

Prove: (1) $\bar{z}, \sum z_i^2$ independent

(2) $\bar{z} \sim N(0, \frac{\sigma^2}{n})$

(3) $\frac{\sum z_i^2}{\sigma^2} \sim \chi^2(n-1)$

Proof: (1) $\bar{z} \sim N(0, \bar{B})$, $\bar{A} = (1, 1, \dots, 1)$, $B = \sigma^2 I$

取 C 为 n 个正交向量:

$$C = \begin{bmatrix} \frac{1}{\sqrt{n-2}} & \frac{1}{\sqrt{n-2}} & 0 & \dots \\ \frac{1}{\sqrt{2-3}} & \frac{1}{\sqrt{2-3}} & \frac{-2}{\sqrt{2-3}} & 0 & \dots \end{bmatrix}$$

$$\begin{bmatrix} \dots & & & \\ \frac{1}{\sqrt{(n-1)n}} & \frac{1}{\sqrt{(n-1)n}} & \dots & \frac{-(n-1)}{\sqrt{(n-1)n}} \\ \frac{1}{\sqrt{n}} & \frac{1}{\sqrt{n}} & \dots & \frac{1}{\sqrt{n}} \end{bmatrix}$$

经 (正交) 变换后各分量 $\eta_1, \eta_2, \dots, \eta_n$ 仍独立且有相同方差 σ^2

又因 $E(\eta_1) = E(\eta_2) = \dots = E(\eta_{n-1}) = 0$, $E(\eta_n) = \sqrt{n}\alpha$

由正交变换保范数:

$$\sum_{k=1}^n \eta_k^2 = \|\vec{\eta}\|^2 = \|\vec{\bar{z}}\|^2 = \sum_{k=1}^n \bar{z}_k^2$$

$$\text{又 } \eta_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n \bar{z}_k = \sqrt{n} \bar{\bar{z}}$$

$$\Rightarrow n\bar{s}_n^2 = \sum_{k=1}^{n-1} \eta_k^2$$

$$\eta_1, \eta_2, \dots, \eta_n \text{ 独立} \Rightarrow \bar{z} = \frac{\eta_n}{\sqrt{n}} \text{ 与 } \bar{s}_n^2 = \frac{1}{n} \sum_{k=1}^{n-1} \eta_k^2 \text{ 独立}$$

(2) $\eta_n = \sqrt{n} \bar{\bar{z}}$ 且 $\eta_n \sim N(\alpha, \sigma^2) \Rightarrow \bar{\bar{z}} \sim (\alpha, \frac{\sigma^2}{n})$

(3) $\frac{n\bar{s}_n^2}{\sigma^2} = \sum_{k=1}^{n-1} \left(\frac{\eta_k}{\sigma}\right)^2$, $\frac{\eta_k}{\sigma} \sim N(0, 1)$

$$\Rightarrow \frac{n\bar{s}_n^2}{\sigma^2} \sim \chi^2(n-1)$$