5.8

5. Let X_1, X_2, \ldots be independent N(0, 1) variables. Use characteristic functions to find the distribution of: (a) X_1^2 , (b) $\sum_{i=1}^n X_i^2$, (c) X_1/X_2 , (d) X_1X_2 , (e) $X_1X_2 + X_3X_4$.

(e)

爾·
$$\Psi_{x_1x_2+x_3x_4}(t) = E[e^{it(x_1x_2+x_3x_4)}] = E[e^{itx_1x_2}] E[e^{itx_3x_4}] = (\Psi_{x_1x_2}(t))^2$$

$$\begin{aligned} \varphi_{x,x_{2}}(t) &= E[e^{itx_{1}x_{2}}] = E[E[e^{itx_{1}x_{2}}|x_{2}]] = E[\varphi_{x,(x_{2}t)}] = E[e^{-\frac{1}{2}x_{2}^{2}t^{2}}] \\ &= \int_{-\infty}^{+\infty} e^{-\frac{1}{2}x^{2}t^{2}} \cdot \sqrt{\frac{1}{2\pi}} e^{-\frac{X^{2}}{2}} dx = \frac{1}{\sqrt{1+t^{2}}} \end{aligned}$$

5.8

- **9.** Find the characteristic functions of the following density functions:
- (a) $f(x) = \frac{1}{2}e^{-|x|}$ for $x \in \mathbb{R}$,
- (b) $f(x) = \frac{1}{2}|x|e^{-|x|}$ for $x \in \mathbb{R}$.

$$\hat{\mathbf{p}}_{\mathbf{r}}$$
: (a) $\mathbf{V}(\mathbf{t}) = \mathbf{E}[\mathbf{e}^{i\mathbf{t}}] = \frac{1}{2} \left(\frac{1}{1-i\mathbf{t}} + \frac{1}{1+i\mathbf{t}} \right) = \frac{1}{1+\mathbf{t}^2}$
 $\mathbf{v}_{\mathbf{r}}(\mathbf{t})$ $\mathbf{v}_{\mathbf{r}}(\mathbf{r})$

$$\varphi_2(t) = E[e^{itx}] = \int_0^{t\infty} e^{itx} \cdot \chi e^{-x} dx = \frac{1}{(it-1)^2}$$

$$\varphi(t) = \frac{1}{2} \left(\frac{1}{(1+t)^2} + \frac{1}{(1+t)^2} \right) = \frac{1-t^2}{(1+t^2)^2}$$

5.9

2. Let X_n have distribution function

$$F_n(x) = x - \frac{\sin(2n\pi x)}{2n\pi}, \qquad 0 \le x \le 1.$$

- (a) Show that F_n is indeed a distribution function, and that X_n has a density function.
- (b) Show that, as $n \to \infty$, F_n converges to the uniform distribution function, but that the density function of F_n does not converge to the uniform density function.

 $\nu E \cdot (a) f_n(x) = 1 - \cos(2n\pi x) \ge 0, \int_0^1 f_n(x) dx = \int_0^1 1 - \cos(2n\pi x) dx = 1 - \frac{\sin(2n\pi)}{2n\pi} = 1$ ⇒f_n(X)为窓度函数,F_n(X)为分布函数

(b) n→∞ B寸、Fn(x) → x、1里 fn(x) 不以文定义

5. 95. Use the inversion theorem to show that

$$\int_{-\infty}^{\infty} \frac{\sin(at)\sin(bt)}{t^2} dt = \pi \min\{a, b\}.$$

ib: ig v.v. X 在[-a,a]上tヨヨ分布, v,v. 丫在[-b,b]上tヨヨ分布. 且 X 与 Y 9 E z

$$\Psi_{x}(t) = E[e^{itx}] = \int_{-a}^{a} e^{itx} \cdot \frac{1}{2a} dx = \frac{1}{2a} \cdot \frac{1}{it} e^{itx} \Big|_{-a}^{a} = \frac{\sin(at)}{at}$$

同理,
$$\psi_{Y}(t) = \frac{\sin(bt)}{bt}$$
 $\psi_{X+Y}(t) = \frac{\sin(at)\sin(bt)}{abt^2}$

由反转公式矩, $f(\alpha) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} e^{-it\alpha} \varphi(t) dt$. 代入 $\alpha = 0$.

$$f_{x+y}(0) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \varphi_{x+y}(t) dt = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \frac{\sin(\alpha t) \sin(bt)}{\alpha b t^2} dt$$

$$f_{x+y}(0) = \begin{cases} min\{a,b\} \\ -min\{a,b\} \end{cases} \frac{1}{2a} \cdot \frac{1}{2b} = \frac{min\{a,b\}}{2ab}$$

$$\int_{-\infty}^{+\infty} \frac{\sin(at)\sin(bt)}{t^2} dt = \pi \cdot \min\{a,b\}$$

8. Let X_1, X_2 have a bivariate normal distribution with zero means, unit variances, and correlation ρ . Use the inversion theorem to show that

$$\frac{\partial}{\partial \rho} \mathbb{P}(X_1 > 0, \ X_2 > 0) = \frac{1}{2\pi\sqrt{1 - \rho^2}}.$$

 $(t, t_1)\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}\begin{pmatrix} t_1 \\ t_2 \end{pmatrix}$

Hence find $\mathbb{P}(X_1 > 0, X_2 > 0)$.

ντ: φ(t) = φ(t,t2) = exp(iμt - ± t Σt) = exp (- ± t Σt)

 $\frac{\partial}{\partial \rho} p(x_1 > 0, x_2 > 0) = \frac{\partial}{\partial \rho} \int_0^{\infty} \int_0^{\infty} f(x_1, x_2) dx_1 dx_2$

$$\frac{x + x + x + x}{2} \frac{\partial}{\partial \rho} \frac{\partial}{\partial \rho} \frac{1}{4\pi^2} \int \frac{e^{x} \rho(-\frac{1}{2} + \frac{1}{2} + \frac{1}{2} + \frac{1}{2})}{(i + x)(i + x)} dt_x dt_y$$

$$\frac{\overrightarrow{x} + x \cancel{R} + y}{3 \rho} \xrightarrow{\frac{1}{2} \rho} \int_{\mathbb{R}^{2}} \frac{e^{x} \rho(-\frac{1}{2} + \tau + \Sigma +)}{(it_{1})(it_{2})} dt_{1} dt_{2}$$

$$= \frac{1}{4 \pi^{2}} \int_{\mathbb{R}^{2}} e^{x} \rho(-\frac{1}{2} + \tau + \Sigma +) dt_{1} dt_{2} = \frac{2\pi |\Sigma^{-1}|^{\frac{1}{2}}}{4 \pi^{2}} = \frac{1}{2\pi \sqrt{1-\rho^{2}}}$$

由ex.4.7.5 知 D(X, >0, X2>0)= 1+ = 1 sin P

5.10

1. Prove that, for $x \ge 0$, as $n \to \infty$,

(a)
$$\sum_{\substack{k:\\ |k-\frac{1}{2}n| \le \frac{1}{2}x, \sqrt{n}}} \binom{n}{k} \sim 2^n \int_{-x}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} du,$$

(b)
$$\sum_{\substack{k: \ |k-n| \le x\sqrt{n}}} \frac{n^k}{k!} \sim e^n \int_{-x}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}u^2} du.$$

$$\hat{V}F: (A) \quad X_{1} \sim B(1, \frac{1}{2}) \quad S_{n} = X_{1} + \cdots + X_{n} \quad Var(X_{1}) = \frac{1}{4}, \quad E(X_{1}) = \frac{1}{2}$$

$$P\left(\frac{1S_{n} - \frac{1}{2}NI}{\frac{1}{2}\sqrt{n}} \leq X\right) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K) = \sum_{\substack{|K - \frac{1}{2}N| \\ \frac{1}{2}\sqrt{n}}} P(S_{n} = K)$$

$$\sum \frac{2^{N}-\frac{1}{2}\sqrt{N}}{\frac{7}{2}\sqrt{N}} \leq x \xrightarrow{N\to\infty} \Phi(x) - \Phi(-x) = \int_{-x}^{-x} \frac{\sqrt{2\pi}}{1} e^{-\frac{1}{2}N^{2}} dn$$

$$|K-\frac{2}{3}u| = \frac{2}{3}x\sqrt{u}$$

$$\sum_{k} {k \choose k} \sim 5_{k} \sum_{x} {\sqrt{2u} \choose x} e_{-\frac{2}{3}n_{x}} dn$$

(b)
$$X_{1} \sim Poi(1)$$
 $S_{n} = X_{1} + \dots + X_{N}$ $Var(X_{1}) = 1$, $E(x_{1}) = 1$

$$P\left(\frac{|S_{N} - N|}{\sqrt{N}} \leq X\right) = \sum_{i} P(S_{N} = K) = \sum_{i} e^{-N} \cdot \frac{N^{k}}{K!}$$

$$\frac{|K - N|}{\sqrt{N}} \leq X$$

$$\frac{1}{2} \frac{1}{\sqrt{n}} = \frac{1}{\sqrt{n}} = \frac{1}{2} \frac{1}{\sqrt{$$

$$\sum_{k:} \frac{k!}{n^k} \sim e^n \int_{-x}^{x} \sqrt{\frac{2\pi}{2\pi}} e^{-\frac{1}{2}n^2} du$$

3. Let X have the $\Gamma(1, s)$ distribution; given that X = x, let Y have the Poisson distribution with parameter x. Find the characteristic function of Y, and show that

$$\frac{Y - \mathbb{E}(Y)}{\sqrt{\operatorname{var}(Y)}} \xrightarrow{D} N(0, 1) \quad \text{as } s \to \infty.$$

Explain the connection with the central limit theorem.

$$\frac{\lambda \cdot \mathbb{E}}{\lambda \cdot \mathbb{E}} = \mathbb{E}[e^{it\tau}] = \mathbb{E}[e^{it\tau}] \times \mathbb{E}[e^{it\tau}] \times \mathbb{E}[e^{it\tau}]$$

$$\frac{X \sim \mathbb{E}(1.5)}{1 - (e^{it} - 1)})^{5} = \left(\frac{1}{2 - e^{it}}\right)^{5}$$

E[T] =
$$\frac{1}{7} \phi_{7}(0) = 5$$
, E[T] = $-\phi_{7}(0) = 5^{2} + 25$, Var(Y) = 25

$$\psi_{2}(t) = E[\exp(it \cdot \frac{\tau - s}{\sqrt{2s}})] = e^{-it\sqrt{s}} \psi_{\tau}(\frac{t}{\sqrt{2s}})$$

$$\log (\phi_{\epsilon}(t)) \rightarrow e^{-\frac{1}{2}t^{2}}, \ t \times Z = \frac{\Upsilon - E(\Upsilon)}{\sqrt{V_{\alpha}(\Upsilon)}} \longrightarrow N(0,1)$$

联系:同样也可以用中心极限定理证明:

4. Let X_1, X_2, \ldots be independent random variables taking values in the positive integers, whose common distribution is non-arithmetic, in that $gcd\{n : \mathbb{P}(X_1 = n) > 0\} = 1$. Prove that, for all integers x, there exist non-negative integers r = r(x), s = s(x), such that

$$\mathbb{P}(X_1 + \cdots + X_r - X_{r+1} - \cdots - X_{r+s} = x) > 0.$$

记:设以在〈n,,…,nk}中每点取值的根系学>0.

tolan,对ynzn,neZ,都存在 a,...,ak非负.

$$S.t. N = N_1 \alpha_1 + \cdots + N_K \alpha_K$$

 $P(X_1 + \dots + X_{V-1} - \dots - X_{V+1} = X) \ge P(X_1 + \dots + X_{V+1} = N + X) P(X_{V+1} + \dots + X_{V+1} = N)$ $P(X_{V+1} + \dots + X_{V+1} = N) \ge \frac{K}{1} P(X_1 = N_1)^{\beta_1} > 0.$

$$\begin{cases} P(X_{r+1} + \dots + X_{r+s} = N) > \prod_{i=1}^{K} P(X_i = N_i)^{\beta_i} > 0 \\ P(X_1 + \dots + X_{r+s} = N + X) > \prod_{i=1}^{K} P(X_i = N_i)^{\gamma_i} > 0 \end{cases}$$

7.2

- **1.** (a) Suppose $X_n \xrightarrow{r} X$ where $r \ge 1$. Show that $\mathbb{E}|X_n^r| \to \mathbb{E}|X^r|$.
- (b) Suppose $X_n \xrightarrow{1} X$. Show that $\mathbb{E}(X_n) \to \mathbb{E}(X)$. Is the converse true?
- (c) Suppose $X_n \stackrel{2}{\to} X$. Show that $var(X_n) \to var(X)$.
- **2. Dominated convergence.** Suppose $|X_n| \leq Z$ for all n, where $\mathbb{E}(Z) < \infty$. Prove that if $X_n \stackrel{P}{\to} X$ then $X_n \stackrel{1}{\to} X$.
- 1. vu: (a) 由 Minkowski 不等式、(E[x*]) + (E[x*]) + (E[xn*]) +,

$$(E|X_{N}^{r}|)^{\frac{1}{r}} \leq (E|(X-X_{N})^{r}|)^{\frac{1}{r}} + (E|X_{N}|)^{\frac{1}{r}}$$

$$\therefore Var(Xn) = E[Xn^2] - E[Xn]^2 \longrightarrow E[X^2] - E[X]^2 = Var(X)$$

$$\therefore X_{n} \xrightarrow{p} X \qquad \therefore p(\lim_{n\to\infty} Z_{n} \geq \xi) = p(\lim_{n\to\infty} |X_{n} - X| \geq \xi) = 0.$$

$$\forall \text{270}, \text{E[2n]} = \text{E[2n1}_{\{2n \leq \zeta\}}] + \text{E[2n1}_{\{2n > \zeta\}}] \leq \text{2} + \text{2} \text{E[21}_{\{2n > \zeta\}}]$$