

第九次作业

b.1

设随机变量 $Y = X - \mu$, 那么有 $E(Y) = 0, \text{Var}(Y) = \sigma^2$.

$$\Rightarrow P(X - \mu \leq -\varepsilon) = P(Y \leq -\varepsilon)$$

$$= P(Y - t \leq -\varepsilon - t)$$

$$\leq P((Y+t)^2 \geq (\varepsilon+t)^2). \text{ 由 Markov 不等式.}$$

$$\leq \frac{E((Y+t)^2)}{(\varepsilon+t)^2} = \frac{\sigma^2 + t^2}{(\varepsilon+t)^2}$$

$$\Rightarrow P(X - \mu \leq -\varepsilon) \leq \left(\frac{\sigma^2 + t^2}{(\varepsilon+t)^2} \right)_{\min}$$

$$\frac{\sigma^2 + t^2}{(\varepsilon+t)^2} = \frac{(t+\varepsilon)^2 - 2\varepsilon(t+\varepsilon) + \sigma^2 + \varepsilon^2}{(\varepsilon+t)^2} = 1 - 2\varepsilon \times \frac{1}{\varepsilon+t} + \frac{\sigma^2 + \varepsilon^2}{(\varepsilon+t)^2}$$

$$\text{当 } \frac{1}{\varepsilon+t} = \frac{\varepsilon}{\sigma^2 + \varepsilon^2} \text{ 时取最小值也即 } t = \frac{\sigma^2}{\varepsilon} \text{ 时取最小值 } \frac{\sigma^2}{\varepsilon^2 + \sigma^2}.$$

$$\Rightarrow P(X - \mu \leq -\varepsilon) \leq \frac{\sigma^2}{\varepsilon^2 + \sigma^2}$$

b.2.

$$\text{解: } E(X) = 2, E(Y) = -2 \Rightarrow E(X+Y) = 0.$$

$$\rho_{XY} = \frac{E[(X-E(X))(Y-E(Y))]}{\sqrt{\text{Var}(X)}\sqrt{\text{Var}(Y)}} = -\frac{1}{2}$$

$$\text{Var}(X) = 1, \text{Var}(Y) = 4$$

$$\Rightarrow E[(X-E(X))(Y-E(Y))] = \sqrt{1 \times 4} = 2.$$

$$\Rightarrow \text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2E[(X-E(X))(Y-E(Y))]$$

$$= 4 + 1 - 2 = 3.$$

由 Chebysheve 不等式

$$Pr\{|X+Y| \geq 6\} \leq \frac{\text{Var}(X+Y)}{36} = \frac{1}{12}$$



6.3 由Chebyshev不等式

$$\Pr(|\frac{1}{n} \sum_{i=1}^n X_i - \mu| \geq \varepsilon) \leq \frac{1}{\varepsilon^2} \text{Var}(\frac{1}{n} \sum_{i=1}^n X_i)$$

$$\text{由 } \text{Var}(\frac{1}{n} \sum_{i=1}^n X_i) = \frac{1}{n^2} \text{Var}(\sum_{i=1}^n X_i) = \frac{1}{n} \text{Var}(X_i) \leq \frac{1}{n}$$

$$\Rightarrow \Pr(|\frac{1}{n} \sum_{i=1}^n X_i - \mu| \geq \varepsilon) \leq \frac{1}{\varepsilon^2} \text{Var}(\frac{1}{n} \sum_{i=1}^n X_i) \leq \frac{1}{\varepsilon^2 n}$$

6.4

阐述Chernoff方法:

利用矩生成函数和Markov不等式证明并得到, 对任意随机变 X 和任意 $t > 0, \varepsilon > 0$, 有 $\Pr[X \geq E[X] + \varepsilon] = \Pr[e^{tX} \geq e^{tE[X] + t\varepsilon}] \leq e^{-t\varepsilon - tE[X]} E[e^{tX}]$

6.5

$$\text{令 } \bar{X} = \sum_{i=1}^n X_i, \mu = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n p_i$$

$$\Pr[\frac{1}{n} \sum_{i=1}^n (X_i - E[X_i]) \geq \varepsilon] = \Pr[\frac{1}{n} (\sum_{i=1}^n X_i - \sum_{i=1}^n E[X_i]) \geq \varepsilon]$$

$$= \Pr[\sum_{i=1}^n X_i \geq n\varepsilon + \sum_{i=1}^n E[X_i]]$$

$$= \Pr[\bar{X} \geq n\varepsilon + \mu] \text{ 由Chernoff方法}$$

$$= \Pr[e^{t\bar{X}} \geq e^{t(n\varepsilon + \mu)}]$$

$$\leq e^{-tn\varepsilon - t\mu} E[e^{t\bar{X}}]$$

利用随机变量的独立性和 $1+x \leq e^x$ 有

$$E[e^{t\bar{X}}] = E[e^{t \sum_{i=1}^n X_i}] = \prod_{i=1}^n E[e^{tX_i}]$$

$$= \prod_{i=1}^n [(1-p_i) + p_i e^t] = \prod_{i=1}^n [1 + p_i(e^t - 1)]$$

$$\leq \exp(\sum_{i=1}^n p_i(e^t - 1)) = \exp(\mu(e^t - 1))$$

$$\Rightarrow \Pr[\frac{1}{n} \sum_{i=1}^n (X_i - E[X_i]) \geq \varepsilon] \leq e^{-tn\varepsilon - t\mu} \exp(\mu(e^t - 1)) = e^{-tn\varepsilon - t\mu + \mu e^t}$$

求右式的最小值.

$$\text{令 } F(t) = \mu e^t - (n\varepsilon + \mu)t - \mu$$

$$F'(t) = \mu e^t - n\varepsilon - \mu$$

$$\text{令 } F'(t) = 0 \text{ 得 } e^t = \frac{n\varepsilon + \mu}{\mu} \Rightarrow t = \ln \frac{n\varepsilon + \mu}{\mu}$$

$$\text{当 } t = \ln \frac{n\varepsilon + \mu}{\mu} \text{ 时 } F(t) = \varepsilon n + \mu - (n\varepsilon + \mu) \ln \frac{n\varepsilon + \mu}{\mu} - \mu$$

$$= \varepsilon n - (n\varepsilon + \mu) \ln \frac{n\varepsilon + \mu}{\mu}$$

$$\Rightarrow \Pr[\frac{1}{n} \sum_{i=1}^n (X_i - E[X_i]) \geq \varepsilon] \leq \exp[\varepsilon n - (n\varepsilon + \mu) \ln \frac{n\varepsilon + \mu}{\mu}]$$



同理可得: $P[\frac{1}{n} \sum_{i=1}^n (X_i - E[X_i]) \leq -\varepsilon] \leq \exp[-\varepsilon n + (\varepsilon n - u) \ln \frac{\varepsilon n + u}{u}]$

b.6.

$$\text{令 } Y_i = X_i - \frac{a+b}{2}$$

$$P(X_i = a) = P(X_i = b) = \frac{1}{2} \Rightarrow P(Y_i = \frac{a-b}{2}) = P(Y_i = \frac{b-a}{2}) = \frac{1}{2}$$

$$\begin{aligned} E[e^{tY}] &= E[e^{t(X_i - \frac{a+b}{2})}] = \frac{1}{2} e^{\frac{a-b}{2}t} + \frac{1}{2} e^{\frac{b-a}{2}t} \\ &= e^{\frac{a-b}{2}t} (\frac{1}{2} e^{t^2} + \frac{1}{2} e^{-t^2}) \\ &\leq e^{\frac{(b-a)^2}{4}t^2} \end{aligned}$$

$$P[\frac{1}{n} \sum_{i=1}^n (X_i - \frac{a+b}{2}) \geq \varepsilon] = P[\frac{1}{n} \sum_{i=1}^n Y_i \geq \varepsilon]$$

$$= P[\sum_{i=1}^n Y_i \geq n\varepsilon]$$

$$\text{由Chernoff方法得 } P[\sum_{i=1}^n Y_i \geq n\varepsilon] \leq \exp(-nt\varepsilon) E[\exp(\sum_{i=1}^n tY_i)]$$

$$= \exp(-nt\varepsilon) \prod_{i=1}^n E[\exp(tY_i)]$$

$$\leq \exp(-nt\varepsilon + \frac{(b-a)^2}{4}nt^2)$$

$$\text{当 } t = \frac{2\varepsilon}{b-a} \text{ 时右式取最小值 } \exp(-\frac{2\varepsilon^2 n}{b-a} + \frac{n\varepsilon^2}{b-a}) = \exp(-\frac{n\varepsilon^2}{b-a})$$

$$\Rightarrow P[\frac{1}{n} \sum_{i=1}^n (X_i - \frac{a+b}{2}) \geq \varepsilon] \leq \exp(-\frac{n\varepsilon^2}{b-a})$$

$$\text{同理可得 } P[\frac{1}{n} \sum_{i=1}^n (X_i - \frac{a+b}{2}) \leq -\varepsilon] \leq \exp(-\frac{n\varepsilon^2}{b-a})$$

b.7

$$\text{令 } \bar{X} = \sum_{i=1}^n X_i, u = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n p_i$$

$$P[\sum_{i=1}^n X_i \geq (1+\varepsilon) \sum_{i=1}^n p_i]$$

$$= P[\bar{X} \geq (1+\varepsilon)u] \text{ 由Chernoff方法}$$

$$= P[e^{t\bar{X}} \geq e^{t(1+\varepsilon)u}]$$

$$\leq e^{-tu - t\varepsilon u} E[e^{t\bar{X}}]$$

利用随机变量的独立性以及 $1+x \leq e^x$ 有

$$E[e^{t\bar{X}}] = \exp(u(e^t - 1))$$

$$\Rightarrow P[\bar{X} \geq (1+\varepsilon)u] \leq e^{-tu - t\varepsilon u} \times e^{u(e^t - 1)} = e^{-u(t + t\varepsilon - e^t + 1)}$$

$$\text{令 } f(t) = t + t\varepsilon - e^t + 1, f'(t) = 1 + \varepsilon - e^t$$

$$\text{令 } f'(t) = 0 \text{ 得 } t = \ln(1+\varepsilon)$$

$$\begin{aligned} \Rightarrow P[\bar{X} \geq (1+\varepsilon)u] &\leq \exp(-u \ln(1+\varepsilon) (1+\varepsilon) - 1 + \varepsilon + 1) \\ &= \exp(-u \ln(1+\varepsilon) (1+\varepsilon) - \varepsilon) = \left(\frac{e^{-\varepsilon}}{(1+\varepsilon)^{1+\varepsilon}} \right)^u \leq e^{-\frac{u\varepsilon^2}{3}} \end{aligned}$$

$$\text{此时 } \varepsilon \in (0, 1) \text{ 令 } g(\varepsilon) = \ln\left(\frac{e^{-\varepsilon}}{(1+\varepsilon)^{1+\varepsilon}}\right) + \frac{\varepsilon}{3}$$

$$= -\ln(1+\varepsilon) + \frac{\varepsilon}{3}$$



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$$\text{令 } g'(\varepsilon) = 0 \text{ 得 } \varepsilon = e^{-\frac{2}{3}} - 1$$

$$g(\varepsilon) \text{ 在 } (0, 1) \downarrow$$

$$g(\varepsilon) < g(0) < 0$$

$$\text{即证右式} \leq \frac{\varepsilon^2 u}{2}$$

\Rightarrow 结论得证.

6.8.

$$\text{令 } Y = \frac{X-a}{b-a} \text{ 即 } X = (b-a)Y + a, E(Y) = u.$$

$$E[e^{tx}] = E[e^{t((b-a)Y+a)}] = e^{at} E[e^{(b-a)tY}]$$

由凸函数性质得

$$e^{(b-a)tY} = e^{t((b-a)Y + (b-a)(1-Y))} \leq Y e^{(b-a)t} + (1-Y) e^0$$

$$\text{两边同时取期望有 } E(e^{tY}) \leq 1 - u + u e^{(b-a)t} \\ = \exp(\ln(1-u + u e^{(b-a)t}))$$

$$\text{令 } f(t) = \ln(1-u + u e^{(b-a)t})$$

$$f'(t) = \frac{u(b-a)e^{(b-a)t}}{1-u + u e^{(b-a)t}} \Rightarrow f'(0) = u(b-a)$$

$$f''(t) = \frac{u(b-a)^2 e^{(b-a)t}}{(1-u + u e^{(b-a)t})^2} - \frac{u^2(b-a)^2 e^{2(b-a)t}}{(1-u + u e^{(b-a)t})^3} \leq -\frac{1}{4}(b-a)^2$$

由Taylor中值定理得.

$$f(t) = f(0) + t f'(0) + \frac{f''(\xi)}{2} t^2 \leq (b-a)t u + \frac{(b-a)^2 t^2}{8}$$

$$\Rightarrow E(e^{tx}) \leq \exp(ut + \frac{t^2(b-a)^2}{8}).$$

