

$$5.11 \quad VarS = E(S - \bar{S})^2 \geq 0$$

$$\text{而 } VarS = ES^2 - (ES)^2 \Rightarrow ES \leq \sqrt{ES^2} = \sigma$$

若  $ES^2 > 0$  此时假设  $ES = \sigma$

$$\text{记 } \mathcal{A} = \{X | f(X) > 0\}$$

$$\text{则 } VarS = 0 \text{ 即 } E(S - ES)^2 = 0$$

$$S - ES = 0 \text{ a.e in } \mathcal{A} \Rightarrow S^2 = (ES)^2 \text{ a.e in } \mathcal{A}$$

由于每个随机变量是独立的, 因此  $VarX_n = 0 = \sigma^2$  与条件矛盾

因此  $ES > \sigma$

$$5.14 \quad (a) \text{ 记 } \tilde{U}_i = \sum_{j=1}^n a_{ij} X_j = \sum_{j=1}^n a_{ij} \sigma_j Z_j + \sum_{j=1}^n a_{ij} \mu_j$$

$$\stackrel{\text{记作}}{=} \sum_{j=1}^n \tilde{a}_{ij} Z_j + \tilde{\mu}_j \stackrel{\text{记作}}{=} U_i + \tilde{\mu}_i$$

$$\text{采用同样的记号 } \tilde{V}_r = \sum_{j=1}^n \tilde{b}_{rj} Z_j + \tilde{\mu}_r = V_r + \tilde{\mu}_r$$

$$Cov(\tilde{U}_i, \tilde{V}_r) = 0$$

$$\Rightarrow Cov(U_i, V_r) = 0$$

$$\Rightarrow U_i, V_r \text{ 独立}$$

$$\Rightarrow U_i + \tilde{\mu}_i, V_r + \tilde{\mu}_r \text{ 独立}$$

$$\Rightarrow \tilde{U}_i, \tilde{V}_r \text{ 独立}$$

$$(b) Cov(\tilde{U}_i, \tilde{V}_r) = Cov(U_i, V_r)$$

$$\because EU_i = 0, EV_r = 0, EZ_j^2 = 1$$

$$\begin{aligned}\therefore \text{Cov}(\tilde{U}_i, \tilde{V}_r) &= EU_i V_r = E \sum_{j,k} \tilde{a}_{ij} \tilde{b}_{rk} Z_j Z_k = \sum_{j=1}^n \tilde{a}_{ij} \tilde{b}_{rj} Z_j^2 = \sum_{j=1}^n \tilde{a}_{ij} \tilde{b}_{rj} \\ \therefore \text{Cov}(\tilde{U}_i, \tilde{V}_r) &= \sum_{j=1}^n a_{ij} b_{rj} \sigma_j^2\end{aligned}$$

$$\mathbf{5.15} \quad (\text{a}) \bar{X}_{n+1} = \sum_{i=1}^{n+1} \frac{X_i}{n+1} = \frac{X_{n+1} + \sum_{i=1}^n X_i}{n+1} = \frac{X_{n+1} + n\bar{X}_n}{n+1}$$

(b)

$$\begin{aligned}nS_{n+1}^2 &= \sum_{i=1}^{n+1} (X_i - \bar{X}_{n+1})^2 \\ &= \sum_{i=1}^{n+1} (X_i - \bar{X}_n + \bar{X}_n - \bar{X}_{n+1})^2 \\ &= \sum_{i=1}^{n+1} (X_i - \bar{X}_n)^2 + \sum_{i=1}^{n+1} (\bar{X}_n - \bar{X}_{n+1})^2 + \sum_{i=1}^{n+1} 2(X_i - \bar{X}_n)(\bar{X}_n - \bar{X}_{n+1}) \\ &= nS_n^2 + (X_{n+1} - \bar{X}_n)^2 - (n+1)(\bar{X}_{n+1} - \bar{X}_n)^2\end{aligned}$$

$$\text{而 } \bar{X}_{n+1} - \bar{X}_n = \frac{X_{n+1} + n\bar{X}_n}{n+1} - \bar{X}_n = \frac{X_{n+1} - \bar{X}_n}{n+1}$$

$$\text{因此 } nS_{n+1}^2 = nS_n^2 + \left(1 - \frac{1}{n+1}\right)(X_{n+1} - \bar{X}_n)^2 = nS_n^2 + \frac{n}{n+1}(X_{n+1} - \bar{X}_n)^2$$

这两个式子可以用于求与均值或方差相关的一些量关于  $n$  的递推关系。例如书中的关于 Theorem 5.3.1(c)  $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$  的证明，便是利用上式并使用数学归纳法证明的。其他还可以利用上面两式来得到关于  $\bar{X}_n, S_n^2$  均值与方差的递推关系式，从而给出均值方差通项的另一种算法。