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第六章

7. $T \sim P(\chi^2, \lambda)$

9. $T \sim N(\mu, \sigma^2)$

(1) (1). $X_1 - 2X_2 + 3X_3 \sim N(0, 6)$
 $4X_4 - 5X_5 \sim N(0, 9)$

$(\sqrt{a}(X_1 - 2X_2 + 3X_3))^2 + (\sqrt{b}(4X_4 - 5X_5))^2 \sim \chi^2(2)$

$\sqrt{a} = \frac{1}{\sqrt{6}} \therefore a = \frac{1}{6} \quad b = \frac{1}{9} \quad n=2$

(2) $\frac{\sqrt{3}d(X_1 + X_2 + X_3)}{\sqrt{2}\sqrt{\frac{X_4^2 + X_5^2}{2}}} \quad n=2 \quad X_4^2 + X_5^2 \sim \chi^2(2)$
 $\frac{X_1 + X_2 + X_3}{\sqrt{3}} \sim N(0, 1)$

$\therefore \frac{\sqrt{3}}{\sqrt{2}}c = 1 \therefore c = \frac{\sqrt{2}}{\sqrt{3}} = \frac{\sqrt{6}}{3} \quad n=2$

(3) $\frac{3d}{2} \frac{\frac{X_1^2 + X_2^2 + X_3^2}{3}}{\frac{X_4^2 + X_5^2}{2}} \quad \therefore \frac{3}{2}d = 1 \therefore d = \frac{2}{3}$
 $\chi^2(3, 2)$



第2次

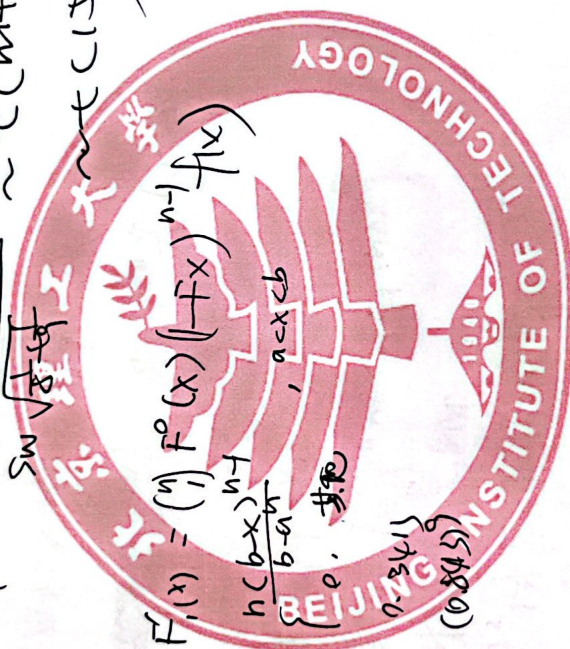
11. $S_n^2 = \frac{(n-1)S_1^2 + (n-1)S_2^2}{n+n-2} = \frac{0.5485 \times 2}{15}$

$m=8 \quad n=9$

$\bar{X} = \frac{\sum_{i=1}^m \bar{X}_i + \sum_{i=1}^n \bar{X}_i}{m+n-2} \sim t(m+n-2)$

16. $\phi_{f,1}(x) = F_{1,1}(x) = \binom{n}{1} F^0(x) (-F(x))^{n-1} f(x)$
 $= \frac{h(b-x)^{n-1}}{b-a^n}, a < x < b$
 f_0, f_1

19. (1) 0.1587
(2) (0.1587, 0.8413)





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第12

$$1.11) \mu_1 = \bar{X} = \int_0^1 x(\theta+1)x^\theta dx \quad \theta = \frac{2\bar{X}-1}{1-\bar{X}}$$

$$\pi(x|\theta) = \prod_{i=1}^n (\theta+1)x_i^\theta = (\theta+1)^n \prod_{i=1}^n x_i^\theta$$

$$\ln L(\theta) = n \ln(\theta+1) + \theta \sum_{i=1}^n \ln x_i$$

$$\frac{d \ln L(\theta)}{d\theta} = \frac{n}{\theta+1} + \sum_{i=1}^n \ln x_i$$

$$\hat{\theta} = -\frac{\sum_{i=1}^n \ln x_i}{1}$$

(2) 矩估计

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n \ln x_i$$

$$\hat{\theta}^2 = \frac{1}{n} \sum_{i=1}^n \left(\ln x_i - \frac{1}{n} \sum_{i=1}^n \ln x_i \right)^2$$

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n \ln x_i$$

$$\hat{\theta}^2 = \frac{1}{n} \sum_{i=1}^n \left(\ln x_i - \frac{1}{n} \sum_{i=1}^n \ln x_i \right)^2$$

(3) 矩估计

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n \ln x_i$$

$$(4) \text{矩估计 } \hat{\mu} = \bar{X} - \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{X})^2}$$

$$\hat{\mu} = \sqrt{\frac{1}{n} \sum_{i=1}^n (x_i - \bar{X})^2}$$

$$\hat{\mu} = \mu_0(X_1, \dots, X_n)$$

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



张华

(5) 矩估计: $\hat{\theta} = \bar{X}$.
最大似然估计: $\hat{\theta} = \bar{X}$.

2. $\hat{\theta} = \bar{X}$

$$\hat{\theta} = \sqrt{\frac{2}{k} \sum_{i=1}^k (x_i - \bar{x})^2}$$

5. 矩估计 $\hat{\theta} = 1.75$
最大似然估计 $\hat{\theta} = 1.75$

