

(1) $L(p) = P(X_1=x_1, \dots, X_n=x_n, p) = p^{\sum x_i} (1-p)^{n-\sum x_i}$ 记 X 有 p 概率取 1, $(1-p)$ 概率取 0

$$\ln L(p) = \sum x_i \ln p + (n - \sum x_i) \ln(1-p)$$

$$\frac{\partial \ln L(p)}{\partial p} = \frac{\sum x_i}{p} + \frac{n - \sum x_i}{p-1} = 0, \text{ 得: } \hat{p} = \frac{\sum x_i}{n} = \bar{x}$$

$$\therefore \Lambda(x_1, \dots, x_n) = \frac{P(x_1, \dots, x_n, \hat{p})}{P(x_1, \dots, x_n, p_0)} = \left(\frac{\bar{x}}{p_0}\right)^{\sum x_i} \left(\frac{1-\bar{x}}{1-p_0}\right)^{n-\sum x_i}$$

\therefore 拒绝域 $W =$

$$\left[\Lambda(x_1, \dots, x_n) \geq c \right], \text{ 满足 } P_{p_0}(\Lambda(x_1, \dots, x_n) \geq c) \leq \alpha, \text{ 且 } p = p_0$$

它的显著水平为 α

(2) ① 当 $H_0: \mu = \mu_0$ 时: $\theta = \mu, \sigma^2$ 的 MLE 分别为 $\mu_0, \frac{1}{n} \sum_{i=1}^n (x_i - \mu_0)^2$

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$$\therefore \Lambda(x_1, \dots, x_n) = \left(\frac{\sum (x_i - \bar{x})^2}{\sum (x_i - \mu_0)^2} \right)^{\frac{n}{2}} = \left(1 + \frac{t^2}{n-1} \right)^{-\frac{n}{2}}, \quad t = \frac{\sqrt{n}(\bar{x} - \mu_0)}{s}$$

② 当 $H_0: \sigma^2 = \sigma_0^2$ 时: $\theta = \mu, \sigma^2$ 的 MLE 分别为 $\bar{x}, \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$

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它们的拒绝域均如下: $[\Lambda(x_1, \dots, x_n) \geq c]$

对 ① 来说, 相当于 $[1+t^2 \leq d], \quad d = 1 - \alpha/2 (n-1), \quad c = \left[1 + \frac{\alpha^2}{n-1} \right]^{\frac{n}{2}}, \quad \alpha \text{ 为显著水平}$

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(3) (a) 指数分布密度函数为 $\frac{\theta^n v^{\theta n}}{(\prod x_i)^{\theta+1}} = L(\theta, v)$

$$\ln L(\theta, v) = n \ln \theta + n \theta \ln v - (\theta+1) \sum \ln x_i$$

$$\frac{\partial \ln L(\theta, v)}{\partial \theta} = \frac{n}{\theta} + n \ln v - \sum \ln x_i = 0; \quad \frac{\partial \ln L(\theta, v)}{\partial v} = \frac{n \theta}{v} > 0$$

$\Rightarrow v$ 在边界上, $\therefore \hat{v} = x_{(n)} \quad (x \geq v)$

$$\Rightarrow \theta = \frac{n}{\sum \ln \frac{x_i}{x_{(n)}}} = \frac{n}{\sum \ln \frac{x_i}{x_{(n)}}} = \frac{n}{T(x)}$$

$$(b) \text{ 由 (a): } L(\theta, v)_{\max} = \frac{\theta^n v^{\theta n}}{(\prod x_i)^{\theta+1}} = \frac{\theta^n \cdot \frac{\prod x_i^{\theta}}{e^{\theta n}}}{(\prod x_i)^{\theta+1}} = \frac{(\frac{\theta}{e})^n}{\prod x_i} =$$

$$\therefore \Lambda(x_1, \dots, x_n) = \frac{\frac{\theta^n v^{\theta n}}{(\prod x_i)^{\theta+1}}}{\frac{x_{(n)}^n}{(\prod x_i)^{\theta+1}}} = \frac{\theta^n x_{(n)}^{n(\theta-1)}}{(\prod x_i)^{\theta-1}} = \frac{(\frac{n}{T(x)})^n}{[e^{T(x)}]^{\theta-1}} = \frac{n^n}{T(x)^n \cdot e^{n-T(x)}}$$

拒绝域为 $\Lambda(x_1, \dots, x_n) \geq b$

$$\frac{\partial \Lambda}{\partial T} = \frac{-n^n}{T^n e^{n-T}} [n T^{n-1} e^{n-T} - e^{n-T} \cdot T^n]$$

$$= \frac{-n^n \cdot n^n e^{n-T} T^{n-1}}{T^{2n} e^{2n-2T}} (n-T) \quad T=n \text{ 时有极大值}$$

\therefore 拒绝域的形式为 $T \leq c$ 或 $T \geq d$

$$\text{满足 } \Lambda(T=c) = \Lambda(T=d) = b$$

$$(4) \text{ 统计量 } \chi^2 = \sum_{i=1}^k \frac{(n_i - n p_{i0})^2}{n p_{i0}} = \frac{(10-10)^2}{10} + \frac{(9-10)^2}{10} + \frac{(12-10)^2}{10} + \frac{(7-10)^2}{10} + \frac{(13-10)^2}{10} + \frac{(9-10)^2}{10}$$

$$= \frac{24}{10} = 2.4$$

$$\chi^2_{1-\alpha}(1) = \chi^2_{0.95}(1) = 11.07 \quad \chi^2 < 11.07 \text{ 不在拒绝域内}$$

\therefore 可以认为该硬币均匀

(5) Z 为品种与列对应数值 1, 2, 3

$$\therefore \bar{x} = p^2 + 4p(1-p) + 3(1-p)^2 = 3-2p \quad \text{为假设 } p \text{ 的 MLE 为 } \frac{3-\bar{x}}{2}$$

证明如下: $L(p) = (p^2)^{n_1} \times [4p(1-p)]^{n_2} \times (1-p)^{2n_3} = p^{2n_1+4n_2} (1-p)^{n_2+2n_3} \cdot 2^{n_2}$, 其中 n_i 为第 i 种的数目

$$\ln L(p) = (2n_1 + n_2) \ln p + (n_2 + 2n_3) \ln(1-p) + n_2 \ln 2$$

$$\frac{1}{2} \frac{d \ln L(p)}{dp} = \frac{(2n_1 + n_2)}{p} + \frac{(n_2 + 2n_3)}{p-1} = 0 \quad \chi^2:$$

$$(2n_1 + 2n_2 + 2n_3)p = (2n_1 + n_2)$$

$$\begin{cases} n_1 + n_2 + n_3 = n \\ n_1 + 2n_2 + 3n_3 = n\bar{x} \end{cases} \Rightarrow 2n_1 + n_2 = n(-\bar{x} + 3) \Rightarrow 2np = n(-\bar{x} + 3) \Rightarrow \boxed{p = \frac{3-\bar{x}}{2}} \text{ 为 MLE}$$

$$\bar{x} = \frac{10 + 53 \times 2 + 46 \times 3}{10 + 53 + 46} = \frac{254}{109} \approx 2.33 \quad \therefore p \approx \frac{0.33486}{2} = 0.16743, \quad p_1 = 0.112, \quad p_2 = 0.446, \quad p_3 = 0.442$$

$$\therefore \chi^2 = \frac{(10 - 12.208)^2}{12.208} + \frac{(53 - 48.614)^2}{48.614} + \frac{(46 - 48.178)^2}{48.178} = 0.39935 + 0.3937 + 0.0985$$

$$\text{而 } \chi^2_{0.95}(2) = 5.9915 \quad \chi^2 < 5.9915, \text{ 不在拒绝域内} = 0.89355$$

\therefore 数据与模型相符合

$$(6) p_{11} = \frac{66}{150} \times \frac{38}{150} = 0.1115$$

$$p_{12} = \frac{66}{150} \times \frac{34}{150} = 0.0997$$

$$p_{13} = \frac{66}{150} \times \frac{78}{150} = 0.2288$$

$$p_{21} = \frac{84}{150} \times \frac{38}{150} = 0.14187$$

$$p_{22} = \frac{84}{150} \times \frac{34}{150} = 0.12693$$

$$p_{23} = \frac{84}{150} \times \frac{78}{150} = 0.2912$$

$$\therefore \chi^2 = \sum_{j=1}^3 \sum_{i=1}^2 \frac{(n_{ij} - n p_{ij})^2}{n p_{ij}}$$

$$= \frac{(18 - 16.72)^2}{16.72} + \frac{(15 - 14.96)^2}{14.96} + \frac{(33 - 34.32)^2}{34.32} + \frac{(20 - 21.28)^2}{21.28} + \frac{(19 - 19.04)^2}{19.04} + \frac{(174 - 174.68)^2}{174.68}$$

$$= 0.09799 + 0.00017 + 0.05077 + 0.07699 + 0.00 + 0.03989$$

$$= 0.26564$$

$$\chi^2_{0.95}(5) = 11.0705 > \chi^2 \quad \therefore \text{原假设, 臂部大小与智商不相关}$$