```
I \cdot P(X=0) = \frac{e^{-\lambda} \lambda^0}{0!} = e^{-\lambda} = \frac{8!}{2500} \cdot \lambda = \ln \frac{2500}{8!} = 3.43
2. E(x) = \int_{1}^{\theta} x \cdot \frac{2\theta^{2}}{(\theta^{2} - 1)x^{3}} dx = \frac{2\theta}{\theta + 1} \Rightarrow \theta = \frac{2}{2 - E(x)} - 1 = \frac{E(x)}{2 - E(x)}
                      F(x) = \lambda , \quad \hat{\lambda} = \overline{Y} = \frac{1}{5} + \frac{3}{5} + \frac{3}{25} + \frac{2}{25} = 1
L(\lambda) = \prod_{i=1}^{50} P(x_{i}; \lambda) = (e^{-\lambda})^{17} \cdot (e^{-\lambda} \cdot \lambda^{1})^{20} \cdot (\frac{e^{-\lambda}}{2})^{10} \cdot (\frac{e^{-\lambda}}{3})^{2} \cdot (\frac{e^{
                   \overline{\ln \mathsf{L}(\lambda)} = -17\lambda - 20\lambda + 20\ln \lambda + 10(-\lambda + 2\ln \lambda - \ln 2) + 2(-\lambda + 3\ln \lambda - \ln 6)
                                                                                                + (-)+ 4/n) - 1724)
          \frac{d \ln L(\lambda)}{d \lambda} = -17 - 20 + \frac{20}{\lambda} - 10 + \frac{20}{\lambda} - 2 + \frac{6}{\lambda} - 1 + \frac{4}{\lambda} = -50 + \frac{50}{\lambda} = 0
      4.E(X) = 20(1-0) + 20^2 + 3(1-20) = -40 + 3
                                      \theta = -\frac{E(x)-3}{4}, \delta = -\frac{x-3}{4}, x = 4+4+\frac{3}{2}=2
                                         \frac{(\theta) = \frac{1}{11} P(x_{1}, \theta) = (1-2\theta)^{4} \cdot [2\theta(1-\theta)]^{\frac{1}{2}} \cdot \theta^{2} \cdot \theta^{2}}{|nL|} = \frac{1}{1-2\theta} + \frac{4}{2\theta} - \frac{2}{1-\theta} + \frac{4}{2\theta} = 0 \quad |\theta| = \frac{7-13}{12}
         5.L(\lambda) = \frac{10}{10}e^{-\lambda(1050+...+1150)} = \lambda^{10}e^{-\lambda\cdot 11680}
                                 dInl 1 1 1680 = 0 > 1 = 0.000856
```

4 Inc1-28) 102. (20)2.



6.(1) E(X) = kp,  $\hat{p} = \frac{X}{k}$  $\frac{L(\lambda) = \prod_{i=1}^{n} p(x_i, p) = C_{x_i}^{x_i} p^{x_i} (1-p)^{(k-x_i)} \cdot C_{x_i}^{x_2} p^{x_2} (1-p)^{(k-x_2)} \cdot \dots \cdot C_{x_n}^{x_n} p^{x_n} (1-p)^{k_x}}{d \ln L} = \frac{1}{N} \sum_{i=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{j=1}^{n} \sum_{k=1}^{n} \sum_{$ (5)  $E(X) = \int_{\infty}^{\infty} z \cdot \frac{e^{-\frac{\pi}{2}}}{2} e^{-\frac{\pi}{2}} dx = 20$ ,  $e = \frac{\sqrt{2}}{2}$  $L(\theta) = \prod_{i=1}^{n} f(x_{i,\theta}) = \frac{\prod_{x_i}}{\theta^{2n}} e^{-\frac{\sum_{x_i}}{\theta}}$  $\frac{d\ln L}{d\theta} = 0 \Rightarrow -2n \frac{1}{\theta} + \frac{\sum x_i}{\partial x_i} = 0 \Rightarrow \hat{\theta} = \frac{x_i}{2}$ (3)  $F(x) = \int_0^1 x \sqrt{\theta} x^{1/\theta-1} dx = \int_0^1 \sqrt{\theta} x^{1/\theta} dx = \frac{\sqrt{\theta}}{\sqrt{\theta}+1}, \hat{\theta} = (\frac{x}{1-x})$  $L(\theta) = \frac{\pi}{1 - 1} f(x_{i}, \theta) = (\overline{I}\theta)^{n} \cdot (\pi x_{i})^{\overline{I}\theta - 1}$  $\frac{d\ln L(\theta)}{d\theta} = \frac{1}{2} \cdot \frac{1}{\theta} = \frac{1}{210} \cdot \frac{1$ (4)  $E(x) = \int_{\Omega}^{\infty} \frac{2\theta^2}{x^3} dx = 2\theta$ ,  $\hat{\theta} = \frac{\overline{X}}{x^3}$  $\frac{L(\theta) = \frac{\pi}{10} f(X_i, \theta) = \frac{(2\theta)^n}{(2\theta)^n}, \text{ if } \theta \leq \min\{X_1, \dots, X_n\}$   $\frac{d\ln L}{d\theta} = \frac{4\theta}{2\theta^2} = \frac{2n}{\theta}, \frac{\pi}{10} = \min\{X_1, \dots, X_n\}$  $= \frac{1}{1} \frac{1}{1} \frac{1}{1} = \frac{1}{1} \frac{1}{1} = \frac{1}{1} \frac{1}{1} \frac{1}{1} \frac{1}{1} = \frac{1}{1} \frac{1}{1} \frac{1}{1} \frac{1}{1} = \frac{1}{1} \frac{1}{1$ 

7. X=eInx · 全Inx为y  $E(x) = E(e_{\lambda}) = \int_{-\infty}^{\infty} e_{\lambda} \cdot t^{\lambda}(\lambda) d\lambda = e_{M} + f_{\alpha}$ E(x2)= E(e2x)= [= e2x fx(x) dy = e2m+202  $D(X) = E(X_5) - [E(X)]_5 = 6_{5W+5a_5} - 6_{5W+a_5} = 6_{5W+a_5} (6_{a_5})$  $=E^{2}(X)(e^{\alpha_{2}}-1)$ 由 Y~ N(h, 52) ⇒  $\hat{\beta} = \hat{\pi} \stackrel{\stackrel{\leftarrow}{\xi}}{\xi} \ln X_{\bar{1}}$   $\hat{\sigma}^2 = \hat{\pi} \stackrel{\stackrel{\leftarrow}{\xi}}{\xi} (\ln X_{\bar{1}} - \hat{\mu})^2$