

Statistics & Inferencing Homework #02

(20)

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03) (a) Estimate λ $N = 400$

$$C_1 : \bar{X}_1 = [213(0) + 128(1) + \dots + 1(5)] / 400 = 0.6825$$

$$C_2 : \bar{X}_2 = 1.3225$$

$$C_3 : \bar{X}_3 = 1.8$$

$$C_4 : \bar{X}_4 = 4.68$$

$$\hat{\lambda}_1 = 0.6825$$

$$\hat{\lambda}_2 = 1.3225$$

$$\hat{\lambda}_3 = 1.8$$

$$\hat{\lambda}_4 = 4.68$$

(b) 95% CI

$$\hat{\lambda} \pm z_{\alpha/2} \sqrt{\hat{\lambda}/N}$$

$$\alpha = 0.05$$

$$\alpha/2 = 0.025$$

$$z_{\alpha/2} = 1.96$$

$$C_1 : 0.6825 \pm 1.96 \sqrt{0.6825/400} \Rightarrow 0.6825 \pm 0.08096$$

$$C_2 : \Rightarrow 1.3225 \pm 0.1127$$

$$C_3 : \Rightarrow 1.8 \pm 0.131$$

$$C_4 : \Rightarrow 4.68 \pm 0.212$$

(c) $P(X=k) = \lambda^k / k! \cdot e^{-\lambda}$, $k \in N_0$

$$E_{k,i} = P(X=k) \cdot N = \lambda^k / k! \cdot e^{-\lambda_i} \cdot N$$

$i \in \{1, 2, 3, 4\}$, $N = 400$

$$E_{0,1} = 0.6825^0 / 0! \cdot e^{-0.6825} \cdot 400 = 202.14$$

The table attached below summarizes the expected & observed outcomes.

n	0	1	2	3	4	5	6	7	8	9	10	11	12
Observed 1	213	128	37	18	3	1	0	0	0	0	0	0	0
Expected 1	202.14	137.96	47.08	10.71	1.83	0.25	0.03	0	0	0	0	0	0
Observed 2	103	143	98	42	8	4	2	0	0	0	0	0	0
Expected 2	106.59	140.96	93.21	41.09	13.59	3.59	0.79	0.15	0.02	0	0	0	0
Observed 3	75	103	121	54	30	13	2	1	0	1	0	0	0
Expected 3	66.12	119.02	107.11	64.27	28.92	10.41	3.12	0.8	0.18	0.04	0.01	0	0
Observed 4	0	20	43	53	86	70	54	37	18	10	5	2	2
Expected 4	3.71	17.37	40.65	63.41	74.19	69.44	54.16	36.21	21.18	11.02	5.16	2.19	0.86

$$05) X \rightarrow \text{DRV} \quad P(X=1) = \theta \quad P(X=2) = 1-\theta \quad x_1=1, x_2=2, x_3=2$$

$$E(X) = \sum k \cdot P(X=k) = 1 \cdot \theta + 2(1-\theta) = 2-\theta$$

$$\hat{\theta} = 2 - E[X]$$

$$\bar{x} = (1+2+2)/3 = 5/3$$

$$\hat{\theta} = 2 - 5/3 = 1/3$$

$$\text{lik}(\theta) = f(x_1, \dots, x_n | \theta) = f(x_1 | \theta) \dots f(x_n | \theta)$$

$$= \theta \cdot (1-\theta)^2$$

$$\log(\text{lik}(\theta)) = \ln(\theta) + 2 \cdot \ln(1-\theta)$$

$$l'(\theta) = 1/\theta - 2/(1-\theta) = (1-3\theta)/(\theta(1-\theta))$$

$$l'(\theta) = 0 \Rightarrow \theta = 1/3$$

$$l''(\theta) = (-3\theta^2 + 2\theta - 1)/[\theta \cdot (1-\theta)]^2$$

$$\hat{\theta} = 1/3$$

$$f_{\theta|x}(\theta|x) = [f_{x|\theta}(x|\theta) \cdot f_{\theta}(\theta)] / f_x(x)$$

$$f_{x|\theta}(x|\theta) = \theta(1-\theta)^2$$

$$f_x(x) = \int_0^1 f_{x|\theta}(x|\theta) d\theta$$

$$f_{\theta}(\theta) = 1, \theta \in [0,1]$$

$$\text{Then } \int_0^1 x^{a-1} \cdot (1-x)^{b-1} dx = \Gamma(a) \cdot \Gamma(b) / \Gamma(a+b)$$

$$\Gamma(a) = (a-1)!$$

$$f_x(x) = \int_0^1 \theta \cdot (1-\theta)^2 \cdot d\theta = 1/12$$

$$f_{\theta|x}(\theta|x) = 12\theta(1-\theta)^2$$

$$\hat{\theta} = 1/3 \quad \text{lik}(\theta) = \theta(1-\theta)^2 \quad \hat{\theta} = 1/3$$

$$f_{\theta|x}(\theta|x) = 12\theta(1-\theta)^2$$

$$07) E(X) = 1/p \quad p = 1/E(X) \quad \hat{p} = 1/\bar{x}$$

$$\text{lik}(p) = p^n (1-p)^{\sum x_i - n}$$

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

$$l'(p) = \frac{n}{p} - (\sum x_i - n) \frac{1}{1-p}$$

$$l'(p) = 0 \quad p = 1/\bar{x}$$

$$l''(p) = -\frac{n}{p^2} - (\sum x_i - n) \left(\frac{1}{1-p}\right)^2 \quad \tilde{p} = 1/\bar{x}$$

$$\text{Var}(\tilde{p}) \approx 1/n \cdot I(p) \quad I(p) = E\left[\left(\frac{\partial}{\partial p} \ln f(x|p)\right)^2\right]$$

$$\text{Var}(\tilde{p}) = -1/E(l''(p))$$

$$E(l''(p)) = -n/p^2(1-p)$$

$$\text{Var}(\tilde{p}) \approx p^2(1-p)/n$$

$$f_{x|p}(x|p) = p^n (1-p)^{\sum x_i - n} = \int_0^1 f_{x|p}(x|p) dx$$

$$\int_0^1 x^{a-1} (1-x)^{b-1} \cdot dx$$

$$E(Y) = a/b + b$$

$$E(p|x) = n/\sum x_i$$

$$\hat{p} = 1/\bar{x}$$

$$\tilde{p} = 1/\bar{x}$$

$$\text{Var}(\tilde{p}) = p^2(1-p)/n$$

Q13) $\hat{\alpha} = 3\bar{X}$ $E(X_i) = \alpha/3$

$E(\hat{\alpha}) = \frac{3}{n} \sum E(X_i) = \frac{3}{n} \sum \frac{\alpha}{3} = \frac{3}{n} \cdot n \cdot \frac{\alpha}{3} = \alpha$

$Var(X_i) = E(X_i^2) - [E(X_i)]^2$

$f(x) = \frac{1+\alpha x}{2}$, $x \in [-1, 1]$

$E(X_i^2) = \int_{-1}^1 x^2 \cdot f(x) dx = 1/3$

$Var(X_i) = 1/3 - (\alpha/3)^2 = 3 - \alpha^2/9$

$Var(\hat{\alpha}) = 3 - \alpha^2/n$

$\hat{\alpha} - E(\hat{\alpha}) / \sqrt{Var(\hat{\alpha})} \sim N(0, 1)$

$\hat{\alpha} \sim N(\alpha, 3 - \alpha^2/n)$

$P(|\hat{\alpha}| > 0.5) = 2 - 2\Phi(0.5 / \sqrt{3 - \alpha^2/n}) = 0.498$

Q16) $E(X) = \int x \cdot f(x) dx = \int \frac{1}{2\sigma} x e^{-\frac{|x|}{\sigma}} dx = 0$

$E(X^2) = \int x^2 f(x) dx = \int \frac{1}{2\sigma} x^2 e^{-\frac{|x|}{\sigma}} dx = \frac{1}{\sigma} \int x^2 e^{-\frac{|x|}{\sigma}} dx$
 $= 2\sigma^2$

$\sigma^2 = \frac{1}{2} E(X^2)$ $\sigma = \sqrt{\frac{1}{2} E(X^2)}$

$\hat{\sigma} = \sqrt{\frac{1}{2n} \sum X_i^2}$

$\ell(\hat{\sigma}) = (\frac{1}{2\sigma})^n \cdot e^{-\frac{|x_1|}{\sigma} - \dots - \frac{|x_n|}{\sigma}}$

$\ell(\hat{\sigma}) = n \ln(2\sigma) - \frac{|x_1|}{\sigma} - \dots - \frac{|x_n|}{\sigma}$

$\ell'(\hat{\sigma}) = 0 \Rightarrow \hat{\sigma} = \frac{1}{n} \sum |x_i|$

$\tilde{\sigma} = \frac{1}{n} \sum |x_i|$

$Var(\tilde{\sigma}) \approx \frac{1}{n I(\hat{\sigma})}$ $Var(\tilde{\sigma}) = -1/E(\ell''(\hat{\sigma}))$

$\ell''(\hat{\sigma}) = \frac{n}{\sigma^2} - 2 \frac{|x_1| + \dots + |x_n|}{\sigma^3}$

$E(\ell''(\hat{\sigma})) = \frac{n}{\sigma^2} - 2 E(|x_1| + \dots + |x_n|) / \sigma^3$

$E(|x|) = \sigma \Gamma(2) = \sigma$ $Var(\tilde{\sigma}) \approx \sigma^2/n$

$\alpha(\hat{\sigma}) = -1/\sigma$ $T(x) = |x|$ $\ell(\hat{\sigma}) = n \ln(\frac{1}{2\sigma})$, $S(n) = 0$

$T = \sum T(X_i) = \sum |x_i|$

Q17) $E(X^2) = Var(X) + [E(X)]^2$ $E(X^2) = \alpha+1/2(2\alpha+1)$

$\alpha = \frac{2E(X^2)-1}{1-4E(X^2)}$

$\hat{\alpha} = \frac{[\frac{2}{n} \sum x_i^2 - 1]}{[1 - \frac{4}{n} \sum x_i^2]}$

$\ell(\alpha) = n \ln(\frac{\Gamma(2\alpha)}{\Gamma(\alpha)^2}) + (\alpha-1) \sum \ln(x_i(1-x_i))$

$\ell'(\alpha) = 0 = -\frac{1}{n} \sum \ln(x_i(1-x_i))$

$Var(\hat{\alpha}) \approx \frac{1}{n I(\hat{\alpha})}$

$\frac{2n \frac{\Gamma'(\alpha) \Gamma(2\alpha) - \Gamma'(\alpha)^2}{\Gamma(\alpha)^2}}{\Gamma(2\alpha)^2} - \frac{4n \frac{\Gamma''(2\alpha) \Gamma(2\alpha) - \Gamma'(2\alpha)^2}{\Gamma(2\alpha)^2}}{\Gamma(2\alpha)^2}$

$T = \sum T(X_i) = \sum \ln(x_i \cdot (1-x_i))$

$$Q19) f(n) = \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-(n-\mu)^2/2\sigma^2}$$

$$l(\theta) = \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{(x_1-\mu)^2 + \dots + (x_n-\mu)^2}{2\sigma^2}}$$

$$l(\sigma) = n \ln\left(\frac{1}{\sigma\sqrt{2\pi}}\right) - \frac{1}{2\sigma^2} \sum (x_i - \mu)^2$$

$$l'(\sigma) = -\frac{n}{\sigma} + \frac{1}{\sigma^3} \sum (x_i - \mu)^2$$

$$l'(\sigma) = 0 \Rightarrow \hat{\sigma} = \sqrt{\frac{1}{n} \sum (x_i - \mu)^2}$$

$$l(\mu) = n \ln\left(\frac{1}{\sigma\sqrt{2\pi}}\right) - \frac{1}{2\sigma^2} \sum (x_i - \mu)^2$$

$$l'(\mu) = 0 \Rightarrow \mu = \frac{1}{n} \sum x_i$$

$$\hat{\mu} = \frac{1}{n} \sum x_i$$

$$I(\mu) = \frac{1}{\sigma^2} \Rightarrow \frac{1}{n I(\mu)} = \frac{\sigma^2}{n}$$

$$Q21) E(X) = \theta + 1 \Rightarrow \theta = E(X) - 1 \quad \hat{\theta} = \bar{X} - 1$$

$$l(\theta) = n\theta - \sum x_i$$

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

$$T = X_1 = \min X_i \Rightarrow T = X_1 \rightarrow \text{sufficient.}$$

$$Q32) \mu = \bar{X} = \frac{1}{16} \sum x_i = 3.36109$$

$$\hat{\sigma}^2 = \frac{16-1}{16} S^2 = \frac{1}{16} \sum (x_i - \mu)^2 = 3.2045$$

$$N(3.36109, 3.2045)$$

$$\mu: CI: \bar{X} \pm t_{n-1, \alpha/2} \cdot \frac{S}{\sqrt{n}} \quad S = \sqrt{\frac{1}{16-1} \sum (x_i - \bar{X})^2} = 1.8488$$

$$90\% CI: 3.6109 \pm 0.8102$$

$$95\% CI: 3.6109 \pm 0.9849$$

$$99\% CI: 3.6109 \pm 1.3621$$

$$\sigma^2: 90\% CI: [2.0509, 7.0623]$$

$$95\% CI: [1.8651, 8.1904]$$

$$99\% CI: [1.05632, 11.1461]$$

$$\Rightarrow \sigma = \sqrt{\sigma^2} \Rightarrow 90\% CI: [1.4321, 2.6575]$$

$$95\% CI: [1.3657, 2.8619]$$

$$99\% CI: [1.2503, 3.3386]$$

$$m = 4n$$

$$Q47) E(X) = \int x f(x|x_0, \theta) = \int \theta \cdot x_0^\theta \cdot x^{\theta-1} \cdot dx$$

$$E(X) = \theta x_0^\theta \cdot \frac{x_0^{\theta+1}}{\theta+1} = \frac{\theta}{\theta+1} \cdot x_0$$

$$\Rightarrow \theta = \frac{E(X)}{E(X) - x_0} \quad \hat{\theta} = \frac{\bar{x}}{\bar{x} - x_0}$$

$$l(\theta) = n \ln(\theta) + n\theta \ln(x_0) - (\theta+1) \sum \ln(x_i)$$

$$l'(\theta) = \frac{n}{\theta} + n \ln(x_0) - \sum \ln(x_i)$$

$$l'(\theta) = 0 \Rightarrow \theta = \frac{1}{\frac{1}{n} \sum \ln(x_i) - \ln(x_0)}$$

$$l''(\theta) = -\frac{n}{\theta^2} < 0 \quad \tilde{\theta} = \frac{1}{\frac{1}{n} \sum \ln(x_i) - \ln(x_0)}$$

$$\text{Var}(\tilde{\theta}) \approx \frac{1}{n I(\theta)} \quad E(l''(\theta)) = -n/\theta^2$$

$$\text{Var}(\tilde{\theta}) \approx \theta^2/n$$

$$T = \sum \ln(x_i) \text{ is sufficient}$$

$$Q50) E(X) = \theta \sqrt{\pi/2} \Rightarrow \theta = E(X) \sqrt{2/\pi} \Rightarrow \hat{\theta} = \bar{x} \sqrt{2/\pi}$$

$$l(\theta) = \sum \ln(x_i) - 2n \ln(\theta) - \frac{1}{2\theta^2} \sum x_i^2$$

$$l'(\theta) = -\frac{2n}{\theta} + \frac{1}{\theta^3} \sum x_i^2$$

$$l'(\theta) = 0 \Rightarrow \theta = \sqrt{\frac{1}{2n} \sum x_i^2}$$

$$\tilde{\theta} = \sqrt{\frac{1}{2n} \sum x_i^2}$$

$$\text{Var}(\tilde{\theta}) = -1/E(l''(\theta))$$

$$l''(\theta) = \frac{2n}{\theta^2} - \frac{3}{\theta^4} \sum x_i^2$$

$$E(x^2) = 2\theta^2 \Rightarrow \text{Var}(\tilde{\theta}) \approx \theta^2/4n$$

$$Q52) E(X) = \frac{(\theta+1)(\theta+2)}{(\theta+1)(\theta+2)}$$

$$\theta = (2E(X)-1)/(1-E(X))$$

$$\hat{\theta} = \frac{2\bar{x}-1}{1-\bar{x}}$$

$$l(\theta) = n \ln(\theta+1) + \theta \sum \ln(x_i)$$

$$l'(\theta) = \frac{n}{\theta+1} + \sum \ln(x_i) \Rightarrow l'(\theta) = 0$$

$$\Rightarrow \theta = -(1 + \frac{1}{n} \sum \ln(x_i))$$

$$l''(\theta) = -\frac{n}{(\theta+1)^2} < 0$$

$$\tilde{\theta} = -(1 + \frac{1}{n} \sum \ln(x_i))$$

$$\text{Var}(\tilde{\theta}) = (\theta+1)^2/n$$

$$T = \sum \ln(x_i)$$

$$Q53) E(X) = \theta/2 \Rightarrow \theta = 2E(X)$$

$$\hat{\theta} = 2\bar{X} \quad E(\hat{\theta}) = \theta$$

$$\text{Var}(\hat{\theta}) = 4/n \text{Var}(X_1)$$

$$\text{Var}(X_1) = E(X_1^2) - [E(X_1)]^2$$

$$= \theta^2/12 \Rightarrow \text{Var}(\hat{\theta}) = \theta^2/3n$$

$$Lik(\theta) = 1/\theta^n \quad \hat{\theta} = \max X_i$$

$$F_{X_n}(x) = x/\theta$$

$$f_X(x) = n \cdot x^{n-1}/\theta^n$$

$$E(X_n) = \frac{n}{n+1}\theta$$

$$E(X_n^2) = \frac{n}{n+2}\theta^2$$

$$\text{Var}(X_n) = \frac{n\theta^2}{(n+1)^2(n+2)}$$

$$\text{Bias} = E(\hat{\theta}) - \theta = -\theta/(n+1)$$

$$\text{MSE}(\hat{\theta}) = \text{Var}(\hat{\theta}) + \text{Bias}^2 = \frac{\theta^2}{3n} + \theta^2 = \theta^2/3n$$

$$\text{MSE}(\tilde{\theta}) = \text{Var}(\tilde{\theta}) + \text{Bias}^2 = \frac{n\theta^2}{(n+1)^2(n+2)} + \left(-\frac{\theta}{n+1}\right)^2$$

$$= \frac{2\theta^2}{(n+1)(n+2)}$$

$$E(\tilde{\theta}) = \frac{n}{n+1}\theta \Rightarrow \theta = \frac{n+1}{n}E(\tilde{\theta}) \Rightarrow \theta = E\left(\frac{n+1}{n}\tilde{\theta}\right)$$

$$\tilde{\theta}_2 = \frac{n+1}{n}\tilde{\theta}$$

$$Q57) S^2 = \frac{1}{n-1} \sum (X_i - \bar{X})^2 = \frac{1}{n-1} (\sum X_i^2 - n\bar{X}^2)$$

$$\text{Var}(X_i) = E(X_i^2) - [E(X_i)]^2 \quad E[X_i^2] = \sigma^2 + \mu^2$$

$$E(\bar{X}^2) = \text{Var}(\bar{X}) + [E(\bar{X})]^2$$

$$\text{Var}(\bar{X}) = \sigma^2/n \quad E(\bar{X}) = \mu \quad E(\bar{X}^2) = \frac{\sigma^2}{n} + \mu^2$$

$$E(S^2) = \sigma^2 \quad \hat{\sigma}^2 = \frac{n-1}{n} S^2$$

$$E(\hat{\sigma}^2) = E\left(\frac{n-1}{n} S^2\right) = \frac{n-1}{n} E(S^2) = \frac{n-1}{n} \sigma^2 \neq \sigma^2$$

$\hat{\sigma}^2$ is biased for σ^2

$$\text{MSE}(\hat{\sigma}^2) = \text{Var}(\hat{\sigma}^2) + \text{Bias}^2(\hat{\sigma}^2)$$

$$Y = \frac{(n-1)S^2}{\sigma^2}$$

$$\text{Var}(Y) = 2(n-1)$$

$$\text{Var}(S^2) = \frac{2\sigma^4}{n-1}$$

$$\text{MSE}(S^2) = \frac{2\sigma^4}{n-1} + 0 = \frac{2\sigma^4}{n-1}$$

$$\text{Var}(\hat{\sigma}^2) = \frac{2(n-1)\sigma^4}{n^2}$$

$$\text{Bias}(\hat{\sigma}^2) = E(\hat{\sigma}^2) - \sigma^2 = -\frac{1}{n}\sigma^2$$

$$\text{MSE}(\hat{\sigma}^2) = (2n-1)\sigma^4/n^2$$

$$\text{MSE}(\hat{\sigma}^2) < \text{MSE}(S^2)$$

$$P = e^{Z(X_i - \bar{X})^2}$$

$$P = e^{(n-1)S^2}$$

$$\text{Var}(P) = 2e^2(n-1)\sigma^4$$

$$E[P] = e^{(n-1)\sigma^2}$$

$$\text{Bias}(P) = \sigma^2[p(n-1) - 1]$$

$$\text{MSE}(P) = \frac{2e^2(n-1)\sigma^4}{e^{2(n-1)\sigma^2}} (1 + 2e - 2ne - e^2 + n^2e^2)$$

$$f(e) = \sigma^4(1 + 2e - 2ne - e^2 + n^2e^2)$$

$$f'(e) = 0 \Rightarrow e = 1/(n+1) \quad f''(e) = 2\sigma^4(n^2-1) > 0$$

$$e = 1/(n+1)$$

$$Q60) \text{ Lik}(\tau) \propto 1/\tau^n e^{-\frac{x_1 + \dots + x_n}{\tau}}$$

$$\ell(\tau) \propto n \ln(\tau) - \frac{1}{\tau} \sum x_i$$

$$\ell'(\tau) \propto -n/\tau + 1/\tau^2 \sum x_i$$

$$\ell'(\tau) = 0 \Rightarrow \tau = \frac{1}{n} \sum x_i$$

$$\hat{\tau} = \frac{1}{n} \sum x_i = \bar{x}$$

$$S = \sum x_i \sim \Gamma(n, \frac{1}{\tau}) \quad \bar{x} = S/n$$

$$F_{\bar{x}}(n) = F_S(n \cdot n) \quad f_{\bar{x}}(x) = n \cdot \tau^n / \Gamma(n) \cdot (n \cdot x)^{n-1} e^{-\tau n x}$$

$$= \frac{(n\tau)^n}{\Gamma(n)} x^{n-1} e^{-(n\tau)x} \quad \bar{x} \sim \Gamma(n, 1/\tau)$$

$$E(x_i) = \tau \quad \text{Var}(x_i) = \tau^2$$

$$((\bar{x} - \tau)/\tau) \sqrt{n} \approx N(0, 1)$$

$$\bar{x} \approx N(\tau, \tau^2/n)$$

$$E(\bar{x}) = \tau \quad \text{Var}(\bar{x}) = \frac{1}{n^2} n \text{Var}(x_i) = \frac{1}{n} \text{Var}(x_i) = \tau^2/n$$

$$I(\tau) \propto 1/\tau^2 \Rightarrow 1/n I(\tau) = \tau^2/n$$

$$\bar{x} \approx N(\tau, \tau^2/n)$$

$$\text{Then interval} \Rightarrow [\bar{x}/(1 + z_{\alpha/2}/\sqrt{n}), \bar{x}/(1 - z_{\alpha/2}/\sqrt{n})]$$

$$\bar{x} \sim \Gamma(n, \frac{1}{\tau}) \quad P(\gamma_{n, n\tau(1-\alpha/2)} \leq \bar{x} \leq \gamma_{n, n\tau(\alpha/2)}) = 1 - \alpha,$$

$$[\gamma_{n, n\tau(1-\alpha/2)}, \gamma_{n, n\tau(\alpha/2)}]$$

$$Q68) T = \sum x_i \quad P(X = k) = \lambda^k / k! \cdot e^{-\lambda}$$

$$f_{X|T}(x|t) \propto t!$$

$$x_1! \dots x_{n-1}! \cdot (t - x_1 - \dots - x_{n-1})! \cdot n!$$

$$f_{X_1|X_1}(x_1|k) \propto \frac{\lambda^{x_2 + \dots + x_n} e^{-\lambda(x_2 + \dots + x_n)}}{x_2! \dots x_n!}$$

$$f(x_1, \dots, x_n | \lambda) \propto \lambda^{T(x_1, \dots, x_n)} e^{-\lambda T} \cdot \frac{1}{x_1! \dots x_n!}$$

$$Q70) f(x_1, \dots, x_n | \theta) \propto g(T(x_1, \dots, x_n), \theta) \cdot h(x_1, \dots, x_n)$$

$$f(x) \propto \lambda e^{-\lambda x}$$

$$\rightarrow = \lambda^n \cdot e^{-\lambda(x_1 + \dots + x_n)}$$

$$T = \sum x_i \rightarrow \text{is sufficient.}$$

$$Q73) f(x|\theta) = \pi/\sigma^2 \cdot e^{-\pi^2 x^2/\sigma^2} \quad f(x|\theta) = \frac{\pi}{\sigma^2} e^{-\frac{\pi^2}{\sigma^2} x^2}$$

$$f(x|\theta) = e^{-\frac{1}{2\sigma^2} x^2 - \ln \sigma + \ln \pi}$$

$$T = \sum x_i^2 \text{ is sufficient.}$$