Statistics 105 Midterm Exam 1 (April 30, 2012)

(Total 4 problems and 30 points)

Score 24-Discussion (3 UID

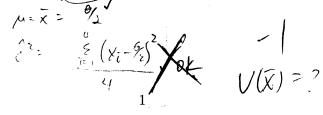
- 1. (12 points) Answer the question briefly or fill in the blanks.
 - (a) Let X and Y be two independent random variables with E(X) = 1, V(X) = 3, E(Y) = 3,and V(Y) = 4. Then

$$E(2X-3Y+1) = \underbrace{\begin{array}{c} -6 \\ }_{\zeta(z)-\zeta(z)+1} \\ V(2X-3Y+1) = \underbrace{\begin{array}{c} -6 \\ }_{\zeta(z)-\zeta(z)+1} \\ \hline \end{array}}_{\zeta(z)-\zeta(z)+1} \\ \text{(b) A random sample of } n=16 \text{ results in } \bar{x}=13 \text{ and } s=5. \text{ An estimate of the population} \\ \end{array}$$

- mean μ is ______ and its estimated standard error is ______ 1.75
- (c) Let X_1, X_2, X_3, X_4 be a random sample from a normal distribution with mean 4 and variance 9. What is sampling distribution of sample mean \bar{X} ? $(x) = \frac{1}{4} \left(\frac{1}{4} \left(\frac{1}{4} + \frac{1}{4} \right) \right) \right) \right) \right)$
- (d) When we make a normal probability plot for data from a normal population, we expect a straight but from lotton lott to by vigit
- (e) For a normal random variable X with mean 5 and variance 4, find the 85th percentile, that is, find x such that $P(X \le x) = 0.85$.

(f) Let X_1, X_2, X_3, X_4 be a random sample from a normal distribution with mean 3 and variance 16. Find $P(\bar{X} > 2)$. $\sqrt{\omega} = 4$ So = 2

- 2. (4 points) Let X_1, \ldots, X_4 be a random sample of size n=4 from a population with mean $\theta/2$ and variance θ^2 .
 - (a) Find the moment estimator of θ . 51 M2 = & (x1/2 => C2 = 12 (x1 - x2)
 - (b) What are the mean and variance of sample mean \bar{X} ?



- 3. (4 points) Clearly circle your answer for each statement.
 - (a) Which measure of location is resistant to outliers? (circle one) sample mean or sample median)
 - (b) For the normal population $N(\mu, \sigma^2)$, the sample mean \bar{X} is an unbiased estimator of μ and has the minimum variance among all unbiased estimators. (circle one True or False
 - (c) For the normal population $N(\mu, \sigma^2)$, the maximum likelihood estimator of the variance σ^2 is not unbiased. (circle one True or False
 - (d) An unbiased estimator always has smaller mean square error than a biased estimator. (circle one) **True** or **False**
- 4. (10 points) Suppose X has probability density function

$$f(x,\theta) = \left(\frac{2\theta}{\pi}\right) e^{-\theta^2 x^2/\pi}, \text{ for } x > 0$$

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where $\theta > 0$ is an unknown parameter.

$$L(6): (-10) = \frac{6^2 \tilde{\epsilon} \times i^3}{10^2 \tilde{\epsilon} \times i^3}$$

$$L(L(0)) = n \ln(\frac{i6}{10}) - \frac{6^2 \tilde{\epsilon} \times i^3}{10^2 \tilde{\epsilon} \times i^3} (x_i)^2$$

$$\frac{1}{10^2 \tilde{\epsilon} \times i^3} (x_i)^2 = 0$$

$$n = \frac{1}{10^2 \tilde{\epsilon} \times i^3} (x_i)^2$$

$$\hat{\theta} = \sqrt{\frac{1}{10^2 \tilde{\epsilon} \times i^3}} (x_i)^2$$

(b) Find the Fisher information $I(\theta)$ of parameter θ . Show your work.

$$\frac{\partial^{2}}{\partial \theta^{2}} \left(\frac{\partial u}{\partial \theta^{2}} \left(\frac{\partial u}{\partial \theta^{2}} - \frac{\partial u}{\partial \theta^{2}} - \frac{\partial u}{\partial \theta^{2}} \right) \right) = \frac{2}{\pi} \left(\frac{\partial u}{\partial \theta^{2}} + \frac{\partial u}{\partial \theta^{2}} \right)$$

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Statistics 105 Midterm Exam Solution (Spring 2012)

1. (a)
$$E(2X - 3Y + 1) = (2)(1) - (3)(3) + 1 = -6$$
, $V(2X - 3Y + 1) = (2^2)(3) + (3^2)(4) = 48$.

(b)
$$\bar{x} = 13$$
 and $s/\sqrt{n} = 5/4 = 1.25$

(c)
$$\bar{X} \sim N(\mu = 4, \sigma^2/n = 9/4)$$

(e)
$$z = 1.04$$
 so $x = \mu + z\sigma = 5 + (1.04)(2) = 7.08$.

(f)
$$\bar{X} \sim N(\mu = 3, \sigma^2/n = 16/4)$$
. $P(\bar{X} > 2) = P(Z > (2-3)/\sqrt{16/4}) = P(Z > -0.5) = 1 - 0.3085 = 0.6915$

2. (a) Let
$$\bar{X} = \theta/2$$
 and solve for θ . The moment estimator is $\hat{\theta} = 2\bar{X}$.

(b)
$$E(\bar{X}) = \theta/2$$
 and $V(\bar{X}) = \sigma^2/n = \theta^2/4$.

$$L(\theta) = \prod_{i=1}^{n} f(x_i, \theta) = \prod_{i=1}^{n} \left(\frac{2\theta}{\pi}\right) e^{-\theta^2 x_i^2/\pi} = \left(\frac{2\theta}{\pi}\right)^n e^{-\frac{\theta^2}{\pi} \sum_{i=1}^{n} x_i^2}$$

$$\ln L(\theta) = n \ln (2\theta) - n \ln \pi - \frac{\theta^2}{\pi} \sum_{i=1}^{n} x_i^2$$

$$\frac{d \ln L(\theta)}{d\theta} = \frac{n}{\theta} - \frac{2\theta}{\pi} \sum_{i=1}^{n} x_i^2$$

Let $\frac{d \ln L(\theta)}{d \theta} = 0$. We obtain MLE

$$\hat{\theta} = \sqrt{\frac{n\pi}{2\sum_{i=1}^{n} X_i^2}}$$

(b)
$$\ln f(x,\theta) = \ln (2\theta) - \ln \pi - \frac{\theta^2}{\pi} x^2$$
. We have

$$\frac{\partial \ln f(x,\theta)}{\partial \theta} = \frac{1}{\theta} - \frac{2\theta}{\pi} x^2 \text{ and } \frac{\partial^2 \ln f(x,\theta)}{\partial \theta^2} = -\frac{1}{\theta^2} - \frac{2}{\pi} x^2.$$

$$I(\theta) = -E(\frac{\partial^2 \ln f(X,\theta)}{\partial \theta^2}) = -E(-\frac{1}{\theta^2} - \frac{2}{\pi} X^2) = \frac{1}{\theta^2} + \frac{2}{\pi} E(X^2) = \frac{1}{\theta^2} + \frac{2}{\pi} \frac{1}{\theta^2} \frac{\pi}{2} = \frac{2}{\theta^2}$$
because $E(X^2) = V(X) + (E(X))^2 = \frac{1}{\theta^2} (\frac{\pi}{2} - 1) + \left(\frac{1}{\theta}\right)^2 = \frac{1}{\theta^2} \frac{\pi}{2}.$

Summary of the scores: n=80, min=8, q1=21, q2=24, q3=27, max=30, mean=23, sd=4.3