

MATH 450, Fall 2018

Instructor: Vu Dinh

Practice problems

10/18/18

Time Limit: 60 Minutes

Name (Print): _____

This exam contains 4 pages (including this cover page) and 3 problems. Check to see if any pages are missing. Enter all requested information on the top of this page, and put your initials on the top of every page, in case the pages become separated.

You are allowed to bring a one-sided A4-sized hand-written note as reference.

You may use calculator.

You are required to show your work on each problem on this exam. The following rules apply:

- **Organize your work**, in a reasonably neat and coherent way, in the space provided. Work scattered all over the page without a clear ordering will receive very little credit.
- **Mysterious or unsupported answers will not receive full credit.** A correct answer, unsupported by calculations, explanation, or algebraic work will receive no credit; an incorrect answer supported by substantially correct calculations and explanations might still receive partial credit.

Problem	Points	Score
1	0	
2	0	
3	0	
Total:	0	

Do not write in the table to the right.

1. For each part, write a brief explanation in the space provided. For true/false questions, also circle true/false.

- (a) **True or false?** For an unbiased estimator $\hat{\theta}$, the mean squared error of $\hat{\theta}$ is just the variance of the estimator $\hat{\theta}$.

True. We have:

$$MSE(\hat{\theta}) = Var(\hat{\theta}) + (bias)^2$$

(the variance-bias decomposition). For an unbiased estimator, we know that $bias = 0$, which implies $MSE(\hat{\theta}) = Var(\hat{\theta})$.

- (b) Consider the distribution P with the following probability mass function

x	30	35	40
p(x)	0.2	0.3	0.5

Let X_1, X_2 be a random sample of size 2 from P , and $T = X_1 - X_2$.

Compute the expected value and the standard deviation of T .

We first compute the expected value and variance of X :

$$E[X] = 36.5, \quad Var(X) = 15.25$$

Thus

$$E[T] = E[X_1] - E[X_2] = 0,$$

$$Var[T] = Var[X_1] + Var[X_2] = 30.5$$

and

$$\sigma_T = \sqrt{30.5}$$

2. Let X_1, X_2, \dots, X_n represent a random sample from a distribution with pdf

$$f(x, \theta) = \frac{2x}{\theta + 1} e^{-x^2/(\theta+1)}, \quad x > 0$$

- (a) It can be shown that

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

Use this fact to construct an estimator of θ based on the method of moments. **We note that**

$$E[X^2 - 1] = E[X^2] - 1,$$

thus

$$E[X^2] = \theta + 1$$

Using the method of moments

$$\frac{X_1^2 + X_2^2 + \dots + X_n^2}{n} = E[X^2] = \theta + 1$$

The estimator is

$$\hat{\theta} = \frac{X_1^2 + X_2^2 + \dots + X_n^2}{n} - 1.$$

- (b) Derive the maximum-likelihood estimator for parameter θ based on the following dataset with $n = 10$

17.85, 11.23, 14.43, 19.27, 5.59, 6.36, 9.41, 6.31, 13.78, 11.95

The joint density function is

$$L(\theta) = f(x_1, x_2, \dots, x_n, \theta) = x_1 x_2 \dots x_n \left(\frac{2}{\theta + 1} \right)^n e^{-\frac{x_1^2 + x_2^2 + \dots + x_n^2}{\theta + 1}}$$

Transform to log-scale

$$\ell(\theta) = \ln(L(\theta)) = \ln(x_1 x_2 \dots x_n) + n \ln \left(\frac{2}{\theta + 1} \right) - \frac{x_1^2 + x_2^2 + \dots + x_n^2}{\theta + 1}$$

Take derivative with respect to θ :

$$\ell'(\theta) = -\frac{n}{\theta + 1} + \frac{x_1^2 + x_2^2 + \dots + x_n^2}{(\theta + 1)^2}$$

Set $\ell'(\theta) = 0$ to obtain

$$\hat{\theta} = \frac{x_1^2 + x_2^2 + \dots + x_n^2}{n} - 1.$$

Plug the numbers about in to obtain the point estimate.

3. Let X equal the amount of orange juice (in grams per day) consumed by an American. Suppose it is known that the standard deviation of X is $\sigma = 16$. To estimate the mean μ of X , an orange growers association took a random sample of $n = 76$ Americans and found that they consumed, on the average, $\bar{x} = 133$ grams of orange juice per day.

(a) Construct a 90% confidence interval for μ .

$$\left(\bar{x} - z_{0.05} \frac{\sigma}{\sqrt{n}}, \bar{x} + z_{0.05} \frac{\sigma}{\sqrt{n}} \right)$$

In our case, $n = 76$, $\bar{x} = 133$, $\sigma = 16$, $z_{0.05} = 1.645$

(b) Find a 90% one-sided confidence interval for μ that provides an upper bound for μ .

$$\left(-\infty, \bar{x} + z_{0.01} \frac{\sigma}{\sqrt{n}} \right)$$

Note: $z_{0.1} = 1.28$

(c) Suppose that X is approximately normal. Another single observation, X_{77} , is to be selected from the same distribution. Construct a 95% prediction interval for X_{77} .

$$\left(\bar{x} - z_{0.025} \sigma \sqrt{1 + \frac{1}{n}}, \bar{x} + z_{0.025} \sigma \sqrt{1 + \frac{1}{n}} \right)$$

Note: $z_{0.025} = 1.96$