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**THE UNIVERSITY OF BRITISH COLUMBIA**  
**DEPARTMENT OF STATISTICS**  
**Stat 305 Introduction to Statistical Inference**  
**Lab 5**

Group Number: \_\_\_\_\_

Members of the group (please print clearly):

	Full Name	Student ID Number
1.		
2.		
3.		
4.		

The goal of this laboratory is to practice the main procedures and concepts involved in Maximum Likelihood Estimation.

1. (2 marks) Let  $Y_i$ , for  $i = 1, \dots, n$ , be independent random variables such that:

$$Y_i \sim N(\theta x_i, 1)$$

Assume that  $\{x_i\}$  are **known constants** and they are not all 0, while  $\theta$  is an **unknown parameter** that we want to estimate. Notice that the  $Y_i$  are **NOT** identically distributed.

- (a) Show that the joint likelihood function can be written as  $f_{Y_1, \dots, Y_n}(y_1, \dots, y_n | \theta) = ce^{\sum -\frac{1}{2}(y_i - \theta x_i)^2}$ , where  $c$  is a constant that does not depend on  $\theta$ .

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(b) Hence, argue that we can write the joint log likelihood function as:

$$l(\theta) = \ln f_{Y_1, \dots, Y_n}(y_1, \dots, y_n | \theta) = c' - \frac{1}{2} \sum (y_i - \theta x_i)^2,$$

where  $c'$  is a constant that does not depend on  $\theta$ .

2. (2 marks) Let  $\hat{\theta}$  be the Maximum Likelihood **estimate** of  $\theta$ . Show that:

$$\hat{\theta} = \frac{\sum x_i y_i}{\sum x_i^2}$$

Using the **appropriate notation** seen in class, write down the Maximum Likelihood **Estimator** (MLE) of  $\theta$ .

3. (2 marks) The MLE of  $\theta$ , denoted  $\tilde{\theta}$  is a random variable and, as such, has properties such as expected value and variance. We are interested in computing variance of  $\tilde{\theta}$  in this question.

(a) Find  $\text{Var}(\tilde{\theta})$ . You can use basic properties of the variance operator to find it.

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- (b) Now use the Fisher Information to find an approximation for  $\text{Var}(\tilde{\theta})$ . Compare the approximation for  $\text{Var}(\tilde{\theta})$  from the Fisher information and the exact value for  $\text{Var}(\tilde{\theta})$  obtained in the previous question. Do they coincide *in this case*?

4. (1 mark) Find a 95% CI for  $\theta$ . Denote  $z_{0.975}$  the 97.5% quantile of a standard normal distribution.

5. (3 marks) In this particular case, we can find more about  $\tilde{\theta}$ .

- (a) Show that  $\tilde{\theta}$  is an unbiased estimator for  $\theta$ . Is it always the case that MLEs are unbiased?

- (b) Use a result seen in class to show that, *in this particular case*, we can find the **exact** distribution of  $\tilde{\theta}$ . Provide its distribution with parameter(s). What does it say about the CI previously obtained?

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6. (optional. 2 bonus marks) Use a property of MLE's to find the MLE for  $\theta^2$ . Name the property used. Is this estimator unbiased?
7. (optional. discuss with your lab partners if time allows) We have seen in an clicker question in class that the recipe above isn't the only way to compute MLE's. In fact, taking derivatives is often not possible or practical. Given  $X_i \sim U(0, \theta)$ , try to do the same we did above to find the MLE for  $\theta$ . What happens when you attempt to proceed as instructed in question 2? How would you find the MLE in this case?