## STAT 415 — Midterm 2 (Spring 2020)

Name: \_\_\_\_\_

## Exam rules:

- You have 2hours 20 minutes to complete the exam, scan it and upload onto Ecampus.
- There are **4** Questions.
- This exam is open book. You are free to use your class notes and HW solutions to solve the problems.

Do not do a brain dump. Answers which are irrelevant will be penalized.

Do not blindly copy answers from the HW solutions.

- State precisely in the derivations all the results that you use.
- Explain your arguments carefully, show all your work, and write out all your steps. Credit will be given for clarity. No credit will be given for a solution without working.
- If you are caught cheating or helping someone to cheat on this exam, you will both receive a grade of zero on the exam and you will be turned in.
- If a problem requires a numerical answer, you may express answer in terms of elementary functions or cdfs (e.g.  $\Gamma(r+1)$  or r!).

(1) Suppose that  $\underline{X} = (X_1, X_2)'$  are jointly normal where

$$\left(\begin{array}{c} X_1\\ X_2 \end{array}\right) \sim N\left(\left(\begin{array}{c} 1\\ 2 \end{array}\right), \left(\begin{array}{c} 1& 0\\ 0& 3 \end{array}\right)\right).$$

Let  $Y = (X_1 - 3^{-1}X_2)^2$  and  $W = (X_1 + X_2)^2$ .

Show that Y and W are independent random variables (stating all the results that you use).

- (2) Suppose that  $\{X_i\}_{i=1}^n$  are iid Poisson distributed random variables with probability mass function  $f(k; \lambda) = \lambda^k \exp(-\lambda)/k!$ .
  - (i) Obtain the maximum likelihood estimator of  $\lambda$ . Denote this estimator as  $\hat{\lambda}$
  - (ii) Obtain the Mean Squared error:  $E[(\hat{\lambda} \lambda)^2]$ . Hint: You can use that

$$\operatorname{var}[X] = \lambda.$$

- (iii) Obtain the asymptotic sampling distribution of the MLE.
- (iv) Construct the approximate 99% confidence interval for  $\lambda$  based on the MLE. The confidence interval should be "feasible" (something that a practitioner can immediately evaluate given the MLE).
- (v) Obtain the Fisher information of  $\lambda$ .
- (vi) For a fixed sample size n, make a sketch of the log-likelihood function corresponding to a large Fisher information and a small Fisher information. Indicate on your plots which log-likelihood will have an MLE with a small variance and which will have a large variance.

(3) Suppose that  $\{X_i\}$  are iid random variables with density

$$f(x;\beta) = \beta (1-x)^{\beta-1}$$
  $x \in [0,1]$   $\beta > 0.$ 

(i) Evaluate the expectation  $E[X_i]$ .

Hint: You may use the two identities

$$\int_{0}^{1} x^{a} (1-x)^{b} dx = \frac{\Gamma(a+1)\Gamma(b+1)}{\Gamma(a+b+2)} \text{ and } \frac{\Gamma(a+1)}{\Gamma(a)} = a,$$

where  $\Gamma(a)$  is the Gamma function. If a is an integer, then  $\Gamma(a) = (a-1)!$  and  $\Gamma(1) = 1$ .

- (ii) Based on your answer in (i) obtain the method of moments estimator of  $\beta$ . Call this estimator  $\hat{\beta}_{MoM}$ .
- (iii) Obtain the asymptotic sampling properties of  $\widehat{\beta}_{MoM}$ . Hint: You can use that

$$\operatorname{var}[X] = \frac{\beta}{(1+\beta)^2(2+\beta)}$$

- (iii) Obtain the maximum likelihood estimator of  $\beta$ . Call this estimator  $\hat{\beta}_{MLE}$ .
- (iv) Obtain the asymptotic sampling properties of  $\hat{\beta}_{MLE}$ .
- (v) By comparing the asymptotic variance of  $\hat{\beta}_{MoM}$  and  $\hat{\beta}_{MLE}$  explain for what values of  $\beta$  the two estimators have a variance which is very close.

(4) Suppose that  $\{X_i\}_{i=1}^n$  are iid random variables with density

$$f(x;\alpha) = \begin{cases} \frac{1}{10-\alpha} & x \in [10-\alpha, 10] \\ 0 & x \notin [10-\alpha, 10] \end{cases}$$

Obtain the maximum likelihood estimator of  $\alpha.$