

Name: _____ USC ID: _____ Date: _____

Signature: _____.

(By signing here, I certify that I have taken this test while refraining from cheating.)

Final Exam

This exam contains 13 pages (including this cover page) and 8 problems. Check to see if any pages are missing. Enter all requested information on the top of this page.

You may *not* use your book or any calculator on this exam. You *cannot* use your homeworks. You are required to show your work on each problem on the exam. The following rules apply:

- You have 120 minutes to complete the exam.
- **If you use a theorem or proposition from class or the notes or the book you must indicate this** and explain why the theorem may be applied. It is okay to just say, “by some theorem/proposition from class.”
- **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.
- If you need more space, use the back of the pages; clearly indicate when you have done this. Scratch paper appears at the end of the document.

Problem	Points	Score
1	10	
2	10	
3	10	
4	10	
5	10	
6	10	
7	10	
8	10	
Total:	80	

Do not write in the table to the right. Good luck!^a

^aMay 3, 2023, © 2023 Steven Heilman, All Rights Reserved.

Reference sheet

Below are some definitions that may be relevant.

We say that a sequence of random variables $X_1, X_2, \dots : \Omega \rightarrow \mathbf{R}$ **converges in probability** to a random variable $X : \Omega \rightarrow \mathbf{R}$ if: for all $\varepsilon > 0$

$$\lim_{n \rightarrow \infty} \mathbf{P}(|X_n - X| > \varepsilon) = 0.$$

We say that a sequence of real-valued random variables X_1, X_2, \dots **converges in distribution** to a real-valued random variable X if, for any $t \in \mathbf{R}$ such that $\mathbf{P}(X \leq t)$ is continuous at t ,

$$\lim_{n \rightarrow \infty} \mathbf{P}(X_n \leq t) = \mathbf{P}(X \leq t).$$

We say that a sequence of random variables $X_1, X_2, \dots : \Omega \rightarrow \mathbf{R}$ **converges in L_2** to a random variable $X : \Omega \rightarrow \mathbf{R}$ if

$$\lim_{n \rightarrow \infty} \mathbf{E}|X_n - X|^2 = 0.$$

We say that a sequence of random variables $X_1, X_2, \dots : \Omega \rightarrow \mathbf{R}$ **converges almost surely** to a random variable $X : \Omega \rightarrow \mathbf{R}$ if

$$\mathbf{P}(\lim_{n \rightarrow \infty} X_n = X) = 1.$$

Suppose $X = (X_1, \dots, X_n)$ is a random sample of size n from a distribution f where $f \in \{f_\theta : \theta \in \Theta\}$ is a family of densities (such as an exponential family). Let $t : \mathbf{R}^n \rightarrow \mathbf{R}^k$, so that $Y := t(X_1, \dots, X_n)$ is a statistic.

We say that Y is a **sufficient statistic** for θ if, for every $y \in \mathbf{R}^k$ and for every $\theta \in \Theta$, the conditional distribution of (X_1, \dots, X_n) given $Y = y$ (with respect to probabilities given by f_θ) does not depend on θ .

We say Y is **minimal sufficient** for θ if Y is sufficient for θ and, for every statistic $Z : \Omega \rightarrow \mathbf{R}^m$ that is sufficient for θ , there exists a function $r : \mathbf{R}^m \rightarrow \mathbf{R}^k$ such that $Y = r(Z)$.

We say Y is **complete** for $\{f_\theta : \theta \in \Theta\}$ if the following holds:

$$\text{For any } f : \mathbf{R}^m \rightarrow \mathbf{R} \text{ such that } \mathbf{E}_\theta f(Y) = 0 \quad \forall \theta \in \Theta, \quad \text{it holds that } f(Y) = 0.$$

We say Y is **ancillary** for θ if the distribution of Y does not depend on θ .

Let $X, Y, Z : \Omega \rightarrow \mathbf{R}$ be discrete or continuous random variables. Let A be the range of Y . Define $g : A \rightarrow \mathbf{R}$ by $g(y) := \mathbf{E}(X|Y = y)$, for any $y \in A$. We then define the **conditional expectation** of X given Y , denoted $\mathbf{E}(X|Y)$, to be the random variable $g(Y)$.

We say that Y is **uniformly minimum variance unbiased (UMVU)** for $g(\theta)$ if, for any other unbiased estimator Z for $g(\theta)$, we have $\text{Var}_\theta(Y) \leq \text{Var}_\theta(Z) \quad \forall \theta \in \Theta$.

1. (10 points)

Let X be a random variable uniformly distributed in $[0, 1]$.

(That is, X has PDF $f_X(x) = 1$ when $x \in [0, 1]$, and $f_X(x) = 0$ when $x \notin [0, 1]$.)

Let Y be a random variable uniformly distributed in $[0, 1]$.

Assume that X and Y are independent.

- Compute $\mathbf{P}(X > 3/4)$.
- Compute $\mathbf{E}X$.
- Compute $\mathbf{P}(X + Y \leq 1/2)$.

2. (10 points)

Let X_1, X_2, \dots be real-valued random variables. Let X be a real-valued random variable with finite second moment. Assume that

$$\lim_{n \rightarrow \infty} \mathbf{E} |X_n - X|^2 = 0.$$

- Prove that X_1, X_2, \dots converges in probability to X .
- Does X_1, X_2, \dots converges in distribution to X ? Justify your answer.

[This was a repeated homework question]

3. (10 points) Let $n \geq 2$ be an integer. Let X_1, \dots, X_n be a random sample from the Gaussian distribution with mean $\mu \in \mathbf{R}$ and variance $\sigma^2 > 0$. That is, X_1 has PDF $\frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{(x-\mu)^2}{2\sigma^2}}$, $\forall x \in \mathbf{R}$.

Let $\bar{X}_n := (X_1 + \dots + X_n)/n$, and let $S_n := \sqrt{\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2}$.

Show: $(n-1)S_n^2/\sigma^2$ is a chi-squared distributed random variable with $n-1$ degrees of freedom.

Hint: you can freely use the following fact:

$$nS_{n+1}^2 = (n-1)S_n^2 + \frac{n}{n+1}(X_{n+1} - \bar{X}_n)^2, \quad \forall n \geq 2.$$

You can also freely use that S_n is independent of \bar{X}_n .

You can freely use that: a chi-squared random variable with k degrees of freedom has the same distribution as a sum squares of k i.i.d. standard Gaussians.

[This was a repeated question from a previous exam.]

4. (10 points) Let $X := (X_1, \dots, X_n)$ be a random sample of size n from a binomial distribution with parameters n and p . Here n is a positive (known) integer and $0 < p < 1$ is unknown. (That is, X_1, \dots, X_n are i.i.d. and X_1 is a binomial random variable with parameters n and p , so that $\mathbf{P}(X_1 = k) = \binom{n}{k} p^k (1-p)^{n-k}$ for all integers $0 \leq k \leq n$.)

You can freely use that $\mathbf{E}X_1 = np$ and $\text{Var}X_1 = np(1-p)$.

- Compute the Fisher information $I_X(p)$ for any $0 < p < 1$.
(Consider n to be fixed.)
- Let Z be an unbiased estimator of p^2 (assume that Z is a function of X_1, \dots, X_n).
State the Cramér-Rao inequality for Z .
- Let W be an unbiased estimator of $1/p$ (assume that W is a function of X_1, \dots, X_n).
State the Cramér-Rao inequality for W .

[This was a repeated question from a previous exam.]

5. (10 points) Let X_1, \dots, X_n be i.i.d. random variables. Let $\theta > 0$ be an unknown parameter. Assume that X_1 is uniform on the interval $[0, \theta]$.

Denote $X := (X_1, \dots, X_n)$.

- Is the Fisher Information $I_{X_1}(\theta)$ well-defined? Justify your answer. If $I_{X_1}(\theta)$ can be computed, simplify $I_{X_1}(\theta)$ to the best of your ability.
- Is the Fisher Information $I_X(\theta)$ well-defined? Justify your answer. If $I_X(\theta)$ can be computed, simplify $I_X(\theta)$ to the best of your ability.

[This was a modified homework question.]

6. (10 points) Let X_1, \dots, X_n be i.i.d. random variables. Let $0 < p < 1$ be an unknown parameter. Assume that

$$\mathbf{P}(X_1 = k) = (1 - p)^{k-1}p,$$

for all positive integers $k \geq 1$.

You can freely use the following facts: $\mathbf{E}X_1 = 1/p$, $\text{Var}(X_1) = (1 - p)/p^2$

- Find a statistic Z such that: Z is a method of moments estimator of p , and Z is a function of X_1, \dots, X_n . Justify your answer.
- Find a statistic W such that W is an MLE for p , and W is a function of X_1, \dots, X_n . (Make sure to justify that an MLE exists. Is an MLE unique in this case?) (Here MLE refers to a Maximum Likelihood Estimator.)

[This was a modified qual question]

7. (10 points) Let X_1, \dots, X_n be i.i.d. random variables. Let $\lambda > 0$ be an unknown parameter. Assume that

$$\mathbf{P}(X_1 = k) = (e^{-\lambda}) \frac{\lambda^k}{k!},$$

for all nonnegative integers $k \geq 0$.

You can freely use the following facts: $\mathbf{E}X_1 = \lambda$, $\text{Var}(X_1) = \lambda$, $I_{X_1}(\lambda) = \frac{1}{\lambda}$.

You may assume that the MLE Y_n for λ exists and is unique, and it is given by:

$$Y_n = \frac{1}{n} \sum_{i=1}^n X_i.$$

- Is Y_n unbiased for λ ?
- Is the sequence of estimators Y_1, Y_2, \dots consistent? That is, does this sequence of random variables converge in probability to λ as $n \rightarrow \infty$? Justify your answer.
- What happens to the quantity $\text{Var}(Y_n)$ as $n \rightarrow \infty$? More specifically, does this quantity asymptotically achieve the lower bound in the Cramér-Rao Inequality (for unbiased estimators of λ)? Justify your answer.
- Describe all unbiased estimators of λ that achieve equality in the Cramér-Rao inequality. (Assume such an estimator is a function of X_1, \dots, X_n , for fixed n .)

8. (10 points) Let X_1, \dots, X_n be i.i.d. random variables. Let $0 < p < 1$ be an unknown parameter. Assume that

$$\mathbf{P}(X_1 = k) = (1 - p)^{k-1}p,$$

for all positive integers $k \geq 1$.

- Find a statistic Z such that: Z is complete for p , Z is sufficient for p , and Z is a function of X_1, \dots, X_n . Justify your answer. (If you want to, you can use a result from the homework to do this part of the question.)
- Find a statistic W such that W is unbiased for p , and such that W is UMVU for p . (W should be a function of X_1, \dots, X_n .) Justify your answer.

(Hint: you can use that a sum of n independent geometric random variables has the same distribution as Y where $\mathbf{P}(Y = m) = \binom{m-1}{n-1} p^n (1-p)^{m-n}$ for all integers $m \geq n$.)

[This was a modified qual question]

(Scratch paper 1)

(Scratch paper 2)

(Scratch paper 3)