

Fall 2022, MATH 3215-J, Final (60 pts)

Name:

GT ID:

- Regular exam time is 170 min.
- Open-book/notes. Calculators are allowed. No communication in any form.
- All the solutions are supposed to be short and can fit in the given space. Please read the problem statements carefully.
- Consider skipping a part if you get stuck somewhere.
- If there is a mistake in a problem statement making it unsolvable, skip it and you will be awarded full points for that problem.

1. In a 5-game series played between two teams, the first team to win a total of 3 games is the winner. Suppose that the outcomes of games are independent, and that team A wins each game with probability 0.8 and team B wins each game with probability 0.2.
 - (a) (2 pts) Conditional on the event that team A leads by 2 to 0, what is the probability that team A wins the series?
 - (b) (2 pts) Conditional on the event that team A leads by 1 to 0, what is the probability that team B wins the series?

The final answers (numerical values) should be computed explicitly and exactly.

Let the probability that team A wins a game be $p = 0.8$.

- (a) If team B wins the series, it must win 3 consecutive games, which happens with probability $(1 - p)^3$. Therefore, the answer is $1 - (1 - p)^3 = 0.992$.
- (b) Let E be the event that team A leads by 1 to 0, and let F be the event that team B wins the series. For the event $E \cap F$, the series can have winners ABBB, AABBB, ABABB, or ABBAB. Therefore,

$$\mathcal{P}(F \mid E) = \frac{\mathcal{P}(E \cap F)}{\mathcal{P}(E)} = \frac{p(1 - p)^3 + 3p^2(1 - p)^3}{p} = (1 - p)^3 + 3p(1 - p)^3 = 0.0272.$$

2. (2 pts) Suppose that the total weight of products produced daily at a factory is X^3 lbs where $X \sim \mathcal{N}(21, 0.25)$. What is the probability that the total weight exceeds 8,000 lbs in a day? Use the CDF Φ of $\mathcal{N}(0, 1)$ to express the answer.

The probability is

$$\mathbb{P}\{X^3 > 8000\} = \mathbb{P}\{X > 20\} = \mathbb{P}\left\{\frac{X - 21}{0.5} > \frac{20 - 21}{0.5}\right\} = 1 - \Phi(-2) = \Phi(2).$$

3. Suppose that the number of accidents on Highway A in a year has the Poisson distribution $\text{Poi}(100)$, and that number on Highway B has distribution $\text{Poi}(300)$, independently from the number on Highway A. Use the CDF Φ of the standard normal $\mathcal{N}(0, 1)$ to approximate the probabilities of the following events (no need to worry about correction for continuity or strict vs. non-strict inequalities):

- (a) (2 pts) The number of accidents on Highway A is at least 110.
 - (b) (2 pts) The total number of accidents on Highway A and Highway B is at most 410.
 - (c) (2 pts) There are more accidents on Highway B than on Highway A.
- (a) Let X be the number of accidents on Highway A. Then $X \sim \text{Poi}(100)$, so $\mathbb{E}[X] = 100$ and $\text{Var}(X) = 100$. It follows that

$$\mathbb{P}\{X \geq 110\} = \mathbb{P}\left\{\frac{X - 100}{10} \geq \frac{110 - 100}{10}\right\} \approx 1 - \Phi(1) = \Phi(-1).$$

- (b) Let Y be the number of accidents on Highway B. Then $Y \sim \text{Poi}(300)$, so $\mathbb{E}[Y] = 300$ and $\text{Var}(Y) = 300$. Hence, $X + Y$ is approximately $\mathcal{N}(400, 400)$. It follows that

$$\mathbb{P}\{X + Y \leq 410\} = \mathbb{P}\left\{\frac{X + Y - 400}{20} \leq \frac{410 - 400}{20}\right\} \approx \Phi(0.5).$$

- (c) We have that $Y - X$ is approximately $\mathcal{N}(200, 400)$, so

$$\mathbb{P}\{Y > X\} = \mathbb{P}\left\{\frac{Y - X - 200}{20} > \frac{-200}{20}\right\} \approx 1 - \Phi(-10) = \Phi(10) \approx 1.$$

4. Let $\text{Exp}(3)$ denote the exponential distribution with PDF $f(x) = \lambda e^{-3x}$ for $x \geq 0$ and $f(x) = 0$ for $x < 0$, where λ is a fixed constant. Let X and Y be i.i.d. $\text{Exp}(3)$ random variables.

(a) (2 pts) What is λ ?

(b) (1 pt) Denote the joint PDF of (X, Y) by $\tilde{f}(x, y)$. What is $\tilde{f}(x, y)$?

(The domain of the function needs to be clearly indicated.)

(c) (2 pts) Denote the CDF of $X + Y$ by $F(t)$ for $t \in \mathbb{R}$. What is $F(t)$ for $t \geq 0$?

(d) (2 pts) Denote the PDF of $X + Y$ by $f(t)$ for $t \in \mathbb{R}$. What is $f(t)$ for $t \geq 0$?

(a) We have $1 = \int_0^\infty \lambda e^{-3x} dx = \lambda/3$, so $\lambda = 3$.

(b) We have

$$\tilde{f}(x, y) = 9e^{-3(x+y)}$$

for $x, y \geq 0$, and $\tilde{f}(x, y) = 0$ otherwise.

(c) For $t \geq 0$, we have

$$\begin{aligned} F(t) = \mathbb{P}\{X + Y \leq t\} &= \int_0^t \int_0^{t-y} 9e^{-3(x+y)} dx dy \\ &= \int_0^t -3e^{-3(x+y)} \Big|_{x=0}^{t-y} dy \\ &= \int_0^t 3(e^{-3y} - e^{-3t}) dy \\ &= -e^{-3y} \Big|_{y=0}^t - 3te^{-3t} \\ &= 1 - e^{-3t} - 3te^{-3t}. \end{aligned}$$

(d) For $t \geq 0$, we have

$$f(t) = F'(t) = 3e^{-3t} - 3e^{-3t} + 9te^{-3t} = 9te^{-3t}.$$

5. A plane is missing and is equally likely to have gone down in one of three possible regions. For $i = 1, 2, 3$, let p_i be the probability that the plane can be found upon a search of the i th region when the plane is in fact in that region.

- (a) (1 pt) What is the probability that a search in the 2nd region is successful?
- (b) (2 pts) Conditional on the event that a search of the 2nd region is unsuccessful, what is the probability that the plane is in the 1st region?
- (c) (2 pts) Conditional on the event that a search of the 2nd region is unsuccessful, what is the probability that the plane is in the 2nd region?

(Express the answers in terms of p_1 , p_2 , and p_3 .)

Let X be the region the plane is in. Let Y be the indicator that the search in the 2nd region is successful (i.e., $Y = 1$, successful; $Y = 0$, unsuccessful).

- (a) We have

$$\mathbb{P}\{Y = 1\} = \sum_{i=1}^3 \mathbb{P}\{Y = 1 \mid X = i\} \cdot \mathbb{P}\{X = i\} = \frac{p_2}{3}.$$

- (b) We have

$$\mathbb{P}\{X = 1 \mid Y = 0\} = \frac{\mathbb{P}\{Y = 0 \mid X = 1\} \cdot \mathbb{P}\{X = 1\}}{\mathbb{P}\{Y = 0\}} = \frac{1/3}{1 - p_2/3} = \frac{1}{3 - p_2}.$$

- (c) We have

$$\mathbb{P}\{X = 2 \mid Y = 0\} = \frac{\mathbb{P}\{Y = 0 \mid X = 2\} \cdot \mathbb{P}\{X = 2\}}{\mathbb{P}\{Y = 0\}} = \frac{(1 - p_2)/3}{1 - p_2/3} = \frac{1 - p_2}{3 - p_2}.$$

6. Let X have PDF $f_X(x) = x/4$ for $x \in [1, 3]$ and $f_X(x) = 0$ otherwise. Conditional on $X = x$, let Y have the uniform distribution over $[0, x]$.

(a) (1 pt) What is the conditional PDF of Y given $X = x \in [1, 3]$?

(b) (2 pts) What is the joint PDF $f(x, y)$ of (X, Y) ?

(c) (2 pts) What is the marginal PDF $f_Y(y)$ of Y ?

The domains of the above functions (where a function is nonzero) should be clearly specified.

(a) We have

$$f_{Y|X}(y|x) = 1/x, \quad y \in [0, x].$$

(b) We have

$$f(x, y) = f_X(x) \cdot f_{Y|X}(y|x) = 1/4, \quad x \in [1, 3], y \in [0, x].$$

(c) For $y \in [0, 1]$,

$$f_Y(y) = \int_1^3 \frac{1}{4} dx = \frac{1}{2},$$

and for $y \in (1, 3]$,

$$f_Y(y) = \int_y^3 \frac{1}{4} dx = \frac{3-y}{4}.$$

7. Suppose that we are given i.i.d. X_1, \dots, X_n from the Pareto distribution with PDF $f(x) = \lambda \theta^\lambda x^{-(\lambda+1)}$ for $x \geq \theta$ and $f(x) = 0$ otherwise, where $\theta > 0$ and $\lambda > 0$ are both unknown parameters.

(a) (2 pts) What is the joint PDF of (X_1, \dots, X_n) ?

(Write it as a function of x_1, \dots, x_n with the domain clearly specified.)

(b) (1 pt) What is the MLE $\hat{\theta}$ of θ ?

(c) (1 pt) Now fix $\theta = \hat{\theta}$ (no need to plug in its formula from the previous part). What is the log-likelihood for λ ?

(d) (2 pts) What is the MLE $\hat{\lambda}$ of λ ?

(a) The joint PDF is $\lambda^n \theta^{n\lambda} (x_1 \cdots x_n)^{-(\lambda+1)}$ if $x_i \leq \theta$ for all $i = 1, \dots, n$ and is zero otherwise.

(b) The likelihood is the same as the joint PDF. To maximize this function over θ , we should take $\theta = \min_{1 \leq i \leq n} x_i$. Therefore, the MLE of θ is

$$\hat{\theta} = \min_{1 \leq i \leq n} X_i.$$

(c) The log-likelihood is

$$n \log \lambda + n \lambda \log \hat{\theta} - (\lambda + 1) \log(x_1 \cdots x_n).$$

(d) Differentiating the log-likelihood with respect to λ and setting the result to zero, we obtain

$$\frac{n}{\lambda} + n \log \hat{\theta} - \log(x_1 \cdots x_n) = 0.$$

Therefore, the MLE of λ is

$$\hat{\lambda} = \frac{n}{\log(X_1 \cdots X_n) - n \log(\hat{\theta})}.$$

8. The mean breaking strength of a type of fiber is believed to be at least 200 psi. Suppose that the breaking strength is normally distributed with standard deviation equal to 2 psi. A sample of 16 pieces of fiber yielded average breaking strength 190 psi.
- (1 pt) If we would like to show with high confidence that this type of fiber is not as strong as it is believed to be, what is the null hypothesis H_0 and what is alternative hypothesis H_1 ? (Write the hypotheses in mathematical terms like $H_0 : \mu \dots$)
 - (1 pt) What is the test statistic that has the standard normal distribution? Compute its numerical value.
 - (1 pt) What is the test at significance level 0.05? Express the test as rejecting H_0 if some condition holds (in terms of the above statistic and an explicit numerical threshold). (Recall that $\Phi(1.96) \approx 0.975$ and $\Phi(1.645) \approx 0.95$ where Φ is the CDF of $\mathcal{N}(0, 1)$).
 - (1 pt) Do we reject H_0 or not?
 - (1 pt) What is the p -value, expressed using Φ ?
 - (1 pt) What is the p -value approximately? Choose your answer from: (i) 0.1; (ii) 0.05; (iii) 0.01; (iv) 0.005; (v) 0.
- $H_0 : \mu \geq 200; H_1 : \mu < 200$
 - $\frac{\sqrt{n}}{\sigma}(\bar{X} - \mu_0) = \frac{\sqrt{16}}{2}(190 - 200) = -20$
 - Reject H_0 if $\frac{\sqrt{n}}{\sigma}(\bar{X} - \mu_0) < -z_\alpha = \Phi^{-1}(0.95) \approx -1.645$
 - Reject H_0 . Not acceptable.
 - $\mathbb{P}\{Z < -20\} = \Phi(-20)$
 - 0

9. Consider two independent samples of sizes n_1 and n_2 from $\mathcal{N}(\mu_1, \sigma_1^2)$ and $\mathcal{N}(\mu_2, \sigma_2^2)$ respectively. Let S_1^2 and S_2^2 be the respective sample variances. Let \mathcal{F} denote the F -distribution with $n_1 - 1$ and $n_2 - 1$ degrees of freedom. Recall that

$$R := \frac{S_1^2/\sigma_1^2}{S_2^2/\sigma_2^2} \sim \mathcal{F}.$$

Note that R takes values in $(0, \infty)$. Let G denote the CDF of \mathcal{F} . For $\alpha \in (0, 1)$, let $g_\alpha := G^{-1}(1 - \alpha)$ be the quantile of order $1 - \alpha$.

- (a) (1 pt) Give a test statistic (in terms of S_1^2 and S_2^2 only) that has distribution \mathcal{F} if $\sigma_1^2 = 2\sigma_2^2$.
- (b) (2 pts) Suppose that we test $H_0 : \sigma_1^2 \leq 2\sigma_2^2$ (or $\sigma_1^2 = 2\sigma_2^2$) against $H_1 : \sigma_1^2 > 2\sigma_2^2$. What is the one-sided test at significance level $\alpha \in (0, 1)$? Express the test as rejecting H_0 if some condition holds (in terms of the above statistic and a quantile).
- (c) (2 pts) What is the p -value associated with the above test (i.e., associated with the statistic from part (a))? Express it in terms of S_1^2 , S_2^2 , and the CDF G .
- (d) (1 pt) Suppose that we test $H_0 : \sigma_1^2 \geq 2\sigma_2^2$ (or $\sigma_1^2 = 2\sigma_2^2$) against $H_1 : \sigma_1^2 < 2\sigma_2^2$. What is the one-sided test at significance level $\alpha \in (0, 1)$? Express the test as rejecting H_0 if some condition holds (in terms of S_1^2 , S_2^2 , and a quantile).

- (a) If $\sigma_1^2 = 2\sigma_2^2$, then

$$\frac{S_1^2}{2S_2^2} = \frac{S_1^2\sigma_2^2}{S_2^2\sigma_1^2} = R \sim \mathcal{F}.$$

- (b) We have $\mathbb{P} \left\{ \frac{S_1^2}{2S_2^2} > g_\alpha \right\} = \alpha$, so we reject H_0 if $S_1^2 > 2g_\alpha S_2^2$.
- (c) The p -value is

$$\mathbb{P} \left\{ L > \frac{S_1^2}{2S_2^2} \right\} = 1 - G \left(\frac{S_1^2}{2S_2^2} \right),$$

where the probability is with respect to $L \sim \mathcal{F}$.

- (d) We reject H_0 if $S_1^2 < 2g_{1-\alpha} S_2^2$.

10. Suppose that we observe i.i.d. $X_1, \dots, X_n \sim \mathcal{N}(\mu, \sigma^2)$ where σ^2 is unknown. Consider testing $H_0 : \mu \leq \mu_0$ against $H_1 : \mu > \mu_0$. Let S^2 denote the sample variance. For $\alpha \in (0, 1)$, let $t_\alpha := F^{-1}(1 - \alpha)$, where F denotes the CDF of the t -distribution with $n - 1$ degrees of freedom.

- (a) (2 pts) Find a statistic B (a quantity expressed in terms of the quantile t_α and the data) such that the following holds: The test that rejects H_0 if $B > \mu_0$ achieves significance level α exactly.
- (b) (2 pts) Find a p -value P (a quantity expressed in terms of F , μ_0 , and the data) such that the following holds: The test that rejects H_0 if $P < \alpha$ achieves significance level α exactly.
- (c) (1 pt) Are the above two tests equivalent?
- (d) (1 pt) If $\mu = \mu_0$, what is the CDF $\tilde{F}(t)$ of the P ? (Recall that the p -value is expressed in terms of the random data, so it is a random variable and therefore has its own CDF.)

(a) The t -test at significance level α rejects H_0 if $\frac{\sqrt{n}}{S}(\bar{X} - \mu_0) > t_\alpha$. The rejection rule is equivalent to $\bar{X} - \mu_0 > t_\alpha \frac{S}{\sqrt{n}}$, i.e., $\mu_0 < \bar{X} - t_\alpha \frac{S}{\sqrt{n}}$. Therefore, it suffices to set $B = \bar{X} - t_\alpha \frac{S}{\sqrt{n}}$.

(b) The p -value is $1 - F\left(\frac{\sqrt{n}}{S}(\bar{X} - \mu_0)\right)$.

(c) They are equivalent because, by (a) and (b),

$$\mu_0 < B \iff \frac{\sqrt{n}}{S}(\bar{X} - \mu_0) > t_\alpha \iff F\left(\frac{\sqrt{n}}{S}(\bar{X} - \mu_0)\right) > F(t_\alpha) = 1 - \alpha,$$

where the last condition is equivalent to p -value smaller than α .

(d) The test at significance level α rejects H_0 if $P < \alpha$. This literally means that if $\mu = \mu_0$, then $\mathbb{P}\{P < \alpha\} = \alpha$. Therefore, $\tilde{F}(t) = t$ for $t \in (0, 1)$, $\tilde{F}(t) = 0$ for $t \leq 0$, and $\tilde{F}(t) = 1$ for $t \geq 1$.

11. Consider the linear regression model

$$Y_i = \beta x_i + \varepsilon_i, \quad i = 1, \dots, n,$$

where β is the parameter to be estimated, x_i are fixed covariates, and the noise terms ε_i are i.i.d. $\mathcal{N}(0, 1)$ random variables.

- (a) (1 pt) What is the distribution of Y_i ?
(The name and parameters of the distribution all need to be specified.)
 - (b) (1 pt) What is the joint PDF of (Y_1, \dots, Y_n) ?
 - (c) (1 pt) What is the log-likelihood at β ?
 - (d) (2 pts) What is the MLE $\hat{\beta}$ of β ?
 - (e) (2 pts) What is the mean of $\hat{\beta}$?
- (a) We have $Y_i \sim \mathcal{N}(\beta x_i, 1)$.
- (b) The joint PDF is

$$f(y_1, \dots, y_n) = \frac{1}{(2\pi\sigma)^{n/2}} \exp\left(-\frac{1}{2} \sum_{i=1}^n (y_i - \beta x_i)^2\right).$$

- (c) The log-likelihood is

$$-\frac{n}{2} \log(2\pi\sigma) - \sum_{i=1}^n (y_i - \beta x_i)^2.$$

- (d) To find the MLE, we need to minimize

$$\sum_{i=1}^n (Y_i - \beta x_i)^2,$$

take its derivative with respect to β and set the result to zero:

$$2 \sum_{i=1}^n x_i (Y_i - \beta x_i) = 0.$$

This yields

$$\hat{\beta} = \frac{\sum_{i=1}^n x_i Y_i}{\sum_{i=1}^n x_i^2}.$$

- (e) The mean of $\hat{\beta}$ is

$$\frac{\sum_{i=1}^n x_i \mathbb{E}[Y_i]}{\sum_{i=1}^n x_i^2} = \frac{\sum_{i=1}^n x_i \beta x_i}{\sum_{i=1}^n x_i^2} = \beta.$$