

Probability and Statistics

Mock Exam Paper A Reference Answer

1 Fill in the blanks (20 points)

1. The binomial distribution with parameters n and p has mean \underline{np} and variance $\underline{np(1-p)}$.
2. The Poisson distribution with parameter λ has mean $\underline{\lambda}$ and variance $\underline{\lambda}$.
3. The geometric distribution with parameter p has mean $\underline{\frac{1}{p}}$ and variance $\underline{\frac{1-p}{p^2}}$.
4. The uniform distribution on interval $[a, b]$ has mean $\underline{\frac{a+b}{2}}$ and variance $\underline{\frac{(b-a)^2}{12}}$.
5. The exponential distribution with parameter β has mean $\underline{\frac{1}{\beta}}$ and variance $\underline{\frac{1}{\beta^2}}$.

2

(8 points) A box contains 20 good products and 5 defective products. An inspector draws from the box three products one at a time, without replacement.

- (a) What is the probability that the third product drawn is defective?
- (b) What is the probability that the third product drawn is the second defective product drawn?

Solution: (a) $\frac{5}{20+5} = 0.2$.

(b) Let D_i denote the event that the i -th drawn product is defective. Then the desired probability is $P(D_1 D_2^c D_3 \cup D_1^c D_2 D_3)$.

$$\begin{aligned}
 P(D_1 D_2^c D_3 \cup D_1^c D_2 D_3) &= P(D_1 D_2^c D_3) + P(D_1^c D_2 D_3) \\
 &= P(D_1)P(D_2^c|D_1)P(D_3|D_1 D_2^c) + P(D_1^c)P(D_2|D_1^c)P(D_3|D_1^c D_2) \\
 &= \frac{5}{25} \times \frac{20}{24} \times \frac{4}{23} + \frac{20}{25} \times \frac{5}{24} \times \frac{4}{23} = \frac{4}{69}.
 \end{aligned}$$

3

(10 points) Let A and B be two events. Define random variables X and Y as follows:

$$X(s) = \begin{cases} 1 & \text{for } s \in A \\ 0 & \text{otherwise.} \end{cases}, \quad Y(s) = \begin{cases} 1 & \text{for } s \in B \\ 0 & \text{otherwise.} \end{cases}.$$

Show that $\text{Cov}(X, Y) = 0$ if and only if A and B are independent.

Solution:

$$\begin{aligned}
 E(X) &= P(X = 0) \cdot 0 + P(X = 1) \cdot 1 = P(A) \\
 E(Y) &= P(Y = 0) \cdot 0 + P(Y = 1) \cdot 1 = P(B) \\
 E(XY) &= P(XY = 0) \cdot 0 + P(XY = 1) \cdot 1 = P(X = 1, Y = 1) = P(AB)
 \end{aligned}$$

Hence

$$\begin{aligned}
 \text{Cov}(X, Y) &= E(XY) - E(X)E(Y) = P(AB) - P(A)P(B) \\
 \text{Cov}(X, Y) &= 0 \Leftrightarrow P(AB) = P(A)P(B)
 \end{aligned}$$

4

(12 points) Suppose that random variables X, Y are **independent** and $X \sim \text{normal}(0, \sigma_1^2)$ and $Y \sim \text{normal}(0, \sigma_2^2)$. Suppose further that $P(|5X + 12Y| < 13) < P(|4X + 3Y| < 5)$. Compare σ_1 and σ_2 .

Solution: By assumption, $5X + 12Y \sim \text{normal}(0, 25\sigma_1^2 + 144\sigma_2^2)$ and $4X + 3Y \sim \text{normal}(0, 16\sigma_1^2 + 9\sigma_2^2)$. And

$$P(|5X + 12Y| < 13) = P\left(\frac{|5X + 12Y|}{\sqrt{25\sigma_1^2 + 144\sigma_2^2}} < \frac{13}{\sqrt{25\sigma_1^2 + 144\sigma_2^2}}\right) = 2\Phi\left(\frac{13}{\sqrt{25\sigma_1^2 + 144\sigma_2^2}}\right) - 1$$

$$P(|4X + 3Y| < 5) = P\left(\frac{|4X + 3Y|}{\sqrt{16\sigma_1^2 + 9\sigma_2^2}} < \frac{5}{\sqrt{16\sigma_1^2 + 9\sigma_2^2}}\right) = 2\Phi\left(\frac{5}{\sqrt{16\sigma_1^2 + 9\sigma_2^2}}\right) - 1$$

It then follows that

$$\begin{aligned}\frac{13}{\sqrt{25\sigma_1^2 + 144\sigma_2^2}} &< \frac{5}{\sqrt{16\sigma_1^2 + 9\sigma_2^2}} \\ \frac{169}{25\sigma_1^2 + 144\sigma_2^2} &< \frac{25}{16\sigma_1^2 + 9\sigma_2^2} \\ 2704\sigma_1^2 + 1521\sigma_2^2 &< 625\sigma_1^2 + 3600\sigma_2^2 \\ 2079\sigma_1^2 &< 2079\sigma_2^2 \\ \sigma_1 &< \sigma_2.\end{aligned}$$

5

(14 points) Suppose that X and Y are random variables having the following joint p.d.f.:

$$f(x, y) = \begin{cases} 6xy & \text{for } 0 < y < \sqrt{x} < 1, \\ 0 & \text{otherwise.} \end{cases}$$

- (a) Compute $P(X \leq Y)$.
- (b) Find marginal density $f_Y(y)$.
- (c) Find conditional density $f_{X|Y}(x|y)$.

Solution: (a)

$$P(X \leq Y) = \int_0^1 dx \int_x^{\sqrt{x}} 6xy dy = \int_0^1 dx \cdot 3xy^2 \Big|_{y=x}^{y=\sqrt{x}} = \int_0^1 3(x^2 - x^3) dx = \frac{1}{4}.$$

- (b) If $y \leq 0$ or $y \geq 1$, then $f_Y(y) = 0$. Otherwise,

$$f_Y(y) = \int_{y^2}^1 6xy dx = 3x^2 y \Big|_{x=y^2}^{x=1} = 3y - 3y^5.$$

To sum up,

$$f_Y(y) = \begin{cases} 3y - 3y^5 & \text{for } 0 < y < 1, \\ 0 & \text{otherwise.} \end{cases}$$

- (c) If $0 < y < 1$, then

$$f_{X|Y}(x|y) = \frac{f(x, y)}{f_Y(y)} = \begin{cases} \frac{2x}{1-y^4} & \text{for } y^2 < x < 1, \\ 0 & \text{otherwise.} \end{cases}$$

6

(12 points) Let X be the number of white cells in a ml of man's blood. It is known that X has a mean of 7300, and a standard deviation of 700.

(a) Using Chebyshev's inequality, find an interval whose center is the mean so that the probability that X is in that interval is at least $\frac{1}{2}$.

(b) If it is further supposed that X has normal distribution, find the interval whose center is the mean so that the probability that X is in that interval is $\frac{1}{2}$.

Solution: (a) By Chebyshev's inequality, $P(|X - \mu| \geq \kappa\sigma) \leq \frac{1}{\kappa^2}$, i.e. $P(|X - \mu| < \kappa\sigma) \geq 1 - \frac{1}{\kappa^2}$. Let $\kappa = \sqrt{2}$, then

$$P(|X - \mu| < \sqrt{2}\sigma) \geq \frac{1}{2}.$$

Therefore, that interval is $(\mu - \sqrt{2}\sigma, \mu + \sqrt{2}\sigma) = (6310.050506338834, 8289.949493661166)$.

(b) If $X \sim \text{normal}(\mu, \sigma^2)$, then $P(|X - \mu| < \kappa\sigma) = 2\Phi(\kappa) - 1$. To have $2\Phi(\kappa) - 1 = \frac{1}{2}$, $\kappa = \Phi^{-1}(\frac{3}{4}) = 0.6745$. Therefore, that interval is $(\mu - 0.6745\sigma, \mu + 0.6745\sigma) = (6827.85, 7772.15)$.

7

(12 points) When a computer adds up two real numbers and stores the result into an integer, the error has the uniform distribution on the interval $[-0.5, 0.5]$. Use the central limit theorem to approximately evaluate the following quantities:

(a) The probability that when 1501 real numbers are added, the total error is in the range $[-15, 15]$.

(b) How many real numbers can be added so that the probability for the total error to be in the range $[-10, 10]$ is at least 0.9?

Solution: (a) Let E_i be the error of the i -th addition. Let $S_n = \sum_{i=1}^n E_i$. Then $E(X_i) = 0$ and $\text{Var}(X_i) = \frac{1}{12}$, and $E(S_n) = 0$ and $\text{Var}(S_n) = \frac{n}{12}$. The total error of addition of 1501 real numbers is S_{1500} and

$$E(S_{1500}) = 0 \text{ and } \text{Var}(S_{1500}) = 125.$$

By the central limit theorem, S_{1500} approximately $\sim \text{normal}(0, 125)$. Thus

$$P(|S_{1500}| \leq 15) = P\left(\frac{|S_{1500}|}{\sqrt{125}} \leq \frac{15}{\sqrt{125}}\right) \approx 2\Phi\left(\frac{15}{\sqrt{125}}\right) - 1 = 0.820287505121.$$

(b) By the central limit theorem, S_n approximately $\sim \text{normal}(0, \frac{n}{12})$. We want to have

$$0.9 \leq P(|S_n| \leq 10) = P\left(\frac{|S_n|}{\sqrt{\frac{n}{12}}} \leq \frac{10}{\sqrt{\frac{n}{12}}}\right) \approx 2\Phi\left(\sqrt{\frac{1200}{n}}\right) - 1.$$

$$\Phi\left(\sqrt{\frac{1200}{n}}\right) \geq 0.95$$

$$\sqrt{\frac{1200}{n}} \geq 1.645$$

$$n \leq 443.$$

Therefore at most 444 real numbers can be added to allow the desired relation to hold.

(12 points) Suppose that X_1, \dots, X_n form a random sample from a distribution for which the p.d.f. is as follows:

$$f(x; \theta) = \begin{cases} \frac{x}{\theta^2} e^{-\frac{x}{\theta}} & \text{for } x \geq 0, \\ 0 & \text{otherwise.} \end{cases}$$

where $\theta > 0$ is the unknown parameter to be estimated.

- (a) Derive the M.L.E. of θ .
- (b) Is the M.L.E. unbiased?

Solution: (a) For observed values x_1, \dots, x_n , the likelihood function is

$$L(\theta) = \theta^{-2n} \prod_{i=1}^n x_i \cdot \exp\left(-\sum_{i=1}^n x_i/\theta\right).$$

$$\log L(\theta) = -2n \log(\theta) + \log\left(\prod_{i=1}^n x_i\right) - \sum_{i=1}^n x_i/\theta.$$

The maximizer of $\log L(\theta)$ is the solution of the following equation:

$$\frac{d \log L}{d\theta} = -\frac{2n}{\theta} + \frac{\sum_{i=1}^n x_i}{\theta^2} = 0,$$

$$\theta = \frac{\sum_{i=1}^n x_i}{2n}.$$

Thus the M.L.E. of θ is $\hat{\theta} = \frac{\sum_{i=1}^n X_i}{2n}$.

(b)

$$E(X_i) = \int_0^\infty \frac{x^2}{\theta^2} e^{-\frac{x}{\theta}} dx = \theta \int_0^\infty \left(\frac{x}{\theta}\right)^2 e^{-\frac{x}{\theta}} d\frac{x}{\theta} = \theta \Gamma(3) = 2\theta.$$

Hence

$$E(\hat{\theta}) = E\left(\frac{\sum_{i=1}^n X_i}{2n}\right) = \frac{n \cdot 2\theta}{2n} = \theta,$$

which means $\hat{\theta}$ is an unbiased estimator of θ .