Homework Problems

- Note. Always show your work in your homework solutions to receive full points unless it is stated otherwise.
- 1. (6 pts) Suppose that \mathcal{F} is a σ -field on a space Ω and $\{A_n\}_{n=1}^{\infty}$ is a sequence of sets in \mathcal{F} . Show that

$$\left(\bigcup_{n=1}^{\infty} A_n\right)^c = \bigcap_{n=1}^{\infty} (A_n^c).$$

2. (6 pts) Suppose that P is a probability function defined on a σ -field \mathcal{F} . For a set $B \in \mathcal{F}$ such that P(B) > 0, define a function Q on \mathcal{F} by

$$Q(A) = \frac{P(A \cap B)}{P(B)}$$

for $A \in \mathcal{F}$. Verify that Q is a probability function on \mathcal{F} .

3. (6 pts) Suppose that \mathcal{F} is a σ -field on a space Ω and P is a probability function on \mathcal{F} . Suppose that $\{A_n\}_{n=1}^{\infty}$ is a sequence of sets in \mathcal{F} such that $A_n \supset A_{n+1}$ for all $n \in \{1, 2, \ldots\}$. Show that

$$P\left(\cap_{n=1}^{\infty}A_n\right) = \lim_{n \to \infty} P(A_n).$$

You may use the following fact in your proof.

Fact 1 Suppose that \mathcal{F} is a σ -field on a space Ω and P is a probability function on \mathcal{F} . Suppose that $\{A_n\}_{n=1}^{\infty}$ is a sequence of sets in \mathcal{F} such that $A_n \subset A_{n+1}$ for all $n \in \{1, 2, \ldots\}$. Then,

$$P\left(\cup_{n=1}^{\infty}A_n\right) = \lim_{n \to \infty} P(A_n).$$

4. (4 pts) Suppose that \mathcal{F} is a σ -field on $\Omega = (-\infty, \infty)$ such that all open intervals in $(-\infty, \infty)$ are in \mathcal{F} . Suppose that P is a probability function on \mathcal{F} such that for $n \in \{1, 2, \ldots, \}$,

$$P\left(\left(-\frac{1}{n},\frac{1}{n}\right)\right) = 0.4 + \frac{0.6}{n}.$$

Find $P(\{0\})$.

5. (8 pts) Suppose that \mathcal{F} is a σ -field on $\Omega = \{1, 2, 3, 4, 5\}$. Let $A = \{1, 2, 4, 5\}$, $B = \{1, 2, 4\}$ and $C = \{1, 4, 5\}$. Suppose that A, B, C are in \mathcal{F} and P is a probability function defined on \mathcal{F} . Note that each of P(A), P(B) and P(C) can be expressed as linear combinations of $P(\{5\})$, $P(\{1, 4\})$ and $P(\{2\})$. Use the linear relations to express $P(\{5\})$, $P(\{1, 4\})$, $P(\{2\})$ and $P(\{3\})$ in terms of P(A), P(B) and P(C), and explain why we cannot have

$$(P(A), P(B), P(C)) = (0.5, 0.3, 0.1).$$

6. (4 pts) Suppose that A_1, A_2, A_3, A_4 are events in a σ -field on Ω . Suppose that $P(A_i) = 0.1$ and $P(A_i \cap A_j) = 0.05$ for $i, j \in \{1, 2, 3, 4\}$. Find a lower bound and an upper bound for $P(A_1 \cup A_2 \cup A_3 \cup A_4)$. Be sure that the lower bound is greater than 0 and the upper bound is less than 1.

- 7. (8 pts) Suppose that $\Omega = \{1, 2, 3, 4, 5\}$, $A = \{1, 2, 3\}$ and $B = \{3, 4, 5\}$. Let $\mathcal{C} = \{\emptyset, \Omega, A, B\}$. Suppose that \mathcal{F} is the smallest σ -field on Ω and $\mathcal{C} \subset \mathcal{F}$. List all sets that should be included in \mathcal{F} and explain why they should be in \mathcal{F} . You do not have to verify that the collection of sets in your list is a σ -field, but be sure that it is.
- 8. (6 pts) Suppose that F is the CDF of a random variable X. Show that

$$\lim_{x \to a^+} F(x) = F(a)$$

for all $a \in R$. You may use the fact that $\lim_{x\to a^+} F(x)$ exists for all $a \in R$ without proving it.

Hint: find a decreasing sequence $\{A_n\}_{n=1}^{\infty}$ so that $\bigcap_{n=1}^{\infty} A_n = (-\infty, a]$ and then use the continuity of a probability function to establish the result.

9. (6 pts) Suppose that X is a random variable with CDF F, where

$$F(x) = \begin{cases} 0 & \text{if } x < 0; \\ 0.5 + 0.5x & \text{if } 0 \le x < 1; \\ 1 & \text{if } x \ge 1; \end{cases}$$

- (a) (3 pts) Find P(X = a) for every $a \in R$.
- (b) (3 pts) Find $P(0 \le X \le 1)$.
- 10. (6 pts) For $a, b \in R$ such that a < b, define a function $f_{a,b}$ on R as follows: for $x \in R$,

$$f_{a,b}(x) = \begin{cases} 1/(b-a) & \text{if } x \in (a,b); \\ 0 & \text{if } x \notin (a,b). \end{cases}$$

Suppose that X is a random variable with PDF $f_{a,b}$. Show that $f_{0,1}$ is a PDF of (X - a)/(b - a).

Notation.

- In Problem 10, the distribution of X is called the uniform distribution on (a, b), denoted by U(a, b).
- We will write $X \sim U(a, b)$ to indicate the distribution of X is U(a, b).
- We will use the notation I_A to denote the indicator function of A for a given set A, which is defined by

$$I_A(x) = \begin{cases} 1 & \text{if } x \in A; \\ 0 & \text{otherwise} \end{cases}$$

For instance, the $f_{a,b}(x)$ defined in Problem 10 is $I_{(a,b)}(x)$.

11. (6 pts) Suppose that X is a random variable with PDF f_X , where

$$f_X(x) = 2xe^{-x^2}I_{(0,\infty)}(x)$$
 for $x \in R$

Find a PDF of $Y = \sqrt{X}$.

12. (12 pts) Consider the following function F:

$$F(x) = \begin{cases} c_1 & \text{if } x < 0; \\ 0.5 + 0.5x & \text{if } 0 < x < 1; \\ c_2 & \text{if } x > 1; \\ c_3 & \text{if } x = 0; \\ c_4 & \text{if } x = 1, \end{cases}$$

where c_1, c_2, c_3, c_4 are constants. Suppose that F is the CDF of some random variable X.

(a) (8 pts) Use the properties of a CDF given in Page 2 of the handout "Random variables" to find c_1, c_2, c_3, c_4 . The link for the handout is

https://stat.walkup.tw/teaching/math_stat_under/handouts/C01_5_random_variable.pdf

- (b) (4 pts) Explain why X is not a discrete random variable.
- 13. (6 pts) Suppose that X is a discrete random variable with PMF p_X , which is given below:

$$p_X(x) = \begin{cases} 0.2 & \text{if } x = -1; \\ 0.4 & \text{if } x = 0; \\ c \cdot (0.5)^x & \text{if } x \in \{1, 2, 3..., \}; \\ 0 & \text{otherwise,} \end{cases}$$

where c > 0 is a constant.

- (a) (3 pts) Find c.
- (b) (3 pts) Find P(X > 25).
- 14. (6 pts) Suppose that X is a random variable with PDF f_X . Let $S_X = \{x : f_X(x) > 0\}$. Suppose that S_X is an open interval and f_X is continuous on S_X . Let F be the CDF of X, then it can be shown that F' > 0, F is continuous on S_X , and the inverse function F^{-1} is defined on (0,1)and differentiable on (0,1) (but you don't have to prove these results). Suppose that $U \sim U(0,1)$, that is, the $f_{0,1}$ defined in Problem 10 is a PDF of U. Show that f_X is a PDF of $F^{-1}(U)$.

Remark. The result that X and $F^{-1}(U)$ have the same distribution can be established under weaker conditions.

15. (6 pts) Suppose that X is a random variable with CDF F, where

$$F(x) = \begin{cases} 0 & \text{if } x < 0; \\ 1 - e^{-2x} & \text{if } x \ge 0. \end{cases}$$

- (a) (2 pts) Find P(X > 2).
- (b) (4 pts) Find a PDF of X.
- 16. (6 pts) Suppose that X is a random variable with PDF f_X , where

$$f_X(x) = |x|I_{(-1,0)}(x) + 0.5e^{-x}I_{(0,\infty)}(x)$$

Find a PDF of $Y = X^2$.

Hint: find the CDF of Y first, take the derivative of the CDF as a guess of the PDF of Y, and then verify it is the PDF of Y.

17. (6 pts) Suppose that X is a random variable with PDF f_X , where

$$f_X(x) = e^{-x} I_{(0,\infty)}(x)$$
 for $x \in R$.

Suppose that

$$Y = \begin{cases} X & \text{if } X \le 0.5; \\ 0.5 & \text{if } X > 0.5. \end{cases}$$

Find the CDF of Y and explain why Y does not have a PDF.

18. (8 pts) Suppose that X is a random variable with PDF f_X , where

$$f_X(x) = 2xe^{-x^2}I_{(0,\infty)}(x)$$
 for $x \in R$

- (a) (4 pts) Find the CDF of X.
- (b) (4 pts) Find the median and the IQR of the distribution of X.
- 19. (4 pts) Suppose that X is a random variable such that both $E(X^2)$ and E(|X|) are finite. Verify that $Var(X) = E(X^2) (E(X))^2$ using Properties (i)–(iii) listed in Page 6 of the handout "Quantile and expectation".
- 20. (4 pts) Suppose that X is a random variable with finite expectation μ and standard deviation $\sigma > 0$. Let $Y = (X \mu)/\sigma$. Find E(Y) and Var(Y) with $\mu = 1.5$ and $\sigma = 1.2$.
- 21. (4 pts) Suppose that X is a discrete random variable with PMF p_X , where for $x \in R$,

$$p_X(x) = \begin{cases} C_x^3 (0.6)^x (0.4)^{3-x} & \text{if } x \in \{0, 1, 2, 3\}; \\ 0 & \text{otherwise.} \end{cases}$$

Find $E(X^2)$.

22. (8 pts) Suppose that X is a random variable with PDF f_X , where

$$f_X(x) = \frac{1}{2x^2} \cdot I_{(-\infty, -1)\cup(1,\infty)}(x)$$

for $x \in R$.

- (a) (4 ps) Find a PDF of 1/X.
- (b) (4 ps) Find E(1/X).
- 23. (8 pts) Suppose that X and Y are discrete random variables, and

$$P((X,Y) = (x,y)) = \begin{cases} 0.5 & \text{if } (x,y) = (1,2); \\ 0.1 & \text{if } (x,y) = (3,2); \\ 0.3 & \text{if } (x,y) = (3,6); \\ 0.1 & \text{if } (x,y) = (3,7); \\ 0 & \text{otherwise.} \end{cases}$$

It can be shown that for discrete random variables X and Y,

$$E(g(X,Y)) = \sum_{(x,y):P((X,Y)=(x,y))>0} g(x,y)P((X,Y)=(x,y))$$
(1)

if g is nonnegative.

- (a) (4 pts) Find E(XY) using (1).
- (b) (4 pts) Find E(XY) using the PMF of XY.
- 24. (4 pts) Suppose that X has PDF f_X , and g is a function defined by

$$g(x) = \sum_{i=1}^{m} a_i I_{A_i}(x),$$

where a_1, \ldots, a_m are constants and A_1, \ldots, A_m are disjoint intervals. Therefore, g(X) has only *m* possible values a_1, \ldots, a_m . Verify that

$$E(g(X)) = \int_{-\infty}^{\infty} g(x) f_X(x) dx.$$

25. (8 pts) Suppose that X is a discrete random variable with PMF p_X , where

$$p_X(x) = \frac{e^{-\lambda} \cdot \lambda^x}{x!} I_{\{0,1,2,\dots\}}(x)$$
(2)

for $x \in R$, where $\lambda > 0$ is a constant.

- (a) (4 pts) Show that $E(X) = \lambda$ and $E(X(X-1)) = \lambda^2$.
- (b) (4 pts) Find Var(X). You may use the result in Part (a) even if you choose not to do Part (a).

Note. The distribution of X with the PMF p_X given in (2) is called the Poisson distribution with mean λ .

26. (4 pts) Suppose that X is a random variable with MGF M_X , where

$$M_X(t) = 0.6 + 0.4e^{2t}$$

for
$$t \in (-\infty, \infty)$$
. Find $E(X^k)$ for $k \in \{1, 2, 3, 4\}$.

Hint. You may use the following result.

Fact 2 Suppose that

$$f(x) = \sum_{k=0}^{\infty} a_k x^k$$

for |x| < r for some positive constant r, then

$$f^{(k)}(0) = a_k \cdot k!$$

for $k \in \{0, 1, 2, \ldots\}$.

27. (4 pts) Suppose that Y is a discrete random variable with PMF p_Y , where

$$p_Y(y) = \begin{cases} 0.6 & \text{if } y = 0; \\ 0.4 & \text{if } y = 2; \\ 0 & \text{otherwise} \end{cases}$$

Show that the distribution of Y is the same as the distribution of the X in Problem 26 by verifying that X and Y have the same MGF.

- 28. (8 pts) Suppose that X is a random variable whose distribution is the Poisson distribution with mean λ , where $\lambda > 0$. The PMF of X is the function p_X given in (2) in Problem 25.
 - (a) (4 pts) Find the MGF of X.
 - (b) (4 pts) Show that $E(X) = \lambda$ and find Var(X) using the MGF of X.
- 29. (4 pts) Suppose that X is a random variable with MGF M_X , where $M_X(t) < \infty$ for $t \in (-h, h)$ for some positive constant h. For constants a and $b \in R$, let Y = a + bX and let M_Y be the MGF of Y. Show that if $b \neq 0$, then

$$M_Y(t) = e^{ta} M_X(tb) < \infty$$

for $t \in (-h/|b|, h/|b|)$.

30. (12 pts) For constants $\mu \in R$ and $\sigma > 0$, define

$$f_{\mu,\sigma}(x) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-(x-\mu)^2/(2\sigma^2)}$$

for $x \in R$. Suppose that X is a random variable with PDF $f_{\mu,\sigma}$.

(a) (4 pts) Let $Y = (X - \mu)/\sigma$. Find the MGF of Y. You may use the result that

$$\int_{-\infty}^{\infty} e^{tx} f_{0,1}(x) dx = e^{0.5t^2}$$

for $t \in R$, which has been proved in class.

- (b) (4 pts) Find the MGF of X.
- (c) (4 pts) Find $E(Y^6)$.

Note. For Part (c), it is easier to find $E(Y^6)$ using Fact 2 than performing direct differentiation of the MGF 6 times.

31. (8 pts) Suppose that (X, Y) is a random vector with CDF $F_{X,Y}$, where for $(x, y) \in \mathbb{R}^2$,

$$F_{X,Y}(x,y) = 0.5G(x)G(y) + 0.5(1 - e^{-x})(1 - e^{-y})I_{[0,\infty)}(x)I_{[0,\infty)}(y),$$

and the function G is defined by

$$G(x) = xI_{(0,1)}(x) + I_{[1,\infty)}(x)$$

for $x \in R$.

- (a) (4 pts) Find $P(0 < X \le 1 \text{ and } 1 < Y \le 2)$.
- (b) (4 pts) Find the CDF of X.
- 32. (4 pts) Suppose that (X, Y, Z) is a random vector with joint CDF F. Show that

$$P((X, Y, Z) \in (a, b] \times (c, d] \times (e, f])$$

= $F(b, d, f) - F(b, c, f) - F(a, d, f) + F(a, c, f)$
 $-F(b, d, e) + F(b, c, e) + F(a, d, e) - F(a, c, e),$

where a, b, c, d, e, f are constants such that a < b, c < d and e < f.

33. (12 pts) Suppose that (X, Y) has PDF $f_{X,Y}$, where

$$f_{X,Y}(x,y) = cxI_{(0,1)}(x)I_{(0,1)}(y)$$

for $(x, y) \in \mathbb{R}^2$ and c > 0 is a constant.

- (a) (4 pts) Show that c = 2.
- (b) (4 pts) Find $P(X + 2Y \le 1)$.
- (c) (4 pts) Find a PDF of Y.
- 34. (4 pts) Suppose that (X, Y) has joint PDF $f_{X,Y}$ and there exist two nonnegative functions g and h such that

$$f_{X,Y}(x,y) = g(x)h(y)$$

for $(x, y) \in R^2$. Show that there exist f_X : a PDF of X and f_Y : a PDF of Y such that

$$f_{X,Y}(x,y) = f_X(x)f_Y(y)$$
 (3)

for $(x, y) \in \mathbb{R}^2$.

Note that (3) implies that

$$P(\{X \in A\} \cap \{Y \in B\})$$

= $\int_{A \times B} f_X(x) f_Y(y) d(x, y)$
= $\int_A f_X(x) dx \cdot \int_B f_Y(y) dy$
= $P(\{X \in A\}) P(\{Y \in B\})$

for $A, B \in \mathcal{B}(R)$, which implies that X and Y are independent.

35. (10 pts) Suppose that (X, Y) has PDF $f_{X,Y}$, where

$$f_{X,Y}(x,y) = ce^{-(x^2 + y^2)/2} I_{(0,\infty)}(x) I_{(0,\infty)}(y)$$

for $(x, y) \in \mathbb{R}^2$ and c > 0 is a constant. Let $U = \sqrt{X^2 + Y^2}$ and $V = \tan^{-1}(Y/X)$. Note that for $z \in (-\infty, \infty)$, $\tan^{-1}(z)$ is the value $\theta \in (-\pi/2, \pi/2)$ such that $\tan(\theta) = z$.

- (a) (4 pts) Find a PDF of (U, V). Leave the constant c in your answer.
- (b) (4 pts) Find a PDF of V. Leave the constant c in your answer.
- (c) (2 pts) Find c using your answer in Part (b) and the fact that the integral of a PDF of V over $(-\infty, \infty)$ is 1.

Remark. The result from Problem 35(c) can be used for finding $\int_{-\infty}^{\infty} e^{-x^2/2} dx$. To see this, let $I = \int_{0}^{\infty} e^{-x^2/2} dx$, then

$$1 = \int_{R^2} f_{X,Y}(x,y) d(x,y) = cI^2,$$

so $I = 1/\sqrt{c}$ and $\int_{-\infty}^{\infty} e^{-x^2/2} dx = 2I = 2/\sqrt{c}$ (you should be able to obtain $2/\sqrt{c} = \sqrt{2\pi}$ if your answer for c is correct).

36. (10 pts) Let Γ be the function on $(0, \infty)$ defined by

$$\Gamma(a) = \int_0^\infty x^{a-1} e^{-x} dx$$

for a > 0. Suppose that α and β are two positive constants and (X, Y) has joint PDF $f_{X,Y}$, where

$$f_{X,Y}(x,y) = \frac{1}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} e^{-x} I_{(0,\infty)}(x) y^{\beta-1} e^{-y} I_{(0,\infty)}(y)$$

for $(x, y) \in \mathbb{R}^2$.

(a) (4 pts) Let M be the function on $(-\infty, 1) \times (-\infty, 1)$ defined by

$$M(t_1, t_2) = \left(\frac{1}{1-t_1}\right)^{\alpha} \left(\frac{1}{1-t_2}\right)^{\beta}$$

for $(t_1, t_2) \in (-\infty, 1) \times (-\infty, 1)$. Show that M is the joint MGF of (X, Y).

(b) (2 pts) Find the MGF of X.

(c) (4 pts) Find E(XY) and E(X).

- 37. (2 pts) Consider the (X, Y) in Problem 36. The distribution of X is called the gamma distribution with shape parameter α and scale parameter 1, denoted by $\Gamma(\alpha, 1)$. Show that the distribution of (X + Y) is $\Gamma(\alpha + \beta, 1)$. Hint: the MGF of (X+Y) can be easily obtained from the MGF of (X, Y).
- 38. (8 pts) Suppose that (X, Y) has a joint PDF $f_{X,Y}$, where

$$f_{X,Y}(x,y) = cI_S(x,y)$$

for $(x, y) \in \mathbb{R}^2$,

$$S = \{(x, y) : -2 < x + 2y < 2 \text{ and } -2 < x - 2y < 2\},\$$

and $c = 1 / \int_{B^2} I_S(x, y) d(x, y)$.

- (a) (4 pts) Find P((X + 2Y) > 0). You may leave c in your answer.
- (b) (4 pts) Find c.
- 39. (8 pts) Suppose that (X, Y) has a joint PDF $f_{X,Y}$, where

$$f_{X,Y}(x,y) = cI_S(x,y)$$

for $(x, y) \in \mathbb{R}^2$,

$$S = \{(x, y) : -2 < x + 2y < 2 \text{ and } -2 < x - 2y < 2\},\$$

and $c = 1 / \int_{R^2} I_S(x, y) d(x, y)$.

- (a) (4 pts) Find E(X|Y). You may leave c in your answer.
- (b) (4 pts) Find Var(X|Y). You may leave c in your answer.
- 40. (16 pts) Suppose that (X, Y) is a discrete random vector with PMF $p_{X,Y}$, where

$$P((X,Y) = (x,y)) = \begin{cases} 0.5 & \text{if } (x,y) = (1,2); \\ 0.4 & \text{if } (x,y) = (0,-3); \\ 0.1 & \text{if } (x,y) = (1,-3); \\ 0 & \text{otherwise.} \end{cases}$$

- (a) (8 pts) Find E(X|Y=y) and Var(X|Y=y) for $y \in \{2, -3\}$.
- (b) (8 pts) Find Var(E(X|Y)), E(Var(X|Y)) and Var(X). Verify the equality

$$Var(X) = Var(E(X|Y)) + E(Var(X|Y))$$

based on your answers.

- 41. (8 pts) Suppose that (X, Y) has joint PDF $f_{X,Y}$.
 - (a) (4 pts) For random variables g(X, Y) and h(Y), show that

$$E(g(X,Y)h(Y)|Y) = h(Y)E(g(X,Y)|Y)$$

(b) (4 pts) For random variables $g_1(X, Y)$ and $g_2(X, Y)$, show that

$$E((g_1(X,Y) + g_2(X,Y))|Y) = E(g_1(X,Y)|Y) + E(g_2(X,Y)|Y).$$

- 42. (16 pts) For $\mu \in R$ and $\sigma > 0$, let $f_{\mu,\sigma}$ be the PDF of $N(\mu, \sigma^2)$ given in Problem 30. Suppose that X and Y are random variables, Y has PDF $f_{0,1}$, and $\{f_{1+2y,1} : y \in R\}$ is a version of the conditional PDF of X given Y.
 - (a) (6 pts) Find a PDF of X.
 - (b) (6 pts) Find E(Y|X).
 - (c) (4 pts) Determine whether X and Y are independent. Justify your answer.
- 43. (6 pts) Suppose that (X, Y) has a joint PDF and X and Y are independent. Suppose that u and v are functions such that E(u(X)) and E(v(Y)) are finite. Show that

$$E(u(X)v(Y)) = E(u(X))E(v(Y)).$$

- 44. (6 pts) Consider the (X, Y) in Problem 36. Determine whether X + Y and X Y are independent based on the MGF of (X, Y).
- 45. (4 pts) Consider the (X, Y) in Problem 40. Determine whether X and Y are independent.
- 46. (4 pts) Suppose that X and Y are discrete random variables, X has m possible values x_1, \ldots, x_m , and Y has n possible values y_1, \ldots, y_n . Suppose that for $x \in \{x_1, \ldots, x_m\}$,

$$P(X = x | Y = y_1) = P(X = x | Y = y_2) = \dots = P(X = x | Y = y_n).$$

Show that X and Y are independent.

47. (8 pts) Suppose that Z_1 , Z_2 , Z_3 are independent random variables and $Z_i \sim N(0,1)$ for i = 1, 2, 3. Let

$$\begin{pmatrix} Y_1 \\ Y_2 \\ Y_3 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} Z_1 \\ Z_2 \\ Z_3 \end{pmatrix}.$$

- (a) (4 pts) Find a PDF of (Y_1, Y_2, Y_3) .
- (b) (4 pts) Determine whether Y_1 and (Y_2, Y_3) are independent. Justify your answer.
- 48. (6 pts) Suppose that X_1, \ldots, X_n are IID and $X_1 \sim N(\mu, \sigma^2)$. Let $\overline{X} = \sum_{i=1}^n X_i/n$ and $Y = (X_1 \overline{X}, \ldots, X_n \overline{X})^T$. Show that \overline{X} and Y are independent.
- 49. (12 pts) Suppose that $Z_1, ..., Z_m$ are IID and $Z_1 \sim N(0, 1)$. Let $U = \sum_{i=1}^{m} Z_i^2$.
 - (a) (6 pts) Show that $U/2 \sim \Gamma(m/2, 1)$. Note.
 - Recall that for $\alpha > 0$, the MGF of the gamma distribution $\Gamma(\alpha, 1)$ is the MGF of the random variable X in Problem 36.
 - To solve this problem, you may use the results in Problem 36.
 - For $\mu \in R$ and $\sigma > 0$, a PDF of $N(\mu, \sigma^2)$ is the $f_{\mu,\sigma}$ given in Problem 30.

(b) (6 pts) Let

$$f_U(x) = \frac{1}{2^{m/2} \Gamma(m/2)} x^{m/2-1} e^{-x/2} I_{(0,\infty)}(x)$$

for $x \in R$, where Γ is defined in Problem 36. Show that f_U is a PDF of U.

Hint: find a PDF of U/2 by finding a PDF of the random variable X in Problem 36, and then derive a PDF of U using the PDF of U/2.

Note. In Problem 49, the distirbution of U is called the chi-squared distribution with m degrees of freedom, denoted by $\chi^2(m)$.

50. (6 pts) Suppose that Z and W are independent random variables, $Z \sim N(0,1)$ and $W \sim \chi^2(m)$. Let

$$T = \frac{Z}{\sqrt{W/m}},$$

then the distribution of T is called the t distribution with m degrees of freedom, denoted by t(m). Find a PDF of T.

51. (6 pts) Suppose that (X, Y) is a vector of two discrete random variables with joint PMF $p_{X,Y}$, where

$$p_{X,Y}(x,y) = \begin{cases} 0.5 & \text{if } (x,y) = (1,2); \\ 0.4 & \text{if } (x,y) = (0,a); \\ 0.1 & \text{if } (x,y) = (-1,-2); \\ 0 & \text{otherwise}, \end{cases}$$

and a is a constant.

- (a) (4 pts) Express Corr(X, Y) as a function of a.
- (b) (2 pts) Find all a's such that |Corr(X, Y)| = 1.
- 52. (6 pts) Suppose that X and Y are random variables such that Var(X) and Var(Y) are both finite and Var(X) > 0. Define a function S by

$$S(a,b) = E(Y - (a+bX))^2$$

for $(a,b) \in \mathbb{R}^2$. Show that S is minimized when $(a,b) = (a_0,b_0)$, where $b_0 = Cov(X,Y)/Var(X)$ and $a_0 = E(Y) - b_0E(X)$. Also, show that

$$S(a_0, b_0) = \frac{Var(X)Var(Y) - (Cov(X, Y))^2}{Var(X)}.$$
 (4)

53. (4 pts) Suppose that X and Y are random variables, f_Y is a PDF of $Y, S_Y = \{y : f_Y(y) > 0\}$, and $\{f_{X|Y=y} : y \in S_Y\}$ is a version of the conditional PDF of X given Y. Then for g such that E(g(X,Y)) is finite, E(g(X,Y)|Y) can be obtained using

$$E(g(X,Y)|Y=y) = \int g(x,y)f_{X|Y=y}(x)dx \tag{5}$$

for all $y \in S_Y$. Use (5) to show that E(XY|Y) = YE(X|Y) when E(XY) and E(X) are finite.

54. (22 pts) Suppose that (X, Y, Z) is a random vector with joint PDF $f_{X,Y,Z}$, where

$$f_{X,Y,Z}(x,y,z) = ce^{-(x^2 + 4xy + 5y^2)} ze^{-z} I_{(0,\infty)}(z)$$

for $(x, y, z) \in \mathbb{R}^3$, and c > 0 is a constant.

- (a) (2 pts) Find a version of the conditional PDF of Z given (X, Y).
- (b) (6 pts) Find c, a PDF of Y, and a version of the conditional PDF of X given Y.
- (c) (6 pts) Find E(X|Y), $E(X^2|Y)$ and Var(X|Y).
- (d) (2 pts) Find the best linear predictor of X based on Y.
- (e) (8 pts) Find Cov(X, Y) and Corr(X, Y).
- 55. (4 pts) Suppose that X and Y are random variables such that $Y \sim N(0, 1)$ and a version of the conditional PDF of X given Y is $\{g_y : y \in R\}$, where g_y is the function $f_{\mu,\sigma}$ given in Problem 30 with $\mu = y$ and $\sigma = 1$. Find a version of the conditional PDF of Y given X.
- 56. (4 pts) Suppose that W is an $n \times m$ matrix of random variables, and B is an $m \times k$ non-random matrix. Show that E(WB) = E(W)B.
- 57. (4 pts) Suppose that $\mathbf{X} = (X_1, \ldots, X_m)^T$ is a random vector. Let Σ be the covariance matrix of \mathbf{X} . Suppose that A is a $n \times m$ non-random matrix. Show that the covariance matrix of $A\mathbf{X}$ is $A\Sigma A^T$. You may use the fact that for a random vector \mathbf{Z} , the covariance matrix of \mathbf{Z} is $E(\mathbf{Z}_*\mathbf{Z}_*^T)$, where $\mathbf{Z}_* = \mathbf{Z} E(\mathbf{Z})$.
- 58. (6 pts) Suppose that (X, Y) is a random vector with covariance matrix Σ , where

$$\Sigma = \left(\begin{array}{cc} 1 & a \\ a & 0.25 \end{array}\right)$$

and a is a constant.

- (a) (2 pts) Express Corr(X, Y) as a function of a.
- (b) (4 pts) Find a constant b such that Var(X bY) = 0 when a = 0.5.
- 59. (4 pts) Suppose that $\boldsymbol{X} = (X_1, X_2, X_3, X_4)^T$ is a random vector with covariance matrix Σ , where

Find $Var(X_1 + X_2 + X_3)$.

Hint: Apply Fact 2 in the handout "Covariance and correlation".

60. (3 pts) Suppose that U and X are random variables such that $E(X^2) < \infty$ and $E(U^2) < \infty$. Let $S(b) = E(U - bX)^2$, where b is a constant in R. Verify that

$$\frac{d}{db}S(b) = E\left[\frac{d}{db}(U-bX)^2\right].$$

Remark. The result in Problem 60 implies that

$$\frac{\partial}{\partial b_i} E[(Y - (a + b_1 X_1 + \dots + b_k X_k))^2] = E\left[\frac{\partial}{\partial b_i}(Y - (a + b_1 X_1 + \dots + b_k X_k))^2\right]$$

for $i \in \{1, \ldots, k\}$ and

$$\frac{\partial}{\partial a}E[(Y-(a+b_1X_1+\cdots+b_kX_k))^2] = E\left[\frac{\partial}{\partial a}(Y-(a+b_1X_1+\cdots+b_kX_k))^2\right].$$

61. (6 pts) Suppose that Z_1, \ldots, Z_n are independent random variables and $Z_i \sim N(0,1)$ for $i = 1, 2, \ldots, n$. Suppose that A is an $n \times n$ invertible matrix of constants in R and μ_1, \ldots, μ_n are constants in R. Let

$$\left(\begin{array}{c} Y_1\\ \vdots\\ Y_n \end{array}\right) = A \left(\begin{array}{c} Z_1\\ \vdots\\ Z_n \end{array}\right) + \left(\begin{array}{c} \mu_1\\ \vdots\\ \mu_n \end{array}\right)$$

Find a PDF of (Y_1, \ldots, Y_n) that is determined by AA^T and μ_1, \ldots, μ_n . You may use results from linear algebra such as $\det(B^T) = \det(B)$ and

$$\det(BC) = \det(B) \cdot \det(C)$$

for square matrices B and C. Here det(A) denotes the determinant of a square matrix A.

Note. In Problem 61, AA^T and $(\mu_1, \ldots, \mu_n)^T$ are the covariance matrix and the mean vector of $(Y_1, \ldots, Y_n)^T$, respectively.

62. (6 pts) Prove the following result:

Fact 3 Suppose that the distribution of $(X_1, \ldots, X_m, Y_1, \ldots, Y_n)^T$ is multivariate normal. If $Cov(X_i, Y_j) = 0$ for $1 \le i \le m, 1 \le j \le n$, then the two random vectors $(X_1, \ldots, X_m)^T$ and $(Y_1, \ldots, Y_n)^T$ are independent.

63. (6 pts) Suppose that

$$\left(\begin{array}{c} X\\Y\\Z\end{array}\right) \sim N\left(\left(\begin{array}{c} 0\\0\\0\end{array}\right), \left(\begin{array}{ccc} 10&2&5\\2&16&3\\5&3&25\end{array}\right)\right).$$

- (a) (3 pts) Find a constant b such that Y bX is independent of X.
- (b) (3 pts) Find constants c and d such that Z cY dX is independent of (X, Y).
- 64. (15 pts) Suppose that $\boldsymbol{X} = (X_1, X_2, X_3, X_4)^T$ is a random vector with $E(\boldsymbol{X}) = (0, 0, 0, 0)^T$ and covariance matrix Σ , where

$$\Sigma = \left(\begin{array}{rrrrr} 2 & -1 & 0 & 0 \\ -1 & 2 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 3 \end{array}\right)$$

Suppose that $\boldsymbol{X} \sim N(E(\boldsymbol{X}), \Sigma)$.

- (a) (3 pts) Find the best linear predictor of $(X_1, X_4)^T$ based on X_2 and X_3 .
- (b) (3 pts) Find a version of the conditional PDF of X_1 given X_2 .
- (c) (3 pts) Find a version of the conditional PDF of X_1 given $(X_2, X_3, X_4)^T$.
- (d) (3 pts) Find a version of the conditional PDF of $(X_1, X_2)^T$ given $(X_3, X_4)^T$.
- (e) (3 pts) Find $E(X_1X_2|X_3, X_4)$.
- 65. (6 pts) Suppose that \boldsymbol{Y} , \boldsymbol{X} and \boldsymbol{U} are random vectors in \mathbb{R}^m , \mathbb{R}^n , \mathbb{R}^m respectively such that \boldsymbol{X} and \boldsymbol{U} are independent and

$$\boldsymbol{Y} = \boldsymbol{g}(\boldsymbol{X}) + \boldsymbol{U} \tag{6}$$

for some function $g: \mathbb{R}^n \to \mathbb{R}^m$. Suppose that g is differentiable, X has a PDF f_X that is positive on \mathbb{R}^n , and U has a PDF f_U that is positive on \mathbb{R}^m . For $x \in \mathbb{R}^n$, define

$$f_{\boldsymbol{Y}|\boldsymbol{X}=\boldsymbol{x}}(\boldsymbol{y}) = f_{\boldsymbol{U}}(\boldsymbol{y} - g(\boldsymbol{x}))$$
(7)

for $\boldsymbol{y} \in \mathbb{R}^m$. Show that $\{f_{\boldsymbol{Y}|\boldsymbol{X}=\boldsymbol{x}} : \boldsymbol{x} \in \mathbb{R}^n\}$ is a version of the conditional PDF of \boldsymbol{Y} given \boldsymbol{X} .

- 66. (4 pts) Suppose that (X_1, \ldots, X_n) is a random sample from $N(\mu, \sigma^2)$, where $\mu \in R$ and $\sigma > 0$ are unknown. Let $\bar{X} = \sum_{i=1}^n X_i/n$, $\bar{Y} = \sum_{i=1}^n X_i^2/n$, and $\bar{Z} = \sum_{i=1}^n (X_i - \mu)^2/n$. Which of the following statements are true? You may write down your answers directly without justification.
 - (a) \overline{X} is a consistent estimator of μ .
 - (b) $\overline{Y} (\overline{X})^2$ is a consistent estimator of σ^2 .
 - (c) \bar{Z} is a consistent estimator of σ^2 .
 - (d) \overline{Z} converges to σ^2 in probability as $n \to \infty$.
- 67. (4 pts) Suppose that (X_1, \ldots, X_n) is a random sample from $U(0, \theta)$, where $\theta > 0$. Find a consistent estimator of $1/\theta$ and justify your answer.
- 68. (8 pts) Suppose that $((X_1, Y_1), \ldots, (X_n, Y_n))$ is a random sample from the distribution of a random vector (X, Y), where

$$Y = a + bX + \varepsilon,$$

a, b are constants, X and ε are independent, $E(\varepsilon) = 0$ and $Var(\varepsilon) = \sigma^2 < \infty$. Let \bar{X} and \bar{Y} be the sample means of (X_1, \ldots, X_n) and (Y_1, \ldots, Y_n) respectively, let

$$S_X^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2,$$
$$\hat{b} = \frac{\sum_{i=1}^n (X_i - \bar{X})(Y_i - \bar{Y})}{(n-1)S_X^2},$$

and $\hat{a} = \bar{Y} - \hat{b}\bar{X}$. Show that (\hat{a}, \hat{b}) is a consistent estimator of (a, b) assuming that $E(X^2) < \infty$, which implies that E(X) is finite.

69. (6 pts) Prove the following result.

Fact 4 Suppose that $\{X_{1,n}\}_{n=1}^{\infty}, \ldots, \{X_{k,n}\}_{n=1}^{\infty}$ are k sequences of random variables on the same probability space, and $X_{1,n} \xrightarrow{\mathcal{P}} Y_1, \ldots, X_{k,n} \xrightarrow{\mathcal{P}} Y_k$ for some random vector $(Y_1, \ldots, Y_k)^T$. Then

$$(X_{1,n},\ldots,X_{k,n})^T \xrightarrow{\mathcal{P}} (Y_1,\ldots,Y_k)^T.$$

Hint: make use of the equality:

$$||(X_{1,n},\ldots,X_{k,n})^T - (Y_1,\ldots,Y_k)^T||^2 = |X_{1,n} - Y_1|^2 + \cdots + |X_{k,n} - Y_k|^2.$$

70. (6 pts) Suppose that $\theta \in R$ is a parameter to be estimated, and $\hat{\theta}$ is an estimator of θ such that

 $E(\hat{\theta}) = \theta$

and

$$\lim_{n \to \infty} Var(\hat{\theta}) = 0$$

Show that $\hat{\theta}$ is a consistent estimator of θ . Hint: make use of Chebyshev's inequality:

$$P(|X - E(X)| > k) \le \frac{Var(X)}{L^2}$$

for a positive constant k > 0, or Markov's inequality: for a nonnegative random variable X,

$$P(X > M) \le \frac{E(X)}{M}$$

for a positive constant M.

71. (6 pts) Suppose that (X_1, \ldots, X_n) is a random sample from a discrete distribution with three possible values a_1, a_2, a_3 . Let

$$p_j = P(X_1 = a_j)$$

and

$$\hat{p}_j = \frac{1}{n} \sum_{i=1}^n I_{\{a_j\}}(X_i)$$

for j = 1, 2, 3. Let

$$Y_n = \sqrt{n} \left(\left(\begin{array}{c} \hat{p}_1 \\ \hat{p}_2 \\ \hat{p}_3 \end{array} \right) - \left(\begin{array}{c} p_1 \\ p_2 \\ p_3 \end{array} \right) \right).$$

Find the limiting distribution of Y_n . That is, find the distribution D_0 such that Y_n converges to D_0 in distribution as $n \to \infty$.

72. (6 pts) Consider the following result, which has been proved in class:

Fact 5 Suppose that $\boldsymbol{U} \sim N(\boldsymbol{0}, \Sigma)$ and $\Sigma^2 = \Sigma$. Then $\boldsymbol{U}^T \boldsymbol{U} \sim \chi^2(k)$, where $k = \text{trace}(\Sigma)$.

Suppose that Z_1, \ldots, Z_n are IID N(0,1) random variables. Let $\overline{Z} = \sum_{i=1}^n Z_i/n$. Apply Fact 5 to show that $\sum_{i=1}^n (Z_i - \overline{Z})^2 \sim \chi^2(n-1)$.

Hint: note that the distribution of $(Z_1 - \overline{Z}, \dots, Z_n - \overline{Z})$ is a multivariate normal distribution.

- 73. (12 pts) Consider the \hat{p}_1 in Problem 71. Suppose that $p_1 \in (0, 1)$.
 - (a) (3 pts) Find the limiting distribution of $\sqrt{n}(\hat{p}_1 p_1)/\sqrt{p_1(1-p_1)}$. Justify your answer.
 - (b) (3 pts) Find the limiting distribution of $n(\hat{p}_1 p_1)^2/(p_1(1 p_1))$. Justify your answer.
 - (c) (3 pts) Find the limiting distribution of $n(\hat{p}_1 p_1)^2/(\hat{p}_1(1 \hat{p}_1))$. Justify your answer.
 - (d) (3 pts) Suppose that $\{a_n\}_{n=1}^{\infty}$ is a sequence of postive constants such that $\lim_{n\to\infty} a_n/\sqrt{n} = 0$. Show that $a_n(\hat{p}_1 p_1)$ converges to 0 in distribution as $n \to \infty$.
- 74. (6 pts) Suppose that $\{X_n\}_{n=1}^{\infty}$ is a sequence of random variables and $X_n \xrightarrow{\mathcal{D}} c$ as $n \to \infty$, where c is a constant.
 - (a) (2 pts) Let F_n be the CDF of X_n . What can be said about $\lim_{n\to\infty} F_n(x)$ for $x \neq c$?
 - (b) (4 pts) Show that $X_n \xrightarrow{\mathcal{P}} c$ as $n \to \infty$ based on the result from Part (a).
- 75. (4 pts) Suppose that (X_1, \ldots, X_n) is a random sample from $N(\mu, \sigma^2)$. Let $\bar{Y} = \sum_{i=1}^n X_i^2/n$. Find the limiting distribution of $\sqrt{n}(\bar{Y} - E(X_1^2))$. Justify your answer.