

# Solutions to Probability Problems

Xiankang Wang

November 7, 2025

## Contents

<b>1 Problem 1: Event Probabilities</b>	<b>2</b>
1.1 Problem Statement . . . . .	2
1.2 Solution . . . . .	2
<b>2 Problem 2: Exponential Distribution Transformation</b>	<b>3</b>
2.1 Problem Statement . . . . .	3
2.2 Solution . . . . .	3
<b>3 Problem 3: Joint, Marginal, and Conditional Densities</b>	<b>4</b>
3.1 Problem Statement . . . . .	4
3.2 Solution . . . . .	4
<b>4 Problem 4: Functions of Exponential Variables</b>	<b>6</b>
4.1 Problem Statement . . . . .	6
4.2 Solution . . . . .	6
<b>5 Problem 5: Derived Discrete Variables</b>	<b>8</b>
5.1 Problem Statement . . . . .	8
5.2 Solution . . . . .	8

# 1 Problem 1: Event Probabilities

## 1.1 Problem Statement

Let  $A, B$  denote two events.  $\bar{A}$  is the complement set of  $A$ , and  $\bar{B}$  is the complement set of  $B$ . Given that  $P(A) = x$ ,  $P(B) = y$ ,  $P(A \cap B) = z$ , please find the probability of the following using  $x, y, z$ :

- (a)  $P(\bar{A} \cap B)$ .
- (b)  $P(\bar{A} \cap \bar{B})$ .
- (c)  $P(\bar{A} \cup B)$ .
- (d)  $P(\bar{A} \cup \bar{B})$ .

## 1.2 Solution

**(a)  $P(\bar{A} \cap B)$ .** This is the probability that event  $B$  occurs but event  $A$  does not. We can find this by subtracting the probability of both events occurring from the probability of event  $B$ .

$$P(\bar{A} \cap B) = P(B) - P(A \cap B) = \boxed{y - z}.$$

**(b)  $P(\bar{A} \cap \bar{B})$ .** This is the probability that neither  $A$  nor  $B$  occurs. By De Morgan's laws,  $\bar{A} \cap \bar{B} = \overline{A \cup B}$ .

$$P(\bar{A} \cap \bar{B}) = P(\overline{A \cup B}) = 1 - P(A \cup B).$$

Using the principle of inclusion-exclusion for  $P(A \cup B)$ :

$$P(A \cup B) = P(A) + P(B) - P(A \cap B) = x + y - z.$$

Therefore,

$$P(\bar{A} \cap \bar{B}) = 1 - (x + y - z) = \boxed{1 - x - y + z}.$$

**(c)  $P(\bar{A} \cup B)$ .** Using the probability addition rule:

$$P(\bar{A} \cup B) = P(\bar{A}) + P(B) - P(\bar{A} \cap B).$$

We know  $P(\bar{A}) = 1 - P(A) = 1 - x$ , and from part (a),  $P(\bar{A} \cap B) = y - z$ . Substituting these in:

$$P(\bar{A} \cup B) = (1 - x) + y - (y - z) = \boxed{1 - x + z}.$$

**(d)  $P(\bar{A} \cup \bar{B})$ .** By De Morgan's laws,  $\bar{A} \cup \bar{B} = \overline{A \cap B}$ .

$$P(\bar{A} \cup \bar{B}) = P(\overline{A \cap B}) = 1 - P(A \cap B) = \boxed{1 - z}.$$

## 2 Problem 2: Exponential Distribution Transformation

### 2.1 Problem Statement

Let random variable  $X$  have an exponential distribution with  $EX = 1$ . Let  $Y = X^2 - 2$ .

- (a) Find the moment generating function of  $X$ .
- (b) Find the probability density function of  $Y$ .
- (c) Find the mean and the variance of  $Y$ .

### 2.2 Solution

For an exponential distribution with parameter  $\lambda$ , the PDF is  $f(x) = \lambda e^{-\lambda x}$  and the expectation is  $EX = 1/\lambda$ . Given  $EX = 1$ , we have  $1/\lambda = 1$ , so  $\lambda = 1$ . Thus,  $X$  follows a standard exponential distribution with PDF  $f_X(x) = e^{-x}$  for  $x > 0$ .

**(a) Moment generating function (MGF) of  $X$ .** The MGF is defined as  $M_X(t) = E[e^{tX}]$ .

$$M_X(t) = \int_0^\infty e^{tx} f_X(x) dx = \int_0^\infty e^{tx} e^{-x} dx = \int_0^\infty e^{(t-1)x} dx.$$

For the integral to converge, we require  $t - 1 < 0$ , or  $t < 1$ .

$$M_X(t) = \left[ \frac{1}{t-1} e^{(t-1)x} \right]_0^\infty = 0 - \frac{1}{t-1} = \boxed{\frac{1}{1-t}}, \quad \text{for } t < 1.$$

**(b) Probability density function (PDF) of  $Y$ .** We first find the CDF of  $Y$ ,  $F_Y(y) = P(Y \leq y)$ , for  $y$  in the support of  $Y$ . Since  $X > 0$ ,  $X^2 > 0$ , so  $Y = X^2 - 2 > -2$ . For  $y > -2$ :

$$F_Y(y) = P(X^2 - 2 \leq y) = P(X^2 \leq y + 2).$$

Since  $X$  is always positive, this is equivalent to:

$$F_Y(y) = P(0 < X \leq \sqrt{y+2}) = F_X(\sqrt{y+2}).$$

The CDF of  $X \sim \text{Exp}(1)$  is  $F_X(x) = 1 - e^{-x}$  for  $x > 0$ . So,

$$F_Y(y) = 1 - e^{-\sqrt{y+2}}.$$

To find the PDF, we differentiate the CDF with respect to  $y$ :

$$f_Y(y) = \frac{d}{dy} F_Y(y) = \frac{d}{dy} (1 - e^{-\sqrt{y+2}}) = -e^{-\sqrt{y+2}} \cdot \left( -\frac{1}{2\sqrt{y+2}} \right).$$

$$f_Y(y) = \boxed{\frac{e^{-\sqrt{y+2}}}{2\sqrt{y+2}}}, \quad \text{for } y > -2.$$

**(c) Mean and variance of  $Y$ .** We use the properties of expectation and variance. For  $X \sim \text{Exp}(1)$ , the  $k$ -th moment is  $E[X^k] = k!$ . The mean of  $Y$ :

$$E[Y] = E[X^2 - 2] = E[X^2] - 2.$$

$$E[X^2] = 2! = 2.$$

$$E[Y] = 2 - 2 = \boxed{0}.$$

The variance of  $Y$ :

$$\text{Var}(Y) = \text{Var}(X^2 - 2) = \text{Var}(X^2) = E[(X^2)^2] - (E[X^2])^2 = E[X^4] - (E[X^2])^2.$$

$$E[X^4] = 4! = 24.$$

$$\text{Var}(Y) = 24 - (2)^2 = 24 - 4 = \boxed{20}.$$

### 3 Problem 3: Joint, Marginal, and Conditional Densities

#### 3.1 Problem Statement

The joint probability density function of  $X$  and  $Y$  is given by

$$f_{XY}(x, y) = \begin{cases} c(x^2 - y^2)e^{-x}, & 0 < x < \infty, -x < y < x \\ 0, & \text{otherwise.} \end{cases}$$

- (a) What is the value of  $c$ ?
- (b) Find the marginal probability density function of  $X$ .
- (c) Find the  $EY$ .
- (d) Find the conditional density function of  $Y$ , given  $X = x$ .

#### 3.2 Solution

**(a) Value of  $c$ .** The total probability must be 1.

$$\int_0^\infty \int_{-x}^x c(x^2 - y^2)e^{-x} dy dx = 1.$$

First, we solve the inner integral with respect to  $y$ :

$$\int_{-x}^x (x^2 - y^2) dy = \left[ x^2 y - \frac{y^3}{3} \right]_{-x}^x = \left( x^3 - \frac{x^3}{3} \right) - \left( -x^3 + \frac{x^3}{3} \right) = \frac{2x^3}{3} - \left( -\frac{2x^3}{3} \right) = \frac{4x^3}{3}.$$

Now, we solve the outer integral:

$$c \int_0^\infty \frac{4x^3}{3} e^{-x} dx = 1 \implies \frac{4c}{3} \int_0^\infty x^3 e^{-x} dx = 1.$$

The integral is the Gamma function  $\Gamma(4) = 3! = 6$ .

$$\frac{4c}{3} \cdot 6 = 1 \implies 8c = 1 \implies \boxed{c = \frac{1}{8}}.$$

**(b) Marginal PDF of  $X$ .** We integrate the joint PDF over all possible values of  $y$ .

$$f_X(x) = \int_{-x}^x f_{XY}(x, y) dy = \int_{-x}^x \frac{1}{8}(x^2 - y^2)e^{-x} dy.$$

This is  $\frac{1}{8}e^{-x}$  times the inner integral from part (a):

$$f_X(x) = \frac{1}{8}e^{-x} \left( \frac{4x^3}{3} \right) = \boxed{\frac{1}{6}x^3 e^{-x}}, \quad \text{for } x > 0.$$

**(c) Expectation of  $Y$ .**

$$E[Y] = \int_0^\infty \int_{-x}^x y f_{XY}(x, y) dy dx = \int_0^\infty \int_{-x}^x y \frac{1}{8}(x^2 - y^2)e^{-x} dy dx.$$

Consider the inner integral:

$$\int_{-x}^x y(x^2 - y^2) dy = \int_{-x}^x (x^2 y - y^3) dy.$$

The integrand  $g(y) = x^2 y - y^3$  is an odd function of  $y$ , and it is integrated over a symmetric interval  $[-x, x]$ . Therefore, the integral is 0.

$$E[Y] = \int_0^\infty \frac{1}{8}e^{-x}(0) dx = \boxed{0}.$$

**(d) Conditional PDF of  $Y$  given  $X = x$ .** The conditional density is given by  $f_{Y|X}(y|x) = \frac{f_{XY}(x,y)}{f_X(x)}$  for  $f_X(x) > 0$ .

$$f_{Y|X}(y|x) = \frac{\frac{1}{8}(x^2 - y^2)e^{-x}}{\frac{1}{6}x^3 e^{-x}}.$$

The term  $e^{-x}$  cancels out. The domain for  $y$  is  $-x < y < x$ .

$$f_{Y|X}(y|x) = \frac{1/8}{1/6} \cdot \frac{x^2 - y^2}{x^3} = \frac{6}{8} \frac{x^2 - y^2}{x^3} = \boxed{\frac{3}{4x^3}(x^2 - y^2)}, \quad \text{for } -x < y < x.$$

## 4 Problem 4: Functions of Exponential Variables

### 4.1 Problem Statement

Let  $Z$  be an exponentially distributed random variable with mean  $\beta$ .

- (a) Find the probability density function of  $X = -\ln \frac{Z}{\beta}$ .
- (b) Suppose  $X_1$  and  $X_2$  are two independent random variables following the same distribution as  $X$ . Find the probability density function of  $X_1 - X_2$ .
- (c) Suppose  $X_1, X_2, \dots, X_n$  are independent random variables following the same distribution as  $X$ . Show that  $\max(X_2, X_3, \dots, X_n)$  and  $X_1 + \ln(n-1)$  are identically distributed.

### 4.2 Solution

The PDF of  $Z \sim \text{Exp}(\text{mean} = \beta)$  is  $f_Z(z) = \frac{1}{\beta} e^{-z/\beta}$  for  $z > 0$ . The CDF is  $F_Z(z) = 1 - e^{-z/\beta}$ .

- (a) **PDF of  $X = -\ln(Z/\beta)$** . We find the CDF of  $X$  first.

$$\begin{aligned} F_X(x) &= P(X \leq x) = P(-\ln(Z/\beta) \leq x) = P(\ln(Z/\beta) \geq -x) \\ &= P(Z/\beta \geq e^{-x}) = P(Z \geq \beta e^{-x}). \end{aligned}$$

This is  $1 - F_Z(\beta e^{-x}) = 1 - (1 - e^{-(\beta e^{-x})/\beta}) = e^{-e^{-x}}$ . So, the CDF is  $F_X(x) = e^{-e^{-x}}$ . The PDF is the derivative of the CDF:

$$f_X(x) = \frac{d}{dx} \left( e^{-e^{-x}} \right) = e^{-e^{-x}} \cdot (-e^{-x}) \cdot (-1) = \boxed{e^{-x} e^{-e^{-x}}}.$$

This is the standard Gumbel distribution.

**(b) PDF of  $Y = X_1 - X_2$ .** Let  $U_i = Z_i/\beta \sim \text{Exp}(1)$ . Then  $X_i = -\ln U_i$ . So,  $Y = X_1 - X_2 = (-\ln U_1) - (-\ln U_2) = \ln(U_2/U_1)$ . We find the CDF of  $Y$ :  $F_Y(y) = P(Y \leq y) = P(\ln(U_2/U_1) \leq y) = P(U_2 \leq U_1 e^y)$ . We integrate over the distribution of  $U_1$ :

$$\begin{aligned} F_Y(y) &= \int_0^\infty P(U_2 \leq u_1 e^y) f_{U_1}(u_1) du_1 = \int_0^\infty (1 - e^{-u_1 e^y}) e^{-u_1} du_1 \\ &= \int_0^\infty (e^{-u_1} - e^{-(1+e^y)u_1}) du_1 = \left[ -e^{-u_1} \right]_0^\infty - \left[ \frac{e^{-(1+e^y)u_1}}{1+e^y} \right]_0^\infty \\ &= (1) - \left( \frac{1}{1+e^y} \right) = \frac{e^y}{1+e^y} = \frac{1}{1+e^{-y}}. \end{aligned}$$

This is the CDF of the standard logistic distribution. Differentiating to find the PDF:

$$f_Y(y) = \frac{d}{dy} \left( \frac{e^y}{1+e^y} \right) = \frac{e^y(1+e^y) - e^y(e^y)}{(1+e^y)^2} = \boxed{\frac{e^y}{(1+e^y)^2}}.$$

**(c) Show identical distribution.** Let  $V = X_1 + \ln(n - 1)$  and  $M = \max(X_2, \dots, X_n)$ . We show they have the same CDF. First, the CDF of  $V$ :

$$\begin{aligned} F_V(v) &= P(V \leq v) = P(X_1 + \ln(n - 1) \leq v) = P(X_1 \leq v - \ln(n - 1)). \\ &= F_X(v - \ln(n - 1)) = \exp(-\exp(-(v - \ln(n - 1)))) = \exp(-\exp(-v + \ln(n - 1))) \\ &= \exp(-e^{-v} \cdot e^{\ln(n-1)}) = \exp(-(n-1)e^{-v}). \end{aligned}$$

Next, the CDF of  $M$ :

$$F_M(m) = P(M \leq m) = P(\max(X_2, \dots, X_n) \leq m).$$

This is equivalent to  $P(X_2 \leq m, X_3 \leq m, \dots, X_n \leq m)$ . Due to independence and identical distribution:

$$\begin{aligned} F_M(m) &= P(X_2 \leq m) \cdots P(X_n \leq m) = [F_X(m)]^{n-1}. \\ &= (e^{-e^{-m}})^{n-1} = \exp(-(n-1)e^{-m}). \end{aligned}$$

Since  $F_V(x) = F_M(x)$  for all  $x$ , the two random variables are identically distributed.

## 5 Problem 5: Derived Discrete Variables

### 5.1 Problem Statement

Suppose the probability density function of random variable  $Y$  is  $f_Y(y) = e^{-y}$  for  $y > 0$ . Define the random variables  $X_i$  as follows:

$$X_i = \begin{cases} 1, & Y > i \\ 0, & Y \leq i \end{cases} \quad \text{for } i = 1, 2.$$

- (a) Let  $Z = e^{X_1}$ . Find the probability mass function of  $Z$ .
- (b) Find the joint probability mass function of  $(X_1, X_2)$ .

### 5.2 Solution

The given distribution for  $Y$  is standard exponential,  $Y \sim \text{Exp}(1)$ . We have  $P(Y > y) = e^{-y}$  and  $P(Y \leq y) = 1 - e^{-y}$ .

**(a) PMF of  $Z = e^{X_1}$ .** First, we find the PMF of  $X_1$ .  $X_1$  can take values 0 or 1.

$$P(X_1 = 1) = P(Y > 1) = e^{-1}.$$

$$P(X_1 = 0) = P(Y \leq 1) = 1 - e^{-1}.$$

Now, we find the possible values of  $Z$  and their probabilities.

- If  $X_1 = 0$ , then  $Z = e^0 = 1$ . This occurs with probability  $P(X_1 = 0) = 1 - e^{-1}$ .
- If  $X_1 = 1$ , then  $Z = e^1 = e$ . This occurs with probability  $P(X_1 = 1) = e^{-1}$ .

So, the PMF of  $Z$ , denoted  $p_Z(z)$ , is:

$$p_Z(z) = \begin{cases} 1 - e^{-1}, & z = 1 \\ e^{-1}, & z = e \\ 0, & \text{otherwise.} \end{cases}$$

**(b) Joint PMF of  $(X_1, X_2)$ .** The vector  $(X_1, X_2)$  can take four possible values:  $(0, 0), (0, 1), (1, 0), (1, 1)$ . We compute the probability for each case.

- $P(X_1 = 0, X_2 = 0) = P(Y \leq 1 \text{ and } Y \leq 2) = P(Y \leq 1) = [1 - e^{-1}]$ .
- $P(X_1 = 0, X_2 = 1) = P(Y \leq 1 \text{ and } Y > 2)$ . These are disjoint events, so the probability is  $[0]$ .
- $P(X_1 = 1, X_2 = 0) = P(Y > 1 \text{ and } Y \leq 2) = P(1 < Y \leq 2) = P(Y \leq 2) - P(Y \leq 1) = (1 - e^{-2}) - (1 - e^{-1}) = [e^{-1} - e^{-2}]$ .
- $P(X_1 = 1, X_2 = 1) = P(Y > 1 \text{ and } Y > 2) = P(Y > 2) = [e^{-2}]$ .

The joint PMF  $p(x_1, x_2)$  can be summarized in a table:

		$X_2 = 0$	$X_2 = 1$
		$X_1 = 0$	$1 - e^{-1}$
$X_1 = 1$	$e^{-1} - e^{-2}$	$e^{-2}$	