## Solutions to Review Problems for Exam 2

1. Let $f_{X}(x)=\left\{\begin{array}{ll}\frac{1}{x^{2}} & \text { if } 1<x<\infty ; \\ 0 & \text { if } x \leq 1,\end{array}\right.$ be the pdf of a random variable $X$. If $E_{1}$ denote the interval $(1,2)$ and $E_{2}$ the interval $(4,5)$, compute $\operatorname{Pr}\left(E_{1}\right), \operatorname{Pr}\left(E_{2}\right)$, $\operatorname{Pr}\left(E_{1} \cup E_{2}\right)$ and $\operatorname{Pr}\left(E_{1} \cap E_{2}\right)$.
Solution: Compute

$$
\begin{gathered}
\operatorname{Pr}\left(E_{1}\right)=\int_{1}^{2} \frac{1}{x^{2}} d x=-\left.\frac{1}{x}\right|_{1} ^{2}=\frac{1}{2} \\
\operatorname{Pr}\left(E_{2}\right)=\int_{4}^{5} \frac{1}{x^{2}} d x=-\left.\frac{1}{x}\right|_{4} ^{5}=\frac{1}{20} \\
\operatorname{Pr}\left(E_{1} \cup E_{2}\right)=\operatorname{Pr}\left(E_{1}\right)+\operatorname{Pr}\left(E_{2}\right)=\frac{11}{20}
\end{gathered}
$$

since $E_{1}$ and $E_{2}$ are mutually exclusive, and

$$
\operatorname{Pr}\left(E_{1} \cap E_{2}\right)=0
$$

since $E_{1}$ and $E_{2}$ are mutually exclusive.
2. Let $X$ have pdf $f_{X}(x)= \begin{cases}2 x, & \text { if } 0<x<1 ; \\ 0, & \text { elsewhere }\end{cases}$

Compute the probability that $X$ is at least $3 / 4$, given that $X$ is at least $1 / 2$.
Solution: We are asked to compute

$$
\begin{equation*}
\operatorname{Pr}(X \geqslant 3 / 4 \mid X \geqslant 1 / 2)=\frac{\operatorname{Pr}[(X \geqslant 3 / 4) \cap(X \geqslant 1 / 2)]}{\operatorname{Pr}(X \geqslant 1 / 2)} \tag{1}
\end{equation*}
$$

where

$$
\begin{aligned}
\operatorname{Pr}(X \geqslant 1 / 2) & =\int_{1 / 2}^{1} 2 x d x \\
& =\left.x^{2}\right|_{1 / 2} ^{1} \\
& =1-\frac{1}{4}
\end{aligned}
$$

so that

$$
\begin{equation*}
\operatorname{Pr}(X \geqslant 1 / 2)=\frac{3}{4} \tag{2}
\end{equation*}
$$

and

$$
\begin{aligned}
\operatorname{Pr}[(X \geqslant 3 / 4) \cap(X \geqslant 1 / 2)] & =\operatorname{Pr}(X \geqslant 3 / 4) \\
& =\int_{3 / 4}^{1} 2 x d x \\
& =\left.x^{2}\right|_{3 / 4} ^{1} \\
& =1-\frac{9}{16}
\end{aligned}
$$

so that

$$
\begin{equation*}
\operatorname{Pr}[(X \geqslant 3 / 4) \cap(X \geqslant 1 / 2)]=\frac{7}{16} \tag{3}
\end{equation*}
$$

Substituting (3) and (2) into (1) then yields

$$
\operatorname{Pr}(X \geqslant 3 / 4 \mid X \geqslant 1 / 2)=\frac{\frac{7}{16}}{\frac{3}{4}}=\frac{7}{12}
$$

3. Divide a segment at random into two parts. Find the probability that the largest segment is at least three times the shorter.
Solution: Assume the segment is the interval $(0,1)$ and let $X \sim \operatorname{Uniform}(0,1)$. Then $X$ models a random point in $(0,1)$. Let $A$ denote the event that the longer of $X$ or $1-X$ is at least three times the length of the shorter of the two.
We have two possibilities: Either $X \leqslant 1-X$ or $X>1-X$; or, equivalently, $X \leqslant \frac{1}{2}$ or $X>\frac{1}{2}$. Define the events $E_{1}=\left(X \leqslant \frac{1}{2}\right)$ and $E_{2}=\left(X>\frac{1}{2}\right)$. Observe that $E_{1}$ and $E_{2}$ are mutually exclusive. Observe also that if $E_{1}$ occurs, then $A$ corresponds to $1-X \geqslant 3 X$, or $X \leqslant 1 / 4$; and, if $E_{2}$ occurs, then $A$ corresponds to $X \geqslant 3(1-X)$, or $X \geqslant 3 / 4$. Thus, by the Law of Total Probability,

$$
\begin{aligned}
\operatorname{Pr}(A) & =\operatorname{Pr}\left(A \cap E_{1}\right)+\operatorname{Pr}\left(A \cap E_{2}\right) \\
& =\operatorname{Pr}\left(X \leqslant \frac{1}{2}, X \leqslant \frac{1}{4}\right)+\operatorname{Pr}\left(X>\frac{1}{2}, X \geqslant \frac{3}{4}\right),
\end{aligned}
$$

so that

$$
\operatorname{Pr}(A)=\frac{1}{4}+\frac{1}{4}=\frac{1}{2}
$$

Thus, the probability that the largest segment is at least three times the shorter is $1 / 2$.
4. Let $X$ have pdf $f_{X}(x)= \begin{cases}x^{2} / 9, & \text { if } 0<x<3 ; \\ 0, & \text { elsewhere. }\end{cases}$

Find the pdf of $Y=X^{3}$.
Solution: First, compute the cdf of $Y$

$$
\begin{equation*}
F_{Y}(y)=\operatorname{Pr}(Y \leqslant y) . \tag{4}
\end{equation*}
$$

Observe that, since $Y=X^{3}$ and the possible values of $X$ range from 0 to 3 , the values of $Y$ will range from 0 to 27 . Thus, in the calculations that follow, we will assume that $0<y<27$.
From (4) we get that

$$
\begin{aligned}
F_{Y}(y) & =\operatorname{Pr}\left(X^{3} \leqslant y\right) \\
& =\operatorname{Pr}\left(X \leqslant y^{1 / 3}\right) \\
& =F_{X}\left(y^{1 / 3}\right)
\end{aligned}
$$

Thus, for $0<y<27$, we have that

$$
\begin{equation*}
f_{Y}(y)=f_{X}\left(y^{1 / 3}\right) \cdot \frac{1}{3} y^{-3 / 2} \tag{5}
\end{equation*}
$$

where we have applied the Chain Rule. It follows from (5) and the definition of $f_{X}$ that

$$
\begin{equation*}
f_{Y}(y)=\frac{1}{9}\left[y^{1 / 3}\right]^{2} \cdot \frac{1}{3} y^{-3 / 2}=\frac{1}{27}, \quad \text { for } 0<y<27 . \tag{6}
\end{equation*}
$$

Combining (6) and the definition of $f_{X}$ we obtain the pdf for $Y$ :

$$
f_{Y}(y)= \begin{cases}\frac{1}{27}, & \text { for } 0<y<27 \\ 0 & \text { elsewhere }\end{cases}
$$

in other words $Y \sim \operatorname{Uniform}(0,27)$.
5. Let $X$ and $Y$ be independent $\operatorname{Normal}(0,1)$ random variables. Put $Z=\frac{Y}{X}$. Compute the distribution functions $F_{Z}(z)$ and $f_{Z}(z)$.
Solution: Since $X, Y \sim \operatorname{Normal}(0,1)$, their pdfs are given by

$$
f_{X}(x)=\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2}, \quad \text { for } x \in \mathbb{R}
$$

and

$$
f_{Y}(y)=\frac{1}{\sqrt{2 \pi}} e^{-y^{2} / 2}, \quad \text { for } y \in \mathbb{R}
$$

respectively. The joint pdf of $(X, Y)$ is then

$$
\begin{equation*}
f_{(X, Y)}(x, y)=\frac{1}{2 \pi} e^{-\left(x^{2}+y^{2}\right) / 2}, \quad \text { for }(x, y) \in \mathbb{R}^{2} \tag{7}
\end{equation*}
$$

We compute the cdf of $Z$,

$$
F_{z}(z)=\operatorname{Pr}(Z \leqslant z)=\operatorname{Pr}\left(\frac{y}{x} \leqslant z\right)
$$

or

$$
\begin{equation*}
F_{Z}(z)=\iint_{\frac{y}{x} \leqslant z} f_{(X, Y)}(x, y) d x d y \tag{8}
\end{equation*}
$$

where the integrand in (8) is given in (7) and the integration is done over the region

$$
R=\left\{(x, y) \in \mathbb{R}^{2} \left\lvert\, \frac{y}{x} \leqslant z\right.\right\} .
$$

Make the change variables

$$
\begin{aligned}
u & =x \\
v & =\frac{y}{x},
\end{aligned}
$$

so that

$$
\begin{align*}
& x=u \\
& y=u v, \tag{9}
\end{align*}
$$

in the integral in (8) to obtain

$$
\begin{equation*}
F_{Z}(z)=\int_{-\infty}^{z} \int_{-\infty}^{\infty} f_{(X, Y)}(u, u v)\left|\frac{\partial(x, y)}{\partial(u, v)}\right| d u d v \tag{10}
\end{equation*}
$$

where

$$
\frac{\partial(x, y)}{\partial(u, v)}=\operatorname{det}\left(\begin{array}{ll}
1 & 0  \tag{11}\\
v & u
\end{array}\right)=u
$$

is the Jacobian determinant of the transformation in (9). It then follows from (10) and (11) that

$$
\begin{equation*}
F_{Z}(z)=\int_{-\infty}^{z} \int_{-\infty}^{\infty} f_{(X, Y)}(u, u v)|u| d u d v \tag{12}
\end{equation*}
$$

Differentiating with respect to $z$ and using the definition of the joint pdf of $(X, Y)$ in (7) we obtain from (12) that

$$
\begin{equation*}
f_{Z}(z)=\frac{1}{2 \pi} \int_{-\infty}^{\infty}|u| e^{-\left(1+z^{2}\right) u^{2} / 2} d u \tag{13}
\end{equation*}
$$

where we have also used the Fundamental Theorem of Calculus.
Since the integrand in (13) is an even function of $u$, we can rewrite the expression for $f_{z}$ in (13) as

$$
\begin{equation*}
f_{z}(z)=\frac{1}{\pi} \int_{0}^{\infty} u e^{-\left(1+z^{2}\right) u^{2} / 2} d u \tag{14}
\end{equation*}
$$

Integrating the right-hand side of equation in (14) we obtain

$$
\begin{equation*}
f_{z}(z)=\frac{1}{\pi} \cdot \frac{1}{1+z^{2}}, \quad \text { for } z \in \mathbb{R} \tag{15}
\end{equation*}
$$

The cdf of $Z$ is then obtained by integrating (15) to get

$$
F_{z}(z)=\int_{-\infty}^{z} f_{Z}(z) d z=\frac{1}{2}+\frac{1}{\pi} \arctan (z), \quad \text { for } z \in \mathbb{R}
$$

6. A random point $(X, Y)$ is distributed uniformly on the square with vertices $(-1,-1),(1,-1),(1,1)$ and $(-1,1)$.
(a) Give the joint pdf for $X$ and $Y$.
(b) Compute the following probabilities:
(i) $\operatorname{Pr}\left(X^{2}+Y^{2}<1\right)$,
(ii) $\operatorname{Pr}(2 X-Y>0)$,
(iii) $\operatorname{Pr}(|X+Y|<2)$.

Solution: The square is pictured in Figure 1 and has area 4.


Figure 1: Sketch of square in Problem 6
(a) Consequently, the joint pdf of $(X, Y)$ is given by

$$
f_{(X, Y)}(x, y)= \begin{cases}\frac{1}{4}, & \text { for }-1<x<1,-1<y<1  \tag{16}\\ 0 & \text { elsewhere }\end{cases}
$$

(b) Denoting the square in Figure 1 by $R$, it follows from (16) that, for any subset $A$ of $\mathbb{R}^{2}$,

$$
\begin{equation*}
\operatorname{Pr}[(x, y) \in A]=\iint_{A} f_{(X, Y)}(x, y) d x d y=\frac{1}{4} \cdot \operatorname{area}(A \cap R) \tag{17}
\end{equation*}
$$

that is, $\operatorname{Pr}[(x, y) \in A]$ is one-fourth the area of the portion of $A$ in $R$.
We will use the formula in (17) to compute each of the probabilities in (i), (ii) and (iii).
(i) In this case, $A$ is the circle of radius 1 around the origin in $\mathbb{R}^{2}$ and pictured in Figure 2.
Note that the circle $A$ in Figure 2 is entirely contained in the square $R$ so that, by the formula in (17),

$$
\operatorname{Pr}\left(X^{2}+Y^{2}<1\right)=\frac{\operatorname{area}(A)}{4}=\frac{\pi}{4}
$$



Figure 2: Sketch of $A$ in Problem 6(i)
(ii) The set $A$ in this case is pictured in Figure 3 on page 8. Thus, in this case, $A \cap R$ is a trapezoid of area $2 \cdot \frac{\frac{1}{2}+\frac{3}{2}}{2}=2$, so that, by the formula in (17),

$$
\operatorname{Pr}(2 X-Y>0)=\frac{1}{4} \cdot \operatorname{area}(A \cap R)=\frac{1}{2}
$$

(iii) In this case, $A$ is the region in the $x y$-plane between the lines $x+y=2$ and $x+y=-2$ (see Figure 4 on page 9 ). Thus, $A \cap R$ is $R$ so that, by the formula in (17),

$$
\operatorname{Pr}(|X+Y|<2)=\frac{\operatorname{area}(R)}{4}=1
$$

7. Prove that if the joint cdf of $X$ and $Y$ satisfies

$$
F_{(X, Y)}(x, y)=F_{X}(x) F_{Y}(y),
$$

then for any pair of intervals $(a, b)$ and $(c, d)$,

$$
\operatorname{Pr}(a<X \leq b, c<Y \leq d)=\operatorname{Pr}(a<X \leqslant b) \operatorname{Pr}(c<Y \leqslant d)
$$



Figure 3: Sketch of $A$ in Problem 6(ii)

Solution: First show that
$\operatorname{Pr}(a<X \leq b, c<Y \leq d)=F_{(X, Y)}(b, d)-F_{(X, Y)}(b, c)-F_{(X, Y)}(a, d)+F_{(X, Y)}(a, c)$ (see Problem 1 in Assignment \#15). Then,

$$
\begin{aligned}
\operatorname{Pr}(a<X \leq b, c<Y \leq d)= & F_{X}(b) F_{Y}(d)-F_{X}(b) F_{Y}(c) \\
& -F_{X}(a) F_{Y}(d)+F_{X}(a) F_{Y}(c) \\
= & \left(F_{X}(b)-F_{X}(a)\right) F_{Y}(d) \\
& -\left(F_{X}(b)-F_{X}(a)\right) F_{Y}(c) \\
= & \left(F_{X}(b)-F_{X}(a)\right)\left(F_{Y}(d)-F_{Y}(c)\right) \\
= & \operatorname{Pr}(a<X \leqslant b) \operatorname{Pr}(c<Y \leqslant d)
\end{aligned}
$$

which was to be shown.
8. The random pair $(X, Y)$ has the joint distribution shown in Table 1 on page 9.


Figure 4: Sketch of $A$ in Problem 6(iii)

| $X \backslash Y$ | 2 | 3 | 4 |
| :---: | :---: | :---: | :---: |
| 1 | $\frac{1}{12}$ | $\frac{1}{6}$ | 0 |
| 2 | $\frac{1}{6}$ | 0 | $\frac{1}{3}$ |
| 3 | $\frac{1}{12}$ | $\frac{1}{6}$ | 0 |

Table 1: Joint Probability Distribution for $(X, Y), p_{(X, Y)}$, in Problem 8
(a) Show that $X$ and $Y$ are not independent.

Solution: Table 2 shows the marginal distributions of $X$ and $Y$ on the margins on page 10.
Observe from Table 2 that

$$
p_{(X, Y)}(1,4)=0
$$

while

$$
p_{X}(1)=\frac{1}{4} \quad \text { and } \quad p_{Y}(4)=\frac{1}{3}
$$

Thus,

$$
p_{X}(1) \cdot p_{Y}(4)=\frac{1}{12}
$$

| $X \backslash Y$ | 2 | 3 | 4 | $p_{X}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1 | $\frac{1}{12}$ | $\frac{1}{6}$ | 0 | $\frac{1}{4}$ |
| 2 | $\frac{1}{6}$ | 0 | $\frac{1}{3}$ | $\frac{1}{2}$ |
| 3 | $\frac{1}{12}$ | $\frac{1}{6}$ | 0 | $\frac{1}{4}$ |
| $p_{Y}$ | $\frac{1}{3}$ | $\frac{1}{3}$ | $\frac{1}{3}$ | 1 |

Table 2: Joint pdf for $X$ and $Y$ and marginal distributions $p_{X}$ and $p_{Y}$ so that

$$
p_{(X, Y)}(1,4) \neq p_{X}(1) \cdot p_{Y}(4),
$$

and, therefore, $X$ and $Y$ are not independent.
(b) Give a probability table for random variables $U$ and $V$ that have the same marginal distributions as $X$ and $Y$, respectively, but are independent.
Solution: Table 3 on page 10 shows the joint pmf of $(U, V)$ and the marginal distributions, $p_{U}$ and $p_{V}$.

| $U \backslash V$ | 2 | 3 | 4 | $p_{U}$ |
| :---: | :---: | :---: | :---: | :---: |
| 1 | $\frac{1}{12}$ | $\frac{1}{12}$ | $\frac{1}{12}$ | $\frac{1}{4}$ |
| 2 | $\frac{1}{6}$ | $\frac{1}{6}$ | $\frac{1}{6}$ | $\frac{1}{2}$ |
| 3 | $\frac{1}{12}$ | $\frac{1}{12}$ | $\frac{1}{12}$ | $\frac{1}{4}$ |
| $p_{V}$ | $\frac{1}{3}$ | $\frac{1}{3}$ | $\frac{1}{3}$ | 1 |

Table 3: Joint pdf for $U$ and $V$ and their marginal distributions.
9. Let $X$ denote the number of trials needed to obtain the first head, and let $Y$ be the number of trials needed to get two heads in repeated tosses of a fair coin. Are $X$ and $Y$ independent random variables?
Solution: $X$ has a geometric distribution with parameter $p=\frac{1}{2}$, so that

$$
\begin{equation*}
p_{X}(k)=\frac{1}{2^{k}}, \quad \text { for } k=1,2,3, \ldots \tag{18}
\end{equation*}
$$

On the other hand,

$$
\begin{equation*}
\operatorname{Pr}[Y=2]=\frac{1}{4} \tag{19}
\end{equation*}
$$

since, in two repeated tosses of a coin, the events are $H H, H T, T H$ and $T T$, and these events are equally likely.

Next, consider the joint event $(X=2, Y=2)$. Note that

$$
(X=2, Y=2)=[X=2] \cap[Y=2]=\emptyset
$$

since $[X=2]$ corresponds to the event $T H$, while $[Y=2]$ to the event $H H$. Thus,

$$
\operatorname{Pr}(X=2, Y=2)=0
$$

while

$$
p_{X}(2) \cdot p_{Y}(2)=\frac{1}{4} \cdot \frac{1}{4}=\frac{1}{16},
$$

by (18) and (19). Thus,

$$
p_{(X, Y)}(2,2) \neq p_{X}(2) \cdot p_{X}(2) .
$$

Hence, $X$ and $Y$ are not independent.
10. Let $X \sim \operatorname{Normal}(0,1)$ and put $Y=X^{2}$. Find the pdf for $Y$.

Solution: The pdf of $X$ is given by

$$
f_{X}(x)=\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2}, \quad \text { for }-\infty<x<\infty
$$

We compute the pdf for $Y$ by first determining its cdf:

$$
\begin{aligned}
\operatorname{Pr}(Y \leqslant y) & =P\left(X^{2} \leqslant y\right) \quad \text { for } y \geqslant 0 \\
& =\operatorname{Pr}(-\sqrt{y} \leqslant X \leqslant \sqrt{y}) \\
& =\operatorname{Pr}(-\sqrt{y}<X \leqslant \sqrt{y}), \quad \text { since } X \text { is continuous. }
\end{aligned}
$$

Thus,

$$
\begin{aligned}
\operatorname{Pr}(Y \leqslant y) & =\operatorname{Pr}(X \leqslant \sqrt{y})-\operatorname{Pr}(X \leqslant-\sqrt{y}) \\
& =F_{X}(\sqrt{y})-F_{X}(-\sqrt{y}) \text { for } y>0 .
\end{aligned}
$$

We then have that the cdf of $Y$ is

$$
F_{Y}(y)=F_{X}(\sqrt{y})-F_{X}(-\sqrt{y}) \quad \text { for } y>0
$$

from which we get, after differentiation with respect to $y$,

$$
\begin{aligned}
f_{Y}(y) & =F_{X}^{\prime}(\sqrt{y}) \cdot \frac{1}{2 \sqrt{y}}+F_{X}^{\prime}(-\sqrt{y}) \cdot \frac{1}{2 \sqrt{y}} \\
& =f_{X}(\sqrt{y}) \frac{1}{2 \sqrt{y}}+f_{X}(-\sqrt{y}) \frac{1}{2 \sqrt{y}} \\
& =\frac{1}{2 \sqrt{y}}\left\{\frac{1}{\sqrt{2 \pi}} e^{-y / 2}+\frac{1}{\sqrt{2 \pi}} e^{-y / 2}\right\} \\
& =\frac{1}{\sqrt{2 \pi}} \cdot \frac{1}{\sqrt{y}} e^{-y / 2},
\end{aligned}
$$

for $y>0$, where we have applied the Chain Rule. Hence,

$$
f_{Y}(y)= \begin{cases}\frac{1}{\sqrt{2 \pi}} \cdot \frac{1}{\sqrt{y}} e^{-y / 2}, & \text { for } y>0 \\ 0 & \text { for } y \leqslant 0\end{cases}
$$

11. Let $X$ and $Y$ be independent $\operatorname{Normal}(0,1)$ random variables. Compute

$$
P\left(X^{2}+Y^{2}<1\right)
$$

Solution: Since $X, Y \sim \operatorname{Normal}(0,1)$, their pdfs are given by

$$
f_{X}(x)=\frac{1}{\sqrt{2 \pi}} e^{-x^{2} / 2}, \quad \text { for } x \in \mathbb{R}
$$

and

$$
f_{Y}(y)=\frac{1}{\sqrt{2 \pi}} e^{-y^{2} / 2}, \quad \text { for } y \in \mathbb{R}
$$

respectively. The joint pdf of $(X, Y)$ is then

$$
\begin{equation*}
f_{(X, Y)}(x, y)=\frac{1}{2 \pi} e^{-\left(x^{2}+y^{2}\right) / 2}, \quad \text { for }(x, y) \in \mathbb{R}^{2} \tag{20}
\end{equation*}
$$

Thus,

$$
\begin{equation*}
P\left(X^{2}+Y^{2}<1\right)=\iint_{x^{2}+y^{2}<1} f_{(X, Y)}(x, y) d x d y \tag{21}
\end{equation*}
$$

where the integrand is given in (20) and the integral in (21) is evaluated over the disc of radius 1 centered around the origin in $\mathbb{R}^{2}$.
We evaluate the integral in (21) by changing to polar coordinates to get

$$
\begin{aligned}
P\left(X^{2}+Y^{2}<1\right) & =\frac{1}{2 \pi} \int_{0}^{2 \pi} \int_{0}^{1} e^{-r^{2} / 2} r d r d \theta \\
& =\int_{0}^{1} e^{-r^{2} / 2} r d r \\
& =\left[-e^{-r^{2} / 2}\right]_{0}^{1} \\
& =1-e^{-1 / 2}
\end{aligned}
$$

or $\operatorname{Pr}\left(X^{2}+Y^{2}<1\right)=1-\frac{1}{\sqrt{e}}$.
12. Suppose that $X$ and $Y$ are independent random variables such that $X \sim$ $\operatorname{Uniform}(0,1)$ and $Y \sim \operatorname{Exponential}(1)$.
(a) Let $Z=X+Y$. Find $F_{Z}$ and $f_{Z}$.

Solution: Since $X \sim \operatorname{Uniform}(0,1)$ and $Y \sim \operatorname{Exponential(1),~their~pdfs~}$ are given by

$$
f_{X}(x)= \begin{cases}1 & \text { if } 0<x<1 \\ 0 & \text { elsewhere }\end{cases}
$$

and

$$
f_{Y}(y)= \begin{cases}e^{-y} & \text { if } y>0 \\ 0 & \text { if } y \leqslant 0\end{cases}
$$

respectively. The joint pdf of $(X, Y)$ is then

$$
f_{(X, Y)}(x, y)= \begin{cases}e^{-y} & \text { if } 0<x<1, y>0  \tag{22}\\ 0 & \text { elsewhere }\end{cases}
$$

We compute the cdf of $Z$,

$$
F_{z}(z)=\operatorname{Pr}(X \leqslant u)=\operatorname{Pr}(X+Y \leqslant z)
$$

or

$$
\begin{equation*}
F_{U}(u)=\iint_{x+y \leqslant z} f_{(X, Y)}(x, y) d x d y \tag{23}
\end{equation*}
$$

where the integrand in (23) is given in (22) and the integration is done over the region

$$
R=\left\{(x, y) \in \mathbb{R}^{2} \mid x+y \leqslant z\right\}
$$

Make the change variables

$$
\begin{aligned}
& u=x \\
& v=x+y,
\end{aligned}
$$

so that

$$
\begin{align*}
& x=u  \tag{24}\\
& y=v-u,
\end{align*}
$$

in the integral in (23) to obtain

$$
\begin{equation*}
F_{Z}(z)=\int_{-\infty}^{z} \int_{-\infty}^{\infty} f_{(X, Y)}(u, v-u)\left|\frac{\partial(x, y)}{\partial(u, v)}\right| d u d v \tag{25}
\end{equation*}
$$

where

$$
\frac{\partial(x, y)}{\partial(u, v)}=\operatorname{det}\left(\begin{array}{cc}
1 & 0  \tag{26}\\
-1 & 1
\end{array}\right)=1
$$

is the Jacobian determinant of the transformation in (24). It then follows from (25) and (26) that

$$
\begin{equation*}
F_{Z}(z)=\int_{-\infty}^{z} \int_{-\infty}^{\infty} f_{(X, Y)}(u, v-u) d u d v \tag{27}
\end{equation*}
$$

Differentiating with respect to $z$ and using the definition of the joint pdf of $(X, Y)$ in (22) we obtain from (27) that

$$
\begin{equation*}
f_{Z}(z)=\int_{-\infty}^{\infty} f_{(X, Y)}(u, z-u) d u \tag{28}
\end{equation*}
$$

where we have also used the Fundamental Theorem of Calculus.
Next, use the definition of $f_{(X, Y)}$ in (22) to rewrite (28) as

$$
\begin{equation*}
f_{Z}(z)=\int_{0}^{1} f_{(X, Y)}(u, z-u) d u, \quad \text { for } z>0 \tag{29}
\end{equation*}
$$

We consider two cases, (i) $0<z \leqslant 1$, and (ii) $z>1$.
(i) For $0<z \leqslant 1$, use (22) to obtain from (29) that

$$
\begin{aligned}
f_{Z}(z) & =\int_{0}^{z} e^{u-z} d u \\
& =e^{-z} \int_{0}^{z} e^{u} d u \\
& =1-e^{-z}
\end{aligned}
$$

so that

$$
\begin{equation*}
f_{z}(z)=1-e^{-z}, \quad \text { for } 0<z \leqslant 1 \tag{30}
\end{equation*}
$$

(ii) For $z>1$, use (22) to obtain from (29) that

$$
\begin{aligned}
f_{Z}(z) & =\int_{0}^{1} e^{u-z} d u \\
& =e^{-z} \int_{0}^{1} e^{u} d u \\
& =(e-1) e^{-z}
\end{aligned}
$$

so that

$$
\begin{equation*}
f_{z}(z)=(e-1) e^{-z}, \quad \text { for } z>1 \tag{31}
\end{equation*}
$$

Combining (30) and (31) we obtain the cdf

$$
f_{Z}(z)= \begin{cases}0 & \text { for } z \leqslant 0  \tag{32}\\ 1-e^{-z}, & \text { for } 0<z \leqslant 1 \\ (e-1) e^{-z}, & \text { for } z>1\end{cases}
$$

Finally, integrating (32) yields the cdf

$$
F_{Z}(z)= \begin{cases}0 & \text { for } z \leqslant 0 \\ z+e^{-z}-1, & \text { for } 0<z \leqslant 1 \\ e^{-1}+(e-1)\left(e^{-1}-e^{-z}\right), & \text { for } z>1\end{cases}
$$

(b) Let $U=Y / X$. Find $F_{U}$ and $f_{U}$.

Solution: Since $X \sim \operatorname{Uniform}(0,1)$ and $Y \sim \operatorname{Exponential(1),~their~pdfs~}$ are given by

$$
f_{X}(x)= \begin{cases}1 & \text { if } 0<x<1 \\ 0 & \text { elsewhere }\end{cases}
$$

and

$$
f_{Y}(y)= \begin{cases}e^{-y} & \text { if } y>0 \\ 0 & \text { if } y \leqslant 0\end{cases}
$$

respectively. The joint pdf of $(X, Y)$ is then

$$
f_{(X, Y)}(x, y)= \begin{cases}e^{-y} & \text { if } 0<x<1, y>0  \tag{33}\\ 0 & \text { elsewhere }\end{cases}
$$

We compute the cdf of $U$,

$$
F_{U}(u)=\operatorname{Pr}(U \leqslant u)=\operatorname{Pr}\left(\frac{Y}{X} \leqslant u\right)
$$

or

$$
\begin{equation*}
F_{U}(u)=\iint_{\frac{y}{x} \leqslant u} f_{(X, Y)}(x, y) d x d y \tag{34}
\end{equation*}
$$

where the integrand in (34) is given in (33) and the integration is done over the region

$$
R=\left\{(x, y) \in \mathbb{R}^{2} \left\lvert\, \frac{y}{x} \leqslant u\right.\right\} .
$$

Make the change variables

$$
\begin{aligned}
w & =x \\
v & =\frac{y}{x}
\end{aligned}
$$

so that

$$
\begin{align*}
& x=w  \tag{35}\\
& y=w v,
\end{align*}
$$

in the integral in (34) to obtain

$$
\begin{equation*}
F_{U}(u)=\int_{-\infty}^{u} \int_{-\infty}^{\infty} f_{(X, Y)}(w, w v)\left|\frac{\partial(x, y)}{\partial(w, v)}\right| d w d v \tag{36}
\end{equation*}
$$

where

$$
\frac{\partial(x, y)}{\partial(w, v)}=\operatorname{det}\left(\begin{array}{cc}
1 & 0  \tag{37}\\
v & w
\end{array}\right)=w
$$

is the Jacobian determinant of the transformation in (35). It then follows from (36) and (37) that

$$
\begin{equation*}
F_{U}(u)=\int_{-\infty}^{u} \int_{-\infty}^{\infty} f_{(X, Y)}(w, w v)|w| d w d v \tag{38}
\end{equation*}
$$

Differentiating with respect to $u$ and using the definition of the joint pdf of $(X, Y)$ in (33) we obtain from (38) that

$$
\begin{equation*}
f_{U}(u)=\int_{-\infty}^{\infty} f_{(X, Y)}(w, w u)|w| d w \tag{39}
\end{equation*}
$$

where we have also used the Fundamental Theorem of Calculus.
Next, use the definition of $f_{(X, Y)}$ in (33) to rewrite (39) as

$$
\begin{equation*}
f_{U}(u)=\int_{0}^{1} e^{-u w} w d w, \quad \text { for } u>1 \tag{40}
\end{equation*}
$$

We evaluate the integral in (40) by integration by parts to get

$$
\begin{align*}
f_{U}(u) & =\left[-\frac{w}{u} e^{-u w}-\frac{1}{u^{2}} e^{-u w}\right]_{0}^{1}  \tag{41}\\
& =\frac{1}{u^{2}}-\frac{1}{u} e^{-u}-\frac{1}{u^{2}} e^{-u}, \quad \text { for } u>0
\end{align*}
$$

In order to compute the cdf, $F_{U}$, we can integrate (34) in Cartesian coordinates to get

$$
\begin{aligned}
F_{U}(u) & =\int_{0}^{1} \int_{0}^{u x} e^{-y} d y d x \\
& =\int_{0}^{1}\left[1-e^{-u x}\right] d x \\
& =1+\frac{1}{u}\left[e^{-u}-1\right]
\end{aligned}
$$

so that

$$
F_{U}(u)= \begin{cases}1+\frac{1}{u}\left[e^{-u}-1\right], & \text { for } u>0  \tag{42}\\ 0 & \text { for } u \leqslant 0\end{cases}
$$

Note that differentiating $F_{U}(u)$ in (42) with respect to $u$, for $u>0$, leads to (41). We then have that

$$
f_{U}(u)= \begin{cases}\frac{1}{u^{2}}\left(1-e^{-u}\right)-\frac{1}{u} e^{-u}, & \text { for } u>0 \\ 0 & \text { for } u \leqslant 0\end{cases}
$$

13. Let $X \sim$ Exponential(1), and define $Y$ to be the integer part of $X+1$; that is, $Y=i+1$ if and only if $i \leqslant X<i+1$, for $i=0,1,2, \ldots$ Find the pmf of $Y$, and deduce that $Y \sim \operatorname{Geometric}(p)$ for some $0<p<1$. What is the value of $p$ ?
Solution: Compute

$$
\operatorname{Pr}[Y=i+1]=\operatorname{Pr}[i \leqslant X<i+1]=\operatorname{Pr}[i<X \leqslant i+1],
$$

since $X$ is continuous; so that

$$
\begin{equation*}
\operatorname{Pr}[Y=i+1]=\int_{i}^{i+1} f_{X}(x) d x \tag{43}
\end{equation*}
$$

where

$$
f_{X}(x)= \begin{cases}e^{-x} & \text { if } x>0  \tag{44}\\ 0 & \text { if } x \leqslant 0\end{cases}
$$

since $X \sim$ Exponential(1).

Evaluating the integral in (43), for $i \geqslant 0$ and $f_{X}$ as given in (44), yields

$$
\begin{aligned}
\operatorname{Pr}[Y=i+1] & =\int_{i}^{i+1} e^{-x} d x \\
& =\left[-e^{-x}\right]_{i}^{i+1} \\
& =e^{-i}-e^{-i-1}
\end{aligned}
$$

so that

$$
\begin{equation*}
\operatorname{Pr}[Y=i+1]=\left(\frac{1}{e}\right)^{i}\left(1-\frac{1}{e}\right) \tag{45}
\end{equation*}
$$

It follows from (45) that $Y \sim \operatorname{Geometric}(p)$ with $p=1-\frac{1}{e}$.
14. Let $X_{1}, X_{2}, X_{3}, \ldots, X_{n}$ be independent identically distributed Bernoulli random variables with parameter $p$, with $0<p<1$. Define

$$
Y=X_{1}+X_{2}+\cdots+X_{n}
$$

Use moment generating functions to determine the distribution of $Y$.
Solution: Compute the moment generation function of $Y$ to get

$$
\begin{aligned}
\psi_{Y}(t) & =E\left(e^{t Y}\right) \\
& =E\left(e^{t\left(X_{1}+X_{2}+\cdots+X_{n}\right)}\right) \\
& =E\left(e^{t X_{1}+t X_{2}+\cdots+t X_{n}}\right) \\
& =E\left(e^{t X_{1}} e^{t X_{2}} \cdots e^{t X_{n}}\right)
\end{aligned}
$$

so that

$$
\psi_{Y}(t)=E\left(e^{t X_{1}}\right) \cdot E\left(e^{t X_{2}}\right) \cdots E\left(e^{t X_{n}}\right)
$$

since the random variables $X_{1}, X_{2}, \ldots, X_{n}$ are mutually independent.
It then follows that

$$
\begin{aligned}
\psi_{Y}(t) & =\psi_{X_{1}}(t) \cdot \psi_{X_{2}}(t) \cdots \psi_{X_{2}}(t) \\
& =\left(p e^{t}+1-p\right) \cdot\left(p e^{t}+1-p\right) \cdots\left(p e^{t}+1-p\right)
\end{aligned}
$$

since each of the $X_{i}$ has a $\operatorname{Bernoulli}(p)$ distribution.

We then have that

$$
\psi_{Y}(t)=\left(p e^{t}+1-p\right)^{n}, \quad \text { for all } t \in \mathbb{R}
$$

which is the moment generating function for a $\operatorname{Binomial}(n, p)$ distribution. It then follows from the Uniqueness Theorem for moment generating functions that $Y$ has a $\operatorname{Binomial}(n, p)$ distribution. Hence, the pmf for $Y$ is

$$
p_{Y}(k)= \begin{cases}\binom{n}{k} p^{k}(1-p)^{n-k}, & \text { for } k=0,1,2, \ldots, n \\ 0, & \text { elsewhere }\end{cases}
$$

