2.4.2Discrete Random Vector

■ Probability Mass Function or, PMF: The collection of positive numbers $\{p_{ij}\}$ satisfy-

 $P(X = x_i, Y = y_j) = p_{ij}$, for all i, j and $\sum_{i,j=1}^{\infty} p_{ij} = 1$, is called the joint probability mass

- $\underbrace{\text{function}}_{\text{function}} \text{ (pmf) of } (X,Y).$ $\blacktriangleright P(X=x_i,Y=y_j) = p_{ij} \Rightarrow F(x,y) = P(X \leq x,Y \leq y) = \sum_{\substack{x_i \leq x,y_i \leq y \\ \infty}} p_{ij}.$
- ▶ If $\{p_{ij}\}$ be a collection of nonnegative real numbers with $\sum_{i,j=1}^{\infty} p_{ij} = 1 \Rightarrow \{p_{ij}\}$ is the

PMF of some discrete RV (X,Y).

Example 2.13. A fair die is rolled, and a fair coin is tossed independently. Let X be the face value on the die, and let Y be the number of head appears. Find the DF F(x,y) of the bivariate random variable (X, Y).

⇒ Here X can take values 1, 2, · · · , 6 and Y can take values 0, 1. The bivariate random variable (X,Y) can take values: $\{(i,j): i=1,\cdots,6; j=0,1\}.$

Since both are fair: $P(X=x,Y=y)=\frac{1}{12}$ where $x=1,\cdots,6;\ y=0,1.$ Therefore $F(x,y) = P(X \le x, Y \le y)$ is

$$F(x,y) = \begin{cases} 0 & \text{if } x < 1, \ y \in \mathbb{R} \\ 0 & \text{if } x \in \mathbb{R}, \ y < 0 \\ 1/12 & \text{if } 1 \leq x < 2, \ 0 \leq y < 1 \\ 1/6 & \text{if } 1 \leq x < 2, \ 1 \leq y \\ 1/6 & \text{if } 2 \leq x < 3, \ 0 \leq y < 1 \\ 1/3 & \text{if } 2 \leq x < 3, \ 1 \leq y; \\ 1/6 & \text{if } 3 \leq x < 4, \ 0 \leq y < 1 \\ 1/2 & \text{if } 3 \leq x < 4, \ 0 \leq y < 1 \\ 1/3 & \text{if } 4 \leq x < 5, \ 0 \leq y < 1 \\ 2/3 & \text{if } 4 \leq x < 5, \ 1 \leq y \\ 5/12 & \text{if } 5 \leq x < 6, \ 0 \leq y < 1 \\ 1/2 & \text{if } 6 \leq x, \ 0 \leq y < 1 \\ 1/2 & \text{if } 6 \leq x, \ 0 \leq y < 1 \\ 1/2 & \text{if } 6 \leq x, \ 1 \leq y \end{cases}$$

- Marginal PMF: Let (X,Y) be a two-dim RV with PMF: $p_{ij} = P(X = x_i, Y = y_j)$.

 The marginal PMF of X is $p_{i.} = \sum_{j=1}^{\infty} p_{ij} = \sum_{j=1}^{\infty} P(X = x_i, Y = y_j) = P(X = x_i)$.
- ▶ The marginal PMF of Y is $p_{.j} = \sum_{i=1}^{\infty} p_{ij} = \sum_{i=1}^{\infty} P(X = x_i, Y = y_j) = P(Y = y_j)$. ▶ Marginal PMF's are <u>univariate</u>, so we can easily find the DF's of marginal distributions.
- ▶ The RV's X and Y are said to be independent if P(X = x, Y = y) = P(X = x)P(Y = y)i.e. $p_{ij} = p_{i.}p_{.j}$.

Example 2.14. Consider the Example 2.13, write down the joint PMF of (X,Y) in tabular form or, matrix form. Hence find the marginal PMF of X and Y.

 \Rightarrow Here X can take values 1, 2, \cdots , 6 and Y can take values 0, 1. The bivariate random variable (X,Y) can take values: $\{(i,j): i=1,\cdots,6; j=0,1\}$.

Since both are fair: $P(X = x, Y = y) = \frac{1}{12}$ where $x = 1, \dots, 6$; y = 0, 1. So the table is:

Table 2.1: Joint probability distribution

□ The marginal distribution table is as follows:

YX	1	2	3	4	5	6	P(Y=y)
0	12	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
1	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
P(X=x)	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	1

Table 2.2: Marginal probability distribution

The marginal PMF of X, shown in the row representing column totals and the marginal PMF of Y is shown in the column representing row totals.

$$P(X=x) = \begin{cases} 1/6 & \text{if } x=1,2,3,4,5,6 \\ 0 & \text{O.w.} \end{cases}; \ P(Y=y) = \begin{cases} 1/2 & \text{if } y=0,1 \\ 0 & \text{O.w.} \end{cases}$$

[Do It Yourself] 2.69. A fair coin is tossed three times. Let X = number of heads in three tossings, and Y = difference, in absolute value, between number of heads and number of tails. Write down the joint PMF of (X,Y) in tabular form. Hence find the marginal PMF of X and Y. Are X and Y independent?

[Do It Yourself] 2.70. Verify if $f(x,y) = \frac{e^{-2}}{x!(y-x)!}$; $x = 0, 1, \dots, y, \ y = 0, 1, \dots, \infty$ is a joint pmf of (X,Y) or, not. If it is a joint pmf then find the marginal pmf's of X and Y. Are X and Y independent?

$$[\underline{Hint}: \sum_{y=0}^{\infty} \sum_{x=0}^{y} \frac{e^{-2}}{x!(y-x)!} = \sum_{y=0}^{\infty} \sum_{x=0}^{y} \frac{e^{-2}}{y!} \binom{y}{x} = e^{-2} \sum_{y=0}^{\infty} \frac{1}{y!} 2^{y}]$$

[Do It Yourself] 2.73. Let X and Y have the joint probability mass function $P(X=n,Y=k)=\left(\frac{1}{2}\right)^{n+2k+1}$; $n=-k,-k+1,\cdots$; $k=1,2,\cdots$. Then E(Y) equals (A) 1 (B) 2 (C) 3 (D) 4.

$$\left[\underbrace{Hint}: P(Y=y) = \sum_{n=-k}^{\infty} \left(\frac{1}{2} \right)^{n+2k+1} = \sum_{n=0}^{\infty} \left(\frac{1}{2} \right)^{n+k+1} \right]$$

Continuous Random Vector 2.4.3

- Properties of PDF: $i) \ f \ge 0, \ \forall (x,y); \ ii) \ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) \ dx \ dy = 1.$ $\bigstar \ i)$ Follows from the definition. ii) Follows from $F(\infty,\infty) = 1.$ If F is abs. continuous and f is continuous at $(x,y) \Rightarrow f(x,y) = \frac{\partial^2 F(x)}{\partial x \partial y}$
- ▶ Every non-negative function f that is integrable over \mathbb{R}^2 with $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = 1$ is the PDF of some bivariate continuous RV (X, Y).

Example 2.15. Let (X,Y) be a bivariate RV with joint PDF: $f(x,y) = e^{-(x+y)}$, x,y > 0. Then find the DF F(x, y).

 \Rightarrow We know that, $F(x,y) = \int_{-\infty}^{x} \left[\int_{-\infty}^{y} f(u,v)dv \right] du$. Therefore,

$$F(x,y) = \begin{cases} 0 & \text{if } x \leq 0, \ y \in \mathbb{R} \\ 0 & \text{if } x \in \mathbb{R}, \ y \leq 0 \\ \int_0^x e^{-u} du \int_0^y e^{-v} dv & \text{if } 0 < x < \infty, \ 0 < y < \infty \end{cases}$$

$$F(x,y) = \begin{cases} (1 - e^{-x})(1 - e^{-y}) & \text{if } x > 0, \ y > 0 \\ 0 & \text{Otherwise} \end{cases}$$

- Marginal PDF: Let (X,Y) be a two-dim RV with PDF: f(x,y)
- The marginal PDF of X is $f_x(x) = \int_{y=-\infty}^{\infty} f(x,y) dy$. The marginal PDF of Y is $f_y(y) = \int_{x=-\infty}^{\infty} f(x,y) dx$.
- ► Marginal PDF's are <u>univariate</u>, so we can easily find the DF's of marginal distributions.
- ▶ The RV's X and Y are said to be independent if $f(x,y) = f_x(x)f_y(y)$

[Do It Yourself] 2.74. Verify if f(x,y) = 2; 0 < x < y < 1 is a joint pdf of (X,Y)or, not. If it is a joint pdf then find the marginal pdf's of X and Y. Are X and Y independent?

$$[\underline{Hint}: \int_{x=0}^{1} \int_{y=x}^{1} 2 \ dy \ dx = 1; \ f_X(x) = \int_{y=-\infty}^{\infty} f(x,y) dy = \int_{x}^{1} 2 dy = 2(1-x); \ 0 < x < 1. \ f_Y(y) = \int_{x=-\infty}^{\infty} f(x,y) dx = \int_{0}^{y} 2 dx = 2y; \ 0 < y < 1. \ Not \ independent]$$

- [Do It Yourself] 2.75. Verify if $f(x,y) = 2e^{-x-y}$; $0 < x < y < \infty$ is a joint pdf of (X,Y) or, not. If it is a joint pdf then find the marginal pdf's of X and Y. Are X and Y independent?
- [Do It Yourself] 2.78. Let (X,Y) have the joint PDF f defined by $f(x,y) = \frac{1}{2}$ inside the square with corners at the points (1,0),(0,1),(-1,0) and (0,-1) in the (x,y) plane, and = 0 otherwise. Find the marginal PDFs of X and Y.

$$[\underline{Hint}: f(x,y) = \frac{1}{2}, if |x| + |y| \le 1]$$

2.4.4Conditional distribution

- \blacksquare Conditional PMF: Let (X,Y) be a discrete random variable.
- ▶ If P(Y = y) > 0, the function $P(X = x|Y = y) = \frac{P(X = x, Y = y)}{P(Y = y)}$, for fixed y is known as the conditional PMF of X, given Y = y.

 ▶ If P(X = x) > 0, the function $P(Y = y|X = x) = \frac{P(X = x, Y = y)}{P(X = x)}$, for fixed x is known
- as the conditional PMF of Y, given X = x.
- \square If X, Y are independent P(X = x | Y = y) = P(X = x) for P(Y = y) > 0.
- \square If X, Y are independent P(Y = y | X = x) = P(Y = y) for P(X = x) > 0.
- ▶ Combining: If X, Y are independent then P(X = x, Y = y) = P(X = x)P(Y = y).

[Do It Yourself] 2.81. Consider the Example 2.13, write down the conditional PMFs: P(X = x|Y = 0), P(X = x|Y = 1), P(Y = y|X = 1), P(Y = y|X = 5).

[Do It Yourself] 2.82. Consider the Example 2.70, write down the conditional PMFs: P(X = x|Y = y) for fixed y, and P(Y = y|X = x) for fixed x. $[\underline{Hint}: Easy]$

[Do It Yourself] 2.85. For the trinomial RV (X,Y) with PMF as follows: $P(X = x, Y = y) = \frac{n!}{x!y!(n-x-y)!}p_1^xp_2^y(1-p_1-p_2)^{n-x-y}$, where $x, y = 0, 1, \dots, n$ (with $x+y \le n$), $0 < p_1 < 1$, $0 < p_2 < 1$ so that $p_1 + p_2 \le 1$. Show that it is a PMF. Find the marginal PMFs of X and Y and the conditional PMFs. $[\underline{Hint}: Easy]$

- \blacksquare Conditional PDF: Let (X,Y) be a continuous RV [Note that, P(Y=y)=0].
- ▶ The conditional DF of a random variable X, given Y = y, is defined as the limit $F_{X|Y}(x|y) = \lim_{h \to 0+} P(X \le x|Y \in (y-h,y+h])$, provided the limit exists. We define the conditional density function of X, given Y = y by $f_{X|Y}(x|y)$ as a nonnegative function satisfying $F_{X|Y}(x|y) = \int_{-\infty}^{x} f_{X|Y}(t|y) \ dt$ for all $x \in \mathbb{R}$. Note that: $\int_{-\infty}^{\infty} f_{X|Y}(x|y) \ dx = F_{X|Y}(\infty|y) = 1$.
- Taking h → 0+, we can show that: The conditional PDF of X|Y = y is

$$f_{X|Y}(x|y) = \frac{f_{X,Y}(x,y)}{f_{Y}(y)}$$

▶ Similarly, the conditional PDF of Y|X=x is $f_{Y|X}(y|x) = \frac{f_{X,Y}(x,y)}{f_X(x)}$.

Example 2.16. Let (X,Y) have the joint PDF

$$f(x,y) = \begin{cases} c[xy + \frac{x^2}{2}] & \text{if } 0 < x < 1, 0 < y < 2\\ 0 & \text{Otherwise} \end{cases}$$

Find c, P(X < 1/2), P(Y < 1/3), P(0.5 < X < 1, 0 < Y < 1), P(Y < 1|X < 1/2), $\begin{array}{l} P(X=Y),\ P(X<Y),\ P(X+Y<1),\ P(XY<1/2).\\ \Rightarrow Since\ f\ is\ a\ pdf \Rightarrow \int_{y=0}^2 \int_{x=0}^1 f(x,y) dx dy = 1 \Rightarrow c = \frac{3}{4}. \end{array}$ So, $f(x,y) = \frac{3}{4}[xy + \frac{x^2}{2}], 0 < x < 1, 0 < y < 2.$

 \square Marginal distributions are $f_X(x) = \int_{y=0}^2 f(x,y) dy = \frac{3}{4} (2x + x^2), \ 0 < x < 1.$ and $f_Y(y) = \int_{x=0}^1 f(x,y) dy = \frac{3}{4} (\frac{y}{2} + \frac{1}{6}), \ 0 < y < 2.$

Therefore, $P(X < 1/2) = \int_{x=0}^{1/2} f_X(x) dx = \frac{7}{32}$, $P(Y < 1/3) = \int_{y=0}^{1/3} f_Y(y) dy = \frac{1}{16}$.

 $\square P(0.5 < X < 1, 0 < Y < 1) = \int_{y=0}^{1} \int_{x=0.5}^{1} f(x, y) dx dy = 0.25.$

 $\begin{array}{l} P(Y < 1 | X < 1/2) = \frac{P(X < 1/2, Y < 1)}{P(X < 1/2)} = \frac{0.0625}{7/32} = 0.2857. \\ \square \ P(X = Y) = 0, \ since \ continuous \ distribution. \\ Draw \ region, \ P(X < Y) = \int_{y=0}^{1} \int_{x=0}^{y} f(x,y) dx dy + \int_{y=1}^{2} \int_{x=0}^{1} f(x,y) dx dy = \frac{1}{8} + \frac{11}{16} = \frac{13}{16}. \end{array}$ $P(X + Y < 1) = \int_{y=0}^{1} \int_{x=0}^{y} f(x, y) dx dy = \frac{1}{8}.$

 $P(XY < 1/2) = \int_{y=0}^{1/2} \int_{x=0}^{1} f(x,y) dx dy + \int_{y=1/2}^{2} \int_{x=0}^{1/2y} f(x,y) dx dy = 0.11 + 0.16 = 0.27.$

[Do It Yourself] 2.87. Let X and Y be continuous random variables with the joint probability density function

$$f(x,y) = \begin{cases} x+y, & \textit{if } 0 < x < 1, 0 < y < 1 \\ 0, & \textit{Otherwise} \end{cases}$$

Then $P(X + Y > \frac{1}{2})$ equals

(A) 23/24. (B) 1/12. (C) 11/12. (D) 1/24.

[Do It Yourself] 2.95. Let the joint density function of (X,Y) be

 $f(x,y) = \begin{cases} c(x+y), & \text{if } -x < y < x, 0 < x < 1, \\ 0, & \text{otherwise.} \end{cases}.$ Then find the value of c.

 $[\underline{Hint}: Easy]$

[Do It Yourself] 2.96. Let the joint probability mass function of random variable X and Y be given by $P(X=m,Y=n)=\frac{e^{-1}}{(n-m)!m!2^n},\ m=0,1,2,\cdots,n;\ n=0,1,2,\cdots$. Find the marginal probability mass functions of X and Y. Also, find the conditional probability

mass function of X given Y = 5, and that of Y given X = 6. $[\underline{Hint}: \ P(X = m) = \sum_{n=0}^{\infty} \frac{e^{-1}}{(n-m)!m!2^n} = \sum_{n=m}^{\infty} \frac{e^{-1}}{(n-m)!m!2^n} = \frac{e^{-1}}{m!} \sum_{n=0}^{\infty} \frac{1}{n!2^{n+m}}]$

 $\begin{aligned} & \textbf{[Do It Yourself] 2.97.} \ \ \textit{Let X and Y have the joint probability density function} \\ & f(x,y) = \begin{cases} cxye^{-(x^2+2y^2)}, & \textit{if $x>0,y>0$}, \\ 0, & \textit{otherwise}. \end{cases} \end{aligned} . \textit{Evaluate the constant c and $P(X^2>Y^2)$}.$

2.4.5Independent Random Variables

- Two RVs X, Y are independent iff $F(x,y) = F_X(x)F_Y(y), \ \forall (x,y) \in \mathbb{R}^2$.

 ▶ The RV's X and Y are independent iff $P(X \in A_x, Y \in A_y) = P(X \in A_x)P(Y \in A_y)$, for all Borel sets A_x on the x axis and A_y on the y axis.

 ▶ The DRV's X and Y are independent iff P(X = x, Y = y) = P(X = x)P(Y = y).

 ▶ The RV's X and Y are said to be independent iff $f(x,y) = f_X(x)f_Y(y)$.

- \square If X, Y are independent P(X = x|Y = y) = P(X = x) for P(Y = y) >
- \square If X, Y are independent P(Y = y | X = x) = P(Y = y) for P(X = x) > 0.
- A degenerate RV is independent of any RV.
- ▶ If X_1, X_2, \dots, X_n are independent \Rightarrow every sub-collection $X_{i_1}, X_{i_2}, \dots, X_{i_k}$ of X_1, X_2, \dots, X_n is also independent.

[Do It Yourself] 2.99. Let X and Y are jointly distributed with pdf

$$f(x,y) = \begin{cases} \frac{1+xy}{4} & \text{if } |x| < 1, \ |y| < 1 \\ 0 & \text{Otherwise} \end{cases}$$

Then show that X and Y are not independent but X^2 and Y^2 are independent. Explain the reason.

 $[\underline{Hint}: Easy; P(X^2 \le x, Y^2 \le y) = P(-\sqrt{x} \le X \le \sqrt{x}, -\sqrt{y} \le Y \le \sqrt{y}) =$ $\int_{-\sqrt{x}}^{\sqrt{x}} \int_{-\sqrt{y}}^{\sqrt{y}} f(x,y) dy \ dx = \sqrt{x} \sqrt{y}; \ Show \ F_X^2(x) = \sqrt{x}$

- Identically Distributed: Two RVs X, Y are identically distributed if X and Y have the same DF i.e. $F_X(x) = F_Y(y)$.
- ▶ Two RVs X, Y are independent and identically distributed (iid) if they are independent and identically distributed.
- ▶ If P(X = Y) = 1, we say that X and Y are equivalent RVs.
- Note that, equivalent RVs has equal event sets whereas for identical distribution the probability of events are equal.
- ▶ Two RVs X, Y are exchangeable if $(X, Y) \stackrel{d}{=} (Y, X)$ i.e (X, Y) and (Y, X) are identically distributed.
- If X, Y are exchangeable RVs ⇒ X − Y has a symmetric distribution.
- [Do It Yourself] 2.100. Let X_1, X_2 be iid RVs with common PMF: $P(X = \pm 1) = \frac{1}{2}$. Take $X_3 = X_1X_2$, show that X_1, X_2, X_3 are pairwise independent but not independent. $[\underline{Hint}: P(X_3=1) = P(X_1=1, X_2=1) + P(X_1=-1, X_2=-1) = \frac{1}{2}, P(X_3=-1) = \frac{1}{2}$ $\frac{1}{2}$; $P(X_1 = 1, X_3 = 1) = P(X_1 = 1, X_1 X_2 = 1) = P(X_1 = 1, X_2 = 1) = \frac{1}{4} = P(X_1 = 1, X_2 = 1)$ 1) $P(X_3 = 1)$
- [Do It Yourself] 2.103. Let X_1, X_2, \dots, X_n be a set of exchangeable RVs. Then show

that
$$E\left(\frac{X_1+X_2\cdots+X_p}{X_1+X_2\cdots+X_n}\right) = \frac{p}{n}$$
, for $1 \le p \le n$.

$$[\underline{Hint}: E\left(\frac{X_1+X_2\cdots+X_n}{X_1+X_2\cdots+X_n}\right) = 1 \Rightarrow E\left(\frac{X_1}{X_1+X_2\cdots+X_n}\right) + E\left(\frac{X_2}{X_1+X_2\cdots+X_n}\right) + \cdots + E\left(\frac{X_n}{X_1+X_2\cdots+X_n}\right) = 1 \Rightarrow nE\left(\frac{X_1}{X_1+X_2\cdots+X_n}\right) = 1 \Rightarrow E\left(\frac{X_1}{X_1+X_2\cdots+X_n}\right) = \frac{1}{n}]$$

[Do It Yourself] 2.104. Let (X_1, X_2, X_3) be a RV with joint PMF: $f(x_1, x_2, x_3) = \frac{1}{4}$ if $(x_1, x_2, x_3) \in A$, where $A = \{(1, 0, 0), (0, 1, 0), (0, 0, 1), (1, 1, 1)\}$. Are X_1, X_2, X_3 independent? Are X_1, X_2, X_3 pairwise independent? Are $X_1 + X_2$ and X_3 independent?

 $[\underline{Hint}: P(X_1=0)=f(0,1,0)+f(0,0,1)=\frac{1}{2}; P(X_1=1)=f(1,0,0)+f(1,1,1)=\frac{1}{2}]$