

## Review:

① joint cdf.

Def. For two random variables  $X$  and  $Y$  that are defined over the same probability space:

$$F_{X,Y}(u,v) = P(X \leq u, Y \leq v) = P\{\omega \in \Omega: X(\omega) \leq u, Y(\omega) \leq v\}$$

Properties:

$$\bullet P(a < X \leq b, c < Y \leq d) = F_{X,Y}(b,d) - F_{X,Y}(a,d) - F_{X,Y}(b,c) + F_{X,Y}(a,c)$$

$$\bullet \lim_{u \rightarrow +\infty} F_{X,Y}(u,v) = F_Y(v), \quad \lim_{v \rightarrow +\infty} F_{X,Y}(u,v) = F_X(u)$$

Proposition: A function  $F$  is a joint cdf if and only if:

$$JF.1: 0 \leq F(u,v) \leq 1 \text{ for any } (u,v) \in \mathbb{R}^2$$

$$JF.2: F(u,v) \text{ is nondecreasing in } u \text{ and nondecreasing in } v.$$

$$JF.3: F(u,v) \text{ is right-continuous in } u \text{ and right-continuous in } v.$$

$$JF.4: \lim_{u \rightarrow -\infty} F(u,v) = 0 \text{ and } \lim_{v \rightarrow -\infty} F(u,v) = 0$$

$$JF.5: \lim_{u \rightarrow \infty} \lim_{v \rightarrow \infty} F(u,v) = 1$$

$$JF.6: \text{For any } a < b \text{ and } c < d, F(b,d) - F(a,d) - F(b,c) + F(a,c) \geq 0$$

② joint pmf

If  $X$  and  $Y$  are discrete-type,

$$P_{X,Y}(u,v) = P(X=u, Y=v) = P(\{\omega \in \Omega: X(\omega) = u, Y(\omega) = v\})$$

Properties:

$$P_X(u) = \sum_i P_{X,Y}(u, v_i), \text{ where } P(Y \in \{v_1, v_2, \dots\}) = 1.$$

## Today:

① joint pmf

② joint pdf

① joint pmf (continued)

Def. Suppose that  $X$  and  $Y$  are discrete-type random variables defined over a same probability space  $(\Omega, \mathcal{F}, P)$ .

**Def.** Suppose that  $X$  and  $Y$  are discrete-type random variables defined over a same probability space  $(\Omega, \mathcal{F}, P)$ . The conditional pmf of  $Y$  given  $X$  is denoted by  $P_{Y|X}$  and defined as:

$$P_{Y|X}(v|u) = P(Y=v|X=u) = \frac{P(X=u, Y=v)}{P(X=u)} = \frac{P_{X,Y}(u,v)}{P_X(u)}$$

**Proposition:** A function  $p$  is joint pmf if and only if:

JPM1.  $p$  is non-negative

JPM2. There are finite or countably infinite sets  $\{u_1, u_2, \dots\}$  and  $\{v_1, v_2, \dots\}$  such that  $p(u,v) = 0$  if  $u \notin \{u_1, u_2, \dots\}$  or  $v \notin \{v_1, v_2, \dots\}$

$$\text{JPM3. } \sum_i \sum_j p(u_i, v_j) = 1$$

**Proof.** If  $p$  satisfies above then  $F(u,v) = \sum_{i: u_i \leq u} \sum_{j: v_j \leq v} p(u_i, v_j)$  satisfies JF1-JF6. If  $p$  is joint pmf, then above are trivial.

**Important remark:**

• Recall that two random variables  $X$  and  $Y$  are independent if and only if  $P(X \in A, Y \in B) = P(X \in A)P(Y \in B)$  for any  $A, B$ .

Recall that if  $X$  and  $Y$  are discrete-type, then  $X$  and  $Y$  are independent if

$$P_{X,Y}(u,v) = P(X=u, Y=v) = P(X=u)P(Y=v) = P_X(u)P_Y(v)$$

• Recall that given  $E_1, E_2, \dots$  for which  $E_1 \cup E_2 \cup \dots = \Omega$ ,  $E_i \cap E_j = \emptyset$  for all  $i \neq j$ , and  $P(E_i) \neq 0$  for all  $i$ , we have the law of total probability:

$$P(A) = \sum_{i=1}^{\infty} P(A \cap E_i) = \sum_{i=1}^{\infty} P(E_i) P(A|E_i)$$

Similarly, we can write a version of law of total probability for conditional pmf.

$$P_Y(v) = \sum_i P_{X,Y}(u_i, v) = \sum_i P_X(u_i) \cdot P_{Y|X}(v|u_i)$$

#### 4.1. [A joint pmf]

The joint pmf  $p_{X,Y}(u,v)$  of  $X$  and  $Y$  is shown in the table below.

	u=0	u=1	u=2	u=3
v=4	0	0.1	0.1	0.2
v=5	0.2	0	0	0
v=6	0	0.2	0.1	0.1

- Find the marginal pmfs  $p_X(u)$  and  $p_Y(v)$ .
- Let  $Z = X + Y$ . Find  $p_Z$ , the pmf of  $Z$ .
- Are  $X$  and  $Y$  independent random variables? Justify your answer.
- Find  $p_{Y|X}(v|3)$  for all  $v$  and find  $E[Y|X = 3]$ .

**Solution:**

(a)  $P_X$  :

	u=0	u=1	u=2	u=3
$P_X(u)$	0.2	0.3	0.2	0.3

,  $P_Y$  :

	v=4	v=5	v=6
$P_Y(v)$	0.4	0.2	0.4

(b)  $Z = X + Y$ . Notice that  $Z$  can take values between 4 and 9

	w=4	w=5	w=6	w=7	w=8	w=9
$P_Z(w)$	0	0.3	0.1	0.4	0.1	0.1

(c) No, since  $P_{X,Y}(0,4) = 0$ , however,  $P_X(0)P_Y(4) = 0.2 \times 0.4$

(d)

	v=4	v=5	v=6
$P_{Y X}(v 3)$	$\frac{2}{3}$	0	$\frac{1}{3}$

$$E[Y|v=3] = 4 \cdot P_{Y|X}(4|3) + 5 \cdot P_{Y|X}(5|3) + 6 \cdot P_{Y|X}(6|3) = \frac{14}{3}$$

② joint pdf

**Def.** We say random variables  $X$  and  $Y$  (that are defined over the same probability space) are jointly continuous-type if there exists a function  $f_{X,Y}$ , called the joint probability density function (pdf), such that

$$F_{X,Y}(u_0, v_0) = \int_{-\infty}^{u_0} \int_{-\infty}^{v_0} f_{X,Y}(u,v) \, dv \, du.$$

• Recall that for any  $A \subset \mathbb{R}$  ( $A$  measurable), we have  $P(X \in A) = \int_A f_X(u) \, du$

Similarly, for any  $R \subset \mathbb{R}^2$  ( $R$  measurable), we have  $P((X,Y) \in R) = \iint_R f_{X,Y}(u,v) du dv$ . In particular,

$$P(a < X \leq b, c < Y \leq d) = F_{X,Y}(b,d) - F_{X,Y}(a,d) - F_{X,Y}(b,c) + F_{X,Y}(a,c)$$

$$= \int_c^d \int_a^b f_{X,Y}(u,v) du dv$$

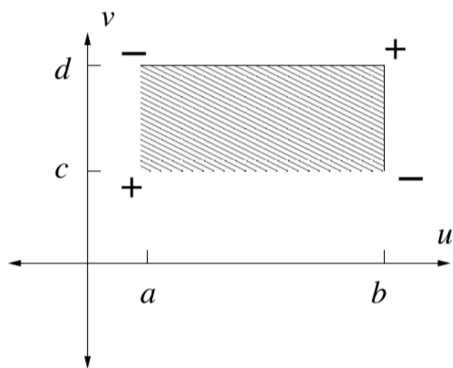


Figure 4.2:  $P\{(X,Y) \in \text{shaded region}\}$  is equal to  $F_{X,Y}$  evaluated at the corners with signs shown.

which also equals  $\int_c^d \int_a^b f_{X,Y}(u,v) du dv$ , given  $X$  and  $Y$  are jointly continuous.

Suppose that  $\epsilon > 0$  is very small and assume that  $f_{X,Y}(u,v)$  is continuous in both  $X$  and  $Y$ .

Using above equality, we have

$$P(X \in (u_0 - \epsilon, u_0 + \epsilon), Y \in (v_0 - \epsilon, v_0 + \epsilon)) \approx f_{X,Y}(u_0, v_0) (2\epsilon)^2$$

In particular,

$$f_{X,Y}(u_0, v_0) = \lim_{\epsilon \rightarrow 0} \frac{P(X \in (u_0 - \epsilon, u_0 + \epsilon), Y \in (v_0 - \epsilon, v_0 + \epsilon))}{(2\epsilon)^2}$$

This gives a probabilistic interpretation of joint pdf  $f_{X,Y}$ .

**Important remark:**

Since  $X$  and  $Y$  are continuous type,  $P(X=u) = P(Y=v) = 0$ , for any  $u, v \in \mathbb{R}$ . In particular,

$$P(a < X \leq b, c < Y \leq d) = P(a < X < b, c < Y < d) \quad \text{for any } a < b \text{ and } c < d$$