

# Stat 260 Lecture Notes

## Set 18 - Joint Probability Distributions

Suppose  $X$  and  $Y$  are two discrete random variables on a sample space  $\mathcal{S}$ . The **joint probability mass function**, (**joint pmf**)  $p(x, y)$  is defined as

$$p(x, y) = P(X = x \cap Y = y) = P(X = x, Y = y)$$

**Rules for joint pmfs:**

- $p(x, y) \geq 0$  for all  $x$  and  $y$
- $\sum_x \sum_y p(x, y) = 1$

The **marginal pmf** of the random variable  $X$ , denoted by  $p_X(x)$  is  $p_X(x) = P(X = x) = \sum_y p(x, y)$ . In other words, it's the probability where we focus on a specific  $x$  value and add up over all cases of  $y$ . The marginal pmf of  $X$  is the same as just the pmf of  $X$ .

**Example 1:** Suppose we have random variables  $X$  and  $Y$  with joint pmf:

		$Y$			
		5	10	15	
$X$	4	0.20	0.16	0.03	0.39
	14	0.15	0.14	0.02	0.31
	26	0.12	0.14	0.04	0.30
		$p(14, y)$			

(a) Find  $P(X = 14 \cap Y = 10)$ .  $= 0.14$

(b) Find  $P(Y = 10)$ .  $= 0.16 + 0.14 + 0.14 = 0.44$

(c) Find  $P(Y \geq 10)$ .  $= P(X=10) + P(X=15) = 0.16 + 0.14 + 0.14 + 0.03 + 0.02 + 0.04 = 0.53$

(d) Find  $E(X)$ .  $= \sum x \cdot f(x) = 13.7$

marginal pmf for  $X$

$x$	4	14	26
$f(x)$	0.39	0.31	0.30

Recall that we calculate conditional probabilities as  $P(A|B) = \frac{P(A \cap B)}{P(B)}$ . We can calculate conditional probabilities with joint probability distributions in the same way.

The **conditional pmf** of  $Y$  given  $X = x$  is

$$P(Y=y | X=x) = \frac{p(x, y)}{p_X(x)} = \frac{P(X=x \cap Y=y)}{P(X=x)}$$

**Example 2:** Use the joint pmf from Example 1 to find  $P(Y = 10 | X = 14)$ .

$p(x, y)$		$Y$			
		5	10	15	
X	4	0.20	0.16	0.03	0.39
	14	0.15	0.14	0.02	0.31
	26	0.12	0.14	0.04	0.30
		0.47	0.30		

$$P(Y=10 | X=14) = \frac{P(Y=10 \cap X=14)}{P(X=14)} = \frac{0.14}{0.31} = 0.4516$$

$$P(Y=10 | X \geq 14) = \frac{P(Y=10 \cap X \geq 14)}{P(X \geq 14)} = \frac{0.14 + 0.02}{0.31 + 0.30} =$$

Two random variables  $X$  and  $Y$  are **independent** if for all pairs of values for  $x$  and  $y$  we have that  $P(X = x \cap Y = y) = P(X = x) \cdot P(Y = y)$ . If there is any pair of values for  $x$  and  $y$  that does not satisfy this, then the random variables  $X$  and  $Y$  are not independent.

**Example 3:** Use the joint pmf from Examples 1 and 2. Look at the values  $x = 4$  and  $y = 5$ .

$$P(X=4 \cap Y=5) = 0.20$$

$$P(X=4) \cdot P(Y=5) = (0.39)(0.47) = 0.1833$$

$\therefore$  not independent

This definition of independence says that if random variables  $X_1, X_2, \dots, X_n$  are all independent, then the way to find the joint pmf is to multiply the marginal probabilities. In other words,

$$p(x_1, x_2, \dots, x_n) = P(X_1 = x_1) \cdot P(X_2 = x_2) \cdot \dots \cdot P(X_n = x_n)$$

$$P(X_1 = x_1 \cap X_2 = x_2 \cap \dots \cap X_n = x_n)$$