Joint Distributions

• Two Discrete Random Variables. Let X and Y be two discrete r.v.'s defined on the sample space S of an experiment. The *joint probability mass function* p(x, y) is defined for each pair of numbers (x, y) by

$$p(x, y) = P(X = x \text{ and } Y = y)$$

Let A be any set consisting of pairs of (x, y) values. Then the probability $P[(X, Y) \in A]$ is obtained by summing the joint pmf over pairs in A:

$$P[(X,Y) \in A] = \sum \sum_{(x,y) \in A} p(x,y)$$

Example 5.1 (on page 194)

• Marginal Distributions. The marginal probability mass functions of X and of Y, denoted by $p_X(x)$ and $p_Y(y)$, are given by

$$p_X(x) = \sum_y p(x,y)$$
 and $p_Y(y) = \sum_x p(x,y)$

Example 5.1 continued

• Two Continuous Random Variables. Let X and Y be two continuous r.v.'s. Then f(x, y) is the *joint probability density function* for X and Y if for any two-dimensional set A

$$P[(X,Y) \in A] = \int_A \int f(x,y) dy dx$$

In particular, if A is a two-dimensional rectangle $\{(x, y) : a \le x \le b, c \le y \le d\}$, then

$$P[(X,Y) \in A] = P(a \le x \le b, c \le y \le d) = \int_a^b \int_c^d f(x,y) dy dx$$

Example 5.3 (on page 196)

• Marginal Density. The marginal probability density functions of X and Y, denoted by $f_X(x)$ and $f_Y(y)$, respectively, are given by

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) \, dy$$
 and $f_Y(y) = \int_{-\infty}^{\infty} f(x, y) \, dx$

Example 5.3 continued

Example 5.5 (on page 197)

• Independence. Two random variables X and Y are said to be *independent* if for every pair of x and y values,

$$p(x,y) = p_X(x)p_Y(y)$$

or

$$f(x,y) = f_X(x)f_Y(y)$$

Otherwise, X and Y are said to be *dependent*. Example 5.8 (on page 200)

• Homework problems: Section 5.1: pp. 203-206; # 1, 3, 7, 9, 13, 19.