

# Two Discrete Random Variables

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## Joint Probability Distributions

The **joint probability mass function** of the discrete random variables  $X$  and  $Y$ , denoted as  $f_{XY}(x, y)$ , satisfies

$$(1) \quad f_{XY}(x, y) \geq 0$$

$$(2) \quad \sum_x \sum_y f_{XY}(x, y) = 1$$

$$(3) \quad f_{XY}(x, y) = P(X = x, Y = y) \quad (5-1)$$

# Two Discrete Random Variables

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## Marginal Probability Distributions

- The individual probability distribution of a random variable is referred to as its **marginal probability distribution**.
- In general, the marginal probability distribution of  $X$  can be determined from the joint probability distribution of  $X$  and other random variables.

# Two Discrete Random Variables

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## Definition: Marginal Probability Mass Functions

If  $X$  and  $Y$  are discrete random variables with joint probability mass function  $f_{XY}(x, y)$ , then the **marginal probability mass functions** of  $X$  and  $Y$  are

$$f_X(x) = P(X = x) = \sum_y f_{XY}(x, y) \quad \text{and} \quad f_Y(y) = P(Y = y) = \sum_x f_{XY}(x, y) \quad (5-2)$$

where the first sum is over all points in the range of  $(X, Y)$  for which  $X = x$  and the second sum is over all points in the range of  $(X, Y)$  for which  $Y = y$

# Two Discrete Random Variables

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## Conditional Probability Distributions

Given discrete random variables  $X$  and  $Y$  with joint probability mass function  $f_{XY}(x, y)$  the **conditional probability mass function** of  $Y$  given  $X = x$  is

$$f_{Y|x}(y) = f_{XY}(x, y)/f_X(x) \quad \text{for } f_X(x) > 0 \quad (5-3)$$

# Two Discrete Random Variables

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## Conditional Probability Distributions

Because a conditional probability mass function  $f_{Y|x}(y)$  is a probability mass function for all  $y$  in  $R_x$ , the following properties are satisfied:

$$(1) \quad f_{Y|x}(y) \geq 0$$

$$(2) \quad \sum_y f_{Y|x}(y) = 1$$

$$(3) \quad P(Y = y | X = x) = f_{Y|x}(y) \tag{5-4}$$

# Two Discrete Random Variables

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## Definition: Conditional Mean and Variance

The **conditional mean** of  $Y$  given  $X = x$ , denoted as  $E(Y|x)$  or  $\mu_{Y|x}$ , is

$$E(Y|x) = \sum_y y f_{Y|x}(y) \quad (5-5)$$

and the **conditional variance** of  $Y$  given  $X = x$ , denoted as  $V(Y|x)$  or  $\sigma_{Y|x}^2$ , is

$$V(Y|x) = \sum_y (y - \mu_{Y|x})^2 f_{Y|x}(y) = \sum_y y^2 f_{Y|x}(y) - \mu_{Y|x}^2$$

### Example 1

In a group of 100 sports car buyers, 40 bought alarm systems, 30 purchased bucket seats, and 20 purchased an alarm system and bucket seats. If a car buyer chosen at random bought an alarm system, what is the probability they also bought bucket seats?

# Two Discrete Random Variables

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## Independence

For discrete random variables  $X$  and  $Y$ , if any one of the following properties is true, the others are also true, and  $X$  and  $Y$  are **independent**.

- (1)  $f_{XY}(x, y) = f_X(x)f_Y(y)$  for all  $x$  and  $y$
- (2)  $f_{Y|x}(y) = f_Y(y)$  for all  $x$  and  $y$  with  $f_X(x) > 0$
- (3)  $f_{X|y}(x) = f_X(x)$  for all  $x$  and  $y$  with  $f_Y(y) > 0$
- (4)  $P(X \in A, Y \in B) = P(X \in A)P(Y \in B)$  for any sets  $A$  and  $B$  in the range of  $X$  and  $Y$ , respectively. (5-6)



# Problem

The joint PMF of two random variables  $X$  and  $Y$  is given by

$$P_{XY}(x,y) = \begin{cases} k(2x + y) & x = 1,2; y = 1,2 \\ 0 & \text{otherwise} \end{cases}$$

where  $k$  is constant

- a) What is the value of  $k$ ?
- b) Find the marginal PMFs of  $X$  and  $Y$
- c) Are  $X$  and  $Y$  independent?

# Problem

- A fair coin is tossed three times. Let  $X$  be a random variable that takes the value 0 if the first toss is a tail and the value 1 if the first toss is a head. Also, let  $Y$  be a random variable that defines the total number of heads in the three tosses.
  - a) Determine the joint PMF of  $X$  and  $Y$
  - b) Are  $X$  and  $Y$  independent?

# Problem

The joint PMF of two random variables  $X$  and  $Y$  is given by

$$P_{XY}(x,y) = \begin{cases} \frac{1}{18} (2x + y) & x = 1,2; y = 1,2 \\ 0 & \text{otherwise} \end{cases}$$

where  $k$  is constant

- a) What is the conditional PMF of  $Y$  given  $X$ ?
- b) What is the conditional PMF of  $X$  given  $Y$ ?

# Two Continuous Random Variables

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## Joint Probability Distribution

### Definition

A **joint probability density function** for the continuous random variables  $X$  and  $Y$ , denoted as  $f_{XY}(x, y)$ , satisfies the following properties:

(1)  $f_{XY}(x, y) \geq 0$  for all  $x, y$

(2) 
$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f_{XY}(x, y) dx dy = 1$$

(3) For any region  $R$  of two-dimensional space

$$P((X, Y) \in R) = \iint_R f_{XY}(x, y) dx dy \quad (5-14)$$

# Two Continuous Random Variables

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## Marginal Probability Distributions

### Definition

If the joint probability density function of continuous random variables  $X$  and  $Y$  is  $f_{XY}(x, y)$ , the **marginal probability density functions** of  $X$  and  $Y$  are

$$f_X(x) = \int_y f_{XY}(x, y) dy \quad \text{and} \quad f_Y(y) = \int_x f_{XY}(x, y) dx \quad (5-15)$$

where the first integral is over all points in the range of  $(X, Y)$  for which  $X = x$  and the second integral is over all points in the range of  $(X, Y)$  for which  $Y = y$

# Two Continuous Random Variables

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## Conditional Probability Distributions

### Definition

Given continuous random variables  $X$  and  $Y$  with joint probability density function  $f_{XY}(x, y)$ , the **conditional probability density function** of  $Y$  given  $X = x$  is

$$f_{Y|x}(y) = \frac{f_{XY}(x, y)}{f_X(x)} \quad \text{for} \quad f_X(x) > 0 \quad (5-16)$$

# Two Continuous Random Variables

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## Independence

### Definition

For continuous random variables  $X$  and  $Y$ , if any one of the following properties is true, the others are also true, and  $X$  and  $Y$  are said to be **independent**.

- (1)  $f_{XY}(x, y) = f_X(x) f_Y(y)$  for all  $x$  and  $y$
- (2)  $f_{Y|x}(y) = f_Y(y)$  for all  $x$  and  $y$  with  $f_X(x) > 0$
- (3)  $f_{X|y}(x) = f_X(x)$  for all  $x$  and  $y$  with  $f_Y(y) > 0$
- (4)  $P(X \in A, Y \in B) = P(X \in A)P(Y \in B)$  for any sets  $A$  and  $B$  in the range of  $X$  and  $Y$ , respectively. (5-19)

# Problem

Let  $X$  and  $Y$  two continuous random variables whose joint PDF is given by

$$f_{XY}(x,y) = \begin{cases} x + Cy^2 & 0 \leq x \leq 1 ; 0 \leq y \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

(a) Find the constant  $C$

(b) Find  $P(0 \leq x \leq \frac{1}{2}, 0 \leq y \leq \frac{1}{2})$



# Problem

X and Y two continuous random variables whose joint PDF is given by

$$f_{XY}(x,y) = \begin{cases} e^{-(x+y)} & 0 \leq x < \infty ; 0 \leq y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Are X and Y independent?

# Problem

The joint PMF of two random variables  $X$  and  $Y$  is given by

$$f_{XY}(x,y) = \begin{cases} xe^{-x(y+1)} & 0 \leq x < \infty ; 0 \leq y < \infty \\ 0 & \text{otherwise} \end{cases}$$

- a) What is the conditional PMF of  $Y$  given  $X$ ?
- b) What is the conditional PMF of  $X$  given  $Y$ ?

# Covariance

The *covariance* between two rv's  $X$  and  $Y$  is

$$\text{Cov}(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$$

$$= \begin{cases} \sum_x \sum_y (x - \mu_X)(y - \mu_Y) p(x, y) & \text{discrete} \\ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_X)(y - \mu_Y) f(x, y) dx dy & \text{continuous} \end{cases}$$

# Short-cut Formula for Covariance

$$\text{Cov}(X, Y) = E(XY) - \mu_X \cdot \mu_Y$$

Theorem:  $X$  and  $Y$  indep implies  $\text{Cov}(X, Y) = 0$ .

Proof:

$$\begin{aligned}\text{Cov}(X, Y) &= E[XY] - E[X]E[Y] \\ &= E[X]E[Y] - E[X]E[Y] \quad (X, Y \text{ indep}) \\ &= 0.\end{aligned}$$

$\text{Cov}(X, Y) = 0$  *does not imply*  $X$  and  $Y$  are independent!!

Definition: The **correlation** between  $X$  and  $Y$  is

$$\rho = \text{Corr}(X, Y) \equiv \frac{\text{Cov}(X, Y)}{\sqrt{\text{Var}(X)\text{Var}(Y)}} = \frac{\sigma_{XY}}{\sigma_X \sigma_Y}.$$

Remark: Cov has “square” units; corr is unitless.

Corollary:  $X, Y$  indep implies  $\rho = 0$ .

# Correlation Proposition

1. If  $a$  and  $c$  are either both positive or both negative,  $\text{Corr}(aX + b, cY + d) = \text{Corr}(X, Y)$
2. For any two rv's  $X$  and  $Y$ ,  
 $-1 \leq \text{Corr}(X, Y) \leq 1.$

Theorem:  $\text{Var}(X+Y) = \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y)$ ,  
*whether or not  $X$  and  $Y$  are indep.*

Remark: If  $X, Y$  are indep, the Cov term goes away.

Proof: By the work we did on a previous proof,

$$\begin{aligned}\text{Var}(X + Y) &= E[X^2] - (E[X])^2 + E[Y^2] - (E[Y])^2 \\ &\quad + 2(E[XY] - E[X]E[Y]) \\ &= \text{Var}(X) + \text{Var}(Y) + 2\text{Cov}(X, Y).\end{aligned}$$



Theorem:  $\text{Cov}(aX, bY) = ab\text{Cov}(X, Y)$ .

Proof:

$$\text{Cov}(aX, bY) = \mathbb{E}[aX \cdot bY] - \mathbb{E}[aX]\mathbb{E}[bY]$$

$$= ab\mathbb{E}[XY] - ab\mathbb{E}[X]\mathbb{E}[Y]$$

$$= ab\text{Cov}(X, Y).$$

Example:  $\text{Var}(X - Y) = \text{Var}(X) + \text{Var}(Y) - 2\text{Cov}(X, Y)$ .

Example:

$$\begin{aligned}\text{Var}(X - 2Y + 3Z) \\ &= \text{Var}(X) + 4\text{Var}(Y) + 9\text{Var}(Z) \\ &\quad - 4\text{Cov}(X, Y) + 6\text{Cov}(X, Z) - 12\text{Cov}(Y, Z).\end{aligned}$$

The proportions  $X$  and  $Y$  of two chemicals found in samples of an insecticide have the joint probability density function

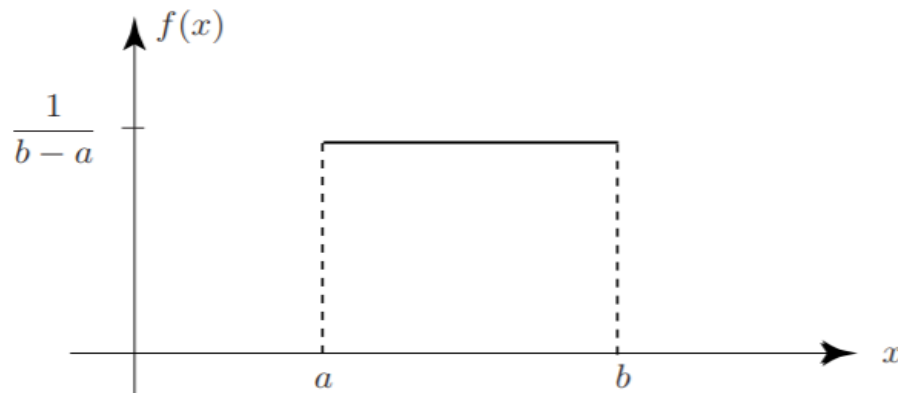
$$f(x, y) = \begin{cases} 2, & 0 \leq x \leq 1, 0 \leq y \leq 1, 0 \leq x + y \leq 1 \\ 0, & \text{elsewhere} \end{cases}$$

The random variable  $Z = X + Y$  denotes the proportion of the insecticide due to both chemicals combined.

1) Find  $E(Z)$  and  $V(Z)$

# Uniform Distribution (continuous)

The Uniform or Rectangular distribution has random variable  $X$  restricted to a finite interval  $[a, b]$  and has  $f(x)$  a constant over the interval.



The function  $f(x)$  is defined by:

$$f(x) = \begin{cases} \frac{1}{b-a}, & a \leq x \leq b \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned}
\mathbb{E}(X) &= \int_{-\infty}^{\infty} x f(x) dx = \int_a^b x \frac{1}{b-a} dx = \frac{1}{2(b-a)} \left[ x^2 \right]_a^b \\
&= \frac{b^2 - a^2}{2(b-a)} \\
&= \frac{b+a}{2}
\end{aligned}$$

Using the formula for the variance, we may write:

$$\begin{aligned}
V(X) &= \mathbb{E}(X^2) - [\mathbb{E}(X)]^2 \\
&= \int_a^b x^2 \cdot \frac{1}{b-a} dx - \left( \frac{b+a}{2} \right)^2 = \frac{1}{3(b-a)} \left[ x^3 \right]_a^b - \left( \frac{b+a}{2} \right)^2 \\
&= \frac{b^3 - a^3}{3(b-a)} - \left( \frac{b+a}{2} \right)^2 \\
&= \frac{b^2 + ab + a^2}{3} - \frac{b^2 + 2ab + a^2}{4} \\
&= \frac{(b-a)^2}{12}
\end{aligned}$$

## Problem

You arrive into a building and are about to take an elevator to your floor. Once you call the elevator it will take between 0 and 40 seconds to arrive to you. We will assume that the elevator arrives uniformly between the 0 and 40 seconds after you press the button. To calculate the probability that takes less than 15 seconds to arrive.

## The Discrete Uniform Distribution

Suppose the possible values of a random variable from an experiment are a set of integer values occurring with the same frequency. That is, the integers 1 through  $k$  occur with equal probability. Then the probability of obtaining any particular integer in that range is  $1/k$  and the probability distribution can be written

$$p(y) = 1 / k, \quad y = 1, 2, \dots, k.$$

$k$  depending on the range of existing values of the variable

The mean and variance

$$\mu = (k + 1) / 2$$

and

$$\sigma^2 = (k^2 - 1) / 12.$$

## Problem

Let  $Y$  be the random variable describing the number of spots on the top face of the die. Then

$$p(y) = 1/6, \quad y = 1, 2, \dots, 6,$$

which is the discrete uniform distribution with  $k = 6$ . The mean of  $Y$  is

$$\mu = (6 + 1) / 2 = 3.5,$$

and the variance is

$$\sigma^2 = (36 - 1) / 12 = 2.917.$$



