



### **Introduction Into Probability Theory**

MTH 231 Lecture 4 Chapter III

Joint Probability Distributions



## Today's lecture

- Joint Density Functions
- Marginal Distributions
- Conditional Probability
- Independence
- ☐ The Covariance

## Joint Probability Distributions

- Definition: The function P(x, y) is a joint probability distribution function of two *discrete r.v's*. X and Y if:
- 1.  $P(x, y) \ge 0$  for all (x, y)
- $\sum_{x}\sum_{y}P(x,y)=1$
- 3. P(X = x, Y = y) = P(x, y)
- 4. For any region A in the XY plane,

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

For the discrete case.

### Joint *PMF*

 $\square$  Example: Let the joint *pmf* of X and Y be give by

$$P(x,y) = \begin{cases} \alpha(x^2 + y^2), & \text{if } (x,y) = (1,1), (1,2), (2,3), (3,3) \\ 0, & \text{otherwise.} \end{cases}$$

- a) Find the value of  $\alpha$ ?
- b) Calculate P(X > Y),  $P(X + Y \le 4)$ , and  $P(Y \ge X)$ .

#### Joint PMF

■ Solution:

a) 
$$\sum_{x} \sum_{y} P(x, y) = \sum_{(x, y)} P(x, y) = 1$$

$$= \alpha [P(1, 1) + P(1, 2) + P(2, 3) + P(3, 3)]$$

$$= 38 \alpha,$$

$$\Rightarrow \alpha = 1/38.$$

b) P(X > Y) = 0,  $P(X + Y \le 4) = 7/38$ , and  $P(Y \ge X) = 1$ . Why?

## Joint Density Functions

- Definition: The function f(x,y) is a joint probability density function of the *continuous random variables* X and Y if:
- 1.  $f(x, y) \ge 0$  for all (x, y)
- $\int_{-\infty-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dxdy = 1$ 
  - 3. P(X < x, Y < y) = f(x, y)
  - 4.  $P[(X,Y) \in A] = \int \int f(x,y) dxdy$

For the continous case.

- Example: A candy company distributed boxes of chocolates with a mixture of creams, toffees, and nuts coated in both light and dark chocolate. Select a box randomly, let X and Y, be the proportions of the light and dark chocolates that are creams respectively. Suppose that the joint density function of X and Y is  $\left(\frac{2}{3}(2x+3y) 0 \le x \le 1.0 \le y \le 1.0 \le 1.0 \le y \le$ 
  - is  $f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere} \end{cases}$
- a) Verify whether  $\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} f(x, y) dx dy = 1$ ?
- b) Find P[(X,Y)  $\in$  A], where A is the region  $\{(x,y) \mid 0 \le x \le \frac{1}{2}, \frac{1}{4} \le y \le \frac{1}{2}\}$ ?

#### ☐ Solution:

a) 
$$\int_{-\infty-\infty}^{\infty} \int_{-\infty}^{\infty} f(x,y) dx dy = \int_{0}^{1} \int_{0}^{1} \frac{2}{5} (2x + 3y) dx dy =$$



$$= \int_{0}^{1} \frac{2x^{2}}{5} + \frac{6xy}{5} \bigg|_{x=0}^{x=1} dy$$

$$= \int_{0}^{1} \left( \frac{2}{5} + \frac{6y}{5} \right) dy = \frac{2y}{5} + \frac{3y^{2}}{5} \Big|_{0}^{1}$$

$$= \frac{2}{5} + \frac{3}{5} = 1$$

#### Continue

b) 
$$P[(X,Y) \in A] = P(0 < X < \frac{1}{2}, \frac{1}{4} < Y < \frac{1}{2}) =$$

$$= \int_{\frac{1}{4}}^{\frac{1}{2}} \int_{0}^{\frac{1}{2}} \frac{2}{5} (2x + 3y) dx dy$$

$$= \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{2x^{2}}{5} + \frac{6xy}{5} \Big|_{x=0}^{x=\frac{1}{2}} dy$$

$$= \int_{\frac{1}{4}}^{\frac{1}{2}} \left( \frac{1}{10} + \frac{3y}{5} \right) dy = \frac{y}{10} + \frac{3y^{2}}{10} \Big|_{\frac{1}{4}}^{\frac{1}{2}}$$

$$= \frac{1}{10} \left[ \left( \frac{1}{2} + \frac{3}{4} \right) - \left( \frac{1}{4} + \frac{3}{16} \right) \right] = \frac{13}{160}$$

## Marginal Distributions

■ Definition: The individual (marginal) probability mass functions of *X* alone and of *Y* alone are:

$$P(x) = \sum_{y} p(x, y)$$
 and

$$P(y) = \sum_{x} p(x, y)$$

For the discrete

## Marginal Distributions

**Definition:** The individual (marginal) probability density functions of *X* alone and of *Y* alone are

For the continuous case

## Statistical Independence

Definition: Let X and Y be two discrete (or continous) random variables, with joint probability mass (or density) function P(x, y) (or f(x,y)). The random variables X and Y are said to be statistically independent if and only if:

$$P(x, y) = P_X(x) P_Y(y)$$
 or  $f_{XY}(x,y) = f_X(x) f_Y(y)$ 

For the discrete case.

For the continuous case

### Conditional Probability Distributions

- Let X and Y be two discrete random variables, with joint PMF P(x,y) and marginal probability mass functions  $P_X(x)$  and  $P_Y(y)$ .
- The conditional probability mass function of the random variable Y, given that X = x, is  $P(y \mid x) = \frac{P(x, y)}{P_{x}(x)}$
- Similarly, the conditional PDF of the random variable X, given that Y = y, is  $P(x \mid y) = \frac{P(x, y)}{P_{v}(y)}$

#### Covariance

#### ☐ Definition:

The covariance between any two jointly distributed random variables X and Y, denoted by Cov(X, Y), is defined by

$$Cov(X, Y) = E[(X - \mu_X)(Y - \mu_Y)]$$
$$= E[XY] - \mu_X \mu_Y$$

where  $\mu_X = E[X]$  and  $\mu_Y = E[Y]$ 

 $\square$  Example: If X and Y have the joint density function

$$f(x,y) = \begin{cases} \frac{x(1+3y^2)}{4}, & 0 < x < 2, 0 < y < 1 \\ 0, & \text{O.W.} \end{cases}$$

- a) Find g(x), h(y), f(y|x), f(x|y),
- b) Calculate  $P(\frac{1}{4} < X < \frac{1}{2} | Y = 1/3)$ ,
- c) Evaluate Var(X),
- d) Evaluate Cov(X, Y).

a) Solution: By definition,

$$g(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_{0}^{1} \frac{x(1+3y^{2})}{4} dy = \frac{xy}{4} + \frac{xy^{3}}{4} \bigg|_{y=0}^{y=1} = \frac{x}{2}, \quad 0 < x < 2$$

and 
$$h(y) = \int_{-\infty}^{\infty} f(x, y) dx = \int_{0}^{2} \frac{x(1+3y^2)}{4} dx = \frac{x^2}{8} + \frac{3x^2y^2}{8} \Big|_{x=0}^{x=2} = \frac{1+3y^2}{2}, \quad 0 < y < 1$$

Then, 
$$f(y \mid x) = \frac{f(x, y)}{g(x)} = \frac{x(1+3y^2)/4}{x/2} = \frac{1}{2}(1+3y^2), \quad 0 < y < 1$$

and 
$$f(x|y) = \frac{f(x,y)}{h(y)} = \frac{x(1+3y^2)/4}{(1+3y^2)/2} = \frac{x}{2}, \quad 0 < x < 2$$

b) Since X and Y are independent, then

$$P\left(\frac{1}{4} < X < \frac{1}{2} \mid Y = \frac{1}{3}\right) = \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{x}{2} dx = \frac{3}{64}.$$

c) Since 
$$E(X) = \int_{0}^{2} x \cdot g(x) dx = \int_{0}^{2} \frac{x^{2}}{2} dx = \frac{4}{3}$$
,

$$E(X^2) = \int_0^2 x \cdot g(x) dx = \int_0^2 \frac{x^3}{2} dx = 2,$$

then 
$$Var(X) = E(X^2) - (E(X))^2 = \frac{2}{3}$$
.

d) 
$$Cov(X,Y) = 0$$
. Why?

 $\square$  Example: If X and Y have the joint density function

$$f(x,y) = \begin{cases} 10 x y^2, & 0 < x < y < 1, \\ 0, & otherwise \end{cases}$$

- a) Find the marginal densities g(x) and h(y).
- b) Find f(y|x).
- c) Find  $P(Y > \frac{1}{2} \mid X=0.25)$ .

#### Solution

a) 
$$g(x)=\int_{x}^{1} f(x,y)dy = \int_{x}^{1} 10 x y dy = \frac{10}{3}x(1-x^{3}).$$

and

$$h(y) = \int_0^y f(x, y) dx = 5 y^4.$$

b) 
$$f(y|x) = \frac{f(y,x)}{g(x)} = \frac{3y^2}{(1-x^3)^2}$$

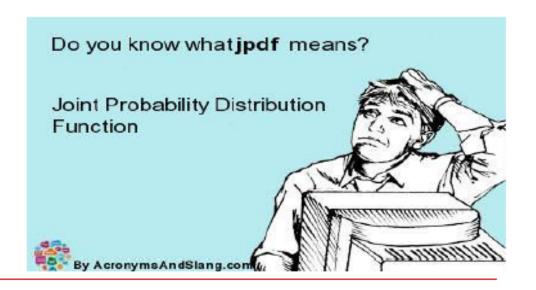
c) 
$$P\left(Y > \frac{1}{2} \middle| X = 0.25\right) = \int_{\frac{1}{2}}^{1} f(Y | X) dy = \int_{\frac{1}{2}}^{1} \frac{3y^{2}}{(1 - 0.25^{3})} dy = \frac{8}{9}.$$

#### Problems:

1) Consider the joint density function of X and Y is

$$f(x,y) = \begin{cases} \frac{2}{5}(2x+3y), & 0 \le x \le 1, 0 \le y \le 1\\ 0, & \text{elsewhere} \end{cases}$$

- a) Find g(x), h(y),
- b) Find f(y|x), f(x|y),
- c) Evaluate Var(X) and Var(Y),
- d) Evaluate Cov(X, Y).

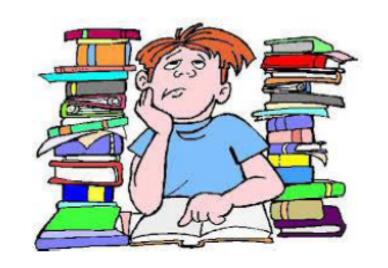


#### Problems:

2) Let f(x,y) be the joint pdf of two random variables X and Y. If f(x,y) is given by:

$$f(x,y) = \begin{cases} \frac{12}{5}x(2-x-y) & \text{if } 0 \le x \le 1, \ 0 \le y \le 1\\ 0 & \text{otherwise.} \end{cases}$$

- a) Find P(X | Y = y), where  $0 \le y \le 1$ ,
- b) Calculate  $P(X > 0.5 \mid Y = 0.5)$ ,
- c) Evaluate  $E(X \mid Y = 0.5)$ ,
- d) Evaluate Var(X | Y = 0.5).



## Questions!

# PRACTICE!



The expert in anything once a