

## 5. Joint probability Distribution:

Defines the simultaneous behavior of more than one random variable:

$$P(x_1 \leq X \leq x_2 \text{ AND } y_1 \leq Y \leq y_2)$$

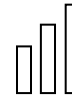
Ex: length and diameter of a piece of plastic material from injection molding machine

Ex: X: no. of bars on cell phone

1, 2, 3

Y: no. of times city name is stated

1,2,3,4



$y \backslash x$	1	2	3	Sum
4	.15	.1	.05	.30
3	.02	.1	.05	.17
2	.02	.03	.2	.25
1	.01	.02	.25	.28
Sum	.2	.25	.55	1.0

- Each entry in the table specifies the joint probability distribution of X and Y
- Range of (X,Y): set of points (x,y) in 2-Dimensional space for which  $P(X=x)$  and  $P(Y=y)$  is positive
- The joint probability of two variables is called bivariate probability distribution OR bivariate distribution.

### Discrete Variates X and Y :

Joint probability distribution of X and Y is a distribution of the points (x,y) in the range of (X,Y) along with the probability of each point.

Joint probability Mass function of (X,Y) is  $f_{XY}(x,y)$  and must satisfy

- 1)  $f_{XY}(x,y) \geq 0$
- 2)  $\sum_x \sum_y f_{XY}(x,y) = 1.0$
- 3)  $f_{XY}(x,y) = P(X = x, Y = y)$

$f_{XY}(x,y) = 0.0$  for values of (x,y) outside the range of X and Y

## Continuous Variates X and Y

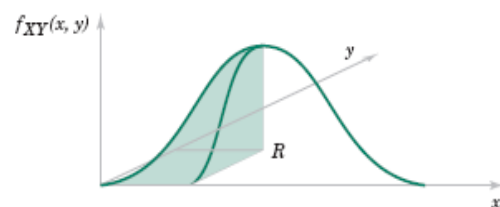
Probability distribution of two continuous random variables X,Y can be specified by providing a method for calculating the probability that X and Y assume a value in the region R of two dimensional space.

Joint Probability density function  $f_{XY}(x,y)$  must satisfy

- 1)  $f_{XY}(x, y) \geq 0$  for all x and y
- 2)  $\int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f_{XY}(x, y) dx dy = 1.0$
- 3) For any region R of two dimensional space

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

= Probability X,Y assume values in R  
= Vol under surface  $f_{XY}(x, y)$  over R



Probability that  $(X, Y)$  is in the region R is determined by the volume of  $f_{XY}(x, y)$  over the region R.

## Marginal Probability Distributions:

It is the individual Probability distribution of a random variable in cases in which more than one variable is defined in a random experiment.

## Discrete Random Variable

Mass function for X and Y is  $f_{XY}(x,y)$

$$f_{XY}(x_i, y_j) \begin{cases} \longrightarrow f_X(x_i) = \sum_j f_{xy}(x_i, y_j) \\ \longrightarrow f_Y(y_j) = \sum_i f_{xy}(x_i, y_j) \end{cases}$$

$f_X(x_i) = P(X=x_i) = \text{Sum of } P(X=x_i; Y=y_j) \text{ for all points in the range of X,Y for which } X=x_i$

$y \backslash x$	1	2	3	$f_Y(y)$
4	.15	.1	.05	0.3
3	.02	.1	.05	0.17
2	.02	.03	.2	0.25
1	.01	.02	.25	0.28
$f_X(x)$	0.2	0.25	0.55	1.0
				1.0

## Continuous Random Variable

Marginal Probability density function

$$f_X(x) = \int_y f_{xy}(x,y) dy \quad ; \quad f_Y(y) = \int_x f_{xy}(x,y) dx$$

$\uparrow$   $\uparrow$   
x is constant      y is constant

$$P(a < X < b) = \int_a^b f_X(x) dx = \int_a^b \left[ \int_{all\ y} f_{XY}(x,y) dy \right] dx$$

### Example:

Let X denote the temperature ( $^{\circ}\text{C}$ ) and let Y denote the time in minutes that it takes for a diesel engine on an automobile to get ready to start. Assume that the joint density for (X,Y) is given by:

$$f_{xy}(x,y) = c(4x + 2y + 1) \quad \begin{matrix} 0 \leq x \leq 40 \\ 0 \leq y \leq 2 \end{matrix}$$

- Find the value of c that makes this a density.
- Find the probability that on a randomly selected day, the air temperature will exceed  $20^{\circ}\text{C}$  and it will take at least one minute for the car to be ready to start.
- Find the marginal densities for X and Y.
- Find the probability that on a randomly selected day it will take at least one minute for the car to be ready to start.
- Are X and Y independent? explain on a mathematical basis.

### Solution:

- a) For  $f_{XY}(x,y)$  to be a density, the condition  $\int_0^2 \int_0^{40} c(4x + 2y + 1) dx dy = 1$  must be satisfied. Therefore, c can be found upon integration and solving for it.  
Upon integration we obtain:

$$\begin{aligned} (2x^2y + xy^2 + xy) \Big|_{x=0}^{x=40} \Big|_{y=0}^{y=2} &= \frac{1}{c} \\ (2 \times 40^2 \times 2 + 40 \times 2^2 + 40 \times 2) &= \frac{1}{c} \\ (6400 + 160 + 80) &= \frac{1}{c} \quad ; \quad c = \frac{1}{6640} \quad ; \quad = 1.506 \times 10^{-4} \end{aligned}$$

b)  $P(X > 20 \text{ and } Y \geq 1) = ?$

$$P(X > 20 \text{ and } Y \geq 1) = \int_1^2 \int_{20}^{40} \frac{1}{6640} (4x + 2y + 1) dx dy$$

$$= \frac{2480}{6640} = 0.3735$$

c)  $f_X(x) = \int_0^2 f_{XY}(x, y) dy$

$$f_X(x) = \int_0^2 c(4x + 2y + 1) dy$$

$$= c [4xy + y^2 + 2]_0^2$$

$$f_X(x) = c(8x + 6)$$

$$f_Y(y) = \int_0^{40} f_{XY}(x, y) dx$$

$$f_Y(y) = \int_0^{40} c(4x + 2y + 1) dx$$

$$= c [2x^2 + 2xy + x]_0^{40}$$

$$f_Y(y) = c(3240 + 80y)$$

d)  $P(Y \geq 1) = \int_1^2 f_Y(y) dy$

$$= \int_1^2 c(3240 + 80y) dy$$

$$= c[3240y + 40y^2]_1^2$$

$$= \frac{3360}{6640}$$

e) *check if*  $f_{XY}(x, y) = f_X(x)f_Y(y)$

$$f_X(x)f_Y(y) = c(8x + 6) c(3240 + 80y)$$

$$= c^2 (25920x + 640xy + 480y + 19440) \neq f_{XY}(x, y)$$

Therefore, X and Y are NOT independent

## Conditional Probability Distribution:

For two events A and B

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

This is the conditional probability of B given A provided A and B are dependent

## Discrete RV:

A: event defined to be  $X=x$


B: event defined to be  $Y=y$

$$P(Y=y | X=x) = \frac{P(X=x, Y=y)}{P(x)} = \frac{\text{Joint}}{\text{Marginal}}$$

This is a new Probability Distribution of Y defined for each X.

$\begin{matrix} x \\ y \end{matrix}$	1	2	3	$f_Y(y)$
4	.15	.1	.05	0.3
3	.02	.1	.05	0.17
2	.02	.03	.2	0.25
1	.01	.02	.25	0.28
$f_X(x)$	0.2	0.25	0.55	1.0

Conditional



$$f_{Y|X}(y) = \frac{\text{Joint}}{\text{Marginal}}$$

$\begin{matrix} x \\ y \end{matrix}$	1	2	3
4	.75	.4	.091
3	.1	.4	.091
2	.1	.12	.364
1	.05	.08	.454
$f_{Y X}(y)$	1.0	1.0	1.0

Each column represents a conditional Probability Mass Function

**Ex:** Signal bars

$$P(Y=1 | X=3) = \frac{P(X=3, Y=1)}{P(X=3)} = \frac{f_{XY}(3,1)}{f_X(3)} = \frac{0.25}{0.55} = 0.454$$

The conditional PMF must satisfy the condition:

$$\sum_j P(Y = y_j | X = x_i) = 1.0$$

## Continuous RV

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

It is a probability density function for all Y in Rx

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

$$2) \int f_{Y|x}(y) dy = 1$$

$$3) P(Y \in B | X=x) = \int_B f_{Y|x}(y) dy \quad \text{For any set B in the range of Y.}$$

Similar results are for X can be written

Using the results in the example of the diesel engine:

$$\begin{aligned} f_{Y|x}(y) &= \frac{f_{xy}(x,y)}{f_X(x)} \\ &= \frac{c(4x + 2y + 1)}{c(8x + 6)} \\ &= \frac{(4x + 2y + 1)}{(8x + 6)} \end{aligned}$$

Find probability  $Y > 1$  given  $X=20$

$$\begin{aligned} P(Y > 1, X = 20) &= \int_1^2 \frac{(4x + 2y + 1)}{(8x + 6)} dy \quad ; \quad \text{for } x = 20 \\ &= \frac{1}{166} \int_1^2 (4x + 2y + 1) dy \\ &= \frac{1}{166} (84) = 0.51 \end{aligned}$$

## Conditional Mean and Variance:

### Mean:

$$\text{Discrete: } E(Y|x) = \mu_{Y|x} = \sum_y f_{y|x}(y) \cdot y$$

$$\text{Continuous: } E(Y|x) = \mu_{Y|x} = \int_y f_{Y|x}(y) \cdot y dy$$

### Variance:

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

$$\text{Continuous: } V(Y|x) = \sigma_{Y|x}^2 = \int_y f_{Y|x}(y) \cdot (y - \mu_{Y|x})^2 dy$$

### Ex:

$$E(Y|_1) = \mu_{Y|1} = 1 \times 0.05 + 2 \times 0.1 + 3 \times 0.1 + 4 \times 0.75 = 3.55$$

$$V(Y|_1) = \sigma_{Y|1}^2 = (1 - 3.55)^2 \times 0.05 + (2 - 3.55)^2 \times 0.1 + (3 - 3.55)^2 \times 0.1 + (4 - 3.55)^2 \times 0.75$$