





AGENDA

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





RANDOM VARIABLES

3 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

DEFINITION: A discrete random variable is a function X(s) from a finite or countably infinite sample space S to the real numbers:

$$X(\cdot)$$
 : $\mathcal{S} \rightarrow \mathbb{R}$.

EXAMPLE: Toss a coin 3 times in sequence. The sample space is

$$S = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\},$$

and examples of random variables are

- X(s) = the number of Heads in the sequence; e.g., X(HTH) = 2,
- $Y(s) = \text{The index of the first } H \; ; \quad e.g., \quad Y(TTH) = 3 \; ,$ $0 \quad \text{if the sequence has no } H \; , \; i.e., \; Y(TTT) \; = \; 0 \; .$

NOTE: In this example X(s) and Y(s) are actually *integers*.



Value-ranges of a random variable correspond to events in S.

4 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES CONDITIONAL

EXPECTATION

DISTRIBUTIONS

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: For the sample space

$$\mathcal{S} = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\},$$

with

$$X(s)$$
 = the number of Heads,

the value

$$X(s) = 2$$
, corresponds to the event $\{HHT, HTH, THH\}$,

and the values

$$1 < X(s) \le 3$$
, correspond to $\{HHH, HHT, HTH, THH\}$.

NOTATION: If it is clear what S is then we often just write X instead of X(s).

RANDOM VARIABLES

(5) JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

NOTATION: We will also write $p_X(x)$ to denote P(X=x).

EXAMPLE: For the sample space

$$\mathcal{S} = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\},$$
 with

X(s) = the number of Heads,

we have

$$p_X(0) \equiv P(\{TTT\}) = \frac{1}{8}$$

$$p_X(1) \equiv P(\{HTT, THT, TTH\}) = \frac{3}{8}$$

$$p_X(2) \equiv P(\{HHT, HTH, THH\}) = \frac{3}{8}$$

$$p_X(3) \equiv P(\{HHH\}) = \frac{1}{8}$$

where

$$p_X(0) + p_X(1) + p_X(2) + p_X(3) = 1$$
. (Why?)

RANDOM VARIABLES

6 JOINT DISTRIBUTIONS

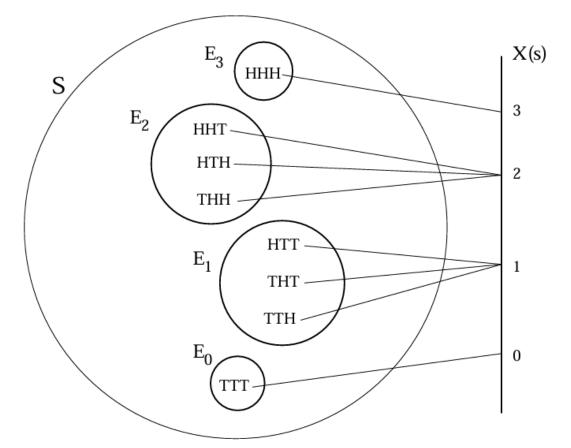
INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE



Graphical representation of X.

The events E_0, E_1, E_2, E_3 are disjoint since X(s) is a function! $(X : S \to \mathbb{R} \text{ must be defined for all } s \in S \text{ and must be single-valued.})$



RANDOM VARIABLES

7 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

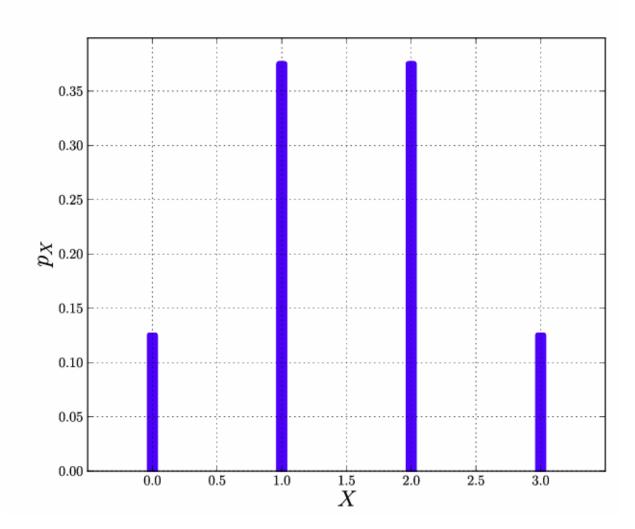
CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





The graph of p_X .



RANDOM VARIABLES

DEFINITION:

 $p_X(x) \equiv P(X=x)$,

is called the *probability mass function*.

8 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

DEFINITION:

 $F_X(x) \equiv P(X \le x)$,

is called the (cumulative) probability distribution function.

PROPERTIES:

- $F_X(x)$ is a non-decreasing function of x. (Why?)
- $F_X(-\infty) = 0$ and $F_X(\infty) = 1$. (Why?)
- $P(a < X \le b) = F_X(b) F_X(a)$. (Why?)

NOTATION: When it is clear what X is then we also write

$$p(x)$$
 for $p_X(x)$ and $F(x)$ for $F_X(x)$.

RANDOM VARIABLES

9 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: With X(s) = the number of Heads, and

$$S = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\}$$

$$p(0) = \frac{1}{8}$$
 , $p(1) = \frac{3}{8}$, $p(2) = \frac{3}{8}$, $p(3) = \frac{1}{8}$,

we have the probability distribution function

$$F(-1) \equiv P(X \le -1) = 0$$

$$F(0) \equiv P(X \le 0) = \frac{1}{8}$$

$$F(1) \equiv P(X \le 1) = \frac{4}{8}$$

$$F(2) \equiv P(X \le 2) = \frac{7}{8}$$

$$F(3) \equiv P(X \le 3) = 1$$

$$F(4) \equiv P(X \le 4) = 1$$

We see, for example, that

$$P(0 < X \le 2) = P(X = 1) + P(X = 2)$$
$$= F(2) - F(0) = \frac{7}{8} - \frac{1}{8} = \frac{6}{8}.$$





RANDOM VARIABLES

IO JOINT DISTRIBUTIONS

INDEPENDENT
RANDOM VARIABLES

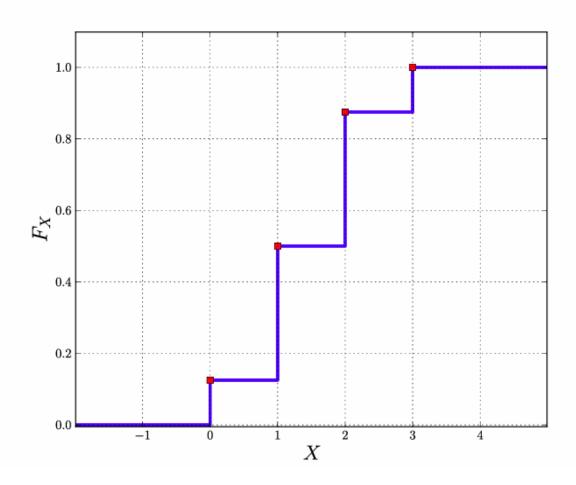
CONDITIONAL
DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





The graph of the probability distribution function F_X .



11 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL **DISTRIBUTIONS**

EXPECTATION

VARIANCE AND **STANDARD DEVIATION**

COVARIANCE

EXAMPLE: Toss a coin until "Heads" occurs.

Then the sample space is *countably infinite*, namely,

$$\mathcal{S} = \{H, TH, TTH, TTTH, \dots \}.$$

The random variable X is the number of rolls until "Heads" occurs:

$$X(H) = 1$$
 , $X(TH) = 2$, $X(TTH) = 3$, \cdots

Then

$$p(1) = \frac{1}{2}$$
 , $p(2) = \frac{1}{4}$, $p(3) = \frac{1}{8}$, \cdots (Why?)

and
$$p(1) = \frac{1}{2}$$
, $p(2) = \frac{1}{4}$, $p(3) = \frac{1}{8}$, \cdots (Why?)
 $F(n) = P(X \le n) = \sum_{k=1}^{n} p(k) = \sum_{k=1}^{n} \frac{1}{2^k} = 1 - \frac{1}{2^n}$,

and, as should be the case,

$$\sum_{k=1}^{\infty} p(k) = \lim_{n \to \infty} \sum_{k=1}^{n} p(k) = \lim_{n \to \infty} \left(1 - \frac{1}{2^n}\right) = 1.$$

NOTE: The outcomes in \mathcal{S} do not have equal probability!

EXERCISE: Draw the *probability mass* and *distribution functions*.



RANDOM VARIABLES

12 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

X(s) is the *number of tosses* until "Heads" occurs \cdots

REMARK: We can also take $S \equiv S_n$ as all ordered outcomes of length n. For example, for n = 4,

$$S_4 = \{ \tilde{H}HHHH, \tilde{H}HHT, \tilde{H}HTH, \tilde{H}HTT,$$

$$\tilde{\boldsymbol{H}}THH$$
 , $\tilde{\boldsymbol{H}}THT$, $\tilde{\boldsymbol{H}}TTT$, $\tilde{\boldsymbol{H}}TTT$,

$$T\tilde{H}HH$$
, $T\tilde{H}HT$, $T\tilde{H}TH$, $T\tilde{H}TT$,

$$TT\tilde{H}H$$
, $TT\tilde{H}T$, $TTT\tilde{H}$, $TTTT$ }.

where for each outcome the first "Heads" is marked as $ilde{H}$.

Each outcome in S_4 has equal probability 2^{-n} (here $2^{-4} = \frac{1}{16}$), and $p_X(1) = \frac{1}{2}$, $p_X(2) = \frac{1}{4}$, $p_X(3) = \frac{1}{8}$, $p_X(4) = \frac{1}{16}$...,

$$p_X(1) = \frac{1}{2}$$
, $p_X(2) = \frac{1}{4}$, $p_X(3) = \frac{1}{8}$, $p_X(4) = \frac{1}{16}$

independent of n.



Joint distributions

The probability mass function and the probability distribution function can also be functions of more than one variable.

13 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: Toss a coin 3 times in sequence. For the sample space

 $\mathcal{S} = \{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT\},$ we let

X(s) = # Heads, Y(s) = index of the first H (0 for TTT).

Then we have the joint probability mass function

$$p_{X,Y}(x,y) = P(X = x, Y = y).$$

For example,

$$p_{X,Y}(2,1) = P(X=2, Y=1)$$

= $P(2 \text{ Heads }, 1^{\text{st}} \text{ toss is Heads})$
= $\frac{2}{8} = \frac{1}{4}$.

48



RANDOM VARIABLES

14 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: (continued \cdots) For

 $\mathcal{S} \; = \; \left\{ HHH \; , \; HHT \; , \; HTH \; , \; HTT \; , \; THH \; , \; THT \; , \; TTH \; , \; TTT \right\} \; ,$

X(s) = number of Heads, and Y(s) = index of the first H,

we can list the values of $p_{X,Y}(x,y)$:

Joint probability mass function $p_{X,Y}(x,y)$

	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
x = 2	0	$\frac{2}{8}$	$\frac{1}{8}$	0	3 8
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	1

NOTE:

- The marginal probability p_X is the probability mass function of X.
- The marginal probability p_Y is the probability mass function of Y.



EXAMPLE: (continued \cdots)

X(s) = number of Heads, and Y(s) = index of the first H.

	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
$\mathbf{x} = 2$	0	$\frac{2}{8}$	$\frac{1}{8}$	0	$\frac{3}{8}$
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	1

INDEPENDENT
RANDOM VARIABLES
CONDITIONAL
DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

For example,

- X = 2 corresponds to the *event* $\{HHT, HTH, THH\}$.
- Y = 1 corresponds to the event $\{HHH, HHT, HTH, HTT\}$
- (X = 2 and Y = 1) corresponds to the event $\{HHT, HTH\}$.

QUESTION: Are the events X = 2 and Y = 1 independent?





RANDOM VARIABLES

16 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

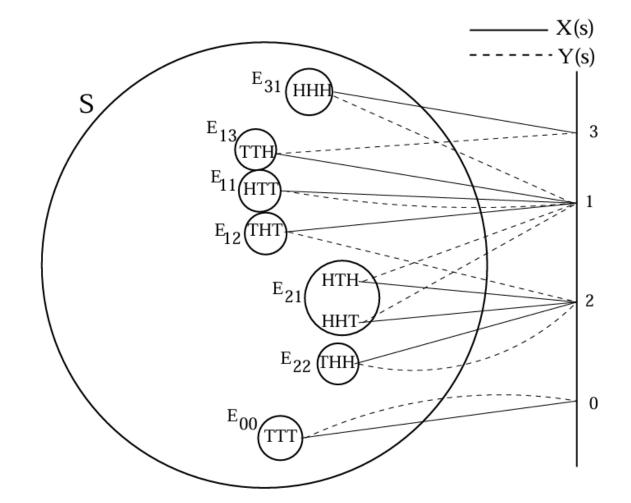
CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





The events $E_{i,j} \equiv \{ s \in S : X(s) = i, Y(s) = j \}$ are disjoint.

QUESTION: Are the events X = 2 and Y = 1 independent?



RANDOM VARIABLES

17 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

DEFINITION:

$$p_{X,Y}(x,y) \equiv P(X=x, Y=y),$$

is called the *joint probability mass function*.

DEFINITION:

and

$$F_{X,Y}(x,y) \equiv P(X \le x, Y \le y),$$

is called the joint (cumulative) probability distribution function.

NOTATION: When it is clear what X and Y are then we also write

$$p(x,y)$$
 for $p_{X,Y}(x,y)$,

F(x,y) for $F_{X,Y}(x,y)$.

48



RANDOM VARIABLES

18 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: Three tosses: X(s) = # Heads, $Y(s) = \text{index } 1^{\text{st}}$ H.

Joint probability mass function $p_{X,Y}(x,y)$

	_	-		1 , -	(, 0 ,
	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
x=2	0	$\frac{2}{8}$	$\frac{1}{8}$	0	$\frac{3}{8}$
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	<u>2</u> 8	$\frac{1}{8}$	1

Joint distribution function $F_{X,Y}(x,y) \equiv P(X \leq x, Y \leq y)$

	y = 0	y = 1	y = 2	y = 3	$F_X(\cdot)$
x = 0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$
x = 1	$\frac{1}{8}$	$\frac{2}{8}$	$\frac{3}{8}$	$\frac{4}{8}$	$\frac{4}{8}$
x=2	$\frac{1}{8}$	$\frac{4}{8}$	<u>6</u> 8	$\frac{7}{8}$	$\frac{7}{8}$
x = 3	$\frac{1}{8}$	<u>5</u> 8	$\frac{7}{8}$	1	1
$F_Y(\cdot)$	$\frac{1}{8}$	<u>5</u> 8	$\frac{7}{8}$	1	1

Note that the distribution function F_X is a copy of the 4th column, and the distribution function F_Y is a copy of the 4th row. (Why?)



RANDOM VARIABLES

19 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

In the preceding example:

Joint probability mass function $p_{X,Y}(x,y)$

	y = 0	y = 1	y = 2	y=3	$p_X(x)$
x = 0	1/8	0	0	0	1 8
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
x=2	0	<u>2</u> 8	$\frac{1}{8}$	0	<u>3</u> 8
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	1

Joint distribution function $F_{X,Y}(x,y) \equiv P(X \leq x, Y \leq y)$

	y = 0	y = 1	y=2	y = 3	$F_X(\cdot)$
x = 0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$
x = 1	$\frac{1}{8}$	$\frac{2}{8}$	<u>3</u> 8	$\frac{4}{8}$	$\frac{4}{8}$
x=2	$\frac{1}{8}$	$\frac{4}{8}$	<u>6</u> 8	$\frac{7}{8}$	$\frac{7}{8}$
x = 3	$\frac{1}{8}$	<u>5</u> 8	$\frac{7}{8}$	1	1
$F_Y(\cdot)$	$\frac{1}{8}$	<u>5</u> 8	$\frac{7}{8}$	1	1

QUESTION: Why is

$$P(1 < X \le 3, 1 < Y \le 3) = F(3,3) - F(1,3) - F(3,1) + F(1,1)$$
?



EXERCISE:

Roll a four-sided die (tetrahedron) two times.

(The sides are marked 1, 2, 3, 4.)

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

(20) IOINT DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

Suppose each of the four sides is equally likely to end facing down.

Suppose the outcome of a $single \ roll$ is the side that faces down (!).

Define the random variables X and Y as

 $X = \text{result of the } first \; roll$, $Y = sum \; \text{of the two rolls.}$

- What is a good choice of the sample space S?
- How many outcomes are there in S?
- List the values of the joint probability mass function $p_{X,Y}(x,y)$.
- List the values of the joint cumulative distribution function $F_{X,Y}(x,y)$.

RANDOM VARIABLES

21 JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

EXERCISE:

Three balls are selected at random from a bag containing

2 red, 3 green, 4 blue balls.

Define the random variables

$$R(s)$$
 = the number of **red** balls drawn,

and

$$G(s)$$
 = the number of *green* balls drawn.

List the values of

- the joint probability mass function $p_{R,G}(r,g)$.
- the marginal probability mass functions $p_R(r)$ and $p_G(g)$.
- the joint distribution function $F_{R,G}(r,g)$.
- the marginal distribution functions $F_R(r)$ and $F_G(g)$.



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT
RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

Independent random variables

Two discrete random variables X(s) and Y(s) are independent if $P(X = x, Y = y) = P(X = x) \cdot P(Y = y)$, for all x and y,

or, equivalently, if their probability mass functions satisfy

$$p_{X,Y}(x,y) = p_X(x) \cdot p_Y(y)$$
, for all x and y,

or, equivalently, if the *events*

$$E_x \equiv X^{-1}(\{x\}) \text{ and } E_y \equiv Y^{-1}(\{y\}),$$

are independent in the sample space S, i.e.,

$$P(E_x E_y) = P(E_x) \cdot P(E_y)$$
, for all x and y.

NOTE:

- In the current discrete case, x and y are typically integers.
- $X^{-1}(\{x\}) \equiv \{ s \in \mathcal{S} : X(s) = x \}$.



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT
RANDOM VARIABLES

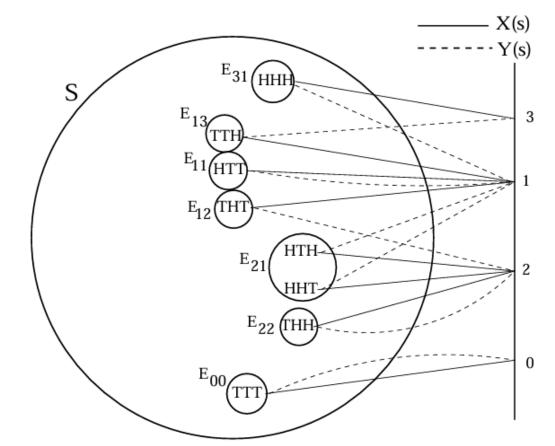
CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





Three tosses: $X(s) = \# \text{ Heads}, Y(s) = \text{ index } 1^{\text{st}} H$.

- What are the values of $p_X(2)$, $p_Y(1)$, $p_{X,Y}(2,1)$?
- Are X and Y independent?



RECALL:

X(s) and Y(s) are *independent* if for all x and y:

$$p_{X,Y}(x,y) = p_X(x) \cdot p_Y(y) .$$

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXERCISE:

Roll a die two times in a row.

Let

X be the result of the 1st roll,

and

Y the result of the 2^{nd} roll.

Are X and Y independent, i.e., is

$$p_{X,Y}(k,\ell) = p_X(k) \cdot p_Y(\ell), \quad \text{for all } 1 \leq k,\ell \leq 6$$
?



EXERCISE:

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

Are these random variables X and Y independent?

Joint probability mass function $p_{X,Y}(x,y)$

	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
x=2	0	$\frac{2}{8}$	$\frac{1}{8}$	0	3 8
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	1



JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXERCISE: Are these random variables X and Y independent?

Joint probability mass function $p_{X,Y}(x,y)$

	·			. 21,1
	y=1	y = 2	y = 3	$p_X(x)$
x = 1	$\frac{1}{3}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
x=2	$\frac{2}{9}$	$\frac{1}{18}$	$\frac{1}{18}$	$\frac{1}{3}$
x = 3	$\frac{1}{9}$	$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{6}$
$p_Y(y)$	2 3	$\frac{1}{6}$	$\frac{1}{6}$	1

Joint distribution function $F_{X,Y}(x,y) \equiv P(X \leq x, Y \leq y)$

	y=1	y = 2	y = 3	$F_X(x)$
x = 1	$\frac{1}{3}$	$\frac{5}{12}$	$\frac{1}{2}$	$\frac{1}{2}$
x=2	<u>5</u>	$\frac{25}{36}$	$\frac{5}{6}$	<u>5</u>
x = 3	$\frac{2}{3}$	<u>5</u> 6	1	1
$F_Y(y)$	$\frac{2}{3}$	<u>5</u> 6	1	1

QUESTION: Is $F_{X,Y}(x,y) = F_X(x) \cdot F_Y(y)$?



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

PROPERTY:

The joint distribution function of independent random variables X and Y satisfies

$$F_{X,Y}(x,y) = F_X(x) \cdot F_Y(y)$$
, for all x,y .

PROOF:

$$F_{X,Y}(x_k, y_\ell) = P(X \le x_k , Y \le y_\ell)$$

$$= \sum_{i \le k} \sum_{j \le \ell} p_{X,Y}(x_i, y_j)$$

$$= \sum_{i \le k} \sum_{j \le \ell} p_X(x_i) \cdot p_Y(y_j) \quad \text{(by independence)}$$

$$= \sum_{i \le k} \left\{ p_X(x_i) \cdot \sum_{j \le \ell} p_Y(y_j) \right\}$$

$$= \left\{ \sum_{i \le k} p_X(x_i) \right\} \cdot \left\{ \sum_{j \le \ell} p_Y(y_j) \right\}$$

$$= F_X(x_k) \cdot F_Y(y_\ell) .$$



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

Conditional distributions

Let X and Y be discrete random variables with joint probability $mass\ function$

$$p_{X,Y}(x,y)$$
.

For given x and y, let

$$E_x = X^{-1}(\{x\})$$
 and $E_y = Y^{-1}(\{y\})$,

be their corresponding *events* in the sample space S.

Then

$$P(E_x|E_y) \equiv \frac{P(E_xE_y)}{P(E_y)} = \frac{p_{X,Y}(x,y)}{p_Y(y)}.$$

Thus it is natural to define the conditional probability mass function

$$p_{X|Y}(x|y) \equiv P(X = x \mid Y = y) = \frac{p_{X,Y}(x,y)}{p_Y(y)}.$$



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

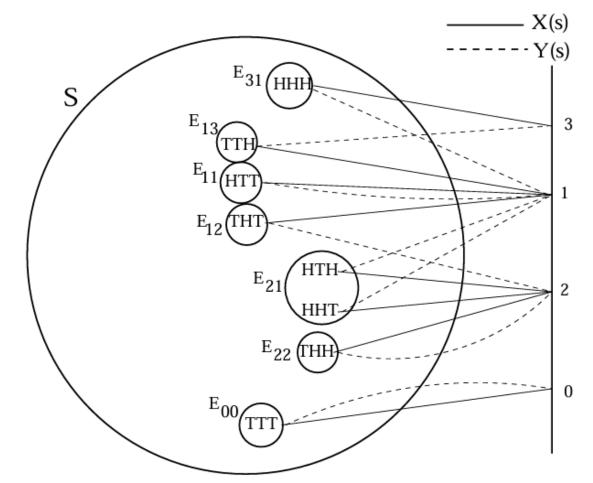
CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE





Three tosses: $X(s) = \# \text{ Heads}, Y(s) = \text{index } 1^{\text{st}} H$.

What are the values of $P(X = 2 \mid Y = 1)$ and $P(Y = 1 \mid X = 2)$?





JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: (3 tosses: X(s) = # Heads, $Y(s) = \text{index } 1^{\text{st}}$ H.)

Joint probability mass function $p_{X,Y}(x,y)$

	_	-		- /	(, 0)
	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{3}{8}$
x = 2	0	$\frac{2}{8}$	$\frac{1}{8}$	0	3 8
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	<u>1</u> 8	$\frac{4}{8}$	<u>2</u> 8	$\frac{1}{8}$	1

Conditional probability mass function $p_{X|Y}(x|y) = \frac{p_{X,Y}(x,y)}{p_Y(y)}$.

	y = 0	y = 1	y = 2	y=3
x = 0	1	0	0	0
x = 1	0	$\frac{2}{8}$	$\frac{4}{8}$	1
x = 2	0	$\frac{4}{8}$	$\frac{4}{8}$	0
x = 3	0	$\frac{2}{8}$	0	0
	1	1	1	1

EXERCISE: Also construct the Table for $p_{Y|X}(y|x) = \frac{p_{X,Y}(x,y)}{p_X(x)}$.



RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE:

Joint probability mass function $p_{X,Y}(x,y)$

	y = 1	y = 2	y = 3	$p_X(x)$
x = 1	1 3	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
x=2	$\frac{2}{9}$	$\frac{1}{18}$	$\frac{1}{18}$	$\frac{1}{3}$
x = 3	$\frac{1}{9}$	$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{6}$
$p_Y(y)$	2 3	$\frac{1}{6}$	$\frac{1}{6}$	1

Conditional probability mass function $p_{X|Y}(x|y) = \frac{p_{X,Y}(x,y)}{p_Y(y)}$

	y = 1	y=2	y = 3
x = 1	$\frac{1}{2}$	$\frac{1}{2}$	$\frac{1}{2}$
x=2	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
x = 3	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$
	1	1	1

QUESTION: What does the last Table tell us?

EXERCISE: Also construct the Table for P(Y = y | X = x).



Expectation

The expected value of a discrete random variable X is

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

$$E[X] \equiv \sum_{k} x_k \cdot P(X = x_k) = \sum_{k} x_k \cdot p_X(x_k) .$$

Thus E[X] represents the weighted average value of X.

(E[X] is also called the *mean* of X.)

EXAMPLE: The expected value of rolling a die is

$$E[X] = 1 \cdot \frac{1}{6} + 2 \cdot \frac{1}{6} + \cdots + 6 \cdot \frac{1}{6} = \frac{1}{6} \cdot \sum_{k=1}^{6} k = \frac{7}{2}$$

EXERCISE: Prove the following:

- $\bullet \quad E[aX] = a E[X] ,$
- $\bullet \quad E[aX+b] = a E[X] + b.$



JOINT DISTRIBUTIONS

INDEPENDENT
RANDOM VARIABLES

CONDITIONAL
DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

EXAMPLE: Toss a coin until "Heads" occurs. Then

$$\mathcal{S} = \{H, TH, TTH, TTTH, \dots \}.$$

The $random\ variable\ X$ is the $number\ of\ tosses$ until "Heads" occurs:

$$X(H) = 1$$
 , $X(TH) = 2$, $X(TTH) = 3$.

Then

$$E[X] = 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{8} + \cdots = \lim_{n \to \infty} \sum_{k=1}^{n} \frac{k}{2^k} = 2.$$

n	$\sum_{k=1}^{n} k/2^k$
1	0.50000000
2	1.00000000
3	1.37500000
10	1.98828125
40	1.99999999

REMARK:

Perhaps using $S_n = \{\text{all sequences of } n \text{ tosses}\}$ is better \cdots



JOINT DISTRIBUTIONS

INDEPENDENT
RANDOM VARIABLES
CONDITIONAL

EXPECTATION

DISTRIBUTIONS

VARIANCE AND STANDARD DEVIATION

COVARIANCE

The expected value of a function of a random variable is

$$E[g(X)] \equiv \sum_{k} g(x_k) p(x_k)$$
.

EXAMPLE:

The pay-off of rolling a die is k^2 , where k is the side facing up.

What should the *entry fee* be for the betting to break even?

SOLUTION: Here $g(X) = X^2$, and

$$E[g(X)] = \sum_{k=1}^{6} k^2 \frac{1}{6} = \frac{1}{6} \frac{6(6+1)(2\cdot 6+1)}{6} = \frac{91}{6} \cong \$15.17.$$

RANDOM VARIABLES

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

The expected value of a function of two random variables is

$$E[g(X,Y)] \equiv \sum_{k} \sum_{\ell} g(x_k, y_\ell) p(x_k, y_\ell) .$$

EXAMPLE:

:		y = 1	y = 2	y = 3	$p_X(x)$
	x = 1	$\frac{1}{3}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
	x = 2	$\frac{2}{9}$	$\frac{1}{18}$	$\frac{1}{18}$	$\frac{1}{3}$
	x = 3	$\frac{1}{9}$	$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{6}$
	$p_Y(y)$	$\frac{2}{3}$	$\frac{1}{6}$	$\frac{1}{6}$	1

$$E[X] = 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{3} + 3 \cdot \frac{1}{6} = \frac{5}{3},$$

$$E[Y] = 1 \cdot \frac{2}{3} + 2 \cdot \frac{1}{6} + 3 \cdot \frac{1}{6} = \frac{3}{2},$$

$$E[XY] = 1 \cdot \frac{1}{3} + 2 \cdot \frac{1}{12} + 3 \cdot \frac{1}{12}$$

$$+ 2 \cdot \frac{2}{9} + 4 \cdot \frac{1}{18} + 6 \cdot \frac{1}{18}$$

$$+ 3 \cdot \frac{1}{9} + 6 \cdot \frac{1}{36} + 9 \cdot \frac{1}{36} = \frac{5}{2}.$$



PROPERTY:

If X and Y are independent then E[XY] = E[X] E[Y].

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL **DISTRIBUTIONS**

EXPECTATION

VARIANCE AND **STANDARD DEVIATION**

COVARIANCE

PROOF:

$$E[XY] = \sum_{k} \sum_{\ell} x_k y_{\ell} p_{X,Y}(x_k, y_{\ell})$$

$$=\sum_{k}\sum_{\ell} x_{k} y_{\ell} p_{X}(x_{k}) p_{Y}(y_{\ell})$$
 (by independence)

$$= \sum_{k} \{ x_k p_X(x_k) \sum_{\ell} y_{\ell} p_Y(y_{\ell}) \}$$

$$= \{\sum_k x_k p_X(x_k)\} \cdot \{\sum_\ell y_\ell p_Y(y_\ell)\}$$

$$= E[X] \cdot E[Y]$$
.

EXAMPLE: See the preceding example!



PROPERTY: E[X+Y] = E[X] + E[Y]. (Always!)

PROOF:

$$E[X+Y] = \sum_{k} \sum_{\ell} (x_k + y_\ell) p_{X,Y}(x_k, y_\ell)$$

$$= \sum_{k} \sum_{\ell} x_{k} p_{X,Y}(x_{k}, y_{\ell}) + \sum_{k} \sum_{\ell} y_{\ell} p_{X,Y}(x_{k}, y_{\ell})$$

$$= \sum_{k} \sum_{\ell} x_{k} p_{X,Y}(x_{k}, y_{\ell}) + \sum_{\ell} \sum_{k} y_{\ell} p_{X,Y}(x_{k}, y_{\ell})$$

$$= \sum_{k} \{x_k \sum_{\ell} p_{X,Y}(x_k, y_{\ell})\} + \sum_{\ell} \{ y_{\ell} \sum_{k} p_{X,Y}(x_k, y_{\ell})\}$$

$$= \sum_{k} \{x_k \ p_X(x_k)\} + \sum_{\ell} \{y_{\ell} \ p_Y(y_{\ell})\}$$

$$= E[X] + E[Y].$$

EXPECTATION

DISTRIBUTIONS

VARIANCE AND STANDARD DEVIATION

COVARIANCE

NOTE: X and Y need not be independent!



INDEPENDENT
RANDOM VARIABLES
CONDITIONAL
DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

EXERCISE:

Probability mass function $p_{X,Y}(x,y)$

			_ , , , , , , , , , , , , , , , , , , ,		
	y = 6	y = 8	y = 10	$p_X(x)$	
x = 1	$\frac{1}{5}$	0	$\frac{1}{5}$	$\frac{2}{5}$	
x = 2	0	$\frac{1}{5}$	0	$\frac{1}{5}$	
x = 3	$\frac{1}{5}$	0	$\frac{1}{5}$	<u>2</u> 5	
$p_Y(y)$	$\frac{2}{5}$	$\frac{1}{5}$	$\frac{2}{5}$	1	

Show that

•
$$E[X] = 2$$
 , $E[Y] = 8$, $E[XY] = 16$

• X and Y are not independent

Thus if

$$E[XY] = E[X] E[Y] ,$$

then it does not necessarily follow that X and Y are independent!





Variance and Standard Deviation

Let X have mean

$$\mu = E[X] .$$

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL **DISTRIBUTIONS**

EXPECTATION

VARIANCE AND **STANDARD DEVIATION**

COVARIANCE

Then the *variance* of X is

$$Var(X) \equiv E[(X - \mu)^2] \equiv \sum_k (x_k - \mu)^2 p(x_k)$$
,

which is the average weighted square distance from the mean.

We have

$$Var(X) = E[X^{2} - 2\mu X + \mu^{2}]$$

$$= E[X^{2}] - 2\mu E[X] + \mu^{2}$$

$$= E[X^{2}] - 2\mu^{2} + \mu^{2}$$

$$= E[X^{2}] - \mu^{2}.$$



$$\sigma(X) \ \equiv \ \sqrt{Var(X)} \ = \ \sqrt{E[\ (X-\mu)^2]} \ = \ \sqrt{E[X^2]\ - \ \mu^2} \ .$$

which is the average weighted *distance* from the mean.

EXAMPLE: The variance of rolling a die is

$$Var(X) = \sum_{k=1}^{6} [k^2 \cdot \frac{1}{6}] - \mu^2$$

$$= \frac{1}{6} \frac{6(6+1)(2\cdot 6+1)}{6} - (\frac{7}{2})^2 = \frac{35}{12}.$$

The standard deviation is

$$\sigma = \sqrt{\frac{35}{12}} \cong 1.70 .$$

INDEPENDENT RANDOM VARIABLES

CONDITIONAL **DISTRIBUTIONS**

EXPECTATION

VARIANCE AND **STANDARD DEVIATION**

COVARIANCE



JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

41) COVARIANCE

•

Covariance

Let X and Y be random variables with mean

$$E[X] = \mu_X , \qquad E[Y] = \mu_Y .$$

Then the *covariance* of X and Y is defined as

$$Cov(X,Y) \equiv E[(X-\mu_X)(Y-\mu_Y)] = \sum_{k,\ell} (x_k-\mu_X)(y_\ell-\mu_Y) p(x_k,y_\ell).$$

We have

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

$$= E[XY - \mu_X Y - \mu_Y X + \mu_X \mu_Y]$$

$$= E[XY] - \mu_X \mu_Y - \mu_Y \mu_X + \mu_X \mu_Y$$

$$= E[XY] - E[X] E[Y].$$



JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

42) COVARIANCE

48

We defined

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

= $\sum_{k,\ell} (x_k - \mu_X)(y_\ell - \mu_Y) p(x_k, y_\ell)$
= $E[XY] - E[X] E[Y]$.

NOTE:

Cov(X,Y) measures "concordance" or "coherence" of X and Y:

- If $X > \mu_X$ when $Y > \mu_Y$ and $X < \mu_X$ when $Y < \mu_Y$ then Cov(X,Y) > 0.
- If $X > \mu_X$ when $Y < \mu_Y$ and $X < \mu_X$ when $Y > \mu_Y$ then Cov(X,Y) < 0.



JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

43) COVARIANCE

4 Q

EXERCISE: Prove the following:

$$\bullet \quad Var(aX+b) = a^2 \ Var(X) \ ,$$

$$\bullet \quad Cov(X,Y) = Cov(Y,X) ,$$

•
$$Cov(cX,Y) = c Cov(X,Y)$$
,

•
$$Cov(X, cY) = c Cov(X, Y)$$
,

$$\bullet \quad Cov(X+Y,Z) = Cov(X,Z) + Cov(Y,Z) ,$$

$$\bullet \quad Var(X+Y) = Var(X) + Var(Y) + 2 Cov(X,Y) .$$



JOINT DISTRIBUTIONS

RANDOM VARIABLES
CONDITIONAL

DISTRIBUTIONS

INDEPENDENT

EXPECTATION

VARIANCE AND STANDARD DEVIATION

COVARIANCE

48

PROPERTY:

If X and Y are independent then Cov(X,Y) = 0.

PROOF:

We have already shown (with $\mu_X \equiv E[X]$ and $\mu_Y \equiv E[Y]$) that

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

and that if X and Y are independent then

$$E[XY] = E[X] E[Y] .$$

from which the result follows.



JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

45) COVARIANCE

48

EXERCISE: (already used earlier \cdots)

Probability mass function $p_{X,Y}(x,y)$

			- / / /		
	y = 6	y = 8	y = 10	$p_X(x)$	
x = 1	1 5	0	$\frac{1}{5}$	<u>2</u> 5	
x = 2	0	$\frac{1}{5}$	0	$\frac{1}{5}$	
x = 3	$\frac{1}{5}$	0	$\frac{1}{5}$	$\frac{2}{5}$	
$p_Y(y)$	<u>2</u> 5	$\frac{1}{5}$	<u>2</u> 5	1	

Show that

•
$$E[X] = 2$$
 , $E[Y] = 8$, $E[XY] = 16$

$$\bullet \quad Cov(X,Y) = E[XY] - E[X] E[Y] = 0$$

• X and Y are not independent

Thus if

$$Cov(X,Y) = 0$$
,

then it does not necessarily follow that X and Y are independent!



PROPERTY:

PROOF:

If X and Y are independent then

$$Var(X+Y) = Var(X) + Var(Y)$$
.

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL **DISTRIBUTIONS**

EXPECTATION

STANDARD DEVIATION

VARIANCE AND

COVARIANCE

We have already shown (in an exercise!) that

$$Var(X+Y) = Var(X) + Var(Y) + 2 Cov(X,Y),$$

and that if X and Y are independent then

$$Cov(X,Y) = 0$$
,

from which the result follows.





JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND STANDARD DEVIATION

47) COVARIANCE

48

EXERCISE:

Compute

$$E[X]$$
 , $E[Y]$, $E[X^2]$, $E[Y^2]$

$$E[XY]$$
 , $Var(X)$, $Var(Y)$

for

Joint probability mass function $p_{X,Y}(x,y)$

	1	0		$I^{-}I^{1}$, I	(, 0)
	y = 0	y = 1	y = 2	y = 3	$p_X(x)$
x = 0	$\frac{1}{8}$	0	0	0	$\frac{1}{8}$
x = 1	0	$\frac{1}{8}$	$\frac{1}{8}$	$\frac{1}{8}$	3 8
x = 2	0	$\frac{2}{8}$	$\frac{1}{8}$	0	3 8
x = 3	0	$\frac{1}{8}$	0	0	$\frac{1}{8}$
$p_Y(y)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{2}{8}$	$\frac{1}{8}$	1





EXERCISE:

Compute

E[X] , E[Y] , $E[X^2]$, $E[Y^2]$

E[XY] , Var(X) , Var(Y)

Cov(X,Y)

JOINT DISTRIBUTIONS

INDEPENDENT RANDOM VARIABLES

CONDITIONAL DISTRIBUTIONS

EXPECTATION

VARIANCE AND **STANDARD DEVIATION**

COVARIANCE

for

Joint probability mass function $p_{X,Y}(x,y)$

	y=1	y = 2	y = 3	$p_X(x)$
x = 1	$\frac{1}{3}$	$\frac{1}{12}$	$\frac{1}{12}$	$\frac{1}{2}$
x = 2	$\frac{2}{9}$	$\frac{1}{18}$	$\frac{1}{18}$	$\frac{1}{3}$
x = 3	$\frac{1}{9}$	$\frac{1}{36}$	$\frac{1}{36}$	$\frac{1}{6}$
$p_Y(y)$	$\frac{2}{3}$	$\frac{1}{6}$	$\frac{1}{6}$	1



THANK YOU

SUBRATA SAHA SUBRATAISTATAMIKARANA.CO.IN