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①

## Discrete random variables (r.v's)

$(\Omega, \mathcal{F}, P)$  — probability space

A r.v.  $X$  is a function on  $\Omega$

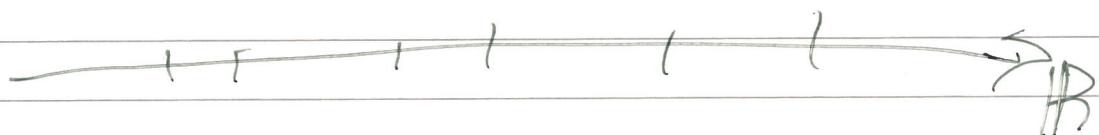
$$X: \Omega \rightarrow \mathbb{R}$$

Def-n. Discr. r.v's are r.v's with the following properties.

(1) The values of  $X$  belong to a discrete set of numbers.

$P(X = x) > 0$  for each  $x$  from this set.

This set can be finite or infinite countable.



$$\sum_{x \in S} P(X = x) = 1$$

(2)

## Expectation and Variance

Def.:  $E(X) \stackrel{\text{def}}{=} \sum_x x P(X=x)$ , if  $\sum_x |x| P(X=x) < \infty$

Consider a r.v. which is a function of  $X$ :

$$Y = g(X), \quad g: \mathbb{R} \rightarrow \mathbb{R}$$

Then, by def-n

$$E(Y) = \sum_y y P(Y=y), \quad \text{if } \sum_y |y| P(Y=y) < \infty$$

Theorem, If  $X$  is discrete, r.v.,

$g: \mathbb{R} \rightarrow \mathbb{R}$  is a function, and

$$\sum_x |g(x)| P(X=x) < \infty, \quad Y = g(X), \text{ then}$$

$$E(Y) = E(g(X)) = \sum_x g(x) P(X=x)$$

Exercise. Prove this theorem for the case

when  $X$  take values  $x_1, x_2, \dots, x_k$ .

Remark. The  $E(X) = x_1 P(X=x_1) + \dots + x_k P(X=x_k)$

$$\left(= \sum_{x_j} x_j P(X=x_j)\right)$$

(3)

Remark. This theorem is important because, in particular, it allows one to compute  $E(Y)$ ,  $Y = g(x)$ , without computing  $P(Y = y)$ .

### Continuous R.V.s

For cont. r.v.,  $P(X = x) = 0$  for all  $x$ .

We consider the probabilities

$$P(X \in [a, b])$$

Def-n. A r.v.  $X$  is continuous if

there is a function  $f_X(x)$  which has two properties:

$$(i) f_X(x) \geq 0$$

$$(ii) \int_{-\infty}^{\infty} f_X(x) dx = 1.$$

and

$$P(X \in [a, b]) = \int_a^b f_X(x) dx.$$

for any interval  $[a, b]$ .

$f_X(x)$  is the probability density function (pdf)

(4)

## Expectation.

Def-n. If  $X$  is a cont. r.v. with pdf  
 (probability density function)  $f_X(x)$  then

$$E(X) = \int_{-\infty}^{\infty} xf_X(x)dx \text{ if } \int_{-\infty}^{\infty} |x|f_X(x)dx < \infty$$

If  $g: \mathbb{R} \rightarrow \mathbb{R}$ , we can define

$$Y = g(X) - \text{a new r.v.}$$

Theorem. In this setting

$$E(Y) = \int_{-\infty}^{\infty} g(x)f_X(x)dx,$$

$$\text{if } \int_{-\infty}^{\infty} |g(x)|f_X(x)dx < \infty.$$

$$\text{Remark, } E(Y) = \int_{-\infty}^{\infty} y f_Y(y) dy$$

This theorem allows one to compute  $E(Y)$  without computing  $f_Y(y)$  first.

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## Conditional Probability (CP)

### 1. Conditional probability of events

Def-n. Suppose that A and B are events. Given that  $P(B) > 0$

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

Def-n. Events  $B_1, B_2, \dots, B_n$  form a partition if

$$(i) \quad B_i \cap B_j = \emptyset \quad \text{for all } i \neq j$$

$$(ii) \quad \bigcup_{j=1}^n B_j = \Omega$$

(Recall: events are subsets of  $\Omega$ ).

If  $\omega$  is an elementary event,  $\omega \in \Omega$ , then it belongs to exactly one of the  $B_j$ 's.

Theorem. (The Total Probability Th-m)

Suppose that A is an event and

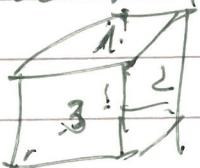
$B_1, B_2, \dots, B_n$  form a partition. Then

$$P(A) = \sum_{j=1}^n P(B_j) P(A|B_j)$$

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Example. You roll two dice.

What is the pr-tiy that the product is even?



1, 2, ..., 6

$B_1, B_2, \dots, B_6$  - form a partition.

$B_j$  = the event of obtaining  $j$  at 1<sup>st</sup> roll.

$$P(B_j) = \frac{1}{6}.$$

A - the event that the product is even.

$$P(A|B_1) = \frac{3}{6} = \frac{1}{2}, = P(A|B_3) = P(A|B_5)$$

$$P(A|B_2) = 1. = P(A|B_4) = P(A|B_6)$$

Hence

$$P(A) = \frac{1}{6} \sum_{j=1}^6 P(A|B_j) = \frac{1}{6} \left( 3 \cdot \frac{1}{2} + 3 \cdot \frac{1}{2} \right)$$

$$= \frac{1}{6} (4.5) = \frac{4.5}{6} = \frac{9}{12} = \underline{\underline{\frac{3}{4}}}.$$

another approach:

$B_1$  - the first result is odd.

$B_2$  ————— is even,  
(see notes).

Example. Alice and Bob play a game. They roll a fair die. If it comes up 1 or 2, Alice wins. If it comes up 3, Bob wins. If it comes up 4, 5, or 6, they continue rolling until one of them wins.

What is the probability that Alice wins?

Solution. Consider events

$B_1 = \{1, 2\}$ ,  $B_2 = \{3\}$ ,  $B_3 = \{4, 5, 6\}$ ,  
(at first roll).

$B_1, B_2, B_3$  form a partition.

$A$  = event that Alice wins.

$$P(B_1) = \frac{2}{6} = \frac{1}{3}, \quad P(B_2) = \frac{1}{6}, \quad P(B_3) = \frac{1}{2}.$$

$$P(A|B_1) = 1, \quad P(A|B_2) = 0,$$

$$P(A|B_3) = P(A) \text{ (1). Hence}$$

$$P(A) = \frac{1}{3} \times 1 + \frac{1}{6} \times 0 + \frac{1}{2} P(A) \Rightarrow \\ \frac{1}{2} P(A) = \frac{1}{3} \Rightarrow \boxed{P(A) = \frac{2}{3}}.$$

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# Discrete

## Conditional Distribution of R.V.'s.

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

Let  $X$  be a discrete r.v.

$P(X=x)$  is given

Question. What can we say about the distribution of  $X$ , given that  $B$  occurs?

Answer. We can compute

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

Definition  $Y = X|B$  is a new r.v. whose distribution is given

by  $P(Y=x) = P(X=x|B)$

Example We toss a fair coin 2 times.

$X = \#$  of heads. What is the conditional distribution of  $X$  given that the 1st toss is tails? ... is heads?

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Answer.  $B_1 = \{1^{\text{st}} \text{ toss is heads}\}$

$B_2 = \{1^{\text{st}} \text{ toss is tails}\}$

(i)  $Y_1 = X | B_1$ ,  $X = 0, 1, 2$  - possible values of  $X$ .

$$P(Y_1 = 0) = P(X = 0 | B_1) = 0$$

$$P(Y_1 = 1) = P(X = 1 | B_1) = \frac{1}{2}.$$

$$P(Y_1 = 2) = P(X = 2 | B_1) = \frac{1}{2}.$$

(ii)  $Y_2 = X | B_2$

$$P(Y_2 = 0) = P(X = 0 | B_2) = \frac{1}{2}$$

$$P(Y_2 = 1) = \frac{1}{2}, \quad P(Y_2 = 2) = 0$$

In general,  $Y$  is a r.v. and hence

$$\sum_x P(Y=x) = 1.$$

Defn

$$E(X|B) = E(Y) = \sum_x x P(Y=x) =$$

$$\sum_x x \frac{P(X=x \cap B)}{P(B)} = \sum_x x P(X=x | B_i)$$

In the above example.

$$E(Y_1) = 0 \times 0 + 1 \times \frac{1}{2} + 2 \times \frac{1}{2} = 1.5$$

$$E(Y_2) = 0 \times \frac{1}{2} + 1 \times \frac{1}{2} + 2 \times 0 = 0.5$$

## The Theorem of Total Probability for Expectations (3)

(TPTE) theorem Let  $X$  be a discrete r.v., let  $B_1, \dots, B_n$  be a partition, and suppose that  $P(B_i) > 0$  for all  $i = 1, 2, \dots, n$ . Then

$$E(X) = \sum_{i=1}^n E(X|B_i) P(B_i)$$

Proof.  $E(X) = \sum_x x P(X=x)$ .

Since  $B_i$  form a partition.

$$P(X=x) = \sum_{i=1}^n P(X=x|B_i) P(B_i) \quad (\text{by the total prob. Th-m})$$

Hence

$$\begin{aligned} E(X) &= \sum_x x \sum_{i=1}^n P(X=x|B_i) P(B_i) \\ &= \sum_x \sum_{i=1}^n x P(X=x|B_i) P(B_i) \\ &= \sum_{i=1}^n \underbrace{\sum_x x P(X=x|B_i)}_{\text{---}} P(B_i) \\ &= \sum_{i=1}^n P(B_i) \underbrace{\sum_x x P(X=x|B_i)}_{E(X|B_i)} \end{aligned}$$

$$= \sum_i E(X|B_i) P(B_i)$$



(4)

Example. Alice and Bob play the following game. They roll a fair dice.

If it comes up 1 or 2, Alice wins.

If it comes up 3, Bob wins.

If it comes up 4, 5, 6, they play again.

What is the expected number of rolls?  
(They play until one of them wins).

Solution  $X$  = number of rolls.

$X$  takes values  $1, 2, 3, \dots, n, \dots$

and  $E(X) = \sum_{n=1}^{\infty} n P(X=n)$  - this

can be done. However, let us use the TPTE.

$B_1$  = the game ends after the 1<sup>st</sup> roll

$B_2 = B_1^c$  = the game does not end after the 1<sup>st</sup> roll.

$$P(B_1) = \frac{1}{2}, \quad P(B_2) = \frac{1}{2}.$$

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$$E(X) = P(B_1)E(X|B_1) + P(B_2)E(X|B_2) -$$

by the TPT. Next

$$E(X|B_1) = 1 \times 1 = 1.$$

$$E(X|B_2) = 1 + E(X). \text{ Hence}$$

$$E(X) = \frac{1}{2} \cdot 1 + \frac{1}{2} (1 + E(X)) = 1 + \frac{1}{2} E(X)$$

$$\frac{1}{2} E(X) = 1 \text{ and } \boxed{E(X) = 2}.$$

Example

geometrie distr-n. Geom(p).

X = the # of trials until 1<sup>st</sup> success.

p = Prob of success in each trial.

$$P(X=1) = p.$$

$$P(X=2) = (1-p)p = qp, q \stackrel{\text{def}}{=} 1-p$$

$$P(X=3) = q^2 p$$

$$P(X=n) = q^{n-1} p.$$

In the example,  $p = \frac{1}{2}$ ,  $q = \frac{1}{2}$ . So, we proved that

$$E(X) = 1 \cdot \frac{1}{2} + 2 \cdot \frac{1}{2^2} + 3 \cdot \frac{1}{2^3} + \dots + n \cdot \frac{1}{2^n} + \dots = 2.$$

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Exercise. Prove that if  $X \sim \text{Geom}(p)$ ,

then  $E(X) = \frac{1}{p}$ .

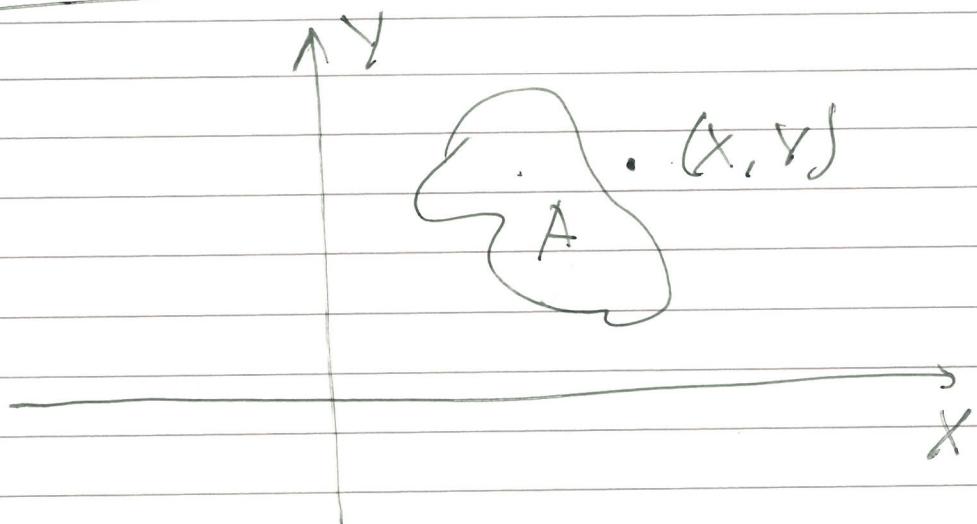
Hint. Use the TPTE.

Jointly continuous r.v.'s

Def-n. Let  $X, Y$  be r.v.'s. We say that  $X$  and  $Y$  are jointly continuous if there is a function  $f_{X,Y}(x,y)$  such that

$$P((X, Y) \in A) = \iint_A f_{X,Y}(x, y) dx dy,$$

where  $A \subset \mathbb{R}^2$  is any subset of  $\mathbb{R}^2$ .



$f_{X,Y}(x,y)$  is the joint probability density function (for  $X, Y$ ), or joint pdf.

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## Properties of the joint pdf of $X, Y$ .

$$(i) \iint_{-\infty}^{\infty} f_{X,Y}(s,t) ds dt = 1.$$

$\underbrace{\quad}_{P(-\infty < X < \infty, -\infty < Y < \infty)} = 1$

$$(ii) f_{X,Y}(s,t) \geq 0 \text{ for all } s, t.$$

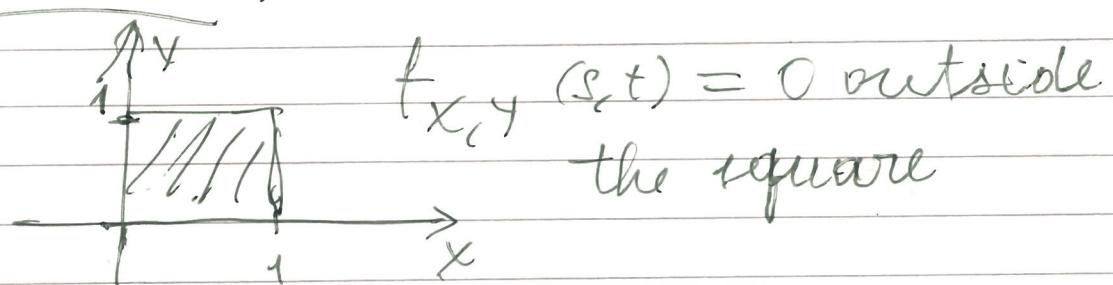
Example. Let  $X, Y$  have a joint pdf

given by

$$f_{X,Y}(x,y) = \begin{cases} c & \text{if } 0 < x < 1, 0 < y < 1 \\ 0 & \text{otherwise.} \end{cases}$$

Find  $c$ .

Solution. We use (i)



Here

$$\begin{aligned} 1 &= \iint_{-\infty}^{\infty} f_{X,Y}(s,t) ds dt = \iint_0^1 f_{X,Y}(s,t) ds dt \\ &= c \iint_0^1 ds dt = c \int_0^1 \left( \int_0^1 ds \right) dt = c \int_0^1 dt = c \\ &\quad \boxed{\int_0^1 dt = 1 \Rightarrow c = 1} \end{aligned}$$

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Example. Let  $X, Y$  be jointly cont. r.v.'s

$$f_{X,Y}(s,t) = \begin{cases} e^{-s-t} & \text{if } s, t > 0 \\ 0 & \text{otherwise.} \end{cases}$$

Find:

- (i)  $P(3 < X < 4 \text{ and } 2 < Y < 5)$
- (ii)  $P(-1 < X < 3 \text{ and } 2 < Y < 5)$
- (iii)  $P(Y > X > 0)$ .