

# Probability and statistics

## (September - 20)

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## Random variable:

R - random expt.

pmf/pdf  
cdf

$(\Omega, \mathcal{F}, P)$  -

$X$  r.v.  $X : \Omega \rightarrow \mathbb{R}$

$$x^{-1}(x) = \{ \omega : X(\omega) = x \} \in \mathcal{F}$$

## Random vector:

$(\Omega, \mathcal{F}, P)$

let  $X_1, X_2, \dots, X_r$  be  $r$  discrete random variables on  $(\Omega, \mathcal{F}, P)$ .

Define  $\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_r \end{bmatrix}$  is an  $r$ -dimensional vector.

For an  $w \in \Omega$

$$\underline{x}(w) = \begin{bmatrix} x_1(w) \\ x_2(w) \\ \vdots \\ x_r(w) \end{bmatrix} \in \mathbb{R}^r$$

Suppose  $x_1 = x_1(w); x_2 = x_2(w), \dots, x_r = x_r(w)$

$$\underline{x}(w) = \begin{bmatrix} x_1 \\ \vdots \\ x_r \end{bmatrix} \in \mathbb{R}^r$$

Definition:

Let  $\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_r \end{pmatrix}$

be a vector where

$x_i$  is a random variable on  $(\Omega, \mathcal{F}, P)$ .

For  $\underline{x} \in \mathbb{R}^r$  the set

$$\{\omega \in \Omega \mid \underline{x}(\omega) = \underline{x}\} \subset \mathcal{F}.$$

Then  $x: \Omega \rightarrow \mathbb{R}^r$  is called as an  $r$ -dimensional random vector.

We are interested in  $\text{Prob}(\underline{x} = \underline{x})$

Let  $\underline{x} = \begin{pmatrix} x_1(\omega) \\ \vdots \\ x_r(\omega) \end{pmatrix}$  be a random vector.

Let  $\underline{x}$  denote the value assumed by the random vector  $\underline{x}$ , then if

$\{\underline{x} \mid \text{Prob}(\underline{x}(\omega) = \underline{x}) > 0\}$  is finite or

countably infinite,

Then  $\underline{x}$  is a discrete random vector.

Definition: The discrete joint p.m.f. of the random vector  $\underline{x}$  is defined as:

$$f(x_1, \dots, x_r) = \text{Prob}(X_1 = x_1, X_2 = x_2, \dots, X_r = x_r)$$

where  $\underline{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_r \end{bmatrix}$  and  $\underline{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_r \end{bmatrix}$ .

In the vector notation

$$f(\underline{x}) = P(\underline{X} = \underline{x}) \quad \forall \underline{x} \in \mathbb{R}^r$$

For a subset  $A \subseteq \mathbb{R}^r$

$$P(\underline{X} \in A) = \sum_{\underline{x} \in A} f(\underline{x})$$

### Definition :

A function  $f$  is called as discrete joint p.m.f. if

- i)  $f(x) \geq 0$  &  $x \in \mathbb{R}^*$
  - ii)  $\{x \mid f(x) > 0\}$  is finite or countably infinite. This set  $x_1, x_2, x_3, \dots$
  - iii)  $\sum_i f(x_i) = 1$
- $\leftarrow R_x : \text{range of } x$

Ex: A discrete r.v.  $X_1$

$x_i$	1	2
$f(x_i)$	$\frac{1}{2}$	$\frac{1}{2}$

$$R_{X_1} = \{1, 2\}$$

Ex:

$x_1$	$x_2$	1	2	3	4
1		$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{16}$
2		$\frac{1}{16}$	$\frac{1}{16}$	$\frac{1}{4}$	$\frac{1}{8}$

Is this function a joint pmf??

All the three conditions are satisfied

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = x ; R_x = \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 1 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 4 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ 3 \end{pmatrix}, \begin{pmatrix} 2 \\ 4 \end{pmatrix} \right\}$$

$$\begin{aligned}
 \text{Prob}(x_1 > x_2) &= \text{Prob} \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 2 \end{pmatrix} \right\} \\
 &= \frac{1}{4} + \frac{1}{16} + \frac{1}{16} \\
 &= \frac{3}{8}
 \end{aligned}$$

$$\begin{aligned}
 \text{Prob}(x_1 = 1) &= \text{Prob}(x_1 = 1, x_2 = 1) \\
 &\quad + \text{Prob}(x_1 = 1, x_2 = 2) + \text{Prob}(x_1 = 1, x_2 = 3) \\
 &\quad + \text{Prob}(x_1 = 1, x_2 = 4) \\
 &= \frac{1}{4} + \frac{1}{8} + \frac{1}{16} + \frac{1}{16} \\
 &= \frac{1}{2}
 \end{aligned}$$

$$p_{rob}(x_1=1) = \sum_{x_2} f(1, x_2) = \frac{1}{2}$$

} pmf of  $x_1$

$$p_{rob}(x_1=2) = \sum_{x_2} f(2, x_2) = \frac{1}{2}$$

$$f_{x_1}(x_1) = \sum_{x_2} f(x_1, x_2) \quad \text{for } x_1 \in R_{x_1}$$

↑  
marginal pmf of  $x_1$

$x_1$	$f(x_1)$
1	1/2
2	1/2

: marginal pmf of  $x_1$

$x_2$	1	2	3	4
$f(x_2)$	5/16	3/16	5/16	3/16

: marginal pmf of  $x_2$

$$f_{x_2}(x_2) = \sum_{x_1} f(x_1, x_2) \quad \forall x_2 \in \mathcal{X}_{x_2}$$

marginal pmf of  $x_2$

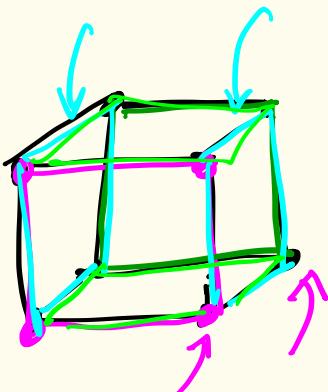
Ex: Toss three coins simultaneously.

Heads - 1

Tails - 0

$$\left\{ \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}, \right. \\ \left. \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}$$

joint pmf.



tensor

$$f_{x_1}(x_1) \sim \sum_{x_2, x_3} f(x_1, \boxed{x_2, x_3})$$

$x_i \in \mathbb{R}_{+}$

## Independent random variables:

Let  $x_1, x_2, \dots, x_r$  be  $r$  discrete random variables with p.m.f.s  $f_1, f_2, \dots, f_r$  respectively.

Then the random variables  $x_1, \dots, x_r$  are called as mutually independent if their joint p.m.f.  $f(x_1, \dots, x_r)$  is given as

$$f(x_1, \dots, x_r) = f_1(x_1) \cdot f_2(x_2) \cdot \dots \cdot f_r(x_r)$$

[Joint pmf can be written as the product of marginal p.m.f.s.]

### Notation:

$$f(x_1, \dots, x_r) = \text{Prob}(X_1 = x_1, X_2 = x_2, \dots, X_r = x_r)$$

Define  $A_i \subseteq \Omega$  s.t.  $A_i = [x_i = x_i]$   
 $= \{\omega \in \Omega \mid x_i(\omega) = x_i\}$

for  $i = 1, 2, \dots, r$

$$\begin{aligned} f(x_1, \dots, x_r) &= \text{Prob}(X_1 = x_1, X_2 = x_2, \dots, X_r = x_r) \\ &= \text{Prob}(A_1 \cap A_2 \cap \dots \cap A_r) \quad \text{independence of events} \\ &= \text{Prob}(A_1) \cdot \text{Prob}(A_2) \cdot \dots \cdot \text{Prob}(A_r) \\ &= \text{Prob}(x_1 = x_1) \cdot \text{Prob}(x_2 = x_2) \cdot \dots \cdot \text{Prob}(x_r = x_r) \end{aligned}$$

$$f(a_1, \dots, a_r) = f_1(x_1) \cdot f_2(x_2) \cdot \dots \cdot f_r(x_r)$$

Ex:

$x_1 \setminus x_2$	0	1	
0	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
1	$\frac{1}{4}$	$\frac{1}{4}$	$\frac{1}{2}$
	$\frac{1}{2}$	$\frac{1}{2}$	

Are  $x_1$  and  $x_2$  indep.??

$$f(x_1, x_2) = f_{x_1}(x_1) f_{x_2}(x_2)$$

Ex:

$x_1 \setminus x_2$	0	1	
0	$(1-p_1)(1-p_2)$	$(1-p_1)p_2$	
1	$(1-p_2)p_1$	$p_1 p_2$	

$$p_1 = p_2 = \frac{1}{2}$$

$\leftarrow$  independent  $x_1$  &  $x_2$  ??

$$f_{x_1}(0) = 1-p_1 \\ f_{x_1}(1) = p_1$$

$$f_{x_2}(0) = 1-p_2 \\ f_{x_2}(1) = p_2$$

Ex: Let  $x_1$  and  $x_2$  be two independent  
random variable each with geometric  
distribution with parameter  $p$ .

Find the distribution of  $\min(x_1, x_2) = z$ .

$$R_z = \{0, 1, 2, \dots\}$$

$$\begin{aligned} \text{Prob}(\min(x_1, x_2) \geq z) &= \text{Prob}(x_1 \geq z, x_2 \geq z) \\ &= \text{Prob}(x_1 \geq z) \cdot \text{Prob}(x_2 \geq z) \\ &= (1-p)^z \cdot (1-p)^z \\ &= ((1-p)^2)^z \end{aligned}$$