

## Expectation and Variance of Random Variables

But  $S \rightarrow$  Sample space

$X: S \rightarrow \mathbb{R}$  be a random variable.

$X =$  values of  $X$ .

$F_x: \Omega \rightarrow$  "Cumulative distribution function"  
cdf of  $X$ .

$P_x(x) \text{ or } f_x(x) \rightarrow$  "probability mass/density function"  
pmf/pdf of  $X$ .

We say  $X$  is

Discrete if

$F_x(x)$  is a step function

$\sum$

$X$  is a discrete subset  
of  $\mathbb{R}$

$\sum$

$F_x(x) = \sum_{y \in X} p_x(y)$

$\sum$

Calculations will usually

involve discrete math and

combinatorics.

$\sum$

$F_x(x) = P(X \leq x)$

$\sum$

$P(a \leq X \leq b) = F(b) - F(a)$

$a \rightarrow$  largest possible value of  $X$   
strictly less than  $a$ .

$$P(a \leq X \leq b) = \int_a^b f_x(x) dx = F_x(b) - F_x(a)$$

Note:

① If  $X$  is discrete and  $c \in X$ ,

$$P(X=c) = p_x(c)$$

If  $X$  is continuous,

$$P(X=c) = P(c \leq X \leq c) = \int_c^c f_x(x) dx = 0$$

for all  $c \in \mathbb{R}$ .

## Expected Value of Random Variable

Average or mean value of the random variable  $X$   
 $\sum$  denoted as

$E(X)$  or  $\mu_X$

If  $X$  is Discrete

$$E(X) = \sum_{x \in X} x \cdot p_x(x)$$

Suppose  $h(x)$  is a function of  $X$ .

$$E(h(x)) = \sum_{x \in X} h(x) \cdot p_x(x)$$

Note: The definition of "expected value" make sense as long as

$$\sum_{x \in X} x \cdot p_x(x)$$

or  $\int_{-\infty}^{\infty} x f_x(x) dx$  exists.

This comment exists because it is possible that

$$\sum_{x \in X} x \cdot p_x(x) \text{ or } \int_{-\infty}^{\infty} x f_x(x) dx \text{ might not exist!}$$

## Variance of a Random Variable

Captures the variation from the expected value  $\mu_X$

Denoted by  $V(X)$  or  $\sigma_X^2$

If  $X$  is Discrete

$$V(X) = E[(X - \mu_X)^2]$$

$$\sigma_X^2 = \sum_{x \in X} (x - \mu_X)^2 p_x(x)$$

If  $X$  is Continuous

$$V(X) = E[(X - \mu_X)^2]$$

$$\sigma_X^2 = \int_{-\infty}^{\infty} (x - \mu_X)^2 f_x(x) dx$$

The "standard deviation" is defined as  $\sigma_X = \sqrt{\sigma_X^2}$

## Properties of $\mu_X$ , $\sigma_X^2$ , and $\sigma_X$

① Suppose  $h(x) = ax + b$   $\rightarrow$  "linear function" of the r.v.  $X$ .

then

$$E[h(x)] = E[ax + b] = aE(x) + b$$

$$\text{i.e. } \mu_{ax+b} = a\mu_X + b.$$

$$V(h(x)) = a^2 V(x) \rightarrow \sigma_{ax+b}^2 = a^2 \sigma_X^2$$

and

$$\sigma_{ax+b} = |a| \sigma_X$$

② Shortcut Formula for Variance

$$V(X) = E[(X - \mu_X)^2] = E[X^2 - 2X\mu_X + \mu_X^2]$$

$$= E(X^2) - 2\mu_X E(X) + E(\mu_X^2)$$

$$\therefore V(X) = E(X^2) - (E(X))^2$$

Example:

Consider the experiment  $\rightarrow$  toss a fair coin four times.

$X =$  # of heads in four tosses.

Then the probability distribution table of  $X$  is

$n$	0	1	2	3	4
$P(X=n)$	$\frac{1}{16}$	$\frac{4}{16}$	$\frac{6}{16}$	$\frac{4}{16}$	$\frac{1}{16}$

$$\begin{aligned} E(X) &= 0 \cdot \frac{1}{16} + 1 \cdot \frac{4}{16} + 2 \cdot \frac{6}{16} + 3 \cdot \frac{4}{16} + 4 \cdot \frac{1}{16} \\ &= 0 \cdot \frac{1}{16} + 1 \cdot \frac{4}{16} + 2 \cdot \frac{6}{16} + 3 \cdot \frac{4}{16} + 4 \cdot \frac{1}{16} \\ &= \frac{4+12+12+4}{16} = \frac{32}{16} \\ &= 2 \end{aligned}$$

$\therefore$  If we toss a fair coin 4 times repeatedly on average we would expect to get 2 heads in 4 tosses.

## To calculate Variance

$$\begin{aligned} V(X) &= \sum_{x \in X} (x - \mu_X)^2 \cdot p_x(x) \\ &= (0-2)^2 \cdot \frac{1}{16} + (1-2)^2 \cdot \frac{4}{16} + (2-2)^2 \cdot \frac{6}{16} + (3-2)^2 \cdot \frac{4}{16} + (4-2)^2 \cdot \frac{1}{16} \\ &= \frac{4}{16} + \frac{4}{16} + \frac{6}{16} + \frac{4}{16} \\ &= \frac{16}{16} = 1. \end{aligned}$$

Note that,

$$\begin{aligned} E(X^2) &= 0 \cdot \frac{1}{16} + 1 \cdot \frac{4}{16} + 2^2 \cdot \frac{6}{16} + 3^2 \cdot \frac{4}{16} + 4^2 \cdot \frac{1}{16} \\ &= 4 + 4 \cdot 6 + 9 \cdot 4 + 16 \cdot 1 = 4 + 24 + 36 + 16 \\ &= \frac{80}{16} = 5 \end{aligned}$$

$$\text{Also } (E(X))^2 = (2)^2 = 4.$$

$$\therefore V(X) = E(X^2) - (E(X))^2 = 5 - 4 = 1$$

$\therefore$  Same as calculated above!!!