

# Lecture 14:

## Expectation and Variance - Part I

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# Expectation

- ▶ Let  $X$  be a random variable describing the amount won in a game.
- ▶ A question is: How much does a person win “on average” in a game?
- ▶ For example, Let  $X$  be the amount won and is equal to the outcome of a fair dice. Then how much does a person win “on average”? or What is the “expected value” of winning amount?

$$\frac{1+2+3+4+5+6}{6} = \frac{21}{6} = 3.5.$$

- ▶ This is the same as the arithmetic mean of  $n$  real numbers  $x_1, \dots, x_n$  defined as

$$\frac{x_1 + \dots + x_n}{n}.$$

- ▶ If the dice is biased (not fair) then how much a person wins “on average” in a game?

# Expectation

- ▶ If  $X$  is a discrete r.v. taking on the possible values  $x_1, x_2, \dots$ , then the **expectation** or **expected value** or **mean** of  $X$ , denoted by  $E(X)$ , is defined as

$$E(X) = \sum_i x_i P(X = x_i).$$

- ▶ **Example:** If the dice is biased (not fair) such that

$$P(X = 1) = \underline{0}, P(X = 2) = P(X = 3) = P(X = 4) = \underline{1/6},$$

$$P(X = 5) = P(X = 6) = \underline{1/4},$$

then how much a person wins “on average” in a game?

$$\begin{aligned} E(X) &= 0 \cdot 1 + (2 + 3 + 4) \cdot \frac{1}{6} + (5 + 6) \cdot \frac{1}{4} \\ &= \frac{3}{2} + \frac{11}{4} = \frac{17}{4} = 4.25. \end{aligned}$$

# Expectation

- **Example:** Find  $E(X)$  if  $X \sim \text{Bern}(p)$ .

$$E(X) = 1 \cdot p + 0 \cdot (1-p) = p.$$

- **Example:** What is the expected value of a binomially distributed r.v.?

- Recall:  $X \sim \text{Bin}(n, p)$  if

$$P(X=k) = \binom{n}{k} p^k (1-p)^{n-k} \text{ for } k=0, 1, \dots, n$$
$$= 0 \text{ otherwise.}$$

- We will use the following identity:

$$k \binom{n}{k} = k \frac{n!}{k! (n-k)!} = \frac{\cancel{k} \cdot n \cdot (n-1)!}{\cancel{k} (k-1)! (n-k)!} = n \cdot \frac{(n-1)!}{(k-1)! (n-k)!} = n \binom{n-1}{k-1}$$

# Expectation

- Now,  $E(X) = \sum_{k=0}^n k P(X=k)$

$$= \sum_{k=1}^n k \binom{n}{k} p^k (1-p)^{n-k} \quad (\because k=0 \Rightarrow k P(X=k)=0)$$

$$= n \sum_{k=1}^n \binom{n-1}{k-1} p^k (1-p)^{n-k} \quad (\because \text{using the identity})$$

$$= np \sum_{k=1}^n \binom{n-1}{k-1} p^{k-1} (1-p)^{n-k}$$

$$= np \sum_{j=0}^{n-1} \underbrace{\binom{n-1}{j} p^j (1-p)^{n-1-j}}_{\text{PMF of Bin}(n-1, p)}$$

$$= np. \quad \underbrace{\text{PMF of Bin}(n-1, p)}_{\text{PMF always sums up to 1}} \text{ for } j=0, 1, \dots, n-1.$$

# Properties of expectation

law of the unconscious statistician

- **Expectation of a function of r.v.:** If  $X$  is a discrete r.v. and  $g : \mathbb{R} \rightarrow \mathbb{R}$  is a function, then

$$E(g(X)) = \sum_x g(x)P(X = x) = \sum_x g(x)p_X(x) \quad \}$$

where the sum is taken over all possible values of  $X$ .

- **How?:** Recall that an r.v.  $X$  is a function  $X : S \rightarrow \mathbb{R}$ .

$$\begin{aligned} \text{Hence, } E(g(X)) &= \sum_{s \in S} g(X(s)) P(\{s\}) \\ &= \sum_{x \in \mathbb{R}} \sum_{s : X(s) = x} g(x) P(\{s\}) \\ &= \sum_{x \in \mathbb{R}} g(x) \sum_{s : X(s) = x} P(\{s\}) \\ &= \sum_x g(x) P(X = x). \end{aligned}$$

See the example in  
the last slide for  
understanding /  
demonstration.

# Properties of expectation

- ▶ Similarly, for a function of two r.v.s, we have

$$\begin{aligned} E(g(X, Y)) &= \sum_x \sum_y g(x, y) P(X = x, Y = y) \\ &= \sum_x \sum_y g(x, y) p_{X,Y}(x, y). \end{aligned}$$

Example:

- ▶ Suppose  $X$  has the following PMF

$$p_X(0) = .2, p_X(1) = .5, p_X(2) = .3.$$

Find  $E[X^2]$ .

Let  $Y = g(X) = X^2$ . Then,

$$\begin{aligned} E(X^2) &= E(Y) = 0^2 \cdot (0.2) + 1^2 \cdot (0.5) + 2^2 \cdot (0.3) \\ &= 1.7. \end{aligned}$$

# Properties of expectation

► **Linearity**: Expectation is linear, i.e., for r.v.s  $X$  and  $Y$  and a constant  $c$ ,

1.  $E(X + Y) = E(X) + E(Y)$ , Let  $g(X, Y) = X + Y$ .
2.  $E(cX) = cE(X)$ .

$$\begin{aligned} E(X + Y) &= E(g(X, Y)) = \sum_x \sum_y g(x, y) p_{X,Y}(x, y) \\ &= \sum_x \sum_y (x + y) p_{X,Y}(x, y) \end{aligned}$$

$$\begin{aligned} E(cX) &= \sum_x c x p_X(x) \\ &= c \sum_x x p_X(x) \\ &= c E(X). \end{aligned}$$

$$\begin{aligned} &= \sum_x \sum_y x p_{X,Y}(x, y) + \sum_x \sum_y y p_{X,Y}(x, y) \\ &= \sum_x x \sum_y p_{X,Y}(x, y) + \sum_y y \sum_x p_{X,Y}(x, y) \\ &= \sum_x x p_X(x) + \sum_y y p_Y(y) \\ &= E(X) + E(Y) \end{aligned}$$



# Properties of expectation

- **Example:** A construction firm has sent in bids for 3 jobs worth (in profits) 10, 20, and 40 (lakh) rupees. If its probabilities of winning the jobs are respectively .2, .8, and .3, what is the firm's expected total profit?

- Let  $X_i$  be the profit from job  $i$ ,  $i \in \{1, 2, 3\}$ .

- Total profit =  $X_1 + X_2 + X_3$ .

$$\begin{aligned} \text{- Expected total profit} &= E(X_1 + X_2 + X_3) \\ &= E(X_1) + E(X_2) + E(X_3) \\ &= 10 \cdot 0.2 + 20 \cdot 0.8 + 40 \cdot 0.3 \\ &= 2 + 16 + 12 \\ &= 30 \text{ Lakh rupees.} \end{aligned}$$

Example:

Let  $S = \{a, b, c, d, e, f\}$

$X: S \rightarrow \mathbb{R}$  such that  $X(a) = 0, X(b) = 0$   
 $X(c) = 1, X(d) = 2$   
 $X(e) = 3, X(f) = 3$

Let  $P: S \rightarrow [0, 1]$  such that  $P(s) = \frac{1}{6}, s \in S$ .

Then,  $P_X(0) = \frac{1}{3} = P_X(3), P_X(1) = P_X(2) = \frac{1}{6}$ .

Let  $g$  be the function of  $X$  such that

$$g(0) = g(1) = 10$$

$$g(2) = g(3) = 20$$

$$\text{Then, } E(g(X)) = \sum_{s \in S} g(X(s)) P(s)$$

$$= g(X(a))P(a) + g(X(b))P(b) \\ + g(X(c))P(c) + g(X(d))P(d) \\ + g(X(e))P(e) + g(X(f))P(f)$$

$$= g(0)P(a) + g(0)P(b) \\ + g(1)P(c) + g(2)P(d) \\ + g(3)P(e) + g(3)P(f)$$

$$= \sum_{x=0}^3 \sum_{s: X(s)=x} g(x) P(s)$$

$$= \sum_{x=0}^3 g(x) \underbrace{P(X=x)}_{\sum_{s: X(s)=x} P(s)}$$