

# Recitation 12

# Random Variables

A random variable depends on the outcome of an uncertain event and takes specific values with specific probabilities (called a distribution)

Example: Let  $X$  denote the outcome of the roll of a fair die. Then  $X$  is a random variable with values  $\{1, 2, 3, 4, 5, 6\}$  each with a probability of  $1/6$ .

Note: "Random" does not necessarily mean random with equal chance, we can have  $X$  be  $\{1, 2, 3, 4, 5, 6\}$  with probabilities  $\{1/2, 1/10, 1/10, 1/10, 1/10, 1/10\}$  (This can be written as  $P(X=1) = 1/2$ ,  $P(X=2) = 1/10$ , etc. Where  $P$  is the Probability Mass Function)

# Expectation

The Expected Value of a RV  $X$ , denoted as  $E[X]$  is the "average" value attained by the RV if we repeat  $X$  a large number of times. We denote the "average" or mean of  $X$  as  $\mu_x$

$$E[X] = \sum_{\text{all possible } x} xP(X = x).$$

Example: The expected value of the roll of a die is

$$1\left(\frac{1}{6}\right) + 2\left(\frac{1}{6}\right) + 3\left(\frac{1}{6}\right) + \cdots + 6\left(\frac{1}{6}\right) = 21/6 = 3.5.$$

# Expectation

We also define the expected value for a function of a random variable. If  $g$  is a function (for example,  $g(x) = x^2$ ), then the expected value of  $g(X)$  is

$$E[g(X)] = \sum_{\text{all possible } x} g(x)P(X = x).$$

For example,

$$E[X^2] = \sum_{\text{all possible } x} x^2 P(X = x).$$

In general,  $E[g(X)]$  is not the same as  $g(E[X])$ . In particular,  $E[X^2]$  is not the same as  $(E[X])^2$ .

Ex: we roll a four-sided die,  $X = \{1, 2, 3, 4\}$ ,  $P = \{\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}\}$

$$E[X]^2 = [(1 * (\frac{1}{4})) + (2 * (\frac{1}{4})) + (3 * (\frac{1}{4})) + (4 * (\frac{1}{4}))]^2$$

$$E[X^2] = [(1^2 * (\frac{1}{4})) + (2^2 * (\frac{1}{4})) + (3^2 * (\frac{1}{4})) + (4^2 * (\frac{1}{4}))]$$

**Theorem 2.1 [Linearity of Expectations]:** For any finite collection of discrete random variables  $X_1, X_2, \dots, X_n$  with finite expectations,

$$\mathbf{E}\left[\sum_{i=1}^n X_i\right] = \sum_{i=1}^n \mathbf{E}[X_i].$$

**Proof:** We prove the statement for two random variables  $X$  and  $Y$ ; the general case follows by induction. The summations that follow are understood to be over the ranges of the corresponding random variables:

$$\begin{aligned}\mathbf{E}[X + Y] &= \sum_i \sum_j (i + j) \Pr((X = i) \cap (Y = j)) \\ &= \sum_i \sum_j i \Pr((X = i) \cap (Y = j)) + \sum_i \sum_j j \Pr((X = i) \cap (Y = j)) \\ &= \sum_i i \sum_j \Pr((X = i) \cap (Y = j)) + \sum_j j \sum_i \Pr((X = i) \cap (Y = j)) \\ &= \sum_i i \Pr(X = i) + \sum_j j \Pr(Y = j) \\ &= \mathbf{E}[X] + \mathbf{E}[Y].\end{aligned}$$

# Variance

The **variance** of a random variable  $X$  is denoted by either  $Var[X]$  or  $\sigma_X^2$ . ( $\sigma$  is the Greek letter sigma.) The variance is defined by

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

this is the expected value of the squared difference between  $X$  and its mean. For a discrete distribution, we can write the variance as

$$\sigma_X^2 = \sum_x (x - \mu_X)^2 P(X = x).$$

An alternative expression for the variance (valid for both discrete and continuous random variables) is

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

This is the difference between the expected value of  $X^2$  and the square of the mean of  $X$ .

The **standard deviation** of a random variable is the square-root of its variance and is denoted by  $\sigma_X$ . Generally speaking, the greater the standard deviation, the more spread-out the possible values of the random variable.

# Problem 1

Let  $X$  be a discrete random variable with the following PMF

$$P_X(k) = \begin{cases} \frac{1}{4} & \text{for } k = -2 \\ \frac{1}{8} & \text{for } k = -1 \\ \frac{1}{8} & \text{for } k = 0 \\ \frac{1}{4} & \text{for } k = 1 \\ \frac{1}{4} & \text{for } k = 2 \\ 0 & \text{otherwise} \end{cases}$$

I define a new random variable  $Y$  as  $Y = (X + 1)^2$ .

- Find the range of  $Y$ .
- Find the PMF of  $Y$ .

# Problem 1

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I define a new random variable  $Y$  as  $Y = (X + 1)^2$ .

- Find the range of  $Y$ .
- Find the PMF of  $Y$ .

Here, the random variable  $Y$  is a function of the random variable  $X$ . This means that we perform the random experiment and obtain  $X = x$ , and then the value of  $Y$  is determined as  $Y = (x + 1)^2$ . Since  $X$  is a random variable,  $Y$  is also a random variable.

- a. To find  $R_Y$ , we note that  $R_X = \{-2, -1, 0, 1, 2\}$ , and

$$\begin{aligned} R_Y &= \{y = (x + 1)^2 \mid x \in R_X\} \\ &= \{0, 1, 4, 9\}. \end{aligned}$$

- b. Now that we have found  $R_Y = \{0, 1, 4, 9\}$ , to find the PMF of  $Y$  we need to find  $P_Y(0), P_Y(1), P_Y(4)$ , and  $P_Y(9)$ :

$$\begin{aligned} P_Y(0) &= P(Y = 0) = P((X + 1)^2 = 0) \\ &= P(X = -1) = \frac{1}{8}; \end{aligned}$$

$$\begin{aligned} P_Y(1) &= P(Y = 1) = P((X + 1)^2 = 1) \\ &= P((X = -2) \text{ or } (X = 0)); \end{aligned}$$

$$P_X(-2) + P_X(0) = \frac{1}{4} + \frac{1}{8} = \frac{3}{8};$$

$$\begin{aligned} P_Y(4) &= P(Y = 4) = P((X + 1)^2 = 4) \\ &= P(X = 1) = \frac{1}{4}; \end{aligned}$$

$$\begin{aligned} P_Y(9) &= P(Y = 9) = P((X + 1)^2 = 9) \\ &= P(X = 2) = \frac{1}{4}. \end{aligned}$$

Again, it is always a good idea to check that  $\sum_{y \in R_Y} P_Y(y) = 1$ . We have

$$\sum_{y \in R_Y} P_Y(y) = \frac{1}{8} + \frac{3}{8} + \frac{1}{4} + \frac{1}{4} = 1.$$



## Problem 2

Let  $X$  be a discrete random variable with the following PMF

$$P_X(k) = \begin{cases} 0.1 & \text{for } k = 0 \\ 0.4 & \text{for } k = 1 \\ 0.3 & \text{for } k = 2 \\ 0.2 & \text{for } k = 3 \\ 0 & \text{otherwise} \end{cases}$$

- Find  $EX$ .
- Find  $\text{Var}(X)$ .
- If  $Y = (X - 2)^2$ , find  $EY$ .

# Problem 2a

Let  $X$  be a discrete random variable with the following PMF

$$P_X(k) = \begin{cases} 0.1 & \text{for } k = 0 \\ 0.4 & \text{for } k = 1 \\ 0.3 & \text{for } k = 2 \\ 0.2 & \text{for } k = 3 \\ 0 & \text{otherwise} \end{cases}$$

- a. Find  $EX$ .
- b. Find  $\text{Var}(X)$ .
- c. If  $Y = (X - 2)^2$ , find  $EY$ .

$$\begin{aligned} EX &= \sum_{x_k \in \mathcal{R}_X} x_k P_X(x_k) \\ &= 0(0.1) + 1(0.4) + 2(0.3) + 3(0.2) \\ &= 1.6 \end{aligned}$$

## Problem 2b

Let  $X$  be a discrete random variable with the following PMF

$$P_X(k) = \begin{cases} 0.1 & \text{for } k = 0 \\ 0.4 & \text{for } k = 1 \\ 0.3 & \text{for } k = 2 \\ 0.2 & \text{for } k = 3 \\ 0 & \text{otherwise} \end{cases}$$

- a. Find  $EX$ .
- b. Find  $\text{Var}(X)$ .
- c. If  $Y = (X - 2)^2$ , find  $EY$ .

We can use  $\text{Var}(X) = EX^2 - (EX)^2 = EX^2 - (1.6)^2$ . Thus we need to find  $EX^2$ . Using LOTUS, we have

$$EX^2 = 0^2(0.1) + 1^2(0.4) + 2^2(0.3) + 3^2(0.2) = 3.4$$

Thus, we have

$$\text{Var}(X) = (3.4) - (1.6)^2 = 0.84$$

# Problem 2c

Let  $X$  be a discrete random variable with the following PMF

$$P_X(k) = \begin{cases} 0.1 & \text{for } k = 0 \\ 0.4 & \text{for } k = 1 \\ 0.3 & \text{for } k = 2 \\ 0.2 & \text{for } k = 3 \\ 0 & \text{otherwise} \end{cases}$$

- a. Find  $EX$ .
- b. Find  $\text{Var}(X)$ .
- c. If  $Y = (X - 2)^2$ , find  $EY$ .

$$Y = (X-2)*(X-2) = X^2 - 4X + 4$$

$$E[Y] = E[X^2 - 4X + 4] = E[X^2] - E[4X] + E[4] = E[X^2] - 4E[X] + 4 \text{ using linearity of expectation}$$

$$E[X] = 0*(0.1) + 1*(0.4) + 2*(0.3) + 3*(0.2) = (0.4) + 2*(0.3) + 3*(0.2)$$

$$E[X^2] = (0^2)*(0.1) + (1^2)*(0.4) + (2^2)*(0.3) + (3^2)*(0.2) = (0.4) + 4*(0.3) + 9*(0.2)$$

$$E[Y] = [(0.4) + 2*(0.3) + 3*(0.2)] + [(0.4) + 4*(0.3) + 9*(0.2)] + 4$$

# Challenge Problem

You take an exam that contains 20 multiple-choice questions. Each question has 4 possible options. You know the answer to 10 questions, but you have no idea about the other 10 questions so you choose answers randomly. Your score  $X$  on the exam is the total number of correct answers. Find the PMF of  $X$ . What is  $P(X > 15)$ ?

# Challenge Problem

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Let's define the random variable  $Y$  as the number of your correct answers to the 10 questions you answer randomly. Then your total score will be  $X = Y + 10$ . First, let's find the PMF of  $Y$ . For each question your success probability is  $\frac{1}{4}$ . Hence, you perform 10 independent Bernoulli( $\frac{1}{4}$ ) trials and  $Y$  is the number of successes. Thus, we conclude  $Y \sim \text{Binomial}(10, \frac{1}{4})$ , so

$$P_Y(y) = \begin{cases} \binom{10}{y} \left(\frac{1}{4}\right)^y \left(\frac{3}{4}\right)^{10-y} & \text{for } y = 0, 1, 2, 3, \dots, 10 \\ 0 & \text{otherwise} \end{cases}$$

Now we need to find the PMF of  $X = Y + 10$ . First note that  $R_X = \{10, 11, 12, \dots, 20\}$ . We can write

$$\begin{aligned} P_X(10) &= P(X = 10) = P(Y + 10 = 10) \\ &= P(Y = 0) = \binom{10}{0} \left(\frac{1}{4}\right)^0 \left(\frac{3}{4}\right)^{10-0} = \left(\frac{3}{4}\right)^{10}; \\ P_X(11) &= P(X = 11) = P(Y + 10 = 11) \\ &= P(Y = 1) = \binom{10}{1} \left(\frac{1}{4}\right)^1 \left(\frac{3}{4}\right)^{10-1} = 10 \left(\frac{1}{4}\right) \left(\frac{3}{4}\right)^9. \end{aligned}$$

So, you get the idea. In general for  $k \in R_X = \{10, 11, 12, \dots, 20\}$ ,

$$\begin{aligned} P_X(k) &= P(X = k) = P(Y + 10 = k) \\ &= P(Y = k - 10) = \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k}. \end{aligned}$$

# Challenge Problem

You take an exam that contains 20 multiple-choice questions. Each question has 4 possible options. You know the answer to 10 questions, but you have no idea about the other 10 questions so you choose answers randomly. Your score  $X$  on the exam is the total number of correct answers. Find the PMF of  $X$ . What is  $P(X > 15)$ ?

To summarize,

$$P_X(k) = \begin{cases} \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k} & \text{for } k = 10, 11, 12, \dots, 20 \\ 0 & \text{otherwise} \end{cases}$$

In order to calculate  $P(X > 15)$ , we know we should consider  $y = 6, 7, 8, 9, 10$

$$P_Y(y) = \begin{cases} \binom{10}{y} \left(\frac{1}{4}\right)^y \left(\frac{3}{4}\right)^{10-y} & \text{for } y = 6, 7, 8, 9, 10 \\ 0 & \text{otherwise} \end{cases}$$

$$P_X(k) = \begin{cases} \binom{10}{k-10} \left(\frac{1}{4}\right)^{k-10} \left(\frac{3}{4}\right)^{20-k} & \text{for } k = 16, 17, \dots, 20 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} P(X > 15) &= P_X(16) + P_X(17) + P_X(18) + P_X(19) + P_X(20) \\ &= \binom{10}{6} \left(\frac{1}{4}\right)^6 \left(\frac{3}{4}\right)^4 + \binom{10}{7} \left(\frac{1}{4}\right)^7 \left(\frac{3}{4}\right)^3 + \binom{10}{8} \left(\frac{1}{4}\right)^8 \left(\frac{3}{4}\right)^2 \\ &\quad + \binom{10}{9} \left(\frac{1}{4}\right)^9 \left(\frac{3}{4}\right)^1 + \binom{10}{10} \left(\frac{1}{4}\right)^{10} \left(\frac{3}{4}\right)^0. \end{aligned}$$

# FINAL QUIZ!!

- You will have 25 minutes to complete the Quiz
- Final Exam will be on Dec 20th