

# Chapter 4 part 1

## Discrete Random Variables

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MATH 241

# Outline

- 1 Discrete random variables
- 2 Expectation

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1 Discrete random variables

2 Expectation

# Random Variables

## Definition

*Random Variable  $X$  is a real-valued function on the sample space  $S$ .*

- Random variable is a number associated with a random experiment.
- Random variables are in essence a fancy way of describing an event.

## Question

Two fair six-sided dice are rolled. Let the random variable  $X$  denote the product of the 2 dice. What are possible values of  $X$  and their associated probabilities? Just give a few examples and you do not need to calculate the associated probability for all possible values.

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$$P(X = 1) = P(1, 1) = 1/6 \times 1/6$$

$$P(X = 2) = P(1, 2) + P(2, 1) = 1/6 \times 1/6 \times 2$$

$$P(X = 3) = P(1, 3) + P(3, 1) = 1/6 \times 1/6 \times 2$$

.....

$$P(X = 36) = P(6, 6) = 1/6 \times 1/6$$

# Cumulative distribution function (cdf)

## Definition

For a random variable  $X$ , the function  $F$  defined by

$$F(x) = P(X \leq x), \quad -\infty < x < \infty$$

is called the *cumulative distribution function* of  $X$ .

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Note that

- Capital  $X$ : random variable
- Little  $x$ : a real-valued number
- $\leq$ : smaller than or equal to



# Properties of the cdf (Ch 4.10)

- $P(a < X \leq b) = F(b) - F(a)$ , for all  $a < b$
- $P(X < b)$  does not necessary equal to  $P(X \leq b)$ .
- ①  $F(x)$  is a non-decreasing function; i.e., if  $x_1 < x_2$ , then

$$F(x_1) \leq F(x_2)$$

- ② 
$$\lim_{x \rightarrow \infty} F(x) = 1 \iff P(X \leq \infty) = 1$$

- ③ 
$$\lim_{x \rightarrow -\infty} F(x) = 0 \iff P(X > -\infty) = 1$$

- ④  $F(x)$  is right continuous; i.e., for any decreasing sequence  $\{x_n : n = 1, 2, \dots\}$  that converges to  $x$ ,

$$\lim_{n \rightarrow \infty} F(x_n) = F(x) \implies \lim_{n \rightarrow \infty} P(X \leq x + \frac{1}{n}) = P(X \leq x)$$

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Part (a)

$$\begin{aligned} P(X \leq 2) &= P(X = 1) + P(X = 2) \\ &= P(1, 1) + P(1, 2) + P(2, 1) = 1/6 \times 1/6 \times 3 \end{aligned}$$

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Part (b)

$$\begin{aligned}P(X \leq 35) &= P(X = 1) + P(X = 2) + \cdots + P(X = 35) \\&= 1 - P(X = 36) \\&= 1 - P(6, 6) \\&= 1 - 1/6 \times 1/6\end{aligned}$$

# Discrete random variables

## Definition

- 👉 A random variable  $X$  that can take on at most a countable number of possible values is a *discrete random variable*.
- 👉 For a discrete random variable  $X$ , we define the *probability mass function* (pmf) by

$$p(x) = P(X = x)$$

Note: pmf is also written as  $f(x)$ .

# pmf and cdf

- For a discrete random variable  $X$ , there exists a countable sequence  $x_1, x_2, \dots$ , such that

$$p(x_i) > 0 \quad \text{for } i = 1, 2, \dots$$

$$p(x) = 0 \quad \text{for all other values of } x$$

and

$$\sum_{i=1}^{\infty} p(x_i) = 1$$

- Relationship between pmf and cdf (for discrete random variable)

$$F(a) = \sum_{\text{all } x \leq a} p(x)$$

- If we know pmf, we can compute cdf. And vice versa.

## Question

Three fair coins are tossed. Let the random variable  $X$  denote the number of heads. Write out the pmf and cdf of  $X$ .

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For pmf:

$$P(X = 0) = P(TTT) = (1/2)^3 = 1/8$$

$$P(X = 1) = P(HTT) + P(THT) + P(TTH) = (1/2)^3 \times 3 = 3/8$$

$$P(X = 2) = P(HHT) + P(HTH) + P(THH) = (1/2)^3 \times 3 = 3/8$$

$$P(X = 3) = P(HHH) = (1/2)^3 = 1/8$$



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For pmf:

$$F(a) = \begin{cases} 1/8, & 0 \leq a < 1 \\ 1/2, & 1 \leq a < 2 \\ 7/8, & 2 \leq a < 3 \\ 1, & 3 \leq a \end{cases}$$

## Question

Suppose that the distribution function of  $X$  is given by

$$F(b) = \begin{cases} 0, & b < 0 \\ \frac{b}{4}, & 0 \leq b < 1 \\ \frac{1}{2} + \frac{b-1}{4}, & 1 \leq b < 2 \\ \frac{11}{12}, & 2 \leq b < 3 \\ 1, & 3 \leq b \end{cases}$$

Find  $P(X = i), i = 1, 2, 3$ .

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Find  $P(X = i), i = 1, 2, 3$ .

- $P(X = 1) = P(X \leq 1) - P(X < 1) = \frac{1}{2} - \frac{1}{4} = \frac{1}{4}$

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- $P(X = 2) = P(X \leq 2) - P(X < 2) = \frac{11}{12} - \left(\frac{1}{2} + \frac{2-1}{4}\right) = \frac{11}{12} - \frac{3}{4} = \frac{1}{6}$

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- $P(X = 3) = P(X \leq 3) - P(X < 3) = 1 - \frac{11}{12} = \frac{1}{12}$

# Recap

## Random Variable

- Random Variable  $X$  is a real-valued function on the sample space  $S$ .

$$X : S \longrightarrow \mathbb{R}$$

- Cumulative distribution function (cdf)

$$F_X(x) = P(X \leq x), \quad \text{for any } x \in \mathbb{R}$$

## Discrete random variable

- can only take at most a countable number of possible values.
- probability mass function (pmf)

$$p_X(x) = P(X = x)$$

$$\bullet \sum_{i=1}^{\infty} p(x_i) = 1$$

$$\bullet F(a) = \sum_{\text{all } x \leq a} p(x)$$

# Outline

1 Discrete random variables

2 **Expectation**

# Expected value

## Definition

The *expected value* (or *mean*) of a discrete random variable is defined as

$$E[X] = \sum_{x:p(x)>0} x \cdot P(X = x) = \sum_{x:p(x)>0} xp(x)$$

- $E[X]$  is a weighted average of the possible values  $x$  that  $X$  can take on, each value being weighted by the probability  $p(x)$ .



When she told me I was average,  
she was just being mean.



## Question

Toss a coin. Suppose the probability of a head is  $p$ . Let  $X$  be a 0-1 indicator random variable s.t.

$$X = \begin{cases} 1 & \text{if head is obtained} \\ 0 & \text{otherwise} \end{cases}$$

Compute  $\mu = E[X]$ .

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

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- In general, for indicator variable  $X = \delta_A$  or  $\mathbf{1}_A$ , denoted as

$$X = \begin{cases} 1 & \text{if event } A \text{ occurs} \\ 0 & \text{otherwise} \end{cases}$$

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$$X = \begin{cases} 1 & \text{if event } A \text{ occurs} \\ 0 & \text{otherwise} \end{cases}$$

The expected value of  $X$  equals the probability that  $A$  occurs.

$$E[X] = 1 \cdot P(A) + 0 \cdot P(A^c) = P(A)$$

## Question

Let the random variable  $X$  denote the GP a certain student will earn in this class. Suppose its pmf is

$$p(0) = 0.05, \quad p(1) = 0.05, \quad p(2) = 0.3, \quad p(3) = 0.4$$

Calculate their expected GP  $E[X]$ .

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Calculate their expected GP  $E[X]$ .

Since the domain of  $X$  (all possible values  $X$  can take) is  $\{0, 1, 2, 3, 4\}$ , first compute  $p(4)$ .

$$p(4) = 1 - p(0) - p(1) - p(2) - p(3) = 0.2$$

Then compute  $E[X]$

$$E[X] = 0 \times 0.05 + 1 \times 0.05 + 2 \times 0.3 + 3 \times 0.4 + 4 \times 0.2 = 2.65$$

## Expectation of a function of a random variable

- If  $X$  is a discrete random variable, and  $g$  is a real-valued function then the expectation (or expected value) of  $Y = g(X)$  is

$$E[g(X)] = \sum_{x:p_X(x)>0} g(x) \cdot p_X(x)$$

## Question

Let  $X$  denote a random variable that takes on any of the values  $-1$ ,  $0$ , and  $2$  with respective probability:  $P(X = -1) = \frac{1}{5}$ ,  $P(X = 0) = \frac{1}{5}$ ,  $P(X = 2) = \frac{3}{5}$ . Compute  $E[X^3]$ .



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Let  $Y = X^3$ , then

$$\begin{aligned} E[X^3] &= E[Y] = \sum_{\text{all } x} x^3 \cdot p_X(x) \\ &= (-1)^3 \times \frac{1}{5} + 0^3 \times \frac{1}{5} + 2^3 \times \frac{3}{5} \\ &= -\frac{1}{5} + 0 + \frac{24}{5} \\ &= \frac{24}{5} \end{aligned}$$

# Properties of expected values

If  $a$  and  $b$  are constants, then

$$E[aX + b] = aE[X] + b$$

- Holds for all random variable  $X$  (not necessary discrete random variable).
- Special cases of linear transformation  $E[aX + b] = aE[X] + b$ 
  - ▶ constant factor

$$E[aX] = aE[X]$$

- ▶ constant

$$E[b] = b$$

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## Question

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Let  $Y = X^2$ , then

$$\begin{aligned}
 E[X^2] &= E[Y] = \sum_{\text{all } x} x^2 \cdot p_X(x) \\
 &= (-1)^2 \times \frac{1}{5} + 0^2 \times \frac{1}{5} + 2^2 \times \frac{3}{5} \\
 &= \frac{1}{5} + 0 + \frac{12}{5} \\
 &= \frac{13}{5}
 \end{aligned}$$

Part (a):  $E[2X^2] = 2E[X^2] = 2E[X^2] = 26/5$

Part (b):  $E[4X^2 - 1] = 4E[X^2] - 1 = 52/5 - 1 = 47/5$

# Recap

Expectation  $\mu$

- **For discrete random variable:**  $E[X] = \sum_{\text{all } x} x \cdot p(x)$
- **Functions:**  $E[g(X)] = \sum_{\text{all } x} g(x) p(x)$
- **Indicators:**  $E[\delta_A] = P(A)$  where  $\delta_A$  is an indicator function
- **Linear function:**  $E[aX + b] = aE[X] + b$
- **Constants:**  $E[c] = c$  if  $c$  is constant

For two random variable's  $X$  and  $Y$

$$E[X + Y] = E[X] + E[Y]$$