

# STA 103 Lecture 3: Discrete Random Variables

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# Announcement

## Update on Assignment Due Dates

Hi everyone,

To give you time to review solutions before each exam, all assignments will now be due **one day earlier** than originally scheduled. This allows me to post solutions **24 hours before the exam**, so you can use them as an additional study resource.

To balance the shorter timeline, I've removed one problem from Assignment 1:

- **Before:** 6 problems in 7 days
- **Now:** 5 problems in 6 days (similar to having 12 days in a regular quarter as summer session moves twice as fast as a regular quarter)

The removed problem will **not** appear on the exam.

### Revised Due Dates:

- Assignment 1: **Aug 12 (Tue) Noon**
- Assignment 2: **Aug 25 (Mon) Noon**
- Assignment 3: **Sep 8 (Mon) Noon**

The syllabus and assignment have been updated accordingly, **please check the latest versions**. Thanks for your understanding, and I hope this helps you better prepare for the exams!

# Bernoulli Random Variables

- **Definition:** A Bernoulli random variable takes on only two values: 0 and 1, with probabilities  $1 - p$  and  $p$ , respectively.

- PMF in table form:

$x$	$p(x)$
0	$1 - p$
1	$p$

- PMF in Function Form:

$$p(x) = \begin{cases} p^x(1-p)^{1-x} & \text{if } x = 0 \text{ or } x = 1, \\ 0 & \text{otherwise.} \end{cases}$$

- The Bernoulli random variable can be denoted by  $X \sim \text{Bernoulli}(p)$ .

# Bernoulli Random Variables

For Bernoulli random variables:

- Expectation:

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

- Variance:

$$E(X^2) = \sum_x x^2 p(x) = 0^2 \cdot (1 - p) + 1^2 \cdot p = p,$$

$$\text{Var}(X) = E(X^2) - (E(X))^2 = p - p^2 = p(1 - p).$$

# Binominal Random Variables

- **Definition:** A binominal random variable  $X$  counts the number of successes in  $n$  independent Bernoulli trials.
- Each trial has only two outcomes: “success” with probability  $p$ , and “failure” with probability  $1 - p$ .
- The total number of trials is fixed and denoted by  $n$ .
- The random variable  $X$  can take values  $x = 0, 1, 2, \dots, n$ .

# Binominal Random Variables

- A binomial random variable is the sum of  $n$  independent and identically distributed (i.i.d.) Bernoulli random variables with the same success probability  $p$ :

$$Y = X_1 + X_2 + \cdots + X_n, \quad X_i \stackrel{\text{i.i.d.}}{\sim} \text{Bernoulli}(p).$$

- The PMF of Binomial RV, getting exactly  $x$  successes in  $n$  trials, is

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}.$$

- The term  $\binom{n}{x} = \frac{n!}{(n-x)!x!}$  counts the number of different ways to choose  $x$  successes out of  $n$  trials.
- The expression  $p^x (1 - p)^{n-x}$  is the probability of any particular sequence with  $x$  successes and  $n - x$  failures.

# Binominal Random Variables

- How to calculate  $E(X)$  and  $\text{Var}(X)$ ? Isn't it too complicated?
- Luckily here is the trick.
  - ▶ **Expectation:**

$$\begin{aligned} E(Y) &= E(X_1 + X_2 + \cdots + X_n) \\ &= E(X_1) + E(X_2) + \cdots + E(X_n). \end{aligned}$$

The summation expansion of expectation **always** holds.

- ▶ **Variance:**

$$\begin{aligned} \text{Var}(Y) &= \text{Var}(X_1 + X_2 + \cdots + X_n) \\ &= \text{Var}(X_1) + \text{Var}(X_2) + \cdots + \text{Var}(X_n). \end{aligned}$$

The summation expansion of variance holds **only if  $X_1, \dots, X_n$  are independent**. (We will discuss independence in the next lecture!)

# Binominal Random Variables

Then,

- **Expectation:**

$$\begin{aligned} E(Y) &= E(X_1 + X_2 + \cdots + X_n) \\ &= E(X_1) + E(X_2) + \cdots + E(X_n) \\ &= n \times E(X_1) = np. \end{aligned}$$

- **Variance:**

$$\begin{aligned} \text{Var}(Y) &= \text{Var}(X_1 + X_2 + \cdots + X_n) \\ &= \text{Var}(X_1) + \text{Var}(X_2) + \cdots + \text{Var}(X_n) \\ &= n \times \text{Var}(X_1) = np(1 - p). \end{aligned}$$

- We can denote it by  $Y \sim \text{Binomial}(n, p)$ . The Bernoulli( $p$ ) is a special case of Binomial( $n, p$ ) when  $n = 1$ .



# Binominal Random Variables

- **Example** (Market Research Analyst): You are analyzing customer response behavior in an online survey conducted by a retail economics team. From past data, it is known that the probability a randomly selected customer responds to the survey is  $p = 0.2$ . You randomly contact  $n = 50$  customers. Let the random variable  $X$  denote the number of customers who respond.
- **Question 1:** Identify the distribution of  $X$ . Justify your answer.

# Binominal Random Variables

- **Question 2:** What is the expected number of responses and the standard deviation?
- **Question 3:** What is the probability that exactly 10 customers respond?
- **Question 4:** What is the probability that at most 2 customers respond?

# Poisson Random Variable

- **Definition:** A Poisson random variable models the number of occurrences of an event in a fixed interval of time or space, under the assumptions that:
  - ▶ Events occur **independently**.
  - ▶ The average rate (events per interval) is **constant**.
  - ▶ Two events **cannot** occur at exactly the same time.
- Let  $X$  denote the number of events in a given interval and let  $\lambda > 0$  be the expected number of events in that interval. Then  $X$  is said to follow a Poisson distribution with parameter  $\lambda$ , and its probability mass function is

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}, \text{ for } x = 0, 1, 2, \dots$$

# Poisson Random Variable

- We can denote it by  $X \sim \text{Poisson}(\lambda)$ .
- **Expectation and Variance:**

$$E(X) = \text{Var}(X) = \lambda.$$

# Poisson Random Variable

**Identifying a Poisson Random Variable:** A random variable  $X$  can be modeled using a Poisson distribution if it satisfies the following conditions

- Counts of Events:  $X$  represents the number of times an event occurs in a fixed interval of time or space.
- Non-negative Integer Values:  $X = 0, 1, 2, \dots$
- Constant Rate: Events occur at a constant average rate  $\lambda$  over time or space.
- Rare Events: In a very small interval, the probability of more than one event occurring is negligible.

# Poisson Random Variable

Examples that likely follow a Poisson distribution:

- Number of customer arrivals at a bank per hour.
- Number of website clicks or page views per second.
- Number of insurance claims filed by a policyholder per year.

# Poisson Random Variable

- **Example:** A local bank observes that, on average, 6 customers arrive at the counter per hour. Assume customer arrivals follow a Poisson process. Let  $X$  be the number of customer arrivals in each hour. Then  $X \sim \text{Poisson}(\lambda = 6)$ .
- **Question 1:** What is the probability that exactly 4 customers arrive in one hour?
- **Question 2:** What is the probability that at most 2 customers arrive?

# Poisson Random Variable

- **Example:** A local bank observes that, on average, 6 customers arrive at the counter per hour. Assume customer arrivals follow a Poisson process. Let  $X$  be the number of customer arrivals in each hour. Then  $X \sim \text{Poisson}(\lambda = 6)$ .
- **Question 3:** What is the probability that more than 1 customers arrive?
- **Question 4:** What are the mean, standard deviation and variance of  $X$ ?



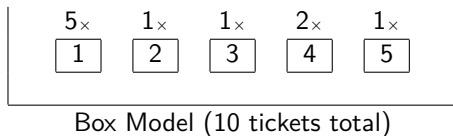
# Conditional PMF

Suppose  $X$  is a random variable with PMF:

- PMF in a table form:

$x$	$p(x)$
1	0.5
2	0.1
3	0.1
4	0.2
5	0.1
	1

- Equivalently, we have the following box model:



# Conditional PMF

- What if I sampled  $X$  but did not tell you which number it was but instead told you that  $X \geq 3$ .
- Then, the actual value of  $X$  is still random but now the possible values are 3, 4, 5 instead of 1, 2, 3, 4, 5.
- The conditional PMF given  $X \geq 3$ , denoted  $p(x \mid X \geq 3)$ , characterizes the randomness in  $X$  after observing  $X \geq 3$ .
- To find  $p(x \mid X \geq 3)$ , just zero out  $p(x)$  for  $x$ 's that are not observed and renormalize to 1.

# Conditional PMF

- The original PMF in a table form:

$x$	$p(x)$
1	0.5
2	0.1
3	0.1
4	0.2
5	0.1
	1

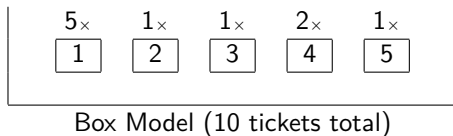
- After observing  $X \geq 3$ ,

$x$	$p(x \mid X \geq 3)$
1	0
2	0
3	0.1/0.4
4	0.2/0.4
5	0.1/0.4
	$(0.1 + 0.2 + 0.1)/0.4$

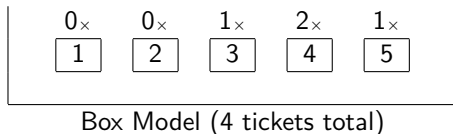
# Conditional PMF

Equivalently,

- The original box model:



- After observing  $X \geq 3$ ,



# Conditional PMF

- **Question 1:** Find  $P(X \geq 4)$ ?

$x$	$p(x)$
1	0.5
2	0.1
3	0.1
4	0.2
5	0.1
	1

- **Question 2:** Find  $P(X \geq 4 \mid X \geq 3)$ ?

$x$	$p(x \mid X \geq 3)$
1	0
2	0
3	$0.1/0.4$
4	$0.2/0.4$
5	$0.1/0.4$
	$(0.1 + 0.2 + 0.1)/0.4$

## Conditional PMF

- We can also compute the “new” expected value  $E(X)$  after observing  $X \geq 3$ .

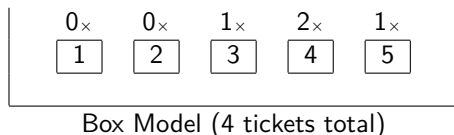
$x$	$p(x \mid X \geq 3)$
1	0
2	0
3	$0.1/0.4$
4	$0.2/0.4$
5	$0.1/0.4$
$(0.1 + 0.2 + 0.1)/0.4$	

$$\begin{aligned} E(X \mid X \geq 3) &= \sum_x xp(x \mid X \geq 3) \\ &= 3 \cdot \frac{1}{4} + 4 \cdot \frac{1}{2} + 5 \cdot \frac{1}{4} = \frac{3 + 8 + 5}{4} = 4. \end{aligned}$$

- **Interpretation** of  $E(X \mid X \geq 3)$ : Adjusted prediction after given new information that  $X \geq 3$ .

# Conditional PMF

- Box average after observing  $X \geq 3$ .



$$\text{Box Average} = \frac{3 + 4 + 4 + 5}{4} = 4.$$

- Interpretation:** Box Average after observing  $X \geq 3$  is identical to  $E(X \mid X \geq 3)$ .

## Conditional PMF

- Similarly, the  $\text{Var}(X)$ , and  $\text{sd}(X)$  after observing  $X \geq 3$  are:

$x^2$	$x$	$p(x \mid X \geq 3)$
$1^2$	1	0
$2^2$	2	0
$3^2$	3	$0.1/0.4 = 1/4$
$4^2$	4	$0.2/0.4 = 1/2$
$5^2$	5	$0.1/0.4 = 1/4$
		$(0.1 + 0.2 + 0.1)/0.4$

$$\begin{aligned}\text{Var}(X \mid X \geq 3) &= E(X^2 \mid X \geq 3) - (E(X \mid X \geq 3))^2 \\ &= 3^2 \cdot \frac{1}{4} + 4^2 \cdot \frac{1}{2} + 5^2 \cdot \frac{1}{4} - (4)^2 \\ &= \frac{9 + 32 + 25}{4} - 16 = 1/2.\end{aligned}$$

$$\text{sd}(X \mid X \geq 3) = \sqrt{\text{Var}(X \mid X \geq 3)} = \frac{1}{\sqrt{2}} = \frac{\sqrt{2}}{2} \approx 0.7071.$$