

# Chapter 5: Discrete Probability Distribution

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# Review

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- Random variable
- Discrete: pmf, CDF
- Continuous: pdf, CDF
- Mean
- Variance

# Learning Objectives

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- Binomial distribution
- Hypergeometric distribution
- Poisson distribution and the Poisson process

# Review of Discrete Random Variables

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- A **discrete** random variable has possible values that form a finite set or can be listed in an infinite sequence ("countable").
- The probability distribution of a random variable  $X$  describes how total probability 1 is distributed among outcomes for  $X$ .

$$\sum_x P(X=x) = 1$$

# The Bernoulli Process

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- The experiment consists of  $n$  repeated trials.
- Each trial results in success or failure.
- The probability of success,  $p$ , is constant from trial to trial.
- The repeated trials are independent.

# Bernoulli Random Variable

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Let  $X$  indicate the outcome of a single trial:

$$X = \begin{cases} 1, & \text{success} \\ 0, & \text{failure} \end{cases} \quad \text{denoted } X \sim \text{Bernoulli}(p).$$

The pmf is

$$f(a) = P(X = a) = \begin{cases} p, & a = 1, \\ 1 - p, & a = 0, \end{cases} \quad 0 \leq p \leq 1.$$

# Motivating Binomial Examples

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Consider:

- Flip a coin 10 times;  $X$  = number of heads.
- A machine produces 1% defective parts;  $X$  = number defective in next 25.
- In the next 20 births;  $X$  = number of female births.

# Binomial Experiment Conditions

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An experiment is **binomial** if:

1. It consists of  $n$  trials (fixed in advance).
2. Trials are identical; each trial is success (S) or failure (F).
3. Trials are independent.
4. Probability of success is constant:  $p$ .

## Example (Setup)

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Assume the probability that a stolen car will be recovered is 60%. If 3 cars are stolen, what is the probability that exactly 1 of them will be recovered?

## Example (Solution)

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Let  $X$  = number of recovered cars. Then  $X \sim \text{Bin}(n = 3, p = 0.6)$ .

$$\begin{aligned} P(X=1) &= \binom{3}{1} 0.6 \times (0.4)^2 \\ &= .288 \end{aligned}$$

# Binomial Distribution

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If  $X \sim \text{Bin}(n, p)$ , then

$$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}, \quad x = 0, 1, \dots, n.$$

## Example (Binomial Probabilities)

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If  $P(\text{pass}) = 0.8$  and  $n = 10$  students:

(a)  $P(\text{all 10 pass})$

$$P(X=10) = \binom{10}{10} 0.8^{10} 0.2^0 = 0.8^{10}$$

(b)  $P(\text{exactly 3 pass})$

$$P(X=3) = \binom{10}{3} 0.8^3 0.2^7$$

(c)  $P(\text{at least 8 pass})$

$$\begin{aligned} P(X \geq 8) &= P(X=8) + P(X=9) + P(X=10) \\ &\sum_{x=8}^{10} \binom{10}{x} 0.8^x 0.2^{10-x} \end{aligned}$$

## CDF for $X \sim \text{Bin}(n, p)$

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For an integer  $a$  with  $0 \leq a \leq n$ ,

$$P(X \leq a) = \sum_{x=0}^a P(X = x).$$

# Mean and Variance for Binomial

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For  $X \sim \text{Bin}(n, p)$ ,

$$\mu = E[X] = np, \quad \sigma^2 = \text{Var}(X) = np(1 - p) = npq, \quad q = 1 - p.$$

## Example (Credit Card Purchases)

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If 70% of purchases are made with a credit card and  $X$  is the number among 10 purchases made with a credit card:

- (a) What is the distribution of  $X$ ?
- (b) Find  $E[X]$ .
- (c) Find  $\text{Var}(X)$ .
- (d) Find  $P(|X - \mu| \leq \sigma)$ .

$$X \sim \text{Bin}(10, 0.7)$$

$$E(X) = 7$$

$$\sigma^2 = npq = 10(0.7)(0.3) = 2.1$$

$$\sigma = \sqrt{2.1} = 1.4$$

$$P(X - 7 \leq 1.4 \text{ or } 7 - X \leq 1.4) = P(6 \leq X \leq 8) = \sum_{x=6}^{10} \binom{10}{x} 0.7^x 0.3^{10-x}$$

# Binomial Formula Examples

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1. A die is rolled 10 times. What is the chance of getting exactly 2 aces (1's)?
2. A die is rolled until it first lands six. If this can be done using the binomial formula, find the chance of getting 2 aces. If not, why not?
3. Ten draws are made with replacement from a box:

[1] [1] [1] [1] [1] [0]

Before the last draw, one ticket labeled 1 is removed. True or False:

$$P(\text{exactly 2 ones}) = \binom{10}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^8$$

4. Four draws are made **without replacement** from the same box. True or False:

$$P(\text{exactly 2 ones}) = \binom{4}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^2$$

# Binomial Formula: Example 1

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## Example 1.

A die is rolled 10 times. What is the chance of getting exactly 2 aces (1's)?

$$P(X=2) = \binom{10}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^8$$

## Binomial Formula: Example 2

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**Example 2.** A die is rolled until it first lands six. If this can be done using the binomial formula, find the chance of getting 2 aces. If not, why not?

no,  $n$  is not fixed

# Binomial Formula: Example 3

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## Example 3.

Ten draws are made with replacement from a box:

[1] [1] [1] [1] [1] [0]

Before the last draw, one ticket labeled 1 is removed.

True or False:

$$P(\text{exactly 2 ones}) = \binom{10}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^8$$

# Binomial Formula: Example 4

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## Example 4.

Four draws are made **without replacement** from the same box.

True or False:

$$P(\text{exactly 2 ones}) = \binom{4}{2} \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^2$$

*False*

# When Can We Use the Binomial Formula?

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We can use the binomial model when:

1. Number of trials is fixed
2. Probability is constant
3. Trials are independent
4. Only two outcomes

# Hypergeometric Distribution (When?)

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Hypergeometric assumptions:

- Population size  $N$  is finite.
- Each item is success (S) or failure (F); there are  $k$  successes.
- Sample  $n$  items **without replacement** (all subsets equally likely).

# Hypergeometric pmf

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If  $X \sim \text{Hypergeom}(N, n, k)$ , then

$$P(X = x) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}},$$

for

$$\max\{0, n - (N - k)\} \leq x \leq \min\{n, k\}.$$

## Example (Lot Acceptance)

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You ship a lot of  $N = 24$  cameras. An inspector samples  $n = 4$  and rejects the lot if any defective is found. You know  $k = 3$  are defective. Find  $P(\text{lot accepted})$ .

$$P(\text{accepted}) = P(X=0) = \frac{\binom{3}{0} \binom{21}{4}}{\binom{24}{4}}$$

# Mean and Variance (Hypergeometric)

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For  $X \sim \text{Hypergeom}(N, n, k)$ ,

$$\mu = E[X] = n \frac{k}{N},$$

$$\sigma^2 = \text{Var}(X) = \frac{N-n}{N-1} n \frac{k}{N} \left(1 - \frac{k}{N}\right).$$

What is the mean and variance of the previous example?

$$E[X] = n \frac{k}{N} = 4 \frac{3}{24} = \frac{1}{2}$$

$$\text{Var}(X) = \frac{N-n}{N-1} n \frac{k}{N} \left(1 - \frac{k}{N}\right) = \frac{24-4}{24-1} \times 4 \frac{3}{24} \left(1 - \frac{3}{24}\right)$$
$$= \frac{35}{92}$$

# Hypergeometric vs Binomial: Example

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A box contains 4 red balls and 6 blue balls.

- (a) Four balls are drawn **with replacement**.
- (b) Four balls are drawn **without replacement**.

What is the probability of drawing exactly 2 red balls?

$$(a) P(X=2) = \binom{4}{2} (.4)^2 (.6)^2$$

$$(b) P(X=2) = \frac{\binom{4}{2} \binom{6}{2}}{\binom{10}{4}}$$

# Example Solution

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# Poisson Experiments (Counts)

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A Poisson experiment records the number of events in a time interval or a region of space. Examples:

- Calls received per hour
- Days school is closed
- Accidents in a week
- Typing errors on a page

# Poisson Process Properties

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- Counts in disjoint intervals/regions are independent.
- Probability of one event in a very short interval is proportional to interval length/region size.
- Probability of more than one event in a very short interval is negligible.

# Poisson Distribution

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If events occur at average rate  $\lambda$  per unit time (or space), then the number of events in interval length  $t$  is

$$X \sim \text{Poisson}(\lambda t),$$

with pmf

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

## Example (Poisson)

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Let  $X$  be the number of creatures captured in a week. Suppose that  $X$  has a Poisson distribution with  $\lambda = 4.5$ , so on average traps will contain 4.5 creatures. Suppose  $X \sim \text{Poisson}(\lambda t = 4.5)$ .

- (a)  $P(X = 5)$
- (b)  $P(X \leq 5)$
- (c)  $P(X \geq 5)$

$$P(X=5) = e^{-4.5} \frac{(4.5)^5}{5!}$$

$$P(X \leq 5) = \sum_{x=0}^5 e^{-4.5} \frac{(4.5)^x}{x!}$$

$$P(X \geq 5) = 1 - \cancel{P(X \leq 4)}$$

# Mean and Variance (Poisson)

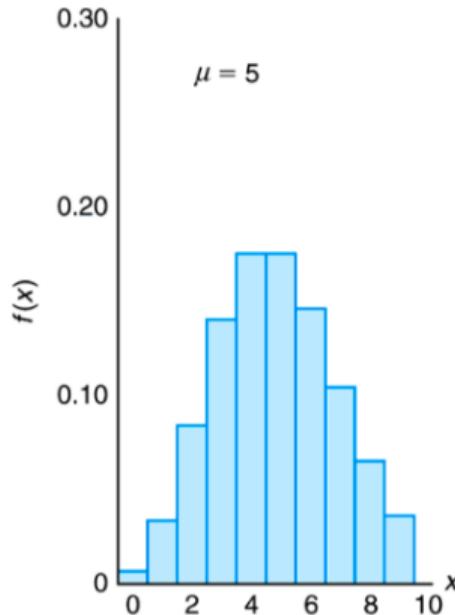
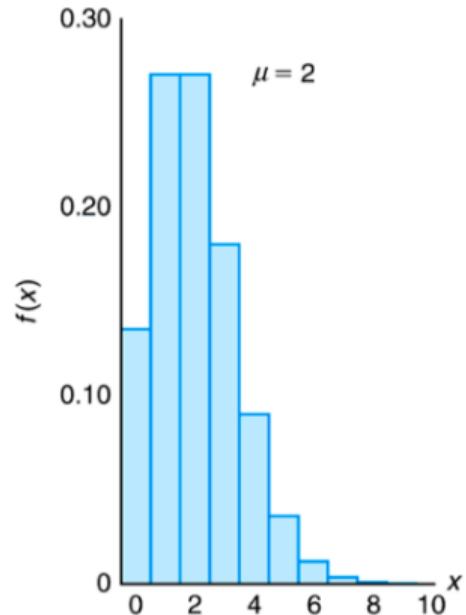
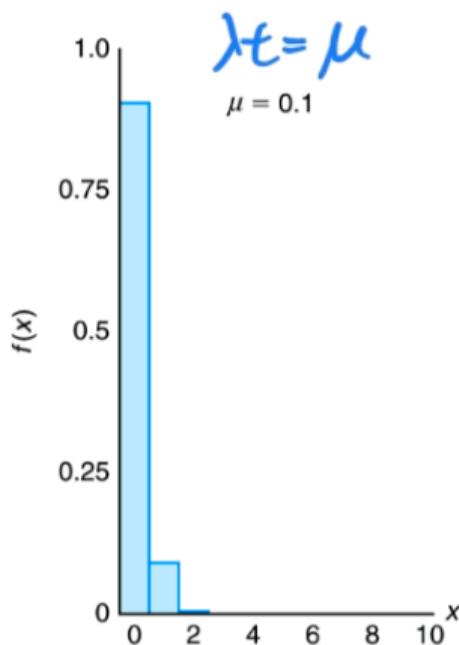
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If  $X \sim \text{Poisson}(\lambda t)$ , then

$$\mu = E[X] = \lambda t, \quad \sigma^2 = \text{Var}(X) = \lambda t.$$

# Nature of the Poisson Probability function

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## Example (Arrivals at a Counter)

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Suppose pulses arrive at a counter at an average rate of 6 per minute, so  $\lambda = 6$ .

- (a) Find the probability that in a 0.5-minute interval at least one pulse is received.
- (b) Find the mean number of pulses received in the 30-second interval.

# Poisson Example

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A factory produces screws, and 0.1% are defective.

A box contains 2000 screws.

What is the probability the box has exactly 3 defective screws?

# Poisson Approximation: Example 3 (Solution)

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## Poisson Example

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A store receives customers in bursts: when one arrives, their friends often arrive immediately after.

Is Poisson a good model for the number of arrivals per minute? Why or why not?

# Chapter Summary: Discrete Distributions

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- **Bernoulli Process:** Repeated independent trials with constant success probability  $p$ .
- **Binomial Distribution:**

$$X \sim \text{Bin}(n, p), \quad P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}$$

$$E[X] = np, \quad \text{Var}(X) = np(1 - p)$$

- **Hypergeometric Distribution:** Sampling **without replacement** from finite population.

$$X \sim \text{Hypergeom}(N, n, k)$$

- **Poisson Distribution:** Counts of events over time/space.

$$X \sim \text{Poisson}(\lambda t), \quad E[X] = \text{Var}(X) = \lambda t$$

- **Key Idea:** Choose the distribution based on independence, replacement, and event rate.