

# **STAT 234 Lecture 5**

## **Binomial Distributions**

### **Section 3.5**

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Lecture 5 covers Section 3.5 of MMSA

Please skip

- Moment General Function for Binomial on p.135 in Section 3.5
- Section 3.4 Moment Generating Functions
- Section 3.6 Hypergeometric and Negative Binomial Distributions
- Section 3.7 Poisson Distributions

# The Binomial Distribution

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What is the probability that the first two draws are Red and the next 3 are Green?

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$$\begin{aligned}P(R R G G G) &= P(R) \cdot P(R) \cdot P(G) \cdot P(G) \cdot P(G) \\ &= 0.1 \times 0.1 \times 0.9 \times 0.9 \times 0.9.\end{aligned}$$

As the draws are made with replacement, the outcomes of the 5 draws are **independent**. The multiplication rule for independent events can be applied.

What is the probability of getting **exactly two Reds in 5 draws**?  
Is it also equal to

$$0.1 \times 0.1 \times 0.9 \times 0.9 \times 0.9?$$

There are 10 possible orderings of the 2 Reds and the 3 Greens.

Possible Orders	Probability
RRGGG	$0.1 \times 0.1 \times 0.9 \times 0.9 \times 0.9 = (0.1)^2(0.9)^3$
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RGGRG	$0.1 \times 0.9 \times 0.9 \times 0.1 \times 0.9 = (0.1)^2(0.9)^3$
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GGGRR	$0.9 \times 0.9 \times 0.9 \times 0.1 \times 0.1 = (0.1)^2(0.9)^3$

P(exactly 2 Reds in 5 draws) is the sum of the probabilities of the 10 cases above because the 10 cases are *disjoint*. So

$$P(\text{exactly 2 Reds in 5 draws}) = 10 \times (0.1)^2(0.9)^3.$$

What is  $P(\text{getting exactly } k \text{ Reds in } n \text{ draws})$ ?



## What is P(getting exactly $k$ Reds in $n$ draws)?

Consider all possible ways to order the  $k$  Reds and the  $n - k$  Greens.

Possible Orders	Probability
$\underbrace{RRR \dots R}_k \underbrace{G \dots G}_{n-k}$	$\underbrace{0.1 \times \dots \times 0.1}_k \times \underbrace{0.9 \times \dots \times 0.9}_{n-k} = (0.1)^k (0.9)^{n-k}$
RGR...RG...G	$0.1 \cdot 0.9 \cdot 0.1 \dots 0.1 \times 0.9 \dots 0.9 = (0.1)^k (0.9)^{n-k}$
$\vdots$	$\vdots$

Note

- the events for different orderings are *disjoint*, and
- each occurs with identical probability  $(0.1)^k (0.9)^{n-k}$ .

By the Addition Rule, P(exactly  $k$  Reds in  $n$  draws) equals

$$(\# \text{ of ways to order } k \text{ Reds and } n - k \text{ Greens}) \times (0.1)^k (0.9)^{n-k}$$

# Factorial

The notation  $n!$ , read *n factorial*, is defined as

$$n! = 1 \times 2 \times 3 \times \dots \times (n - 1) \times n$$

e.g.,

$$1! = 1,$$

$$3! = 1 \times 2 \times 3 = 6,$$

$$2! = 1 \times 2 = 2,$$

$$4! = 1 \times 2 \times 3 \times 4 = 24.$$

By convention,

$$0! = 1.$$

## Binomial Coefficients

The number of ways to order  $k$  Reds and  $n - k$  Greens equals

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

- $\binom{n}{k}$  is read as “ $n$  choose  $k$ ”, also denoted as  ${}_n C_k$ , or  $C_k^n$ .

e.g.,

$$\binom{5}{2} = \frac{5!}{2! \times (5-2)!} = \frac{5 \times 4 \times 3 \times 2 \times 1}{(2 \times 1)(3 \times 2 \times 1)} = \frac{5 \times 4}{2 \times 1} = 10,$$

$$\binom{n}{n} = \frac{n!}{n! \times 0!} = \frac{n!}{n! \times 1} = 1$$

You can also use R for these calculations:

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choose(5, 2)
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 $\Rightarrow$  there are  $n$  ways to order 1 Red and  $n - 1$  Greens
- $\binom{n}{n-1} = \frac{n!}{(n-1)! \times 1!} = n$   
 $\Rightarrow$  there are  $n$  ways to order  $n - 1$  Reds and 1 Green

## Summary of All the Calculations So Far

When  $n$  draws are made at random with replacement from a box contains one red ball and 9 green ones,



the probability to get exactly  $k$  Reds (and  $n - k$  Greens) equals

$$\binom{n}{k}(0.1)^k(0.9)^{n-k}.$$

Such calculations can be generalized to other similar problems and the general formula is called the *Binomial Formula*.

A random trial having only 2 possible outcomes (Success, Failure) is called a *Bernoulli trial*, e.g.,

- whether a coin lands heads or tails when tossing a coin
- whether one gets a six or not a six when rolling a die
- whether a drug works on a patient or not
- whether a electronic device is defected
- whether a subject answers Yes or No to a survey question

# Binomial Formula

Suppose  $n$  **independent** Bernoulli trials are to be performed, each of which results in

- a *success* with probability  $p$  and
- a *failure* with probability  $1 - p$ .

The probability of getting  $k$  successes and  $n - k$  failures in  $n$  Bernoulli trials is given by

$$\begin{aligned} & (\# \text{ of ways to order the } k \text{ successes and } n - k \text{ failures}) \times p^k(1 - p)^{n-k} \\ &= \binom{n}{k} p^k(1 - p)^{n-k} \end{aligned}$$

# Binomial Distribution

Suppose  $n$  **independent** Bernoulli trials are to be performed, each of which results in

- a *success* with probability  $p$  and
- a *failure* with probability  $1 - p$ .

If we define

$X =$  the number of successes that occur in the  $n$  trials,

then  $X$  is said to have a *binomial distribution* with parameters  $(n, p)$ , denoted as

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

with the probability mass function (pmf)

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

## Does the Binomial Probabilities Add Up to 1?

Recall the *Binomial expansion* in math:

$$(a + b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

which is valid for all real numbers  $a$  and  $b$ .

Applying the Binomial expansion with  $a = p$  and  $b = 1 - p$ , we get

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

which means that the pmf

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

is a valid pmf.

## Conditions Required to be Binomial

Conditions required to apply the binomial formula:

1. each trial outcome must be classified as a *success* or a *failure*
2. the probability of success,  $p$ , must be the same for each trial
3. the number of trials,  $n$ , must be fixed
4. the trials must be independent

## Binomial or Not? — 10 Rolls of a Die

Rolling a die 10 times, what is the probability of getting exactly 3 aces?



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- number of trials  $n = 10$

## Binomial or Not? — 10 Rolls of a Die

Rolling a die 10 times, what is the probability of getting exactly 3 aces?

- a trial: whether one gets an ace when rolling a die once
- prob. of success  $p = 1/6$
- number of trials  $n = 10$
- the trials (rolls) are independent

So, it's okay to use the Binomial formula.

$$\begin{aligned}P(3 \text{ aces in } 10 \text{ rolls}) &= \frac{10!}{3!7!} \left(\frac{1}{6}\right)^3 \left(1 - \frac{1}{6}\right)^7 \\&= \frac{10 \times 9 \times 8 \times (7!)}{(3 \times 2 \times 1)(7!)} \left(\frac{1}{6}\right)^3 \left(\frac{5}{6}\right)^7 \\&= 120 \left(\frac{1}{6}\right)^3 \left(\frac{5}{6}\right)^7 \approx 0.155\end{aligned}$$

## Binomial or Not? — Rolling a Die Until 3rd Ace

Rolling a die continuously, is the probability of getting the 3rd aces in the 10th roll equals to

$$\frac{10!}{3!7!} \left(\frac{1}{6}\right)^3 \left(1 - \frac{1}{6}\right)^7 ?$$

## Binomial or Not? — Rolling a Die Until 3rd Ace

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No. The number of trials (sample size) is not determined in advance.

## Binomial or Not? — Restaurants

Suppose an inspector randomly selects **5** restaurants from the 20 restaurants in a town, of which 10 currently have health code violation(s) and the other 10 have no violations. Let  $X$  be the number of selected restaurants with violations. Is  $X$  binomial?

- a trial: whether a randomly selected restaurant has violation(s)

$S$  = violation,     $F$  = no violation

- number of trials:  $n = 5$

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- prob. of success?

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- $P(\text{S on 5th trial} \mid \text{FFFF}) = \frac{10}{16} = 0.625$
- $P(\text{S on 5th trial} \mid \text{SSSS}) = \frac{6}{16} = 0.375$
- Trials are NOT independent since the selection are made **without replacement.**

## Binomial or Not — UC Undergrads

50 UC undergrads are randomly selected and each of them is asked whether he/she has a driver's licence. Suppose 4000 of the 5000 UC undergrads have driver's licence. Let  $X$  be the number who reply yes. Is  $X$  binomial?

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- a trial: a randomly selected student reply yes = S or no = F
- number of trials  $n = 50$
- Strictly speaking, NOT binomial, since selection are made without replacement — trials are dependent. However,...

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- Since the sample size 50 is only 1% of the population size (5000), the 50 draws has little effect on the makeup of the population.  $P(S)$  stays close to 0.8 regardless of the outcome of prior draws. Trials are nearly independent
- So  $X$  is approx. binomial,  $\text{Bin}(n = 50, p = 0.8)$ .



## Rule: Approx. Binomial or Not When Sampling w/o Replacement

Consider sampling without replacement from a dichotomous population of size  $N$ . If the sample size (number of trials)  $n$  is at most 5% of the population size, the experiment can be analyzed as though it were exactly a binomial experiment.