# The Binomial and Geometric Distributions

Chapter 7.5

### 7.5 The Binomial Distribution

- A **binomial experiment** is <u>statistical experiment</u> that has the following properties:
  - The experiment consists of *n* repeated trials.
  - Each trial can result in just two possible outcomes. We call one of these outcomes a **success** and the other, a **failure**.
  - The probability of success, denoted by **P**, is the same on every trial.
  - The trials are <u>independent</u>; that is, the outcome on one trial does not affect the outcome on other trials.
  - \*NOTE: For discrete random variables only

## Use the B.I.N.S. acronym:

- **B**: Binary? Outcomes classified as "success" or "failure"
- I: Independent? Knowing the result of one trial must not have any effect on the result of another
- N:Number? Number of trials n of the chances process must be fixed
- S: Success? The probability p of success must be the same on each trial Acronym: "BINS"

### Example

- Consider the following statistical experiment. You flip a coin 2 times and count the number of times the coin lands on heads. This is a binomial experiment because:
  - The experiment consists of repeated trials. We flip a coin 2 times.
  - Each trial can result in just two possible outcomes - heads or tails.
  - The probability of success is constant 0.5 on every trial.
  - The trials are independent; that is, getting heads on one trial does not affect whether we get heads on other trials.

#### Notation

- x: The number of successes that result from the binomial experiment.
- n: The number of trials in the binomial experiment.
- P: The probability of success on an individual trial.
- q: The probability of failure on an individual trial. (This is equal to 1 P.)
- b(x; n, P): Binomial probability the probability that an n-trial binomial experiment results in exactly x successes, when the probability of success on an individual trial is P.

#### Binomial or not?

- Tossing 20 coins and counting the number of heads.
  - Yes-1. Success is a heads, failure is a tails. 2. n = 20.
    3. Independence is true coins have no influence on each other. 4. p = .5. So X is B(20, .5). The possible values of X are the integers from 0 to 20.
- Picking 5 cards from a standard deck and counting the number of cards that are hearts. We replace the card each time and reshuffle.
  - Yes-Success is a heart, failure is anything but a heart.
    2. n = 5. 3. Independence is true. 4. p = .25. So X is B(5, .25). The possible values of X are the integers from 0 to 5.

- Picking 5 cards from a standard deck and counting the number of hearts without replacing after each pick.
  - No, b/c of independence issue
- Choosing a card from a standard deck until you get a heart.
  - No, b/c there are not a fixed number of observations

#### Binomial or not?

- It is estimated that 87% of computers users use Chrome as their default web browser. We choose 15 computer users and ask their default browser.
  - B: Binary? Success is Chrome, failure is anything else.
  - I: Independent? Independence seems logical.
  - **N**:Number fixed? n = 15.
  - **S**: Success? p = .87.
  - So X is B(15, .87). The possible values of X are the integers from 0 to 15.

### What is the probability?

- Taking the previous binomial setting, what is the probability that 10 of the computers will use chrome as their default?
  - **N**:Number of fixed trials: n = 15.
  - **S**: Success of each trail is p = .87.

• 
$$P(x = 10) = {15 \choose 10} \cdot (.87)^{10} (.13)^5 = .0276990$$

15  $C_{10}$ 

# Is it Binomial...? Use the B.I.N.S. acronym:

- **B**: Binary? Outcomes classified as "success" or "failure"
- I: Independent? Knowing the result of one trial must not have any effect on the result of another
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# Complete the Intro to Binomial Distributions WS

- Consider three aspects:
  - 1. Number of combinations,
  - 2. Probability of success, # of successes
  - 3. Probability of failure, # of failures
- $\circ$   ${}_{n}C_{k} \times (probability \ of \ success)^{k \ times} \times (probability \ of \ failure)^{n-k \ times}$
- $\circ$   ${}_{n}C_{k} \times p^{k} \times (1-p)^{n-k}$
- Plug and CHUG!

# Mean and SD of a Binomial Random Variable

- ullet Formulas: mean of the binomial random variable  $\mu_{\chi}=np$
- S.D. of the binomial random variable  $\sigma_{\chi} = \sqrt{np(1-p)}$

## Large samples

\*This concept holds true in real world experiments and expected values!

#### Sampling Distribution of a Count

Choose an SRS of size n from a population with proportion p of successes. When the population is much larger than the sample, the count X of successes in the sample has approximately the binomial distribution with parameters n and p.

**Example 1**: A University of 10,000 students has 1,000 scholarship students. We choose 8 students and count the number of scholarship students.

• Success is a scholarship student. 2. n = 8 3. It could be argued we don't have independence, as choosing the first student as a scholarship student changes the probability of the second being a scholarship student. But the probabilities change so little that we still consider this an independent situation. 4. p = .1. So X is B(8, .1). Possible values of X are integers from 0 to 8.

- Example 2: An engineer chooses a SRS of 10 switches from a shipment of 10,000 switches. Suppose that (unknown to the engineer) 10% of the switches in the shipment are bad. The engineer counts the number X of bad switches in the sample.
  - This is not quite a binomial setting-just as removing one card in changes the makeup of the deck, removing one switch changes the proportion of bad switches remaining in the shipment. So the state of the second switch chosen is not independent of the first. BUT removing one switch from a shipment of 10,000 changes the makeup of the remaining 9999 switches very little. In practice, the distribution of X is very close to the binomial distribution with n = 10 and p = .1

#### Binomial Distribution

- As discussed, binomial random variable is the number of successes x in n repeated trials of a binomial experiment.
- The <u>probability distribution</u> of a binomial random variable is called a **binomial** distribution
- Suppose we flip a coin two times and count the number of heads (successes). The binomial random variable is the number of heads, which can take on values of 0, 1, or 2.

Mean and SD taken the same way as random variable

Number of heads	Probability			
0	0.25			
1	0.50			
2	0.25			

#### Warm-UP

1) The fastidious restaurant manager, and dilettante statistician defines a continuous random variable *X* by:

X = the weight of the crumbs in ounces left on the floor of a restaurant during a one hour period, for the interval from 2 to 6 ounces. If this is a uniform distribution,

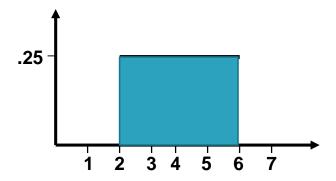
- (a) what will the height of the density function be? (interval)x(height) = 1
- (b) Make a sketch of the density function.

Define a continuous random variable X by

X = the weight of the crumbs in ounces left on the floor of a restaurant during a one hour period, for the interval from 2 to 6 ounces. If this is a uniform distribution

Suppose that x has a probability distribution with density function

$$f(x) = \begin{cases} .25 & 2 < x < 6 \\ 0 & \text{otherwise} \end{cases}$$



## Binomial Probability

- The **binomial probability** refers to the probability that a binomial experiment results in <u>exactly</u> *X* successes. For example, in the previous table, we see that the binomial probability of getting exactly one head in two coin flips is 0.50.
- Given x, n, and P, we can compute the binomial probability based on the following formula (ick!):

**Binomial Formula.** Suppose a binomial experiment consists of *n* trials and results in *x* successes. If the probability of success on an individual trial is *P*, then the binomial probability is:

$$b(x; n, P) = {}_{n}C_{x} * P^{x} * (1 - P)^{n - x}$$

#### Calculator!

• A **Binomial PDF** (Probability Density function) allows you to find the probability that *X* is any value in a binomial distribution. It is found in the Distribution Menu:

o 2nd VARS A: binompdf(... 5



- Its form is: Binompdf(n, p, X). (There are 3 important variables: n is the number of observations, p is the probability of success, and X is the number of successes you want.
- If you don't specify X, it will give you the probability for all values of X, from 0 to n as a list.

#### Calculator

### trials:N p:P x value:X**=**

binompdf

trials:10

⇒:.2

x value:5

Paste

Older software of TI-83 simply shows:

$$_{n}C_{k} \times p^{k} \times (1-p)^{n-k}$$

$$_{5}C_{1} \times (1)^{.2} \times (4)^{.8}$$

### Examples

1. We want to compare the probability of getting 3 heads from 5 tosses of a coin with 4 heads on 5 tosses.

- 2. Bob takes a true-false test of 6 questions and has absolutely no idea of any of the answers so he guesses on all of them. If 4 questions correct is passing, what is the probability that he passes the exam?
  - this is a binomial distribution with n=6 and p=.5 and we need to add the probabilities of getting 4, 5, or 6

$$P(X = 4) = {6 \choose 4}.5^4 (1 - .5)^{6-4} = \frac{6!}{4!(2)!}.5^4 (.5)^2 = .234$$

$$P(X = 5) = {6 \choose 5}.5^5 (1 - .5)^{6-5} = \frac{6!}{5!(1)!}.5^5 (.5)^1 = .094$$

$$P(X = 6) = {6 \choose 6}.5^6 (1 - .5)^{6-6} = \frac{6!}{6!(0)!}.5^6 (.5)^9 = .016$$
The probability of passing is .234 + .094 + .016 = .344

### Examples continued

- Suppose the test is now multiple choice with 4 answers per problem and again, Bob guesses. Find the probability that he passes the test and the expected number of passing students in a school of 1,500 if they all guessed.
  - Again, this is a binomial distribution with n = 6 and p = .25.
- Using the formula that Expected Value (mean number of passing students) = np, we get that 3.8% of 1,500 students or about 57 of them would pass the test by sheer guessing.

```
binompdf(6,.25,4
)+binompdf(6,.25
,5)+binompdf(6,.
25,6)
.038
```

# Cumulative Binomial Probability

- A **cumulative binomial probability** refers to the probability that the binomial random variable falls within a specified range (e.g., is greater than or equal to a stated lower limit and less than or equal to a stated upper limit).
- Ex: In a particular city, 63% of the adults own their home and 37% rent. A sample of 20 adults is taken. Find the probability that the sample will have at least half homeowners.
  - This is binomial with n = 20 and p = .63. We want the value of X = 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20.
  - That is a lot of work, even with the Binompdf function!

### Calculator

- To solve it, we turn to the **Binomcdf** formula found in the same menu. This gives the **cumulative probabilities** starting at X = 0. For instance, Binomcdf(20,.63,3) would give P(X = 0) + P(X = 1) + P(X = 2) + P(X = 3).
  - In our case we can find the sum of the probabilities that X = 0, 1, 2, 3, 4, 5, 6, 7, 8, and 9 then subtract that from 1. That will give us the probability that X = 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, or 20.

### More Examples

- 1. What is the probability of obtaining 45 or fewer heads in 100 tosses of a coin?
  - The sum of all these probabilities is the answer we seek: b(x = 0; 100, 0.5) + b(x = 1; 100, 0.5) + . . . + b(x = 45; 100, 0.5)
  - $b(x \le 45; 100, 0.5) = 0.184$  \*try on calc!
- 2. The probability that a student is accepted to a prestigious college is 0.3. If 5 students from the same school apply, what is the probability that at most 2 are accepted?
  - $b(x \le 2; 5, 0.3) = 0.8369$  on calc: binomcdf(5, .3, 2)

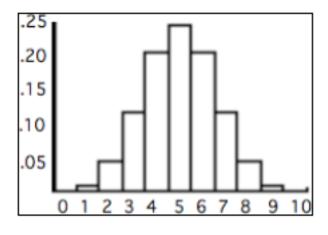
# Mean and SD of a Binomial Random Variable

- Formulas:  $\mu_{\chi} = np$  and  $\sigma_{\chi} = \sqrt{np(1-p)}$
- A basketball player is traditionally a 72% foul shooter. In a season, he takes 427 foul shots. Find the mean and standard deviation of the distribution.
- M = 307.44 SD = 9.278

# Probability Distribution Histogram

A coin is tossed 10 times, a head is a success.
 Construct a probability distribution histogram

X	0	1	2	3	4	5	6	7	8	9	10
P(X)	0.001	0.010	0.044	0.117	0.205	0.246	0.205	0.117	0.044	0.010	0.001



This distribution appears normal, but it's not, it's binomial- normal distributions are for continuous variables where there are an infinite number of outcomes. Binomial distributions are for discrete data where there is only a finite number of outcomes. However, as n gets larger, a binomial distribution starts to appear more and more normal and each one is a good approximation for the other.

# Binomial distribution histograms

- To do this on your calculator:
  - Enter the values of X in L1
  - Enter the binomial probabilities into list L2: Highlight L2 and press 2<sup>nd</sup> VARS(DISTR) – binompdf(n, p, L1) and hit enter. L2 should populate
  - Define Plot1 to be a histogram with Xlist: L1 and Freq: L2
  - Set your X and Y viewing window to cover all your values of X and the probabilities: X(0,10,1) Y(0,.4,.1)

# Histograms cont.

- To do a cumulative histogram on your calc, make L3 your cumulative probabilities by highlighting L3 and defining it as binomcdf(n, p, L1) ENTER and it will populate.
- Make a histogram with Xlist: L1 and Ylist: L3 and adjust viewing window.

# The Normal Approximation to Binomial Distributions- how large is large?

If we aren't using a calculator, the by-hand formula for binomial probabilities becomes awkward/annoying as the number of trials n increases so here is our alternative: When n is large, we can use Normal probability calculations to approximate hard-to-calculate binomial probabilities

#### Normal Approximation for Binomial Distributions

Suppose that a count *X* has the binomial distribution with *n* trials and success probability p. When n is large, the distribution of *X* is approximately Normal,  $N(np, \sqrt{np(1-p)})$ . As a rule of thumb, we will use the Normal approximation when n and p satisfy  $np \ge 10$  and  $n(1-p) \ge 10$ .

# Example: Are attitudes toward shopping changing?

- A survey asked a random sample of 2500 adults if they agreed or disagreed that "I like buying new clothes, but shopping is often frustrating and time-consuming." The population that the poll wants to draw conclusions about is all US residents aged 18 and over. Suppose that in fact 60% of all adult US residents would say Agree if asked the same question, what is the probability that 1520 or more of the sample agree?
  - B/c there are more than 218 million adults, we can take the responses of 2500 randomly chosen adults to be independent. So the number in our sample who agree that shopping is frustrating is a random variable X having the binomial distribution with n = 2500 and p = .6.
  - To find the probability that at least 1520 of the people in the sample find shopping frustrating, we must add the binomial probabilities of all outcomes from X = 1520 to X = 2500. This isn't practical.
    - Method 1: P(X≥1520( = 1 P(X≤ 1519) on calculator which = .2131
    - Method 2: B/c it's so large, we can use the normal distribution and find area under the curve using NormalCDF! So area under curve N(1500, 24.49)\*\* = .2061 (we're only off from the actual calculation by .0007!)

### Geometric Distributions

- The geometric distribution is a special case of the binomial distribution. It deals with the number of trials required to obtain your first success.
- An example of a geometric distribution would be tossing a coin until it lands on heads. We might ask: What is the probability that the first head occurs on the third flip? That probability is referred to as a **geometric probability** and is denoted by g(x; P).

#### Calculator

- DISTR- geometpdf(p,X) which gives probability of success (p) on the Xth trial.
- Example: It is estimated that 45% of people in Fast-Food restaurants order a diet drink with their lunch. Find the probability that the fourth person orders a diet drink. (7.5%)
  - How could we find the probability that the first diet drinker of the day occurs **before** the 5th person?
  - o This last problem can also be done using the **geometcdf** function which will calculate the probability of success **on or before** the X<sub>th</sub> tr

# Mean and SD of Geometric random variable

- If X is a random variable with probability p on each trial, the mean (or expected value) is μ = 1/p which means that the expected number of trials required for the first success is 1/p
- The probability that its takes more than n trials to see the first success is  $(q)^n$ .
- The variance of X is (q)/p²
  - (SD obviously square root of this)

### Example

- In New York City at rush hour, the chance that a taxicab passes someone and is available is 15%.
  - a) How many cabs can you expect to pass you for you to find one that is free
    - 6.67 so 7 cabs
  - b) what is the probability that more than 10 cabs pass you before you find one that is free.
    - **o** 19.68%

# Histograms of Geometric distributions

- For illustration, we will use the roll of a die with n = 6 likely outcomes, and probability p = 1/6 of rolling a 3 (our success). The random variable X is the number of rolls until a 3 is observed.
- Enter numbers 1-10 in L1
- L2 is geometpdf(1/6, L1) ENTER.
- L3 you can make geometcdf(1/6, L1)ENTER
- Do the histograms as before