# Lecture 16: Chapter 7, Section 2 Binomial Random Variables 

-Definition
$\square$ What if Events are Dependent?
םCenter, Spread, Shape of Counts, Proportions
■Normal Approximation

## Looking Back: Review

## - 4 Stages of Statistics

- Data Production (discussed in Lectures 1-4)
- Displaying and Summarizing (Lectures 5-12)
- Probability
- Finding Probabilities (discussed in Lectures 13-14)
- Random Variables (introduced in Lecture 15)

- Sampling Distributions
- Statistical Inference


## Definition (Review)

- Discrete Random Variable: one whose possible values are finite or countably infinite (like the numbers $1,2,3, \ldots$ )


## Looking Ahead: To perform inference about

 categorical variables, need to understand behavior of sample proportion. A first step is to understand behavior of sample counts. We will eventually shift from discrete counts to a normal approximation, which is continuous.
## Definition

## Binomial Random Variable counts sampled

 individuals falling into particular category;- Sample size $n$ is fixed
- Each selection independent of others
- Just 2 possible values for each individual
- Each has same probability $p$ of falling in category of interest


## Example: A Simple Binomial Random Variable

$\square$ Background: The random variable $X$ is the count of tails in two flips of a coin.
$\square$ Questions: Why is $X$ binomial? What are $n$ and $p$ ?
$\square$ Responses:

- Sample size $n$ fixed?
- Each selection independent of others?
- Just 2 possible values for each?
- Each has same probability $p$ ?


## Example: A Simple Binomial Random Variable

$\square$ Background: The random variable $X$ is the count of tails in two flips of a coin.
$\square$ Question: How do we display $X$ ?

- Response:

> Looking Back: We already discussed and displayed this random variable when learning about probability distributions.

## Example: Determining if R.V. is Binomial

$\square$ Background: Consider following R.V.:

- Pick card from deck of 52, replace, pick another. $X=$ no. of cards picked until you get ace.
$\square$ Question: Is $X$ binomial?
$\square$ Response:


## Example: Determining if R.V. is Binomial

$\square$ Background: Consider following R.V.:

- Pick 16 cards without replacement from deck of 52. $X=$ no. of red cards picked.
$\square$ Question: Is $X$ binomial?
$\square$ Response:


## Example: Determining if R.V. is Binomial

$\square$ Background: Consider following R.V.:

- Pick 16 cards with replacement from deck of 52. $W=$ no. of clubs, $X=$ no. of diamonds, $Y=$ no. of hearts, $Z=$ no. of spades. Goal is to report how frequently each suit is picked.
$\square$ Question: Are $W, X, Y, Z$ binomial?
$\square$ Response:


## Example: Determining if R.V. is Binomial

$\square$ Background: Consider following R.V.:

- Pick with replacement from German deck of 32 (doesn't include numbers 2-6), then from deck of 52 , back to deck of 32 , etc. for 16 selections altogether. $X=$ no. of aces picked.
$\square$ Question: Is $X$ binomial?
$\square$ Response:


## Example: Determining if R.V. is Binomial

- Background: Consider following R.V.:
- Pick 16 cards with replacement from deck of 52. $X=$ no. of hearts picked.
$\square$ Question: Is $X$ binomial?
$\square$ Response:
- fixed $n=16$
- selections independent (with replacement)
- just 2 possible values (heart or not)
- same $p=0.25$ for all selections
$\rightarrow$


## Requirement of Independence

## Snag:

- Binomial theory requires independence
- Actual sampling done without replacement so selections are dependent
Resolution: When sampling without replacement, selections are approximately independent if population is at least $10 n$.


## Example: A Binomial Probability Problem

$\square$ Background: The proportion of Americans who are left-handed is 0.10 . Of 44 presidents, 7 have been left-handed (proportion 0.16).
$\square$ Question: How can we establish if being left-handed predisposes someone to be president?
$\square$ Response: Determine if 7 out of 44 (0.16) is when sampling at random from a population where 0.10 fall in the category of interest.

## Solving Binomial Probability Problems

- Use binomial formula or tables

Only practical for small sample sizes

- Use software

Won't take this approach until later

- Use normal approximation for count $X$

Not quite: more interested in proportions

- Use normal approximation for proportion

Need mean and standard deviation...

## Example: Mean of Binomial Count, Proportion

$\square$ Background: Based on long-run observed outcomes, probability of being left-handed is approx. 0.1. Randomly sample 100 people.
$\square$ Questions: On average, what should be the

- count of lefties?
- proportion of lefties?
$\square$ Responses: On average, we should get
- count of lefties
- proportion of lefties


## Mean and S.D. of Counts, Proportions

Count $X$ binomial with parameters $n, p$ has:

- Mean $n p$
- Standard deviation $\sqrt{n p(1-p)}$ Sample proportion $\hat{p}=\frac{X}{n}$ has:
- Mean $p$
- Standard deviation $\sqrt{\frac{p(1-p)}{n}}$

Looking Back: Formulas for s.d. require independence: population at least $10 n$.

## Example: Standard Deviation of Sample Count

$\square$ Background: Probability of being left-handed is approx. 0.1. Randomly sample 100 people. Sample count has mean $100(0.1)=10$, standard deviation $\sqrt{100(0.1)(1-0.1)}=3$
$\square$ Question: How do we interpret these?
$\square$ Response: On average, expect sample count lefties.
Counts vary; typical distance from 10 is

## Example: S.D. of Sample Proportion

$\square$ Background: Probability of being left-handed is approx. 01 . Randomly sample 100 people.
Sample proportion has mean 0.l, standard deviation $\sqrt{\frac{0.1(1-0.1)}{100}}=0.03$
$\square$ Question: How do we interpret these?
$\square$ Response: On average, expect
sample proportion $=\quad$ lefties.
Proportions vary; typical distance from 0.1 is

## Example: Role of Sample Size in Spread

- Background: Consider proportion of tails in various sample sizes $n$ of coinflips.
- Questions: What is the standard deviation for

$$
n=1 ? n=4 ? n=16 ?
$$

$\square$ Responses:

| - | $\boldsymbol{n}=\mathbf{1}:$ s.d. $=$ |
| :--- | :--- |
| - | $\boldsymbol{n}=\mathbf{4}:$ s.d. $=$ |
|  | $\boldsymbol{n}=\mathbf{1 6 :}$ s.d. $=\square$ |

A Closer Look: Due to $n$ in the denominator of formula for standard deviation, spread of sample proportion as n increases.

## Shape of Distribution of Count, Proportion

Binomial count $X$ or proportion $\hat{p}=\frac{X}{n}$ for repeated random samples has shape approximately normal if samples are large enough to offset underlying skewness. (Central Limit Theorem)
For a given sample size $n$, shapes are identical for count and proportion.

## Example: Underlying Coinflip Distribution

$\square$ Background: Distribution of count or proportion of tails in $n=1$ coinflip ( $p=0.5$ ):


$\square$ Question: What are the distributions' shapes?
$\square$ Response:

## Example: Distribution for 4 Coinflips

- Background: Distribution of count or proportion of tails in $n=4$ coinflips ( $p=0.5$ ):


$X=$ Count of tails in 4 coin flips
$\hat{p}=$ Proportion of tails in 4 coin flips
$\square$ Question: What are the distributions' shapes?
$\square$ Response:


## Shift from Counts to Proportions

Binomial Theory begins with counts

- Inference will be about proportions


## Example: Distribution of $\widehat{p}$ for 16 Coinflips

- Background: Distribution of proportion of tails in $n=16$ coinflips ( $p=0.5$ ):

$\square$ Question: What is the shape?
$\square$ Response:


## Example:Underlying Distribution of Lefties

$\square$ Background: Distribution of proportion of lefties $(p=0.1)$ for samples of $n=1$ :


- Question: What is the shape?
$\square$ Response:


## Example: Dist of $\hat{p}$ of Lefties for $n=16$

$\square$ Background: Distribution of proportion of lefties $(p=0.1)$ for $n=16$ :

$\square$ Question: What is the shape?
$\square$ Response:

## Example: Dist of $\hat{p}$ of Lefties for $n=100$

$\square$ Background: Distribution of proportion of lefties $(p=0.1)$ for $n=100$ :


- Question: What is the shape?
$\square$ Response:


## Rule of Thumb:

## Sample Proportion Approximately Normal

Distribution of $\widehat{p}$ is approximately normal if sample size $n$ is large enough relative to shape, determined by population proportion $p$.
Require $n p \geq 10$ and $n(1-p) \geq 10$

Together, these require us to have larger $n$ for $p$ close to 0 or 1 (underlying distribution skewed right or left).

## Example: Applying Rule of Thumb

$\square$ Background: Consider distribution of sample proportion for various $n$ and $p$ :
$n=4, p=0.5 ; n=20, p=0.5 ; n=20, p=0.1 ; n=20, p=0.9 ; n=100, p=0$.
$\square$ Question: Is shape approximately normal?
$\square$ Response: Normal?

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- \(n=4, p=0.5\)
- \(n=20, p=0.5\)
    \([n p=4(0.5)=2<10]\)
    \([n p=20(0.5)=10=n(1-p)]\)
    - \(n=20, p=0.1 \quad\) No [
    - \(n=20, p=0.9 \quad\) No |
    - \(n=100, p=0.1\)
    \([n p=100(0.1)=10, n(1-p)=100(0.9)=90\) both \(\geq 10]\)
```


## Example: Solving the Left-handed Problem

- Background: The proportion of Americans who are lefties is 0.1. Consider $\mathrm{P}(\widehat{p} \geq 7 / 44=0.16)$ for a sample of 44 presidents.
$\square$ Question: Can we use a normal approximation to find the probability that at least 7 of $44(0.16)$ are left-handed?
$\square$ Response:



## Example: From Count to Proportion and Vice

## Versa

$\square$ Background: Consider these reports:

- In a sample of 87 assaults on police, 23 used weapons.
- 0.44 in sample of 25 bankruptcies were due to med. bills
- Question: In each case, what are $n, X$, and $\widehat{p}$ ?
$\square$ Response:
- First has $n=\quad x=\square \widehat{p}=$
- Second has $n=$

$$
\widehat{p}=\quad x=
$$

## Lecture Summary

(Binomial Random Variables)

- Definition; 4 requirements for binomial
$\square$ R.V.s that do or don't conform to requirements
$\square$ Relaxing requirement of independence
- Binomial counts, proportions
- Mean
- Standard deviation
- Shape
- Normal approximation to binomial

