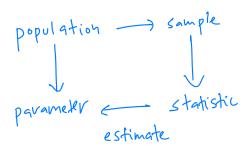
Lecture 6: Sampling Distributions STAT 310, Spring 2021

- ► A **parameter** is a numerical characteristic of the population (fixed number that is usually unknown).
- A **statistic** is a numerical characteristic of the sample (varies depending on sample).
- ► The statistic is also referred to as a **point estimate**, since it is our best guess at the value of a population parameter.





### Notation for the proportion:

- p: population proportion
- $\triangleright$   $\hat{p}$ : sample proportion

#### Notation for the mean:

- $\blacktriangleright \mu$ : population mean
- $ightharpoonup \bar{x}$ : sample mean

The sample size is denoted as n

## Example

A recent Gallup poll found that 66% of Americans are dissatisfied with how the COVID-19 vaccination process is going in the U.S. The survey results were based on random sample of 4,098 adults.<sup>1</sup>

(a) What is the sample proportion?

$$\hat{p} = 0.66$$

(b) What is the sample size?

$$n = 4,098$$

(c) Describe the population proportion?

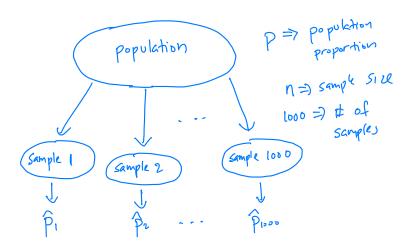
P: population proportion of Americans that

(d) Suppose another poll is conducted with a different random sample of adults. Would you expect the sample proportion to be the same, or slightly different?

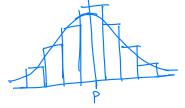
https:

- Sampling Error refers to how much a statistic, such as the sample proportion, will vary from one random sample to the next.
- ▶ For example, the Gallup poll reported a sampling error of  $\pm 2$  percentage points. This means that the population proportion of Americans that are dissatisfied with the vaccine rollout is likely between 64% and 68%.

A **sampling distribution** is the distribution of a statistic when repeatedly taking random samples from a population.



Sampling distribution is histogram of



- When n is large, histogram looks normal and centered around population proportion P
- The standard deviation of the loop sample proportions is called the <u>standard error</u> (SE). It measures the variability, or spread, of the estimates from sample to sample.

- In real-world applications, we never actually observe the sampling distribution, since we usually take a single random sample.
- However, it is useful to always think of a statistic, such as the sample proportion, as coming from such a hypothetical distribution.
- ► The concept of a sampling distribution is very important when trying to quantify sampling error.

# Central Limit Theorem (CLT)

The sampling distribution for  $\hat{p}$  follows an approximate normal distribution centered around the population proportion p, and with standard error  $\sqrt{p(1-p)/n}$ .

### Conditions for CLT

The following conditions should be met to apply the CLT:

- ► The data come from a simple random sample. This is called the independence condition since it implies that the individuals or cases in the data are unrelated.
- ▶  $np \ge 10$  and  $n(1-p) \ge 10$ . This is sometimes called the success-failure condition since np can be interpreted as the expected number of successes and n(1-p) the expected number of failures.

## Example

Suppose that the population proportion of Americans who support the expansion of solar energy is p=0.88, and n=1000 Americans are randomly sampled.

(a) What is the mean, or center, of the sampling distribution for  $\hat{p}$ ?

(b) What is the standard error of the sampling distribution for  $\hat{p}$ ?

$$SE = \int \frac{P(1-P)}{N} = \int \frac{0.88(0.12)}{1000} = 0.01$$

(c) What distribution does  $\hat{p}$  follow?

(d) Are the conditions for the CLT satisfied?

• random sample 
$$\sqrt{}$$
•  $np = 1000(0.88) = 880210 n(1-p) = 1000(0.12)$ 

## Example

Suppose that population proportion of Americans who support the expansion of solar energy is p=0.88, and n=1000 Americans are randomly sampled. What is the probability that the sample proportion  $\hat{p}$  will be greater than 0.9?

By (LT 
$$\hat{p} \sim N(p, \sqrt{p(1-p)}) = N(0.88, 0.01)$$

sampling dist for



$$Z = \frac{0.9 - 0.88}{0.01} = 2$$

$$P(272) = 1 - P(2<2)$$
  
=  $1 - pnorm(2)$   
=  $[0.023]$