

# Class 17

## DATA1220-55, Fall 2024

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# Recap: The Central Limit Theorem

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- ▶ A distribution of multiple sample means approximates a normal distribution as the sample size for each mean gets larger
- ▶ If you take an infinite number of samples of size  $n$  from a population, the *sample statistics* (i.e. means  $\bar{x}_1, \bar{x}_2, \dots, \bar{x}_\infty$ ) have a probability distribution (i.e. the **sampling distribution**) that is about normal

## Recap: CLT Requirements

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- ▶ Requires at least  $n > 30$  for  $\bar{x}$
- ▶ Requires independent observations
- ▶ Requires identically distributed (i.i.d.) observations

# Recap: The CLT in Practice

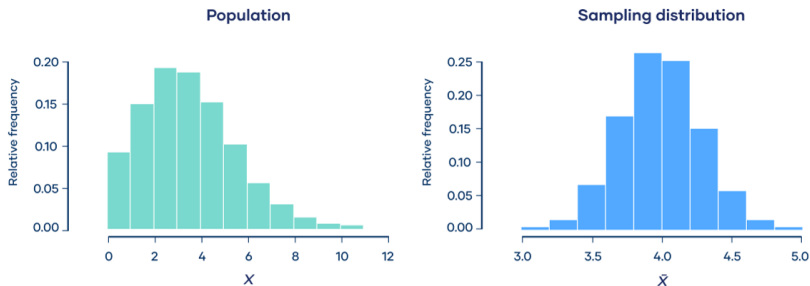


Figure 1: The **sampling distribution** of an infinite number of **sample statistics** from a population approximates a normal distribution.



# Recap: Standard Error of the Sample Statistic

- ▶ **Standard error (SE)** is the *standard deviation* of the *sample statistic* in a theoretical *sampling distribution*
- ▶ If you took an infinite number of samples from a known distribution, the **standard error** is the standard deviation of the means of those samples
- ▶ Describes the scale (i.e. variability, sampling error) of the sampling distribution

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As  $n$  increases, the standard error  $SE$  decreases.

## Recap: Calculating a Z-Score

A **Z-score** indicates how many standard deviations  $\sigma$  away from the mean  $\mu$  a given observation is.

$$\begin{aligned} Z &= \frac{\text{observedvalue} - \text{mean}}{\text{standarddeviation}} \\ &= \frac{x - \mu}{\sigma} \end{aligned}$$

# Recap: Accuracy vs Precision

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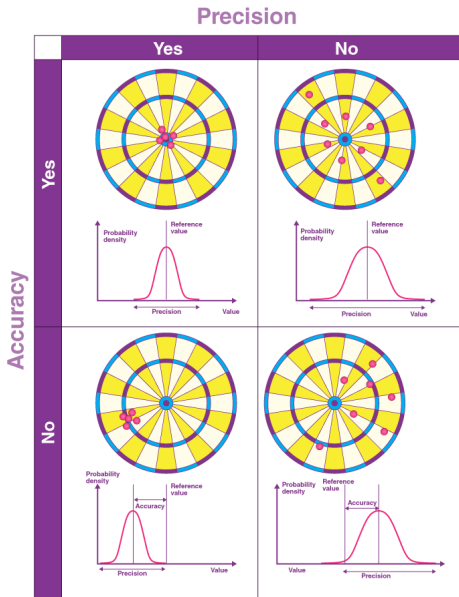
- ▶ Accuracy describes how similar an observation or statistic is to the “true” population parameter

## Recap: Accuracy vs Precision

- ▶ Accuracy describes how similar an observation or statistic is to the “true” population parameter
- ▶ Precision describes how similar the observations or statistics in a distribution are to each other (i.e. the variability of the estimates)



# Recap: Accuracy & Precision of Estimates



# Recap: Point Estimates & Confidence Intervals

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- ▶ A ***confidence interval*** describes the ***scale*** of an estimate or distribution
- ▶ The ***confidence threshold*** or ***confidence level*** describes our uncertainty regarding these values

## Recap: Confidence Intervals

A ***confidence interval*** is a numerical range *inside* which a statistic is expected to occur with a given probability  $1 - \alpha$  (alpha) in any theoretical sample from a given population

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- ▶  $1 - \alpha$  is the ***confidence level*** and is often expressed as a %
- ▶ ***This is only true if your assumptions about the population hold.***

## Recap: Confidence Intervals in Practice

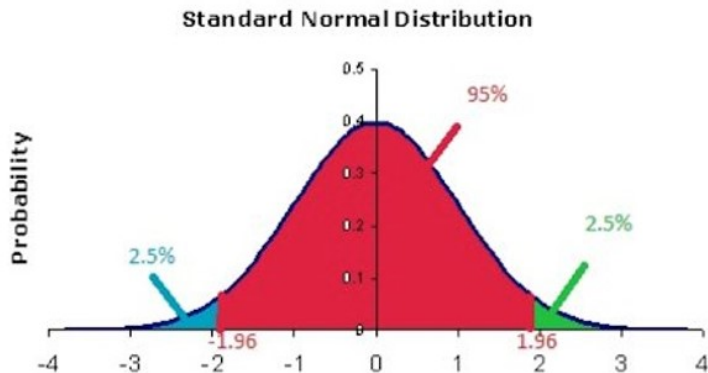


Figure 3: Properties of known distributions, like the 68-95-99.7 Rule, are used to calculate the bounds of a confidence interval.



## Recap: Confidence Intervals & $Z^*$

- ▶ A confidence interval is defined as  $\text{pointestimate} \pm \text{marginoferror}$
- ▶  $\text{marginoferror} = Z^* \times SE$
- ▶  $Z^* = Z\text{-Score}_{\alpha/2}$

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2. Your data is **reliable**, so your sample statistics are **reliable** estimations of your sample population distribution.
3. Your data is **valid**, so a sampling distribution based on your sample statistics is a **valid** estimation of the “true” distribution in the study population.
4. Your data is **generalizable**, so your estimated sampling distribution for your study population is **generalizable** as the “true” sampling distribution for your target population

# Statistical Inference and Hypothesis Testing

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- ▶ We use sample statistics to describe sample populations and estimate the parameters of the study population's sampling distribution
- ▶ We also describe the variability of our measure and quantify our uncertainty regarding our estimate
- ▶ We use the overlap between theoretical distributions to decide how meaningful the differences between groups are



# Hypothesis Testing Framework

- ▶  $H_0$ : The “Null” Hypothesis
  - ▶ Represents a position of skepticism, *nothing* is happening here
  - ▶ “There is *not* an association between process A and B”

# Hypothesis Testing Framework

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  - ▶ Represents a position of skepticism, *nothing* is happening here
  - ▶ “There is *not* an association between process A and B”
- ▶  $H_A$ : The “Alternative” Hypothesis
  - ▶ The complement of  $H_0$ , *something* is happening here
  - ▶ “There *is* an association between process A and B”

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- ▶ Calculate *the probability that you would see results as extreme or more extreme* than what you saw in your study, assuming the distribution under  $H_0$
- ▶ The lower the probability, the less likely it is that we would see these results if  $H_0$  was the “true” state of our population
- ▶ If the probability is sufficiently low, we *reject*  $H_0$  and *accept*  $H_A$

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- ▶ The probability below which you will reject the null hypothesis
- ▶ Predetermined before doing hypothesis test (often  $p < 0.05$ )
- ▶ Also the probability of rejecting the null hypothesis when  $H_0$  is true (i.e. ***Type I Error*** or ***false positive rate***)

# Decision Errors

	Decision	
	fail to reject $H_0$	reject $H_0$
Truth		
$H_0$ true	✓	Type 1 Error
$H_A$ true	Type 2 Error	✓

		Predicted	
		Positive	Negative
Actual	Positive	True Positive (TP)	False Negative (FN)
	Negative	False Positive (FP)	True Negative (TN)

# Inference for a Single Proportion

**Central Limit Theorem for Proportions:** sample proportions  $\hat{p}$  will be nearly normally distributed with the mean equal to the population proportion ( $\mu = p$ ) and the standard deviation equal to the standard error for a proportion ( $\sigma = \sqrt{\frac{p(1-p)}{n}}$ ), such that  $\hat{p} \sim N(\mu = p, \sigma = SE_p)$ .

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**Assumptions:** independence, identically distributed, 10+ successes/failures each

## Example

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***Research Question:*** Is a hurricane more likely to hit the continental US in 2024?



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***What is the study population?***

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***What is the study population?***

*All hurricanes which formed in the Atlantic Ocean with the potential to make landfall in the continental US, for which we have records.*

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***What is the sample population?***

*298 hurricanes which formed in the Atlantic Ocean between 1980-2023 with the potential to make landfall in the continental United States.*

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***What is the target population?***

*Future hurricanes which form in the Atlantic Ocean with the potential to make landfall in the continental US.*

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Is it reasonable to assume that the estimated sampling distribution for the study population will be generalizable to the unobserved distribution in the target population?

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- ▶ Are the observations *independent*?
- ▶ Are the observations *identically distributed*?
- ▶ Is the *sample size sufficient*?

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If we assume our data is valid, then we can use our sample statistics to *infer* the sampling distribution for the study population.

If we assume our data is generalizeable, then we can use our sampling distribution to *test the hypothesis* in the target population.



## Example

Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 1: Calculate the sample statistic.

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$$\hat{p} = \frac{72}{298} = 0.242$$

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 2: Estimate the sampling distribution.

$$SE = \sqrt{\frac{0.242(1 - 0.242)}{298}} = 0.025$$

The sampling distribution for  $\hat{p}$  approximates the normal distribution  $N(24.2, 2.5)$ .

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Step 3: Calculate  $Z^*$  for the confidence threshold  $\alpha = 0.05$ .

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$$Z^* = Z_{\alpha/2}$$

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```
qnorm(0.05 / 2)
```

```
[1] -1.959964
```

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 4: Construct a 95% confidence interval.



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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 4: Construct a 95% confidence interval.

pointestimate  $\pm Z^* \times SE$

```
24.2 - qnorm(0.05 / 2) * 2.5
```

```
[1] 29.09991
```

```
24.2 + qnorm(0.05 / 2) * 2.5
```

```
[1] 19.30009
```

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

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24.2 + qnorm(0.05 / 2) * 2.5
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With 95% confidence, the probability of a hurricane making landfall in the continental US is 19.3% to 29.1%.

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Step 5: Assume the null hypothesis.

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 5: Assume the null hypothesis.

$H_0$ : The probability of a hurricane making landfall in 2024 is 24.2% ( $\hat{p} = 24.2\%$ ).

$H_A$  The probability of a hurricane making landfall in 2024 is *not* 24.2% ( $\hat{p} \neq 24.2\%$ ).

## Example

Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 6: Calculate the sample statistic.

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

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$$\begin{aligned}\hat{p} &= \frac{2}{9} \\ &= 0.222\end{aligned}$$

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 7: Calculate the test statistic under  $H_0$ .

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$$\begin{aligned} Z &= \frac{\hat{p} - p}{SE} \\ &= \frac{22.2 - 24.2}{2.5} \\ &= -0.8 \end{aligned}$$



## Example

Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 8: Calculate the p-value under  $H_0$ .

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```
pnorm(-0.8, mean = 0, sd = 1) * 2
```

```
[1] 0.4237108
```

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

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The p-value for the observed data under the null hypothesis is  $p = 0.423$ . As  $p > \alpha$  ( $\alpha = 0.05$ ), this is *not* sufficient evidence of a difference.

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Based on the data from 1980-2023, what is the average probability that a hurricane makes landfall in the continental US?

Step 9: Reject or fail to reject the null hypothesis.

The p-value for the observed data under the null hypothesis is  $p = 0.423$ . As  $p > \alpha$  ( $\alpha = 0.05$ ), this is *not* sufficient evidence of a difference.

***We fail to reject the null hypothesis that the probability of a hurricane making landfall in 2024 is 24.2%.***

# Example

