

Lecture (chapter 8): Hypothesis testing I: The one-sample case

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Source: Healey, Joseph F. 2015. "Statistics: A Tool for Social Research." Stamford: Cengage Learning. 10th edition. Chapter 8 (pp. 185–215).



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Chapter learning objectives

- Explain the logic of hypothesis testing, including concepts of the null hypothesis, the sampling distribution, the alpha level, and the test statistic
- Explain what it means to “reject the null hypothesis” or “fail to reject the null hypothesis”
- Identify and cite examples of situations in which one-sample tests of hypotheses are appropriate
- Test the significance of single-sample means and proportions using the five-step model, and correctly interpret the results
- Explain the difference between one- and two-tailed tests, and specify when each is appropriate
- Define and explain Type I and Type II errors, and relate each to the selection of an alpha level
- Use the Student’s t distribution to test the significance of a sample mean for a small sample



Significant differences

- Hypothesis testing is designed to detect significant differences
 - Differences that did not occur by random chance
 - Hypothesis testing is also called significance testing
- This chapter focuses on the “one sample” case
 - Compare a random sample against a population
 - Compare a sample statistic to a (hypothesized) population parameter to see if there is a statistically significant difference



Example 1: Question

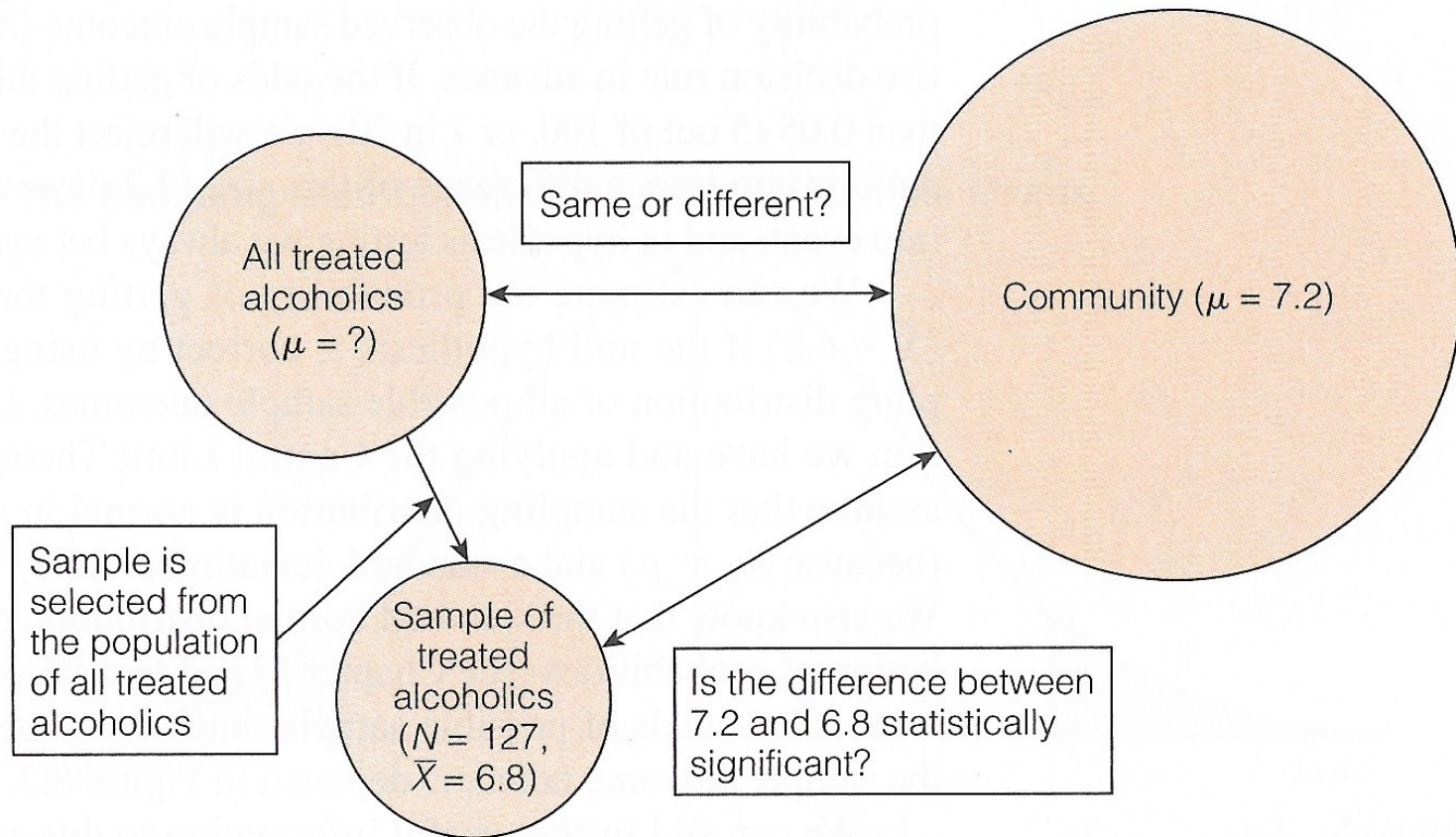
- Are people who have been treated for alcoholism more reliable workers than those in the community?
 - Does the group of all treated alcoholics have different absentee rates than the community as a whole?
 - Effectiveness of rehabilitation center for alcoholics
- Absentee rates for community and sample
 - Don't have resources to gather information of all people who have been treated by the program

Community	Sample of treated alcoholics
$\mu = 7.2$ days per year	$\bar{X} = 6.8$ days per year
$\sigma = 1.43$	$N = 127$

- What causes the difference between 7.2 and 6.8?
 - Real difference? Or difference due to random chance?



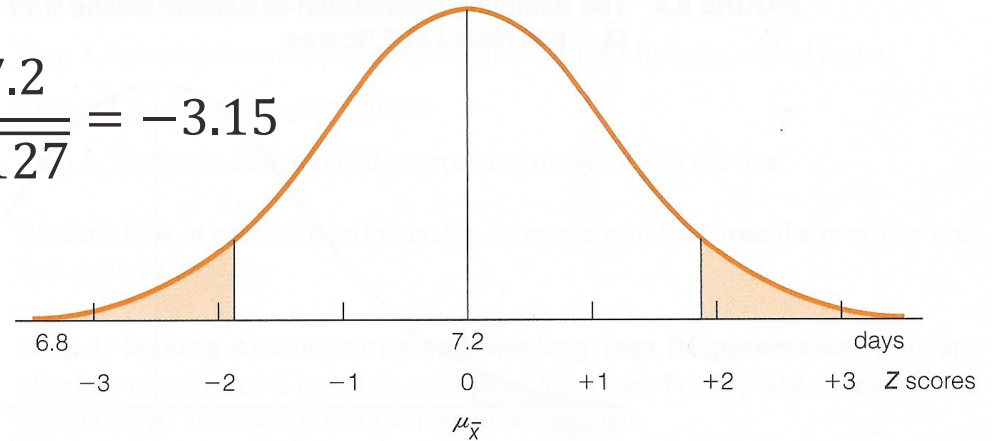
A test of hypothesis for single-sample means



Example 1: Result

- For a known/empirical distribution, we use: $Z = \frac{X_i - \bar{X}}{s}$
- However, we are concerned with the sampling distribution of all possible sample means

$$Z(\text{obtained}) = \frac{\bar{X} - \mu}{\sigma/\sqrt{N}} = \frac{6.8 - 7.2}{1.43/\sqrt{127}} = -3.15$$



- The sample outcome falls in the shaded area
 - $Z(\text{obtained}) = -3.15$
 - Reject $H_0: \mu = 7.2$ days per year
 - The sample of 127 treated alcoholics comes from a population that is significantly different from the community on absenteeism

The five-step model

1. Make assumptions and meet test requirements
2. Define the null hypothesis (H_0)
3. Select the sampling distribution and establish the critical region
4. Compute the test statistic
5. Make a decision and interpret the test results

Example 2: Question

- The education department at a university has been accused of “grade inflation”
 - So education majors have much higher GPAs than students in general
- GPAs of all education majors should be compared with the GPAs of all students
 - There are 1000s of education majors, far too many to interview
 - How can the dispute be investigated without interviewing all education majors?



Example 2: Numbers

- The average GPA for all students is 2.70 (μ)
 - This value is a parameter
- Random sample of education majors
 - Mean = $\bar{X} = 3.00$
 - Standard deviation = $s = 0.70$
 - Sample size = $N = 117$
- There is a difference between parameter ($\mu=2.70$) and statistic ($\bar{X}=3.00$)
 - It seems that education majors do have higher GPAs



Example 2: Explanations

- We are working with a random sample
 - Not all education majors
- Two explanations for the difference
 1. The sample mean ($\bar{X}=3.00$) is the same as the population mean ($\mu=2.70$)
 - The observed difference may have been caused by random chance
 2. The difference is real (statistically significant)
 - Education majors are different from all students



Step 1: Assumptions, requirements

- Make assumptions
 - Random sampling
 - Hypothesis testing assumes samples were selected according to EPSEM
- Meet test requirements
 - The sample of 117 was randomly selected from all education majors
 - Level of measurement is interval-ratio
 - GPA is an interval-ratio level variable, so the mean is an appropriate statistic
 - Sampling distribution is normal in shape
 - This is a large sample ($N \geq 100$)



Step 2: Null hypothesis

- Null hypothesis, $H_0: \mu = 2.7$
 - H_0 always states there is no significant difference
 - The sample of 117 comes from a population that has a GPA of 2.7
 - The difference between 2.7 and 3.0 is trivial and caused by random chance
- Alternative hypothesis, $H_1: \mu \neq 2.7$
 - H_1 always contradicts H_0
 - The sample of 117 comes from a population that does not have a GPA of 2.7
 - There is an actual difference between education majors ($\bar{X}=3.0$) and other students ($\mu=2.7$)



Step 3: Distribution, critical region

- Sampling distribution: standard normal shape
 - Alpha (α) = 0.05
 - Use the 0.05 value as a guideline to identify differences that would be rare if H_0 is true
 - Any difference with a probability less than α is rare and will cause us to reject the H_0
- Use the Z score to determine the probability of getting the observed difference
 - If the probability is less than 0.05, the obtained Z score will be beyond the critical Z score of ± 1.96
 - This is the critical Z score associated with a two-tailed test and $\alpha=0.05$



Step 4: Test statistic

- For a known/empirical distribution, we would use

$$Z = \frac{X_i - \bar{X}}{s}$$

- However, we are concerned with the sampling distribution of all sample means
- We only have the standard deviation for the sample (s), not for the population (σ)

$$Z(\textit{obtained}) = \frac{\bar{X} - \mu}{s/\sqrt{N - 1}} = \frac{3.0 - 2.7}{0.7/\sqrt{117 - 1}} = 4.62$$



Step 5: Decision, interpret

- $Z(\text{obtained}) = 4.62$
 - This is beyond $Z(\text{critical}) = \pm 1.96$
 - The obtained Z score fell in the critical region, so we **reject** the H_0
 - If H_0 was true, a sample GPA of 3.0 would be unlikely
 - Therefore, the H_0 is false and must be rejected
- Education majors have a GPA that is significantly higher than general student body
 - The difference between the parameter ($\mu=2.7$) and the statistic ($\bar{X}=3.0$) was large and unlikely to have occurred by random chance ($p<0.05$)



Five-step model summary

Situation	Decision	Interpretation
The test statistic is in the critical region	Reject the null hypothesis (H_0)	The difference is statistically significant
The test statistic is not in the critical region	Fail to reject the null hypothesis (H_0)	The difference is not statistically significant

- Model is fairly rigid, but there are two crucial choices
 - One-tailed or two-tailed test
 - Alpha (α) level



One or two-tailed test

- Null hypothesis always has the equal sign

$$H_0: \mu = 2.7$$

- Two-tailed test states that population mean is not equal to the value stated in null hypothesis

$$H_1: \mu \neq 2.7$$

- One-tailed test estimates differences in a specific direction (based on theory)

$$H_1: \mu > 2.7$$

$$H_1: \mu < 2.7$$



One or two-tailed test

One- vs. Two-Tailed Tests, $\alpha = 0.05$

If the Research Hypothesis (H_1) Uses	The Test Is	Concern Is on	Z(critical) Is
\neq	Two-tailed	Both tails	± 1.96
$>$	One-tailed	Upper tail	+1.65
$<$	One-tailed	Lower tail	-1.65

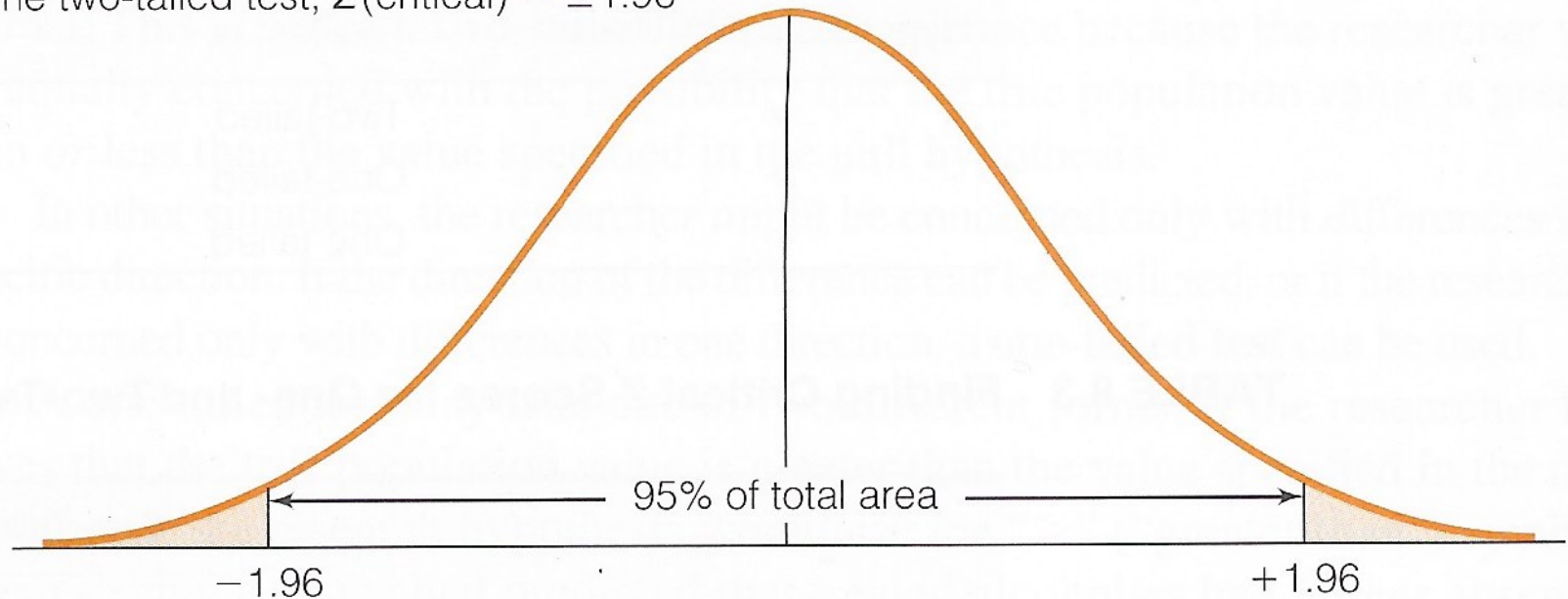
Finding Critical Z Scores for One- and Two-Tailed Tests

Alpha	Two-Tailed Value	One-Tailed Value	
		<i>Upper Tail</i>	<i>Lower Tail</i>
0.10	± 1.65	+1.29	-1.29
0.05	± 1.96	+1.65	-1.65
0.01	± 2.58	+2.33	-2.33
0.001	± 3.32	+3.10	-3.10
0.0001	± 3.90	+3.70	-3.70



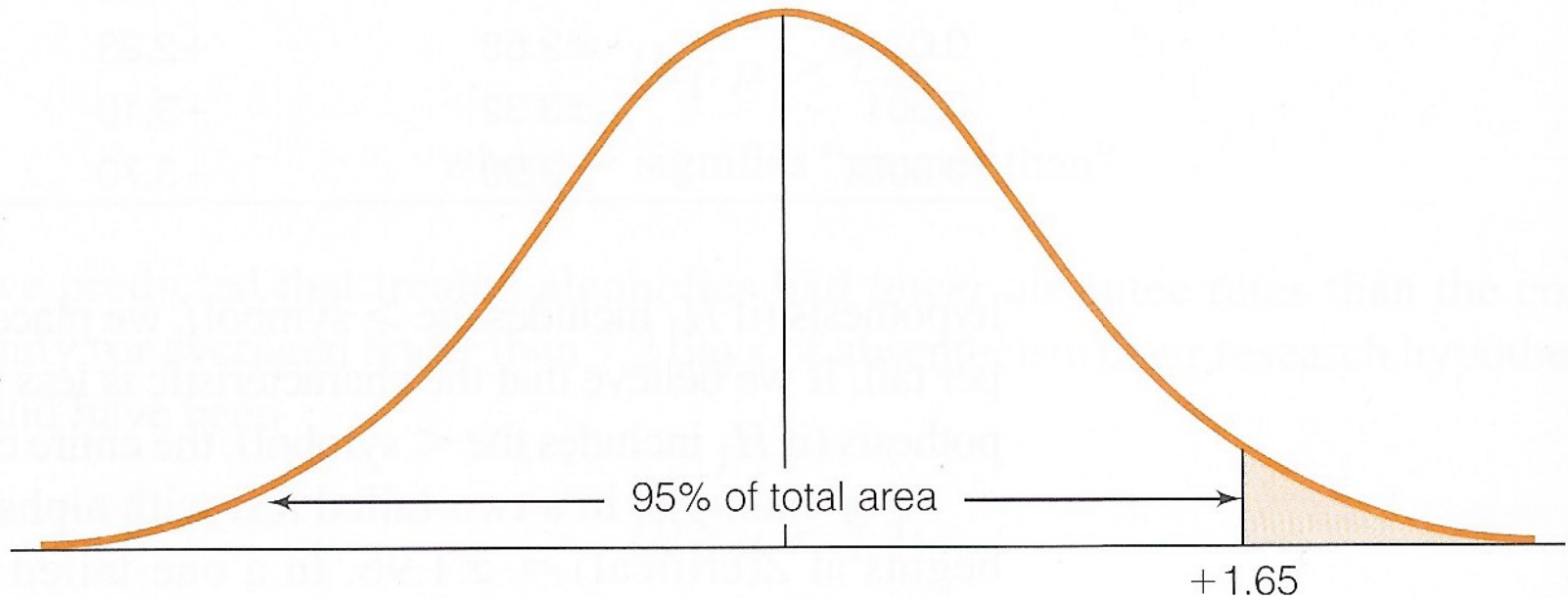
Two-tailed test: $\alpha=0.05$

A. The two-tailed test, $Z(\text{critical}) = \pm 1.96$



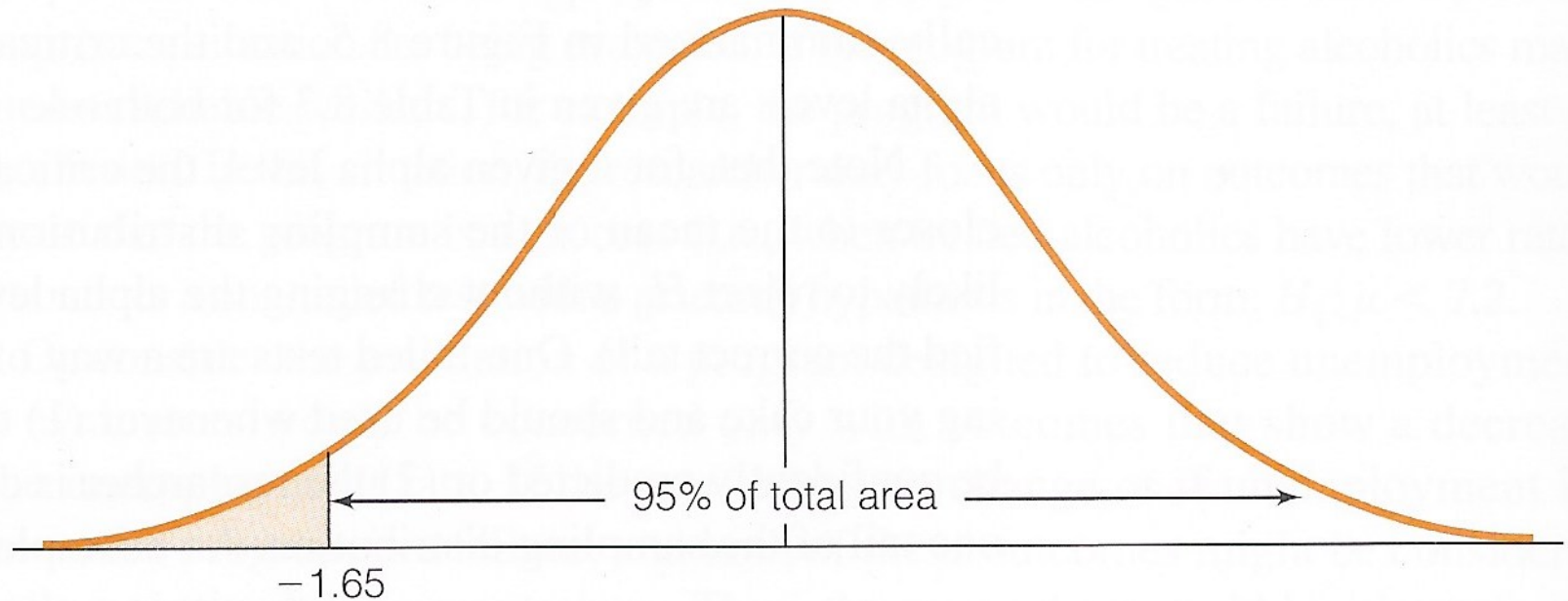
One-tailed test (upper): $\alpha=0.05$

B. The one-tailed test for upper tail, $Z(\text{critical}) = +1.65$



One-tailed test (lower): $\alpha=0.05$

C. The one-tailed test for lower tail, $Z(\text{critical}) = -1.65$



Selecting an alpha level

- By assigning an alpha level, one defines an “unlikely” sample outcome
- Alpha level is the probability that the decision to reject the null hypothesis is incorrect
- Examine this table for critical regions

The Relationship Between Alpha and Z(Critical) for a Two-Tailed Test

If Alpha =	The Two-Tailed Critical Region Will Begin at Z(Critical) =
0.100	± 1.65
0.050	± 1.96
0.010	± 2.58
0.001	± 3.32



Type I and Type II errors

- Type I error (alpha error)
 - Rejecting a true null hypothesis
- Type II error (beta error)
 - Failing to reject a false null hypothesis
- Examine table below for relationships between decision making and errors

Decision Making and the Five-Step Model

	If Our Decision Is to	And H_0 Is Actually	The Result Is
a	Reject H_0	False	OK
b	Fail to reject H_0	True	OK
c	Reject H_0	True	Type I or alpha (α) error
d	Fail to reject H_0	False	Type II or beta (β) error



Decisions about hypotheses

Hypotheses	$p < \alpha$	$p > \alpha$
Null hypothesis (H_0)	Reject	Fail to reject
Alternative hypothesis (H_1)	Accept	Fail to accept

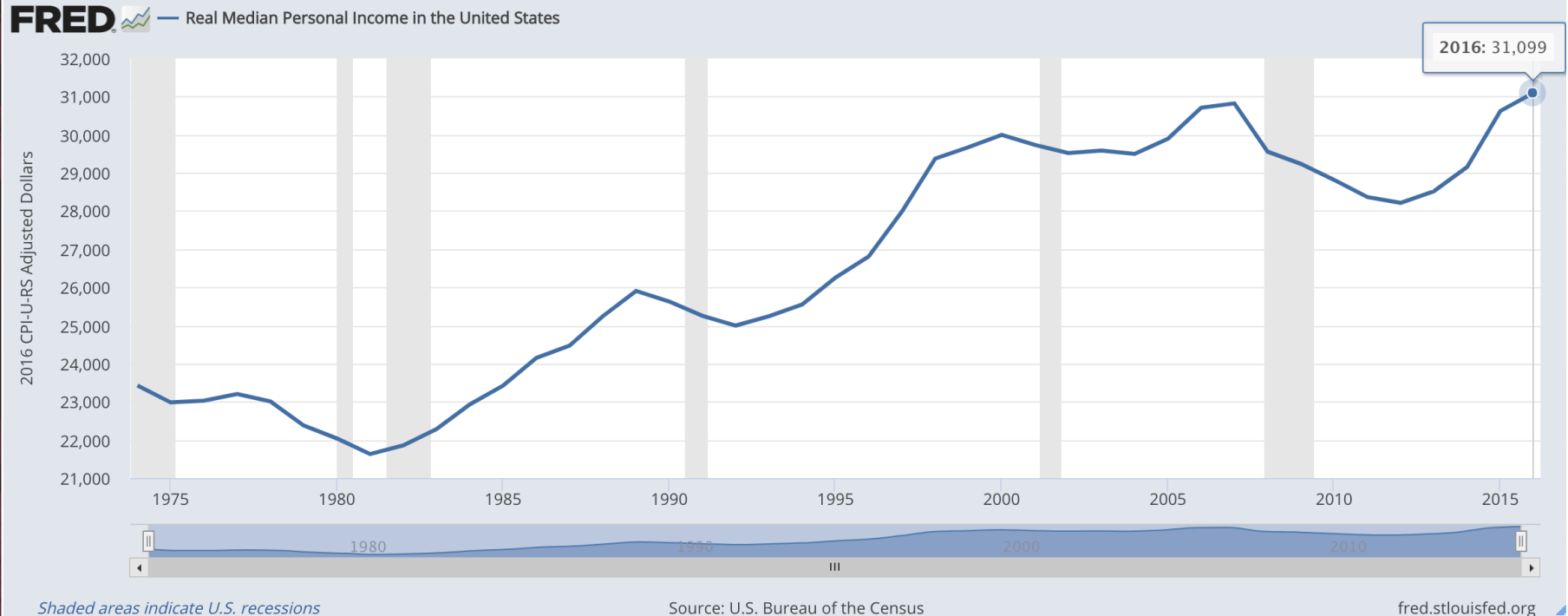
- **p -value** is the probability of failing to reject the null hypothesis
- If a statistical software gives only the two-tailed p -value, divide it by 2 to obtain the one-tailed p -value

Significance level (α)	Confidence level
0.10 (10%)	90%
0.05 (5%)	95%
0.01 (1%)	99%
0.001 (0.1%)	99.9%



Example 3: Income, 2016

- Is the average personal income of the adult population (18+) in the U.S. higher than among the population 15+?
- We know the income for the population 15+



Source: U.S. Bureau of the Census, Real Median Personal Income in the United States [MEPAINUSA672N], retrieved from FRED, Federal Reserve Bank of St. Louis; <https://fred.stlouisfed.org/series/MEPAINUSA672N>, October 17, 2017.



Example 3: Census & GSS

- We know the income for the GSS sample 18+

```
. mean conrinc
```

```
Mean estimation                Number of obs   =      1,632
```

	Mean	Std. Err.	[95% Conf. Interval]	
conrinc	34822.52	897.5571	33062.03	36583

- What causes the difference between \$31,099.00 (pop.15+, Census) and \$34,822.52 (sample 18+, GSS)?
- Real difference? Or difference due to random chance?



Example 3: Result

- Population 18+ has an average income that is significantly higher than the population 15+
 - The difference between the parameter \$31,099.00 and the statistic \$34,822.52 was large and unlikely to have occurred by random chance ($p < 0.05$)

```
. ztest conrinc=31099
```

One-sample z test

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
conrinc	1,632	34822.52	.0247537	1	34822.47	34822.56

mean = mean(**conrinc**)

z = **1.5e+05**

Ho: mean = **31099**

Ha: mean < **31099**
Pr(Z < z) = **1.0000**

Ha: mean != **31099**
Pr(|Z| > |z|) = **0.0000**

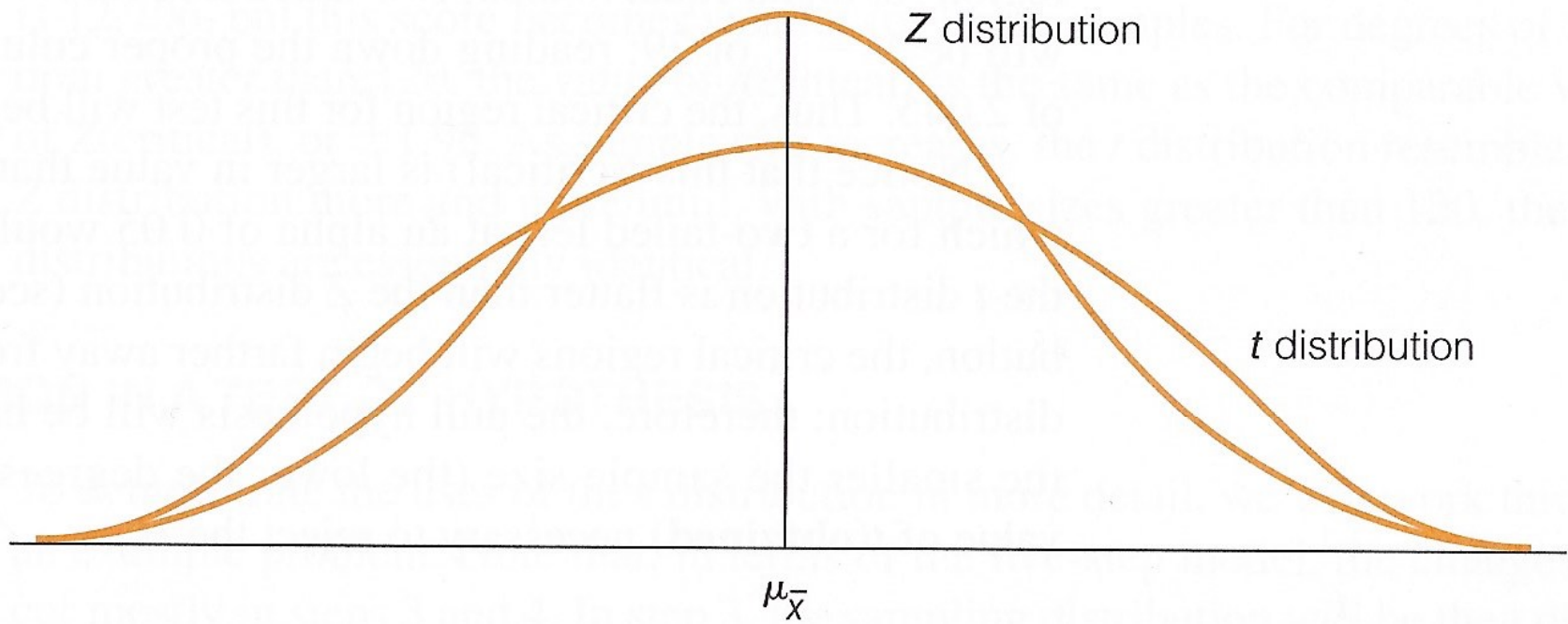
Ha: mean > **31099**
Pr(Z > z) = **0.0000**

The Student's t distribution

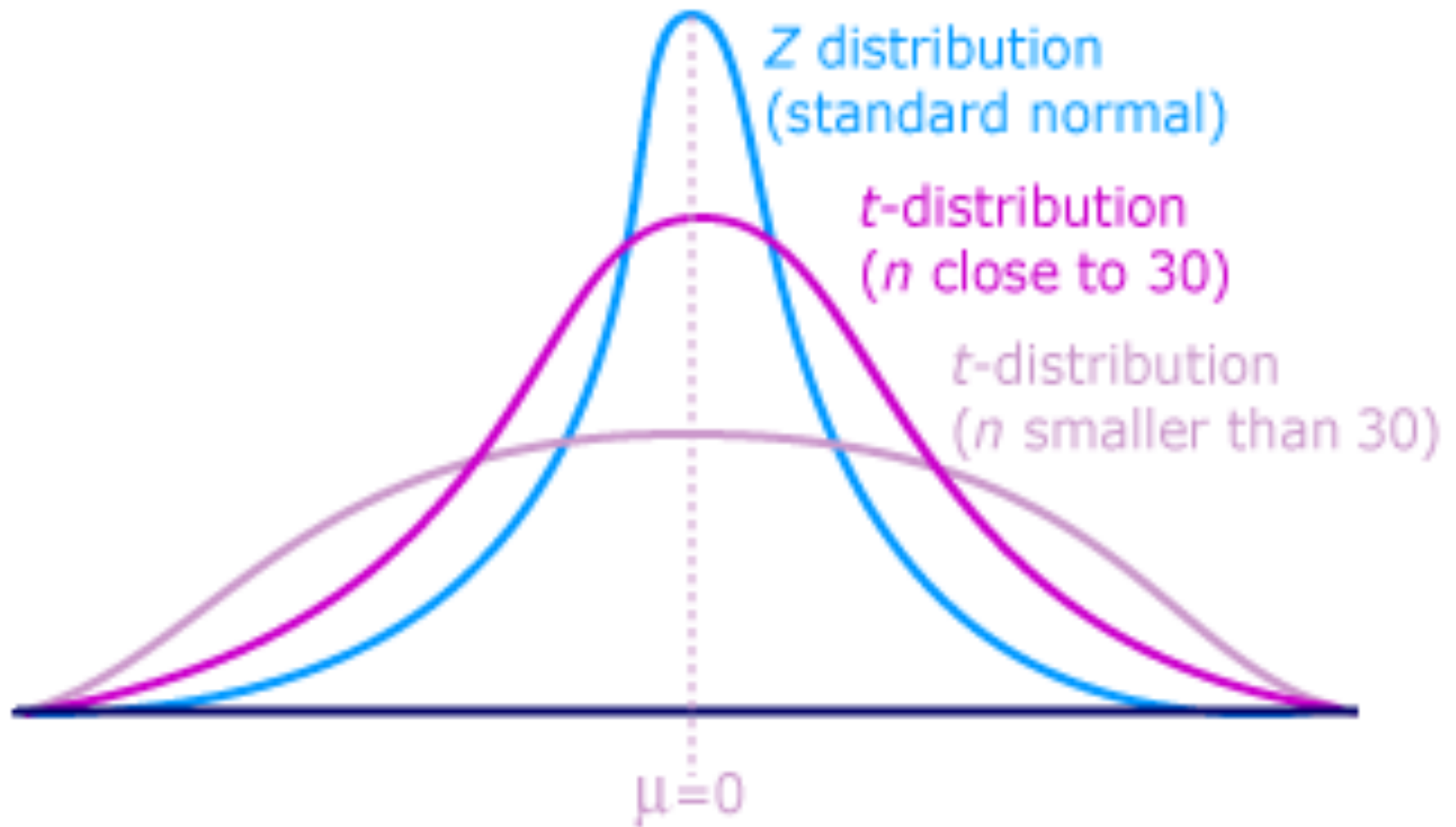
- How can we test a hypothesis when the population standard deviation (σ) is unknown, as is usually the case?
- For large samples ($N \geq 100$), we can use the sample standard deviation (s) as an estimator of the population standard deviation (σ)
 - Use standard normal distribution (Z)
- For small samples, s is too biased to estimate σ
 - Do not use standard normal distribution
 - Use Student's t distribution



t and Z distributions



t and Z distributions



Source: <https://joejeong33.wordpress.com/2013/06/03/t-distribution-in-the-normal-distribution-there-are-enough/>.

Choosing the distribution

- Choosing a sampling distribution when testing single-sample means for significance

If population standard deviation (σ) is	Sampling distribution is the
Known	Z distribution
Unknown and sample size (N) is large	Z distribution
Unknown and sample size (N) is small	t distribution



Example 4: With t -test

- This is the same as example 3, but with t -test
 - GSS has a large sample. This is just an illustration
- Population 18+ has an average income that is significantly higher than the population 15+ ($p < 0.05$)

```
. ttest conrinc=31099
```

One-sample t test

Variable	Obs	Mean	Std. Err.	Std. Dev.	[95% Conf. Interval]	
conrinc	1,632	34822.52	897.5571	36259.53	33062.03	36583

```
mean = mean(conrinc)                                t = 4.1485
Ho: mean = 31099                                    degrees of freedom = 1631
```

```
Ha: mean < 31099
Pr(T < t) = 1.0000
```

```
Ha: mean != 31099
Pr(|T| > |t|) = 0.0000
```

```
Ha: mean > 31099
Pr(T > t) = 0.0000
```


Five-step model for proportions

- When analyzing variables that are not measured at the interval-ratio level
 - A mean is inappropriate
 - We can test a hypothesis on a one sample proportion
- The five step model remains primarily the same, with the following changes
 - The assumptions are: random sampling, nominal level of measurement, and normal sampling distribution
 - The formula for Z is

$$Z = \frac{P_s - P_u}{\sqrt{P_u(1 - P_u)/N}}$$



Example 5: Proportions

- A random sample of 122 households in a low-income neighborhood revealed that 53 of the households were headed by women
 - $P_s = 53 / 122 = 0.43$
- In the city as a whole, the proportion of women-headed households (P_u) is 0.39
- Are households in lower-income neighborhoods significantly different from the city as a whole?
- Conduct a 90% hypothesis test ($\alpha = 0.10$)



Step 1: Assumptions, requirements

- Make assumptions
 - Random sampling
 - Hypothesis testing assumes samples were selected according to EPSEM
- Meet test requirements
 - The sample of 122 was randomly selected from all lower-income neighborhoods
 - Level of measurement is nominal
 - Women-headed households is measured as a proportion
 - Sampling distribution is normal in shape
 - This is a large sample ($N \geq 100$)



Step 2: Null hypothesis

- Null hypothesis, $H_0: P_u = 0.39$
 - The sample of 122 comes from a population where 39% of households are headed by women
 - The difference between 0.43 and 0.39 is trivial and caused by random chance
- Alternative hypothesis, $H_1: P_u \neq 0.39$
 - The sample of 122 comes from a population where the percent of women-headed households is not 39
 - The difference between 0.43 and 0.39 reflects an actual difference between lower-income neighborhoods and all neighborhoods

Step 3: Distribution, critical region

- Sampling distribution
 - Standard normal distribution (Z)
- Alpha (α) = 0.10 (two-tailed)
- Critical region begins at $Z(\text{critical}) = \pm 1.65$
 - This is the critical Z score associated with a two-tailed test and alpha equal to 0.10
 - If the obtained Z score falls in the critical region, we reject H_0



Step 4: Test statistic

- Proportion of households headed by women

City	Sample in a low-income neighborhood
$P_u = 0.39$	$P_s = 0.43$
	$N = 122$

- The formula for Z is

$$Z = \frac{P_s - P_u}{\sqrt{P_u(1 - P_u)/N}} = \frac{0.43 - 0.39}{\sqrt{0.39(1 - 0.39)/122}} = 0.91$$



Step 5: Decision, interpret

- $Z(\text{obtained}) = 0.91$
 - $Z(\text{obtained})$ did not fall in the critical region delimited by $Z(\text{critical}) = \pm 1.65$, so we **fail to reject** the H_0
 - This means that if H_0 was true, a sample outcome of 0.43 would be likely
 - Therefore, the H_0 is not false and cannot be rejected
- The population of women-headed households in lower-income neighborhoods is not significantly different from the city as a whole
 - The difference between the parameter ($P_u=0.39$) and the statistic ($P_s=0.43$) was small and likely to have occurred by random chance ($p>0.10$)



Example 6: Sex, 2016

- Is the female proportion of the adult population (18+) in the U.S. higher than among the overall population?
- We know the percentage of women for the population

Population estimates, July 1, 2016, (V2016)	323,127,513
PEOPLE	
Population	
Population estimates, July 1, 2016, (V2016)	323,127,513
Population estimates base, April 1, 2010, (V2016)	308,758,105
Population, percent change - April 1, 2010 (estimates base) to July 1, 2016, (V2016)	4.7%
Population, Census, April 1, 2010	308,745,538
Age and Sex	
Persons under 5 years, percent, July 1, 2016, (V2016)	6.2%
Persons under 5 years, percent, April 1, 2010	6.5%
Persons under 18 years, percent, July 1, 2016, (V2016)	22.8%
Persons under 18 years, percent, April 1, 2010	24.0%
Persons 65 years and over, percent, July 1, 2016, (V2016)	15.2%
Persons 65 years and over, percent, April 1, 2010	13.0%
Female persons, percent, July 1, 2016, (V2016)	50.8%
Female persons, percent, April 1, 2010	50.8%

Source: U.S. Census Bureau (<https://www.census.gov/quickfacts/fact/table/US/PST045216>).



Example 6: Census & GSS

- The percentage of women for the GSS sample 18+

```
. tab female
```

female	Freq.	Percent	Cum.
0	1,276	44.51	44.51
1	1,591	55.49	100.00
Total	2,867	100.00	

- What causes the difference between 50.8% (population, Census) and 55.5% (sample 18+, GSS)?
- Real difference? Or difference due to random chance?



Example 6: Result

- Population 18+ has a statistically significant higher proportion of women than overall population
 - The difference between the parameter 50.8% and the statistic 55.5% was large and unlikely to have occurred by random chance ($p < 0.05$)

```
. prtest female=.508
```

```
One-sample test of proportion
```

```
female: Number of obs = 2867
```

Variable	Mean	Std. Err.	[95% Conf. Interval]	
female	.5549355	.0092815	.536744	.5731269

```
p = proportion(female)
```

```
z = 5.0269
```

```
Ho: p = 0.508
```

```
Ha: p < 0.508
```

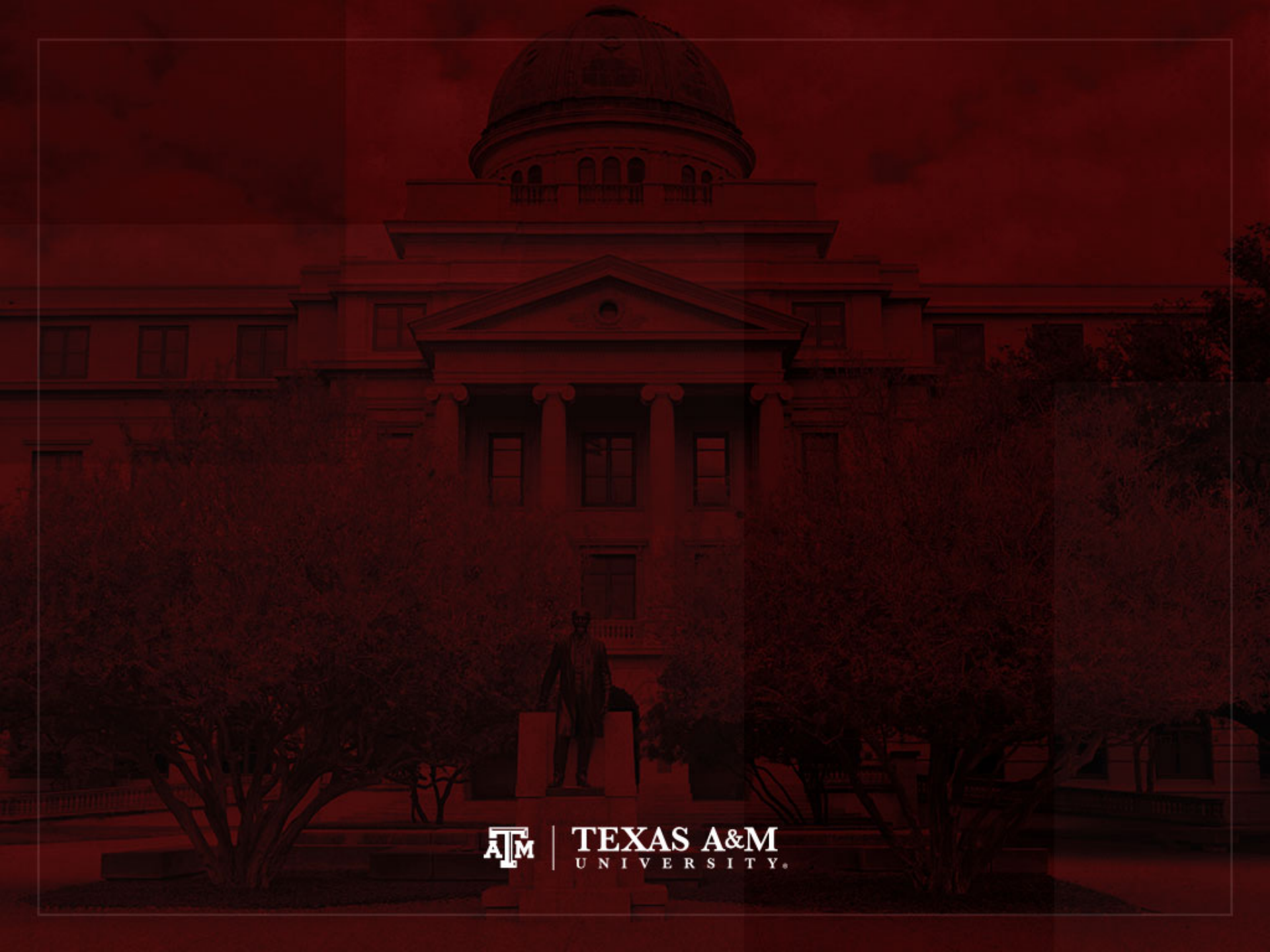
```
Pr(Z < z) = 1.0000
```

```
Ha: p != 0.508
```

```
Pr(|Z| > |z|) = 0.0000
```

```
Ha: p > 0.508
```

```
Pr(Z > z) = 0.0000
```



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