



# Hypothesis Testing

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# Nonstatistical Hypothesis Testing...

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In a trial a jury must decide between two hypotheses.

- $H_0$ : The null hypothesis - The defendant is innocent
- $H_1$ : The alternative hypothesis -The defendant is guilty

The jury does not know which hypothesis is true. They must make a decision on the basis of evidence presented.



# Research Hypothesis Testing...

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The FDA or “science” needs to decide on a new theory, drug, treatment...

- $H_0$ : The null hypothesis - the current theory, drug, treatment, is as good or better
- $H_1$ : The alternative hypothesis - the new theory, drug, treatment, should replace the old one

Researchers do not know which hypothesis is true. They must make a decision on the basis of evidence presented.



# Hypothesis Testing Terms

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- Convicting the defendant is *rejecting the null hypothesis in favor of the alternative hypothesis*.
- There is enough evidence to conclude that the defendant is guilty (i.e., there is enough evidence to support the alternative hypothesis).
- If the jury acquits it is stating that *there is not enough evidence to support the alternative hypothesis*.
- The jury is not saying that the defendant is innocent, only that there is not enough to convict.
- We don't accept the null hypothesis, we just don't have sufficient evidence to reject it.



# Hypothesis Testing...

- Two hypotheses, the null and the alternative
- Begins with the assumption that the null hypothesis is true.
- Is there enough evidence to infer that the alternative hypothesis is true, or the null is not likely to be true.
- There are two possible decisions:
  - Enough evidence to support the alternative hypothesis. Reject the null.
  - Not enough evidence to support the alternative hypothesis. Fail to reject the null.



# Example: Is the Coin Biased?

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- Flip a coin 10 times and get 2 HEADs
- Flip a coin 100 times and get 20 HEADs
- Flip a coin 100 times and get 37 HEADs
- Flip a coin 100 times and get 47 HEADs

We reject the prior assumption that the coin is unbiased if the number of HEADs is very unlikely under that assumption



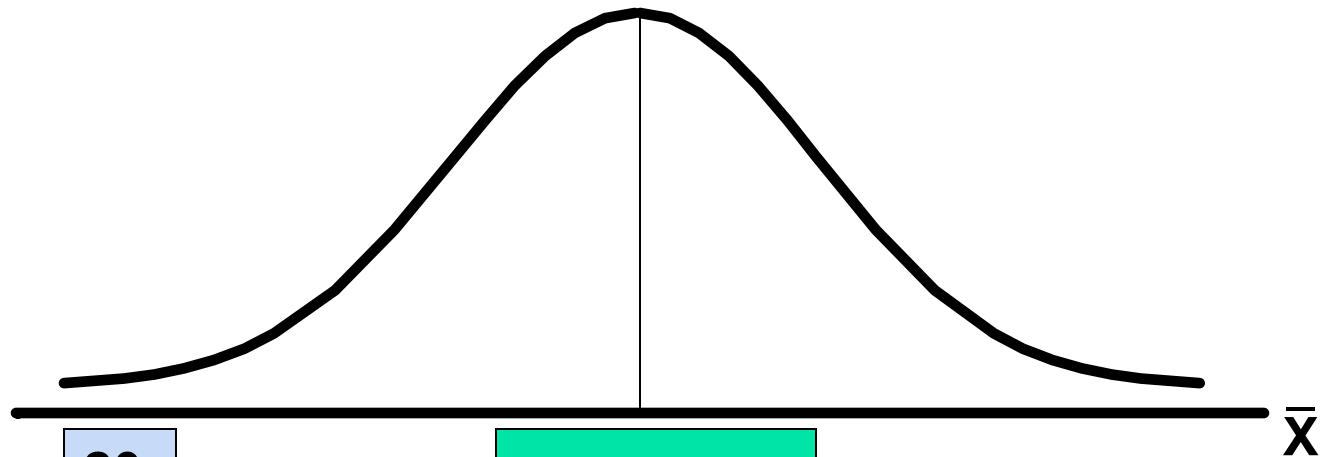
# Example: Is the Coin Biased?

- $H_0$  : Probability of HEAD =  $\frac{1}{2}$
- Possible alternatives:
  - $H_1$  : Probability of HEAD  $\neq \frac{1}{2}$
  - $H_1$  : Probability of HEAD is  $> \frac{1}{2}$
  - $H_1$  : Probability of HEAD is  $< \frac{1}{2}$



# Example: Is the Coin Biased?

Number of HEADs in 100 flipd



20

If it is unlikely that  
we get this value  
...

$\mu = 50$   
If  $H_0$  is true

... if in fact the coin was  
unbiased..

... Therefore  
we reject the  
null hypothesis  
that the coin is  
unbiased



# Level of Significance and the Rejection Region

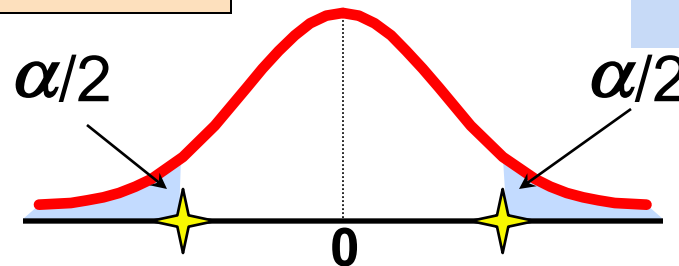
Level of significance =  $\alpha$

★ Represents critical value

$$H_0: \mu = 1/2$$

$$H_1: \mu \neq 1/2$$

Two-tail test

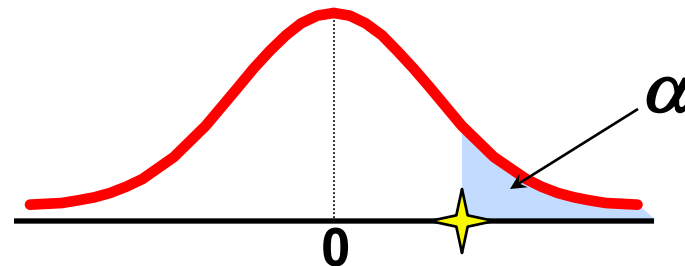


Rejection region is shaded

$$H_0: \mu \leq 1/2$$

$$H_1: \mu > 1/2$$

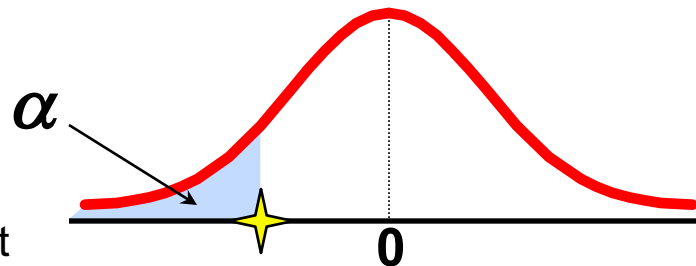
Upper-tail test



$$H_0: \mu \geq 1/2$$

$$H_1: \mu < 1/2$$

Lower-tail test





# What is a Hypothesis?

- A hypothesis is a claim (fact, model, parameter) that can be tested.



- population mean

**Example: The mean monthly cell phone bill of this city is  $\mu = \$42$**

- population proportion

**Example: The proportion of adults in this city with cell phones is  $p = .68$**



# The Null Hypothesis, $H_0$

- States the assumption (numerical) to be tested

**Example:** The average number of TV sets in U.S. Homes is equal to three ( $H_0 : \mu = 3$  )

- Is always about a model, not about a sample statistic

$$H_0 : \mu = 3$$

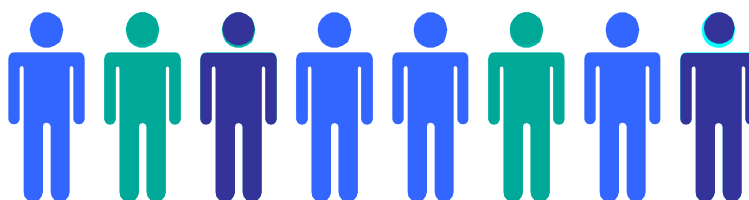
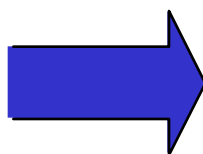
$$H_0 : \bar{X} = 3$$



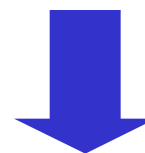


# Hypothesis Testing Process

**Claim:** the  
population  
mean age is 50.  
(Null Hypothesis:  
 $H_0: \mu = 50$ )

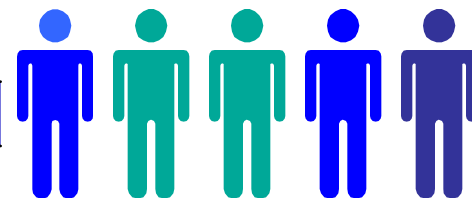
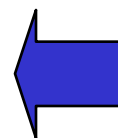


**Population**



Now select a  
random sample

Is  $\bar{X}=20$  likely if  $\mu = 50$ ?



**Sample**

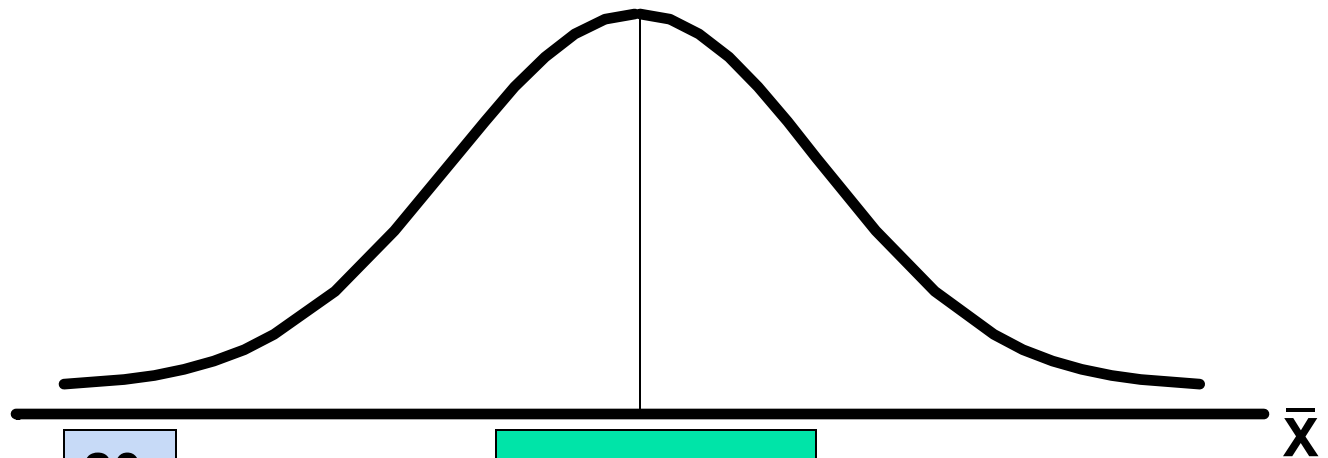
Suppose  
the sample  
mean age  
is 20:  $\bar{X} = 20$

If not likely,  
**REJECT**  
Null Hypothesis



# Reason for Rejecting $H_0$

## Sampling Distribution of $\bar{X}$



20

If it is unlikely that we would get a sample mean of this value ...

$\mu = 50$   
If  $H_0$  is true

... if in fact this were the population mean...

... then we reject the null hypothesis that  $\mu = 50$ .



# Level of Significance, $\alpha$

- **Defines the unlikely values of the sample statistic if the null hypothesis is true**
  - Defines **rejection region** of the sampling distribution
- Is designated by  **$\alpha$**  , (level of significance)
  - Typical values are .01, .05, or .10
- Is selected by the researcher at the beginning
- Provides the **critical value(s)** of the test

# Level of Significance and the Rejection Region

Level of significance =  $\alpha$

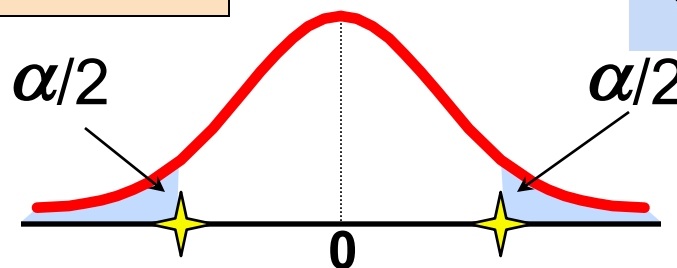
★ Represents critical value

Rejection region is shaded

$$H_0: \mu = 3$$

$$H_1: \mu \neq 3$$

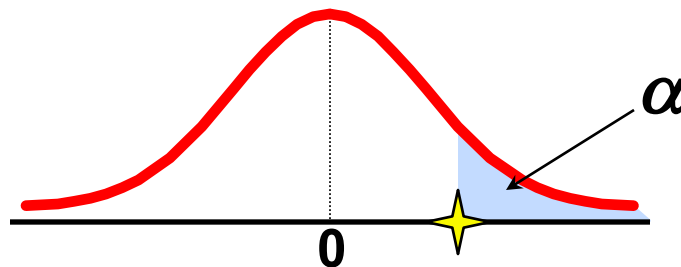
Two-tail test



$$H_0: \mu \leq 3$$

$$H_1: \mu > 3$$

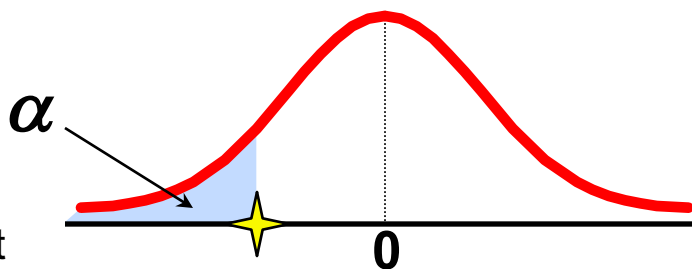
Upper-tail test



$$H_0: \mu \geq 3$$

$$H_1: \mu < 3$$

Lower-tail test





# Errors in Making Decisions

- **Type I Error**

- Reject a true null hypothesis
- Considered a serious type of error

The probability of Type I Error is  $\alpha$

- Called **level of significance** of the test
- Set by researcher in advance





# Errors in Making Decisions

*(continued)*

- **Type II Error**
  - Fail to reject a false null hypothesis

The probability of Type II Error is  $\beta$



# Outcomes and Probabilities

## Possible Hypothesis Test Outcomes


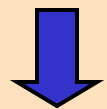
	Actual Situation	
Decision	$H_0$ True	$H_0$ False
Do Not Reject $H_0$	No error ( $1 - \alpha$ )	Type II Error ( $\beta$ )
Reject $H_0$	Type I Error ( $\alpha$ )	No Error ( $1 - \beta$ )

**Key:**  
**Outcome**  
**(Probability)**




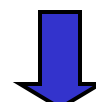

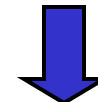



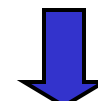
# Type I & II Error Relationship

- Type I and Type II errors can not happen at the same time
  - Type I error can only occur if  $H_0$  is **true**
  - Type II error can only occur if  $H_0$  is **false**

If Type I error probability (  $\alpha$  ) , then  
Type II error probability (  $\beta$  ) 



# Factors Affecting Type II Error

- All else equal,
  - $\beta$   when the difference between hypothesized parameter and its true value 
  
  - $\beta$   when  $\alpha$  
  
  - $\beta$   when  $\sigma$  
  
  - $\beta$   when  $n$  



# Power of the Test

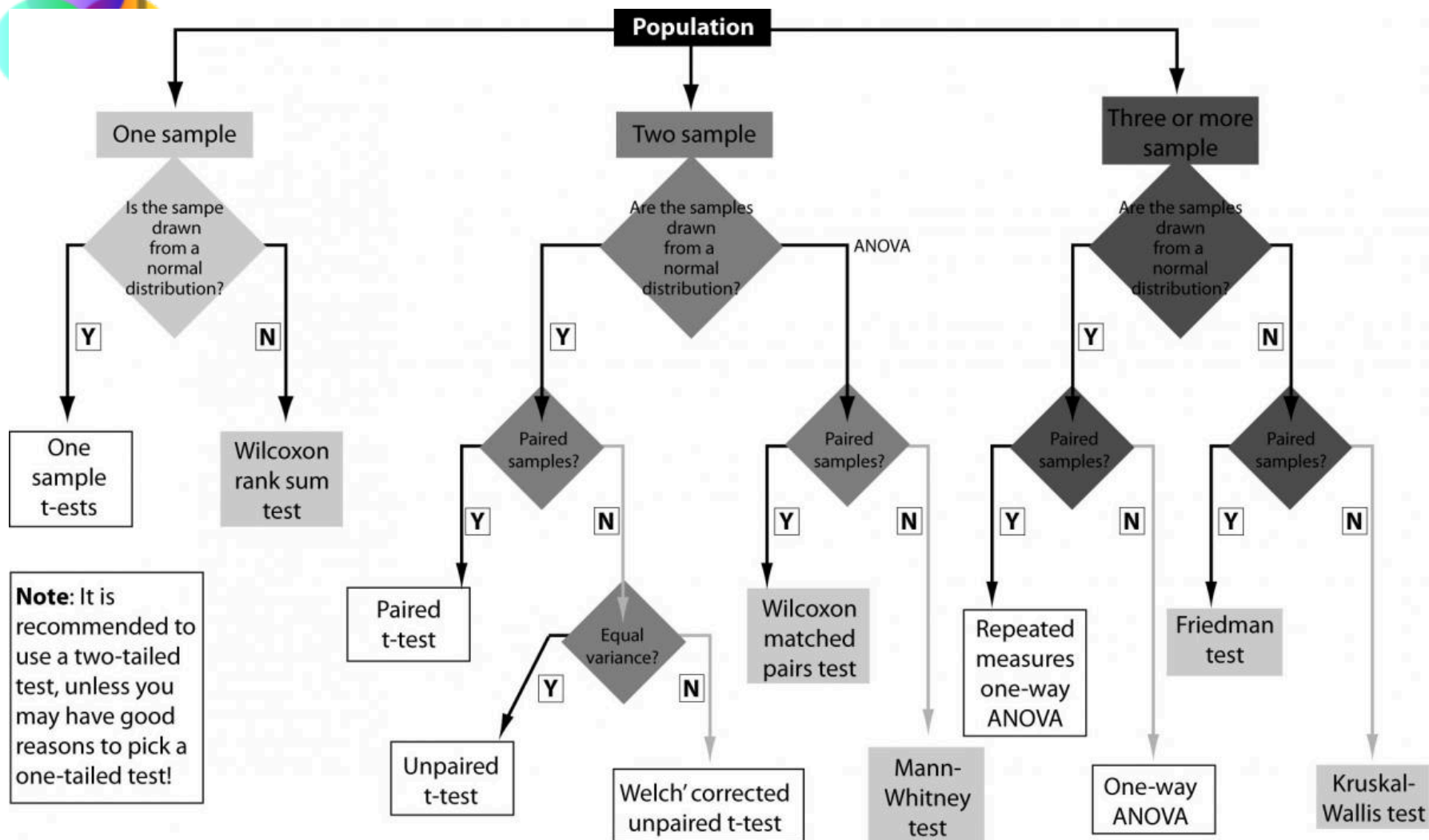
- The **power** of a test is the probability of rejecting a null hypothesis that is false
- i.e.,  $\text{Power} = P(\text{Reject } H_0 \mid H_1 \text{ is true})$ 
  - Power of the test increases as the sample size increases



# Select your test

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- Testing is a bit like finding the right recipe based on these ingredients:
  - Question
  - Data type
  - Sample size
  - Variance known? Variance of several groups equal?
- Good news: Plenty of tables available, e.g.,
  - [http://www.ats.ucla.edu/stat/mult\\_pkg/whatstat/default.htm](http://www.ats.ucla.edu/stat/mult_pkg/whatstat/default.htm) (with examples in R, SAS, Stata, SPSS)
  - [http://sites.stat.psu.edu/~ajw13/stat500\\_su\\_res/notes/lesson14/images/summary\\_table.pdf](http://sites.stat.psu.edu/~ajw13/stat500_su_res/notes/lesson14/images/summary_table.pdf)





# Example of a table of tests

**Summary Table for Statistical Techniques**

	Inference	Parameter	Statistic	Type of Data	Examples	Analysis	Minitab Command	Conditions
1	Estimating a Mean	One Population Mean $\mu$	Sample mean $\bar{y}$	Numerical	<ul style="list-style-type: none"> <li>What is the average weight of adults?</li> <li>What is the average cholesterol level of adult females?</li> </ul>	1-sample t-interval $\bar{y} \pm t_{\alpha/2} \frac{s}{\sqrt{n}}$	Stat >Basic statistics >1-sample t	<ul style="list-style-type: none"> <li>data approximately normal or</li> <li>have a large sample size (<math>n \geq 30</math>)</li> </ul>
2	Test about a Mean	One Population Mean $\mu$	Sample mean $\bar{y}$	Numerical	<ul style="list-style-type: none"> <li>Is the average GPA of juniors at Penn State higher than 3.0?</li> <li>Is the average Winter temperature in State College less than 42° F?</li> </ul>	$H_0: \mu = \mu_0$ $H_a: \mu \neq \mu_0$ or $H_a: \mu > \mu_0$ or $H_a: \mu < \mu_0$ The one sample t test: $t = \frac{\bar{y} - \mu_0}{s/\sqrt{n}}$	Stat >Basic statistics >1-sample t	<ul style="list-style-type: none"> <li>data approximately normal or</li> <li>have a large sample size (<math>n \geq 30</math>)</li> </ul>
3	Estimating a Proportion	One Population Proportion $\pi$	Sample Proportion $\hat{p}$	Categorical (Binary)	<ul style="list-style-type: none"> <li>What is the proportion of males in the world?</li> <li>What is the proportion of students that smoke?</li> </ul>	1-proportion Z-interval $\hat{p} \pm z_{\alpha/2} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$	Stat >Basic statistics >1-sample proportion	<ul style="list-style-type: none"> <li>have at least 5 in each category</li> </ul>
4	Test about a Proportion	One Population Proportion $\pi$	Sample Proportion $\hat{p}$	Categorical (Binary)	<ul style="list-style-type: none"> <li>Is the proportion of females different from 0.5?</li> <li>Is the proportion of students who fail Stat 500 less than 0.1?</li> </ul>	$H_0: \pi = \pi_0$ $H_a: \pi \neq \pi_0$ or $H_a: \pi > \pi_0$ or $H_a: \pi < \pi_0$ The one proportion Z-test: $z = \frac{\hat{p} - \pi_0}{\sqrt{\frac{\pi_0(1-\pi_0)}{n}}}$	Stat >Basic statistics >1-sample proportion	<ul style="list-style-type: none"> <li><math>n\pi_0 \geq 5</math> and <math>n(1-\pi_0) \geq 5</math></li> </ul>



# A common test: One-sample t-test



- **When:** Estimating a mean, comparing mean to a hypothetical value
- **Requirements:** Data approx. normal or sample size  $> 30$

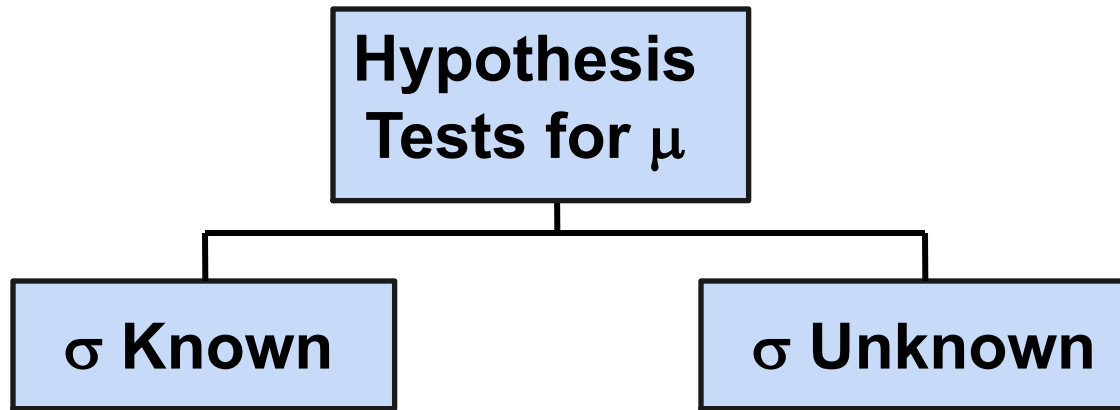
- **Setup:**


$\begin{array}{l} H_0 : \mu \leq \mu_0 \\ H_1 : \mu > \mu_0 \end{array}$	$\begin{array}{l} H_0 : \mu = \mu_0 \\ H_1 : \mu \neq \mu_0 \end{array}$	$\begin{array}{l} H_0 : \mu \geq \mu_0 \\ H_1 : \mu < \mu_0 \end{array}$
$T = \sqrt{n} \frac{\bar{X} - \mu_0}{S} \sim t_{n-1}$		
$t = \sqrt{n} \frac{\bar{x} - \mu_0}{s} \quad \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad s = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$		

- **Evaluation:** Compare t-statistic (table, excel) to your value and accept / reject null hypothesis



# Hypothesis Tests for the Mean





# Test of Hypothesis for the Mean ( $\sigma$ Known)

- Convert sample result ( $\bar{x}$ ) to a **z value**

## Hypothesis Tests for $\mu$

**$\sigma$  Known**

**$\sigma$  Unknown**

Consider the test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu > \mu_0$$

(Assume the population is normal)

The **decision rule** is:

$$\text{Reject } H_0 \text{ if } z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} > z_{\alpha}$$



# Decision Rule

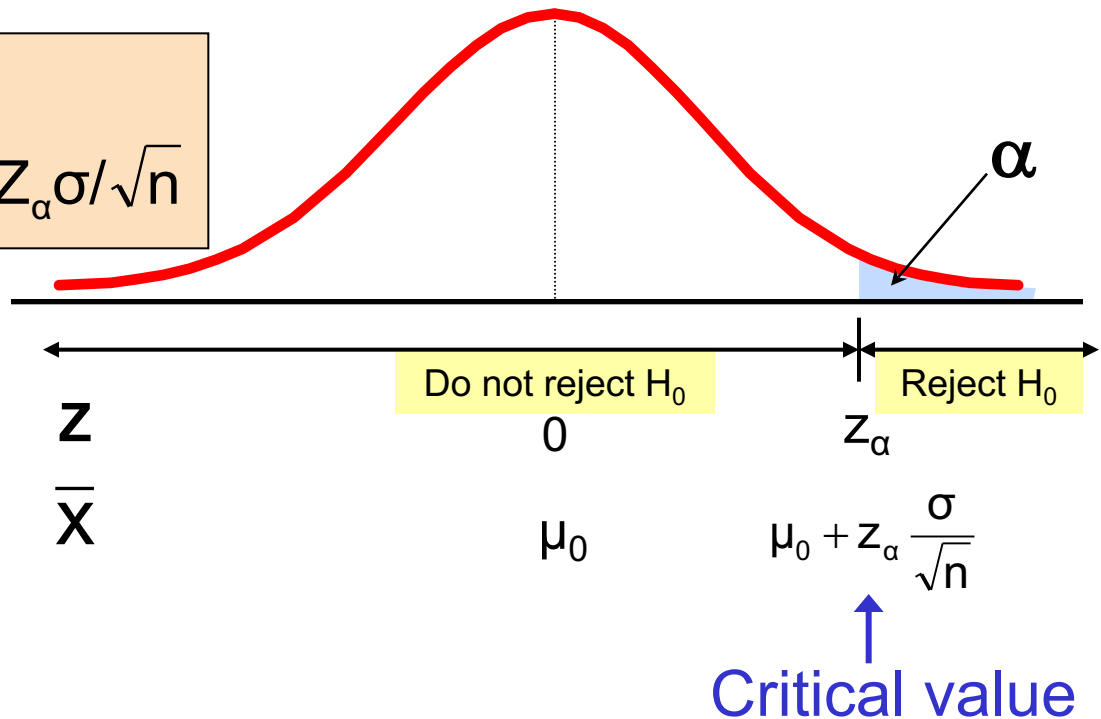
Reject  $H_0$  if  $z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} > z_\alpha$

$$H_0: \mu = \mu_0$$

$$H_1: \mu > \mu_0$$

Alternate rule:

Reject  $H_0$  if  $\bar{X} > \mu_0 + z_\alpha \sigma / \sqrt{n}$





# p-Value Approach to Testing

- **p-value**: Probability of obtaining a test statistic more extreme ( $\leq$  or  $\geq$ ) than the observed sample value **given  $H_0$  is true**
  - Also called **observed level of significance**
  - Smallest value of  $\alpha$  for which  $H_0$  can be rejected



# p-Value Approach to Testing

(continued)

- Convert sample result (e.g.,  $\bar{x}$ ) to test statistic (e.g., z statistic)

- Obtain the **p-value**

- For an upper tail test:

$$\begin{aligned} \text{p-value} &= P\left(Z > \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}}, \text{ given that } H_0 \text{ is true}\right) \\ &= P\left(Z > \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} \mid \mu = \mu_0\right) \end{aligned}$$

- **Decision rule:** compare the **p-value** to  $\alpha$

- If  $\text{p-value} < \alpha$ , reject  $H_0$
- If  $\text{p-value} \geq \alpha$ , do not reject  $H_0$



# Example: Upper-Tail Z Test for Mean ( $\sigma$ Known)

A phone industry manager thinks that customer monthly cell phone bill have increased, and now average over \$52 per month. The company wishes to test this claim. (Assume  $\sigma = 10$  is known)



Form hypothesis test:

$H_0: \mu \leq 52$	the average is not over \$52 per month
$H_1: \mu > 52$	the average <b>is</b> greater than \$52 per month (i.e., sufficient evidence exists to support the manager's claim)

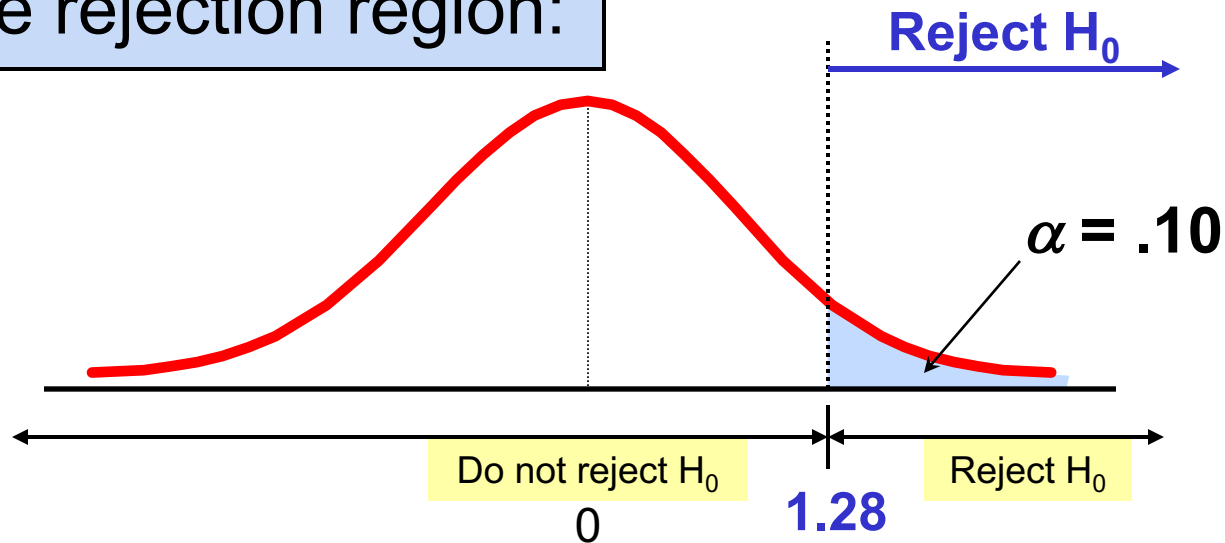


# Example: Find Rejection Region

(continued)

- Suppose that  $\alpha = .10$  is chosen for this test

Find the rejection region:



$$\text{Reject } H_0 \text{ if } z = \frac{\bar{x} - \mu_0}{\sigma/\sqrt{n}} > 1.28$$







# Example: Sample Results

(continued)

Obtain sample and compute the test statistic

Suppose a sample is taken with the following results:  $n = 64$ ,  $\bar{x} = 53.1$  ( $\sigma=10$  was assumed known)

- Using the sample results,

$$z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} = \frac{53.1 - 52}{\frac{10}{\sqrt{64}}} = 0.88$$

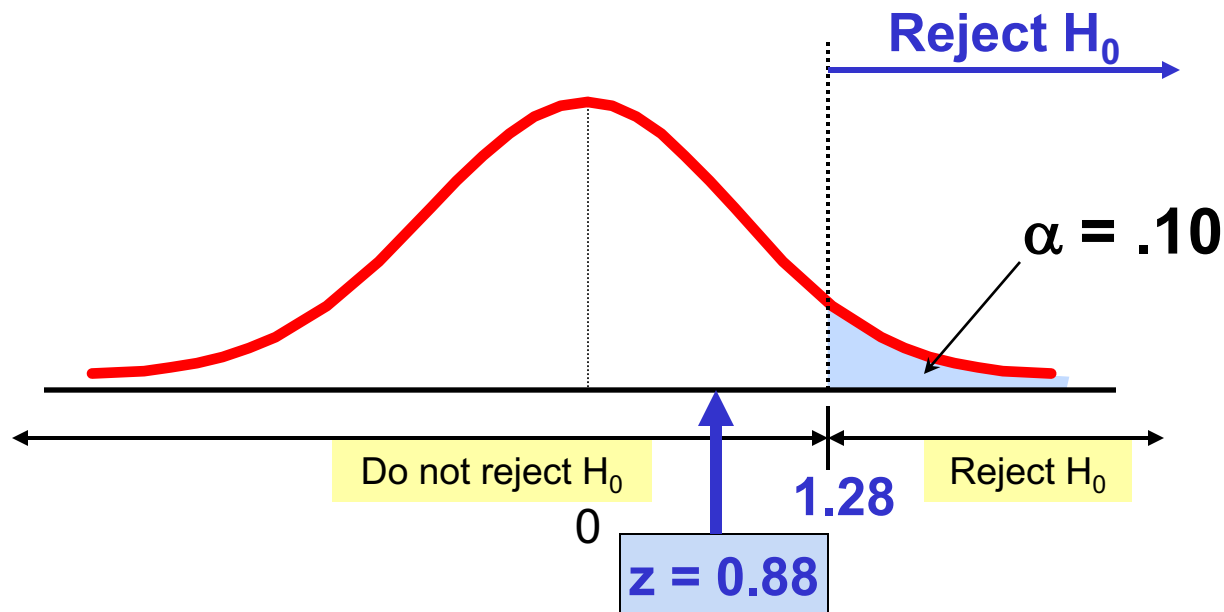




# Example: Decision

(continued)

Reach a decision and interpret the result:



**Do not reject  $H_0$  since  $z = 0.88 < 1.28$**

i.e.: there is not sufficient evidence that the mean bill is over \$52

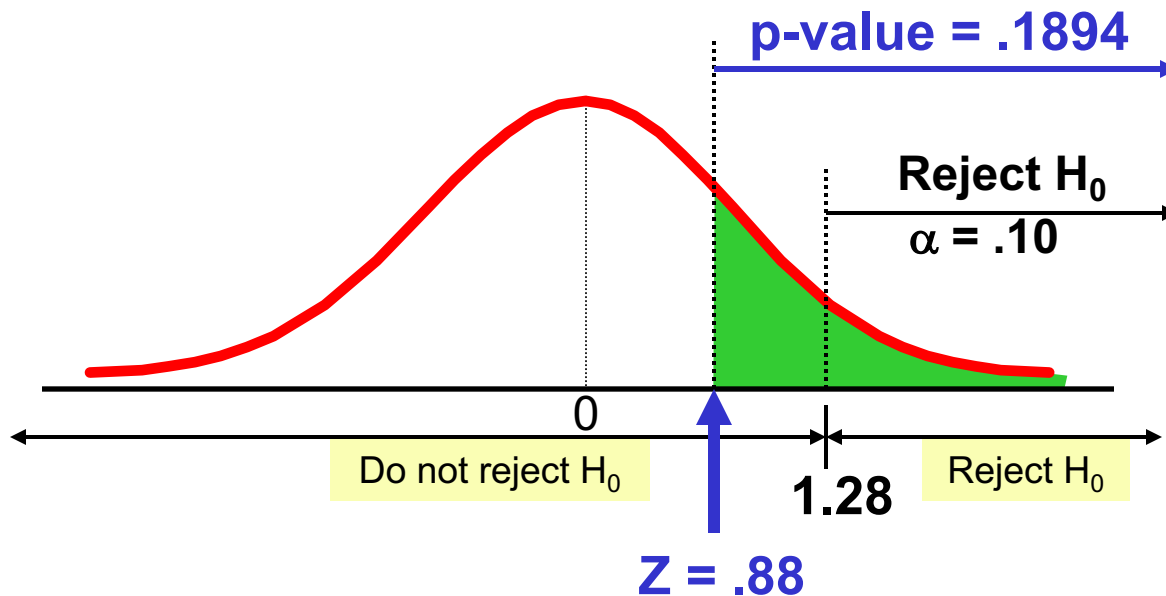




# Example: p-Value Solution

(continued)

Calculate the p-value and compare to  $\alpha$   
(assuming that  $\mu = 52.0$ )



$$P(\bar{x} \geq 53.1 | \mu = 52.0)$$

$$= P\left(z \geq \frac{53.1 - 52.0}{10/\sqrt{64}}\right)$$

$$= P(z \geq 0.88) = 1 - .8106$$

$$= .1894$$

**Do not reject  $H_0$  since  $p\text{-value} = .1894 > \alpha = .10$**

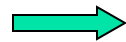


# One-Tail Tests

- In many cases, the alternative hypothesis focuses on one particular direction

$$H_0: \mu \leq 3$$

$$H_1: \mu > 3$$



This is an **upper**-tail test since the alternative hypothesis is focused on the upper tail above the mean of 3

$$H_0: \mu \geq 3$$

$$H_1: \mu < 3$$

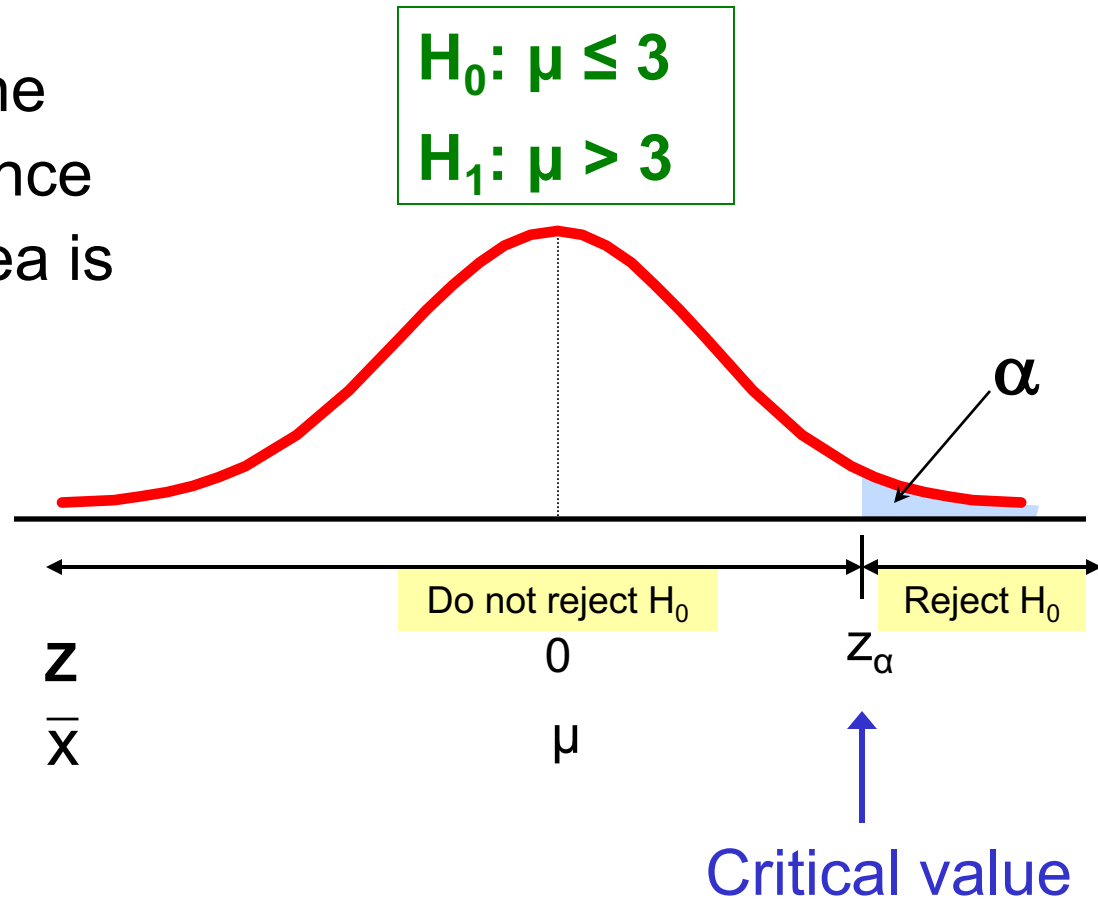


This is a **lower**-tail test since the alternative hypothesis is focused on the lower tail below the mean of 3



# Upper-Tail Tests

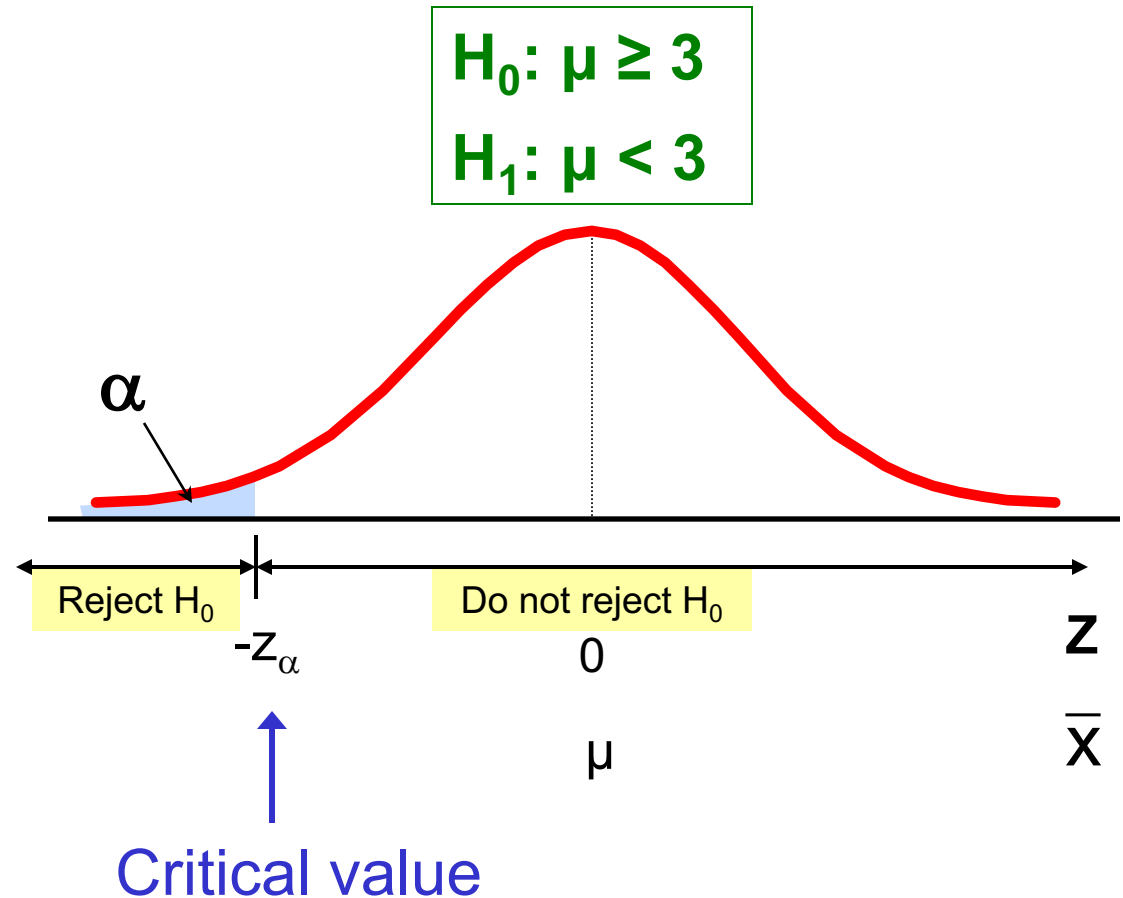
- There is only one critical value, since the rejection area is in only one tail





# Lower-Tail Tests

- There is only one critical value, since the rejection area is in only one tail



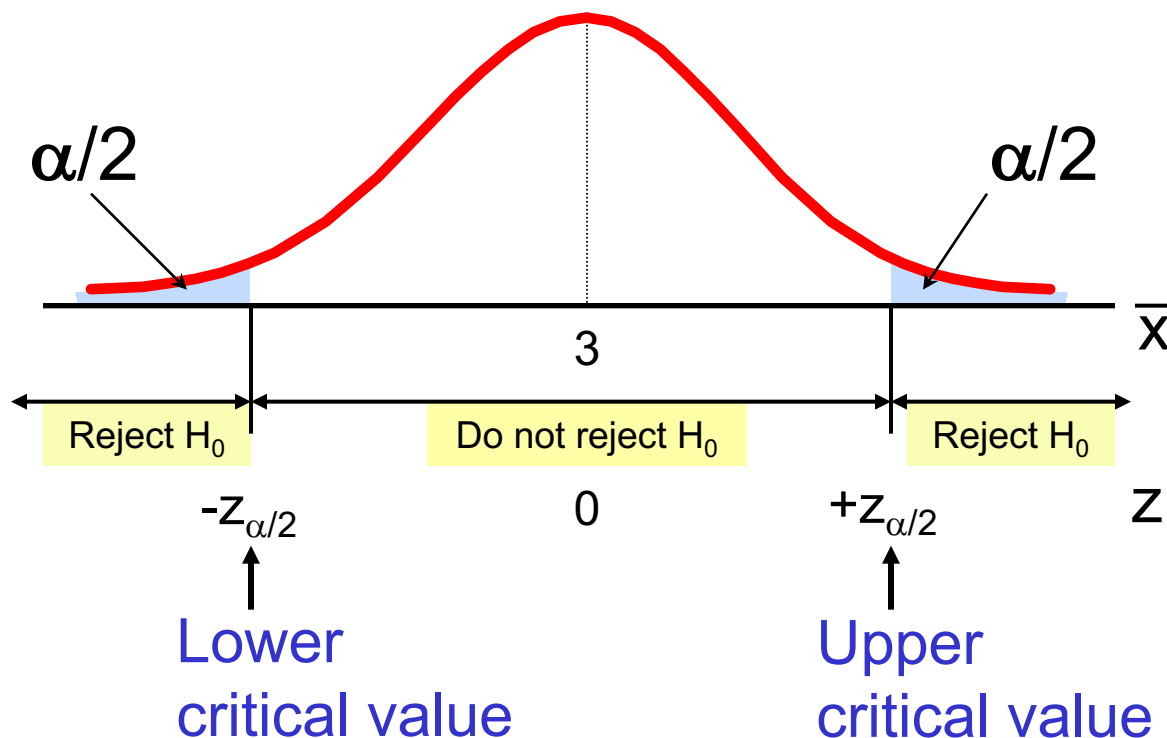


# Two-Tail Tests

- In some settings, the alternative hypothesis does not specify a unique direction

$$\begin{aligned} H_0: \mu &= 3 \\ H_1: \mu &\neq 3 \end{aligned}$$

- There are two critical values, defining the two regions of rejection

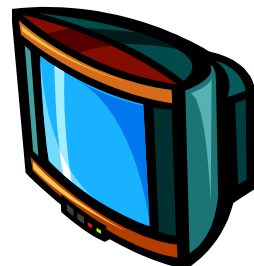




# Hypothesis Testing Example

**Test the claim that the true mean # of TV sets in US homes is equal to 3.  
(Assume  $\sigma = 0.8$ )**

- State the appropriate null and alternative hypotheses
  - $H_0: \mu = 3$  ,  $H_1: \mu \neq 3$  (This is a two tailed test)
- Specify the desired level of significance
  - Suppose that  $\alpha = .05$  is chosen for this test
- Choose a sample size
  - Suppose a sample of size  $n = 100$  is selected







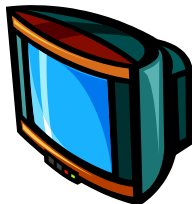
# Hypothesis Testing Example

(continued)

- Determine the appropriate technique
  - $\sigma$  is known so this is a  $z$  test
- Set up the critical values
  - For  $\alpha = .05$  the critical  $z$  values are  $\pm 1.96$
- Collect the data and compute the test statistic
  - Suppose the sample results are  
 $n = 100$ ,  $\bar{x} = 2.84$  ( $\sigma = 0.8$  is assumed known)

So the test statistic is:

$$z = \frac{\bar{X} - \mu_0}{\frac{\sigma}{\sqrt{n}}} = \frac{2.84 - 3}{\frac{0.8}{\sqrt{100}}} = \frac{-.16}{.08} = -2.0$$



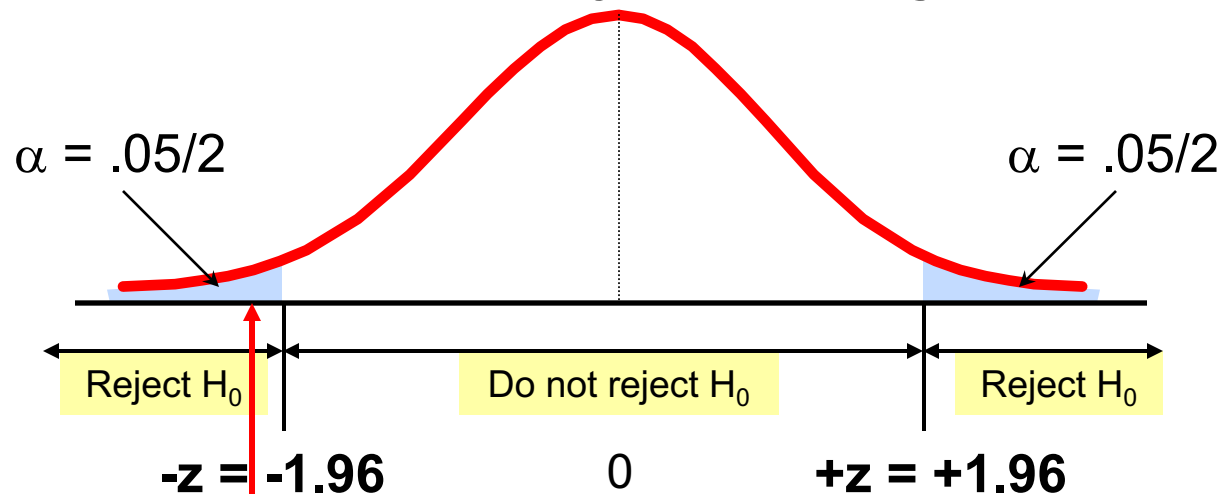


# Hypothesis Testing Example

(continued)

- Is the test statistic in the rejection region?

Reject  $H_0$  if  
 $z < -1.96$  or  
 $z > 1.96$ ;  
otherwise  
do not  
reject  $H_0$



Here,  $z = -2.0 < -1.96$ , so the test statistic is in the rejection region

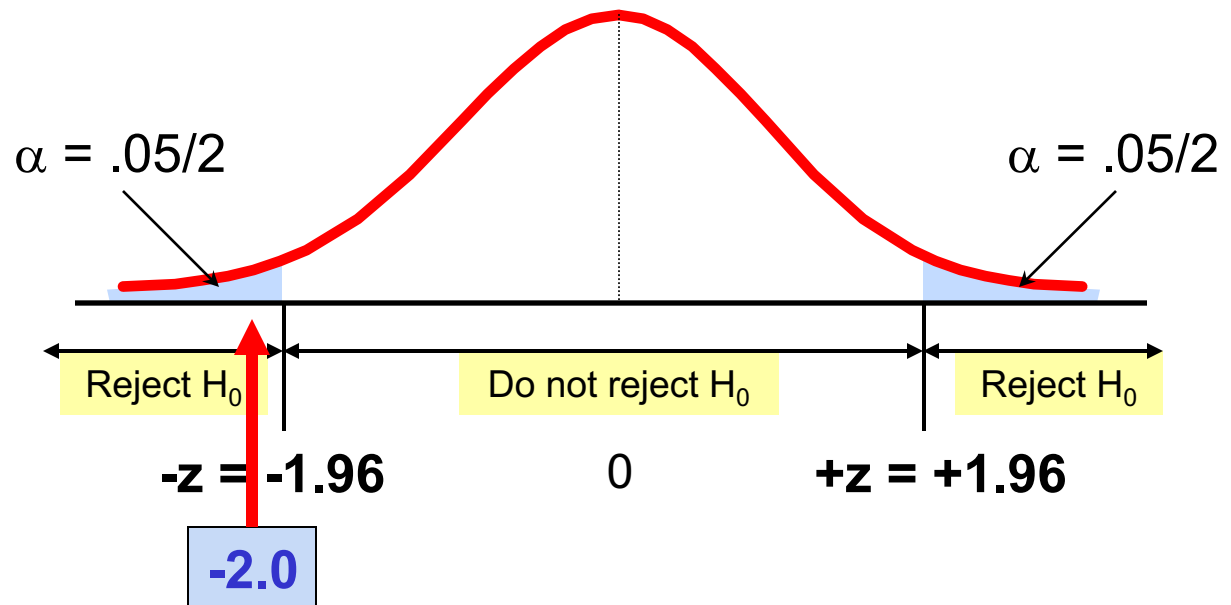




# Hypothesis Testing Example

(continued)

- Reach a decision and interpret the result



Since  $z = -2.0 < -1.96$ , we reject the null hypothesis and conclude that there is sufficient evidence that the mean number of TVs in US homes is not equal to 3





# Example: p-Value

- **Example:** How likely is it to see a sample mean of 2.84 (or something further from the mean, in either direction) if the true mean is  $\mu = 3.0$ ?

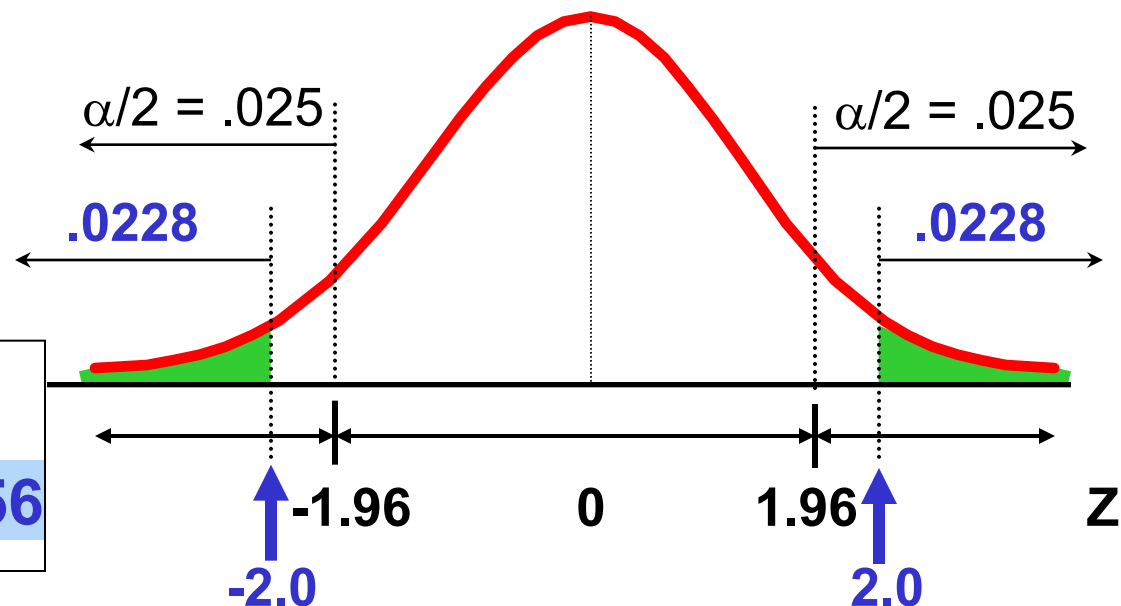
$\bar{x} = 2.84$  is translated to a z score of  $z = -2.0$

$$P(z < -2.0) = .0228$$

$$P(z > 2.0) = .0228$$

**p-value**

$$= .0228 + .0228 = .0456$$





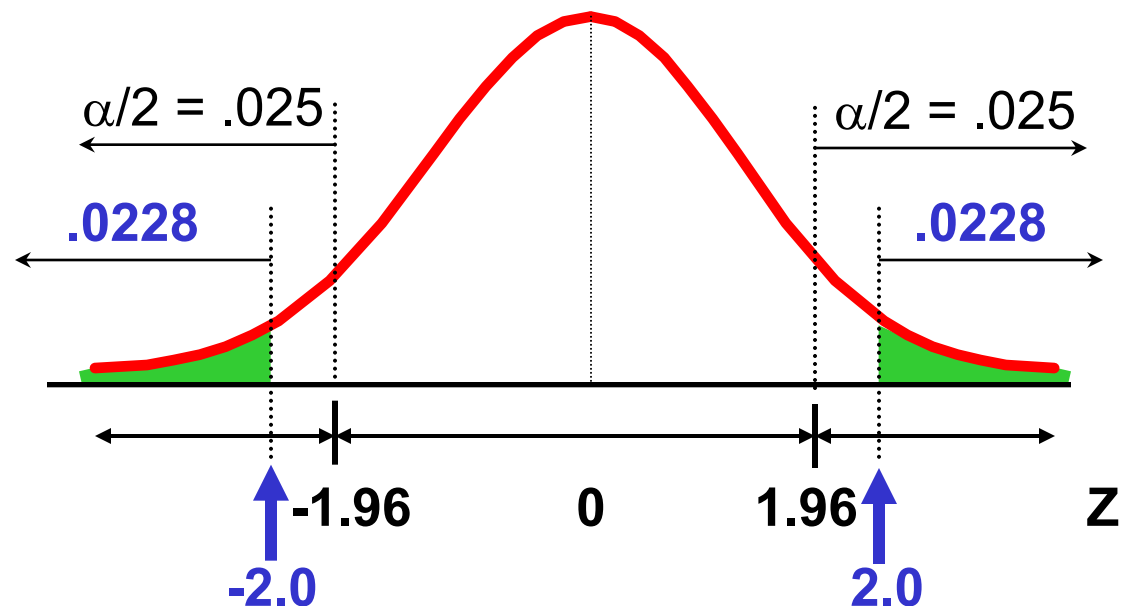
# Example: p-Value

(continued)

- Compare the p-value with  $\alpha$ 
  - If p-value  $< \alpha$ , reject  $H_0$
  - If p-value  $\geq \alpha$ , do not reject  $H_0$

Here: p-value = .0456  
 $\alpha = .05$

Since .0456  $< .05$ , we  
reject the null  
hypothesis



# An Alternative Way

## Beer Consumption XXXXXXXXXX Human Attractiveness to Malaria Mosquitoes

### Beer (25):

27 20 21 26 27 31 24 21 20 19  
23 24 28 19 24 29 18 20 17 31  
20 25 28 21 27

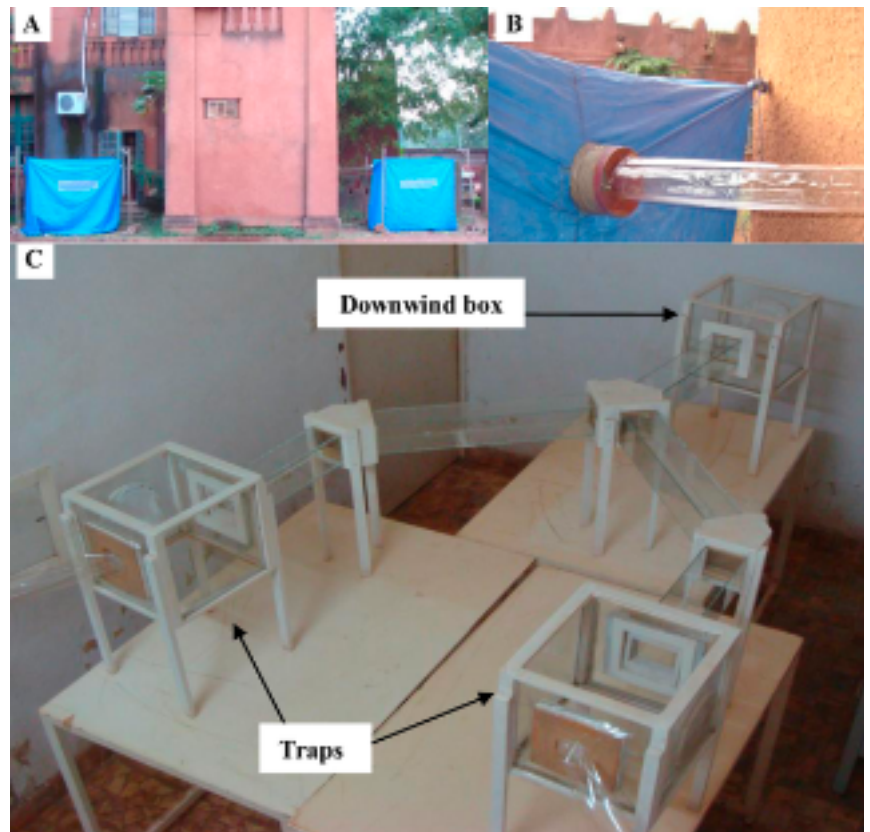
Mean: 23.6

### Water (18):

21 22 15 12 21 16 19 15 22 24  
19 23 13 22 20 24 18 20

Mean: 19.2

**Is a difference of 4.4 significant?**



# Permutation Test

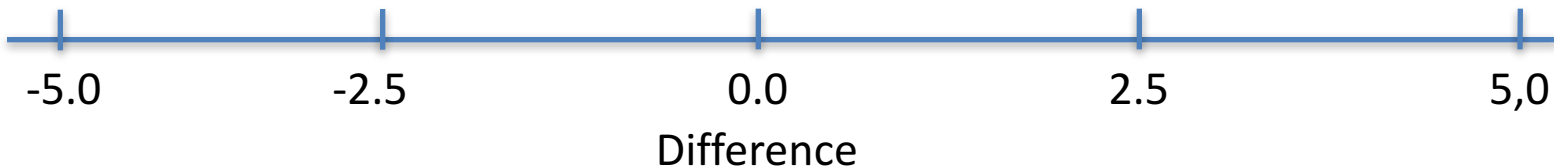
Beer (25)

27	23	20	31	29
20	24	25	24	18
21	28	28	21	20
26	19	21	20	17
27	24	27	19	31

Water (18)

21	19	16	24
22	23	19	18
15	13	15	20
12	22	22	
21	20	24	

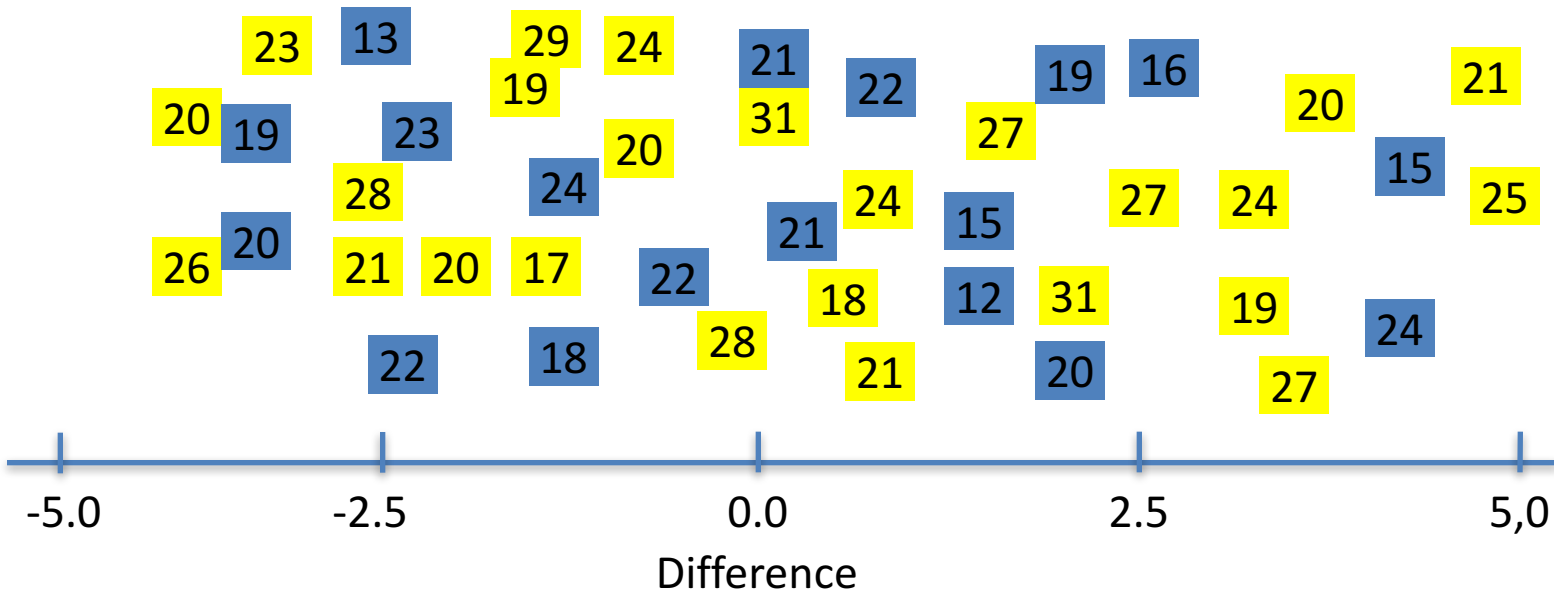
Difference: 4.4



# Permutation Test

Beer (25)

Water (18)

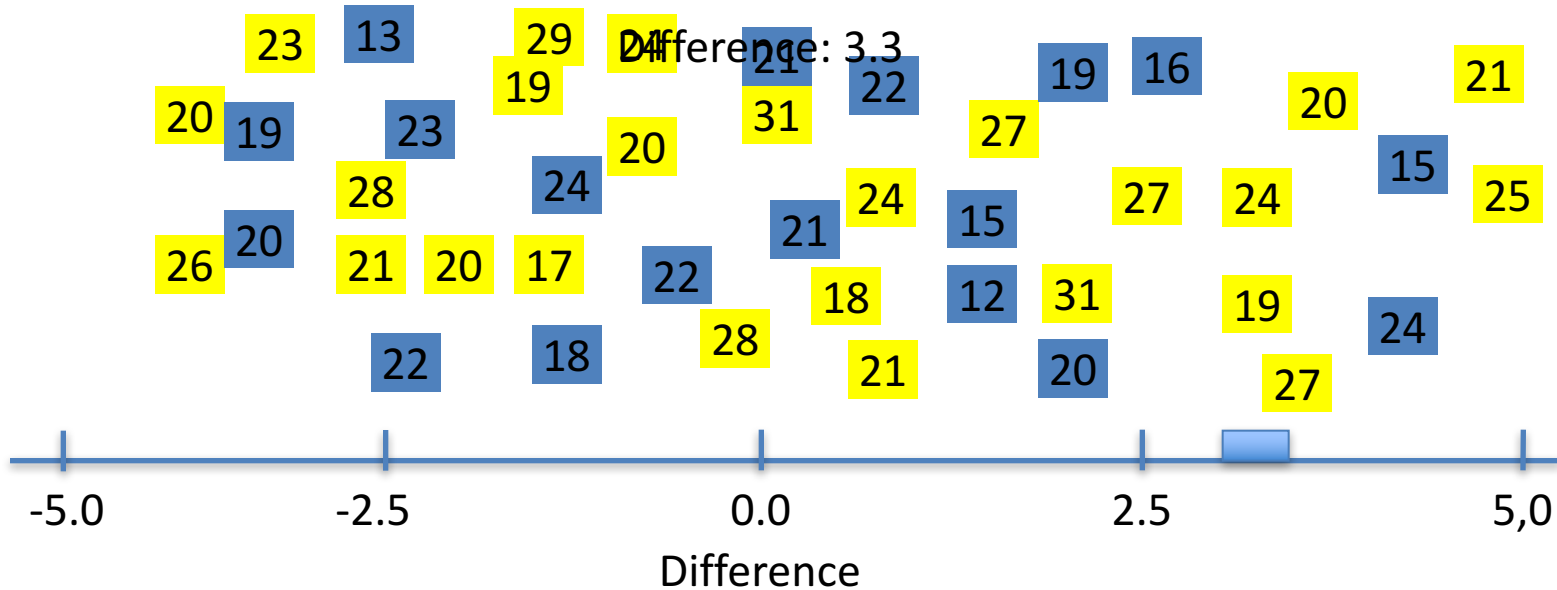




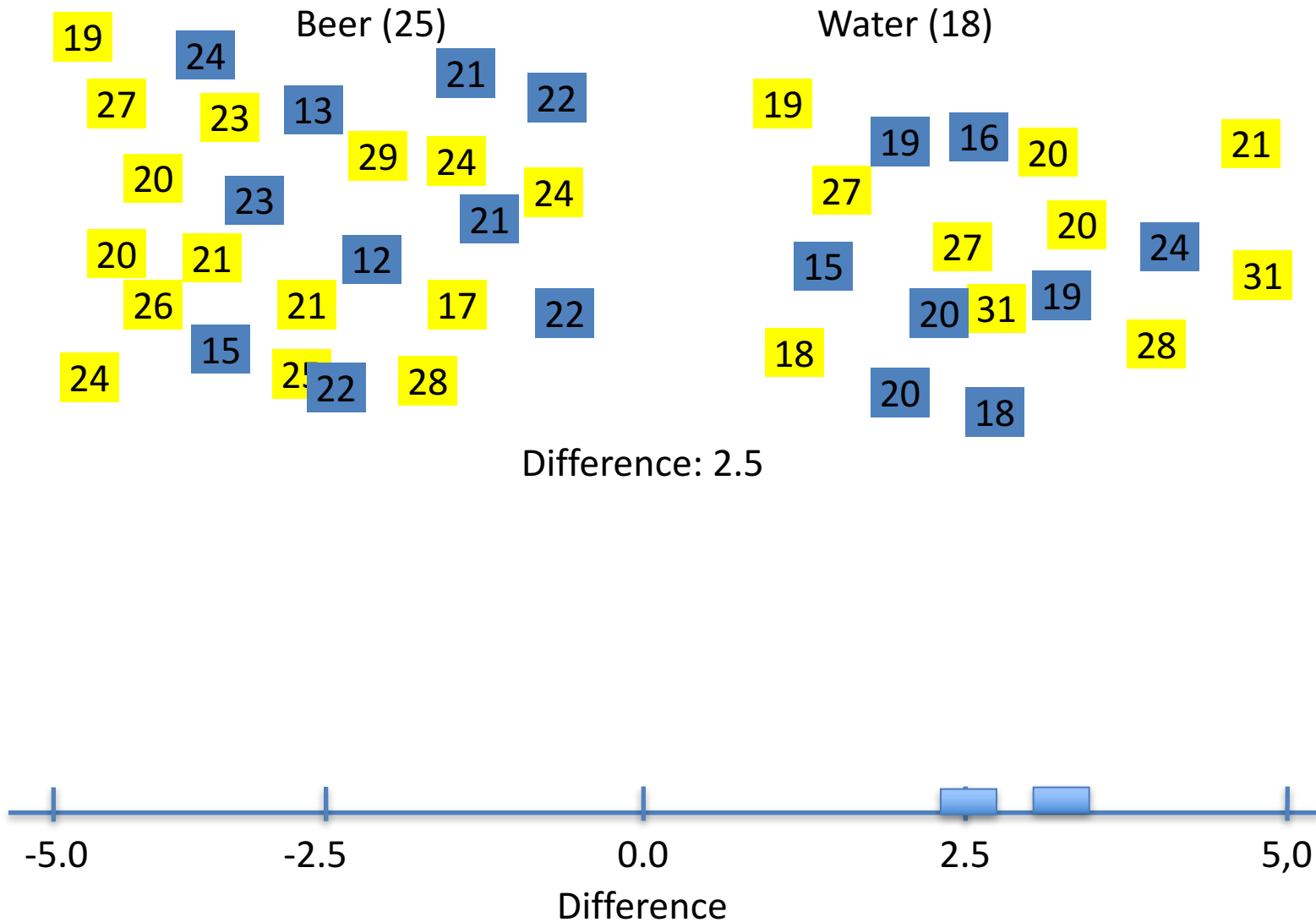
# Permutation Test

Beer (25)

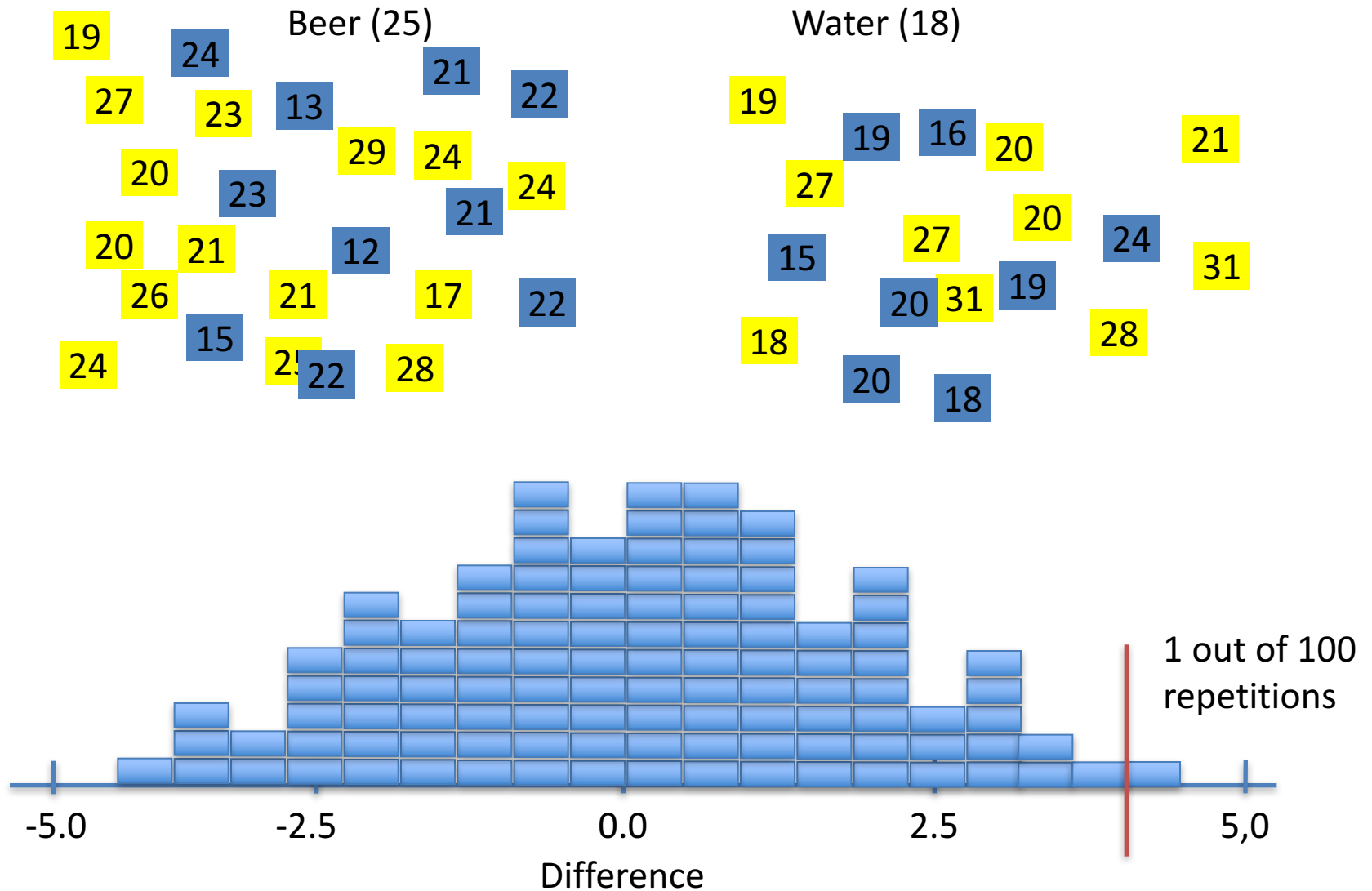
Water (18)



# Permutation Test



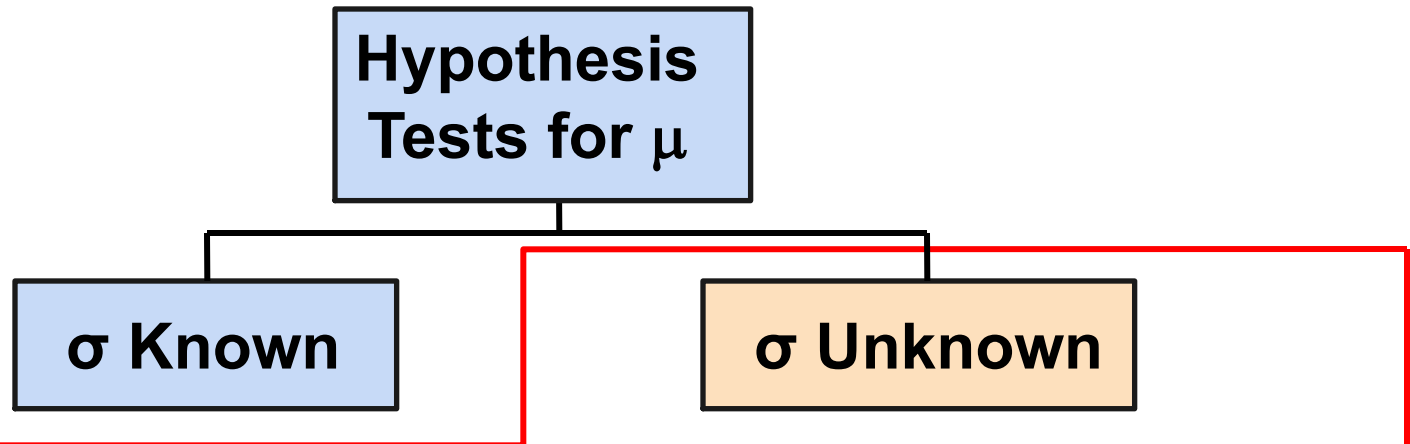
# Permutation Test





# t Test of Hypothesis for the Mean ( $\sigma$ Unknown)

- Convert sample result ( $\bar{x}$ ) to a **t test statistic**



Consider the test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu > \mu_0$$

(Assume the population is normal)

The **decision rule** is:

$$\text{Reject } H_0 \text{ if } t = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}} > t_{n-1, \alpha}$$



# t Test of Hypothesis for the Mean ( $\sigma$ Unknown)

*(continued)*

- For a two-tailed test:

Consider the test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu \neq \mu_0$$

(Assume the population is normal,  
and the population variance is  
unknown)

The **decision rule** is:

Reject  $H_0$  if

$$t = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}} < -t_{n-1, \alpha/2}$$

or if

$$t = \frac{\bar{x} - \mu_0}{\frac{s}{\sqrt{n}}} > t_{n-1, \alpha/2}$$



# What is a T-distribution?

---

- A t-distribution is like a Z distribution, except has slightly fatter tails to reflect the uncertainty added by estimating  $\sigma$ .
- The bigger the sample size (i.e., the bigger the sample size used to estimate  $\sigma$ ), then the closer t becomes to Z.
- If  $n > 100$ , t approaches Z.

# The T probability density function



$$f(x) = \frac{\Gamma\left(\frac{v+1}{2}\right)}{\Gamma\left(\frac{v}{2}\right)} \frac{1}{\sqrt{v\pi}} \left(1 + \frac{x^2}{v}\right)^{-\frac{v+1}{2}}$$

Where:

$v$  is the degrees of freedom

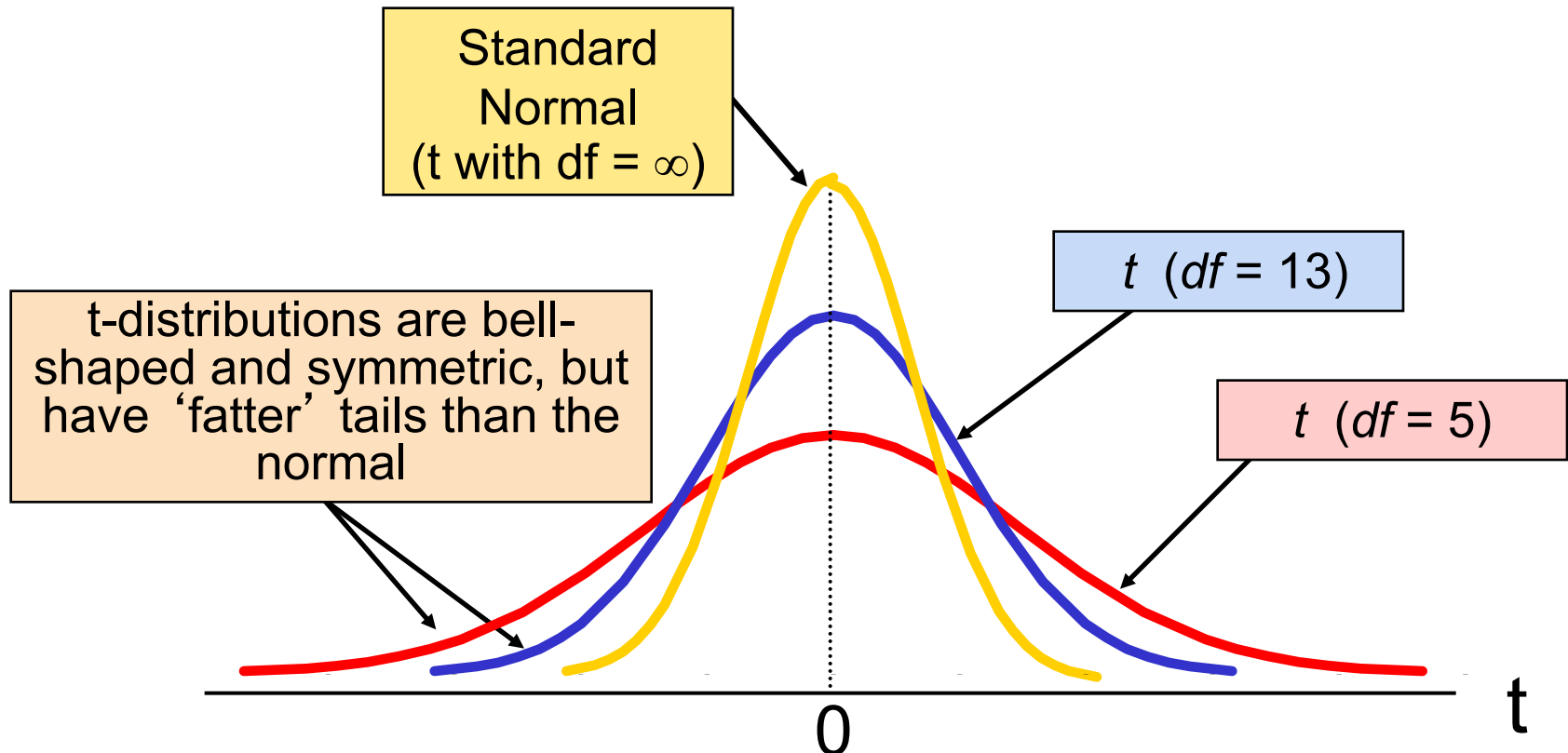


(gamma) is the Gamma function  
is the constant Pi (3.14...)



# Student's t Distribution

Note:  $t \rightarrow Z$  as  $n$  increases



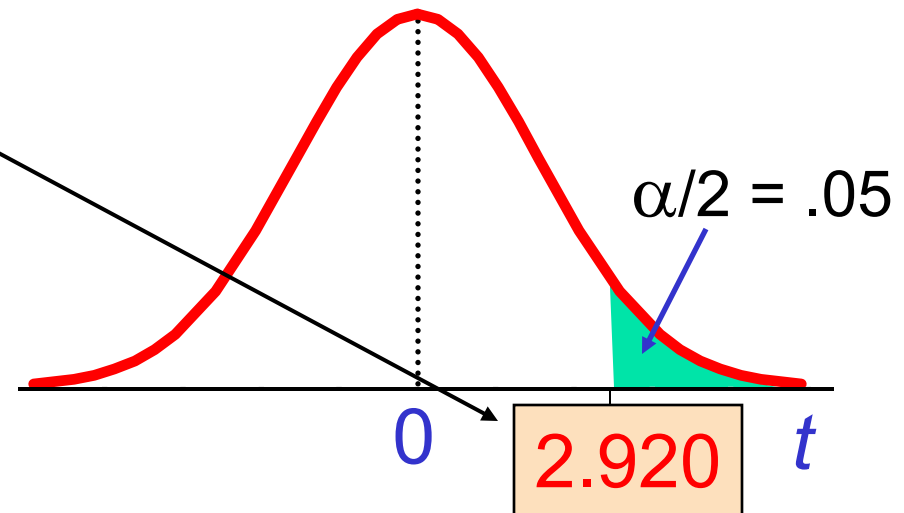


# Student's t Table

df	Upper Tail Area		
	.25	.10	.05
1	1.000	3.078	6.314
2	0.817	1.886	2.920
3	0.765	1.638	2.353

The body of the table contains t values, not probabilities

Let:  $n = 3$   
 $df = n - 1 = 2$   
 $\alpha = .10$   
 $\alpha/2 = .05$





# t distribution values

With comparison to the Z value

Confidence Level	t (10 d.f.)	t (20 d.f.)	t (30 d.f.)	Z
.80	1.372	1.325	1.310	1.28
.90	1.812	1.725	1.697	1.64
.95	2.228	2.086	2.042	1.96
.99	3.169	2.845	2.750	2.58

Note:  $t \rightarrow Z$  as  $n$  increases



# Summary: Single population mean (small n, normality)

- Hypothesis test:

$$t_{n-1} = \frac{\text{observed mean} - \text{null mean}}{\frac{s_x}{\sqrt{n}}}$$

- Confidence Interval

$$\text{confidence interval} = \text{observed mean} \pm t_{n-1, \alpha/2} * \left( \frac{s_x}{\sqrt{n}} \right)$$



# Summary: Single population mean (large n)

- Hypothesis test:

$$Z \cong t_{n-1} = \frac{\text{observed mean} - \text{null mean}}{\frac{s_x}{\sqrt{n}}}$$

- Confidence Interval

$$\text{confidence interval} = \text{observed mean} \pm [t_{n-1, \alpha/2} \cong Z_{\alpha/2}] * \left( \frac{s_x}{\sqrt{n}} \right)$$

# Example: Two-Tail Test ( $\sigma$ Unknown)



The average cost of a hotel room in New York is said to be \$168 per night. A random sample of 25 hotels resulted in  $\bar{x} = \$172.50$  and  $s = \$15.40$ . Test at the  $\alpha = 0.05$  level.

(Assume the population distribution is normal)



$$H_0: \mu = 168$$

$$H_1: \mu \neq 168$$



# Example Solution: Two-Tail Test

$$H_0: \mu = 168$$

$$H_1: \mu \neq 168$$

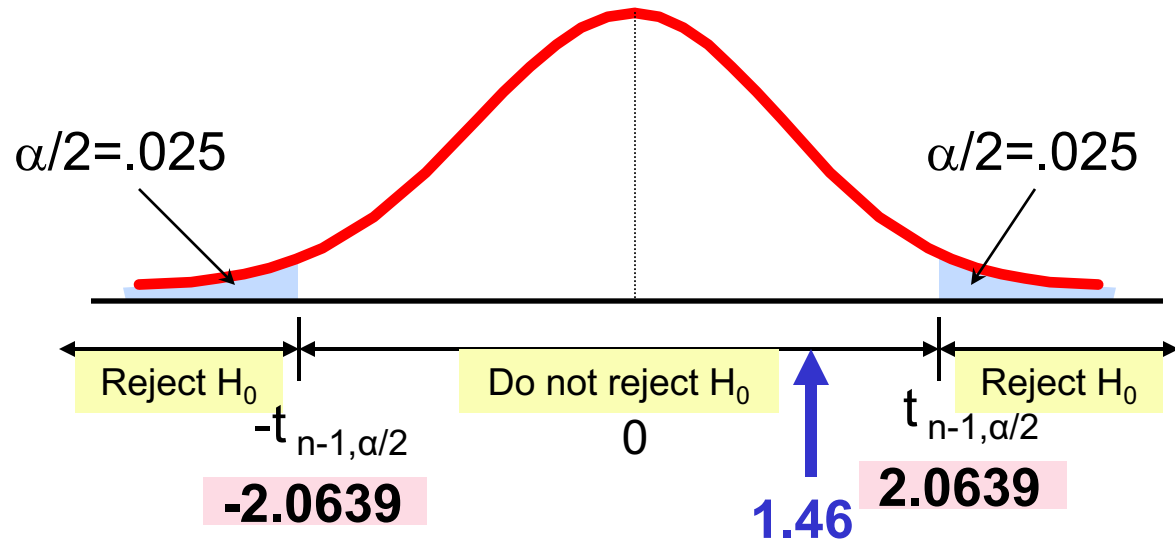
■  $\alpha = 0.05$

■  $n = 25$

■  $\sigma$  is unknown, so  
use a **t statistic**

■ Critical Value:

$$t_{24, .025} = \pm 2.0639$$



$$t_{n-1} = \frac{\bar{x} - \mu}{\frac{s}{\sqrt{n}}} = \frac{172.50 - 168}{\frac{15.40}{\sqrt{25}}} = 1.46$$

**Do not reject  $H_0$ :** not sufficient evidence that true mean cost is different than \$168



# Tests of the Population Proportion

---

- Involves **categorical variables**
- Two possible outcomes
  - “Success” (a certain characteristic is present)
  - “Failure” (the characteristic is not present)
- Fraction or proportion of the population in the “success” category is denoted by  $P$
- Assume sample size is large



# Proportions

(continued)

- Sample proportion in the success category is denoted by  $\hat{p}$

- $$\hat{p} = \frac{x}{n} = \frac{\text{number of successes in sample}}{\text{sample size}}$$

- When  $nP(1 - P) > 9$ ,  $\hat{p}$  can be approximated by a normal distribution with mean and standard deviation

- $$\mu_{\hat{p}} = P$$

$$\sigma_{\hat{p}} = \sqrt{\frac{P(1-P)}{n}}$$

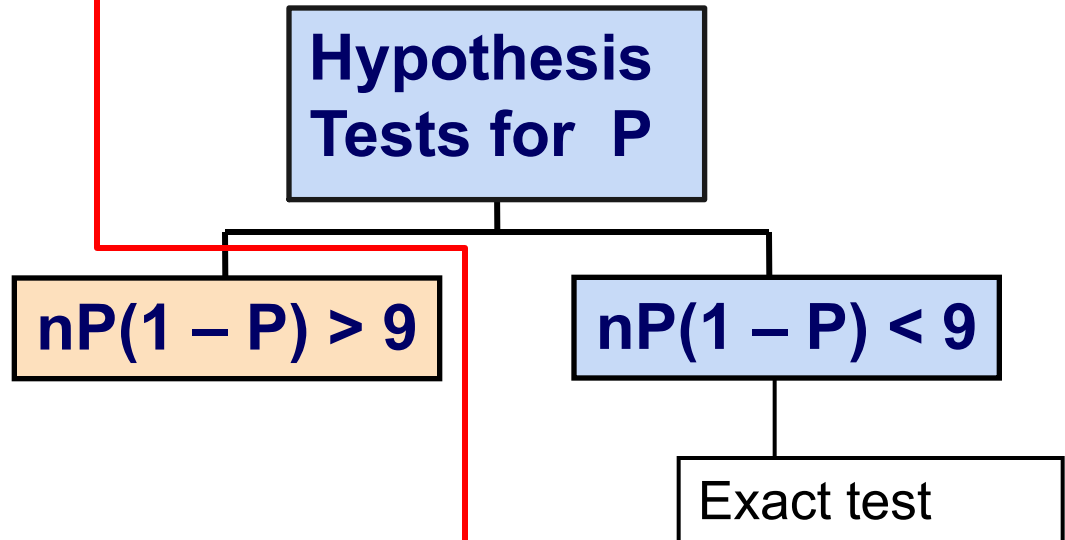




# Hypothesis Tests for Proportions

- The sampling distribution of  $\hat{p}$  is approximately normal, so the test statistic is a z value:

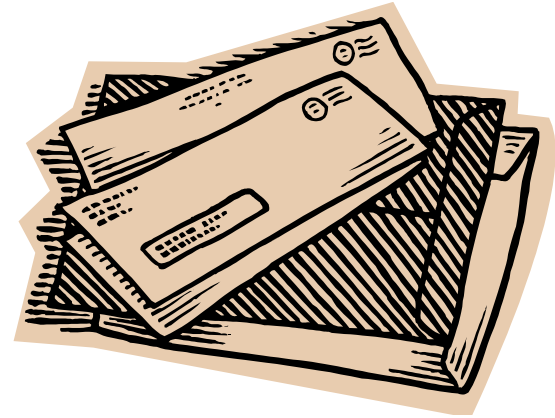
$$z = \frac{\hat{p} - P_0}{\sqrt{\frac{P_0(1 - P_0)}{n}}}$$





# Example: Z Test for Proportion

A marketing company claims that it receives 8% responses from its mailing. To test this claim, a random sample of 500 were surveyed with 25 responses. Test at the  $\alpha = .05$  significance level.



Check:

Our approximation for P is

$$\hat{p} = 25/500 = .05$$

$$\begin{aligned} nP(1 - P) &= (500)(.05)(.95) \\ &= 23.75 > 9 \end{aligned}$$





# Z Test for Proportion: Solution

$$H_0: P = .08$$

$$H_1: P \neq .08$$

$$\alpha = .05$$

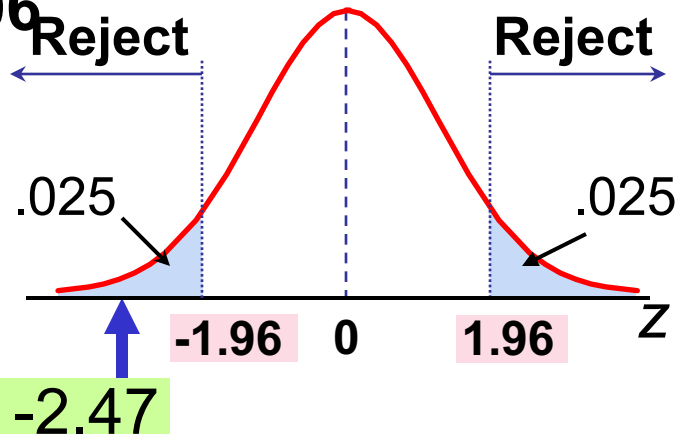
$$n = 500, \hat{p} = .05$$

**Test Statistic:**

$$z = \frac{\hat{p} - P_0}{\sqrt{\frac{P_0(1-P_0)}{n}}} = \frac{.05 - .08}{\sqrt{\frac{.08(1-.08)}{500}}} = -2.47$$

**Critical Values:**  $\pm$

1.96



**Decision:**

Reject  $H_0$  at  $\alpha = .05$

**Conclusion:**

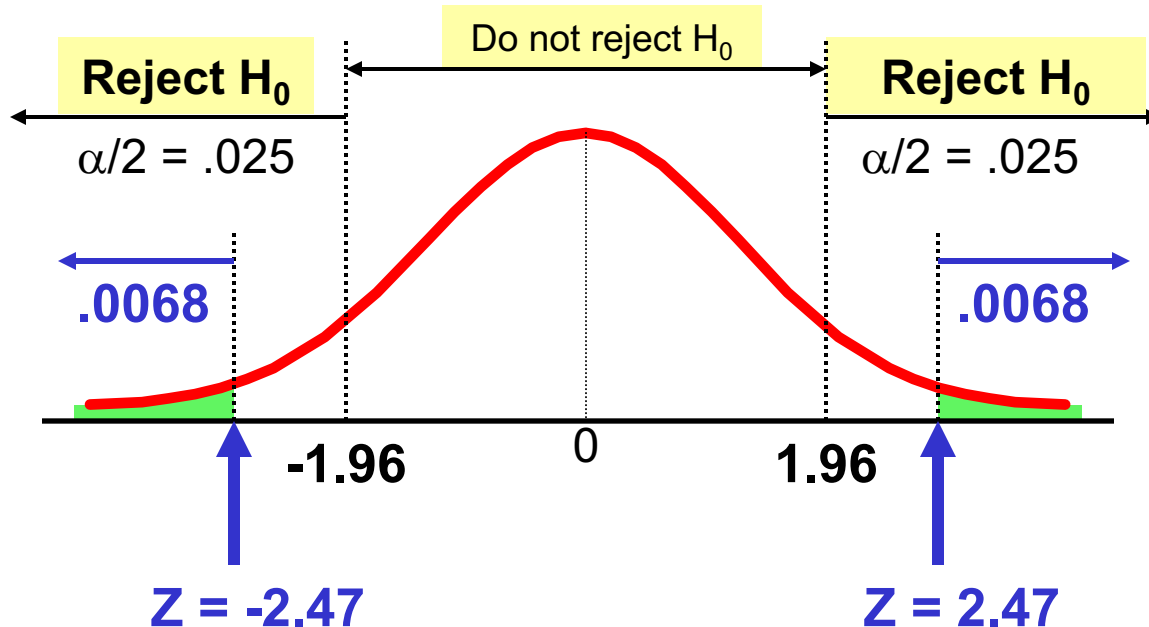
There is sufficient evidence to reject the company's claim of 8% response rate.



# p-Value Solution

(continued)

Calculate the p-value and compare to  $\alpha$   
(For a two sided test the p-value is always two sided)



**p-value = .0136:**

$$\begin{aligned} &P(Z \leq -2.47) + P(Z \geq 2.47) \\ &= 2(.0068) = 0.0136 \end{aligned}$$

**Reject  $H_0$  since p-value = .0136 <  $\alpha$  = .05**



# Power of the Test

- Recall the possible hypothesis test outcomes:

**Key:**  
**Outcome**  
**(Probability)**

	<b>Actual Situation</b>	
<b>Decision</b>	$H_0$ True	$H_0$ False
Do Not Reject $H_0$	<b>No error</b> <b>(<math>1 - \alpha</math>)</b>	<b>Type II Error</b> <b>(<math>\beta</math>)</b>
Reject $H_0$	<b>Type I Error</b> <b>(<math>\alpha</math>)</b>	<b>No Error</b> <b>(<math>1 - \beta</math>)</b>

- $\beta$  denotes the probability of Type II Error
- $1 - \beta$  is defined as the **power of the test**

Power =  $1 - \beta$  = the probability that a false null hypothesis is rejected



# Type II Error

Assume the population is normal and the population variance is known. Consider the test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu > \mu_0$$

The decision rule is:

$$\text{Reject } H_0 \text{ if } z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} > z_\alpha \quad \text{or} \quad \text{Reject } H_0 \text{ if } \bar{x} = \bar{x}_c > \mu_0 + Z_\alpha \sigma / \sqrt{n}$$

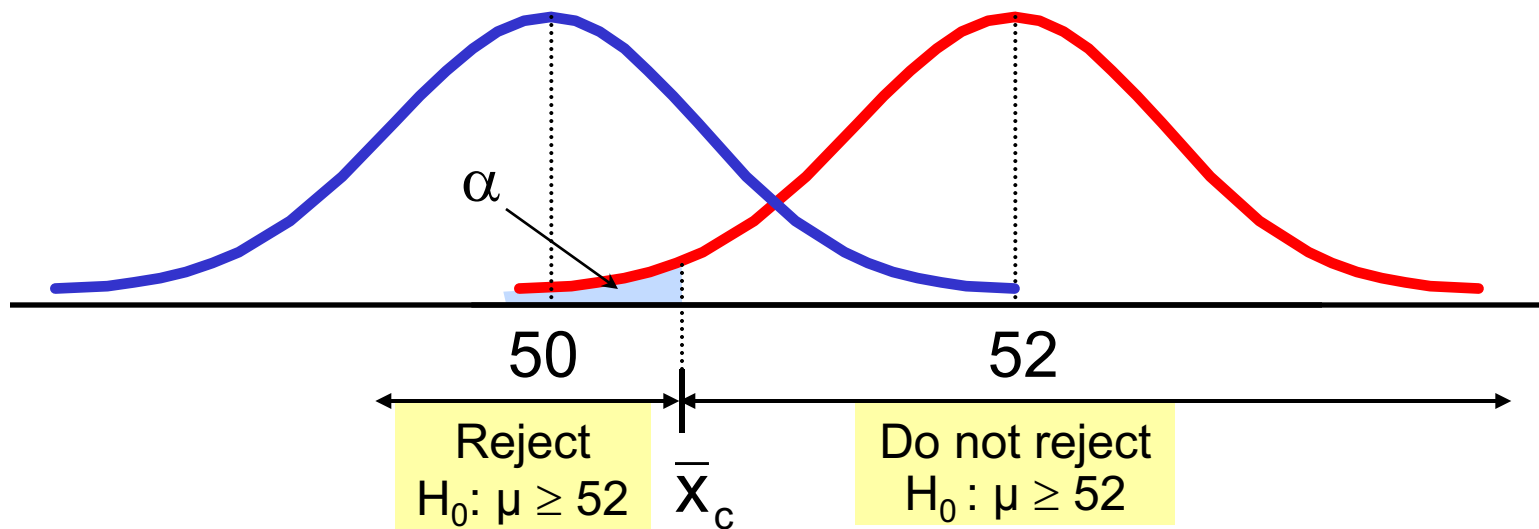
If the null hypothesis is false and the true mean is  $\mu^*$ , then the probability of type II error is

$$\beta = P(\bar{x} < \bar{x}_c \mid \mu = \mu^*) = P\left(z < \frac{\bar{x}_c - \mu^*}{\sigma / \sqrt{n}}\right)$$

# Type II Error Example

- Type II error is the probability of failing to reject a false  $H_0$

Suppose we fail to reject  $H_0: \mu \geq 52$   
when in fact the true mean is  $\mu^* = 50$

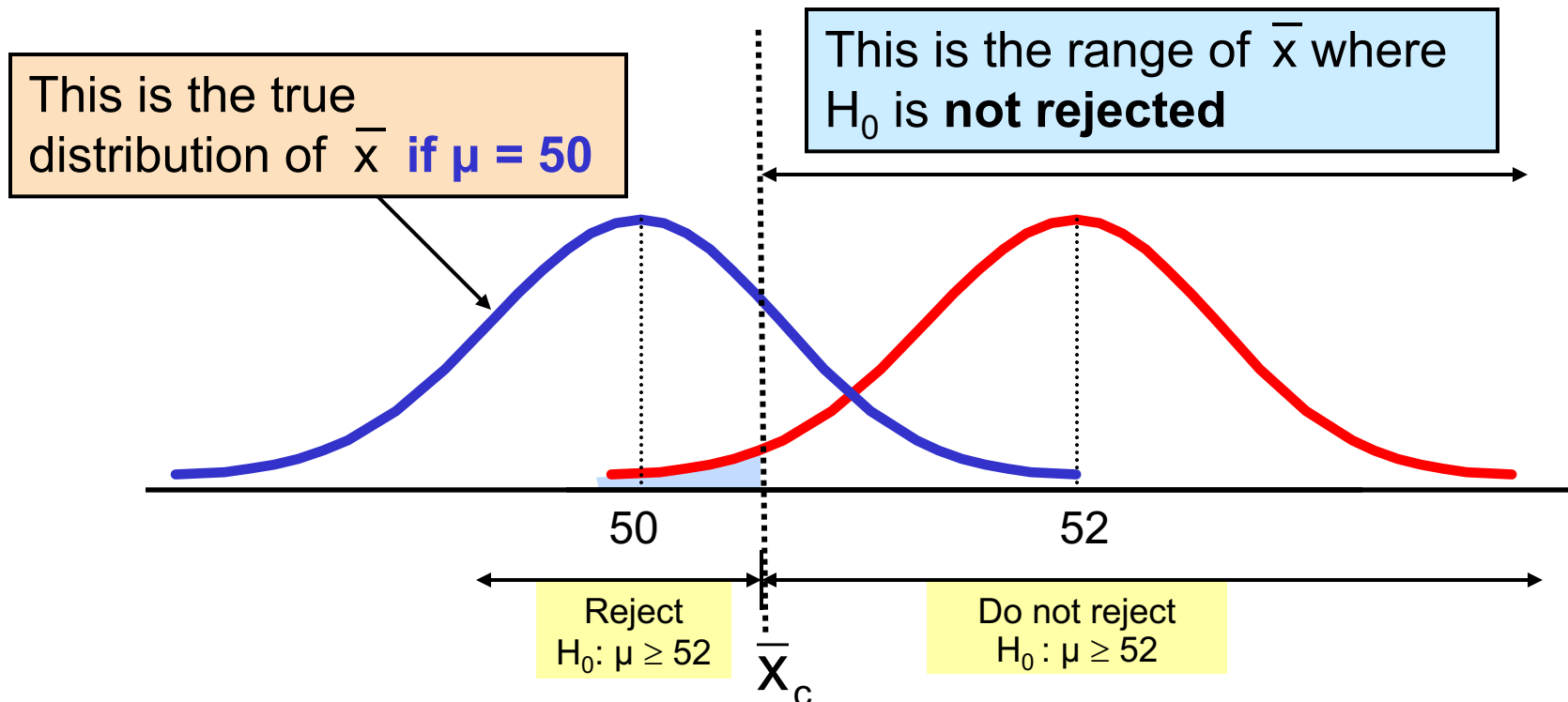




# Type II Error Example

(continued)

- Suppose we do not reject  $H_0: \mu \geq 52$  when in fact the true mean is  $\mu^* = 50$



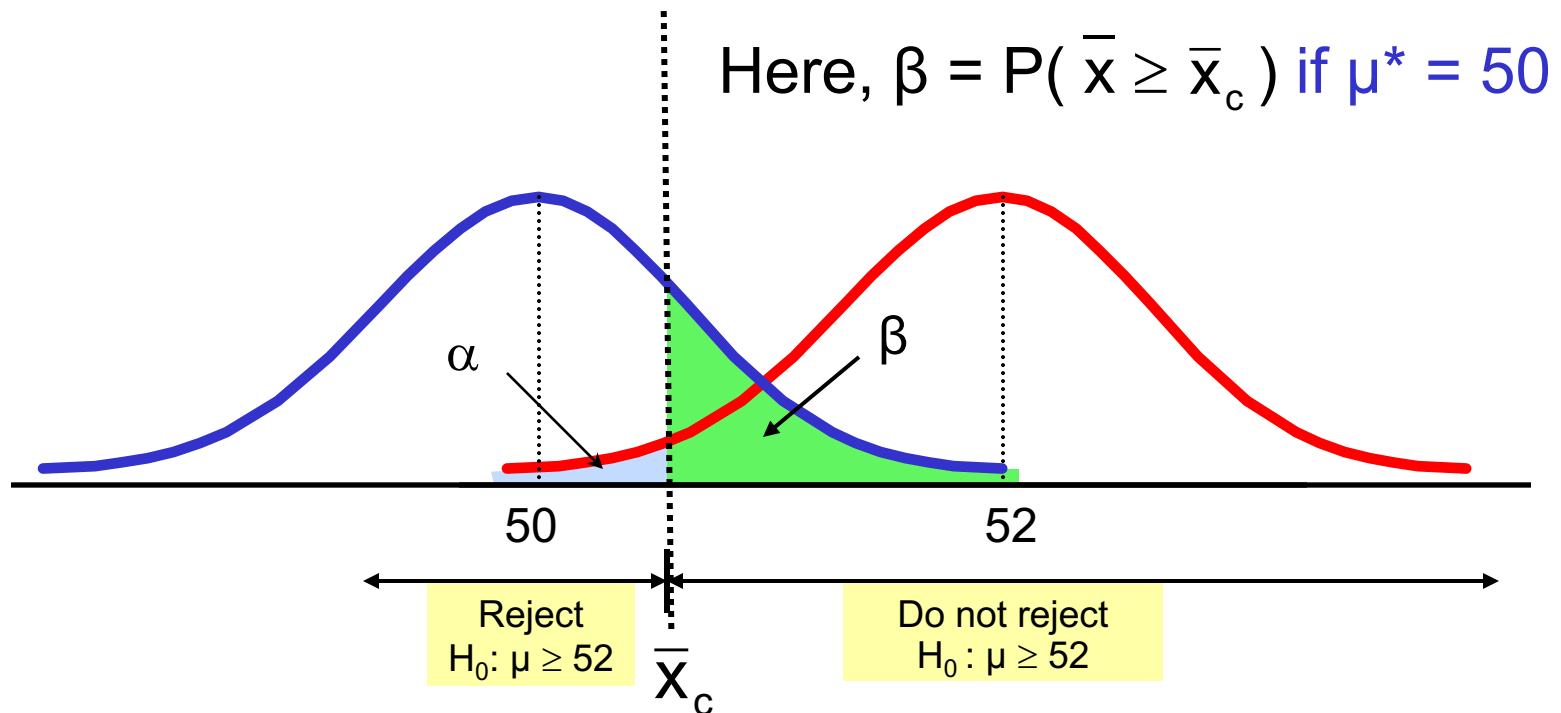




# Type II Error Example

(continued)

- Suppose we do not reject  $H_0: \mu \geq 52$  when in fact the true mean is  $\mu^* = 50$



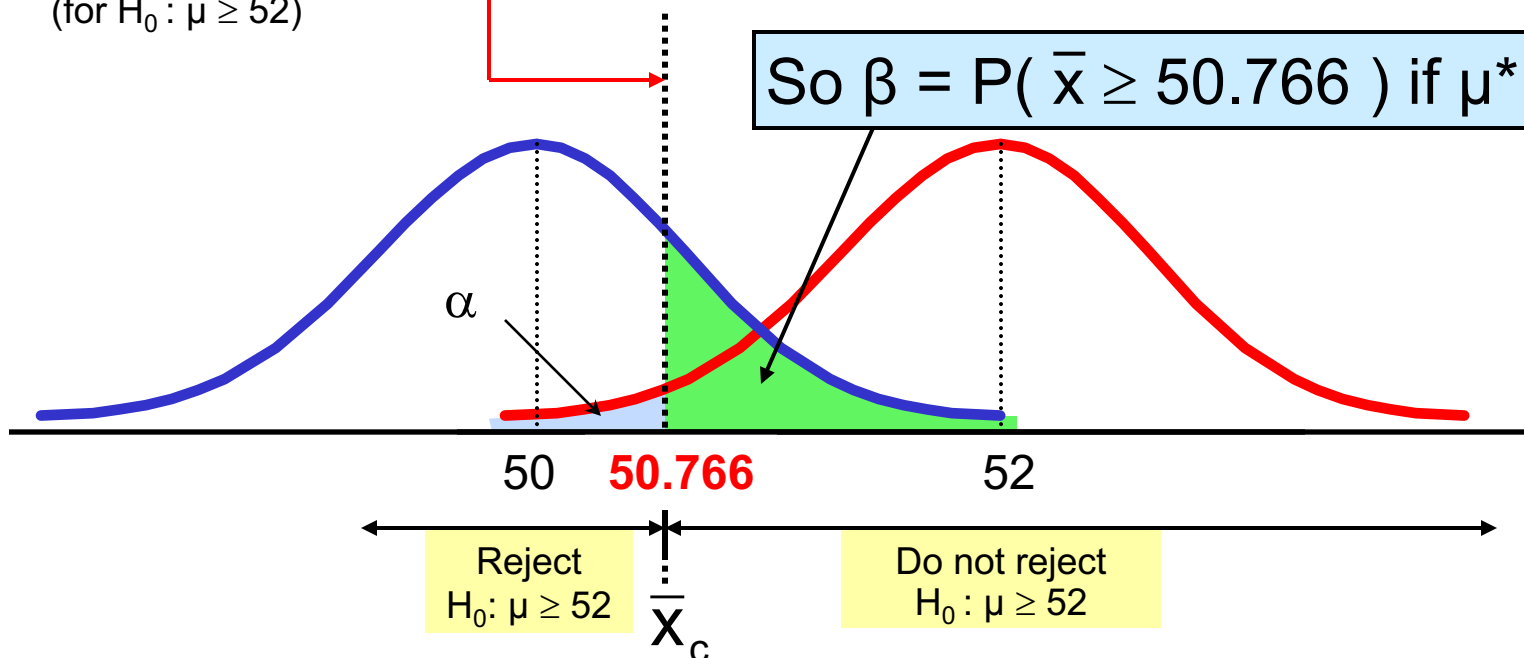


# Calculating $\beta$

- Suppose  $n = 64$ ,  $\sigma = 6$ , and  $\alpha = .05$

$$\bar{X}_c = \mu_0 - z_\alpha \frac{\sigma}{\sqrt{n}} = 52 - 1.645 \frac{6}{\sqrt{64}} = 50.766$$

(for  $H_0: \mu \geq 52$ )



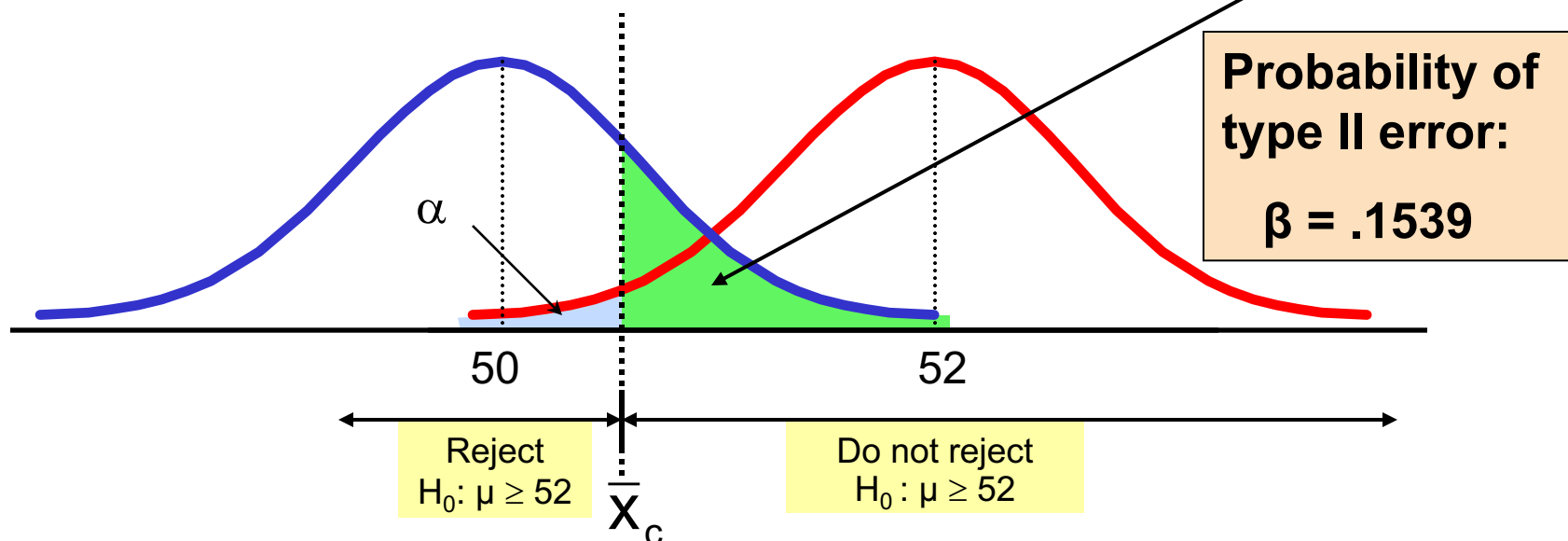


# Calculating $\beta$

(continued)

- Suppose  $n = 64$ ,  $\sigma = 6$ , and  $\alpha = .05$

$$P(\bar{x} \geq 50.766 | \mu^* = 50) = P\left(z \geq \frac{50.766 - 50}{\frac{6}{\sqrt{64}}}\right) = P(z \geq 1.02) = .5 - .3461 = .1539$$





# Power of the Test Example

If the true mean is  $\mu^* = 50$ ,

- The probability of Type II Error =  $\beta = 0.1539$
- The power of the test =  $1 - \beta = 1 - 0.1539 = 0.8461$

**Key:**  
**Outcome**  
**(Probability)**

	<b>Actual Situation</b>	
<b>Decision</b>	$H_0$ True	$H_0$ False
Do Not Reject $H_0$	<b>No error</b> <b><math>1 - \alpha = 0.95</math></b>	<b>Type II Error</b> <b><math>\beta = 0.1539</math></b>
Reject $H_0$	<b>Type I Error</b> <b><math>\alpha = 0.05</math></b>	<b>No Error</b> <b><math>1 - \beta = 0.8461</math></b>

(The value of  $\beta$  and the power will be different for each  $\mu^*$ )