

Statistics for Business and Economics



Chapter 9

Hypothesis Testing: Single Population

What is a Hypothesis?

- A hypothesis is a claim (assumption) about a population parameter:



- 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 household is equal to three ($H_0 : \mu = 3$)

- Is always about a population parameter, not about a sample statistic

$$H_0 : \mu = 3$$

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



The Null Hypothesis, H_0

(continued)

- Begin with the assumption that the null hypothesis is true
 - Similar to the notion of innocent until proven guilty
- Refers to the status quo
- Always contains “=”, “≤” or “≥” sign
- May or may not be rejected



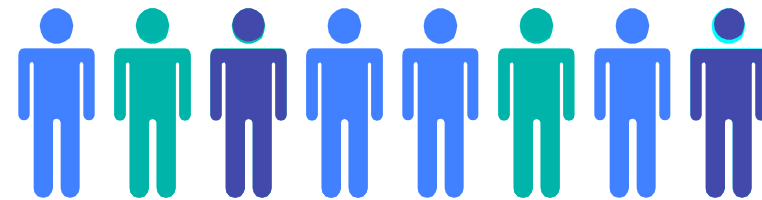


The Alternative Hypothesis, H_1

- Is the opposite of the null hypothesis
 - e.g., The average number of TV sets in U.S. homes is not equal to 3 ($H_1: \mu \neq 3$)
- Challenges the status quo
- Never contains the “=”, “≤” or “≥” sign
- May or may not be supported
- Is generally the hypothesis that the researcher is trying to support

Hypothesis Testing Process

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



Population

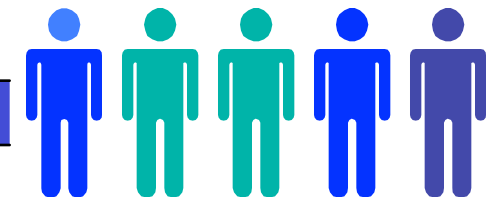
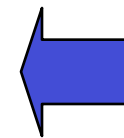
Now select a random sample



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

If not likely,
REJECT
Null Hypothesis

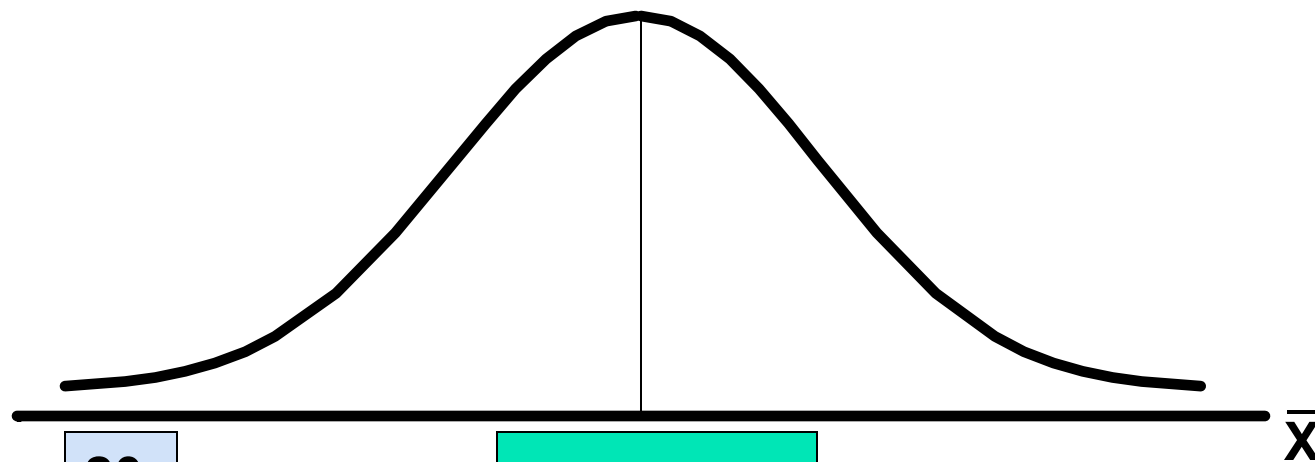
Suppose the sample mean age is 20: $\bar{X} = 20$



Sample

Reason for Rejecting H_0

Sampling Distribution of \bar{X}



20

$\mu = 50$
If H_0 is true

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

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

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



Level of Significance, α

- **Defines the unlikely values of the sample statistic if the null hypothesis is true**
 - Defines **rejection region** of the sampling distribution
- Is designated by **α** , (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

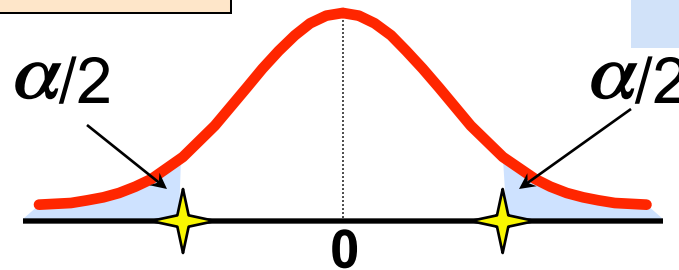
Level of significance = α

✦ Represents critical value

$$H_0: \mu = 3$$

$$H_1: \mu \neq 3$$

Two-tail test

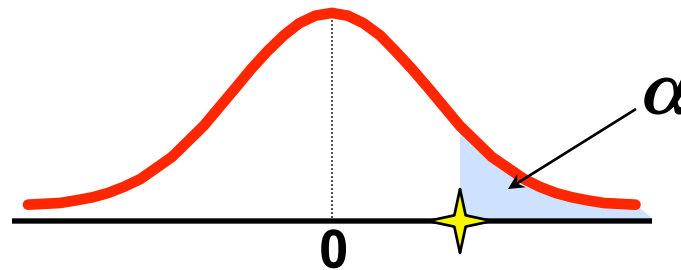


Rejection region is shaded

$$H_0: \mu \leq 3$$

$$H_1: \mu > 3$$

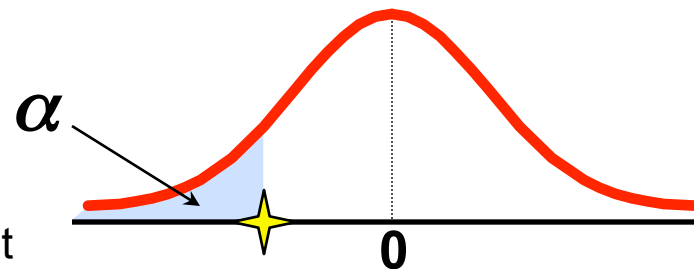
Upper-tail test



$$H_0: \mu \geq 3$$

$$H_1: \mu < 3$$

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 α

- 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 β



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 (β)
Reject H_0	Type I Error (α)	No Error ($1 - \beta$)

Key:
Outcome
(Probability)




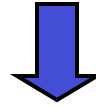





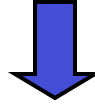
Type I & II Error Relationship

- 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 (α) , then
Type II error probability (β) 



Factors Affecting Type II Error

- All else equal,
 - β  when the difference between hypothesized parameter and its true value 
 - β  when α 
 - β  when σ 
 - β  when n 

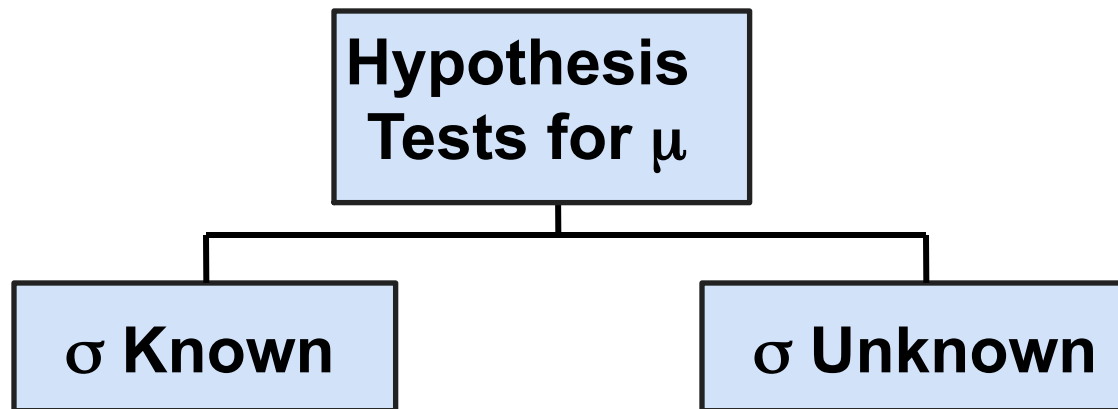


Power of the Test

- 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



Hypothesis Tests for the Mean



Test of Hypothesis for the Mean (σ Known)

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

Hypothesis Tests for μ

σ Known

σ Unknown

Consider the test

$$H_0: \mu = \mu_0$$

$$H_1: \mu > \mu_0$$

(Assume the population is normal)

The **decision rule** is:

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

Decision Rule

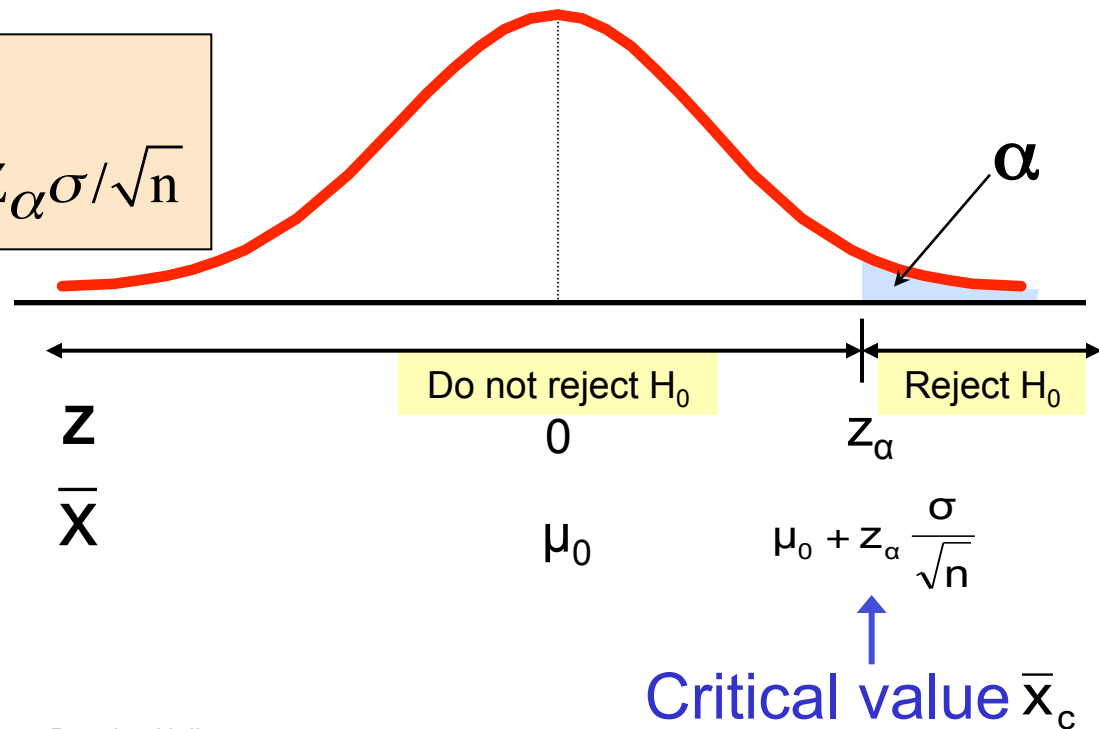
Reject H_0 if $z = \frac{\bar{x} - \mu_0}{\sigma / \sqrt{n}} \geq z_\alpha$

$$H_0: \mu = \mu_0$$

$$H_1: \mu > \mu_0$$

Alternate rule:

Reject H_0 if $\bar{x} \geq \mu_0 + Z_\alpha \sigma / \sqrt{n}$





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 α 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:

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

- **Decision rule:** compare the **p-value** to α

- If $\text{p-value} \leq \alpha$, reject H_0
- If $\text{p-value} > \alpha$, do not reject H_0

Example: Upper-Tail Z Test for Mean (σ 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:

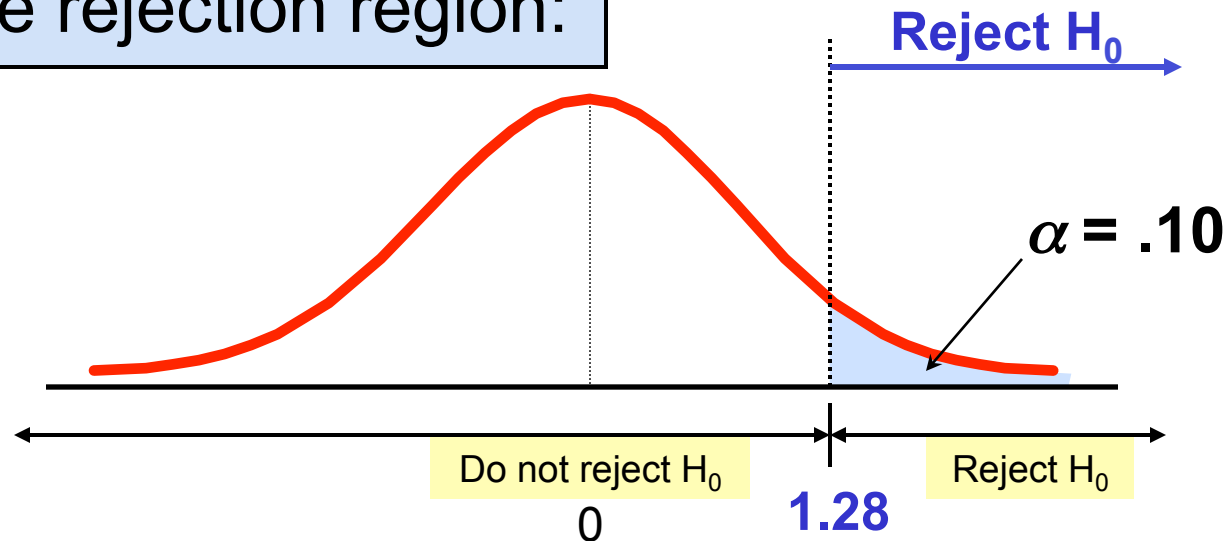
$H_0: \mu \leq 52$	the average is not over \$52 per month
$H_1: \mu > 52$	the average is greater than \$52 per month (i.e., sufficient evidence exists to support the manager's claim)

Example: Find Rejection Region

(continued)

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

Find the rejection region:



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

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,

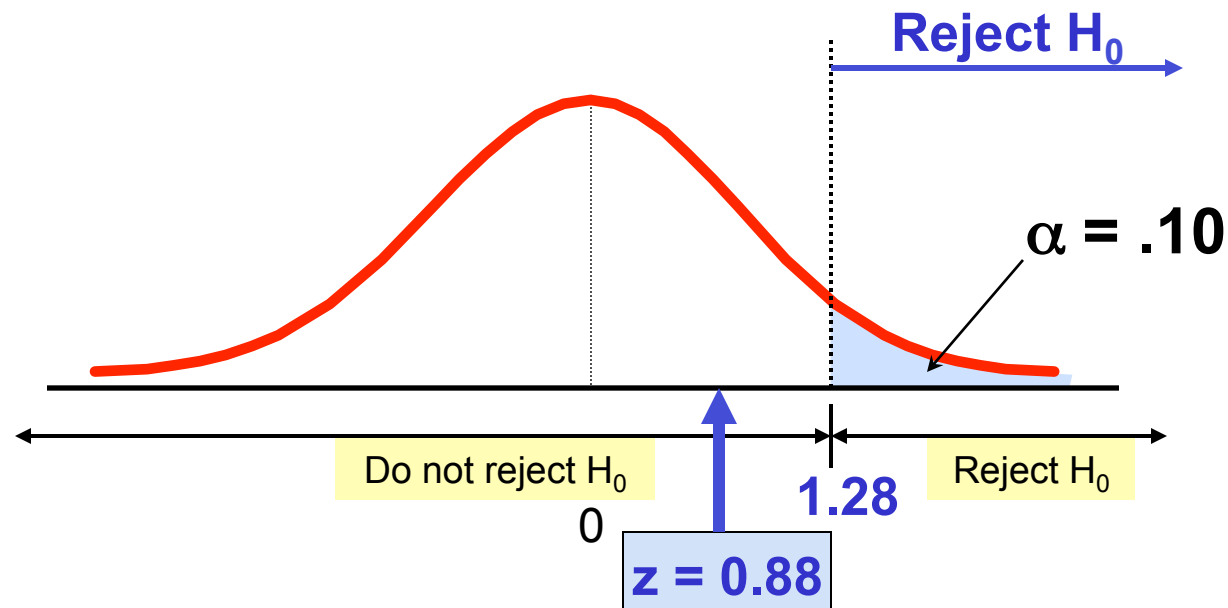
$$z = \frac{\bar{x} - \mu_0}{\frac{\sigma}{\sqrt{n}}} = \frac{53.1 - 52}{\frac{10}{\sqrt{64}}} = 0.88$$



Example: Decision

(continued)

Reach a decision and interpret the result:



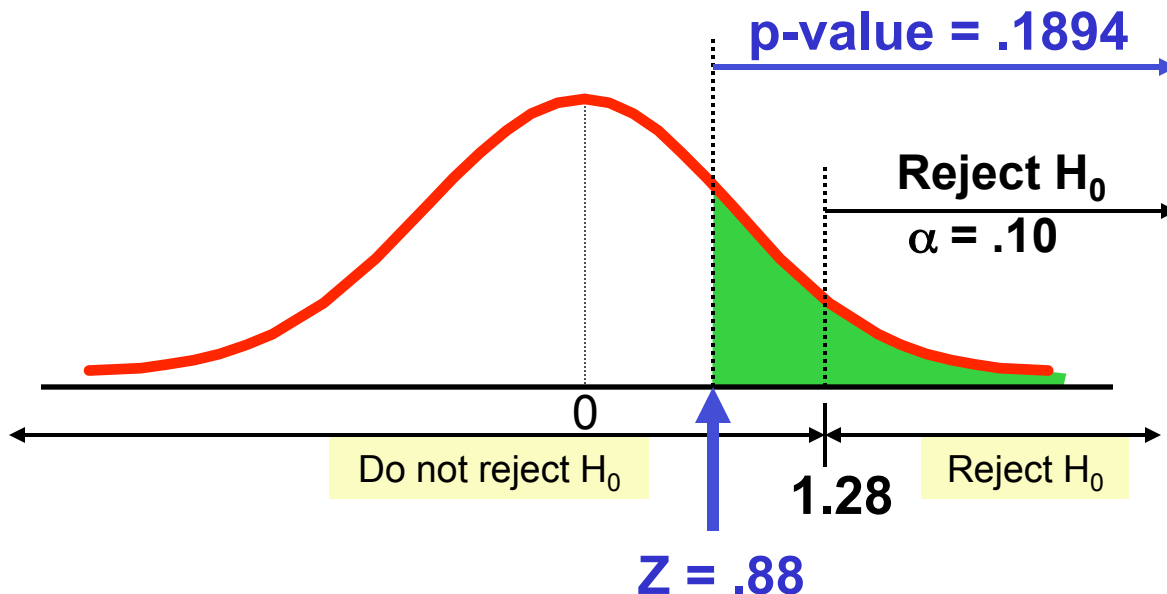
Do not reject H_0 since $z = 0.88 < 1.28$

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 α
(assuming that $\mu = 52.0$)



$$\begin{aligned} P(\bar{x} \geq 53.1 | \mu = 52.0) \\ &= P\left(z \geq \frac{53.1 - 52.0}{10/\sqrt{64}}\right) \\ &= P(z \geq 0.88) = 1 - .8106 \\ &= .1894 \end{aligned}$$

Do not reject H_0 since p-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

$$H_0: \mu \geq 3$$

$$H_1: \mu < 3$$



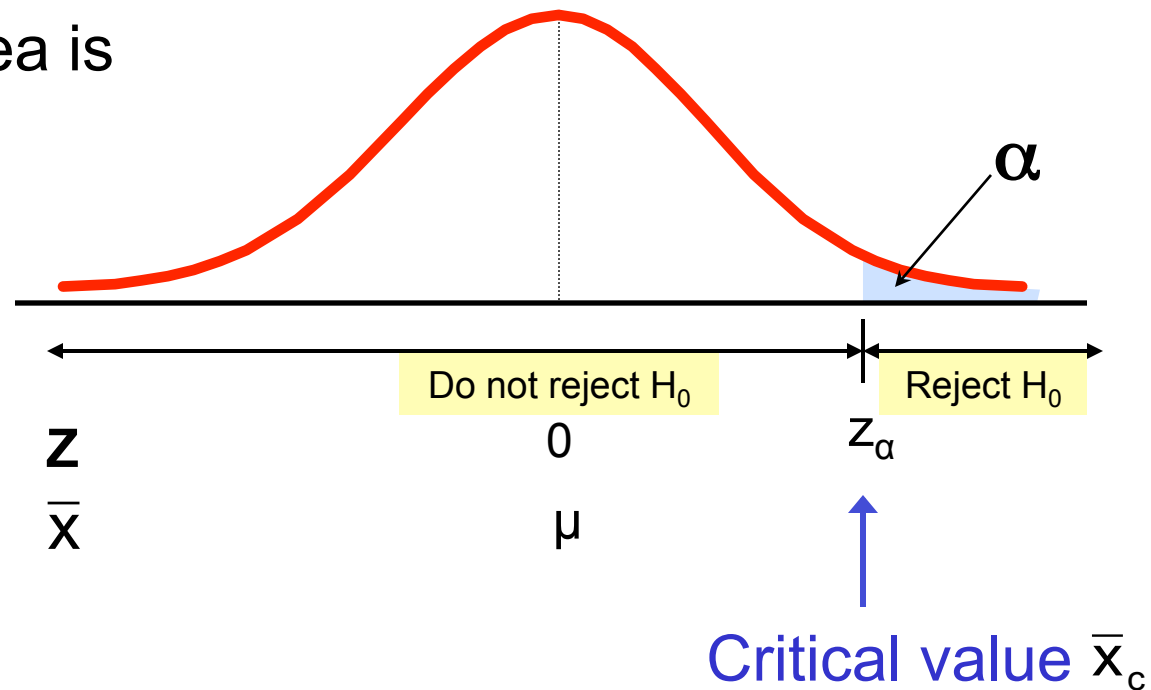
This is a **lower**-tail test since the alternative hypothesis is focused on the lower tail below the mean of 3

Upper-Tail Tests

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

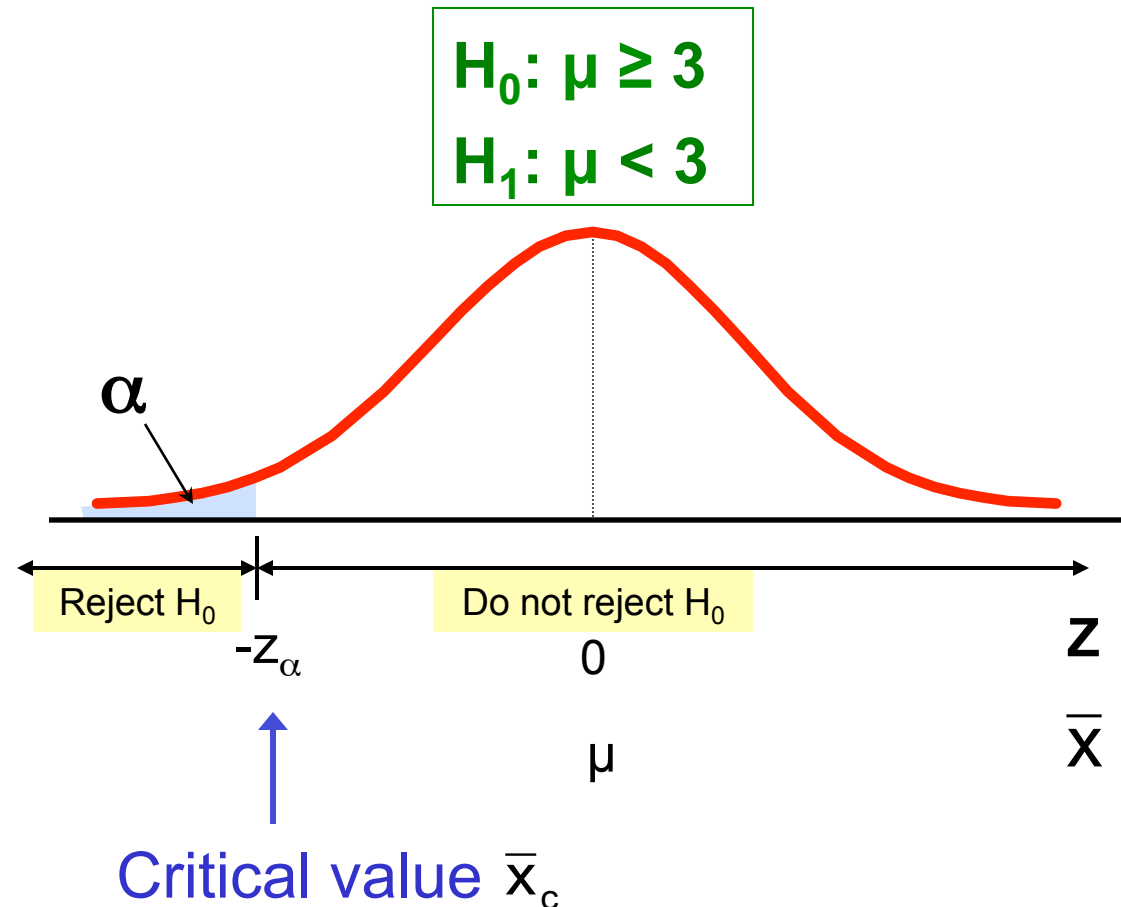
$$H_0: \mu \leq 3$$

$$H_1: \mu > 3$$



Lower-Tail Tests

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

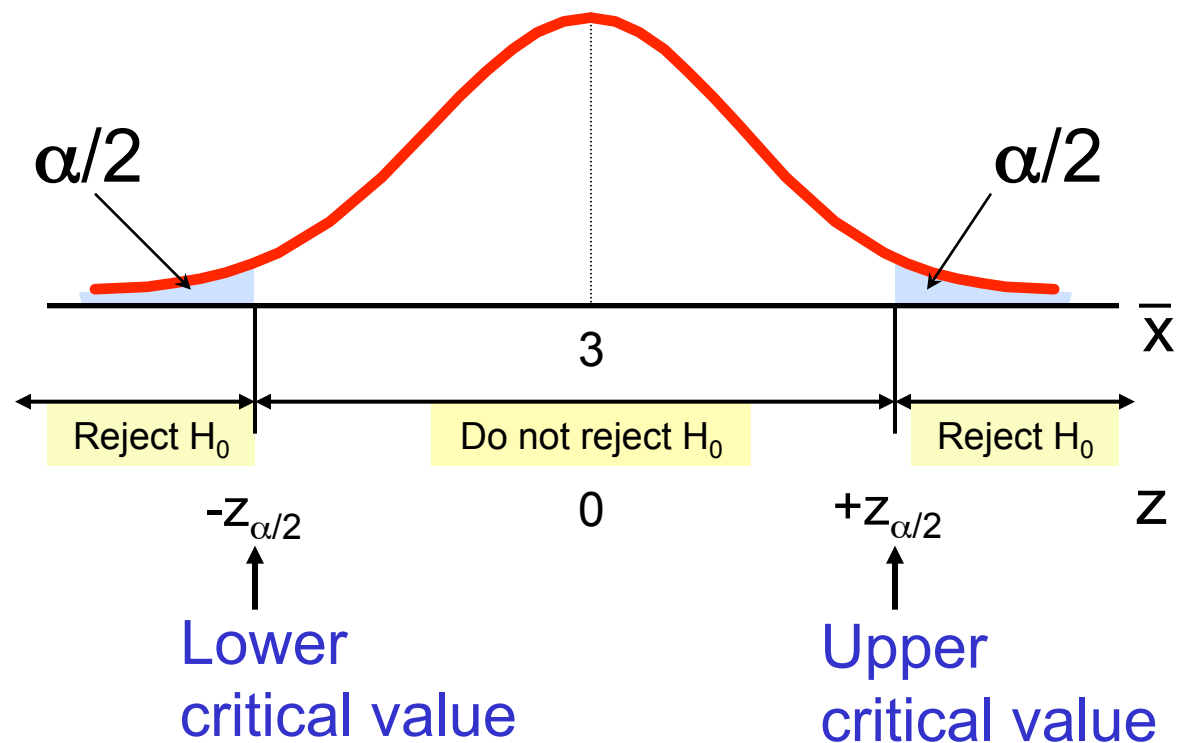


Two-Tail Tests

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

$$H_0: \mu = 3$$
$$H_1: \mu \neq 3$$

- 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 not 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
 - σ is known so this is a z test
- Set up the critical values
 - For $\alpha = .05$ the critical z values are ± 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:

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

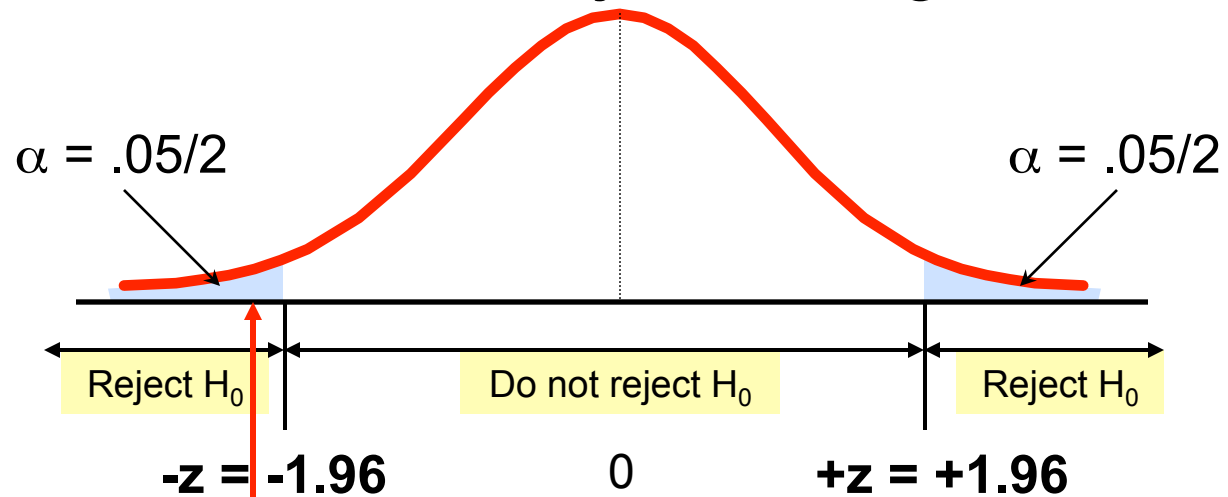


Hypothesis Testing Example

(continued)

- Is the test statistic in the rejection region?

Reject H_0 if
 $z \leq -1.96$ or
 $z \geq 1.96$;
otherwise
do not
reject H_0



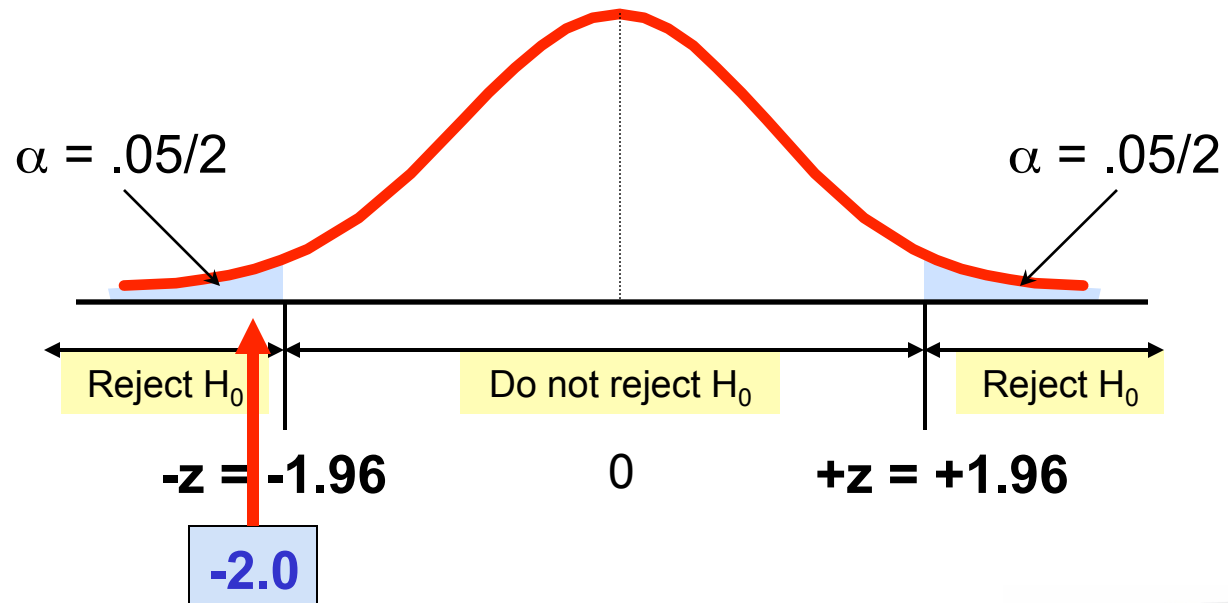
Here, $z = -2.0 < -1.96$, so the test statistic is in the rejection region



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 $\mu = 3.0$?

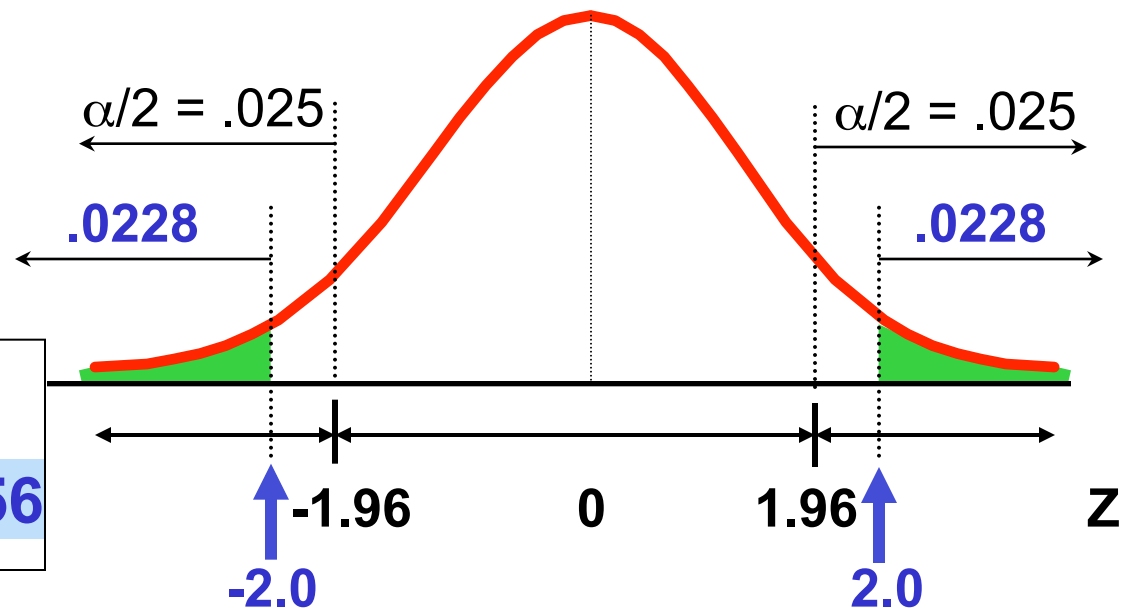
$\bar{x} = 2.84$ is translated to a z score of $z = -2.0$

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

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

p-value

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



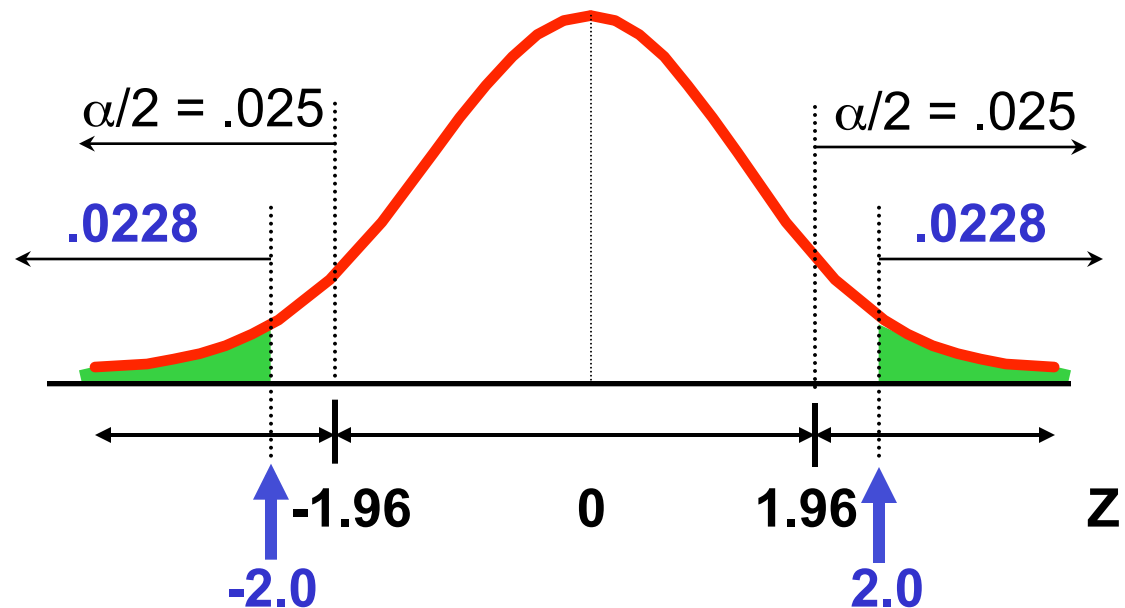
Example: p-Value

(continued)

- Compare the p-value with α
 - If p-value $\leq \alpha$, reject H_0
 - If p-value $> \alpha$, do not reject H_0

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

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



t Test of Hypothesis for the Mean (σ Unknown)

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

Hypothesis Tests for μ

σ Known

σ Unknown

Consider the test

$$H_0 : \mu = \mu_0$$

$$H_1 : \mu > \mu_0$$

(Assume the population is normal)

The **decision rule** is:

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

t Test of Hypothesis for the Mean (σ 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 $t = \frac{\bar{X} - \mu_0}{s / \sqrt{n}} \leq -t_{n-1, \alpha/2}$ or if $t = \frac{\bar{X} - \mu_0}{s / \sqrt{n}} \geq t_{n-1, \alpha/2}$

Example: Two-Tail Test (σ Unknown)

The average cost of a hotel room in Chicago 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)



$$H_0: \mu = 168$$

$$H_1: \mu \neq 168$$

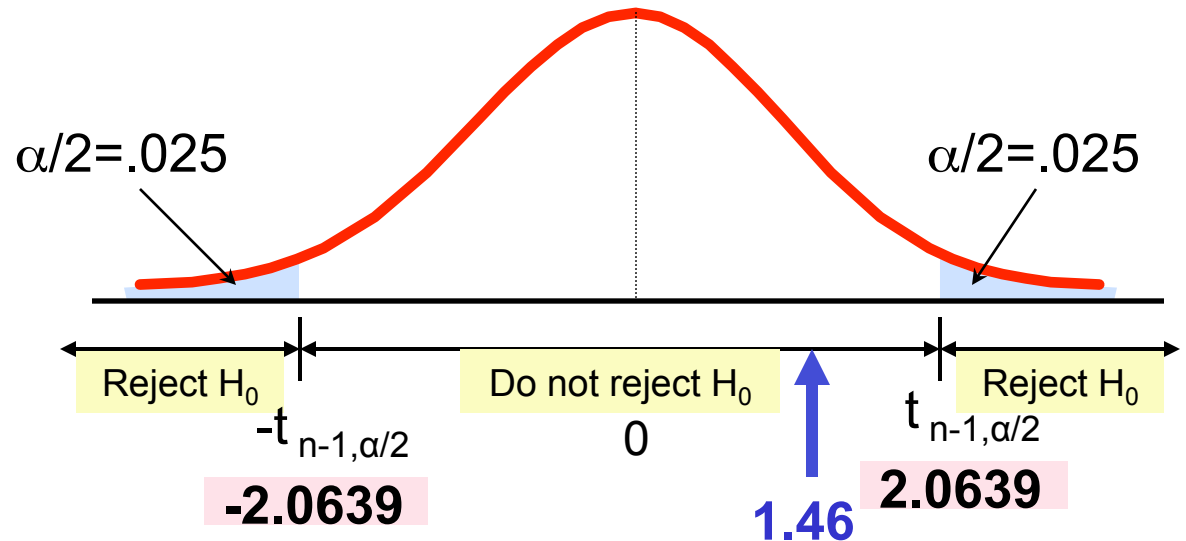
Example Solution: Two-Tail Test

$$H_0: \mu = 168$$

$$H_1: \mu \neq 168$$

- $\alpha = 0.05$
- $n = 25$
- σ 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}

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

- When sample size is large, \hat{p} can be approximated by a normal distribution with mean and standard deviation

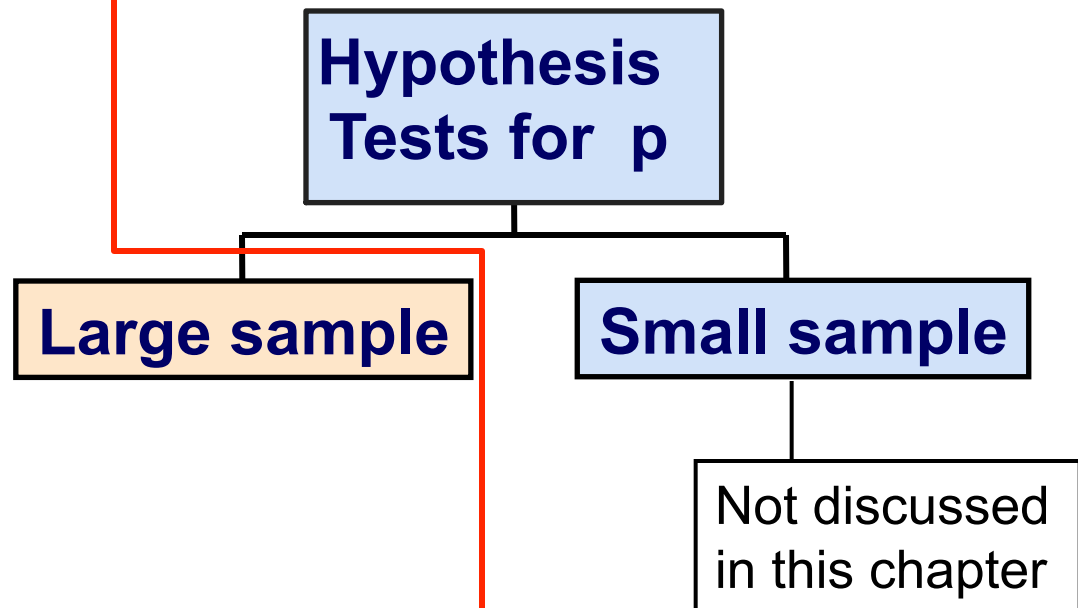
- $$\mu_{\hat{p}} = p$$

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

Hypothesis Tests for Proportions

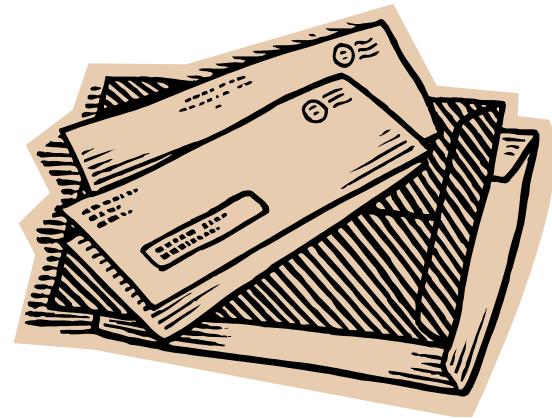
- The sampling distribution of \hat{p} is approximately normal, so the test statistic is a z value:

$$z = \frac{\hat{p} - p_0}{\sqrt{\frac{p_0(1-p_0)}{n}}}$$



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.



Z Test for Proportion: Solution

$$H_0: p = .08$$

$$H_1: p \neq .08$$

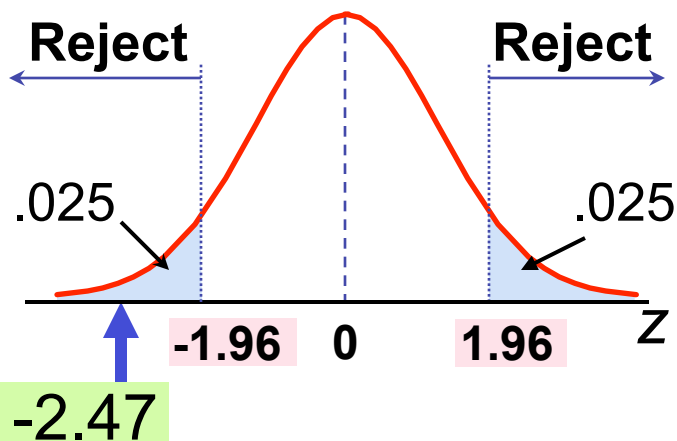
$$\alpha = .05$$

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

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: ± 1.96



Decision:

Reject H_0 at $\alpha = .05$

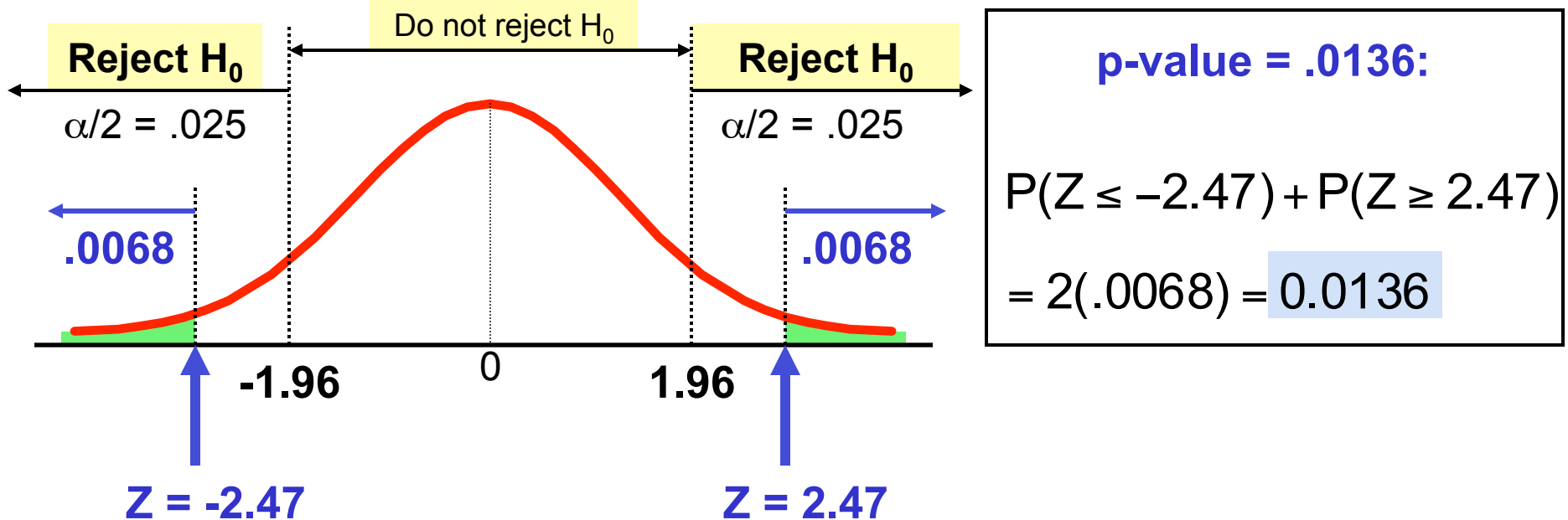
Conclusion:

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 α
(For a two sided test the p-value is always two sided)



Reject H_0 since p-value = .0136 < α = .05

Power of the Test

- Recall the possible hypothesis test outcomes:

Key:
Outcome
(Probability)

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

- β 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

$$H_0 : \mu \geq \mu_0 = 52$$

$$H_1 : \mu < \mu_0 = 52$$

The decision rule is:

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

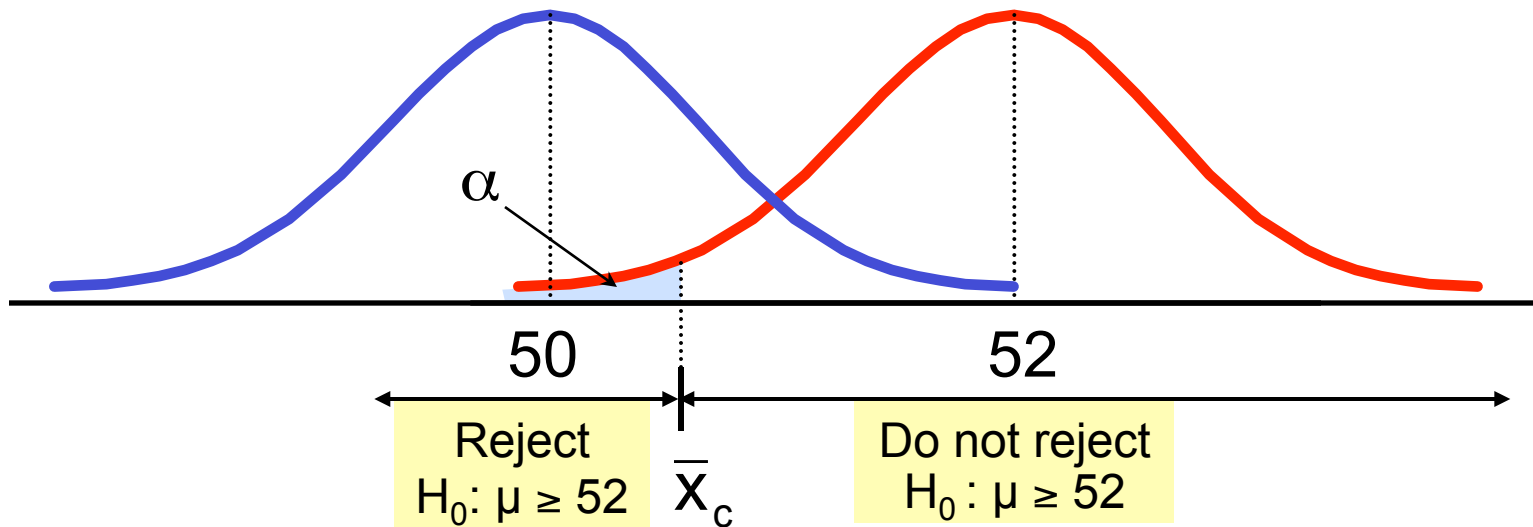
If the null hypothesis is false and the true mean is μ^* , then the probability of type II error is

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

Type II Error Example

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

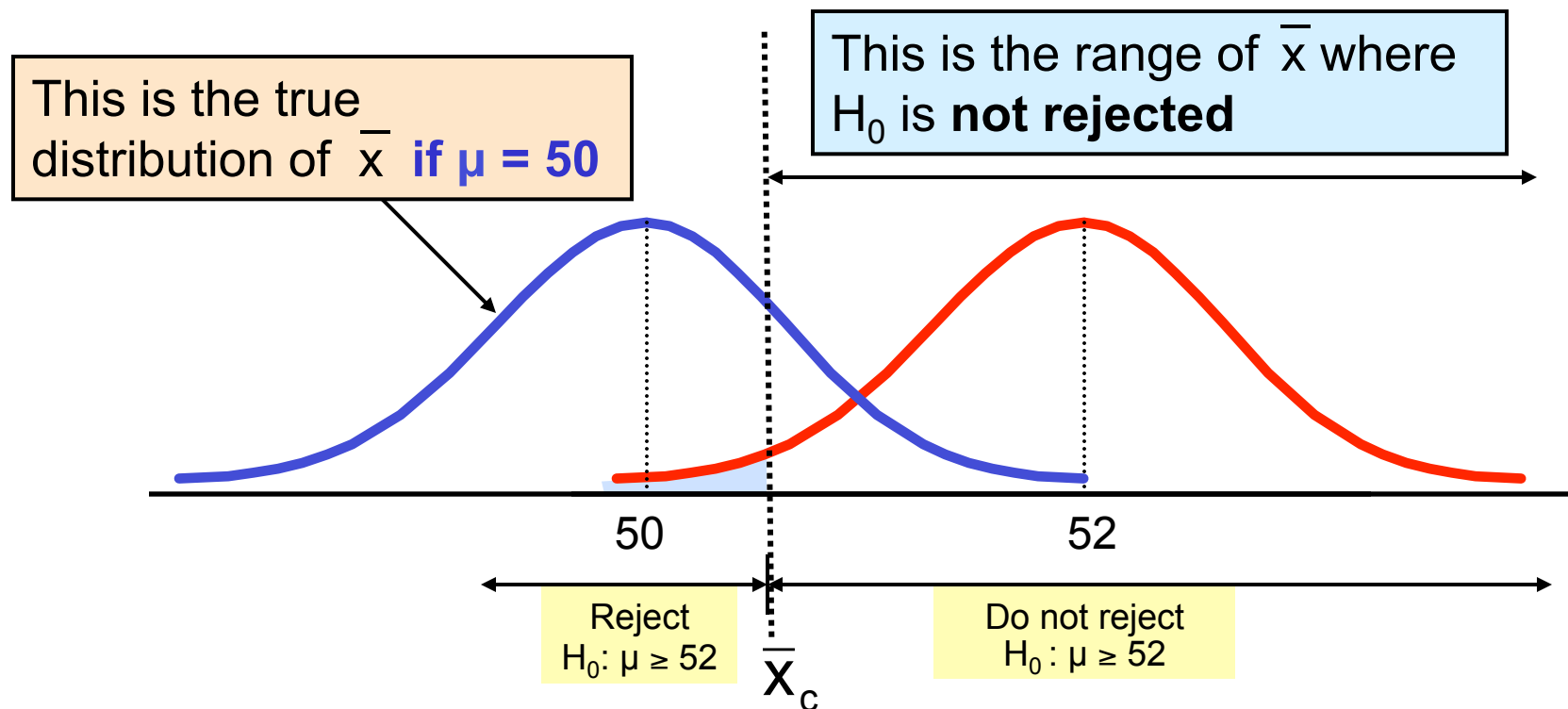
Suppose we fail to reject $H_0: \mu \geq 52$
when in fact the true mean is $\mu^* = 50$



Type II Error Example

(continued)

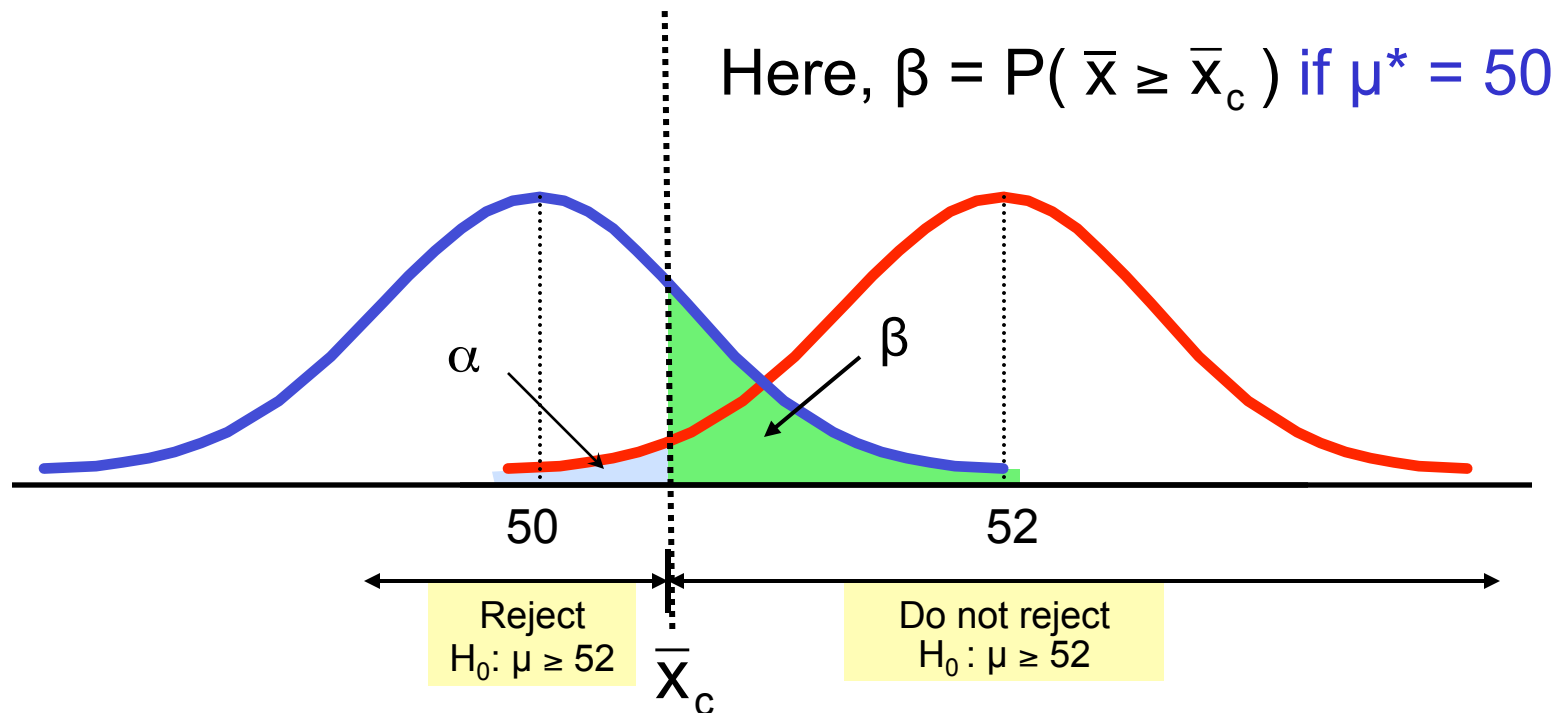
- Suppose we do not reject $H_0: \mu \geq 52$ when in fact the true mean is $\mu^* = 50$



Type II Error Example

(continued)

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



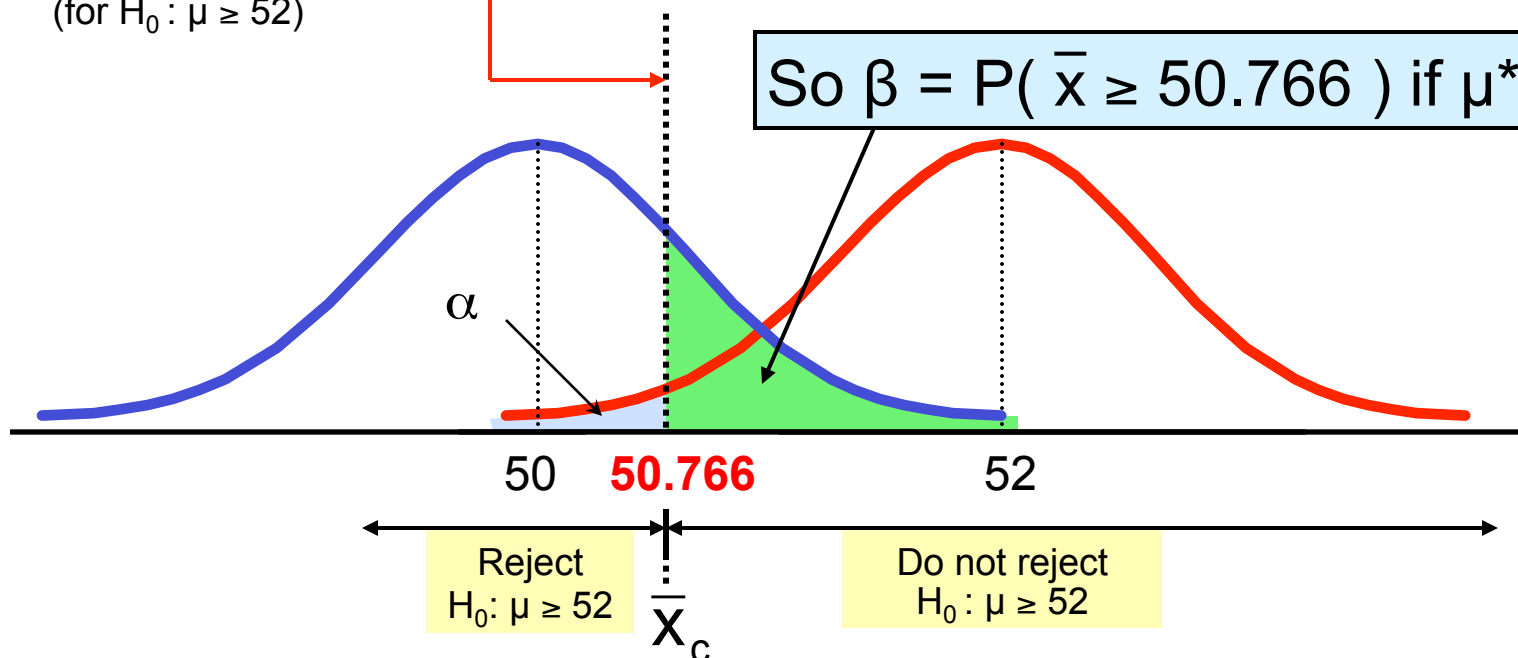
Calculating β

- 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 $H_0: \mu \geq 52$)

So $\beta = P(\bar{x} \geq 50.766)$ if $\mu^* = 50$

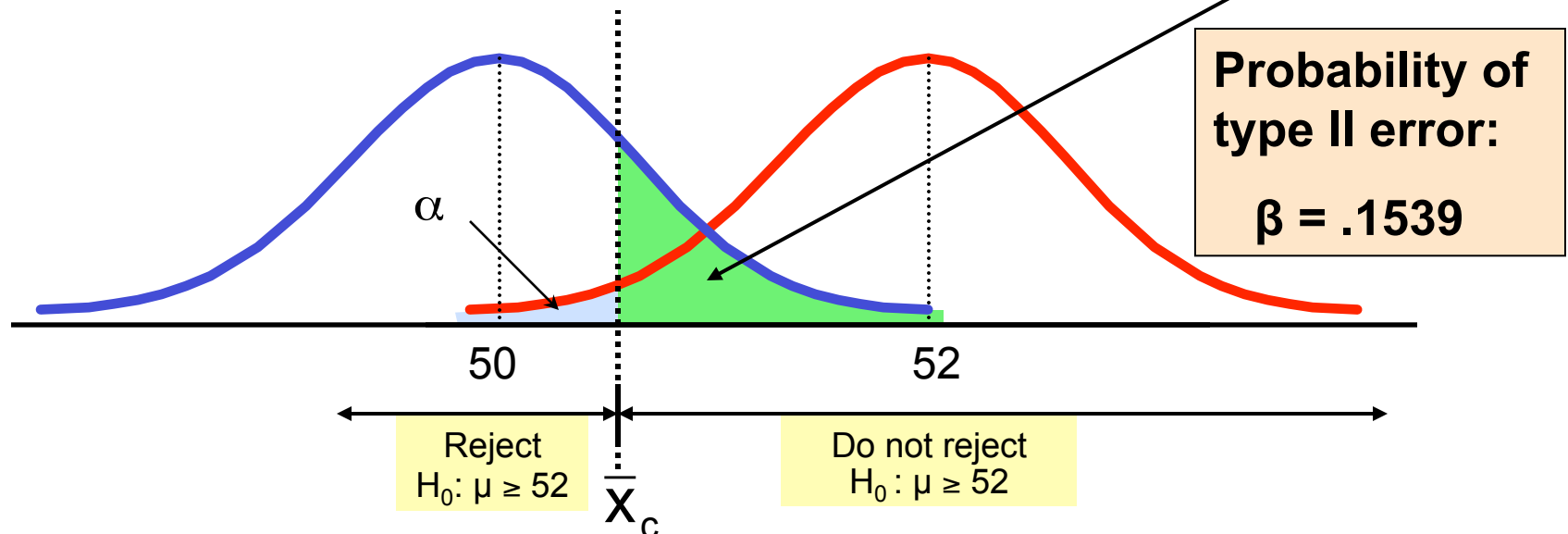


Calculating β

(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)

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

(The value of β and the power will be different for each μ^*)

Hypothesis Tests of one Population Variance

- **Goal:** Test hypotheses about the population variance, σ^2
- If the population is normally distributed,

$$\chi_{n-1}^2 = \frac{(n-1)s^2}{\sigma^2}$$

has a chi-square distribution with $(n - 1)$ degrees of freedom



Hypothesis Tests of one Population Variance

(continued)

The test statistic for hypothesis tests about one population variance is

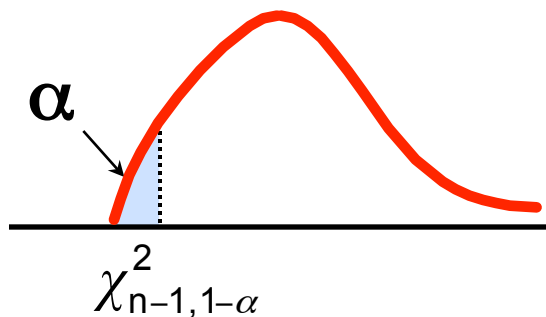
$$\chi_{n-1}^2 = \frac{(n-1)s^2}{\sigma_0^2}$$

Decision Rules: Variance

Population variance

Lower-tail test:

$$H_0: \sigma^2 \geq \sigma_0^2$$
$$H_1: \sigma^2 < \sigma_0^2$$

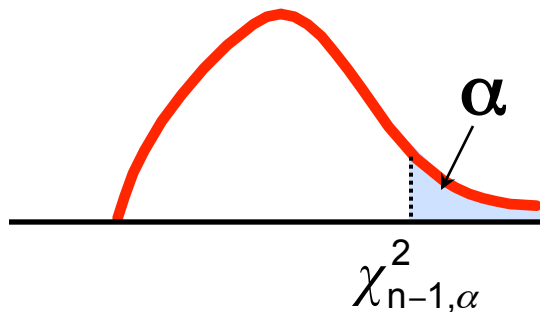


Reject H_0 if

$$\chi_{n-1}^2 \leq \chi_{n-1, 1-\alpha}^2$$

Upper-tail test:

$$H_0: \sigma^2 \leq \sigma_0^2$$
$$H_1: \sigma^2 > \sigma_0^2$$

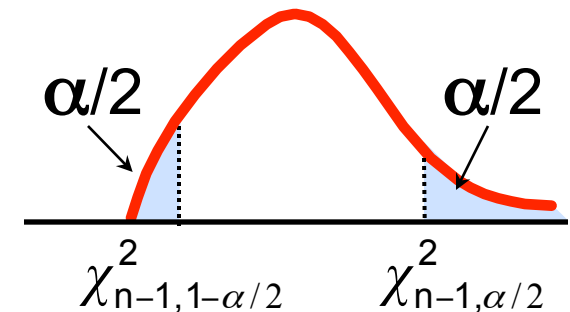


Reject H_0 if

$$\chi_{n-1}^2 \geq \chi_{n-1, \alpha}^2$$

Two-tail test:

$$H_0: \sigma^2 = \sigma_0^2$$
$$H_1: \sigma^2 \neq \sigma_0^2$$



Reject H_0 if

or $\chi_{n-1}^2 \geq \chi_{n-1, \alpha/2}^2$

$$\chi_{n-1}^2 \leq \chi_{n-1, 1-\alpha/2}^2$$