

College of Science  
Department of Statistics & OR

STAT 109  
Biostatistics

Chapter 7

**Using Sample Statistics To  
Test Hypotheses About Population Parameters**

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NOTE: This presentation is based on the presentation prepared thankfully by Professor Abdullah al-Shiha

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## 7.1 Introduction

Consider a population with some **unknown parameter**  $\theta$ . We are interested in testing (confirming or denying) some conjectures about  $\theta$ . For example, we might be interested in testing the conjecture that  $\theta > \theta_0$ , where  $\theta_0$  is a **given value**.

- A **hypothesis** is a statement about one or more populations.
- A **research hypothesis** is the conjecture or supposition that motivates the research.
- A **statistical hypothesis** is a conjecture (or a statement) concerning the population which can be evaluated by appropriate statistical technique.
- For example, if  $\theta$  is an **unknown parameter** of the population, we might be interested in testing the conjecture stating that  $\theta \geq \theta_0$  against  $\theta < \theta_0$  (for **some specific value**  $\theta_0$ ).

# The hypothesis

- We usually test the null hypothesis ( $H_0$ ) against the alternative (or the research) hypothesis ( $H_1$  or  $H_A$ ) by choosing one of the following situations:

(i)	$H_0: \theta = \theta_0$	against	$H_A: \theta \neq \theta_0$
(ii)	$H_0: \theta \geq \theta_0$	against	$H_A: \theta < \theta_0$
(iii)	$H_0: \theta \leq \theta_0$	against	$H_A: \theta > \theta_0$

- Equality sign must appear in the null hypothesis ( $H_0$ ) .
- $H_0$  is the null hypothesis, and  $H_A$  is the alternative hypothesis.  
( $H_0$  and  $H_A$  are **complement** of each other)
- The null hypothesis ( $H_0$ ) is also called "the hypothesis of no difference".
- The alternative hypothesis ( $H_A$ ) is also called the **research hypothesis**.

There are 4 possible situations in testing a statistical hypothesis:

		Condition of Null Hypothesis Ho (Nature/reality)	
		Ho is true	Ho is false
Possible Action (Decision)	Accepting Ho	Correct Decision	Type II error ( $\beta$ )
	Rejecting Ho	Type I error ( $\alpha$ )	Correct Decision

- There are two types of Errors:
  - Type I error = Rejecting Ho when Ho is true
  - $P(\text{Type I error}) = P(\text{Rejecting Ho} \mid \text{Ho is true}) = \alpha$
  - Type II error = Accepting Ho when Ho is false
  - $P(\text{Type II error}) = P(\text{Accepting Ho} \mid \text{Ho is false}) = \beta$

- The level of significance of the test is the probability of rejecting true  $H_0$ :

$$\alpha = P(\text{Rejecting } H_0 \mid H_0 \text{ is true}) = \text{P(Type I error)}$$

- There are 2 types of alternative hypothesis:

- One-sided alternative hypothesis ( $H_A$ ):

- $H_0: \theta \geq \theta_0$  against  $H_A: \theta < \theta_0$

- $H_0: \theta \leq \theta_0$  against  $H_A: \theta > \theta_0$

- Two-sided alternative hypothesis ( $H_A$ ):

- $H_0: \theta = \theta_0$  against  $H_A: \theta \neq \theta_0$

- We will use the terms "accepting" and "not rejecting" interchangeably. Also, we will use the terms "acceptance" and "nonrejection" interchangeably.

We will use the terms "accept" and "fail to reject" interchangeably

# The Procedure of Testing $H_0$ (against $H_A$ ):

The test procedure for rejecting  $H_0$  (accepting  $H_A$ ) or accepting  $H_0$  (rejecting  $H_A$ ) involves the following steps:

1. **Determine the hypothesis :** Null hypothesis ( $H_0$ ) and Alternative hypothesis ( $H_A$ ) .

2. **Determining a test statistic (T.S.)**

We choose the appropriate test statistic based on the point estimator of the parameter.

The test statistic has the following form:

$$\text{Test statistic} = \frac{\text{Estimate} - \text{hypothesized parameter}}{\text{Standard error of the estimate}}$$

3. **Determining the level of significance ( $\alpha$ ):**

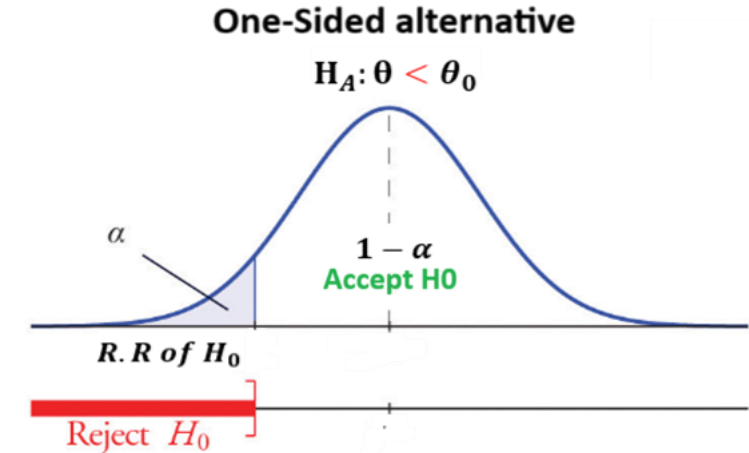
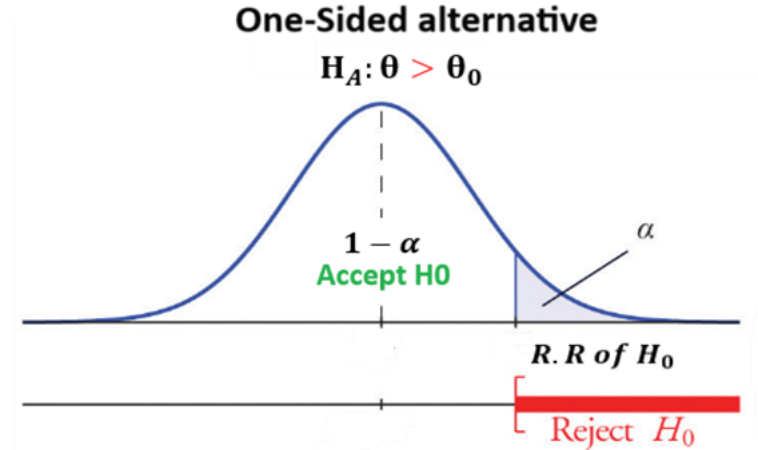
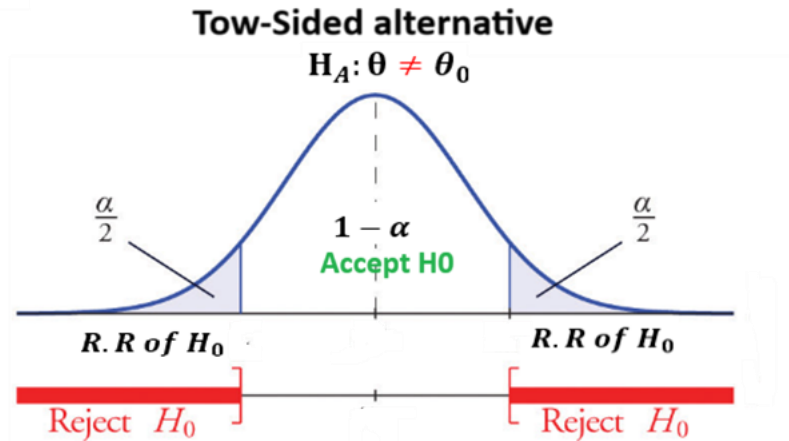
$$\alpha = 0.01, 0.025, 0.05, 0.10$$

4. **Determining the rejection region of  $H_0$  (R.R.) and the acceptance region of  $H_0$  (A.R.).**

The R.R. of  $H_0$  depends on  $H_A$  and  $\alpha$

- $H_A$  determines the **direction** of the R.R. of  $H_0$  .
- $\alpha$  determines the **size** of the R.R. of  $H_0$  . ( $\alpha$  = the size of the R.R. of  $H_0$  = shaded area)

The rejection region(R.R of  $H_0$  ) depends on the sign of  $H_A$  and  $\alpha$  :



## 5. Decision:

We reject  $H_0$  (and accept  $H_A$ ) if the value of the test statistic (T.S.) belongs to the R.R. of  $H_0$  ,and vice versa.

## Notes:

1. The **rejection region of  $H_0$  (R.R.)** is sometimes called "**the critical region**".
2. The values which separate the rejection region (R.R.) and the acceptance region (A.R.) are called the **critical values** or "**Reliability coefficient**".

## 7.2 Hypothesis Testing: A Single Population Mean ( $\mu$ )

Suppose that  $x_1, x_2, \dots, x_n$  is a random sample of size  $n$  from a distribution (or population) with mean  $\mu$  and variance  $\sigma^2$ .

We need to test some hypotheses (make some statistical inference) about the mean ( $\mu$ ).

### (1) First case:

#### Assumptions:

-The variance  $\sigma^2$  is known + Normal distribution with any sample size .

**or**

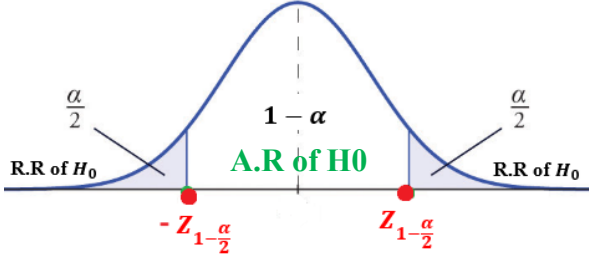
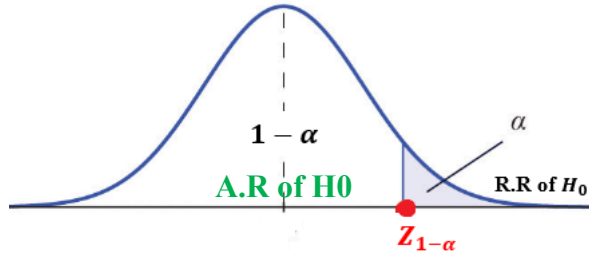
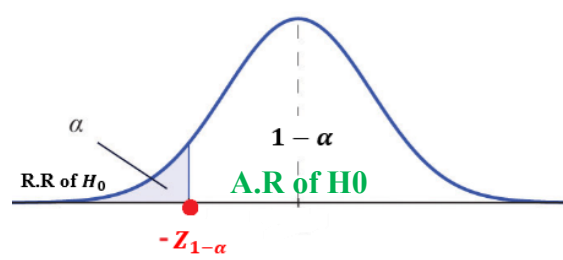
-The variance  $\sigma^2$  is known + Non-normal distribution with large sample size ( $n \geq 30$ ).

### (2) Second case:

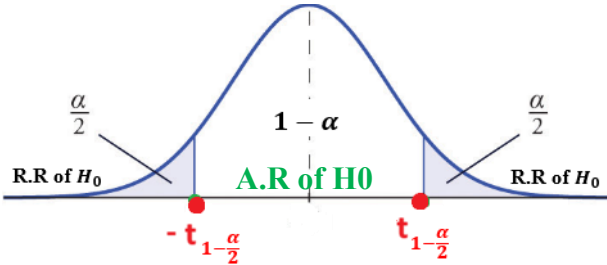
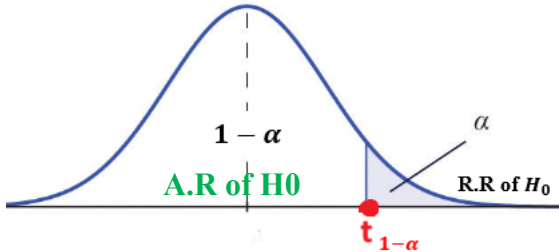
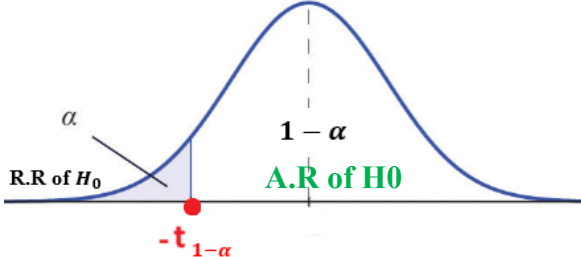
#### Assumptions:

-The variance  $\sigma^2$  is unknown + Normal distribution + Small sample size ( $n < 30$ ).

# Test procedures:

Hypotheses	$H_0 : \mu = \mu_0$ $H_A : \mu \neq \mu_0$	$H_0 : \mu \leq \mu_0$ $H_A : \mu > \mu_0$	$H_0 : \mu \geq \mu_0$ $H_A : \mu < \mu_0$
First case	$\sigma$ is known + Normal distribution or $\sigma$ is known + Non-normal distribution (n large = $n \geq 30$ )		
Test Statistic (T.S.)	$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}} \sim N(0,1)$		
R.R. & A.R. of $H_0$			
Critical value(s)	$Z_{1-\frac{\alpha}{2}}$ and $-Z_{1-\frac{\alpha}{2}}$	$Z_{1-\alpha}$	$-Z_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$Z_{T.S} > Z_{1-\frac{\alpha}{2}}$ or $Z_{T.S} < -Z_{1-\frac{\alpha}{2}}$	$Z_{T.S} > Z_{1-\alpha}$	$Z_{T.S} < -Z_{1-\alpha}$
	T.S. $\in$ R.R. Two - Sided Test	T.S. $\in$ R.R. One - Sided Test	T.S. $\in$ R.R. One - Sided Test

# Test procedures:

Hypotheses	$H_0 : \mu = \mu_0$ $H_A : \mu \neq \mu_0$	$H_0 : \mu \leq \mu_0$ $H_A : \mu > \mu_0$	$H_0 : \mu \geq \mu_0$ $H_A : \mu < \mu_0$
Second case	$\sigma$ is unknown, Normal distribution and n small (n < 30 )		
Test Statistic (T.S.)	$T = \frac{\bar{X} - \mu_0}{S/\sqrt{n}} \sim t_{n-1} \quad df = n - 1$		
R.R. & A.R. of $H_0$			
Critical value(s)	$t_{1-\frac{\alpha}{2}}$ and $-t_{1-\frac{\alpha}{2}}$	$t_{1-\alpha}$	$-t_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$T_{T.S} > t_{1-\frac{\alpha}{2}}$ or $T_{T.S} < -t_{1-\frac{\alpha}{2}}$	$T_{T.S} > t_{1-\alpha}$	$T_{T.S} < -t_{1-\alpha}$
	T.S. $\in$ R.R. Two - Sided Test	T.S. $\in$ R.R. One - Sided Test	T.S. $\in$ R.R. One - Sided Test

## Example : First case (Variance $\sigma^2$ known )

A random sample of 100 recorded deaths in the United States during the past year showed an average of 71.8 years. Assuming a population standard deviation of 8.9 year . Does this seem to indicate that the mean life span today is greater than 70 years? Use a 0.05 level of significance.

### Solution:

$$n = 100 \text{ (large } n \geq 30 \text{ )}$$

$$\sigma = 8.9 \text{ ( } \sigma \text{ known )}$$

$$\bar{X} = 71.8$$

$$\alpha = 0.05 \text{ (level of significance )}$$

$$\mu = \text{average (mean )life span}$$

$$\mu_0 = 70$$

#### First case:

##### Assumptions:

- The variance  $\sigma^2$  is known .
- Non-normal distribution with large sample size

## Hypotheses:

$$H_0 : \mu \leq 70 \quad (\mu_0 = 70)$$
$$H_A : \mu > 70 \quad (\text{research hypothesis})$$

## Test statistics (T.S.)

$$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}} = \frac{71.8 - 70}{8.9 / \sqrt{100}} = 2.02$$

## Rejection Region of $H_0$ (R.R.): (critical region): $(1.645, \infty)$

$$\alpha = 0.05 \gg \text{Critical Value : } Z_{1-\alpha} = Z_{1-0.05} = Z_{0.95} = 1.645$$

## Decision :

Reject  $H_0: \mu \leq 70$  if :

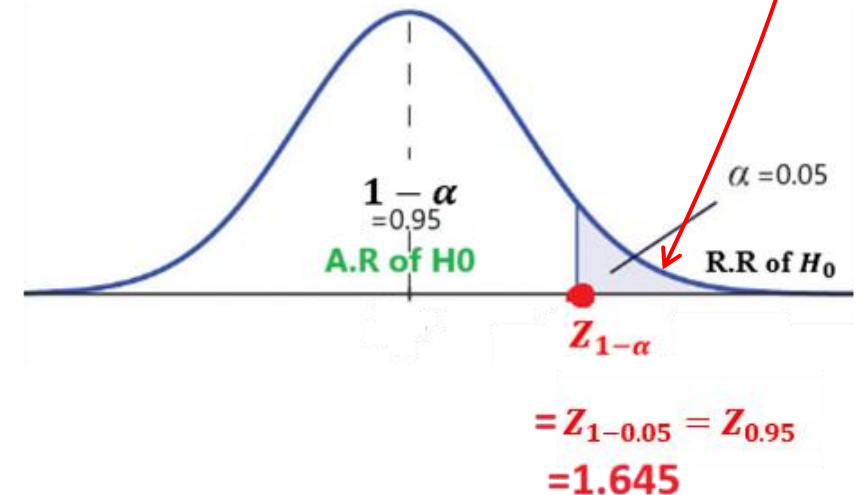
$$Z_{T.S} > Z_{1-\alpha}$$

$$2.02 > 1.645 \quad \text{condition satisfied}$$

**Decision is** Reject  $H_0: \mu \leq 70$  and Accept  $H_A: \mu > 70$

Therefore, we conclude that the mean life span today is greater than 70 years.

**Another way to take decision is by graph :**  
Determine test statistics value on the graph



### Third way we can use to take decision by Using P- Value as a decision tool:

**P-value** : is the smallest value of  $\alpha$  for which we can reject the null hypothesis  $H_0$ .

### Calculate P-value :

- Calculating P-value depends on the **alternative hypothesis  $H_A$** .
- Suppose that  $Z_c = Z_{T.S} = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}}$  is the computed **value of the test Statistic**.
- The following table illustrates how to compute P-value, and how to use P-value for testing the null hypothesis:

Alternative Hypothesis ( $H_A$ )	$H_A : \mu \neq \mu_0$	$H_A : \mu > \mu_0$	$H_A : \mu < \mu_0$
<b>P-Value =</b>	$2 \times P(Z >   Z_c  )$	$P(Z > Z_c )$	$P( Z < Z_c )$
<b>Significance Level =</b>	$\alpha$		
<b>Decision:</b>	Reject $H_0$ if P-value $\leq \alpha$		

## Example : For the previous example

### Hypotheses:

$$H_0 : \mu \leq 70 \quad (\mu_0 = 70)$$

$$H_A : \mu \geq 70 \quad (\text{research hypothesis})$$

### Test statistics (T.S.)

$$Z = \frac{\bar{X} - \mu_0}{\sigma / \sqrt{n}} = \frac{71.8 - 70}{8.9 / \sqrt{100}} = 2.02$$

To calculate P- value    The alternative hypothesis was  $H_A: \mu \geq 70$  ,  $\alpha = 0.05$

$$\text{P-value} = P(Z \geq Z_c) = P(Z > 2.02) = 1 - P(Z < 2.02) = 1 - 0.9783 = 0.0217$$

### Decision :

Reject  $H_0$  if P-value  $< \alpha = 0.05$

$$0.0217 < \alpha = 0.05 \quad (\text{condition satisfied})$$

Then , we Reject  $H_0$  .

## Example: Second case (Variance $\sigma^2$ unknown )

The manager of a private clinic claims that the mean time of the patient-doctor visit in his clinic is 8 minutes. Test the hypothesis that  $\mu=8$  minutes against the alternative that  $\mu \neq 8$  minutes . If a random sample of 25 patient-doctor visits yielded a mean time of 7.8 minutes with a standard deviation of 0.5 minutes. It is assumed that the distribution of the time of this type of visits is normal. Use a 0.01 level of significance.

### Solution:

Normal distribution ,

$$\bar{X} = 7.8 ,$$

$\alpha=0.01$  (level of significance )

$S = 0.5$  Sample standard deviation ( $\sigma$  is unknown)

$n = 25$  (  $n$  small  $n < 30$  )

$\mu$  = mean time of the visit

Second case:

#### Assumptions:

- The variance  $\sigma^2$  is unknown .
- Normal distribution
- Any sample size .

## Hypotheses:

$$H_0 : \mu = 8 \quad (\mu_0 = 8)$$
$$H_A : \mu \neq 8 \quad (\text{research hypothesis})$$

## Test statistics (T.S.)

$$T = \frac{\bar{X} - \mu_0}{S/\sqrt{n}} = \frac{7.8 - 8}{0.5/\sqrt{25}} = -2$$

Rejection Region of  $H_0$  (R.R.): (critical region) :  $(-\infty, -2.797) \cup (2.797, \infty)$

$$\alpha = 0.01 \gg \text{Critical Value : } \pm t_{1-\frac{\alpha}{2}} = \pm t_{1-\frac{0.01}{2}} = \pm t_{0.995} = \pm 2.797$$

$$\text{df} = n-1 = 25-1 = 24$$

## Decision :

Reject  $H_0: \mu = 8$  if :

$$T_{T.S} < -T_{1-\frac{\alpha}{2}} \quad \text{or} \quad T_{T.S} > T_{1-\frac{\alpha}{2}}$$

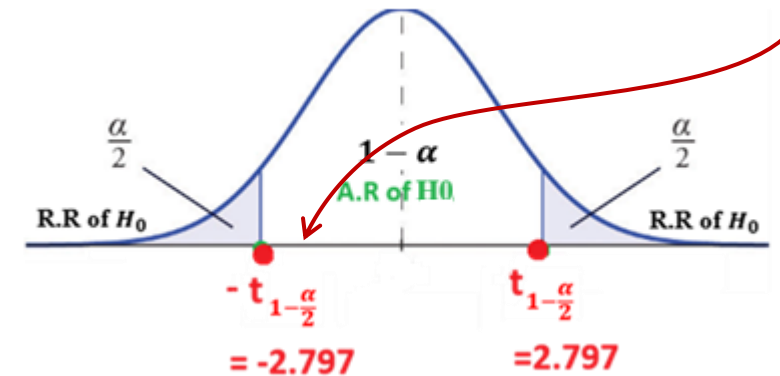
$-2 \not< -2.797$  (First condition not satisfied, then try the second condition)

$-2 \not> 2.797$  (Second condition not satisfied)

**Decision is** Accept  $H_0: \mu = 8$  and Reject  $H_A: \mu \neq 8$

Therefore, we conclude that the claim is correct.

Another way to take decision is by graph :  
Determine test statistics value on the graph



## Special case :

For the case of **non-normal population with unknown variance**, and when the sample size is **large ( $n \geq 30$ )**,

we may use the following test statistic:

$$Z = \frac{\bar{X} - \mu_0}{s/\sqrt{n}} \sim N(0,1)$$

That is, we replace the population standard deviation ( $\sigma$ ) by the sample standard deviation (S), and we conduct the test as described for the first case.

### 7.3 Hypothesis Testing: The Difference Between Two Population Means ( $\mu_1 - \mu_2$ ) : (Independent Populations)

Suppose that we have two (independent) populations:

- 1-st population with mean  $\mu_1$  and variance  $\sigma_1^2$  .
- 2-nd population with mean  $\mu_2$  and variance  $\sigma_2^2$  .
- We are interested in comparing  $\mu_1$  and  $\mu_2$  , or equivalently, making inferences about the difference between the means ( $\mu_1 - \mu_2$ ).
- We independently select a random sample of size  $n_1$  from the 1-st population and another random sample of size  $n_2$  from the 2-nd population:
- Let  $\bar{X}_1$  and  $S_1^2$  be the sample mean and the sample variance of the 1-st sample.
- Let  $\bar{X}_2$  and  $S_2^2$  be the sample mean and the sample variance of the 2-nd sample.
- The sampling distribution of  $\bar{X}_1 - \bar{X}_2$  is used to make inferences about  $\mu_1 - \mu_2$ .

We wish to test some hypotheses comparing the population means.

# Hypotheses:

We choose one of the following situations:

- (i)  $H_0: \mu_1 = \mu_2$  against  $H_A: \mu_1 \neq \mu_2$
- (ii)  $H_0: \mu_1 \geq \mu_2$  against  $H_A: \mu_1 < \mu_2$
- (iii)  $H_0: \mu_1 \leq \mu_2$  against  $H_A: \mu_1 > \mu_2$

or equivalently,

- (i)  $H_0: \mu_1 - \mu_2 = 0(\mu_0)$  against  $H_A: \mu_1 - \mu_2 \neq 0(\mu_0)$
- (ii)  $H_0: \mu_1 - \mu_2 \geq 0(\mu_0)$  against  $H_A: \mu_1 - \mu_2 < 0(\mu_0)$
- (iii)  $H_0: \mu_1 - \mu_2 \leq 0(\mu_0)$  against  $H_A: \mu_1 - \mu_2 > 0(\mu_0)$

## Test Statistic (T.S):

### (1) First Case:

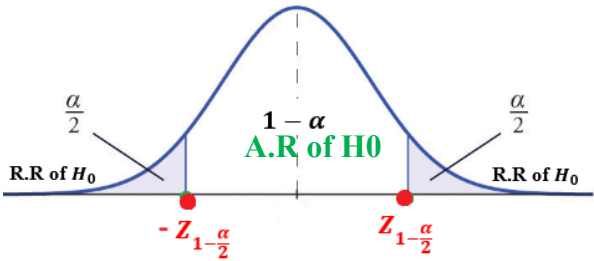
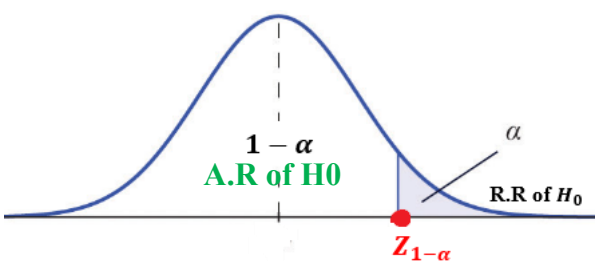
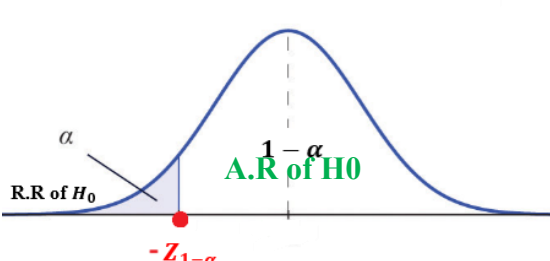
#### Assumptions:

- Normal populations +  $\sigma_1^2$  and  $\sigma_2^2$  are known.
- Non-normal populations with large sample sizes ( $n_1, n_2 \geq 30$ ) +  $\underline{\sigma_1^2}$  and  $\underline{\sigma_2^2}$  are known

The test statistic is:

$$Z = \frac{(\bar{X}_1 - \bar{X}_2) - \mu_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0,1)$$

Test procedures:

Hypotheses	$H_0 : \mu_1 - \mu_2 = \mu_0$ $H_A : \mu_1 - \mu_2 \neq \mu_0$	$H_0 : \mu_1 - \mu_2 \leq \mu_0$ $H_A : \mu_1 - \mu_2 > \mu_0$	$H_0 : \mu_1 - \mu_2 \geq \mu_0$ $H_A : \mu_1 - \mu_2 < \mu_0$
First case	if $\sigma_1^2$ and $\sigma_2^2$ are known , Normal populations or $\sigma_1^2$ and $\sigma_2^2$ are known , non-normal populations with large sample sizes		
Test Statistic (T.S.)	$Z = \frac{(\bar{X}_1 - \bar{X}_2) - \mu_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0,1)$		
R.R. & A.R. of $H_0$			
Critical value(s)	$Z_{1-\frac{\alpha}{2}}$ and $-Z_{1-\frac{\alpha}{2}}$	$Z_{1-\alpha}$	$-Z_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$Z_{T.S} > Z_{1-\frac{\alpha}{2}}$ or $Z_{T.S} < -Z_{1-\frac{\alpha}{2}}$	$Z_{T.S} > Z_{1-\alpha}$	$Z_{T.S} < -Z_{1-\alpha}$
	T.S. $\in$ R.R. Two - Sided Test	T.S. $\in$ R.R. One - Sided Test	T.S. $\in$ R.R. One - Sided Test

## (2) Second Case:

### Assumptions:

- Normal populations .
- $\sigma_1^2$  and  $\sigma_2^2$  are unknown but equal  $\sigma_1^2 = \sigma_2^2 = \sigma^2$ ,
- small sample size ( $n_1, n_2 < 30$ )

then the **test statistic** is:

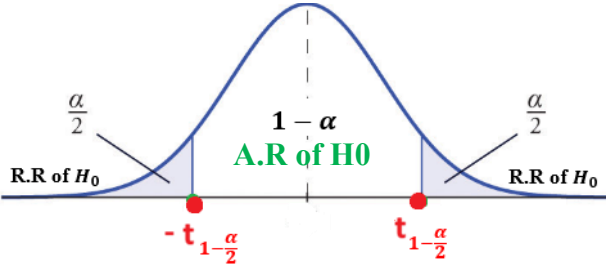
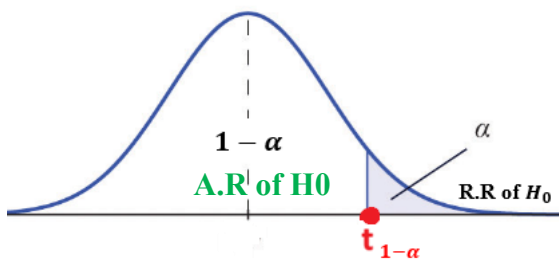
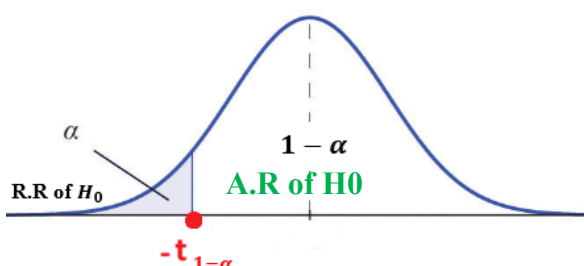
$$T = \frac{(\bar{X}_1 - \bar{X}_2) - \mu_0}{\sqrt{\frac{s_P^2}{n_1} + \frac{s_P^2}{n_2}}} \sim t(n_1 + n_2 - 2)$$

where the pooled variance estimate of  $\sigma^2$  is

$$S_P^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}$$

and the degrees of freedom of is  $df = v = n_1 + n_2 - 2$ .

# Test procedures:

Hypotheses	$H_0 : \mu_1 - \mu_2 = \mu_0$ $H_A : \mu_1 - \mu_2 \neq \mu_0$	$H_0 : \mu_1 - \mu_2 \leq \mu_0$ $H_A : \mu_1 - \mu_2 > \mu_0$	$H_0 : \mu_1 - \mu_2 \geq \mu_0$ $H_A : \mu_1 - \mu_2 < \mu_0$
Second case	if $\sigma_1^2 = \sigma_2^2 = \sigma^2$ is unknown + Normal populations + $n_1, n_2 < 30$		
Test Statistic (T.S.)	$T = \frac{(\bar{X} - \bar{X}_2) - \mu_0}{\sqrt{\frac{S_P^2}{n_1} + \frac{S_P^2}{n_2}}} \sim t_{n_1 + n_2 - 2}, \quad S_P^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}, \quad df = n_1 + n_2 - 2$		
R.R. & A.R. of $H_0$			
Critical value(s)	$t_{1-\frac{\alpha}{2}}$ and $-t_{1-\frac{\alpha}{2}}$	$t_{1-\alpha}$	$-t_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$T_{T.S} > t_{1-\frac{\alpha}{2}}$ or $T_{T.S} < -t_{1-\frac{\alpha}{2}}$	$T_{T.S} > t_{1-\alpha}$	$T_{T.S} < -t_{1-\alpha}$
	T.S. $\in$ R.R. Two - Sided Test	T.S. $\in$ R.R. One - Sided Test	T.S. $\in$ R.R. One - Sided Test

**Example: (  $\sigma_1^2$  and  $\sigma_2^2$  are known )**

Researchers wish to know if the data they have collected provide sufficient evidence to indicate the difference in mean serum uric acid levels between individuals with Down's syndrome(1) and normal individuals(2). The data consist of serum uric acid on 12 individuals with Down's syndrome and 15 normal individuals.

The sample means are  $\bar{X}_1 = 4.5$  mg/100 ml and  $\bar{X}_2 = 3.4$  mg/100 ml. Assume the populations are normal with variances  $\sigma_1^2 = 1$  and  $\sigma_2^2 = 1.5$ . Use significance level  $\alpha=0.05$ .

**Solution:**

$\mu_1$  = mean serum uric acid levels for the individuals with Down's syndrome.

$\mu_2$  = mean serum uric acid levels for the normal individuals.

$$n_1 = 12, \bar{X}_1 = 4.5, \sigma_1^2 = 1 \quad (\sigma_1^2 \text{ known})$$

$$n_2 = 15, \bar{X}_2 = 3.4, \sigma_2^2 = 1.5 \quad (\sigma_2^2 \text{ known})$$

Normal populations

**First case:**

**Assumptions:**

- The variance  $\sigma_1^2, \sigma_2^2$ , is known .
- Normal distribution .

## Hypotheses:

$$\begin{array}{l} H_0 : \mu_1 = \mu_2 \quad \text{OR} \quad (H_0 : \mu_1 - \mu_2 = 0) \\ H_A : \mu_1 \neq \mu_2 \quad (H_0 : \mu_1 - \mu_2 \neq 0) \end{array}$$

## Test statistics (T.S.)

$$Z = \frac{(\bar{X}_1 - \bar{X}_2) - \mu_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} = \frac{4.5 - 3.4}{\sqrt{\frac{1}{12} + \frac{1.5}{15}}} = 2.569$$

Rejection Region of  $H_0$  (R.R.): (critical region):  $(-\infty, -1.96) \cup (1.96, \infty)$

$$\alpha = 0.05 \gg \text{Critical Value : } \pm Z_{1-\frac{\alpha}{2}} = \pm Z_{1-\frac{0.05}{2}} = \pm Z_{0.975} = \pm 1.96$$

## Decision :

Reject  $H_0 : \mu_1 = \mu_2$  if :

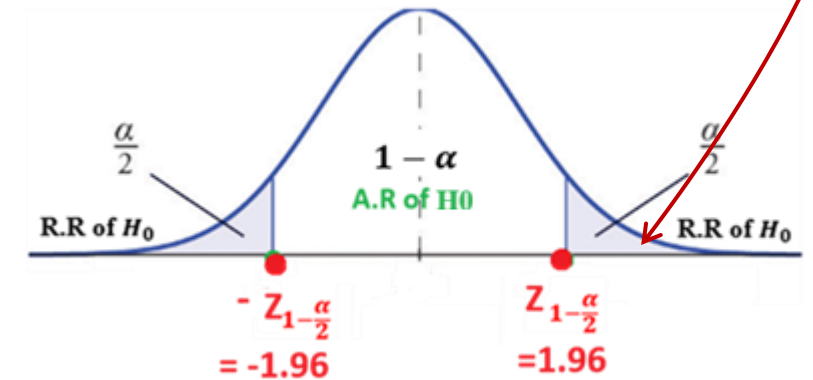
$$Z_{T.S} < -Z_{1-\frac{\alpha}{2}} \quad \text{or} \quad Z_{T.S} > Z_{1-\frac{\alpha}{2}}$$

$2.569 < -1.96$  (First condition not satisfied, then try the second condition)

$2.569 > 1.96$  (Second condition satisfied)

**Decision is** Reject  $H_0 : \mu_1 = \mu_2$  and Accept  $H_A : \mu_1 \neq \mu_2$   
we conclude that the two populations means are not equal.

Another way to take decision is by graph :  
Determine test statistics value on the graph



# Decision by using P-value :

Calculate the P-value

$$\begin{aligned} \text{P-value} &= 2 \times P(Z > |Z_c|) \\ &= 2 \times P(Z > |2.569|) \\ &= 2 \times P(Z > 2.57) \\ &= 2 \times [1 - P(Z < 2.57)] \\ &= 2 \times [1 - 0.99492] \\ &= 2 \times 0.00508 \\ &= 0.01016 \end{aligned} \quad (\text{where } Z_c = Z(\text{Test statistic}) = 2.569)$$

$$\text{P-value} = 0.01016$$

Decision : **Reject  $H_0$  if P-value  $< \alpha = 0.05$**

Since P-value = 0.0106  $< \alpha = 0.05$

We reject  $H_0$  and accept  $H_A$

**Example:** ( $\sigma_1^2 = \sigma_2^2 = \sigma^2$  is unknown)

An experiment was performed to compare the abrasive wear of two different materials used in making artificial teeth. 12 pieces of material(1) were tested by exposing each piece to a machine measuring wear. 10 pieces of material(2) were similarly tested. In each case, the depth of wear was observed. The samples of material(1) gave an average wear of 85 units with a sample standard deviation of 4, while the samples of materials(2) gave an average wear of 81 and a sample standard deviation of 5.

Can we conclude at the 0.05 level of significance that the mean abrasive wear of material(1) is greater than that of material(2)? Assume normal populations with equal variances.

**Solution:**

Normal populations  
 $\alpha=0.05$  (level of significance)  
 $\sigma_1^2 = \sigma_2^2 = \sigma^2$  is unknown

Material(1)	Material(2)
$n_1 = 12$	$n_2 = 10$
$\bar{X}_1 = 85$	$\bar{X}_2 = 81$
$S_1 = 4$	$S_2 = 5$

**Second case:**

**Assumptions:**

- The variance  $\sigma_1^2, \sigma_2^2$ , is Unknown and equal
- Normal distribution .
- $n_1, n_2 < 30$  small

## Hypotheses:

$$\begin{array}{ll} H_0 : \mu_1 \leq \mu_2 & \text{OR} \quad (H_0: \mu_1 - \mu_2 \leq 0 \quad \mu_0 = 0) \\ H_A : \mu_1 > \mu_2 & (H_0: \mu_1 - \mu_2 > 0 \quad \mu_0 = 0) \end{array}$$

$$\text{Test statistics (T.S.) } T = \frac{(\bar{X}_1 - \bar{X}_2) - \mu_0}{\sqrt{\frac{s_p^2}{n_1} + \frac{s_p^2}{n_2}}} = \frac{85 - 81}{\sqrt{\frac{20.05}{12} + \frac{20.05}{10}}} = 2.09$$

$$\text{pooled variance } (S_p^2) : S_p^2 = \frac{(n_1 - 1) \times S_1^2 + (n_2 - 1) \times S_2^2}{n_1 + n_2 - 2} = \frac{(12 - 1) \times (4^2) + (10 - 1) \times (5^2)}{12 + 10 - 2} = 20.05$$

**Rejection Region of  $H_0$  (R.R.):** (critical region :  $H_A$  one sided)  $(1.725, \infty)$

$$\alpha = 0.05 \gg \text{Critical Value : } t_{1-\alpha} = t_{1-0.05} = t_{0.95} = 1.725$$

$$\text{df} = n_1 + n_2 - 2 = 12 + 10 - 2 = 20$$

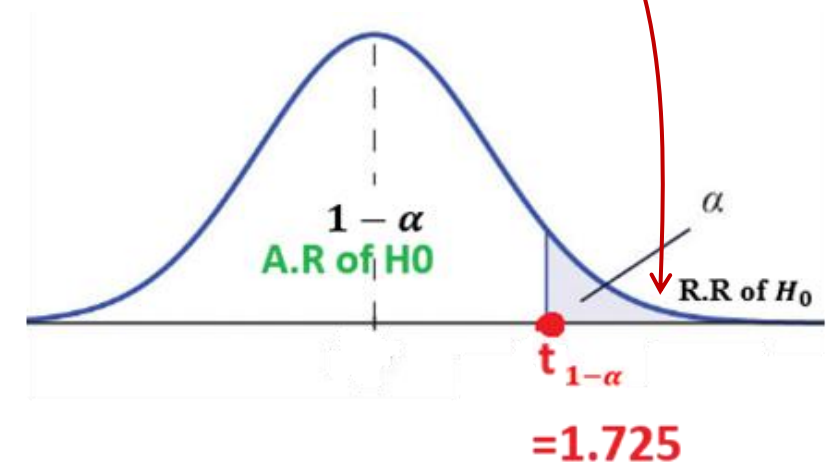
## Decision :

$$\begin{array}{ll} \text{Reject } H_0: \mu_1 \leq \mu_2 \text{ if :} & T_{T.S} > t_{1-\alpha} \\ & 2.09 > 1.725 \text{ (condition satisfied)} \end{array}$$

**Decision is** Reject  $H_0: \mu_1 \leq \mu_2$  and Accept  $H_A: \mu_1 > \mu_2$

Therefore, we conclude that the mean abrasive wear of material (1) is greater than that of material (2).

**Another way to take decision is by graph :**  
Determine test statistics value on the graph



## 7.4 Paired Comparisons:

- In this section, we are interested in comparing the means of two related (non-independent/dependent) normal populations.
- In other words, we wish to make statistical inference for the difference between the means of two related normal populations.
- Paired t-Test concerns about testing the equality of the means of two related normal populations.

### **Examples of related (dependent) populations are:**

1. Height of the father and height of his son.
2. Mark of the student in MATH and his mark in STAT.
3. Pulse rate of the patient before and after the medical treatment.
4. Hemoglobin level of the patient before and after the medical treatment.

# Test procedure

**Let**

$X$  : the values of the first population .

$Y$  : the values of the second population .

$D = \text{Values of } X - \text{Values of } Y$

**Means**

$\mu_1$ : Mean of the first population .

$\mu_2$ : Mean of the second population .

$\mu_D = \text{Mean of } X - \text{Mean of } Y$

$$( \mu_D = \mu_1 - \mu_2 )$$

Confident interval about a difference of two population means (  $\mu_D = \mu_1 - \mu_2$  ) :  
(Dependent , Related populations)

calculate the following quantities: Or (Can use the calculator)	<p>-The differences (D-observations): <math>D_i = X_i - Y_i</math> (i=1, 2, ..., n)</p> <p>-Sample mean of the D-observations: <math>\bar{D} = \frac{\sum_{i=1}^n D_i}{n}</math></p> <p>-Sample variance of the D-observations: <math>S_D^2 = \frac{\sum_{i=1}^n (D_i - \bar{D})^2}{n-1}</math></p> <p>-Sample standard deviation of the D-observations: <math>S_D = \sqrt{S_D^2}</math></p>
Confident interval for $\mu_D = \mu_1 - \mu_2$	
A 100% confidence interval for $\mu_D$	$\bar{D} \pm t_{1-\frac{\alpha}{2}} \frac{S_D}{\sqrt{n}}, \quad df = n-1$

# Testing hypothesis about a difference of two population means ( $\mu_D = \mu_1 - \mu_2$ ) : (Dependent , Related populations)

Hypothesis	$H_o: \mu_1 - \mu_2 = \mu_0$ vs $H_A: \mu_1 - \mu_2 \neq \mu_0$ Or $H_o: \mu_D = \mu_0$ vs $H_A: \mu_D \neq \mu_0$	$H_o: \mu_1 - \mu_2 \leq \mu_0$ vs $H_A: \mu_1 - \mu_2 > \mu_0$ Or $H_o: \mu_D \leq \mu_0$ vs $H_A: \mu_D > \mu_0$	$H_o: \mu_1 - \mu_2 \geq \mu_0$ vs $H_A: \mu_1 - \mu_2 < \mu_0$ Or $H_o: \mu_D \geq \mu_0$ vs $H_A: \mu_D < \mu_0$
Test statistic(T.S)	$T = \frac{\bar{D} - (\mu_0)}{s/\sqrt{n}}$ , df =n-1		
R.R. and A.R. of Ho			
Critical value (s)	$t_{(1-\frac{\alpha}{2})}$ and $-t_{(1-\frac{\alpha}{2})}$	$t_{1-\alpha}$	$-t_{1-\alpha}$
Decision :	Reject Ho (and accept HA) at the significance level $\alpha$ if:		
Reject Ho if the following condition satisfied.	$T_{T.S} > t_{1-\frac{\alpha}{2}}$ or $T_{T.S} < -t_{1-\frac{\alpha}{2}}$	$T_{T.S} > t_{1-\alpha}$	$T_{T.S} < -t_{1-\alpha}$
	T.S. $\in$ R.R. Two-Sided Test	T.S. $\in$ R.R. One-Sided Test	T.S. $\in$ R.R. One-Sided Test

## Example (effectiveness of a diet program)

Suppose that we are interested in study the effectiveness of a certain diet program on ten individual . Let the random variable X and Y given as following table :

Individual (i)	1	2	3	4	5	6	7	8	9	10
Weight before (Xi)	86.6	80.2	91.5	80.6	82.3	81.9	88.4	85.3	83.1	82.1
Weight after (Yi)	79.7	85.9	81.7	82.5	77.9	85.8	81.3	74.7	68.3	69.7

Assume that the populations are normal.

1. Find a 95% confidence interval for the difference between the mean of weights before the diet program ( $\mu_1$ ) and the mean of weights after the diet program ( $\mu_2$ ) [ $\mu_D = \mu_1 - \mu_2$ ].
2. Does these data provide sufficient evidence to allow us to conclude that the diet program is effective ? Use  $\alpha=0.05$  .

## Solution:

Let the random variables  $X$  and  $Y$  are as follows:

$X$  = the weight of the individual **before** the diet program

$Y$  = the weight of the same individual **after** the diet program.

We assume that the distributions of these random variables are normal with means  $\mu_1$  and  $\mu_2$ , respectively .

## Populations:

1-st population ( $X$ ): weights **before** a diet program mean

2-nd population ( $Y$ ): weights **after** the diet program mean

These two variables are related (dependent/non-independent) because they are measured on the same individual.

We select a random sample of  $n$  individuals. At the beginning of the study, we record the individuals' weights before the diet program ( $X$ ). At the end of the diet program, we record the individuals' weights after the program ( $Y$ ). We end up with the following information and calculations:

Individual $I$	Weight before $X_i$	Weight after $Y_i$	Difference $D_i = X_i - Y_i$
1	$X_1$	$Y_1$	$D_1 = X_1 - Y_1$
2	$X_2$	$Y_2$	$D_2 = X_2 - Y_2$
.	.	.	.
.	.	.	.
.	.	.	.
$n$	$X_n$	$Y_n$	$D_n = X_n - Y_n$

Find the following measures by calculator or rules

- The sample mean of the D-observations:  $\bar{D} = \frac{\sum_{i=1}^n D_i}{n} = \frac{54.5}{10} = 5.45$
- Sample variance of the D-observations:  $S_D^2 = \frac{\sum_{i=1}^n (D_i - \bar{D})^2}{n-1} = \frac{(6.9-5.45)^2 + \dots + (12.4-5.45)^2}{10-1} = 50.328$
- Sample standard deviation of the D-observations:  $S_D = \sqrt{S_D^2} = \sqrt{50.328} = 7.09$

**First :**

Calculate difference  $D_i = X_i - Y_i$   
(Last column)

**Then, From calculator,**

-Enter the data in the last column  
( $D_i = X_i - Y_i$ )

- Find the sample mean of the

D-observations ( $\bar{D} = \frac{\sum_{i=1}^n D_i}{n} = \frac{54.5}{10} = 5.45$ )

-Find the sample standard deviation of the  
D-observations ( $S_D = 7.09$ )

$i$	$X_i$	$Y_i$	$D_i = X_i - Y_i$
1	86.6	79.7	6.9
2	80.2	85.9	-5.7
3	91.5	81.7	9.8
4	80.6	82.5	-1.9
5	82.3	77.9	4.4
6	81.9	85.8	-3.9
7	88.4	81.3	7.1
8	85.3	74.7	10.6
9	83.1	68.3	14.8
10	82.1	69.7	12.4
Sum			$\sum_{i=1}^{10} D_i = 54.5$

1. We need to find a 95% confidence interval for  $\mu_D = \mu_1 - \mu_2$

Reliability coefficient  $t_{1-\frac{\alpha}{2}}$  :

$$\alpha = 0.05, \quad t_{1-\frac{\alpha}{2}} = t_{1-\frac{0.05}{2}} = t_{0.975} = 2.262 \quad (\text{df} = n - 1 = 10 - 1 = 9)$$

The 95% C.I for  $\mu_D = \mu_1 - \mu_2$

$$\bar{D} \pm t_{1-\alpha/2} \frac{S_D}{\sqrt{n}}$$

$$5.45 \pm 2.262 \times \frac{7.09}{\sqrt{10}}$$

$$5.45 \pm 5.0715$$

$$(5.45 - 5.0715, 5.45 + 5.0715)$$

$$(0.38, 10.52)$$

$$0.38 < \mu_D < 10.52 \quad \gg \quad \mu_D(0.38, 10.52)$$

**Note :** Since the confidence interval does not include zero,  $0 \notin (0.38, 10.52)$  ( $\mu_1 - \mu_2 \neq 0 \Leftrightarrow \mu_1 \neq \mu_2$ )

Mean of weights **before** a diet program  $\neq$  Mean of weights **after** a diet program mean

## 2. Does these data provide sufficient evidence to allow us to conclude that the diet program is effective?

Use  $\alpha=0.05$  and assume that the populations are normal.

$\mu_1$  =Mean of weights **before** a diet program mean

$\mu_2$  =Mean of weights **after** the diet program mean

$\mu_D$  =Mean of X – Mean of Y ( $\mu_D = \mu_1 - \mu_2$ )

### Hypotheses:

$H_0$ : the diet program has **no effect** on weight

$H_A$ : the diet program has **an effect** on weight

Equivalently,

$H_0: \mu_1 = \mu_2$  (**no effect**) VS  $H_A: \mu_1 \neq \mu_2$  (**There is effect**)

$H_0: \mu_1 - \mu_2 = 0$  VS  $H_A: \mu_1 - \mu_2 \neq 0$

$H_0: \mu_D = 0$  VS  $H_A: \mu_D \neq 0$  (*where :  $\mu_D = \mu_1 - \mu_2$* )

## Hypotheses:

$$\begin{array}{l} H_0 : \mu_1 = \mu_2 \\ H_A : \mu_1 \neq \mu_2 \end{array} \quad \text{OR} \quad \begin{array}{l} (H_0 : \mu_1 - \mu_2 \neq 0) \\ (H_A : \mu_1 - \mu_2 \neq 0) \end{array} \quad \text{OR} \quad \begin{array}{l} H_0 : \mu_D \neq 0 \\ H_A : \mu_D \neq 0 \end{array}$$

$$\text{Test statistics (T.S.) } T = \frac{\bar{D} - \mu_0}{s_D / \sqrt{n}} = \frac{5.45 - 0}{7.09 / \sqrt{10}} = 2.43$$

## Rejection Region of $H_0$ (R.R.): (critical region : $H_A$ Two sided)

$$\alpha = 0.05 \gg \text{Critical Value : } \pm t_{1-\frac{\alpha}{2}} = \pm t_{1-\frac{0.05}{2}} = \pm t_{0.975} = \pm 2.262$$

$$df = n - 1 = 10 - 1 = 9$$

## Decision :

$$\text{Reject } H_0 : \mu_1 = \mu_2 \text{ if : } T_{T.S} < -t_{1-\frac{\alpha}{2}} \text{ or } T_{T.S} > t_{1-\frac{\alpha}{2}}$$

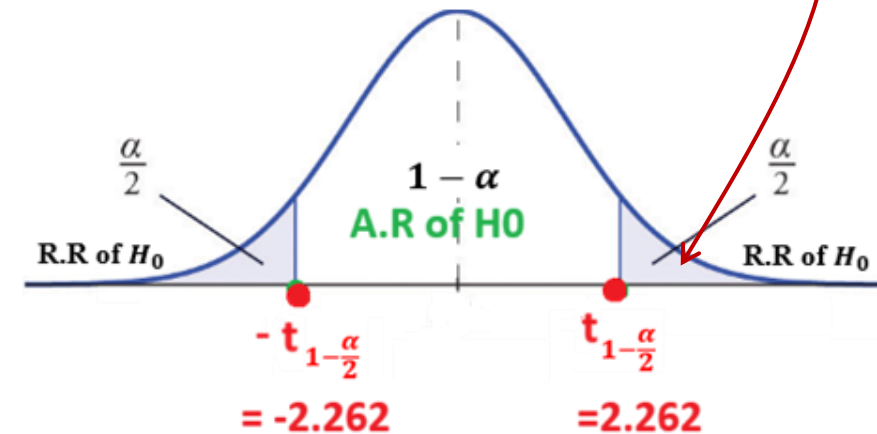
$$2.43 < -2.262 \text{ (First condition not satisfied, then try the second condition)}$$

$$2.43 > 2.262 \text{ (Second condition satisfied)}$$

**Decision is** Reject  $H_0 : \mu_1 = \mu_2$  (no effect) and Accept  $H_A : \mu_1 \neq \mu_2$  (effect)

We conclude that there is effect of the diet program.

Another way to take decision is by graph :  
Determine test statistics value on the graph



**Note:**

The **sample mean** of the weights **before** the program is ( $\bar{X} = 84.2$  )

The **sample mean** of the weights **after** the program is ( $\bar{Y} = 78.75$  )

Since the diet program is effective and since

$$\bar{X} > \bar{Y}$$

$$84.2 > 78.75$$

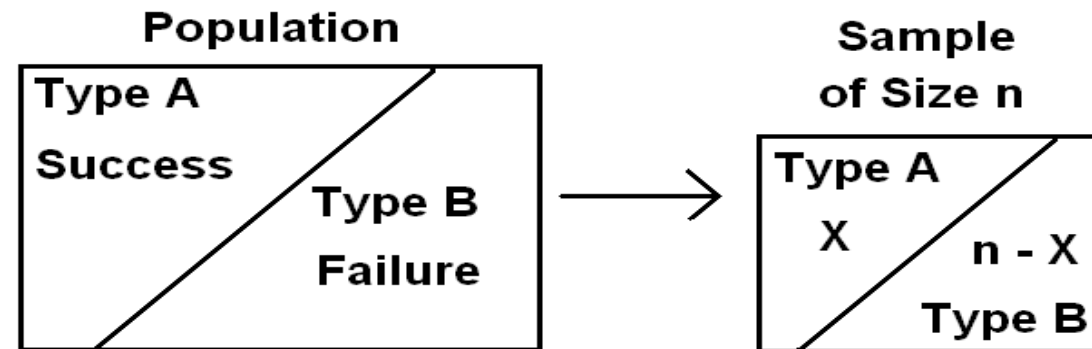
**Weight before > weight after**

we can conclude that the program is effective in reducing the weight (So, we have a good diet program).

$i$	$X_i$	$Y_i$
1	86.6	79.7
2	80.2	85.9
3	91.5	81.7
4	80.6	82.5
5	82.3	77.9
6	81.9	85.8
7	88.4	81.3
8	85.3	74.7
9	83.1	68.3
10	82.1	69.7
Sample Mean	$\bar{X} = 84.2$	$\bar{Y} = 78.75$

## 7.5 Hypothesis Testing: A Single Population Proportion (p):

In this section, we are interested in testing some hypotheses about the population proportion (p).



Recall:

$p$  = Population proportion of elements of Type A in the population.

$$P = \frac{\text{No. of elements of type A in the population}}{\text{Total number of elements in the population}} = \frac{A}{N} \quad (N = \text{Population size})$$

$$\hat{P} = \frac{\text{No. of elements of type A in the sample}}{\text{Total number of elements in the sample}} = \frac{X}{n} \quad (n = \text{Sample size})$$

$\hat{P}$  is a "good" point estimate for  $p$ .

From Chapter 5:

For large  $n$  , (  $n \geq 30$  ,  $np > 5$  ,  $nq > 5$  )

$$Z = \frac{\hat{P} - P}{\sqrt{\frac{P(1-P)}{n}}} \sim N(0,1)$$

# Test procedures:

Hypotheses	$H_0 : P = P_0$ $H_A : P \neq P_0$	$H_0 : P \leq P_0$ $H_A : P > P_0$	$H_0 : P \geq P_0$ $H_A : P < P_0$
Test Statistic (T.S.)	$Z = \frac{\hat{P} - P_0}{\sqrt{\frac{P_0 q_0}{n}}} \sim N(0, 1)$ <p> <math>P_0</math> : be a given known value and <math>q_0 = 1 - P_0</math>  <math>\hat{P}</math> = Sample propotion         </p>		
R.R. & A.R. of $H_0$			
Critical value(s)	$Z_{1-\frac{\alpha}{2}}$ and $-Z_{1-\frac{\alpha}{2}}$	$Z_{1-\alpha}$	$-Z_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$Z_{T.S} > Z_{1-\frac{\alpha}{2}}$ or $Z_{T.S} < -Z_{1-\frac{\alpha}{2}}$	$Z_{T.S} > Z_{1-\alpha}$	$Z_{T.S} < -Z_{1-\alpha}$
	<b>T.S. <math>\in</math> R.R. Two - Sided Test</b>	<b>T.S. <math>\in</math> R.R. One - Sided Test</b>	<b>T.S. <math>\in</math> R.R. One - Sided Test</b>

## Example

A researcher was interested in the proportion of females in the population of all patients visiting a certain clinic. The researcher claims that 70% of all patients in this population are females. Would you agree with this claim if a random survey shows that 24 out of 45 patients are females? Use a 0.10 level of significance.

## Solution:

$p$  = Proportion of female in the population.

$n=45$  (large)

$X$ = no. of female in the sample = 24

$\hat{P}$  = proportion of females in the sample

$$\hat{P} = \frac{X}{n} = \frac{24}{45} = 0.533$$

$\alpha=0.10$  (level of significance )

## Hypotheses:

$$H_0 : P = 0.7 \quad (P_0 = 0.7)$$

$$H_A : P \neq 0.7$$

## Test statistics (T.S.)

$$q_o = 1 - P_0 = 1 - 0.7 = 0.3$$

$$Z = \frac{\hat{P} - P_0}{\sqrt{\frac{P_0 q_0}{n}}} = \frac{0.533 - 0.7}{\sqrt{\frac{0.7 \times 0.3}{45}}} = -2.44$$

Rejection Region of  $H_0$  (R.R.): (critical region) :  $(-\infty, -1.645) \cup (1.645, \infty)$

$$\alpha = 0.10 \gg \text{Critical Value : } \pm Z_{1-\frac{\alpha}{2}} = \pm Z_{1-\frac{0.10}{2}} = \pm Z_{0.95} = \pm 1.645$$

## Decision :

Reject  $H_0 : P = 0.7$  if :

$$Z_{T.S} < -Z_{1-\frac{\alpha}{2}} \text{ or } Z_{T.S} > Z_{1-\frac{\alpha}{2}}$$

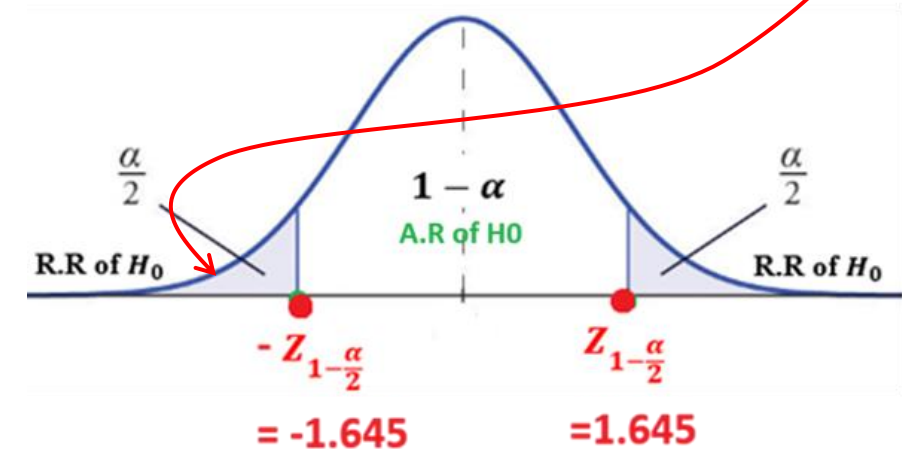
$$-2.44 < -1.645$$

(First condition satisfied, not need to try the second condition)

**Decision is** Reject  $H_0 : P = 0.7$  and Accept  $H_A : P \neq 0.7$

Therefore, we do not agree with the claim stating that 70% of the patients in this population are females.

Another way to take decision is by graph :  
Determine test statistics value on the graph



## Example (Reading)

In a study on the fear of dental care in a certain city, a survey showed that 60 out of 200 adults said that they would hesitate to take a dental appointment due to fear. Test whether the proportion of adults in this city who hesitate to take dental appointment is less than 0.25. Use a level of significance of 0.025.

### Solution:

$p$  = Proportion of adults in the city who hesitate to take a dental appointment.

$n = 200$  (large)

$X$  = no. of adults who hesitate in the sample = 60

$\hat{P}$  = proportion of adults who hesitate in the sample.

$$\hat{P} = \frac{X}{n} = \frac{60}{200} = 0.3$$

$\alpha = 0.025$  (level of significance)

## Hypotheses:

$$H_0 : P \geq 0.25 \quad (P_0 = 0.25)$$

$$H_A : P < 0.25$$

$$q_0 = 1 - P_0 = 1 - 0.25 = 0.75$$

## Test statistics (T.S.)

$$Z = \frac{\hat{P} - P_0}{\sqrt{\frac{P_0 q_0}{n}}} = \frac{0.3 - 0.25}{\sqrt{\frac{0.25 \times 0.75}{200}}} = 1.633$$

## Rejection Region of $H_0$ (R.R.): (critical region): $(-\infty, -1.96)$

$$\alpha = 0.025 \gg \text{Critical Value : } -Z_{1-\alpha} = -Z_{1-0.025} = -Z_{0.975} = -1.96$$

## Decision :

Reject  $H_0 : P = 0.7$  if :

$$Z_{T.S} < -Z_{1-\alpha}$$

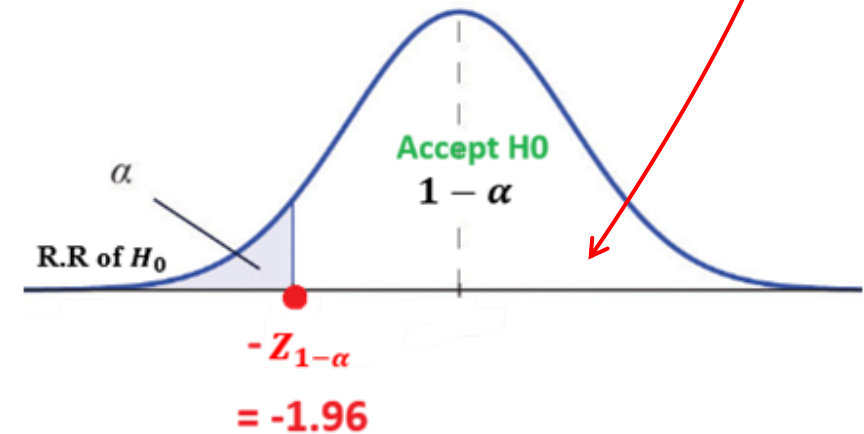
$$1.633 < -1.96$$

condition not satisfied

Decision is Accept  $H_0 : P \geq 0.25$  and Reject  $H_A : P < 0.25$

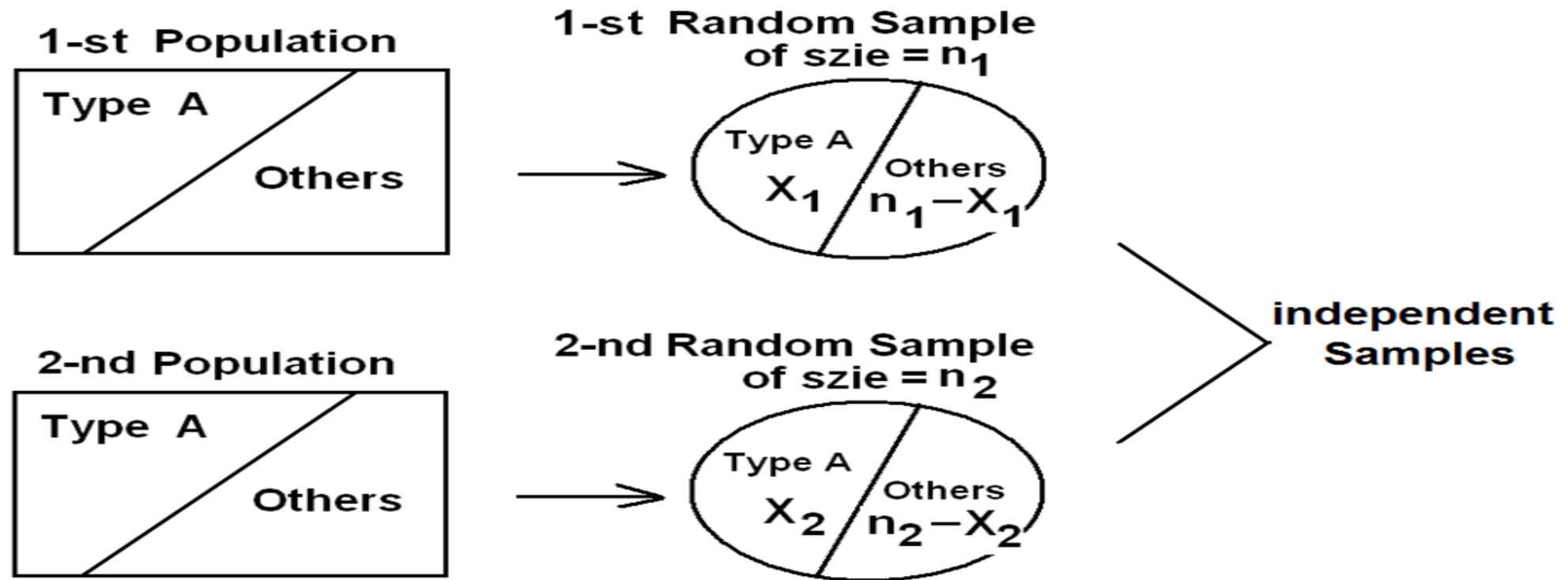
Another way to take decision is by graph :

Determine test statistics value on the graph



## 7.6 Hypothesis Testing: The Difference Between Two Population Proportions ( $P_1 - P_2$ ):

In this section, we are interested in testing some hypotheses about the difference between two population proportions ( $P_1 - P_2$ )



**Suppose that we have two populations:**

- $P_1$  = population proportion of the 1-st population.
- $P_2$  = population proportion of the 2-nd population.
- We are interested in comparing  $P_1$  and  $P_2$ , or equivalently, making inferences about  $P_1 - P_2$ .
- We independently select a random sample of size  $n_1$  from the 1-st population and another random sample of size  $n_2$  from the 2-nd population:
  - Let  $X_1$  = no. of elements of type A in the 1-st sample.
  - Let  $X_2$  = no. of elements of type A in the 2-nd sample.
  - $\hat{P}_1 = \frac{X_1}{n_1}$  the sample proportion of the 1-st sample.
  - $\hat{P}_2 = \frac{X_2}{n_2}$  the sample proportion of the 1-st sample.

## Hypotheses:

We choose one of the following situations:

- (i)  $H_0: P_1 = P_2$  against  $H_A: P_1 \neq P_2$
- (ii)  $H_0: P_1 \geq P_2$  against  $H_A: P_1 < P_2$
- (iii)  $H_0: P_1 \leq P_2$  against  $H_A: P_1 > P_2$

or equivalently,

- (i)  $H_0: P_1 - P_2 = 0$  against  $H_A: P_1 - P_2 \neq 0$
- (ii)  $H_0: P_1 - P_2 \geq 0$  against  $H_A: P_1 - P_2 < 0$
- (iii)  $H_0: P_1 - P_2 \leq 0$  against  $H_A: P_1 - P_2 > 0$

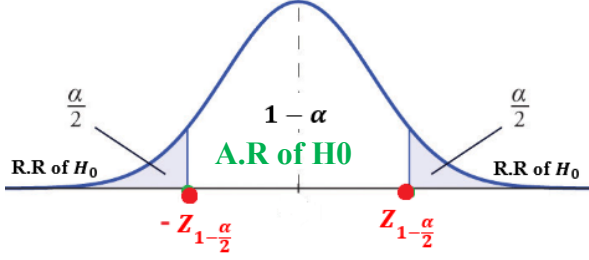
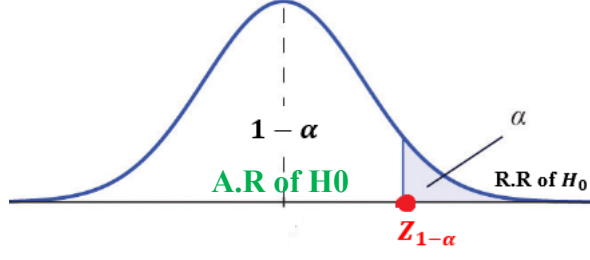
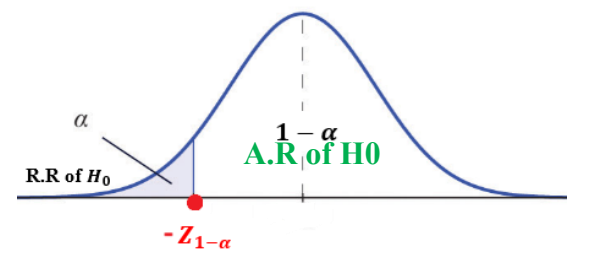
**Note**, under the assumption of the equality of the two population proportions ( $H_0: P_1 = P_2 = P$ ), the pooled estimate of the common proportion  $p$  (**pooled proportion**) is:

$$\bar{P} = \frac{X_1 + X_2}{n_1 + n_2} \quad (\bar{q} = 1 - \bar{P})$$

**The test statistic (T.S.) is**

$$Z = \frac{\hat{P}_1 - \hat{P}_2}{\sqrt{\frac{\bar{P}\bar{q}}{n_1} + \frac{\bar{P}\bar{q}}{n_2}}} \sim N(0,1)$$

# Test procedures:

Hypotheses	$H_0 : P_1 - P_2 = 0$ $H_A : P_1 - P_2 \neq 0$	$H_0 : P_1 - P_2 \leq 0$ $H_A : P_1 - P_2 > 0$	$H_0 : P_1 - P_2 \geq 0$ $H_A : P_1 - P_2 < 0$
Test Statistic (T.S.)	$Z = \frac{\hat{P}_1 - \hat{P}_2}{\sqrt{\frac{\bar{P}\bar{q}}{n_1} + \frac{\bar{P}\bar{q}}{n_2}}} \sim N(0,1)$ Pooled Proportion $\bar{P} = \frac{X_1 + X_2}{n_1 + n_2}$ and $\bar{q} = 1 - \bar{P}$		
R.R. & A.R. of $H_0$			
Critical value(s)	$Z_{1-\frac{\alpha}{2}}$ and $-Z_{1-\frac{\alpha}{2}}$	$Z_{1-\alpha}$	$-Z_{1-\alpha}$
Decision:	We reject $H_0$ (and accept $H_A$ ) at the significance level $\alpha$ if:		
	$Z_{T.S} > Z_{1-\frac{\alpha}{2}}$ or $Z_{T.S} < -Z_{1-\frac{\alpha}{2}}$	$Z_{T.S} > Z_{1-\alpha}$	$Z_{T.S} < -Z_{1-\alpha}$
	<b>T.S. <math>\in</math> R.R. Two - Sided Test</b>	<b>T.S. <math>\in</math> R.R. One - Sided Test</b>	<b>T.S. <math>\in</math> R.R. One - Sided Test</b>

## Example:

In a study about the obesity (overweight), a researcher was interested in comparing the proportion of obesity between males and females. The researcher has obtained a random sample of 150 males and another independent random sample of 200 females. The following results were obtained from this study.

	n	Number of obese people (X)
<b>Males</b>	150	21
<b>Females</b>	200	48

Can we conclude from these data that there is a difference between the proportion of obese males and proportion of obese females? Use  $\alpha = 0.05$  , assume that the two population proportions are equal .

### Solution :

$P_1$  = population proportion of obese males.

$P_2$  = population proportion of obese females .

$\hat{P}_1$  = sample proportion of obese males

$\hat{P}_2$  = sample proportion of obese females

$\alpha=0.05$  (level of significance )

	Sample size (n)	X	Sample proportion $\hat{P}$
Male (1)	150 large	21	$\hat{P}_1 = \frac{X_1}{n_1} = \frac{21}{150} = 0.14$
Female (2)	200 large	48	$\hat{P}_2 = \frac{X_2}{n_2} = \frac{48}{200} = 0.24$

The pooled estimate of the common proportion  $P$  is:

$$\bar{P} = \frac{X_1 + X_2}{n_1 + n_2} = \frac{21 + 48}{150 + 200} = 0.197 \quad (\bar{q} = 1 - \bar{P} = 1 - 0.197 = 0.803)$$

**Hypotheses:**

$$H_0: P_1 = P_2 \quad \text{against} \quad H_A: P_1 \neq P_2$$

OR

$$H_0: P_1 - P_2 = 0 \quad \text{against} \quad H_A: P_1 - P_2 \neq 0$$

Hypotheses:

$$H_0 : P_1 = P_2 \quad \text{OR} \quad (H_0: P_1 - P_2 = 0)$$
$$H_A : P_1 \neq P_2 \quad (H_A: P_1 - P_2 \neq 0)$$

Test statistics (T.S.)

$$Z = \frac{\hat{P}_1 - \hat{P}_2}{\sqrt{\frac{\bar{P}\bar{q}}{n_1} + \frac{\bar{P}\bar{q}}{n_2}}} = \frac{0.14 - 0.24}{\sqrt{\frac{0.196 \times 0.803}{150} + \frac{0.196 \times 0.803}{200}}} = -2.328$$

Rejection Region of  $H_0$  (R.R.): (critical region)  $(-\infty, -1.96) \cup (1.96, \infty)$

$$\alpha = 0.05 \gg \text{Critical Value : } \pm Z_{1-\frac{\alpha}{2}} = \pm Z_{1-\frac{0.05}{2}} = \pm Z_{(0.975)} = \pm 1.96$$

Decision :

Reject  $H_0: P_1 = P_2$  if :

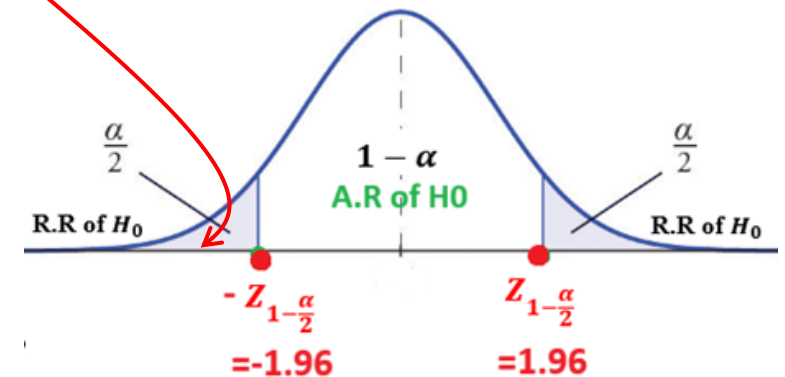
$$Z_{T.S} < -Z_{1-\frac{\alpha}{2}} \quad \text{or} \quad Z_{T.S} > Z_{1-\frac{\alpha}{2}}$$

$-2.328 < -1.96$  (First condition satisfied, then do not try the second condition)

**Decision is** Reject  $H_0: P_1 = P_2$  and Accept  $H_A: P_1 \neq P_2$

Therefore, we conclude that there is a difference between the proportion of obese males and the proportion of obese females

Another way to take decision is by graph :  
Determine test statistics value on the graph



Since  $Z = -2.328 \in R.R.$ , we reject  $H_0: P_1 = P_0$  and accept  $H_A: P_1 \neq P_2$  at  $\alpha = 0.05$ . Therefore,

we conclude that there **is a difference between the proportion of obese males and the proportion of obese females**. Additionally, since,

$$\hat{P}_1 = 0.14 < \hat{P}_2 = 0.24$$

**Sample proportion of male < Sample proportion of female**

We may conclude that the **proportion of obesity for females is larger than that for males**.

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