

```
1 #sample average
2 def Meu(num_of_1,num_of_0):
3     meu=num_of_1/(num_of_1+num_of_0)
4     return meu

1 #standard deviation1
2 def SD1(num_of_1,num_of_0):
3     sum1=num_of_1*((1-Meu(num_of_1,num_of_0))**2)
4     sum0=num_of_0*((0-Meu(num_of_1,num_of_0))**2)
5     sum_of_all=sum1+sum0
6     sd=sum_of_all/(num_of_1+num_of_0)
7     return sd
```

### ▼ standard deviation1

$$\sigma = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}.$$

```
1 #standard deviation2
2 def SD2(num_of_1,num_of_0):
3     sum1=num_of_1*((1-Meu(num_of_1,num_of_0))**2)
4     sum0=num_of_0*((0-Meu(num_of_1,num_of_0))**2)
5     sum_of_all=sum1+sum0
6     sd=sum_of_all/(num_of_1+num_of_0-1)
7     return sd
```

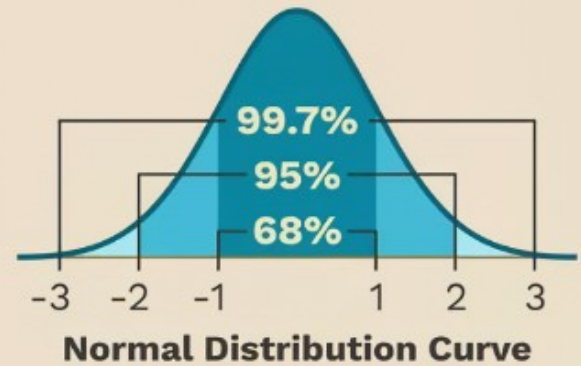
### ▼ standard deviation2

## Calculating Standard Deviation

$$S_x = \sqrt{\frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n-1}}$$



$n$  = The number of data points  
 $X_i$  = Each of the values of the data  
 $\bar{X}$  = The mean of  $X_i$



```
1 #=====Example 1=====
2 #total sms=40
3 #group A=20 :
4 print("mean of group A:",Meu(10,10))
5 print("standard deviation of group A:",SD1(10,10))
6 #group B=20 :
7 print("mean of group B:",Meu(18,2))
8 print("standard deviation of group B:",SD1(18,2))
```

mean of group A: 0.5  
 standard deviation of group A: 0.25  
 mean of group B: 0.9  
 standard deviation of group B: 0.09

## Statistics Kingdom

[Home](#) > [Mean](#) > Two-Sample T (Welch's)

### Two Sample T-Test Calculator (Welch's T-test)

**Unknown unequal** standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:	Two ( $H_1: \mu_1 \neq \mu_2 + d$ )	Significance level ( $\alpha$ ):	0.05
Effect:	Medium	Effect type:	Standardized effect size

Effect Size:

0.5

Outliers:

Included

Difference (d):

0

Enter raw data

or

Paste excel data

or enter summarized data ( $\bar{x}$ ,  $n$ ,  $\sigma$ ,  $S$ ) below

Group name:

Group-1

Group-2

Sample average ( $\bar{x}$ ):

0.5

0.9

Sample size (n):

20

20

Sample SD (S):

0.25

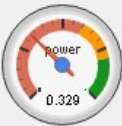
0.09

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

Calculate test

Clear

Load last run



### Two sample t-test (Welch), using T distribution (df=23.8434) (two-tailed) (validation)

#### 1. $H_0$ hypothesis

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

#### 2. P-value

The p-value equals **5.994e-7**, ( $p(x \leq T) = 2.997e-7$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 5.994e-7 (0.00006%).

The smaller the p-value the more it supports  $H_1$ .

#### 3. The statistics

The test statistic T equals **-6.7324**, which is not in the 95% region of acceptance: [-2.0646 : 2.0646].

$x_1 - x_2 = -0.4$ , is not in the 95% region of acceptance: [-0.1227 : 0.1227].

The standard deviation of the difference,  $S'$  equals 0.0594, is used to calculate the statistic.

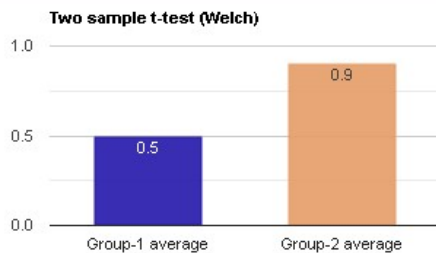
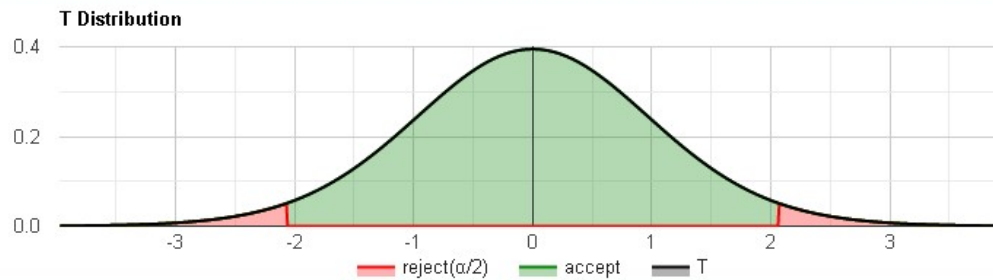
#### 4. Effect size

The observed effect size  $d$  is **large**, **2.13**. This indicates that the magnitude of the difference between the average and average is large.

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(statskingdom@gmail.com)

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). it is likely you chose the right test.

- **Test power**

Although the priori power is low (0.3294), the  $H_0$  is rejected.

- **Equality of variances assumption**

Based on a two-tailed F test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is 0.0000437)

F test assumes **equal** standard deviations, which is **not** your test assumption.

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

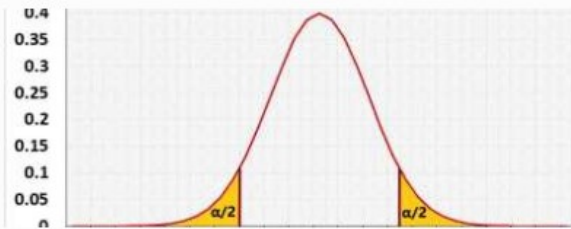
$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2})^2}{\frac{s_1^4}{n_1^2(n_1-1)} + \frac{s_2^4}{n_2^2(n_2-1)}}$$

## Normal distribution



## Assumptions



Normal distribution

**$\sigma$**

The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$

**$d$**

Expected difference  $d$  between the populations's average is known

## Required Sample Data

**$\bar{x}$**

$\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2

**$n$**

$n_1, n_2$  - Sample size of group1 and group2

**$S$**

$S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

We would expect  $\mu_1$  to be equal to  $\mu_2$ . If one of the machines is slower than the other one, it should be serviced. In this case, we would like to know both if  $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ , since either machine could be slower.

2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

1 #=====Example 2=====

2 #total sms=40



```
3 #group A=30 :
4 print("mean of group A:",Meu(10,20))
5 print("standard deviation of group A:",SD1(10,20))
6 #group B=10 :
7 print("mean of group B:",Meu(8,2))
8 print("standard deviation of group B:",SD1(8,2))
```

```
mean of group A: 0.3333333333333333
standard deviation of group A: 0.2222222222222227
mean of group B: 0.8
standard deviation of group B: 0.16
```

# Statistics Kingdom

[Home](#) > [Mean](#) > [Two-Sample T \(Welch's\)](#)

## Two Sample T-Test Calculator (Welch's T-test)

**Unknown unequal** standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:	Two ( $H_1: \mu_1 \neq \mu_2 + d$ )	Significance level ( $\alpha$ ):	0.05
Effect:	Medium	Effect type:	Standardized effect size
Effect Size:	0.5	Outliers:	Included
Difference (d):	0		

[Enter raw data](#) or [Paste excel data](#) or enter summarized data ( $\bar{x}$ , n,  $\sigma$ , S) below

Group name:	Group-1	Group-2
Sample average ( $\bar{x}$ ):	0.3333333333333333	0.8
Sample size (n):	30	10
Sample SD (S):	0.2222222222222227	0.16

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

Calculate test

Clear

Load last run



### Two sample t-test (Welch), using T distribution (df=21.5323) (two-tailed) (validation)

#### 1. $H_0$ hypothesis

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

#### 2. P-value

The p-value equals **3.717e-7**, ( $p(x \leq T) = 1.858e-7$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 3.717e-7 (0.000037%).

The smaller the p-value the more it supports  $H_1$ .

#### 3. The statistics

The test statistic T equals **-7.1956**, which is not in the 95% region of acceptance: [-2.0765 : 2.0765].

$\bar{x}_1 - \bar{x}_2 = -0.47$ , is not in the 95% region of acceptance: [-0.1347 : 0.1347].

The standard deviation of the difference,  $S'$  equals 0.0649, is used to calculate the statistic.

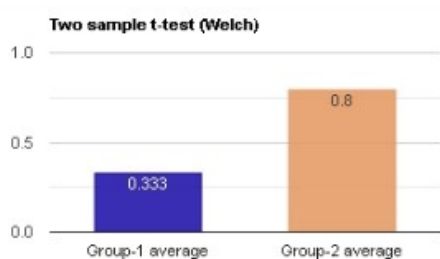
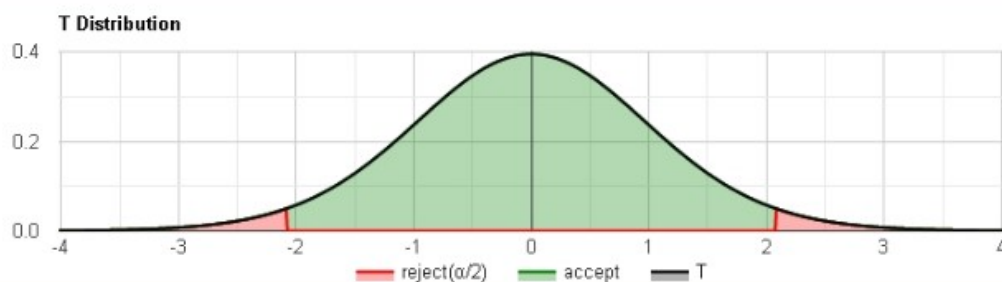
#### 4. Effect size

The observed effect size  $d$  is **large, 2.23**. This indicates that the magnitude of the difference between the average and average is large.

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). it is likely you chose the right test.

- **Test power**

Although the priori power is low (0.2972), the  $H_0$  is rejected.

- **Equality of variances assumption**

Based on a two-tailed F test,  $\sigma_1$  is considered as **equal** to  $\sigma_2$  (p-value is 0.302)

F test assumes **equal** standard deviations, which is **not** your test assumption.

### Recommendations

The F-test can't reject the **unequal** variance assumption, since it is based on the **equal** variance assumption, which is not your preliminary assumption. You should continue with the Welch's t-test, which is also robust to variances equality.

Calculate T equal  $\sigma$

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{s_1^4}{n_1^2(n_1-1)} + \frac{s_2^4}{n_2^2(n_2-1)}}$$

### Normal distribution





## Assumptions



Normal distribution

**$\sigma$**

The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$

**$d$**

Expected difference  $d$  between the populations's average is known

## Required Sample Data

**$\bar{x}$**

$\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2

**$n$**

$n_1, n_2$  - Sample size of group1 and group2

**$S$**

$S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

We would expect  $\mu_1$  to be equal to  $\mu_2$ . If one of the machines is slower than the other one, it should be serviced. In this case, we would like to know both if  $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ , since either machine could be slower.

2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

```
1 #=====Example 3=====
```

```
2 #total sms=1000
```

```
3 #group A=750 :
```

```
4 print("mean of group A:",Meu(98,617))
```

```
5 print("standard deviation of group A:",SD1(98,617))
```

```
6 #group B=250 :
```

```
7 print("mean of group B:",Meu(47,203))
```

```
8 print("standard deviation of group B:",SD1(47,203))
```

```
mean of group A: 0.13706293706293707
```

```
standard deviation of group A: 0.11827668834661842
```

```
mean of group B: 0.188
```

standard deviation of group B: 0.15265600000000001

# Statistics Kingdom

[Home](#) > [Mean](#) > Two-Sample T (Welch's)

## Two Sample T-Test Calculator (Welch's T-test)

**Unknown unequal** standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:	Two ( $H_1: \mu_1 \neq \mu_2 + d$ )	Significance level ( $\alpha$ ):	0.05
Effect:	Medium	Effect type:	Standardized effect size
Effect Size:	0.5	Outliers:	Included
Difference (d):	0		

[Enter raw data](#) or [Paste excel data](#) or enter summarized data ( $\bar{x}$ ,  $n$ ,  $\sigma$ ,  $S$ ) below

Group name:	Group-1	Group-2
Sample average ( $\bar{x}$ ):	0.13706293706293707	0.188
Sample size (n):	750	250
Sample SD (S):	0.11827668834661842	0.15265600000000001

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

[Calculate test](#)

[Clear](#)

[Load last run](#)



**Two sample t-test (Welch), using T distribution (df=353.9094) (two-tailed) (validation)**

**1.  $H_0$  hypothesis**

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

## 2. P-value

The p-value equals **0.000002176**, ( $p(x \leq T) = 0.000001088$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 0.000002176 (0.00022%).

The smaller the p-value the more it supports  $H_1$ .

## 3. The statistics

The test statistic T equals **-4.8159**, which is not in the 95% region of acceptance:  $[-1.9667; 1.9667]$ .

$x_1 - x_2 = -0.051$ , is not in the 95% region of acceptance:  $[-0.0208; 0.0208]$ .

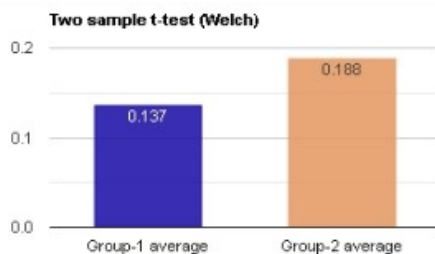
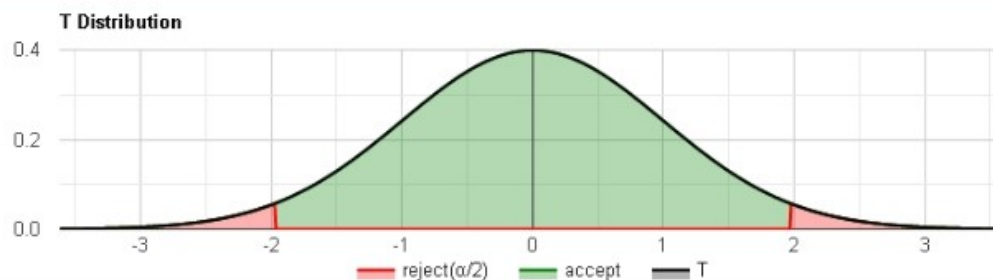
The standard deviation of the difference,  $S'$  equals 0.0106, is used to calculate the statistic.

## 4. Effect size

The observed effect size  $d$  is **medium**, **0.4**. This indicates that the magnitude of the difference between the average and average is medium.

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). It is likely you chose the right test.

### ● Test power

The test priori power is strong: 1

### ● Equality of variances assumption

Based on a two-tailed F test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is  $2.7e-7$ ).

Based on a two-tailed t test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is  $2.76 \times 10^{-7}$ )  
 F test assumes **equal** standard deviations, which is **not** your test assumption.

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

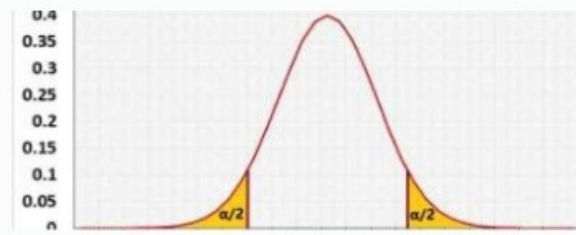
$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{s_1^4}{n_1^2(n_1-1)} + \frac{s_2^4}{n_2^2(n_2-1)}}$$

### Normal distribution



## Assumptions



Normal distribution



The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$



Expected difference  $d$  between the populations's average is known

## Required Sample Data



$\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2



$n_1, n_2$  - Sample size of group1 and group2



$S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same



number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

We would expect  $\mu_1$  to be equal to  $\mu_2$ . If one of the machines is slower than the other one, it should be serviced. In this case, we would like to know both if  $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ , since either machine could be slower.

2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

```
1 #=====Example 4=====
```

```
2 #total sms=1000
```

```
3 #group A=750 :
```

```
4 print("mean of group A:",Meu(39,711))
```

```
5 print("standard deviation of group A:",SD1(39,711))
```

```
6 #group B=250 :
```

```
7 print("mean of group B:",Meu(9,241))
```

```
8 print("standard deviation of group B:",SD1(9,241))
```

```
mean of group A: 0.052
```

```
standard deviation of group A: 0.049296
```

```
mean of group B: 0.036
```

```
standard deviation of group B: 0.034704
```

## Statistics Kingdom

[Home](#) > [Mean](#) > Two-Sample T (Welch's)

### Two Sample T-Test Calculator (Welch's T-test)

**Unknown unequal** standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:

Two ( $H_1: \mu_1 \neq \mu_2 + d$ )



Significance

0.05



level ( $\alpha$ ):

Effect: Medium Effect type: Standardized effect size

Effect Size: 0.5 Outliers: Included

Difference (d): 0

Enter raw data or Paste excel data or enter summarized data ( $\bar{x}$ , n,  $\sigma$ , S) below

Group name:	Group-1	Group-2
Sample average ( $\bar{x}$ ):	0.052	0.036
Sample size (n):	750	250
Sample SD (S):	0.049296	0.034704

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

Calculate test

Clear

Load last run



### Two sample t-test (Welch), using T distribution (df=605.5212) (two-tailed) (validation)

#### 1. $H_0$ hypothesis

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

#### 2. P-value

The p-value equals **2.661e-8**, ( $p(x \leq T) = 1$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 2.661e-8 (0.0000027%).

The smaller the p-value the more it supports  $H_1$ .

#### 3. The statistics

The test statistic T equals **5.6366**, which is not in the 95% region of acceptance: [-1.9639 : 1.9639].

$\bar{x}_1 - \bar{x}_2 = 0.016$ , is not in the 95% region of acceptance: [-0.005575 : 0.005575].

The standard deviation of the difference, S' equals 0.00284, is used to calculate the statistic.

#### 4. Effect size

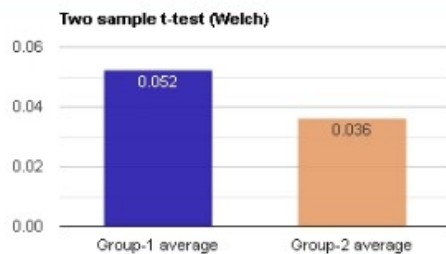
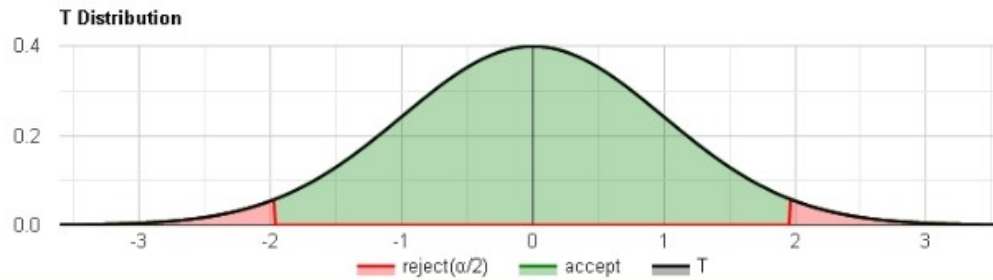
The observed effect size d is **small, 0.35**. This indicates that the magnitude of the difference between the average and average is small.

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(statskingdom@gmail.com)

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). it is likely you chose the right test.

- **Test power**

The test priori power is strong: 1

- **Equality of variances assumption**

Based on a two-tailed F test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is  $2.18e-10$ )

F test assumes **equal** standard deviations, which is **not** your test assumption.

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

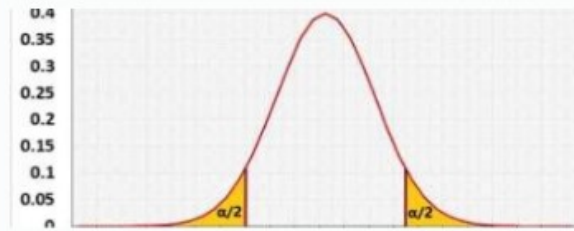
$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{\left(\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}\right)^2}{\frac{S_1^4}{n_1^2(n_1-1)} + \frac{S_2^4}{n_2^2(n_2-1)}}$$

## Normal distribution



## Assumptions



Normal distribution



The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$



Expected difference **d** between the populations's average is known

## Required Sample Data



$\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2



$n_1, n_2$  - Sample size of group1 and group2



$S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

#### 1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

We would expect  $\mu_1$  to be equal to  $\mu_2$ . If one of the machines is slower than the other one, it should be serviced. In this case, we would like to know both if  $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ , since either machine could be slower.

#### 2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

```

1 #=====Example 5=====
2 #total sms=10000
3 #group A=9000 :
4 print("mean of group A:",Meu(2927,6073))
5 print("standard deviation of group A:",SD1(2927,6073))
6 #group B=1000 :
7 print("mean of group B:",Meu(742,258))
8 print("standard deviation of group B:",SD1(742,258))

```

```

mean of group A: 0.3252222222222225
standard deviation of group A: 0.21945272839506172
mean of group B: 0.742
standard deviation of group B: 0.19143599999999997

```

# Statistics Kingdom

[Home](#) > [Mean](#) > Two-Sample T (Welch's)

## Two Sample T-Test Calculator (Welch's T-test)

**Unknown unequal** standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:	Two ( $H_1: \mu_1 \neq \mu_2 + d$ )	Significance level ( $\alpha$ ):	0.05
Effect:	Medium	Effect type:	Standardized effect size
Effect Size:	0.5	Outliers:	Included
Difference (d):	0		

Enter raw data or Paste excel data or enter summarized data ( $\bar{x}$ ,  $n$ ,  $\sigma$ ,  $S$ ) below

Group name:	Group-1	Group-2
Sample average ( $\bar{x}$ ):	0.3252222222222225	0.742
Sample size (n):	9000	1000



Sample SD (S):

0.21945272839506172

0.19143599999999997

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

Calculate test

Clear

Load last run



### Two sample t-test (Welch), using T distribution (df=1308.9351) (two-tailed) (validation)

#### 1. $H_0$ hypothesis

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

#### 2. P-value

The p-value equals **0**, ( $p(x \leq T) = 0$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 0 (0%).

The smaller the p-value the more it supports  $H_1$ .

#### 3. The statistics

The test statistic T equals **-64.3111**, which is not in the 95% region of acceptance: [-1.9618 : 1.9618].

$\bar{x}_1 - \bar{x}_2 = -0.42$ , is not in the 95% region of acceptance: [-0.01271 : 0.01271].

The standard deviation of the difference, S' equals 0.00648, is used to calculate the statistic.

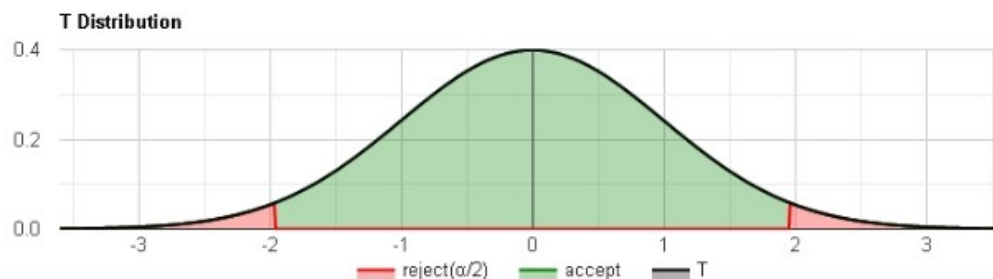
#### 4. Effect size

The observed effect size d is **large, 1.92**. This indicates that the magnitude of the difference between the average and average is large.

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(statskingdom@gmail.com)

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). it is likely you chose the right test.

- **Test power**

The test priori power is strong: 1

- **Equality of variances assumption**

Based on a two-tailed F test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is NaN)

F test assumes **equal** standard deviations, which is **not** your test assumption.

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

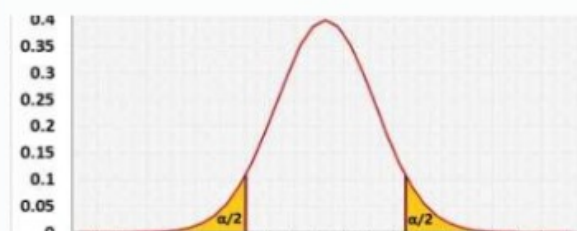
$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{\left(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}\right)^2}{\frac{s_1^4}{n_1^2(n_1-1)} + \frac{s_2^4}{n_2^2(n_2-1)}}$$

### Normal distribution



## Assumptions



Normal distribution

**6** The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$

**d** Expected difference **d** between the populations's average is known

## Required Sample Data

**$\bar{x}$**   $\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2

**n**  $n_1, n_2$  - Sample size of group1 and group2

**S**  $S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

We would expect  $\mu_1$  to be equal to  $\mu_2$ . If one of the machines is slower than the other one, it should be serviced. In this case, we would like to know both if  $\mu_1 < \mu_2$  or  $\mu_1 > \mu_2$ , since either machine could be slower.

2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

```
1 #=====Example 6=====
2 #total sms=10000
3 #group A=9000 :
4 print("mean of group A:",Meu(7967,1033))
5 print("standard deviation of group A:",SD1(7967,1033))
6 #group B=1000 :
7 print("mean of group B:",Meu(617,383))
8 print("standard deviation of group B:",SD1(617,383))
```

```
mean of group A: 0.8852222222222222
standard deviation of group A: 0.10160383950617284
mean of group B: 0.617
standard deviation of group B: 0.236311
```

[Home](#) > [Mean](#) > Two-Sample T (Welch's)

# Two Sample T-Test Calculator (Welch's T-test)

## Unknown unequal standard deviation

Expected difference between two populations' mean

[Video](#) [Information](#) [T-Equal standard deviation](#) [Paired-T](#) [One sample T](#)

Tails:	Two ( $H_1: \mu_1 \neq \mu_2 + d$ )	Significance level ( $\alpha$ ):	0.05
Effect:	Medium	Effect type:	Standardized effect size
Effect Size:	0.5	Outliers:	Included
Difference (d):	0		

[Enter raw data](#) or [Paste excel data](#) or enter summarized data ( $\bar{x}$ ,  $n$ ,  $\sigma$ ,  $S$ ) below

Group name:	Group-1	Group-2
Sample average ( $\bar{x}$ ):	0.8852222222222222	0.617
Sample size (n):	9000	1000
Sample SD (S):	0.10160383950617284	0.236311

When entering raw data, the t test calculator will run the Shapiro-Wilk normality test and calculate outliers, as part of the test calculation, and will generate the R code for your data.

[Calculate test](#)[Clear](#)[Load last run](#)

**Two sample t-test (Welch), using T distribution (df=1040.4126) (two-tailed) (validation)**

### 1. $H_0$ hypothesis

Since  $p\text{-value} < \alpha$ ,  $H_0$  is rejected.

The average of **Group-1's** population is considered to be **not equal to** the average of **Group-2's** population.

In other words, the difference between the sample average of **Group-1** and **Group-2** is big enough to be statistically significant.

## 2. P-value

The p-value equals **0**, ( $p(x \leq T) = 1$ ). It means that the chance of type I error (rejecting a correct  $H_0$ ) is small: 0 (0%).  
The smaller the p-value the more it supports  $H_1$ .

## 3. The statistics

The test statistic T equals **35.53**, which is not in the 95% region of acceptance: [-1.9622 : 1.9622].

$\bar{x}_1 - \bar{x}_2 = 0.27$ , is not in the 95% region of acceptance: [-0.01481 : 0.01481].

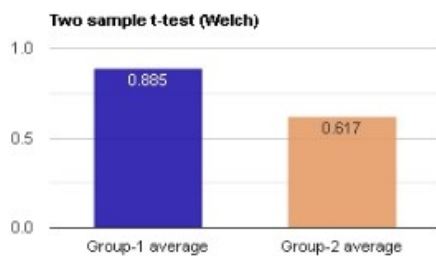
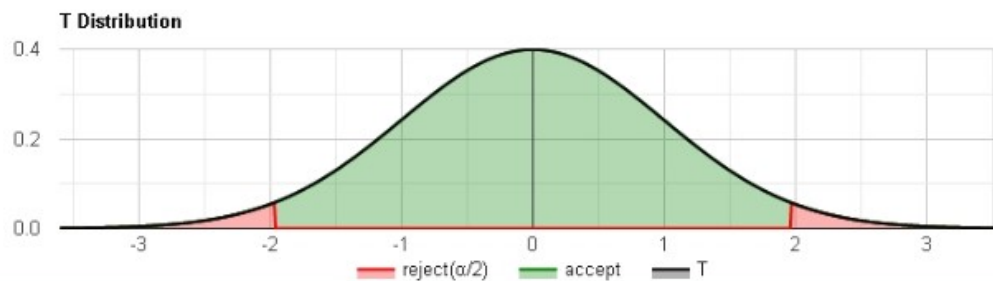
The standard deviation of the difference, S' equals 0.00755, is used to calculate the statistic.

## 4. Effect size

The observed effect size d is **large, 2.2**. This indicates that the magnitude of the difference between the average and average is large.

**If you like the page, please share or like. Questions, comments and suggestions are appreciated.**  
(statskingdom@gmail.com)

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## Test validation

The requested test was calculated, assumes **unequal** standard deviation ( $\sigma$ ). it is likely you chose the right test.

- **Test power**

The test priori power is strong: 1

- **Equality of variances assumption**

Based on a two-tailed F test,  $\sigma_1$  is considered as **unequal** to  $\sigma_2$  (p-value is 2.67e-13)

F test assumes **equal** standard deviations, which is **not** your test assumption.

## Information

**Target:** To check if the difference between the average (mean) of two groups (populations) is significant, using sample data

Example1: A man of average is expected to be 10cm taller than a woman of average ( $d=10$ )

Example2: The average weight of an apple grown in field 1 is expected to be equal in weight to the average apple grown in field 2 ( $d=0$ )

### Hypotheses

$$H_0: \mu_1 = \mu_2 + d$$

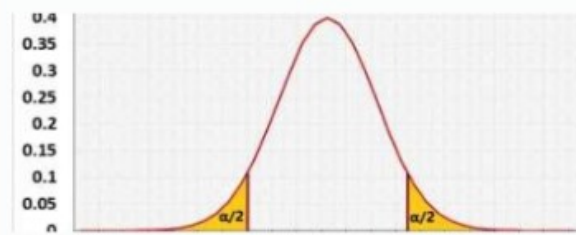
$$H_1: \mu_1 \neq \mu_2 + d$$

### Test statistic

$$t = \frac{\bar{x}_1 - \bar{x}_2 - d}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

$$df = \frac{(\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2})^2}{\frac{s_1^4}{n_1^2(n_1-1)} + \frac{s_2^4}{n_2^2(n_2-1)}}$$

## Normal distribution



## Assumptions



Normal distribution

6

The standard deviations of both populations not necessarily equal, so either  $\sigma_1 = \sigma_2$  or  $\sigma_1 \neq \sigma_2$

d

Expected difference **d** between the populations's average is known

## Required Sample Data

$\bar{x}$

$\bar{x}_1, \bar{x}_2$  - Sample average of group1 and group2

n

$n_1, n_2$  - Sample size of group1 and group2

S

$S_1, S_2$  - Sample standard deviation of group1 and group2

### Examples

1. Two tailed test example:

A factory uses two identical machines to produce plastic plates. You would expect both machines to produce the same number of plates per minute.

Let  $\mu_1$  = average number of plates produced by machine1 per minute.

Let  $\mu_2$  = average number of plates produced by machine2 per minute.

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2. Left tail example.

A farmer uses fertilizer-1 with good results.

A friend told him fertilizer-2 is better than fertilizer #1.

Let  $\mu_1$  = average number of potatoes per square meter in gardens using fertilizer-1.

Let  $\mu_2$  = average number of potatoes per square meter in gardens using fertilizer-2.

The farmer assumes that the fertilizer currently in use (fertilizer-1) is better than the suggested one.(or equal)

He is willing to change fertilizer only if the new one is better.

A one-tailed test is controversial. It increase the type I error. In this example, it may be important to know if fertilizer-2 is less effective than fertilizer-1, and use the two-tailed test.

1

```
1 #=====Example 1=====
2 #total sms=40
3 #group A=20 :
4 print("mean of group A:",Meu(10,10))
5 print("standard deviation of group A:",SD1(10,10))
6 #group B=20 :
7 print("mean of group B:",Meu(18,2))
8 print("standard deviation of group B:",SD1(18,2))
```

```
mean of group A: 0.5
standard deviation of group A: 0.25
mean of group B: 0.9
standard deviation of group B: 0.09
```

```
1 #=====Example 2=====
2 #total sms=40
3 #group A=30 :
4 print("mean of group A:",Meu(10,20))
5 print("standard deviation of group A:",SD1(10,20))
6 #group B=10 :
7 print("mean of group B:",Meu(8,2))
8 print("standard deviation of group B:",SD1(8,2))
```

```
mean of group A: 0.3333333333333333
standard deviation of group A: 0.22222222222222227
mean of group B: 0.8
standard deviation of group B: 0.16
```

```
1 #=====Example 3=====
```

```
2 #total sms=1000
3 #group A=750 :
4 print("mean of group A:",Meu(98,617))
5 print("standard deviation of group A:",SD1(98,617))
6 #group B=250 :
7 print("mean of group B:",Meu(47,203))
8 print("standard deviation of group B:",SD1(47,203))
```

```
mean of group A: 0.13706293706293707
standard deviation of group A: 0.11827668834661842
mean of group B: 0.188
standard deviation of group B: 0.15265600000000001
```

```
1 #=====Example 4=====
```

```
2 #total sms=1000
3 #group A=750 :
4 print("mean of group A:",Meu(39,711))
5 print("standard deviation of group A:",SD1(39,711))
6 #group B=250 :
7 print("mean of group B:",Meu(9,241))
8 print("standard deviation of group B:",SD1(9,241))
```

```
mean of group A: 0.052
standard deviation of group A: 0.049296
mean of group B: 0.036
standard deviation of group B: 0.034704
```

```
1 #=====Example 5=====
```

```
2 #total sms=10000
3 #group A=9000 :
4 print("mean of group A:",Meu(2927,6073))
5 print("standard deviation of group A:",SD1(2927,6073))
6 #group B=1000 :
7 print("mean of group B:",Meu(742,258))
8 print("standard deviation of group B:",SD1(742,258))
```

```
mean of group A: 0.32522222222222225
standard deviation of group A: 0.21945272839506172
mean of group B: 0.742
standard deviation of group B: 0.19143599999999997
```

```
1 #=====Example 6=====
```

```
2 #total sms=10000
```

```
3 #group A=9000 :
4 print("mean of group A:",Meu(7967,1033))
5 print("standard deviation of group A:",SD1(7967,1033))
6 #group B=1000 :
7 print("mean of group B:",Meu(617,383))
8 print("standard deviation of group B:",SD1(617,383))
```

```
mean of group A: 0.8852222222222222
standard deviation of group A: 0.10160383950617284
mean of group B: 0.617
standard deviation of group B: 0.236311
```