## **Small sample tests: t-distribution**

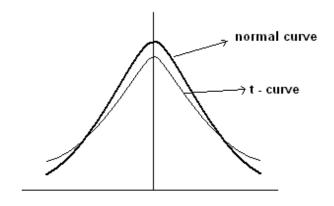
If the sample sizes are small (<30), the above tests (for large samples) do not hold good. Therefore, for estimation of the parameters as well as for testing a hypothesis, we cannot use the above methods.

#### Student's t-distribution

If we take a large number of samples of small (<30) sizes, calculate the mean of each sample, obtain the frequencies and obtain the frequency curve, we will find that the resulting sampling distribution of the mean is the **student's t-distribution**.

#### Properties of t distribution

- 1. The curve extends from  $-\infty$  to  $+\infty$
- 2. Like the normal distribution the t distribution is also symmetrical and has mean zero.
- 3. The variance of t distribution is greater than unity and approaches unity when the degrees of freedom (d.f.) i.e the size of the sample becomes large. The degrees of freedom (d.f.) is given by n-1



t distribution curve

#### Uses of t distribution

Suppose it is not possible to take a sample of large size (due to lack of time or prohibitive costs), we can use the t distribution to do the 't-test' for testing hypothesis about the population mean/s and for estimating the population mean using the sample mean.

## 1. Test of Significance (TOS) of the difference between sample mean and population mean:

Sample size  $n \pmod{n < 30}$ 

Population mean:  $\mu$ 

Sample mean :  $\bar{x}$ 

Population S. D. :  $\sigma$  (always unknown)

Sample S. D.: s

Null Hypothesis :  $H_0$  :  $\mu = \mu_0$ 

Alternate Hypothesis :  $H_1: \mu \neq \mu_0$  (Two tailed)

 $H_1: \mu < \mu_0$  (One tailed (left))

 $H_1: \mu > \mu_0$  (One tailed (right))

#### **Test Statistic**

$$t = \frac{\bar{x} - \mu_0}{s / \sqrt{n-1}} \sim t_{(n-1)} d.f$$

This t statistic is said to follow a t distribution with v = n - 1 degrees of freedom. The t distribution has been derived under the hypothesis that the parent population is distributed normally.

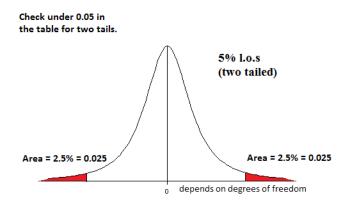
### **Conclusion of the test**

Let LOS =  $\alpha$ 

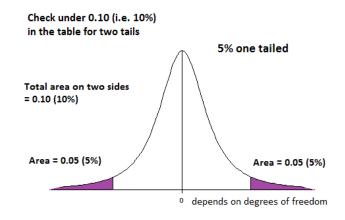
If  $|t_{cal}| > t_{\alpha}$ , Reject H<sub>0</sub> at  $\alpha$  level of significance. Otherwise we have no reason to reject H<sub>0</sub>.

#### How to read the table:

There is a different table given for t-distribution, since the curve is different, the areas will also differ from a normal distribution. The rows in the t-distribution table indicate degrees of freedom, and the columns indicate the most commonly used levels of significance. So if you wanted to find the critical value at 5% l.os.(two tailed) and with 9 degrees of freedom (i.e. the size of the sample is 10), then the value from the table is 2.26. This means that when the area in the critical region is 5% (2.5% on each side, since it is two-tailed, the critical value  $Z_{\infty} = 2.26$ .



Note that the table given to you indicates two-tail areas. In the event of a two-tailed test (depending on the alternative hypothesis), reading the critical values from the table is easy. However, if the test is one tailed, with 5% l.o.s. then you cannot look up the value under 5% in the table because it is the two tailed area, meaning the corresponding one-tailed area is 2.5%. However, if you look up the critical value for the two tailed area 10%, then the area is distributed evenly on both the tails (5% on each side), and the critical value corresponding to this area will be the same as the 5% area on a one-tailed test.



## **Solved Examples:**

1. A machinist is expected to make engine parts with axle diameter of 1.75 cm. A random sample of 10 parts shows a mean diameter of 1.85 cm, with a S.D of 0.1 cm. On the basis of this sample, would you say that the work of the machinist is inferior?

**Solution**: Given n = 10 (small sample)

$$\bar{x} = 1.85$$

Level of significance:  $\alpha = 5\%$ 

$$H_0: \mu = 1.75$$

 $H_1$ :  $\mu \neq 1.75$  (two-tailed test)

Under H<sub>0</sub>, the test statistic is given by

$$t = \frac{\bar{x} - \mu_0}{s / \sqrt{n-1}} \sim t_{(n-1)} = t_0 d.f$$

*i.e* 
$$t = \frac{1.85 - 1.75}{0.1/\sqrt{10 - 1}} = 3$$

From the tables,  $t_{9.0.05} = 2.262$ 

Since |t| > 2.262 we reject H<sub>0</sub> at 5% level of significance. That is, **the work is inferior** at 5% LOS.

**Remark:** But from the tables,  $t_{9,0.01} = 3.250$ . Since |t| < 2.262 we have no reason to reject H<sub>0</sub> at 1% level of significance. That is, we cannot consider the work to be inferior at 1% LOS.

2. A certain injection administered to each of the 12 patients resulted in the following increases of blood pressure: 5, 2, 8, -1, 3, 6, 0, -2, -1, 5, 0, 4 Can it be concluded that the injection will be, in general,

accompanied by an increase in BP?

**Solution**: We have n=12 (small sample).

The mean of the sample is the average increase in B.P.

$$\bar{x} = \frac{1}{n} \sum x = \frac{1}{12} (31) = 2.58$$

The S.D of the sample is given by

$$s^{2} = \frac{1}{n} \sum x^{2} - \left(\frac{1}{n} \sum x\right)^{2} = \frac{1}{12} (185) - (2.58)^{2} = 8.76$$

$$\Rightarrow s = 2.96$$

Level of significance:  $\alpha = 5\%$ 

 $H_0: \mu = 0$  (there is no increase in B.P)

 $H_1: \mu > 0$  (one-tailed test)

Under H<sub>0</sub>, the test statistic is given by

$$t = \frac{\bar{x} - \mu_0}{s / \sqrt{n-1}} \sim t_{(n-1)} = t_{11} d.f$$

*i.e* 
$$t = \frac{2.58 - 0}{2.96 / \sqrt{11}} = 2.89$$

From the tables,  $t_{11,0.05}$  (one – tailed) = 1.796

Since |t| > 1.796 we reject H<sub>0</sub> at 5% level of significance. And conclude that, the injection will, in general, result in increase in B.P

**Exercise**: The annual rainfall at a certain place is normally distributed with mean 30mm. If the rainfalls during the past 8 years (in mm) are 31.1, 30.7, 24.3, 28.1, 27.9, 32.2, 25.4 and 29.1, can we conclude that the average rainfall during the last 8 years is less than the normal rainfall?

## TOS of difference between two sample means (Small samples) and Paired t-tests

## TOS of difference between two sample means (Small samples)

Sample Sizes :  $n_1 < 30$ ,  $n_2 < 30$ 

Population means :  $\mu_{\!\scriptscriptstyle 1}$  ,  $\mu_{\!\scriptscriptstyle 2}$ 

Population s.d.'s :  $\sigma_{\scriptscriptstyle 1}$  ,  $\sigma_{\scriptscriptstyle 2}$  (unknown)

Sample means :  $\bar{x}_1$ ,  $\bar{x}_2$ 

Sample s.d.'s:  $s_1$ ,  $s_2$ 

Null Hypothesis :  $H_0$  :  $\mu_1 = \mu_2$ 

Alternate Hypothesis :  $H_1: \mu_1 \neq \mu_2$  (Two tailed)

 $H_1: \mu_1 < \mu_2$  (One tailed (left))

 $H_{\scriptscriptstyle 1}$  :  $\mu_{\scriptscriptstyle \! 1}$  >  $\mu_{\scriptscriptstyle \! 2}$  (One tailed (right))

# Test Statistic: (for when the two samples are independent)

$$t = \frac{\overline{x_1} - \overline{x_2}}{\sqrt{\frac{n_1 s_1^2 + n_2 s_2^2}{n_1 + n_2^{-2}} \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim t_{(n_1 + n_2 - 2)} d.f$$

Let LOS be  $\alpha$ 

If  $|t| > t_{\alpha}$ ,  $(n_1 + n_2 - 2)$ , Reject H<sub>0</sub> at  $\alpha$  level of significance. Otherwise we have no reason to reject H<sub>0</sub>.

## **Examples**

1. The mean height and S.D of 8 randomly chosen soldiers are 166.9 and 8.29 respectively. The corresponding values of 6 randomly chosen sailors are 170.3 and 8.50 cm respectively. Based on these data, can we conclude that soldiers are, in general, shorter than sailors?

**Solution**: Given  $n_1 = 8$ ;  $n_2 = 6$  (small samples)

$$\bar{x}_1 = 166.9; \ \bar{x}_2 = 170.3; \ s_1 = 8.29; \ s_2 = 8.50$$

Level of significance:  $\alpha = 5\%$ 

$$H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 < \mu_2 \ (one-tailed\ test)$$

Test Statistic: 
$$t = \frac{\overline{x_1} - \overline{x_2}}{\sqrt{\frac{n_1 s_1^2 + n_2 s_2^2}{n_1 + n_2 - 2} \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim t_{(n_1 + n_2 - 2)} d.f$$

We have

$$t = \frac{166.9 - 170.3}{\sqrt{\frac{983.29}{12} \left(\frac{1}{8} + \frac{1}{6}\right)}} = -0.695$$

From the tables,  $t_{12.0.05}$  (one – tailed) = 1.782

Since |t| < 1.782 we have no reason to reject H<sub>0</sub> at 5% level of significance. That is, we accept H<sub>0</sub> and conclude that soldiers are, in general, of the same average height as sailors.

2. Samples of two types of electric bulbs were tested for length of life and the following data were obtained:

	size	Mean	SD
Sample 1	8	1234 h	36 h
Sample 2	7	1036 h	40 h

Is the difference in the means sufficient to warrant that Type 1 bulbs are superior to type 2 bulbs?

**Solution**: Given  $n_1 = 8$ ;  $n_2 = 7$  (small samples)

$$\bar{x}_1 = 1234; \ \bar{x}_2 = 1036; \ s_1 = 36; \ s_2 = 40$$

Level of significance:  $\alpha = 5\%$ 

$$H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 > \mu_2 \ (one-tailed\ test)$$

Test Statistic: 
$$t = \frac{\overline{x_1} - \overline{x_2}}{\sqrt{\frac{n_1 s_1^2 + n_2 s_2^2}{n_1 + n_2 - 2} \left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim t_{(n_1 + n_2 - 2)} d.f$$

We have 
$$t = \frac{1234 - 1036}{21.0807} = 9.39$$

From the tables, 
$$t_{13.0.05}$$
 (one – tailed) = 1.771

Since |t| > 1.782 we reject H<sub>0</sub> at 5% level of significance and conclude that **type 1 bulbs** are superior to type 2 bulbs.

## **Small sample tests** (for dependent samples)

#### Paired t tests

The paired t test is generally used when measurements are taken from the same subject before and after some manipulation such as injection of a drug, extra coaching, adding water treatment etc. For example, you can use a paired t test to determine the significance of a difference in blood pressure before and after administration of an experimental pressure substance or the effectiveness of a water additive in reducing bacterial numbers by sampling water from different sources and comparing bacterial counts in the treated versus untreated water sample. Each different water source would give a different pair of data points.

Let  $x_1$  and  $x_2$  be the variables denoting the first and second series respectively

Let 
$$d = x_1 - x_2$$
  
and  $\overline{d} = \frac{1}{n} \sum (x_1 - x_2)$  (the mean (average) of the differences)

Let 
$$s^2 = \operatorname{var}(d) = \frac{1}{n} \sum d^2 - \left(\overline{d}\right)^2$$

Then the test statistic is given by

$$t = \frac{\overline{d}}{\frac{s}{\sqrt{n-1}}} \sim t_{(n-1)} d.f$$

#### **Example:**

1. The following data relate to marks obtained by 11 students in 2 tests, one held at the beginning of the year and the other at the end of the year after intensive coaching:

Test 1	19	23	16	24	17	18	20	18	21	19	20
Test 2	17	24	20	24	20	22	20	20	18	22	19

Do the data indicate that the students have benefited from coaching?

**Solution**: We have,

Test 1	19	23	16	24	17	18	20	18	21	19	20
Test 2	17	24	20	24	20	22	20	20	18	22	19
$d = x_1 - x_2$	2	-1	-4	0	-3	-4	0	-2	3	-3	1
$d^2$	4	1	16	0	9	16	0	4	9	9	1

$$\overline{d} = \frac{1}{n} \sum (x_1 - x_2)$$

$$\Rightarrow \overline{d} = \frac{1}{11} (-11) = -1$$

$$s^2 = \text{var}(d) = \frac{1}{n} \sum d^2 - (\overline{d})^2$$

$$= \frac{1}{11} (69) - (-1)^2 = 5.27$$

$$\Rightarrow$$
  $s = 2.296$ 

Let the Level of significance:  $\alpha = 5\%$ 

$$H_0: \overline{d} = 0$$
 (no benefit from coaching)  
 $H_1: \overline{d} < 0$  (one-tailed test)

We have the test statistic to be given by

$$t = \frac{\overline{d}}{\frac{s}{\sqrt{n-1}}} t_{(n-1)} d.f$$

$$i.e \ t = \frac{-1}{\frac{2.296}{\sqrt{10}}} = -1.38$$

From the tables,  $t_{10.0.05}$  (one – tailed) = 1.812

Since |t| < 1.812 we have no reason to reject H<sub>0</sub> at 5% level of significance and conclude that the students have not benefited from coaching.