

**Theorem 8.4.** Let  $X_1, X_2, \dots, X_n$  be i.i.d.  $N(\mu, \sigma)$  random variables. Also, let  $S^2$  be the standard variance for this random sample. Then, the random variable  $T$  defined as

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$$

has a  $t$ -distribution with  $n - 1$  degrees of freedom, i.e.,  $T \sim T(n - 1)$ .

**Proof:**

Define the random variable  $Z$  as

$$Z = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}.$$

Then,  $Z \sim N(0, 1)$ . Also, define the random variable  $Y$  as

$$Y = \frac{(n - 1)S^2}{\sigma^2}.$$

Then by Theorem [Theorem 8.3](#),  $Y \sim \chi^2(n - 1)$ . We conclude that the random variable

$$T = \frac{Z}{\sqrt{\frac{Y}{n-1}}} = \frac{\bar{X} - \mu}{S/\sqrt{n}}$$

has a  $t$ -distribution with  $n - 1$  degrees of freedom.

## Confidence Intervals for the Mean of Normal Random Variables

Here, we assume that  $X_1, X_2, X_3, \dots, X_n$  is a random sample from a normal distribution  $N(\mu, \sigma^2)$ , and our goal is to find an interval estimator for  $\mu$ . We no longer require  $n$  to be large. Thus,  $n$  could be any positive integer. There are two possible scenarios depending on whether  $\sigma^2$  is known or not. If the value of  $\sigma^2$  is known, we can easily find a confidence interval for  $\mu$ . This can be done using exactly the same method that we used to estimate  $\mu$  for a general distribution for the case of large  $n$ . More specifically, we know that the random variable

$$Q = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$$

has  $N(0, 1)$  distribution. In particular,  $Q$  is a function of the  $X_i$ 's and  $\mu$ , and its distribution does not depend on  $\mu$ . Thus,  $Q$  is a pivotal quantity, and we conclude that  $\left[\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}, \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}\right]$  is  $(1 - \alpha)100\%$  confidence interval for  $\mu$ .

Assumptions: A random sample  $X_1, X_2, X_3, \dots, X_n$  is given from a  $N(\mu, \sigma^2)$  distribution, where  $\text{Var}(X_i) = \sigma^2$  is known.

Parameter to be Estimated:  $\mu = EX_i$ .

Confidence Interval:  $\left[\bar{X} - z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}, \bar{X} + z_{\frac{\alpha}{2}} \frac{\sigma}{\sqrt{n}}\right]$  is a  $(1 - \alpha)100\%$  confidence interval for  $\mu$ .

---

The more interesting case is when we do not know the variance  $\sigma^2$ . More specifically, we are given  $X_1, X_2, X_3, \dots, X_n$ , which is a random sample from a normal distribution  $N(\mu, \sigma^2)$ , and our goal is to find an interval estimator for  $\mu$ . However,  $\sigma^2$  is also unknown. In this case, using [Theorem 8.4](#), we conclude that the random variable  $T$  defined as

$$T = \frac{\bar{X} - \mu}{S/\sqrt{n}}$$

has a  $t$ -distribution with  $n - 1$  degrees of freedom, i.e.,  $T \sim T(n - 1)$ . Here, the random variable  $T$  is a pivotal quantity, since it is a function of the  $X_i$ 's and  $\mu$ , and its distribution does not depend on  $\mu$  or any other unknown parameters. Now that we have a pivot, the next step is to find a  $(1 - \alpha)$  interval for  $T$ . Using the definition of  $t_{p,n}$ , a  $(1 - \alpha)$  interval for  $T$  can be stated as

$$P\left(-t_{\frac{\alpha}{2}, n-1} \leq T \leq t_{\frac{\alpha}{2}, n-1}\right) = 1 - \alpha.$$

Therefore,

$$P\left(-t_{\frac{\alpha}{2}, n-1} \leq \frac{\bar{X} - \mu}{S/\sqrt{n}} \leq t_{\frac{\alpha}{2}, n-1}\right) = 1 - \alpha,$$

which is equivalent to

$$P\left(\bar{X} - t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}} \leq \mu \leq \bar{X} + t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}}\right) = 1 - \alpha.$$

We conclude that  $\left[\bar{X} - t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}}, \bar{X} + t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}}\right]$  is  $(1 - \alpha)100\%$  confidence interval for  $\mu$ .

Assumptions: A random sample  $X_1, X_2, X_3, \dots, X_n$  is given from a  $N(\mu, \sigma^2)$  distribution, where  $\mu = EX_i$  and  $\text{Var}(X_i) = \sigma^2$  are unknown.

Parameter to be Estimated:  $\mu = EX_i$ .

Confidence Interval:  $\left[\bar{X} - t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}}, \bar{X} + t_{\frac{\alpha}{2}, n-1} \frac{S}{\sqrt{n}}\right]$  is a  $(1 - \alpha)$  confidence interval for  $\mu$ .

### Example 8.20

A farmer weighs 10 randomly chosen watermelons from his farm and he obtains the following values (in lbs):

7.72   9.58   12.38   7.77   11.27   8.80   11.10   7.80   10.17   6.00

Assuming that the weight is normally distributed with mean  $\mu$  and variance  $\sigma$ , find a 95% confidence interval for  $\mu$ .

**Solution**

Using the data we obtain

$$\begin{aligned}\bar{X} &= 9.26, \\ S^2 &= 3.96\end{aligned}$$

Here,  $n = 10$ ,  $\alpha = 0.05$ , so we need

$$t_{0.025, 9} \approx 2.262$$

The above value can be obtained in MATLAB using the command `tinv(0.975, 9)`.

Thus, we can obtain a 95% confidence interval for  $\mu$  as