

Statistics: the t-test is a test of statistical significance

During the first week each compared two sets of data: for example, pulse rate between males and females, or mass-to-length ratios of two different kinds of bean.

How can we quantify our confidence that these sets of data truly reflect a difference?

We call this quantified confidence “statistical significance,” and we use the following parameters to calculate it. The (so-called “two-tailed Student’s”) **t-test** allows us to distinguish between the “null hypothesis” and an “alternative hypothesis.”

H₀: There is **no difference** between the observed data sets (or **no effect** experimentally).

H_A: There **is a** “statistically significant” **difference** between the two data sets.

x, y	Measurements, data points, samples. N is the sample size , the number of subjects or specimens you examined.
$\bar{x} = \frac{\sum x}{N}$	Mean (or average). Σ (Sigma) simply means “sum.”
$x - \bar{x}$	Deviation is simply the difference between a measurement and the mean.
$S_x^2 = \frac{\sum (x - \bar{x})^2}{N - 1}$	Variance S_x^2 is a combined measure of (the square of) all the deviations . $N - 1$ is the degrees of freedom , in this case the number of ways you can compare any measurement with all the others.
$SD_x = \sqrt{S_x^2}$	The standard deviation is the square root of the variance. Think: why would you square the sum of the deviations and then take the square root?
$SE = \sqrt{\frac{S_x^2}{N_x} + \frac{S_y^2}{N_y}}$	The standard error is a measure of the combined variances of two collections of data .
$t_s = \frac{ \bar{x} - \bar{y} }{SE}$	The t value for your two sets of samples is a measure of the difference in means compared to the standard error .
$t_c, \alpha = 0.05, df = 8 = 2.306$	You select the “critical” t value from a table based on the total degrees of freedom of both data sets, and the confidence level that you choose.

If $t_s > t_c$, then it is appropriate, with the selected confidence level, to **reject the null hypothesis**.