

Testing Differences between Two Means

Large Independent Sample Means:

Used to test whether the data from two samples come from the same populations or whether two populations are different.

Assumptions:

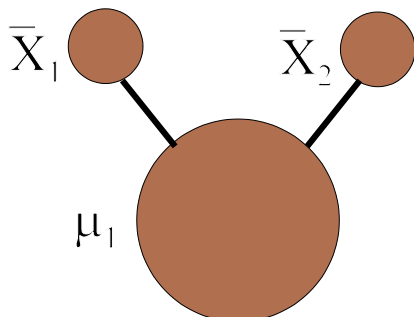
- samples must be independent, i.e., there can be no relationship between the two samples
- populations must be normally distributed and standard deviations known or sample size > 30
- should not be used if more than two means are tested unless adjustments are made to significance levels (e.g., Bonferroni correction, $\alpha_{\text{Bonferroni}} = \alpha/\text{number of tests}$)

Z-test:

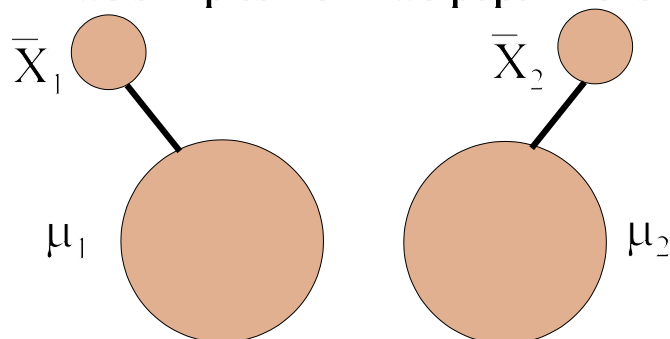
Test value:
$$z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Critical value comes from standard normal (z) distribution. Use one- or two-tailed test. Conservatively, choose the two-tailed test. Values are also available at bottom of t -distribution.

Two samples from one population



Two samples from two populations



The Step-by-Step Approach

Step 1: State hypotheses

Two-tailed:

$$H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 \neq \mu_2$$

One-tailed:

$$H_0: \mu_1 \leq \mu_2 \quad \text{or} \quad H_0: \mu_1 \geq \mu_2$$

$$H_1: \mu_1 > \mu_2 \quad \text{or} \quad H_1: \mu_1 < \mu_2$$

Step 2: Find critical value

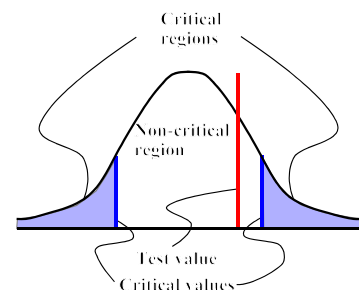
Look up z -score for specified significance (α) level and for one- or two-tailed test (selected in advance). Usually use $\alpha = 0.05$ and two-tailed test, i.e., $z_{\text{critical}} = \pm 1.960$. For one-tailed use $z_{\text{critical}} = \pm 1.645$.

Step 3: Compute test value

$$z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$

Step 4: Make decision

Draw diagram of normal distribution and critical regions. If test value is in critical region reject the null hypothesis otherwise do not reject.



Step 5: Summarize results

Restate hypothesis (null or alternate) accepted in step 4.

If reject null:

There is enough evidence to reject the null hypothesis.

If not reject null:

There is not enough evidence to reject the null hypothesis.

Optionally, reword hypothesis in “lay” terms. E.g., There is/is not a difference between the two populations **or** one population is greater/lesser than the other for the independent variable.

Testing Differences between Two Means

Small Independent Sample Means:

When population standard deviations are unknown **and** sample size is < 30 use the t -distribution for critical values and a t -test for test values. First use an F -ratio to determine whether sample variances are equal or unequal. Then choose the correct t -test.

Assumptions

- two samples must be independent, i.e., different subjects—if not, use “dependent-groups t -test”
- data must be normally distributed

If sample variances are NOT equal:

Use test value:

$$t = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$$

For degrees of freedom (df) use smaller of $n_1 - 1$ and $n_2 - 1$ (i.e., conservative choice, higher critical value)

If sample variances are equal:

Use test value:

$$t = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{(n_1 - 1)s_1^2 + (n_2 - 1)s_2^2}{n_1 + n_2 - 2}} \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

and $df = n_1 + n_2 - 2$

Uses a “pooled” estimate of variance that combined with a larger degree of freedom ($n_1 + n_2 - 2$) increases the test’s power (i.e., ability to find a true difference).

Test for Equal Variances

Also called Homogeneity of Variance

- used primarily to determine which t -test to use
- uses F -distribution and F -test (later used for ANOVA)
- assume variances are equal and test if unequal
- SPSS uses “Levine’s Test for Equality of Variances”
If P (Sig.) $< \alpha$ variances are NOT equal.

Step 1: Always a two-tailed test.

$$\mathbf{H_0: } s_1^2 = s_2^2$$

$$\mathbf{H_1: } s_1^2 \neq s_2^2$$

Step 2:

Find critical value (F_{CV}) from F -distribution. Use degrees of freedom of larger variance ($df_N = n_{\text{larger}} - 1$) as numerator and degrees of freedom of smaller variance as denominator ($df_D = n_{\text{smaller}} - 1$).

Step 3:

Compute test value:
$$F_{TV} = \frac{s_{\text{larger}}^2}{s_{\text{smaller}}^2}$$

Note, F_{TV} will always be ≥ 1 .

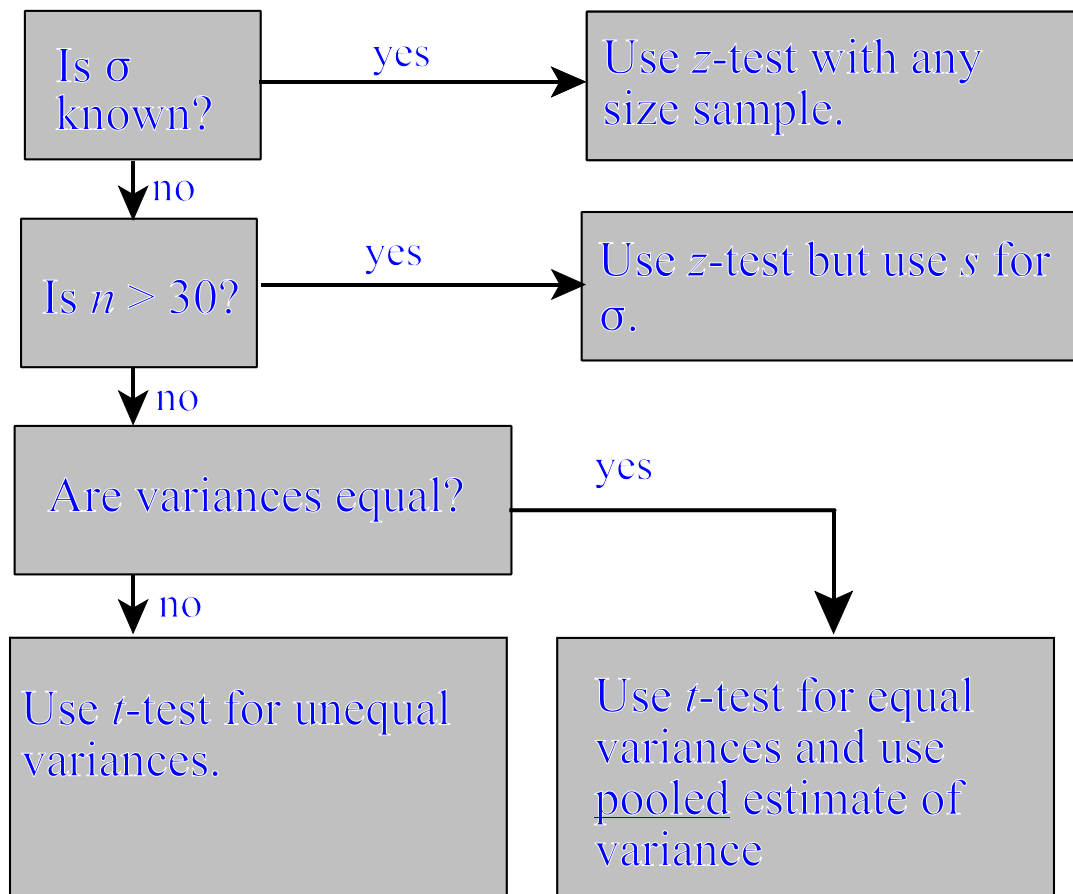
Step 4 and 5:

If $F_{TV} > F_{CV}$ then reject H_0 and conclude variances are unequal.

If $F_{TV} \leq F_{CV}$ then do NOT reject H_0 and conclude variances are equal. I.e., you have **homogeneity of variances**. You can now select the appropriate “Independent-groups t -test”.

Flow Diagram for Choosing the Correct Independent Samples t -Test

Similar to flow diagram used for single sample means. But requires a test for equality of variances (**homogeneity of variance**). Generally the sample's mean and standard deviation are used with the t -distribution. The t -distribution becomes indistinguishable from the z -distribution (normal distribution) when $n > 30$. Samples must be independent.



Testing Differences between Two Means

Dependent Sample Means:

Used when two samples are not independent. More powerful than independent groups t -test and easier to perform (no variance test required). Simplifies research protocol (i.e., fewer subjects) but dependence may limit generalizability.

Examples:

- repeated measures (test/retest, before/after)
- matched pairs t -test (subjects matched by a relevant variable: height, weight, shoe size, IQ score, age)
- twin studies (identical, heterozygotic, living apart)

Step 1: Two-tailed:

$$H_0: \mu_D = 0$$

$$H_1: \mu_D \neq 0$$

One-tailed:

$$H_0: \mu_D \leq 0 \quad \text{or} \quad H_0: \mu_D \geq 0$$

$$H_1: \mu_D > 0 \quad \text{or} \quad H_1: \mu_D < 0$$

Step 2:

Critical value from **t -distribution** with degrees of freedom equal to number of data pairs minus one ($df = n - 1$).

Step 3:

Compute differences between pairs (D) then mean difference (\bar{D}) and s_D :

$$\bar{D} = \frac{\sum D}{n} \quad \text{and} \quad s_D = \sqrt{\frac{\sum D^2 - \frac{(\sum D)^2}{n}}{n-1}}$$

Test value:

$$t = \frac{\bar{D} - \mu_D}{\frac{s_D}{\sqrt{n}}}$$

Step 4 and 5:

If test value $>$ critical value reject H_0 otherwise there is no difference between the two trials/groups.