

Learning to Calculate Cohen's d Effect Size in SPSS

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The Critical Role of Effect Size in Modern Statistical Analysis

In the realm of [inferential statistics](#), researchers habitually utilize hypothesis tests, such as the t-test, to ascertain whether a genuine difference exists between the means of two distinct population groups. This initial assessment relies heavily on the concept of [statistical significance](#), often determined by calculating a [p-value](#).

While the [p-value](#) effectively establishes whether an observed difference is unlikely to be the result of random chance, it fails to quantify the practical utility or magnitude of that difference. To move beyond mere declaration of significance and assess real-world relevance, researchers must quantify the actual size of the observed phenomenon. This necessity leads us directly to the measurement of [effect size](#).

Among the standardized measures used to compare two means, [Cohen's d](#) stands out as the most widely accepted metric. This measure is indispensable because it standardizes the difference between group means, enabling researchers to evaluate the practical impact and generalizability of their findings across diverse studies and contexts, ensuring that the results hold genuine scientific importance.

Defining and Calculating Cohen's d

Fundamentally, [Cohen's d](#) is defined as the difference between two group means divided by the pooled [standard deviation](#) of the data. This crucial standardization step transforms the raw mean difference into a common unit of measure, allowing the resulting value to be interpreted straightforwardly as the number of [standard deviations](#) separating the two group means.

The formula commonly used for calculating [Cohen's d](#), particularly when assuming equal sample sizes and homogeneity of variances (using the pooled standard deviation), is formalized below:

$$\text{Cohen's } d = (x_1 - x_2) / \sqrt{(s_1^2 + s_2^2) / 2}$$

The specific variables used in this calculation represent the following statistical parameters essential for determining the standardized difference:

x_1 , x_2 : The arithmetic [mean](#) of the measurements for sample 1 and sample 2, respectively.

s_1^2 , s_2^2 : The [variance](#) of the data points within sample 1 and sample 2, respectively.

Interpreting the Practical Magnitude of Cohen's d

The numerical output of [Cohen's d](#) offers a direct quantification of the separation or overlap between the distributions of the two populations being compared. A higher absolute value of d signifies a more substantial, non-trivial difference between the groups. The interpretation is

inherently linked to the [standard deviation](#) of the data:

A d of **0.5** indicates that the two group means are separated by exactly half of one [standard deviation](#).

A d of **1** means the group means differ by a full [standard deviation](#).

A d of **2** suggests a highly pronounced separation, where the group means are two [standard deviations](#) apart.

To illustrate the practical implication, an [effect size](#) of 0.5 implies that the average score of an individual in the higher-scoring group surpasses the scores of 69% of the individuals in the lower-scoring group. Essentially, the mean of the first group sits 0.5 [standard deviations](#) above the mean of the second group.

To provide standardized guidance on classifying the strength of the observed effect, researchers commonly rely on the following conventional thresholds, which serve as a helpful rule of thumb for interpreting [Cohen's d](#):

A value of **0.2** is conventionally categorized as a **small effect size**.

A value of **0.5** is conventionally categorized as a **medium effect size**.

A value of **0.8** is conventionally categorized as a **large effect size**.

Preparing for Calculation: The SPSS Workflow



While manual calculation of [Cohen's d](#) is possible, it quickly becomes inefficient when dealing with complex or large datasets. Fortunately, [SPSS](#) (Statistical Package for the Social Sciences) offers an integrated and streamlined method for computing this metric directly alongside the results of an independent-samples t-test.

The most efficient method within [SPSS](#) utilizes the built-in functionality for comparing means. This process is initiated by navigating through the menu: **Analyze**, followed by **Compare Means and Proportions**, and finally selecting **Independent-Samples T Test**. It is essential to confirm that the version of [SPSS](#) being used supports the estimation of effect sizes, a capability that has been standard in all recent releases of the software.

To illustrate this process, we will use a hypothetical botanical study. A researcher is investigating the relative effectiveness of two different fertilizers. The goal is to determine if Fertilizer #1 and Fertilizer #2 produce significantly different average plant growth measurements, recorded in inches. The study involved 24 plants, split into two randomly assigned groups (12 plants per fertilizer type).

The resulting growth data are organized in the [SPSS](#) dataset using two primary variables: **Growth** (a continuous outcome variable) and **Fertilizer** (a categorical grouping variable). The botanist

specifically requires [effect size](#) quantification to understand the magnitude of the difference in growth attributable to the fertilizer type. The structure of the raw data is visualized below:

	 Fertilizer	 Growth	var	var
1	One	8		
2	One	9		
3	One	11		
4	One	11		
5	One	12		
6	One	14		
7	One	15		
8	One	16		
9	One	16		
10	One	18		
11	One	20		
12	One	21		
13	Two	7		
14	Two	9		
15	Two	10		
16	Two	10		
17	Two	11		
18	Two	11		
19	Two	12		
20	Two	14		
21	Two	14		
22	Two	16		
23	Two	20		
24	Two	23		
25				

Executing the Independent-Samples T-Test and Effect Size Estimation

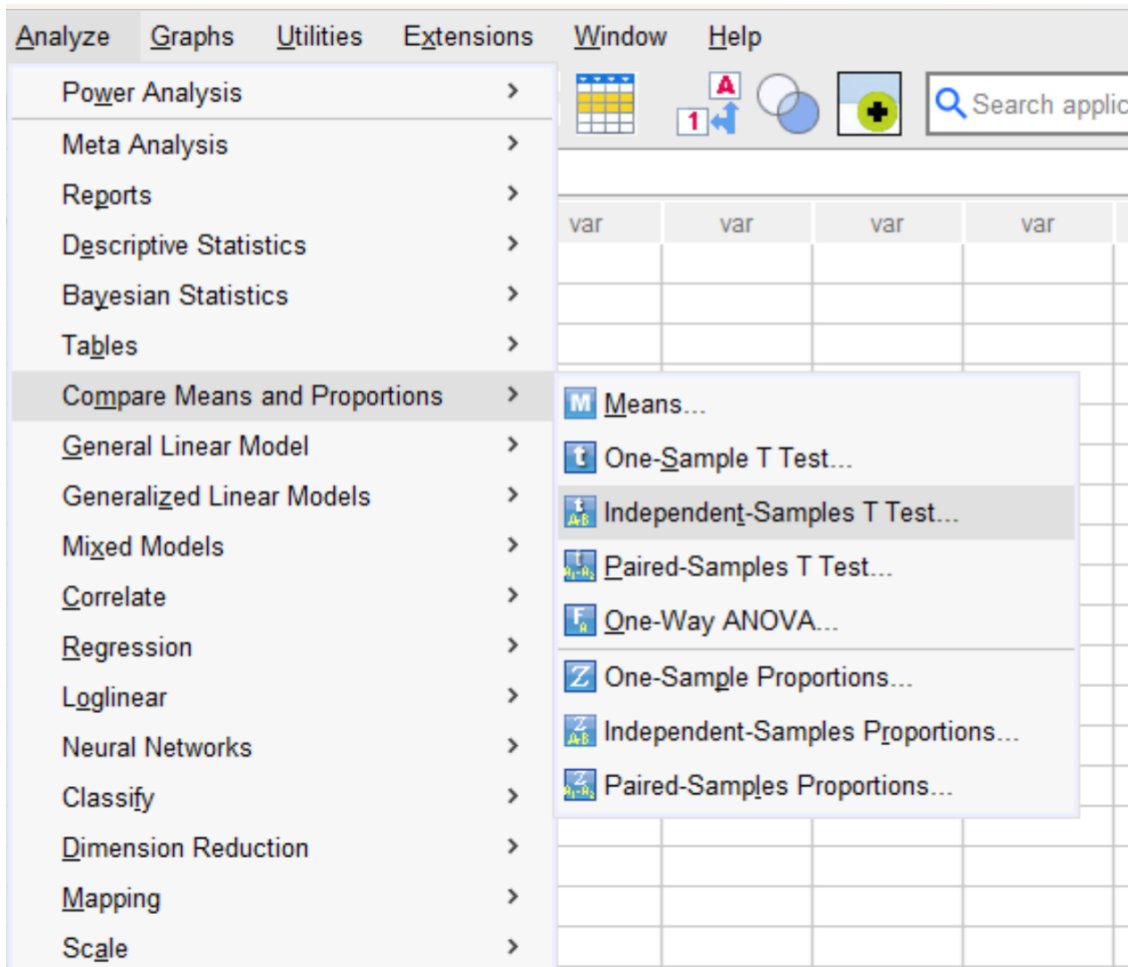
To begin the analysis and ensure the calculation of the effect size measure, follow these procedural steps within the [SPSS](#) interface, ensuring correct variable assignment and activation of the necessary options:

Click the **Analyze** tab located in the main menu bar.

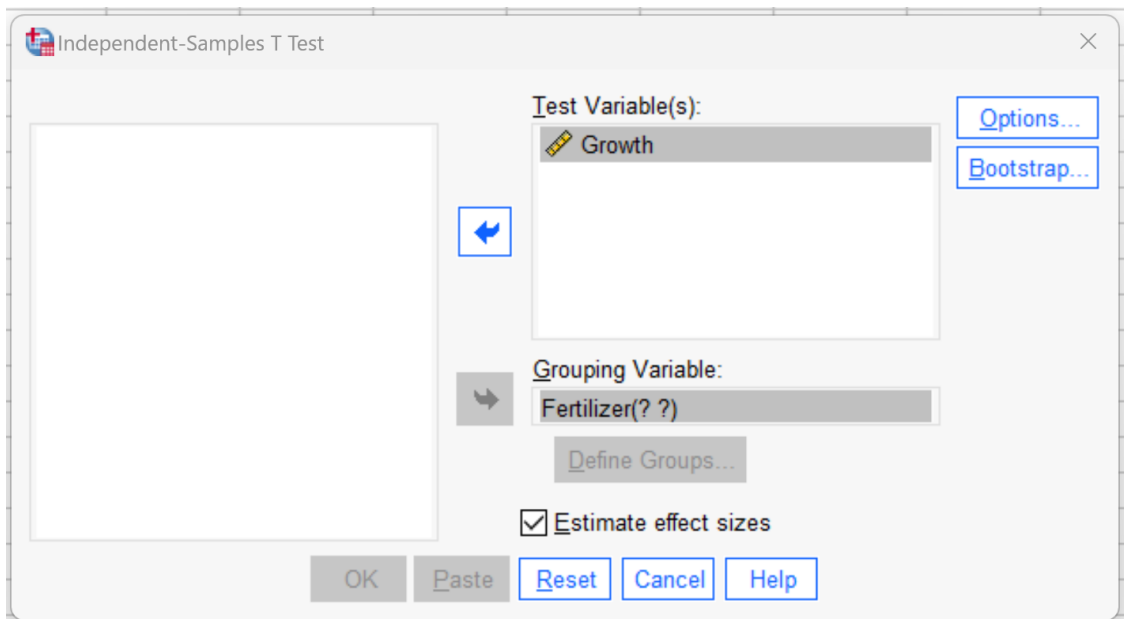
Select **Compare Means and Proportions** from the drop-down menu.

Select **Independent-Samples T Test** to open the primary dialogue box.

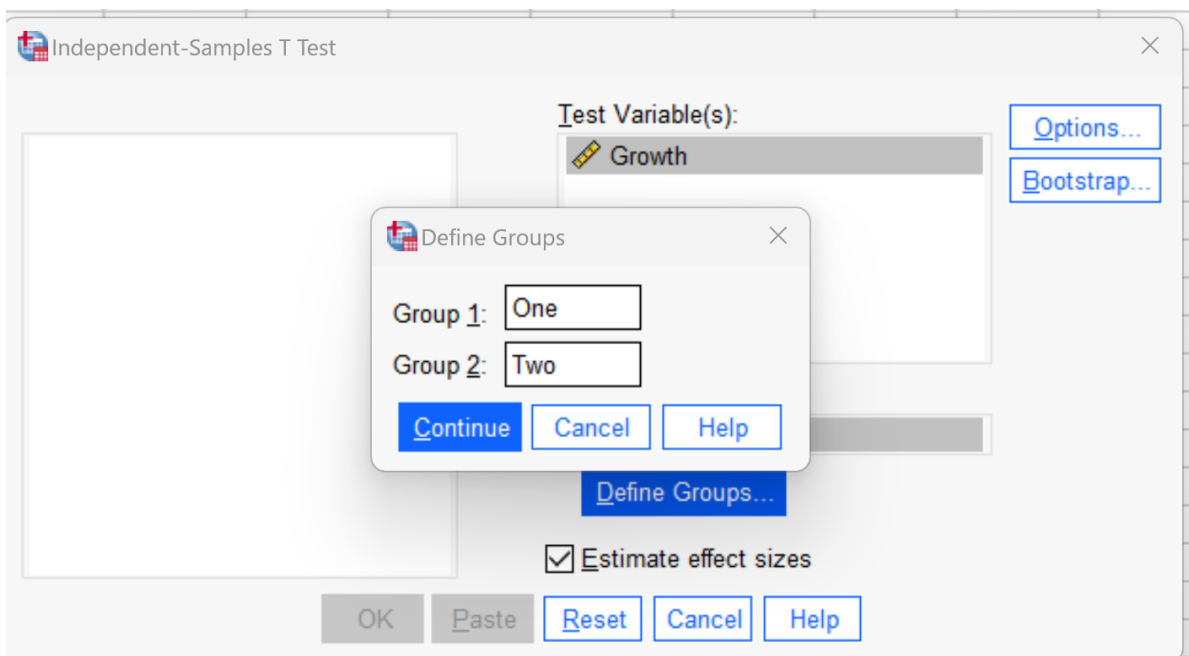
This action opens the main dialogue box for the Independent-Samples T-Test, which requires careful configuration before execution:



Within this dialogue window, assign the variables correctly: drag the continuous dependent variable, **Growth**, into the **Test Variable(s)** field. Next, drag the categorical independent variable, **Fertilizer**, into the **Grouping Variable** field. Critically, locate and activate the checkbox labeled **Estimate effect sizes**. This specific step instructs SPSS to compute [Cohen's d](#) and include it in the output tables.



The next step involves defining the numerical identifiers corresponding to the two groups within the dataset. Click the **Define Groups** button. In the resulting sub-window, input the identifier used for the first category (e.g., type **One** for **Group 1**) and the identifier for the second category (e.g., type **Two** for **Group 2**). These definitions link the analysis back to the levels of the Fertilizer variable.



Once both groups have been defined, click **Continue** to exit the definition window, and then click **OK** in the main Independent-Samples T Test dialogue box to execute the statistical analysis and generate the comprehensive output.

Analyzing the SPSS Output and Final Interpretation

Following the execution, [SPSS](#) generates several output tables. The section of interest for this analysis is specifically the "Effect Sizes" output block, which summarizes the calculated measures of magnitude. The table below represents the effect size output generated for the botanical growth study:

→ **T-Test**

Group Statistics					
	Fertilizer	N	Mean	Std. Deviation	Std. Error Mean
Growth	One	12	14.25	4.181	1.207
	Two	12	13.08	4.660	1.345

Independent Samples Test						
Levene's Test for Equality of Variances						
		F	Sig.	t	df	Sig One-Sided
Growth	Equal variances assumed	.032	.860	.646	22	.26
	Equal variances not assumed			.646	21.745	.26

Independent Samples Effect Sizes					
		Standardizer ^a	Point Estimate	95% Confidence Interval	
				Lower	Upper
Growth	Cohen's d	4.427	.264	-.543	1.064
	Hedges' correction	4.585	.254	-.525	1.028
	Glass's delta	4.660	.250	-.562	1.052

a. The denominator used in estimating the effect sizes.
 Cohen's d uses the pooled standard deviation.
 Hedges' correction uses the pooled standard deviation, plus a correction factor.
 Glass's delta uses the sample standard deviation of the control (i.e., the second) group.

To locate the calculated Cohen's d value, inspect the row labeled **Cohen's d** within the effect size output block. The precise numerical estimate is found in the column designated as **Point Estimate**. In the context of this fertilizer study, the calculated value for Cohen's d is determined to be **0.264**.

Interpreting this finding means that the average growth height of plants treated with Fertilizer #1 is **0.264 standard deviations** higher than the average growth height of plants treated with Fertilizer #2. To contextualize this magnitude, we compare it against the conventional rules of thumb established by Cohen:

0.2 = Small effect size.

0.5 = Medium effect size.

0.8 = Large effect size.

Since the calculated value of 0.264 is only marginally above the threshold for a small effect (0.2), the difference in plant growth attributed to the fertilizer type is concluded to be a **small effect size**. This indicates that even if the primary t-test yielded a [p-value](#) suggesting [statistical significance](#), the practical difference between the mean growth rates of the two fertilizer groups is minor and likely trivial in terms of real-world agricultural application.

Additional Statistical Resources

For researchers seeking to deepen their understanding of effect size calculation, interpretation, and application across diverse statistical methodologies, the following supplementary resources offer valuable guidance: