

Consulting Analytical Chemists and Geochemists

# MAKING SENSE OF SIGNIFICANT DIFFERENCE USING COHEN'S *d*

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**Application Note: 34**

## 1. INTRODUCTION

A  $p$ -value outcome in a statistical test of inference is a measure of the strength of evidence against a null hypothesis. Statistical significance validates that the findings of an experiment are not due to chance, but it does not provide a measure of how much (or how large) a difference there is. In statistical analysis of analytical results, we often want to know not just whether a result is significant, but how large the effect is. Effect size is simply the difference between the sample means being compared. However, effect size needs to be expressed as a standardised mean difference so that any comparisons are based on the same baseline comparison. Essentially, Cohen's  $d$  (Cohen, 1988) effect size represents the degree to which two data sets (samples of a population) do not overlap. Cohen's  $d$  calculates a standardised mean difference between two groups (McLeod, 2019). The less that two samples overlap, the larger the effect size. Cohen developed a system of comparison between two groups as being either "small", "medium", or "large". Cohen's  $d$  is only calculated after rejecting a null hypothesis. If the null is not rejected, effect size has no meaning.

Both effect size and statistical significance are essential for a comprehensive understanding of the results of a statistical experiment and should be reported when quoting an experimental outcome.

There are two Cohen's  $d$  formulas, equation [1] gives the formula for groups with equal numbers of values and equation [3] is used in unequal group sizes.

## 2. APPLICATION

Cohen's  $d$  is a standardised mean difference between two groups, where the mean of one group is subtracted from the other ( $\bar{x}_A - \bar{x}_B$ ) and divided by a pooled standard deviation,  $s_p$  [2], of both sets of data:

$$\text{Cohen's } d = \frac{(\bar{x}_A - \bar{x}_B)}{s_p} \quad [1]$$

$$s_P = \sqrt{\frac{s_A^2 + s_B^2}{2}} \quad [2]$$

For unequal group sizes, Cohen's  $d_s$  is used:

$$\text{Cohen's } d_s = \frac{(\bar{x}_A - \bar{x}_B)}{s_P} \quad [3]$$

$$s_P = \sqrt{\frac{(n_A - 1)s_A^2 + (n_B - 1)s_B^2}{n_A + n_B - 2}} \quad [4]$$

Where,  $n_A$  and  $n_B$  are the number of measurements of data set A and B, respectively.  $s_A$  and  $s_B$  are the standard deviation of data set A and B, respectively.

Cohen's  $d$  and  $d_s$ , represent the number of standard deviations by which the two groups differ; with a  $d$  or  $d_s$  of 1 indicating that the two groups differ by 1 standard deviation, a  $d$  of 2, by 2 standard deviations, and so on. More specifically, Cohen states that if  $d=0.2$  it can be considered a 'small' effect size,  $d=0.5$  represents a 'medium' effect size and  $d=0.8$  a 'large' effect size with  $\geq 1.2$ , as very large effect size.

## 2.1 Worked Example

Two laboratories are sent the same ore sample for copper determination of by ICP-OES. A statistical comparison of the means using a two-sample unequal variance t-test gives a  $p$ -value of 0.002, making the mean value A and B significantly different an alpha of 0.05. Figure 1 presents the results of both labs as Gaussian Distributions and Table 1 the figures of merit for the test work. How large is the difference between the two means? To answer this, Cohen's  $d$  is calculated and equation [1] and [2] are used as there are an equal number of results for each lab, i.e.,  $n=6$  for Lab A and  $n=6$  for Lab B. The pooled standard deviation is calculated followed by Cohen's  $d$ :

$$s_p = \sqrt{\frac{0.0701^2 + 0.0363^2}{2}} = 0.0559$$

Substituting into equation [1]:

$$\text{Cohen's } d = \frac{(12.07 - 11.93)}{0.0559}$$

$$\text{Cohen's } d = \frac{(0.14)}{0.0559} = 2.5$$

In summary, a t-test shows that there is a significant statistical difference between the mean values ( $p=0.002$  at  $\alpha$  of 0.05). In terms of Cohen's  $d$  convention, a 2.5 standard deviation demonstrates that the difference between the means is large.

Table 1. Figures of merit for two mean values for the assay of a copper ore sample for % copper by ICP-OES.

Lab A % Cu	12.07
$n$	6
Lab B % Cu	11.93
$n$	6
Lab A $s$	0.0701
Lab B $s$	0.0363
Spooled	0.0559
$p$ -value ( $\alpha=0.05$ )	0.002
Cohen's $d$	2.5

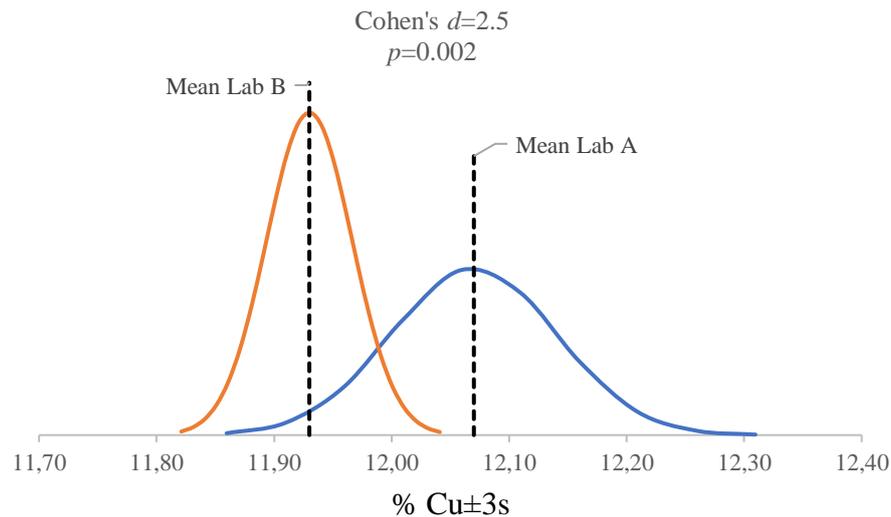


Figure 1. Means A and B are statistically significant ( $p=0.002$  and  $<0.05$ ), with a Cohen's  $d$  of 2.5. There are 2.5 standard deviations difference between the means making the effect size large according to Cohen's effect size convention.

### 3. CONCLUSION

In conclusion, Cohen's  $d$  values quantify the magnitude of the difference between mean values.

### 4. REFERENCES

Cohen, J. (1988). *Statistical power analysis for the behavioural sciences* (2nd ed.). Hillsdale, NJ: Erlbaum.

McLeod, S. A. (2019). What does effect size tell you? *Simply psychology*: <https://www.simplypsychology.org/effect-size.html>

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