

Chapter 6

Compare Two Groups

Section 6.3: Standardized Mean Difference, d

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- Standardized Mean Difference, d
 - Definition and Meaning
 - Separation of Distributions
 - Application

6.3a Definition and Meaning

Statistical Significance

An hypothesis test is the search for statistical significance

- ▶ Does a mean difference exist, that is, does $\mu_1 - \mu_2 \neq 0$?
- ▶ Regarding the relationship of grouping variable X with response variable Y, an hypothesis test results in either
 - “Yes, one true mean is likely larger than the other mean”
 - or
 - “No difference between true means detected”
- ▶ The *p*-value from the hypothesis test provides evidence as to the existence of a mean difference, but . . .
- ▶ **Key Concept:** If significant, less than $\alpha = 0.05$, the *p*-value by itself provides no information regarding the size of the difference between μ_1 and μ_2

Effect Size

Strength of a relationship

- ▶ **Effect Size:** The magnitude of an existing difference
- ▶ The effect size is the separation between the group means, expressed, for example, directly as the mean difference
 - sample: $m_1 - m_2$
 - population: $\mu_1 - \mu_2$, estimated with the confidence interval
- ▶ The effect size answers the question:
 - If the group means are not equal, $\mu_1 - \mu_2 \neq 0$, how large is the difference between them?
- ▶ That is, the difference between μ_1 and μ_2 is real, but . . .
 - Is $\mu_1 - \mu_2$ small and more likely not of interest even though it is nonzero?
 - Is $\mu_1 - \mu_2$ large and more likely of interest?

Practical Importance

Ultimate focus for the decision maker

- ▶ **Practical Importance:** An assessment as to whether the effect size is large enough to be useful
- ▶ Consider the effect of remodeling a retail outlet: compare average sales on a daily basis before and after the remodeling
 - If average sales per day increases, but only by \$0.01, the investment in remodeling is not justified
 - An average increase of \$148.34 per day may have justified the investment, so the effect would be of practical importance
- ▶ The effect size provides more information than provided by just the *p*-value
 - if *p*-value $> \alpha = 0.05$, then conclude a relationship exists
 - effect size indicates the magnitude of this relationship

Standardized Mean Difference

Traditionally called Cohen's d , after Jacob Cohen

- ▶ Express the **mean difference** in the sample as $m_1 - m_2$, and in the population as $\mu_1 - \mu_2$
- ▶ One assessment of effect size is the **mean difference** itself, expressed with units such as \$ USD, length, and hours
- ▶ Effect size can also be expressed in the universal standardized metric, as a **standardized mean difference**, based on the **standard deviation of the data**
- ▶ There are two samples of data, with **two standard deviations**, s_1 and s_2 , so standardize with the **within-group standard deviation**, s_w , calculated as an average from both samples
- ▶ **Standardized mean difference**: Number of standard deviations of the data that separate the group means

$$\text{Sample: } d = \frac{m_1 - m_2}{s_w}$$

$$\text{Population: } \delta = \frac{\mu_1 - \mu_2}{\sigma_w}$$

Standardized Mean Difference is Unitless

d and measurement scales

- ▶ The **size of the expressed mean difference** depends, in part, on the **scale of measurement**
- ▶ For example, the **same physical distance** is expressed with **larger numbers** if in inches instead of centimeters
- ▶ True of all standardized values, d is **unitless**
- ▶ **Unitless Statistic**: The value of a statistic does not depend on the **measurement scale** in which the variable is expressed
- ▶ For d , this unitless property means that, for example, in the analysis of two different average lengths, the **same effect size is obtained** if length is measured in inches or centimeters
- ▶ This unitless property facilitates the **comparison of the strength of an effect across different studies** of variable Y in which Y is measured with different measurement scales

Qualitative Description of Effect Size

Some guidelines that often, but not always, apply

- ▶ In any one situation the following guidelines may not apply, but they provide a useful context in the interpretation of d as long as **flexibility in their interpretation is maintained**
- ▶ Cohen studied many published scientific studies and, when possible from the available information, **calculated and classified the resulting values of d**
- ▶ From this analysis, he formulated some **general guidelines**
 - Trivial Effect: $d < .2$
 - Small Effect: $d > .2$ and $d < .6$
 - Medium Effect: $d > .6$ and $d < .8$
 - Large Effect: $d > .8$

6.3b Separation of Distributions

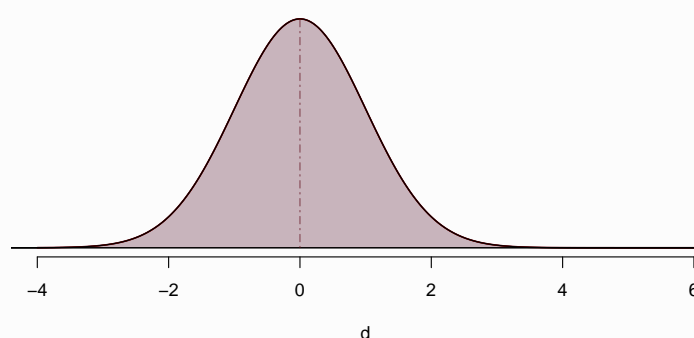
Express d in Terms of the Overlap Between Distributions

Meaning of the Standardized Mean Difference, d

- ▶ A focus on the size of the mean difference naturally leads to a consideration of **how much the two distributions of Y , one for each group, overlap**
- ▶ Consider the **amount of overlap** when Y is normally distributed for each group
- ▶ Express the distributions as **standardized normal distributions**, that is, expressed in terms of standard or z -values
- ▶ The following examples illustrate the **relation of the mean difference to the differences of the entire two standardized normal distributions**
- ▶ The standardized mean difference, d , calibrates the **separation of distributions**, best illustrated for normal distributions
- ▶ Question: How **distinct** are the two distributions depending on the value of d

Meaning of $d = 0.0$ in Terms of Distribution Overlap

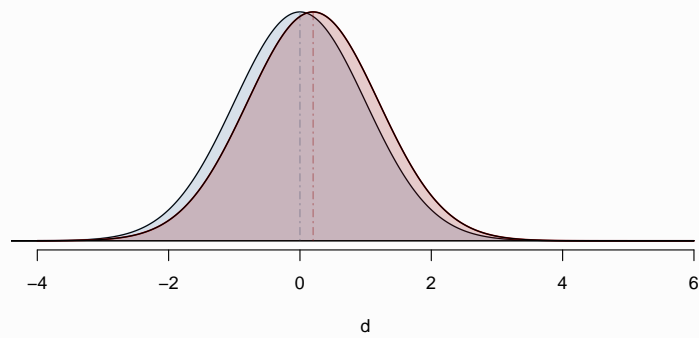
Cohen's $d = 0$, Proportion Overlap = 1



- ▶ Null hypothesis is true, $\mu_1 - \mu_2 = 0$, complete overlap

Meaning of $d = 0.2$ in Terms of Distribution Overlap

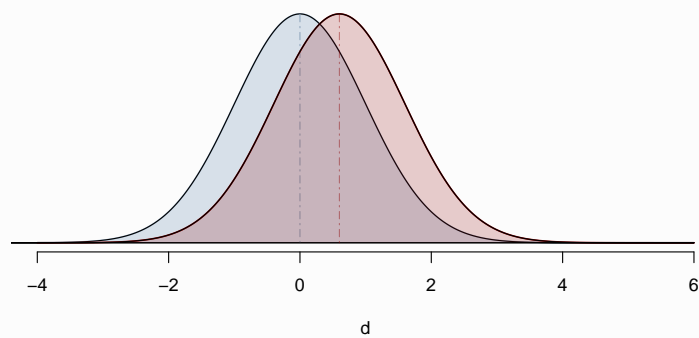
Cohen's $d = 0.2$, Proportion Overlap = 0.85



- $\mu_1 - \mu_2 \neq 0$, Effect is **small**, distributions mostly overlapped

Meaning of $d = 0.6$ in Terms of Distribution Overlap

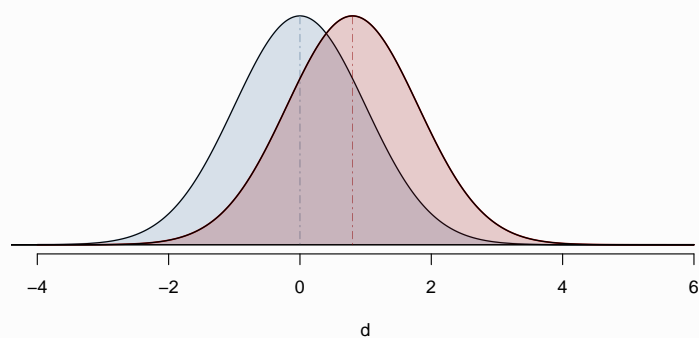
Cohen's $d = 0.6$, Proportion Overlap = 0.62



- $\mu_1 - \mu_2 \neq 0$, Effect is **moderate**, overlap pronounced

Meaning of $d = 0.8$ in Terms of Distribution Overlap

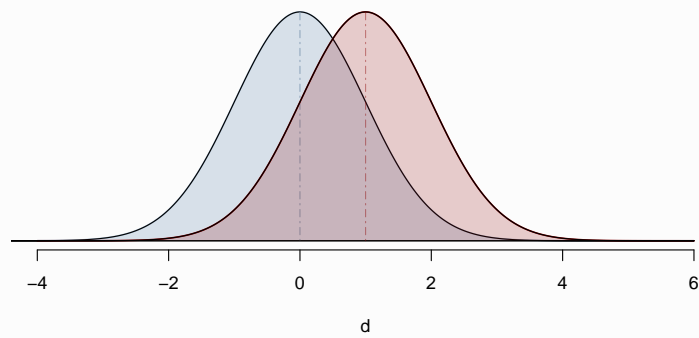
Cohen's $d = 0.8$, Proportion Overlap = 0.53



- $\mu_1 - \mu_2 \neq 0$, Effect is **large**, still more than 50% overlap

Meaning of $d = 1.00$ in Terms of Distribution Overlap

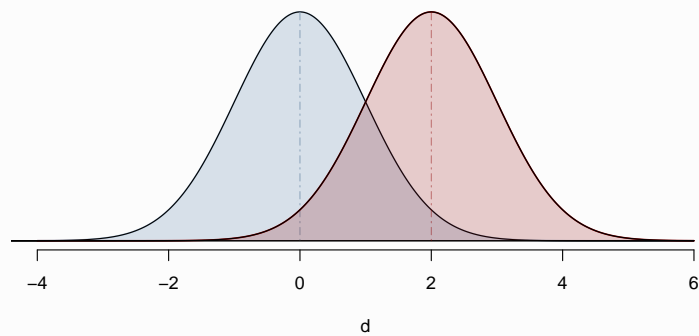
Cohen's $d = 1$, Proportion Overlap = 0.45



► $\mu_1 - \mu_2 \neq 0$, Effect is quite large, overlap still substantial

Meaning of $d = 2.00$ in Terms of Distribution Overlap

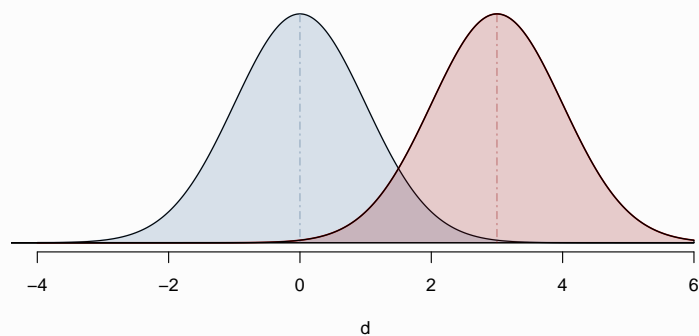
Cohen's $d = 2$, Proportion Overlap = 0.19



► $\mu_1 - \mu_2 \neq 0$, Effect is very large, overlap still considerable

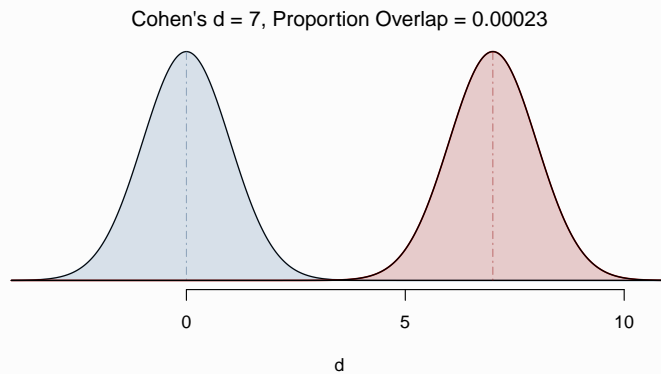
Meaning of $d = 3.00$ in Terms of Distribution Overlap

Cohen's $d = 3$, Proportion Overlap = 0.07



► $\mu_1 - \mu_2 \neq 0$, Effect is huge, still 7% percentage overlap

Meaning of $d = 7.00$ in Terms of Distribution Overlap



- ▶ 7 standard deviations of separation of means finally results in mostly distinct distributions, that is, with very little overlap

A More Complete Consideration of Differences

The mean difference is one part of a more complete scenario

- ▶ The difference between means is just one characteristic of the distinction between two distributions
- ▶ Effect size in terms of d can be expressed as the amount of overlap between the distributions of the response variable Y for each group
- ▶ Focus on the standardized mean difference, d , changes the emphasis from a summary index, the mean difference, to → a more complete consideration of the relation between two distributions
- ▶ Relate the two distributions plotted on the same graph to more completely reveal their differences
- ▶ Even a relatively large mean difference can reflect a substantial overlap between the distributions

6.3c Effect Size and Significance

t -statistic Depends on Size of Effect, d , and Sample Size

d and t can be expressed in terms of each other

- ▶ Divide the sample mean difference, $m_1 - m_2$, by either ...
 - standard deviation of the data, to obtain d
 - standard deviation of the mean difference, to obtain t

$$t_{m_1 - m_2} = \frac{m_1 - m_2}{s_{m_1 - m_2}} = \frac{m_1 - m_2}{s_w} \frac{1}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}} = d \frac{1}{\sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

- ▶ **Key Concept:** The assessment of statistical significance, assessed with the t -statistic, confounds the magnitude of the relationship, assessed with d , with the sample sizes, n_1 and n_2
- ▶ For any given value of effect size d , a significant t -value can be obtained by increasing the sample sizes of the two groups

Statistical Significance vs. Practical Importance

A significant p -value does not imply an important result

- ▶ The t -statistic from an hypothesis test, and its associated p -value, do *not* estimate the size of the population mean difference
- ▶ Even for a very small separation between the distributions of response variable Y for each group, d slightly larger than 0, large sample sizes, n_1 and n_2 , result in statistical significance
- ▶ When d is close to but still larger than zero, a difference from the null exists so that the null hypothesis is false, but the result may be of little or no practical importance
- ▶ Even with a result too small to be of practical importance, with a large enough sample, statistical significance is *always* obtained

Meaning of t vs. d

Move beyond just significance testing

- ▶ t and associated p -value: statistical significance, does a difference between population means exist?
- ▶ Mean difference and standardized mean difference, d : size of the difference as it relates to practical importance
- ▶ A statistically significant difference between means can *always* be found for sufficiently large sample sizes, no matter how small the difference between population means
- ▶ Accordingly, analysis of a mean difference should generally also include a consideration of the minimum mean difference of practical importance
- ▶ Express this minimum effect size in terms of the mean difference, in original units of measurement, mm_d , and/or as the minimum standardized mean difference, $msmd$

6.3d Application

Begin the Analysis

R instructions

- ▶ Invoke the `lessR` *t*-test function, `ttest()`, or `tt_brief()`, with **response variable** `Time` as a function of the **grouping variable** `Supplier`
 - > `tt_brief(Time ~ Supplier)`
- ▶ This application **extends the previous analyses** to a consideration of **effect size and practical importance**
 - **Section 6.1:** Statement of the problem, data, descriptive statistics, evaluation of stable process and normality assumptions, inferential analysis, conclusion
 - **Section 6.2:** Evaluation of homogeneity of variance assumption, standard error, inferential analysis not assuming homogeneity of variance

Effect Size

--- Effect Size ---

-- Assume equal population variances of Time
for each Supplier

Sample Mean Difference of Time: 1.49

Standardized Mean Difference of Time, Cohen's *d*: 0.98

95% Confidence Interval for *smd*: 0.26 to 1.70

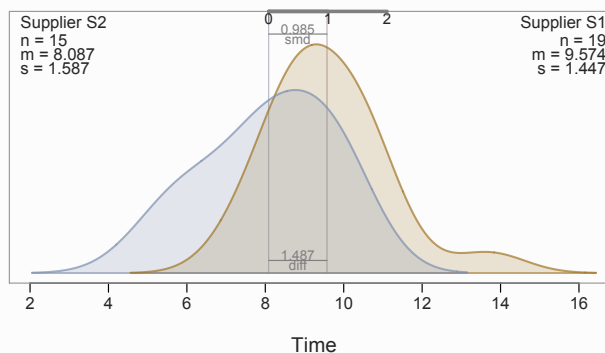
Graphical Display of Effect Size

Compare more than just the means

- ▶ Display the **mean difference in the metric** of
 - **unit of measurement** of the response variable, **Y**
 - **standardized units** of the response variable, **Cohen's d**
- ▶ **Directly display the overlapping distributions**, expressed as smoothed histograms, called density functions
- ▶ Observe the difference between the **entire distributions**
- ▶ An additional parameter for invoking `ttest()` specifies the **minimum effect size**, displayed in the resulting analysis
 - **mmd** for the minimum mean difference, or
 - **msmd** for the minimum standardized mean difference
- ▶ For example,

```
> tt_brief(Time ~ Supplier, mmd=0.75)
```

Overlapping Density Curves



Practical Importance

- ▶ As is true of any statistic, including the standardized mean difference, d , a **confidence interval** provides the range of plausible values for the corresponding **population value**, here δ

--- Practical Importance ---

Minimum Mean Difference of practical importance: mmd

Compare **mmd** = 0.75 to the obtained value of **md** = 1.49

Compare **mmd** to the confidence interval for **md**:

0.42 to 2.55

Minimum Standardized Mean Difference of

practical importance: msmd

Compare **msmd** = 0.5 to the obtained value of

smd = 0.98

Compare **msmd** to the confidence interval for **smd**:

0.26 to 1.7

Extent of Graphics Smoothing

--- Graphics Smoothing Parameter ---

Density bandwidth for Supplier S1: 0.91

Density bandwidth for Supplier S2: 1.05

-
- ▶ Increase the default bandwidth for a smoother density curve
 - ▶ Decrease the default bandwidth for a density curve with more fluctuations
 - ▶ Change the density bandwidth from the default with `bw1` and `bw2` options, for the first and second groups, respectively

Index Subtract 2 from each listed value to get the Slide #

effect size, 5
practical importance, 6

standardized mean difference, 7
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▶ The End