

# PSY 201: Statistics in Psychology

## Lecture 27

Hypothesis testing for dependent sample means

*Within is better than between.*

Greg Francis

Purdue University

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# DEPENDENT SAMPLES

- two samples of data are dependent when each score in one sample is paired with a specific score in the other sample
- e.g.
  - ▶ testing the same set of subjects before and after treatment
  - ▶ matched subjects in two groups (match along IQ before treatment and test mathematics)

# CORRELATED DATA

- when samples are dependent, the scores across samples may be correlated
- suggests that you can (partly) predict one from other
- removes some of the randomness from the samples
- generally a good thing (more control over variables)
- but requires slightly different analysis

# VARIABLE

- the variable of interest for dependent groups is the difference scores

$$d_i = X_{1i} - X_{2i}$$

- where
  - ▶ the  $i$ th scores in each group are matched (same subject)
  - ▶  $X_{1i}$  is the  $i$ th score in the first group
  - ▶  $X_{2i}$  is the  $i$ th score in the second group
- note that for dependent groups  $n_1 = n_2 = n$ , so we can calculate  $n$  difference scores

# HYPOTHESIS

- we can calculate the mean of the difference scores for the sample

$$\bar{d} = \frac{\sum d_i}{n}$$

- which is the same as

$$\bar{d} = \bar{X}_1 - \bar{X}_2$$

- we would like to know if the mean of difference scores for the **population** ( $\mu_1 - \mu_2$ ) is different from zero

$$H_0 : \mu_1 - \mu_2 = 0$$

$$H_a : \mu_1 - \mu_2 \neq 0$$

- This is actually the same as a one-sample  $t$  test!

# STANDARD ERROR

- we estimate the standard error with

$$s_{\bar{d}} = \frac{s_d}{\sqrt{n}}$$

- where  $s_d$  is the standard deviation of the difference scores

$$s_d = \sqrt{\frac{\sum (d_i - \bar{d})^2}{n - 1}}$$

- or in raw score form

$$s_d = \sqrt{\frac{\sum d_i^2 - (\sum d_i)^2 / n}{n - 1}}$$

# TEST STATISTIC

- the test statistic is the same in form as all those before

$$\text{Test statistic} = \frac{\text{Statistic} - \text{Parameter}}{\text{Standard Error}}$$

- for our specific situation it is

$$t = \frac{\bar{d} - (\mu_1 - \mu_2)}{s_{\bar{d}}}$$

- which is used to compute a  $p$ -value in a  $t$ -distribution with  $n - 1$  degrees of freedom

# EXAMPLE

- do your thoughts control your autonomic processes?
- relax and take your pulse for 30 seconds
  - ▶ write the number down ( $X_1$ )
- picture yourself running and then take your pulse for 30 seconds
  - ▶ write the number down ( $X_2$ )
- we want to know if the mean difference across the two measurements (samples) is different from zero

# EXAMPLE

- the measurements are dependent because if you tend to have a high pulse rate, it will be high for both measurements
- but we are interested in the **difference**, so the overall rate is unimportant
- calculate the difference of your pulse rates

$$d_i = X_{1i} - X_{2i}$$

# HYPOTHESIS

- (1) our null hypothesis is that there is no effect of imagination on pulse rate

$$H_0 : \mu_1 - \mu_2 = 0$$

- the alternative hypothesis is that there is an effect

$$H_a : \mu_1 - \mu_2 < 0$$

- note, this is a directional hypothesis because we suspect that thinking about exercise should **increase** heart rate
- we will use a level of significance of  $\alpha = 0.05$

# DATA

- take a sample of pulse rate differences from ten people
- with your sampled data calculate the sample mean

$$\bar{d} = \frac{\sum d_i}{10}$$

- and the sample standard deviation (you can use the on-line calculator)

$$s_d = \sqrt{\frac{\sum d_i^2 - (\sum d_i)^2 / 10}{9}}$$

- and the estimate of the standard error

$$s_{\bar{d}} = \frac{s_d}{\sqrt{10}}$$

# TEST STATISTIC

- (2) now calculate the test statistic as

$$t = \frac{\bar{d} - (\mu_1 - \mu_2)}{s_{\bar{d}}}$$

- since we assume  $\mu_1 - \mu_2 = 0$  this is

$$t = \frac{\bar{d}}{s_{\bar{d}}}$$

## $p$ VALUE

- (3) You can compute the corresponding  $p$ -value with the  $t$ -Distribution Calculator for a one-tailed test
- with your sample of 10 people you have

$$df = n - 1 = 9$$

- if

$$p < \alpha = 0.05$$

- you reject  $H_0$

# DECISION

- (4) if you reject  $H_0$  that means there is evidence that imagination of exercise **does** affect heart rate
- if you do not reject  $H_0$  that means there is no evidence that imagination of exercise affects heart rate
- if you reject, that means that if  $\mu_1 - \mu_2 = 0$ , then the probability of the observed (or a more extreme) sample mean  $\bar{d}$  value is less than 0.05

# SIGNIFICANCE VS. IMPORTANCE

- if you failed to reject  $H_0$ , it may have been because you had too small a sample,  $n$ ,
- or may have been because there was no real difference
- in principle, it is hard to believe that imagined running has **no effect at all** on pulse rate
  - ▶ Surely the brain uses energy differently during imagined running compared to not
- the effect might be very small, so small that our experiment cannot find it

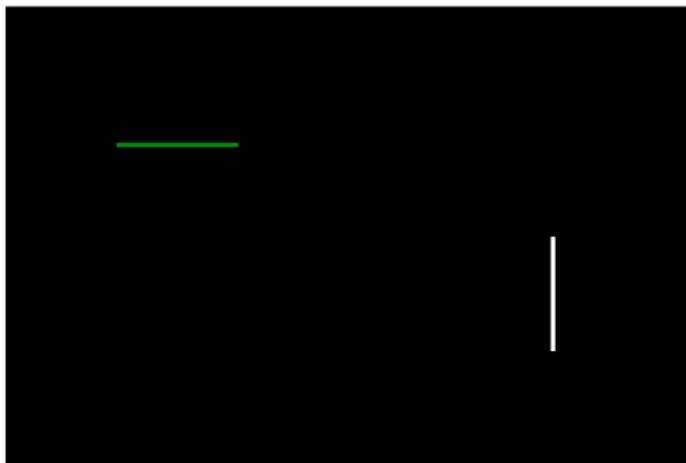
# SIGNIFICANCE VS. IMPORTANCE

- on the other hand
- probably everyone had a sample difference  $\bar{d}$  that was non-zero
- but some people probably did not reject  $H_0$
- we cannot just look at numbers like  $\bar{d}$  and take them at face value
- the statistical procedures keep us from rushing to conclusions that are unwarranted

# POWER

- The computation of power is very similar (actually, identical) to the one-sample  $t$ -test situation
- Consider the STATLAB Horizontal-Vertical illusion. Across the class, we have data from 26 students that reports the mean (for each student) matching length for a horizontal and a vertical line.

Trials to go: 29



Start Next Trial

Submit match

# POWER

- Suppose you want to test for a difference between matching lengths for a horizontal and vertical line.

$$H_0 : \mu_1 - \mu_2 = 0$$

- How many subjects should you use to have 90% power?
- We can use the STATLAB data to estimate power and then compute an appropriate sample size
- From the STATLAB data we find:

$$\bar{X}_1 - \bar{X}_2 = 99.3213 - 105.6313 = -6.31$$

$$s_d = 4.655993$$

# ONLINE CALCULATOR

Specify the population characteristics:

$$H_0 : \mu_1 - \mu_2 = 0$$

$$H_a : \mu_{a1} - \mu_{a2} = -6.31$$

Enter the standard deviation of each population and the correlation between scores...

$$\sigma_1 = \text{NA}$$

$$\sigma_2 = \text{NA}$$

$$\text{Population correlation: } \rho = \text{NA}$$

...or enter the standard deviation of difference scores:

$$\sigma_d = 4.6559$$

Or enter a standardized effect size:

$$\frac{(\mu_{a1} - \mu_{a2}) - (\mu_1 - \mu_2)}{\sigma_d} = \delta = -1.355269$$

Specify the properties of the test:

Type of test

Type I error rate,  $\alpha = 0.05$

Power =

Sample size (pairs of scores)  $n = 8$

Calculate minimum sample size

Calculate power

# ONLINE CALCULATOR

- You get exactly the same numbers using the one-sample calculator:

Specify the population characteristics:

$$H_0 : \mu_0 = 0$$

$$H_a : \mu_a = -6.31$$

$$\sigma = 4.6560$$

Or enter a standardized effect size

$$\frac{\mu_a - \mu_0}{\sigma} = \delta = -1.355240$$

Specify the properties of the test:

Type of test

Type I error rate,  $\alpha = 0.05$

Power =

Sample size,  $n = 8$

Calculate minimum sample size

Calculate power

# ONLINE CALCULATOR

- Instead of using  $s_d$  (the standard deviation of the differences), you could use the standard deviation of each group and the correlation between scores:

Specify the population characteristics:

$$H_0 : \mu_1 - \mu_2 = 0$$

$$H_a : \mu_{a1} - \mu_{a2} = -6.31$$

Enter the standard deviation of each population and the correlation between scores...

$$\sigma_1 = 3.5502$$

$$\sigma_2 = 6.4328$$

$$\text{Population correlation: } \rho = 0.7073$$

...or enter the standard deviation of difference scores:

$$\sigma_d = 4.656027$$

Or enter a standardized effect size:

$$\frac{(\mu_{a1} - \mu_{a2}) - (\mu_1 - \mu_2)}{\sigma_d} = \delta = -1.355232$$

Specify the properties of the test:

Type of test

Type I error rate,  $\alpha = 0.05$

Power =

Sample size (pairs of scores)  $n = 8$

Calculate minimum sample size

Calculate power

# INDEPENDENT TEST

- Suppose you wanted to do the experiment with different subjects assigned to different line orientations. How many subjects do you need?
- Independent Means Power Calculator

Specify the population characteristics:

$$H_0 : \mu_1 - \mu_2 = 0$$

$$H_a : \mu_{a1} - \mu_{a2} = -6.31$$

$$\sigma_1 = 3.5502$$

$$\sigma_2 = 6.4328$$

Or enter a standardized effect size

$$\frac{(\mu_{a1} - \mu_{a2}) - (\mu_1 - \mu_2)}{\sigma} = \delta = \text{NA}$$

Specify the properties of the test:

Type of test

Type I error rate,  $\alpha =$

Power =

Sample size for group 1,  $n_1 =$

Sample size for group 2,  $n_2 =$

Calculate minimum sample size

Calculate power

- 4 times as many subjects for an independent means experiment!

# WITHIN vs. BETWEEN

- A within subjects (dependent test) design is usually more powerful than a between subjects (independent test)
- This is because you are able to remove one source of variability from the standard error
  - ▶ The variability in overall score values
- Standard error reflects variability between conditions
- A between subjects calculation of variability includes variability between conditions and variability across subjects (more variability!)

# CONCLUSIONS

- dependent samples
- very important for lots of interesting tests
- more powerful than independent tests

# NEXT TIME

- two-sample case for independent proportions
- hypothesis testing
- confidence interval
- power

*What is a “margin of error”?*