

PSY 3307

Two Sample t-test Pt2

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Agenda

- Review terms
 - independent samples t-test, homogeneity of variance, pooled variance, standard error of the difference
- Between & Within Designs
- One Tailed Independent Samples t-test
- Paired Sample t-test
- Performing a Paired Sample t-test
- Effect Size

One Sample t-test

- used when we are confident that the direction of the relationship
- if you think one group will have a higher score/value then you would make that statement in your alternative hypothesis

Steps for one-tailed test

- Variable(Groups) = Intervention(Got Intervention/Control)
- Outcome = Fruit & Vegetable Intake

1. decide which sample/group score you think will be larger

- I think that the intervention group will eat more fruits and vegetables

2. decide which condition/group to subtract from the other

- `intervention group - control group`

3. decide whether the difference will be positive or negative

- a positive value should be returned from the previous equation

1. create hypotheses

- The intervention group will eat the same or less fruits and vegetables than the control group
- $H_0: \mu(\text{intervention group}) - \mu(\text{control group}) \leq 0$
- The intervention group will eat more fruits and vegetables than the control group
- $H_1: \mu(\text{intervention group}) - \mu(\text{control group}) > 0$

1. locate regions of rejection

- since we are interested in a positive difference in our hypothesis, we'll only be looking at the upper tail of the distribution

2. conduct your independent samples t-test

- make sure your groups/samples are subtracted the same way as your hypotheses
- **intervention group - control group**

Paired Samples t-test

- within-subjects/groups design
- also called related-samples t-test
 - each participant gets measured in each condition
- an example would be an intervention for weight loss where everyone's weight is measured before (first time point to measure) the intervention and after (second time point to measure)
- **matched-samples design** is when a participant in one condition is matched with a participant in the other condition
- **repeated-measures design** is when each participant gets measured twice
 - can be more for more rigorous statistics, but not the ones we'll learn about in this class

Why we use paired-samples t-tests

- pretest/posttest design
 - before everyone gets the intervention/experiment (pretest or first time point) and then after the intervention/experiment (posttest or second time point)

- we subtract every person's before/after score from the before/score score to get a **difference** score
- you can subtract whatever score you want from the other (before - after or after - before)
- **mean differences** the mean of the differences between the paired scores
 - represented as \bar{d}
- add all the values of differences for before/after to get a sum and then divide by the number of participants (each person has a before and after score or the number of difference scores)

$$\text{Sample mean difference} = \bar{D}$$

$$\text{population mean difference} = \mu_D$$

- now because we have a sample mean from a sample, we can now perform a one-sample t-test

Hypotheses

- our null hypothesis is that there will be no change in scores on both occasions so there should be a difference of zero

$$H_0 : \mu_D = 0$$

- our alternative hypothesis is that either our before or after scores should be higher
 - our hypothesis supports either a positive or negative change

$$H_1 : \mu_D \neq 0$$

- similar to the independent samples t-test, our sampling distribution is of mean differences

Steps to paired-samples t-test

Calculate the estimated variance of the difference scores

$$s_D^2 = \frac{\sum D^2 - \frac{(\sum D)^2}{N}}{N - 1}$$

Note the notation changes from X to D because we are working with differences in scores and not means

Calculate the standard error of the mean differences

$$s_{\bar{D}} = \sqrt{\frac{s_D^2}{N}}$$

Find the obtained t-value

$$t_{obt} = \frac{\bar{D} - \mu_D}{S_{\bar{D}}}$$

The population difference will be zero unless you are testing for a nonzero difference

Calculate the degrees of freedom

$$df = N - 1$$

Calculate effect size

$$d = \frac{\overline{D}}{\sqrt{S_D^2}}$$

OR

$$r_{pb}^2 = \sqrt{\frac{(t_{obt})^2}{(t_{obt})^2 + df}}$$

Discuss Significance

- Is the difference between time point 1 and time point 2 statistically significant?

Example

```
set.seed(100521)
t1 = rnorm(10, mean = 10, sd = 1.5)
t2 = rnorm(10, mean = 5.8, sd = 4)
df <- data.frame(participant = 1:10,
                  t1 = round(t1, 2),
                  t2 = round(t2, 2),
                  score_difference = c("_____", "_____"),
                  score_difference_squared = c("_____", "_____"))
```

```
df$t2
```

```
## [1] 12.75 10.54 7.51 4.32 9.95 3.64 6.45 3.40 3.43 13.77
```

```
df$t1
```

```
## [1] 8.32 13.96 8.17 8.40 8.90 7.58 9.67 11.51 10.13 11.49
```

```
df2 <- df %>%  
  mutate(score_difference = t2 - t1,  
         score_difference_squared = score_difference^2)  
  
df2 %>%  
  summarize(mean = mean(score_difference),  
            variance = sd(score_difference)^2,  
            sd = sd(score_difference))
```

```
##      mean variance      sd  
## 1 -2.237 15.69073 3.961153
```

12.75 - 8.32

[1] 4.43

10.54 - 13.96

[1] -3.42

7.51 - 8.17

[1] -0.66

4.32 - 8.40

[1] -4.08

9.95 - 8.90

[1] 1.05

3.64 - 7.58

[1] -3.94

```
df <- data.frame(participant = 1:10,  
  t1 = round(t1, 2),  
  t2 = round(t2, 2),  
  score_difference = c(4.43, -3.42, -.66, -4.08, 1.05, -3.42, 1.05, -3.42, 1.05, -3.42),  
  score_difference_squared = c("-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----"),  
  "-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----", "-----")
```

```
df$score_difference
```

```
##      [1]  4.43 -3.42 -0.66 -4.08  1.05 -3.94 -3.22 -8.11 -6.70  2.28
```

```
(4.43)^2
```

```
## [1] 19.6249
```

```
(-3.42)^2
```

```
## [1] 11.6964
```

```
(-.66)^2
```

```
## [1] 0.4356
```

```
(-4.08)^2
```

```
## [1] 16.6464
```

```
(1.05)^2
```

```
## [1] 1.1025
```

```
(-3.94)^2
```

```
## [1] 15.5236
```

```
df <- data.frame(participant = 1:10,
  t1 = round(t1, 2),
  t2 = round(t2, 2),
  score_difference = c(4.43, -3.42, -.66, -4.08, 1.05, -3.94, -3.22, -8.11, -6.70, 2.28),
  score_difference_squared = c(19.62, 11.70, .44, 16.64, 1.10, 15.52, 10.37, 65.77, 44.89, 5.20),
  df
```

##	participant	t1	t2	score_difference	score_difference_squared
## 1	1	8.32	12.75	4.43	19.62
## 2	2	13.96	10.54	-3.42	11.70
## 3	3	8.17	7.51	-0.66	0.44
## 4	4	8.40	4.32	-4.08	16.64
## 5	5	8.90	9.95	1.05	1.10
## 6	6	7.58	3.64	-3.94	15.52
## 7	7	9.67	6.45	-3.22	10.37
## 8	8	11.51	3.40	-8.11	65.77
## 9	9	10.13	3.43	-6.70	44.89
## 10	10	11.49	13.77	2.28	5.20


```
df$score_difference
```

```
## [1] 4.43 -3.42 -0.66 -4.08 1.05 -3.94 -3.22 -8.11 -6.70 2.28
```

```
4.43 +-3.42 +-0.66 +-4.08 +1.05 +-3.94 +-3.22 +-8.11 +-6.70 +2.28
```

```
## [1] -22.37
```

```
# sum difference is -22.37
```

```
-22.37/10
```

```
## [1] -2.237
```

```
# mean difference is -2.24
```

Calculate the estimated variance of the difference scores

$$s_D^2 = \frac{\sum D^2 - \frac{(\sum D)^2}{N}}{N - 1}$$

$$s_D^2 = \frac{\sum D^2 - \frac{(-22.37)^2}{10}}{10 - 1}$$

```
19.62+ 11.70+ .44+ 16.64+ 1.10+ 15.52+ 10.37+ 65.77+ 44.89+ 5.20
```

```
## [1] 191.25
```

$$s_D^2 = \frac{191.25 - \frac{(-22.37)^2}{10}}{10 - 1}$$

```
(-22.37)^2
```

```
## [1] 500.4169
```

$$s_D^2 = \frac{191.25 - \frac{500.42}{10}}{10 - 1}$$

```
500.42/10
```

```
## [1] 50.042
```

$$s_D^2 = \frac{191.25 - 50.04}{9}$$

```
191.25 - 50.04
```

```
## [1] 141.21
```

$$s_D^2 = \frac{141.21}{9}$$

```
141.21/9
```

```
## [1] 15.69
```

$$s_D^2 = 15.69$$

Calculate the standard error of the mean differences

$$S_{\bar{D}} = \sqrt{\frac{S_D^2}{N}}$$

$$S_{\bar{D}} = \sqrt{\frac{15.69}{10}}$$

15.69/10

[1] 1.569

$$S_{\overline{D}} = \sqrt{1.57}$$

```
sqrt(1.57)
```

```
## [1] 1.252996
```

$$S_{\overline{D}} = 1.25$$

Find the obtained t-value

$$t_{obt} = \frac{\bar{D} - \mu_D}{S_{\bar{D}}}$$

$$t_{obt} = \frac{-2.24 - 0}{1.25}$$

```
-2.24 - 0
```

```
## [1] -2.24
```

$$t_{obt} = \frac{-2.24}{1.25}$$

```
-2.24/1.25
```

```
## [1] -1.792
```

$$t_{obt} = -1.79$$

Calculate the degrees of freedom

$$df = N - 1$$

```
10 - 1
```

```
## [1] 9
```

$$df = 9$$

Calculate effect size

$$d = \frac{\overline{D}}{\sqrt{S_D^2}}$$

```
mean(df$score_difference)
```

```
## [1] -2.237
```

```
sd(df$score_difference)^2
```

```
## [1] 15.69073
```

```
-2.24
```

```
## [1] -2.24
```

```
15.69
```

```
## [1] 15.69
```

$$d = \frac{-2.24}{\sqrt{15.69}}$$


```
sqrt(15.69)
```

```
## [1] 3.96106
```

```
sd(df$score_difference)
```

```
## [1] 3.961153
```

$$d = \frac{-2.24}{3.96}$$

```
-2.24/3.96
```

```
## [1] -0.5656566
```

$$d = -.57$$

so really this means

$$d = .57$$

$$r_{pb}^2 = \sqrt{\frac{(t_{obt})^2}{(t_{obt})^2 + df}}$$

$$r_{pb}^2 = \sqrt{\frac{(-1.79)^2}{(-1.79)^2 + df}}$$

```
(-1.79)^2
```

```
## [1] 3.2041
```

$$r_{pb}^2 = \sqrt{\frac{3.20}{3.20 + 9}}$$

```
3.20 + 9
```

```
## [1] 12.2
```

$$r_{pb}^2 = \sqrt{\frac{3.20}{12.20}}$$

3.2/12.2

```
## [1] 0.2622951
```

$$r_{pb}^2 = \sqrt{.26}$$

```
sqrt(.26)
```

```
## [1] 0.509902
```

$$r_{pb}^2 = .51$$

Discuss Significance

t-obtained value of -1.79

t-critical value of

One-tailed paired-samples t-test

- choose whether you think your **after** score would be lower/higher than the **before** score
- have your hypotheses reflect that
- if you think after scores should be higher than before (learning intervention) then the difference scores should be positive
 - because you subtracted before scores from after scores

$$H_a : \mu_D > 0$$

$$H_0 : \mu_D \leq 0$$

Reporting

- report in a similar style to all other t-test
 - $t(df) = t \text{ value}, p \text{ value}$
- you'll also report the means of the before and after scores
 - you won't report the difference between the two means

Effect Sizes

- **effect size** is the amount of influence that changing the conditions of the IV has on the DV
- **cohen's d** is a measure of effect size in two-sample studies that reflects the magnitude of difference
 - small = .2
 - medium = .5
 - large = .8
- larger effect size means stronger the strength of the association/relationship between the IV and DV

Effect Size using Proportion of Variance Accounted For

- used to determine how consistently scores change
- **proportion of variance accounted for** is the proportion of differences in DV scores associated with changing the conditions of the IV
 - effect size using **squared point-biserial correlation coefficient**, which indicates the proportion of variance in DV scores that is accounted for by IV variable in two-sample studies, are below
 - small = .09
 - moderate = .10-.25
 - large = .25

Practice Time

```
set.seed(093021)
```

```
mistakes_made_tutor = rnorm(10, mean = 1.5, sd = 1.4)
```

```
mistakes_made_control = rnorm(8, mean = 4.1, sd = 1)
```

```
mistakes_made_tutor
```

```
## [1] 3.0495964 0.7931460 2.1999939 0.4106114 1.3852964 2.2035563 1.8161635
```

```
## [8] 2.4894226 3.1436843 1.0698578
```

```
mistakes_made_control
```

```
## [1] 5.347698 3.962054 5.072715 3.985207 4.945898 2.984944 5.004625 5.405361
```

```
set.seed(093021)
```

```
translating_native_speaker = rnorm(9, mean = 20, sd = 4.7)
```

```
translating_non_native = rnorm(14, mean = 10, sd = .99)
```

```
translating_native_speaker
```

```
## [1] 25.20222 17.62699 22.34998 16.34277 19.61492 22.36194 21.06141 23.32163
```

```
## [9] 25.51808
```

```
translating_non_native
```

```
## [1] 9.695828 11.235221 9.863433 10.962988 9.886355 10.837439 8.896095
```

```
## [8] 10.895579 11.292308 9.102475 12.603552 9.039360 10.388593 9.487347
```

```
set.seed(093021)
```

```
first_gen_bmi = rnorm(6, mean = 22, sd = 2.2)
```

```
second_gen_bmi = rnorm(9, mean = 28, sd = 5)
```

```
first_gen_bmi
```

```
## [1] 24.43508 20.88923 23.09999 20.28810 21.81975 23.10559
```

```
second_gen_bmi
```

```
## [1] 29.12916 31.53365 33.87030 26.46378 34.23849 27.31027 32.86358 27.42604
```

```
## [9] 32.22949
```

```

set.seed(100521)
t1 = rnorm(10, mean = 15, sd = 2.7)
t2 = rnorm(10, mean = 12, sd = 2.1)

df1 <- data.frame(participant = 1:10,
                  t1 = round(t1, 2),
                  t2 = round(t2, 2),
                  score_difference = c("_____", "_____"),
                  score_difference_squared = c("_____", "_____"))

df1$t2

```

```
## [1] 15.65 14.49 12.90 11.22 14.18 10.86 12.34 10.74 10.76 16.19
```

```
df1$t1
```

```
## [1] 11.98 22.12 11.71 12.11 13.02 10.65 14.40 17.73 15.23 17.69
```



```

set.seed(100521)
t1 = rnorm(10, mean = 8, sd = 1.2)

t2 = rnorm(10, mean = 10.5, sd = 4.8)

df11 <- data.frame(participant = 1:10,
                   t1 = round(t1, 2),
                   t2 = round(t2, 2),
                   score_difference = c("_____", "_____"),
                                     "_____", "_____",
                                     "_____", "_____",
                   score_difference_squared = c("_____", "_____",
                                                "_____", "_____",
                                                "_____", "_____",
                                                "_____", "_____",
                                                "_____", "_____"))

```

```
df11$t2
```

```
## [1] 18.83 16.19 12.55 8.72 15.48 7.90 11.28 7.62 7.66 20.07
```

```
df11$t1
```

```
## [1] 6.66 11.17 6.54 6.72 7.12 6.06 7.73 9.21 8.10 9.19
```

```

set.seed(100521)
t1 = rnorm(10, mean = 15, sd = 5.6)

t2 = rnorm(10, mean = 4.3, sd = 5)

df111 <- data.frame(participant = 1:10,
                    t1 = round(t1, 2),
                    t2 = round(t2, 2),
                    score_difference = c("_____", "_____",
                                         "_____", "_____"),
                    score_difference_squared = c("_____", "_____",
                                                  "_____", "_____"))

```

```
df111$t2
```

```
## [1] 12.98 10.22 6.43 2.45 9.49 1.60 5.11 1.30 1.34 14.27
```

```
df111$t1
```

```
## [1] 8.73 29.78 8.18 9.01 10.89 5.97 13.76 20.65 15.49 20.57
```