

hw\_lect18\_1

use  $z$  or  $t$ .

For the data you collected, consider one of the continuous variables (call it  $y$ ), and one of the categorical variables (call it  $x$ ). Let  $\mu_1$  denote the true mean of  $y$  when  $x =$  (first level of  $x$ ), and  $\mu_2$  denote the true mean of  $y$  when  $x =$  (2nd level of  $x$ ).

a) compute a  $t$ -based, 2-sided, 95% C.I. for  $\mu_1 - \mu_2$ .

b) Is there evidence from data that  $\mu_1$  and  $\mu_2$  are different?

Suppose  $y$  contains the cont. data, and  $x$  takes values 1, 2.

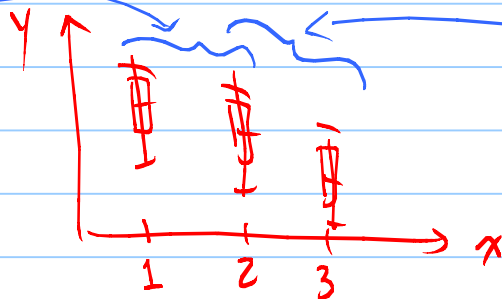
Generally, your code would look like this:

```
t.test(y[x1 == 1], y[x2 == 2])
```

If you want to compare the  $y$  at 2 other levels of  $x$ , e.g., 2, 3:

```
t.test(y[x1 == 2], y[x2 == 3])
```

In a previous hw problem you made a plot like this:



The above  $t$ -tests allow you to test whether there is a difference in the true means of  $y$  at 2 different levels of  $x$ .

Later, in ch-9 we will learn about a more efficient way of testing the means of  $y$  at different  $x$ 's.

## hw\_lect18\_2

Let  $\pi_1$  denote the true proportion of defective bridges in the USA, and  $\pi_2$  .... in Canada. A sample of  $n_1=80$ , and  $n_2=50$  bridges from the two countries, respectively, is taken, and it is found that 21% of the bridges in the USA, and 10% of the bridges in Canada are defective. At 95% confidence level

a) Is there evidence that the true proportions are different?

b) Is there evidence that  $\pi_1$  is larger than  $\pi_2$ ? This type of question requires a 1-sided CI which we are skipping this quarter. So, skip part b.

$$\begin{aligned} \text{a) 2-sided CI : } (P_1 - P_2) \pm 1.96 \sqrt{\frac{P_1(1-P_1)}{n_1} + \frac{P_2(1-P_2)}{n_2}} \\ (0.21 - 0.10) \pm 1.96 \sqrt{\frac{0.21 \cdot 0.79}{80} + \frac{0.10 \cdot 0.90}{50}} \\ 0.11 \pm 1.96(0.0622) = 0.11 \pm 0.1219 \\ \text{Does include zero.} \quad = \underline{(-0.0119, 0.232)} \end{aligned}$$

$\therefore$  There is no evidence That  $\pi_1 \neq \pi_2$

$\therefore$  We cannot conclude That  $\pi_1 \neq \pi_2$  (nor  $\pi_1 = \pi_2$ ).

Part b) is only FYI

$$\begin{aligned} \text{b) } \pi_1 > \pi_2 \Rightarrow \pi_1 - \pi_2 > 0 \Rightarrow \underline{\text{Lower Conf. bound for } \pi_1 - \pi_2 :} \\ (P_1 - P_2) - 1.645 \sqrt{\dots + \dots} = 0.11 - 1.645(0.0622) = \underline{0.007} \end{aligned}$$

$\therefore$  95% Conf. That  $\pi_1 - \pi_2 > .007$ .

This interval does NOT include zero.

$\therefore$  There (is) evidence That  $\pi_1 > \pi_2$ .

Note the "apparent" contradiction in the conclusions!  
Same Data! Different conclusions (from different questions)

## hw\_lect19\_1

use t-dist.

# Consider the following data on x1 and x2 which was collected in a paired design:

x1 = c(-0.27, -0.14, 1.61, 0.09, 0.00, 2.07, 0.56, -1.67, -0.51, -0.54)

x2 = c(-0.32, 0.20, 1.93, 0.54, 0.75, 1.77, 0.84, -0.29, -0.33, 0.17)

# a) Compute a 2-sided, 95% CI for the difference between the two true means. You may use R to do

# simple calculations, but use the CI formulas derived in class.

```
n = length(x1)
```

```
d = x1 - x2
```

```
dbar = mean(d)
```

```
s_d = sd(d)
```

```
dbar - 2.262 * s_d/sqrt(n) # -0.738 (t*=2.262 is from table 4, 2-sided 95% column with df=10-1=9; in R, it  
can be obtained this way: qt(.05/2, df = 10-1, lower.tail=F))
```

```
dbar + 2.262 * s_d/sqrt(n) # -0.073
```

# b) Provide one interpretation of the observed CI, AND state the conclusion in English, i.e., the "corollary."

# We are 95% confident that  $\mu_1 - \mu_2$  is in the interval (-0.7380, -0.073)

# Corollary: There is a difference between the two means

# BTW, you can "test" that x1 and x2 are paired by looking at their scatterplot:

```
plot(x1,x2) # I see a linear association.
```

# c) Consider the following data, which is the same as above, except the cases in x2 have been randomly shuffled. Compute an

# appropriate 95% 2-sided CI.

```
y1 = c(-0.27, -0.14, 1.61, 0.09, 0.00, 2.07, 0.56, -1.67, -0.51, -0.54)
```

```
y2 = c(0.20, 0.54, -0.33, 1.93, -0.32, 1.77, 0.75, 0.17, -0.29, 0.84)
```

```
n1 = length(y1)
```

```
n2 = length(y2)
```

```
y1bar = mean(y1)
```

```
y2bar = mean(y2)
```

```
sd1 = sd(y1)
```

```
sd2 = sd(y2)
```

```
std.err = sqrt(sd1^2/n1 + sd2^2/n2)
```

```
df_welch = (sd1^2/n1 + sd2^2/n2)^2 / ((sd1^2/n1)^(n1-1) + (sd2^2/n2)^(n2-1)) # 16.76825 i.e., about 17
```

```
(y1bar - y2bar) - 2.110 * std.err # -1.309 t* = 2.110 from Table IV or from R: qt(.05/2, df = 17,  
lower.tail=F)
```

```
(y1bar - y2bar) + 2.110 * std.err # +0.498
```

# d) Provide one interpretation of the observed CI, AND state the conclusion in English, i.e., the "corollary."

# We are 95% confident that  $\mu_1 - \mu_2$  is in the interval (-1.309, +0.498)

# Corollary: We cannot tell if there is a difference between the two means.

# e) Which one is narrower?

# The width of the paired interval is

```
2.262 * s_d/sqrt(n) # 0.3326533
```

# The width of the unpaired interval is

```
2.110 * std.err # 0.9035258
```

# The paired one is narrower, i.e., more precise (or more reliable). And that is why the paired CI is able to  
# "see" a difference between the means while the unpaired CI cannot detect the difference.