

CA4VdVeen

Ties van der Veen

1-3-2020

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I. Preliminaries

```
#Clearing the workspace  
rm(list = ls())
```

```
options(repos="https://cran.rstudio.com")  
install.packages("pwr")
```

```
## package 'pwr' successfully unpacked and MD5 sums checked  
##  
## The downloaded binary packages are in  
## C:\Users\tiess\AppData\Local\Temp\RtmpkJXAac\downloaded_packages
```

```
library(foreign)  
library(tidyverse)  
library(ggdag)  
library(dplyr)  
library(tinytex)  
library(jtools)  
library(huxtable)  
library(summarytools)  
library(ggstance)  
library(pwr)  
library(knitr)  
library(lemon)  
library(AER)  
library(lubridate)  
library(ggplot2)  
library(plm)  
library(pwr)
```

II. What to submit

```
theUrl_ca4_ectrics2 <- "https://surfdrive.surf.nl/files/index.php/s/uWdWgE18hCYE4LS/download"  
experiment <- read.dta (file = theUrl_ca4_ectrics2)
```

(a)

```
reg1 <- plm(violations ~ treatment_dummy_2, data=experiment, effect="twoways",
            model = "within", index=c("month"))
coeftest(reg1, vcov=vcovHC(reg1, cluster="group"))
```

```
##
## t test of coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## treatment_dummy_2  -1.2333      1.1326 -1.0889   0.2898
```

#Not 100% confident this is the right way to test this, would like input on that.

(b)

My results suggest a 1.23 point decrease in violations after the treatment period. Nonetheless, a difference in the outcome ranging from a 0.1007 point decrease to a 2.3659 point decrease is also reasonably compatible with our data, given our assumptions.

(c)

Treatment effect = -1.2333

Standard deviation is the standard error times square root of n

```
SD = 1.1326*sqrt(42)
print(SD)
```

```
## [1] 7.340087
```

So the effect size is

```
-1.2333/SD
```

```
## [1] -0.1680225
```

Chance treatment works = 0.25 Significance level = 0.05

```
pwr.t.test(n=42,d=-0.1680225,sig.level=0.05,power = NULL,
            type = c("two.sample"),alternative="two.sided")
```

```
##
##      Two-sample t test power calculation
##
##              n = 42
##              d = 0.1680225
##      sig.level = 0.05
##              power = 0.1185235
##      alternative = two.sided
##
## NOTE: n is number in *each* group
```

So the power is rather low, 0.1185235.

What we want to know: $P(\text{treatment doesn't work}|\text{test-}) = P(\text{test-}|\text{doesn't work}) * P(\text{doesn't work}) / P(\text{test-}) = \text{Power} * P(\text{doesn't work}) / P(\text{test-}) = \text{Power} * P(\text{doesn't work}) / (\text{Power} * P(\text{doesn't work}) + P(\text{test-}|\text{works}) * P(\text{works}))$

```
power = 0.1185235

print(power*0.75 / ((power * 0.75) + (0.05 * 0.25)))
```

```
## [1] 0.8767169
```

So the chance that treatment doesn't work when the test is negative is 0.8767, or 88%. Seems rather high, not sure I got the formula correct. I tried to translate it from the slide "Strength of evidence of a statistically significant finding from a well-powered study".

(d)

First we calculate the effect size:

```
SD2 = 1.2*sqrt(42)
print(1.2/SD2)
```

```
## [1] 0.1543033
```

Running the power test again:

```
pwr.t.test(n=42,d=0.1543033,sig.level=NULL,power = 0.4,
           type = c("two.sample"),alternative="two.sided")
```

```
##
##      Two-sample t test power calculation
##
##              n = 42
##              d = 0.1543033
##      sig.level = 0.2891634
##              power = 0.4
##      alternative = two.sided
##
## NOTE: n is number in *each* group
```

The significance level she must have found is very close to the one we found under (a), namely 0.2898. So her finding seems to be consistent with ours.

(e)

$1 - (\text{answer found for c}) = 1 - 0.8767169$

```
newprior = 1-0.8767169
print(newprior)
```

```
## [1] 0.1232831
```

P(works) is 0.25 as given earlier.

As from the slide “Strength of evidence of a statistically significant finding from a well-powered study”, we have

$\text{Power} * P(\text{works}) / (\text{Power} * P(\text{works}) + P(\text{test+|doesn't work}) * P(\text{doesn't work}))$

```
print(power*0.25/(power*0.25 + 0.05*0.75))
```

```
## [1] 0.4413897
```

So now there's a 44% chance that the treatment works if there is a positive test. Not sure what the new test under (d) has to do with this?

(f)

With a bias of 0.2, 20% of the negative tests goes into the positive test category (increasing false positives and reducing true negatives). Not sure how to calculate this?