

BA 830 Project

Team 10

4/26/2021

Team members: Paula Demacker, Yuchen Feng, Leah Fowlkes, Ryan Veytsman, Xinpeng Yu, Yuzhe Zheng

```
## -- Attaching packages ----- tidyverse 1.3.0 --

## v ggplot2 3.3.3      v purrr  0.3.4
## v tibble  3.0.5      v dplyr  1.0.4
## v tidyr   1.1.2      v stringr 1.4.0
## v readr   1.4.0      v forcats 0.5.1

## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::between()   masks data.table::between()
## x dplyr::filter()    masks stats::filter()
## x dplyr::first()     masks data.table::first()
## x dplyr::lag()       masks stats::lag()
## x dplyr::last()      masks data.table::last()
## x purrr::transpose() masks data.table::transpose()

## Loading required package: Matrix

##
## Attaching package: 'Matrix'

## The following objects are masked from 'package:tidyr':
##
##   expand, pack, unpack
##
## Attaching package: 'lubridate'

## The following objects are masked from 'package:data.table':
##
##   hour, isoweek, mday, minute, month, quarter, second, wday, week,
##   yday, year

## The following objects are masked from 'package:base':
##
##   date, intersect, setdiff, union
##
## Please cite as:
## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary Statistics Tables.
## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer
#import the file
food <- fread('BA830-Project.csv')
```

```
food %>% mutate (outcome_plate = recode (outcome_plate, '0' = 1, '1' = 0))
```

```
##      Person_id  Age      Currently living Gender outcome_plate
##    1:         1 46-55      Europe Woman          0
##    2:         1 46-55      Europe Woman          0
##    3:         1 46-55      Europe Woman          0
##    4:         1 46-55      Europe Woman          1
##    5:         1 46-55      Europe Woman          0
## ---
## 1766:        178 16-25 North America/Central America  Man          0
## 1767:        178 16-25 North America/Central America  Man          0
## 1768:        178 16-25 North America/Central America  Man          1
## 1769:        178 16-25 North America/Central America  Man          1
## 1770:        178 16-25 North America/Central America  Man          1
##      outcome_food outcome_question      food_name Treatment
##    1:         1         1      Burger          0
##    2:         1         1         Taco          0
##    3:         0         0 Fried Chicken          0
##    4:         0         0      Salmon          0
##    5:         1         1         Pasta          0
## ---
## 1766:         1         1         Steak          1
## 1767:         1         1      Dumpling          1
## 1768:         1         1         Sushi          1
## 1769:         1         1    Fried Rice          1
## 1770:         0         0         Salad          1
```

```
#proportion test
prop.test(nrow(food[Treatment == 1]), nrow(food), 0.5)
```

```
##
## 1-sample proportions test with continuity correction
##
## data:  nrow(food[Treatment == 1]) out of nrow(food), null probability 0.5
## X-squared = 13, df = 1, p-value = 0.0004
## alternative hypothesis: true p is not equal to 0.5
## 95 percent confidence interval:
##  0.434 0.481
## sample estimates:
##      p
## 0.458
```

The p value here is lower than 0.05, so we reject the null hypothesis which the randomization was not properly done, so it may cause some problem with the experiment result.

```
reg.plate <- lm(outcome_plate ~ Treatment, data = food)
summary(reg.plate)
```

```
##
## Call:
## lm(formula = outcome_plate ~ Treatment, data = food)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.705 -0.531  0.295  0.469  0.469
##
```

```
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.5313     0.0155   34.28 < 2e-16 ***
## Treatment    0.1737     0.0229    7.58 5.4e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.48 on 1768 degrees of freedom
## Multiple R-squared:  0.0315, Adjusted R-squared:  0.0309
## F-statistic: 57.5 on 1 and 1768 DF, p-value: 5.44e-14

reg.question <- lm(outcome_question ~ Treatment, data = food)
summary(reg.question)

##
## Call:
## lm(formula = outcome_question ~ Treatment, data = food)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.486 -0.486 -0.404  0.513  0.596
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.4865     0.0160   30.4 < 2e-16 ***
## Treatment   -0.0828     0.0237   -3.5 0.00048 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.496 on 1768 degrees of freedom
## Multiple R-squared:  0.00687, Adjusted R-squared:  0.00631
## F-statistic: 12.2 on 1 and 1768 DF, p-value: 0.000481
```

Regression

```
together <- feols(c(outcome_plate, outcome_food) ~ Treatment,
                  data = food, se = 'white')
etable(together)
```

	model 1	model 2
Dependent Var.:	outcome_plate	outcome_food
(Intercept)	0.5312*** (0.0161)	0.4865*** (0.0161)
Treatment	0.1737*** (0.0227)	-0.0828*** (0.0236)
S.E. type	Heteroskedas.-rob.	Heteroskedast.-rob.
Observations	1,770	1,770
R2	0.03150	0.00687
Adj. R2	0.03095	0.00631

In the regression of model 1 we looked at the outcome_plate as the outcome variable first. We get an intercept of 0.5312, which represents the mean of the outcome of our control group. The estimate for the treatment effect is 0.1737 and has a standard error of 0.0227. Therefore, we have a positive treatment effect which indicates a positive relationship between Treatment and outcome_plate.

We know from the stars that this result is statistically significant. This means that people in the treatment group are more likely to choose the red plate.

For model 2 we used `outcome_food` as our outcome variable. The intercept is 0.4865. Our estimated treatment effect is -0.0828 and has a standard error of 0.0236. This indicates that we have a negative relationship between Treatment and `outcome_food`. These results are also statistically significant. It still indicates that people in the treatment group are more likely to choose the food on the red plate. The negative treatment effect shows us that people in the treatment group are more likely to choose the plate that is positioned on the right.

```
fixed_effects <- feols(outcome_plate ~ Treatment
                       | outcome_food + food_name, data = food,
                       se = 'white')
etable(fixed_effects)
```

```
##                               fixed_effects
## Dependent Var.:               outcome_plate
##
## Treatment                    0.1788*** (0.0230)
## Fixed-Effects:  -----
## outcome_food                  Yes
## food_name                     Yes
## -----
## S.E. type                     Heteroskedas.-rob.
## Observations                  1,770
## R2                            0.04394
## Within R2                     0.03349
```

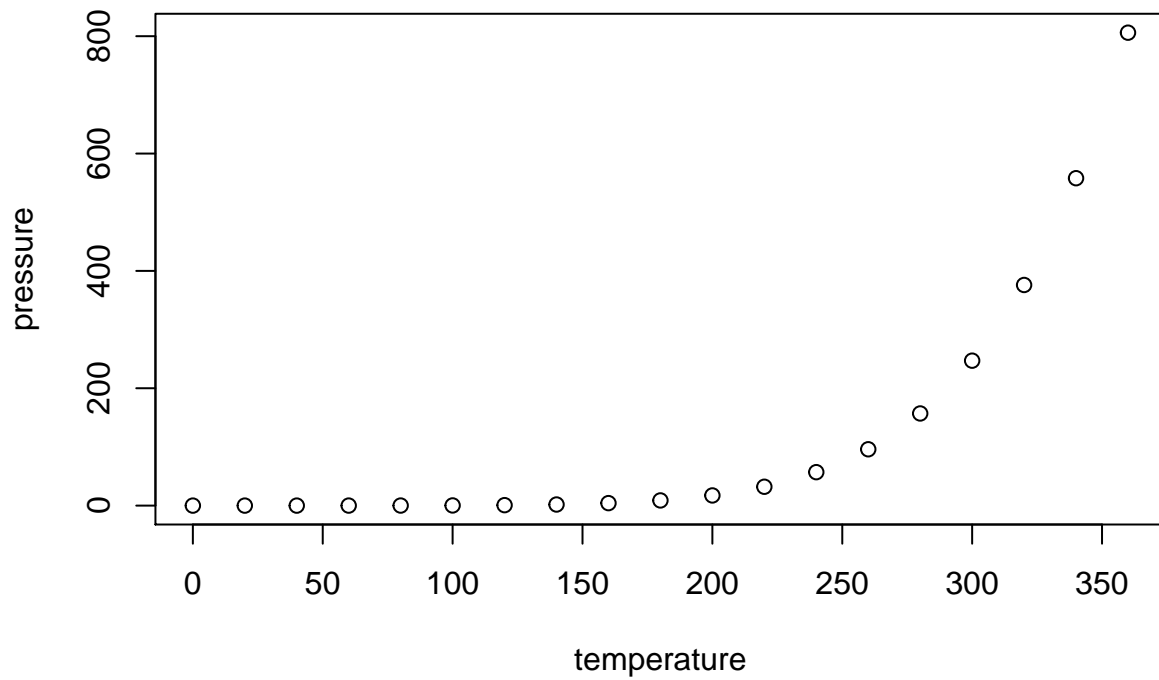
For the regression above we added two fixed effects: `outcome_food` and `food_name` to see if the food has an impact on the outcome. The outcome variable is `outcome_plate` and has an estimated treatment effect of 0.1788 with an standard error of 0.0230, which is really similar to the regression above without the fixed effects. The results of this regression are also statistically significant.

```
fixed_effects2 <- feols(outcome_plate ~ Treatment
                        | outcome_food + food_name + Gender + Age, data = food,
                        se = 'white')
etable(fixed_effects)
```

```
##                               fixed_effects
## Dependent Var.:               outcome_plate
##
## Treatment                    0.1788*** (0.0230)
## Fixed-Effects:  -----
## outcome_food                  Yes
## food_name                     Yes
## -----
## S.E. type                     Heteroskedas.-rob.
## Observations                  1,770
## R2                            0.04394
## Within R2                     0.03349
```

Including Plots

You can also embed plots, for example:



Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.