
title: "Exam One(Redo for HW)" author: "Collin Rafferty" date: "10/22/2021" output: pdf_document

Question One:

$$H_0: P_A - P_B = 0$$

$$H_A: P_A - P_B \neq 0$$

Where P_A is the proportion of people with at least some college in the Northeastern and Western regions of the U.S. that have been vaccinated, P_B is the proportion of people with a high school diploma in the Midwestern and Southern regions of the U.S. that have been vaccinated.

$$P_A \text{ Estimate} = \frac{23,363}{25,907} = 90.18\% \quad P_B \text{ Estimate} = \frac{3,478}{4,678} = 74.35\%$$

$$SE = \sqrt{\frac{\widehat{P}_A(1 - \widehat{P}_A)}{n_1} + \frac{\widehat{P}_B(1 - \widehat{P}_B)}{n_2}} = \sqrt{\frac{.9018(1 - .9018)}{25,907} + \frac{.7435(1 - .7435)}{4,678}} = .009$$

$$t - \text{stat} = \frac{\bar{X}}{SE} = \frac{.9018 - .7435}{.009} = 17.58889$$

I chose to use a .05 significance level. This means the results are significant at $p < .05$, and the p-value is $< .00001$, so the results are highly statistically significant. This means we succeeded in rejecting the null and showing sufficient evidence there is a statistically significant difference between the vaccination rates of the two samples.

$$\text{Confidence Interval: } (.9018 - .7435 \pm .009) = (.1493, .1676)$$

We are 95% confident that the average difference between the vaccination rate of sample A and sample B is between 14.93% and 16.73%

Question Two:

$$H_0: P_A - P_B = 0$$

$$H_A: P_A - P_B \neq 0$$

P_A is the proportion of males with an associate degree or less who have been vaccinated, and P_B is the proportion of people who identify as transgender or other and have a bachelor's degree or great and have been vaccinated.

$$P_A \text{ Estimate} = \frac{8,486}{10,290} = 82.45\% \quad P_B \text{ Estimate} = \frac{313}{389} = 80.46\%$$

$$SE = \sqrt{\frac{\widehat{P}_A(1 - \widehat{P}_A)}{n_1} + \frac{\widehat{P}_B(1 - \widehat{P}_B)}{n_2}} = \sqrt{\frac{.8245(1 - .8245)}{10,290} + \frac{.8046(1 - .8046)}{389}} = .0201$$

$$t - stat = \frac{\bar{X}}{SE} = \frac{.8245 - .8046}{.0201} = .99$$

I chose to use a .10 significance level. This means the results are significant at $p < .10$, and the p-value is .322197, so the results are not statistically significant. This means we failed to reject the null hypothesis, which means there was not sufficient evidence to conclude the vaccination rate between these two groups was statistically significant.

Confidence Interval: $(.8245 - .8046 \pm .0210) = (-.0011, .0409)$

We are 95% confident that the average difference between the vaccination rate of sample A and sample B is between -.11% and 4.09%

These were not expected results. I thought because of their educational attainment levels, sample A would have a much higher vaccination rate than sample B. One explanation for this disparity may be the difference in the sizes of the samples. Sample A is only about 4% the size of sample B

Question Three:

A)

```
load("/cloud/project/Household_Pulse_data.RData")

#Creation of Subgroup
attach(Household_Pulse_data)
use_varb <- ((RRACE=="White") |(RRACE=="Black")) & (INCOME== "HH income $75 - 99.9") & (RHISPANIC== "Not Hispanic") & ((EEDUC=="adv deg") | (EEDUC=="bach d eg"))
sub1 <- subset(Household_Pulse_data,use_varb)
detach(Household_Pulse_data)
attach(sub1)

# Summary Statistics
summary(sub1)
```

##	RHISPANIC	RRACE	EEDUC	MS
##	Not Hispanic:5996	White:5684	less than hs: 0	NA : 20
##	Hispanic : 0	Black: 312	some hs : 0	married :4525
##		Asian: 0	HS diploma : 0	widowed : 204
##		Other: 0	some coll : 0	divorced : 502
##			assoc deg : 0	separated: 52
##			bach deg :3005	never : 693
##			adv deg :2991	
##	EGENID_BIRTH	GENID_DESCRIBE	SEXUAL_ORIENTATION	
##	male :2560	NA : 32	NA : 49	
##	female:3436	male :2525	gay or lesbian: 206	
##		female :3394	straight :5501	
##		transgender: 6	bisexual : 158	
##		other : 39	something else: 46	
##			dont know : 36	

```

##
##          KIDS_LT5Y          KIDS_5_11Y
## NA          :5268  NA          :4984
## Yes children under 5 in HH: 728  Yes children 5 - 11 in HH:1012
##
##
##
##
##          KIDS_12_17Y          ENROLLNONE
## NA          :5074  NA          :5501
## Yes children 12 - 17 in HH: 922  children not in any type of school: 495
##
##
##
##          RECVDVACC          DOSESERV
## NA          : 13  NA          : 330
## yes got vaxx      :5682  yes got all doses      :5570
## no did not get vaxx: 301  yes plan to get all doses: 83
##                  no will not get all doses: 13
##
##
##
##          GETVACRV          KIDDOSES
## NA          :5687  NA          :5078
## definitely will get vaxx: 15  Yes kids got or will get all doses: 701
## probably will get vaxx : 23  no kids did not or will not      : 217
## unsure about vaxx      : 67
## probably not           : 82
## definitely not         : 122
##
##          KIDGETVAC          HADCOVID
## NA          :5776  NA          : 17
## definitely will get vaxx: 31  yes doctor told had covid: 670
## probably will get vaxx : 28  no did not          :5288
## unsure about vaxx      : 46  not sure           : 21
## probably not           : 44
## definitely not         : 58
## dont know yet         : 13
##
##          WRKLOSSRV          ANYWORK
## NA          : 10  NA          : 9
## yes recent HH job loss: 349  yes employment in last 7 days:4113
## no recent HH job loss :5637  no employment in last 7 days :1874
##
##
##
##          KINDWORK          RSNNOWRKRV

```

```

## NA :1916 NA :4144
## work for govt : 943 retired :1378
## work for private co:1937 other : 176
## work for nonprofit : 734 caring for kids: 103
## self employed : 409 did not want : 78
## work in family biz : 57 laid off : 51
## (Other) : 66
## CHLDCARE
## NA :4806
## yes impacts to childcare because pandemic: 287
## no : 903
##
##
##
## CURFOODSUF
## NA : 13
## had enough food :5615
## had enough but not what wanted: 344
## sometimes not enough food : 22
## often not enough food : 2
##
##
## CHILDFOOD
## NA :5814
## often kids not eating enough because couldnt afford: 3
## sometimes kids not eating enough : 22
## kids got enough food : 157
##
##
## ANXIOUS
## NA : 8
## no anxiety over past 2 wks :2721
## several days anxiety over past 2 wks :2088
## more than half the days anxiety over past 2 wks: 543
## nearly every day anxiety : 636
##
##
## WORRY
## NA : 9
## no worry over past 2 wks :3501
## several days worried over past 2 wks :1711
## more than half the days worried over past 2 wks: 388
## nearly every day worry : 387
##
##
## TENURE
## NA : 11
## housing owned free and clear :1752

```

```
## housing owned with mortgage :3548
## housing rented : 664
## housing occupied without rent: 21
##
##
## LIVQTRRV RENTCUR
## live in detached 1 family :4875 NA :5332
## live in bldg w 5+ apts : 470 current on rent: 654
## live in 1 family attached to others: 444 behind on rent : 10
## live in building with 3-4 apts : 80
## live in bldg w 2 apartments : 68
## live in mobile home : 30
## (Other) : 29
## MORTCUR EVICT
## NA :2449 NA :598
6
## current on mortgage:3475 very likely evicted in next 2 months :
1
## behind on mortgage : 72 somewhat likely evicted in next 2 months :
0
## not very likely evicted in next 2 months :
4
## not at all likely evicted in next 2 months:
5
##
##
## FORCLOSE EST_ST
## NA :5924 California: 337
## very likely foreclosed in next 2 months : 5 Texas : 272
## somewhat likely foreclosed in next 2 months : 8 Washington: 233
## not very likely foreclosed in next 2 months : 17 Florida : 201
## not at all foreclosed evicted in next 2 months: 42 Maryland : 201
## Utah : 196
## (Other) :4556
## PRIVHLTH PUBHLTH REGION
## has private health ins:5420 has public health ins:1779 Northeast:1013
## no private health ins : 459 no public health ins :3852 South :1899
## NA : 117 NA : 365 Midwest :1305
## West :1779
##
##
## INCOME Num_kids_Pub_School Num_kids_Priv_School
## HH income $75 - 99.9 :5996 Min. :0.000 Min. :0.000
## NA : 0 1st Qu.:1.000 1st Qu.:0.000
## HH income less than $25k: 0 Median :2.000 Median :1.000
## HH income $25k - $34.9k : 0 Mean :1.748 Mean :1.011
## HH income $35k - 49.9 : 0 3rd Qu.:2.000 3rd Qu.:2.000
## HH income $50k - 74.9 : 0 Max. :4.000 Max. :2.000
## (Other) : 0 NA's :4685 NA's :5727
```

```
## Num_kids_homeschool      Works_onsite      works_remote
## Min. :0.00      NA      : 94      NA      : 182
## 1st Qu.:0.00      worked onsite:3579      worked remotely:3109
## Median :1.00      no      :2323      no      :2705
## Mean :0.72
## 3rd Qu.:1.00
## Max. :2.00
## NA's :5839
## Shop_in_store      eat_in_restaurant
## NA : 86      NA : 104
## shopped in store:5485      eat at restaurant indoors:3394
## no : 425      no :2498
##
##
##
##
```

```
round(prop.table(table(RRACE=="White", ANXIOUS)),2)
```

```
## ANXIOUS
## NA no anxiety over past 2 wks several days anxiety over past 2 w
ks
## FALSE 0.00      0.03      0.
02
## TRUE 0.00      0.43      0.
33
## ANXIOUS
## more than half the days anxiety over past 2 wks
## FALSE 0.00
## TRUE 0.09
## ANXIOUS
## nearly every day anxiety
## FALSE 0.00
## TRUE 0.10
```

```
round(prop.table(table(RRACE=="Black", ANXIOUS)),2)
```

```
## ANXIOUS
## NA no anxiety over past 2 wks several days anxiety over past 2 w
ks
## FALSE 0.00      0.43      0.
33
## TRUE 0.00      0.03      0.
02
## ANXIOUS
## more than half the days anxiety over past 2 wks
## FALSE 0.09
## TRUE 0.00
## ANXIOUS
## nearly every day anxiety
```

##	FALSE	0.10
##	TRUE	0.00

The subgroup I created is revealing because it looks at how upper-middle-class college-educated whites have coped with the pandemic compared to upper-middle-class college-educated African Americans. If you look at the proportion tables of how the anxious levels differ between the two groups, it is very eye-opening. For example, 43% of African Americans in this subgroup reported feeling no anxiety over the past two weeks compared to 43% of whites that reported feeling anxiety over the past two weeks. In the very next factor level, however, these two groups' anxiety levels flip. Much of the pandemic pernicious effects disproportionately affected the poor and underrepresented, but it is also important to remember that it has had a profound impact on every level of society. I think my subgroup helps to show that.

B) I thought it would be interesting to look how the percentage of people with private health insurance differs between the subgroup, and the larger sample.

$$H_0: P_A = P_B$$

$$H_A: P_A \neq P_B$$

Where P_A is the proportion of people in the subgroup who have private health insurance, and P_B is the proportion of people in the greater sample that has private health insurance.

```
summary(sub1$PRIVHLTH)
```

##	has private health ins	no private health ins	NA
##	5420	459	117

```
summary(Household_Pulse_data$PRIVHLTH)
```

##	has private health ins	no private health ins	NA
##	46869	11275	10970

$$P_A \text{ Estimate} = \frac{5,420}{5,996} = 90.39\% \quad P_B \text{ Estimate} = \frac{46,869}{69,114} = 67.81\%$$

$$SE = \sqrt{\frac{\widehat{P}_A(1 - \widehat{P}_A)}{n_1} + \frac{\widehat{P}_B(1 - \widehat{P}_B)}{n_2}} = \sqrt{\frac{.9039(1 - .9039)}{5,996} + \frac{.6781(1 - .6781)}{69,114}} = .0042$$

$$t - stat = \frac{\bar{X}}{SE} = \frac{.9039 - .6781}{.0042} = 53.76$$

I chose to use a .05 significance level. This means the results are significant at $p < .05$, and the p-value is $< .00001$, so the results are highly statistically significant. This means we succeeded in rejecting the null and showing sufficient evidence there is a statistically significant difference between the percentage of people with private health insurance in the subgroup compared to the larger sample.

$$\text{Confidence Interval: } (.9036 - .6781 \pm .0042) = (.2216, .23)$$

We are 95% confident that the average difference between the people with private health insurance in the subgroup and the larger sample is between 22.16% and 23%.

C)

```
##require(tidyverse)

HH1 <-
Household_Pulse_data%>%mutate(INCOME5=as.numeric(INCOME),INCOME5=case_when(IN
COME5==5~NA_integer_,TRUE~as.###.integer(INCOME5)))

HH2 <-
Household_Pulse_data%>%mutate(GEN1=as.numeric(GENID_DESCRIBE),GEN1=case_when(
GEN1==1~NA_integer_,TRUE~as. integer(GEN1)))

HH3 <- Household_Pulse_data%>%mutate(VAX1=as.numeric(RECVDVACC),
##VAX1=case_when(VAX1==1~NA_integer_,TRUE~as.integer(VAX1)))

HH4 <- Household_Pulse_data%>%mutate(ANX2=as.numeric(ANXIOUS),
ANX2=case_when(ANX2==2~NA_integer_,TRUE~as.integer(ANX2)))

H5 <- Household_Pulse_data%>%mutate(EEDUC3=as.numeric(EEDUC),
##EEDUC3=case_when(EEDUC3==3~NA_integer_,TRUE~as.integer(EEDUC3)))

norm_varb <- function(X_in) {(X_in - min(X_in, na.rm = TRUE))/( max(X_in,
##na.rm = TRUE) - min(X_in, na.rm = TRUE) )}

##norm_INCOME5 <- norm_varb(HH1$INCOME5)

##norm_GEN1 <- norm_varb(HH2$GEN1)

##norm_VAX1<- norm_varb(HH3$VAX1)

##norm_ANX2 <- norm_varb(HH4$ANX2)

##norm_EEDUC3<- norm_varb(HH5$EEDUC3)

##data_use <- data.frame(norm_INCOME5,norm_GEN1,norm_VAX1,norm_ANX2,
##norm_EEDUC3)

##good_obs_data_use <- complete.cases(data_use,PUBHLTH)

##dat_use <- subset(data_use,good_obs_data_use)

##y_use <- subset(PUBHLTH,good_obs_data_use)
```



```
##set.seed(12345)
##NN_obs <- sum(good_obs_data_use == 1)
##select1 <- (runif(NN_obs) < 0.8)

##train_data <- subset(dat_use,select1)
##test_data <- subset(dat_use,!select1)
##cl_data <- y_use[select1]
##true_data <- y_use[!select1]
```

```
##summary(cl_data)
has public health ins no public health ins NA
7146 12890 6319
prop.table(summary(cl_data))
has public health ins no public health ins NA
0.2711440 0.4890913 0.2397648
summary(train_data)
norm_INCOME5 norm_GEN1 norm_VAX1 norm_ANX2
norm_EEDUC3
Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.0000
1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.5000 1st Qu.:0.5000
Median :0.2857 Median :0.3333 Median :0.0000 Median :0.5000 Median :0.8333
Mean :0.3629 Mean :0.2307 Mean :0.1101 Mean :0.5475 Mean :0.7538
3rd Qu.:0.7143 3rd Qu.:0.3333 3rd Qu.:0.0000 3rd Qu.:0.7500 3rd Qu.:1.0000
Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.0000
```

```
require(class)

for (indx in seq(1, 9, by= 2)) {

pred_PUBHLTH <- knn(train_data, test_data, cl_data, k = indx, l = 0, prob =
FALSE, use.all = TRUE)}


```

```
num_correct_labels <- sum(pred_PUBHLTH == true_data)

correct_rate <- num_correct_labels/length(true_data)

print(c(indx,correct_rate))

> print(c(indx,correct_rate))

[1] 9.0000000 0.7019365


```

With the predictor variables I included in the classifier, it had an accuracy of 70% of determining if a person has public health insurance or not. 70% is not terrible, but it could be improved. If I were to do another iteration, I would include the variables of ANYWORK,RSNNOWRKRV, and KINDWORK because if a person is a government employee, unemployed, or retired, they would be more likely to have public health insurance.

D)

```
attach(Household_Pulse_data)

## The following objects are masked from sub1:
##
## ANXIOUS, ANYWORK, CHILDFOOD, CHLDCARE, CURFOODSUF, DOSESERV,
## eat_in_restaurant, EEDUC, EGENID_BIRTH, ENROLLNONE, EST_ST, EVICT,
## FORCLOSE, GENID_DESCRIBE, GETVACRV, HADCOVID, INCOME, KIDDOSES,
## KIDGETVAC, KIDS_12_17Y, KIDS_5_11Y, KIDS_LT5Y, KINDWORK, LIVQTRRV,
## MORTCUR, MS, Num_kids_homeschool, Num_kids_Priv_School,
## Num_kids_Pub_School, PRIVHLTH, PUBHLTH, RECDVACC, REGION, RENTCUR,
## RHISPANIC, RRACE, RSNNOWRKRV, SEXUAL_ORIENTATION, Shop_in_store,
## TENURE, Works_onsite, works_remote, WORRY, WRKLOSSRV

reg1<- lm(as.numeric(Num_kids_Pub_School) ~ INCOME+EEDUC+CHLDCARE+Works_onsite)

require(stargazer)

## Loading required package: stargazer

##
## 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
```

```
stargazer(reg1,type = "text", title = "Table 1: Regression Results",out = "table1.txt")
```

```
##
```

```
## Table 1: Regression Results
```

```
## =====  
=====
```

```
##                                     Dependent variable
```

```
:
```

```
##                                     -----
```

```
##                                     as.numeric(Num_kids_Pub_School)
```

```
## -----  
-----
```

```
## 25k                                     -0.087**  
##                                     (0.037)
```

```
##                                     -0.028  
##                                     (0.032)
```

```
## 35k - 49.9                             -0.048*  
##                                     (0.028)
```

```
## 50k - 74.9                             -0.031  
##                                     (0.028)
```

```
## 75 - 99.9                              0.026  
##                                     (0.025)
```

```
## 100k - 149                             0.021  
##                                     (0.031)
```

```
## 150 - 199                              0.040  
##                                     (0.029)
```

```
## EEDUCsome hs                          -0.151  
##                                     (0.101)
```

```
## EEDUCHS diploma                       -0.291***  
##                                     (0.087)
```

```
## EEDUCsome coll                        -0.365***  
##                                     (0.086)
```

```
##
```

```

## EEDUCassoc deg -0.316***
## (0.088)
##
## EEDUCbach deg -0.403***
## (0.086)
##
## EEDUCadv deg -0.412***
## (0.087)
##
## CHLDCAREyes impacts to childcare because pandemic 0.101***
## (0.026)
##
## CHLDCAREno 0.212***
## (0.016)
##
## Works_onsiteworked onsite -0.057**
## (0.029)
##
## Works_onsiteno -0.114***
## (0.030)
##
## Constant 2.059***
## (0.087)
## -----
-----
## Observations 14,006
## R2 0.018
## Adjusted R2 0.017
## Residual Std. Error 0.878 (df = 13988)
## F Statistic 15.008*** (df = 17; 13
988)
## =====
=====
## Note: *p<0.1; **p<0.05; **
*p<0.01

require(ggplot2)

## Loading required package: ggplot2

require(ggthemes)

## Loading required package: ggthemes

Graph1 <-ggplot(Household_Pulse_data, aes(y=Num_kids_Pub_School, x= EEDUC, gr
oup=1))+geom_point()+geom_smooth(method=lm)+labs(x="Educational Attainment",y
="# of Kids in Public School", title = "Graph One for Reg1")

Graph2 <-ggplot(Household_Pulse_data, aes(y=Num_kids_Pub_School, x= INCOME, g
roup=1))+geom_point()+geom_smooth(method=lm)+labs(x="Income($1000s)", y="# of

```

```
Kids in Public School", title = "Graph Two for Reg1") + scale_x_discrete(labels = c('<25', '25-34.49', '35-49.9', '50-74.9', '75-99.9', '100-149', '150-199', '>$200' ))
```

```
gridExtra::grid.arrange(Graph1, Graph2)
```

```
## `geom_smooth()` using formula 'y ~ x'
```

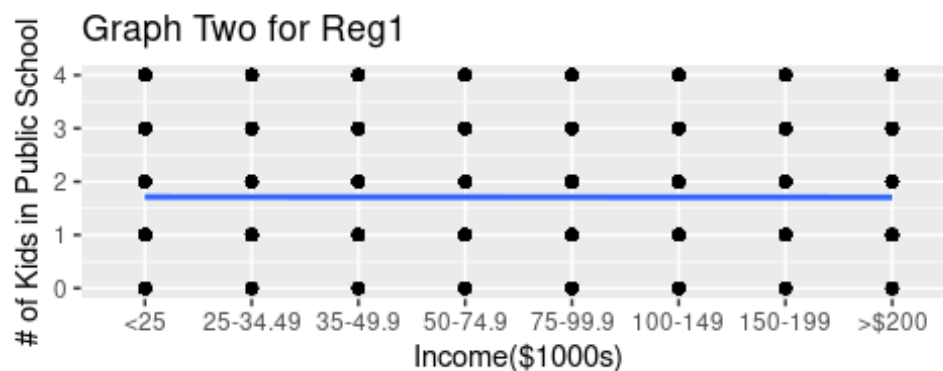
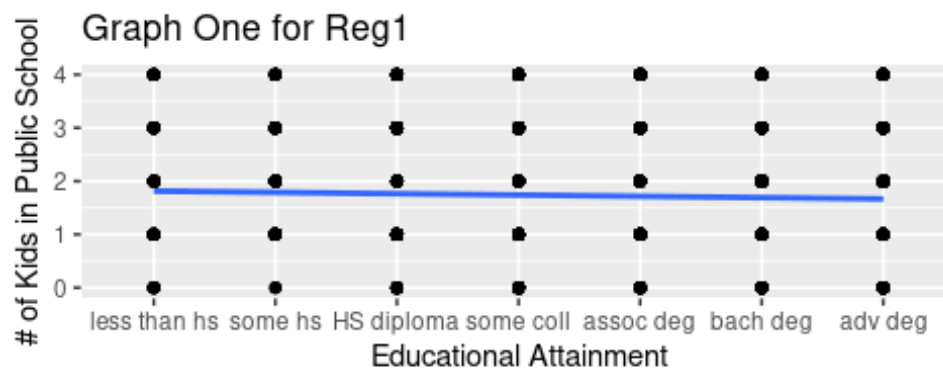
```
## Warning: Removed 55108 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 55108 rows containing missing values (geom_point).
```

```
## `geom_smooth()` using formula 'y ~ x'
```

```
## Warning: Removed 55108 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 55108 rows containing missing values (geom_point).
```



The regression provides some revealing insights about the data. For example, the number of children in public is negatively correlated with educational attainment level after a person graduates high school. A person with an advanced degree has -.412 fewer children in public school than someone who does not have an advanced degree. Of course, you can't have a proportion of a child, but it helps to show the picture. This result was to be expected, but what was interesting was how income affects the number of children a person has in public school. The coefficients of income are negative until 75K than positive for the rest of the income levels. Intuitively, you would think the opposite would be true: as income increases,

the number of children a person has in public school decreases because these high earners have fewer children or send their children to private school.