Jianwen Wu Eco B2000 Homework #4 Prof. Foster

1) Group Members: Keyi Long and Crystal Hernandez.

2)

Model 1

Explanatory Variables: \(\beta 1 = AGE_REF \)
\(\beta 2 = FAM_SIZE \)
\(\beta 3 = VEHQ \)
\(\beta 4 = VEHQL \)
\(\beta 5 = ELCTRCPQ \)
\(\beta 6 = ALLFULPQ \)
\(\beta 7 = GASMOPQ \)
\(\beta 8 = HEALTHPQ \)
\(\beta 9 = HLTHINPQ \)
\(\beta 10 = HLTHINPQ \)

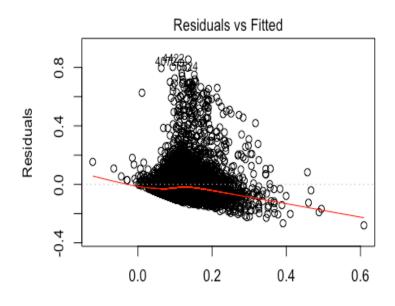
Respond Variable:

 $\beta 11 = PERSCACQ$

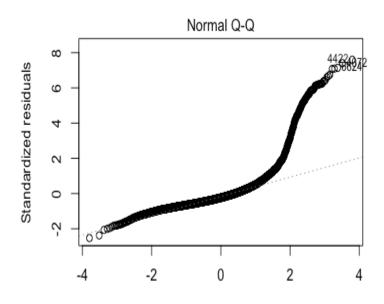
Fraction of transport spent of the total expenditure.

```
rm(list = ls(all = TRUE))
load("cex 2012.RData")
attach(data cex)
fraction_transport <- TRANSPQ/TOTEXPPQ # fraction of spent on transport of to</pre>
talexpenditure
fraction transport[is.infinite(fraction transport)] <- NA</pre>
fraction_transport[fraction_transport<0] <- NA</pre>
summary(fraction_transport)
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                Max.
                                                         NA's
## 0.00000 0.05473 0.09977 0.12840 0.16310 0.99060
                                                            1
summary(fraction_transport[as.logical(FAM_SIZE)])
                                                         NA's
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                Max.
## 0.00000 0.05473 0.09977 0.12840 0.16310 0.99060
                                                            1
FAM_SIZE[is.infinite(FAM_SIZE)] <- NA</pre>
summary(fraction transport[as.logical(VEHQ)])
```

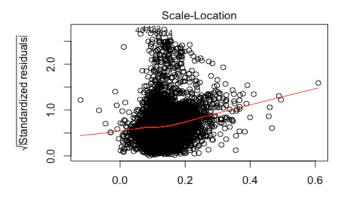
```
Min. 1st Qu. Median Mean 3rd Qu.
                                                    NA's
                                            Max.
## 0.00000 0.06631 0.10930 0.13980 0.17100 0.99060
                                                       1
#make sure we do not have negative transport spent
model1 <- lm(fraction_transport~AGE_REF + FAM_SIZE + VEHQ + VEHQL + ELCTRCPQ
+ ALLFULPQ + GASMOPQ + HEALTHPQ + HLTHINPQ + HLTHINPQ + PERSCACQ)
summary(model1)
##
## Call:
## lm(formula = fraction_transport ~ AGE_REF + FAM_SIZE + VEHQ +
      VEHQL + ELCTRCPQ + ALLFULPQ + GASMOPQ + HEALTHPQ + HLTHINPQ +
##
      HLTHINPQ + PERSCACQ)
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
## -0.28037 -0.05994 -0.02643 0.02302 0.85486
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 1.189e-01 5.532e-03 21.490 < 2e-16 ***
## AGE REF
              -5.016e-04 8.432e-05 -5.948 2.84e-09 ***
## FAM SIZE
              -5.835e-03 1.015e-03 -5.748 9.40e-09 ***
## VEHQ
               1.667e-02 1.086e-03 15.350 < 2e-16 ***
## VEHQL
               6.906e-02 6.601e-03 10.462 < 2e-16 ***
              -5.125e-05 7.121e-06 -7.197 6.82e-13 ***
## ELCTRCPQ
## ALLFULPQ
              -1.763e-05 8.945e-06 -1.971 0.0488 *
## GASMOPQ
               8.496e-05 3.620e-06 23.473 < 2e-16 ***
## HEALTHPQ
              -1.327e-05 2.425e-06 -5.474 4.55e-08 ***
              2.318e-06 3.495e-06
## HLTHINPO
                                    0.663
                                             0.5072
## PERSCACQ
              -5.401e-05 2.512e-05 -2.150
                                             0.0316 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1125 on 6826 degrees of freedom
    (1 observation deleted due to missingness)
## Multiple R-squared: 0.1658, Adjusted R-squared: 0.1646
## F-statistic: 135.7 on 10 and 6826 DF, p-value: < 2.2e-16
plot(model1)
```



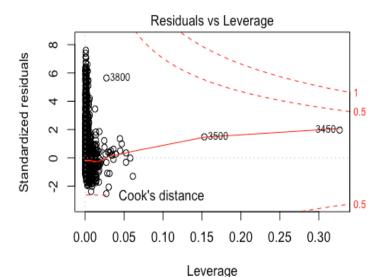
Fitted values
n_transport ~ AGE_REF + FAM_SIZE + VEHQ + VEHQL + ELCTR(



Theoretical Quantiles
n_transport ~ AGE_REF + FAM_SIZE + VEHQ + VEHQL + ELCTR(



Fitted values n_transport ~ AGE_REF + FAM_SIZE + VEHQ + VEHQL + ELCTR(



n_transport ~ AGE_REF + FAM_SIZE + VEHQ + VEHQL + ELCTR(

vcov(model1)# make sure that each explanatory variables are not strongly rela
teive.

```
##
                 (Intercept)
                                   AGE REF
                                                 FAM SIZE
                                                                   VEHQ
## (Intercept)
                3.060075e-05 -3.906128e-07 -2.951579e-06 -4.793648e-07
## AGE_REF
               -3.906128e-07
                              7.110117e-09
                                            2.454770e-08 -7.133136e-09
## FAM SIZE
               -2.951579e-06
                              2.454770e-08
                                            1.030214e-06 -1.980058e-07
## VEHQ
               -4.793648e-07 -7.133136e-09 -1.980058e-07
                                                           1.179018e-06
               -9.357687e-07 -2.304570e-10 -1.585383e-07
## VEHQL
                                                           9.126504e-07
## ELCTRCPQ
               -4.935543e-10 -7.192728e-11 -1.165085e-09 -4.061106e-10
## ALLFULPQ
                2.414568e-09 -4.954670e-11 -1.224377e-11 -1.859742e-10
               -2.750756e-09 4.530921e-11 -4.366865e-10 -1.224637e-09
## GASMOPQ
## HEALTHPQ
                2.093912e-10 -9.727807e-12 5.027021e-11 -6.196866e-11
## HLTHINPQ
                1.059637e-09 -2.615571e-11 -5.701194e-11 -1.295076e-10
## PERSCACQ
               -6.395250e-09 -8.477191e-11 -1.957706e-09 -4.092780e-09
```

```
##
                                  ELCTRCPQ
                       VEHQL
                                                ALLFULPQ
                                                               GASMOPQ
## (Intercept) -9.357687e-07 -4.935543e-10
                                            2.414568e-09 -2.750756e-09
## AGE REF
               -2.304570e-10 -7.192728e-11 -4.954670e-11 4.530921e-11
## FAM SIZE
               -1.585383e-07 -1.165085e-09 -1.224377e-11 -4.366865e-10
## VEHQ
                9.126504e-07 -4.061106e-10 -1.859742e-10 -1.224637e-09
## VEHQL
                4.356988e-05 -8.565519e-10 -9.704407e-10 -1.943148e-09
## ELCTRCPQ
               -8.565519e-10 5.070959e-11 -5.289913e-12 -7.386659e-12
## ALLFULPO
               -9.704407e-10 -5.289913e-12 8.000748e-11 -7.554025e-13
## GASMOPQ
               -1.943148e-09 -7.386659e-12 -7.554025e-13 1.310139e-11
## HEALTHPO
               -1.758725e-10 -1.473246e-12 -6.404387e-13 -6.194293e-13
## HLTHINPQ
               -1.072078e-09 -2.203213e-13
                                            5.770182e-14 -4.717968e-16
## PERSCACQ
               -9.951979e-09 1.978195e-11 -7.456583e-12
                                                          7.151723e-12
##
                    HEALTHPQ
                                  HLTHINPQ
                                                PERSCACQ
## (Intercept)
               2.093912e-10 1.059637e-09 -6.395250e-09
## AGE_REF
               -9.727807e-12 -2.615571e-11 -8.477191e-11
## FAM_SIZE
                5.027021e-11 -5.701194e-11 -1.957706e-09
## VEHQ
               -6.196866e-11 -1.295076e-10 -4.092780e-09
## VEHQL
               -1.758725e-10 -1.072078e-09 -9.951979e-09
## ELCTRCPQ
               -1.473246e-12 -2.203213e-13 1.978195e-11
## ALLFULPQ
               -6.404387e-13 5.770182e-14 -7.456583e-12
## GASMOPQ
               -6.194293e-13 -4.717968e-16 7.151723e-12
## HEALTHPQ
                5.878944e-12 -6.503634e-12 -1.198854e-12
## HLTHINPQ
               -6.503634e-12 1.221193e-11 2.584072e-12
## PERSCACQ
               -1.198854e-12 2.584072e-12 6.312047e-10
```

Explanation(model1):

In this linear model equation, we set the alpha to equal 0.05. In order for the linear regressions to work, we need to determine whether the linear regression model is statistical significant. To do this we must use the null and alternative hypothesis testing for ANOVA and individual coefficient hypothesis test.

Hypothesis test for ANOVA:

```
Null: \beta 1 = \beta 2 = \cdots = B11
```

Alternative: At least one coefficient does not equal to 0

If the all of coefficients are equal to 0, than we can infer that there is no significant linear relationship between the variables, and is therefore not a good candidate for linear regression.

F-statistic: 135.7 on 10 and 6823 DF, p-value: < 2.2e-16

The results indicate that we should reject the null hypothesis and that β 1 through β 11 have no effect on fraction of transport spent. Also, since the P-value < 0.05, and therefore the linear model is statistical significant, with a 5% level of significance.

Hypothesis for individual coefficient:

```
Null: \beta k = 0
Alternative: \beta k \neq 0
```

```
Where k = 1, 2, 3, \dots, 11
```

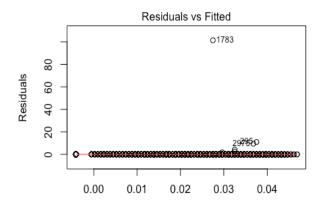
Bases on the result from R, ALLFULPQ(\(\beta\)6) and HLTHINPQ(\(\beta\)10) have T statistic of -1.971 & 0.663, and P-value of 0.0488 & 0.5072. Since the P-value is greater than 0.05, we fail to reject the null hypothesis. Therefore, these two variables are not statistically significant and not useful in predicting the fraction of transport spent of the total expenditure.

The coefficient of determination (R^2) is 0.1658, and it indicates that there is weak linear relationship between the explanatory variables (β 1 to β 11) and the respond variable (fraction of transport spent). The explanatory variables (β 1 to β 11) explain approximate 16.58% of variation in fraction of transport spent of total expenditure, but a much larger 83.42% remain unexplained.

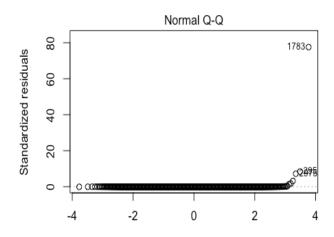
Model 2

```
fraction_babysitting<- BBYDAYPQ/FINCATAX</pre>
fraction_babysitting1<- is.finite(fraction_babysitting)</pre>
model2 <- lm(fraction babysitting[fraction babysitting1] ~ AGE REF[fraction b
abysitting1])
summary(model2)
##
## Call:
## lm(formula = fraction babysitting[fraction babysitting1] ~ AGE REF[fractio
n_babysitting1])
##
## Residuals:
##
      Min
                10 Median
                                30
                                       Max
   -0.125 -0.032 -0.022 -0.012 101.373
##
##
## Coefficients:
##
                                    Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                   0.0583520 0.0502176
                                                          1.162
                                                                    0.245
## AGE_REF[fraction_babysitting1] -0.0007181 0.0009407 -0.763
                                                                    0.445
## Residual standard error: 1.307 on 6158 degrees of freedom
## Multiple R-squared: 9.462e-05, Adjusted R-squared:
## F-statistic: 0.5827 on 1 and 6158 DF, p-value: 0.4453
anova(model2)
## Analysis of Variance Table
##
```

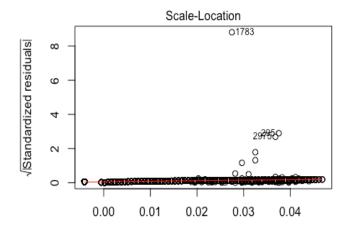
```
## Response: fraction_babysitting[fraction_babysitting1]
##
                                    Df Sum Sq Mean Sq F value Pr(>F)
## AGE_REF[fraction_babysitting1]
                                            1 0.99502 0.5827 0.4453
                                     1
## Residuals
                                  6158 10515 1.70758
coefficients(model2)
##
                      (Intercept) AGE_REF[fraction_babysitting1]
##
                     0.0583519688
                                                    -0.0007180643
vcov(model2)
##
                                    (Intercept)
## (Intercept)
                                   2.521807e-03
## AGE_REF[fraction_babysitting1] -4.456634e-05
##
                                  AGE_REF[fraction_babysitting1]
## (Intercept)
                                                    -4.456634e-05
## AGE_REF[fraction_babysitting1]
                                                     8.848599e-07
plot(model2)
```



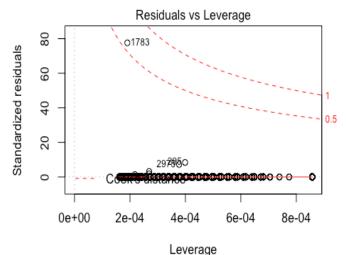
 $\label{lem:fitted_values} Fitted \ values \\ (fraction_babysitting[fraction_babysitting1] \sim AGE_REF[fraction_bab] \\$



 $\label{thm:continuous} Theoretical \ Quantiles \\ (fraction_babysitting[fraction_babysitting1] \sim AGE_REF[fraction_bab] \\$



Fitted values (fraction_babysitting[fraction_babysitting1] ~ AGE_REF[fraction_bab



(fraction_babysitting[fraction_babysitting1] ~ AGE_REF[fraction_bak

Explanation(model2):

we failed to reject the null hypothesis, because the P value of intercept & Age are 0.245 & 0.445, and they are greater than 0.05.

The R^2 is 9.462e-05, it indicates that they have very weak linear relationship between them.

Importation of Coefficient of determination:

- If R^2 is closer to 1, it indicated stronger linear relationship among the variables.
- If R^2 is zero, it indicated that there is no linear relationship among the variables.

• Measure proportion of the variance in the respond variable that is predictable from the explanatory variables.

Limitation of Coefficient of determination:

• The coefficient of determination cannot explain whether the linear relationship is positive or negative.

3)

- a. Fraction of reference person having a college degree spend more than 20% on health insurance is 0.020968357, 2.09%
- b. Fraction of reference person having less than college degree spend more than 20% is 0.041734861, 4.14%
- c. By preforming the hypothesis test for the 2 population statistical test, the difference is 2.05%, the standard error is 0.429 and the t statistic is 4.773, and the P-value is <.0001 for 2 tail test. We reject the null hypothesis at 5% level of significance, Since the P-value is less than 0.05. Therefore, the different in proportion is statistically significant.
- d. The overall share people with any degree is 41%. 26% of people spending more than 20% is made up with any college degree.

e. R result table:

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-0.17430	0.00000	0.02373	0.04640	0.06937	0.78730

The average fraction of spending going to insurance is 4.6%. I think the \pm 0% is not reasonable for the break point, because most people spend approximately 5% of spending on insurance. I think \pm 10% should be better to measure high or low medical spending. Therefore, \pm 10% consider as high medical spending and \pm 10% consider as low medical spending.

f. In the data_cex, there is a sample size of 6838, but in the sample subset provided for this assignment there are only (6289), (548) individuals omitted to state an education background. Those 548 individuals who did not state their educational background could have also spent money on insurances. Therefore, we should create another column for those 548 people who did not state their educational background to be counted in the results. Additionally without additionally information we are not able to remove individuals that are underage, no not have income and who are government social programs, therefore the cex data set makes allows for the data to be used more accurately.