Project

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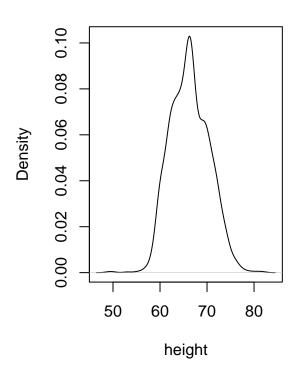
2023-04-20

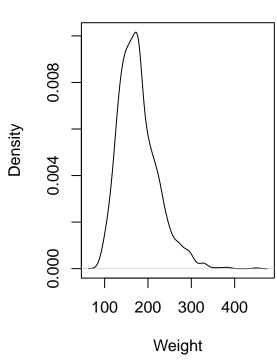
```
library(dplyr)
library(tidyverse)
library(knitr)
library(car)
options(scipen = 999)
df <- read.csv('newdf.csv')</pre>
df <- df %>% rename( "age" = "RIDAGEYR",
                     "gender" = "RIAGENDR",
                      "edu" = "DMDEDUC2",
                      "marry" = "DMDMARTL",
                     "race" = "RIDRETH3",
                     "height" = "WHD010",
                      "weight" = "WHD020",
                      "hi" = "HIQ011",
                      "ogtt" = "LBXGLT") %>% na.omit(df$ogtt)
# Deal with the abnormal value
df$height[df$height > 90] <- NA
df$weight[df$weight > 700] <- NA
df$hi[df$hi != 1] <- 2
# put the missing value with the mean of the data
df$weight[is.na(df$weight)] <- mean(df$weight, na.rm = TRUE)</pre>
df$height[is.na(df$height)] <- mean(df$height, na.rm = TRUE)</pre>
df$bmi <- signif((df$weight / (df$height^2))*703,4)</pre>
summary(df)
##
        height
                         weight
                                         hi
                                                         ogtt
                                                                         age
##
          :49.00
                                          :1.000
                                                    Min. : 35.0
                                                                    Min. :20.0
    Min.
                    Min. : 88
                                   Min.
    1st Qu.:63.00
                    1st Qu.:145
                                   1st Qu.:1.000
                                                    1st Qu.: 92.0
                                                                    1st Qu.:34.0
##
   Median :66.00
                    Median:170
                                   Median :1.000
                                                    Median :111.0
                                                                    Median:47.5
  Mean
##
           :66.31
                    Mean
                            :177
                                   Mean
                                          :1.206
                                                    Mean
                                                          :123.6
                                                                    Mean
                                                                           :48.3
##
    3rd Qu.:69.00
                    3rd Qu.:200
                                   3rd Qu.:1.000
                                                    3rd Qu.:142.0
                                                                    3rd Qu.:62.0
##
    Max.
           :82.00
                    Max.
                            :450
                                   Max.
                                          :2.000
                                                    Max.
                                                           :542.0
                                                                    Max.
                                                                            :80.0
##
        gender
                          edu
                                         marry
                                                           race
##
   Min.
           :1.000
                    Min.
                            :1.000
                                     Min.
                                            :1.000
                                                      Min.
                                                             :1.000
    1st Qu.:1.000
                    1st Qu.:3.000
                                     1st Qu.:1.000
                                                      1st Qu.:2.000
##
  Median :2.000
                    Median :4.000
                                     Median :1.000
                                                      Median :3.000
##
                                            :2.634
## Mean
          :1.509
                    Mean :3.504
                                                             :3.203
                                     Mean
                                                      Mean
   3rd Qu.:2.000
                    3rd Qu.:5.000
                                     3rd Qu.:5.000
                                                      3rd Qu.:4.000
##
  Max.
           :2.000
                    Max.
                           :5.000
                                     Max.
                                            :6.000
                                                      Max.
                                                             :7.000
##
         bmi
##
           :15.83
  \mathtt{Min}.
```

```
## 1st Qu.:23.80
## Median :27.17
## Mean :28.22
## 3rd Qu.:31.41
## Max. :64.23
# Set up the plotting window with two plots side-by-side
par(mfrow = c(1, 2))
plot(density(df$height), main = "Height Density", xlab = "height")
plot(density(df$weight), main = "Weight Density", xlab = "Weight")
```

Height Density

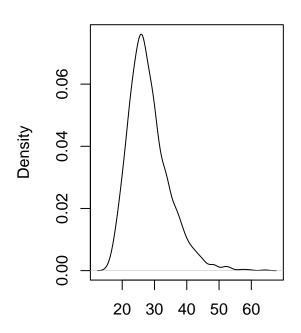
Weight Density



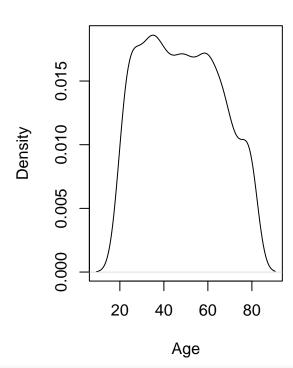


```
par(mfrow = c(1, 2))
plot(density(df$bmi), main = "BMI Density", xlab = "BMI")
plot(density(df$age), main = "Age Density", xlab = "Age")
```





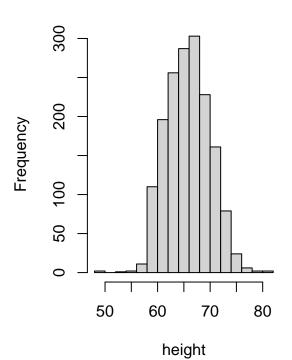
Age Density



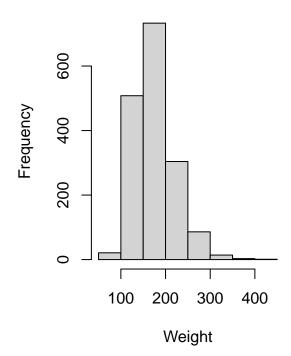
```
par(mfrow = c(1, 2))
hist(df$height, main = "Height Hist", xlab = "height")
hist(df$weight, main = "Weight Hist", xlab = "Weight")
```

Height Hist

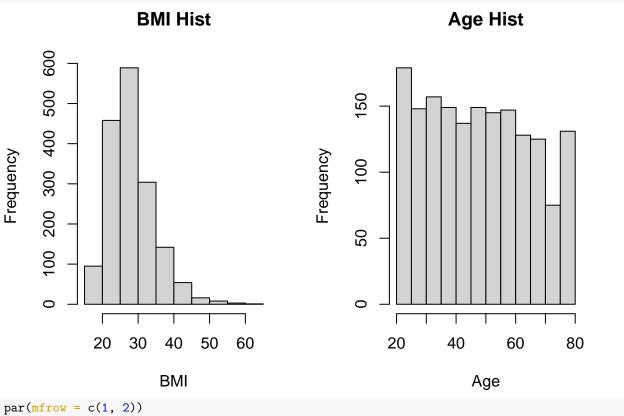
ВМІ



Weight Hist



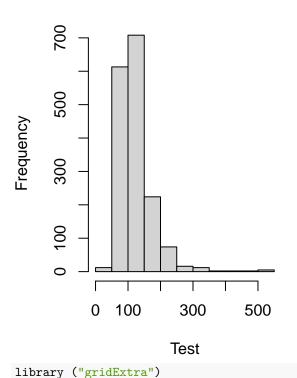
```
par(mfrow = c(1, 2))
hist(df$bmi, main = "BMI Hist", xlab = "BMI")
hist(df$age, main = "Age Hist", xlab = "Age")
```

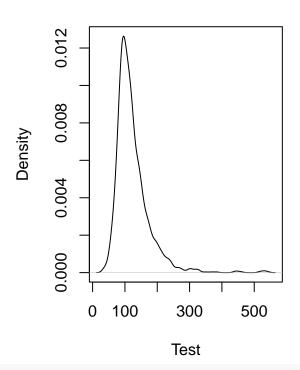


```
par(mfrow = c(1, 2))
hist(df$ogtt, main = "Oral Test Hist", xlab = "Test")
plot(density(df$ogtt), main = "Oral Test Density", xlab = "Test")
```

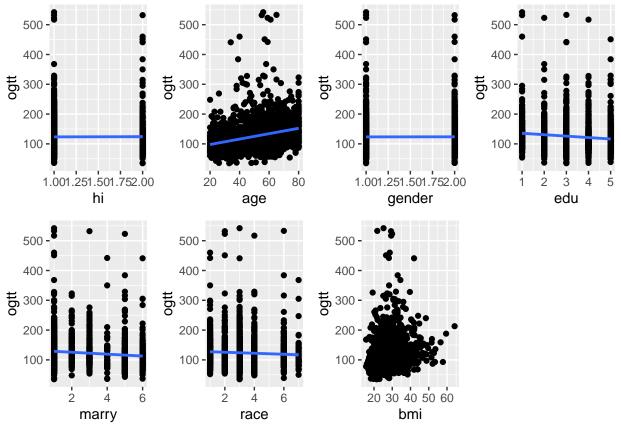
Oral Test Hist

Oral Test Density





```
##
## Attaching package: 'gridExtra'
## The following object is masked from 'package:dplyr':
##
##
       combine
require(ggplot2)
p1 <- ggplot(df, aes(hi, ogtt)) + geom_point() + stat_smooth(method="lm")
p2 <- ggplot(df, aes(age, ogtt)) + geom_point() + stat_smooth(method="lm")</pre>
p3 <- ggplot(df, aes(gender, ogtt)) + geom_point() +stat_smooth(method="lm")
p4 <- ggplot(df, aes(edu, ogtt)) + geom_point() + stat_smooth( method="lm")
p5 <- ggplot(df, aes(marry, ogtt)) + geom_point() + stat_smooth(method="lm")
p6 <- ggplot(df, aes(race, ogtt)) + geom_point() + stat_smooth(method="lm")</pre>
p7 <- ggplot(df, aes(bmi, ogtt)) + geom_point(position = position_jitter(width = .2,height=0))
grid.arrange(p1, p2, p3, p4, p5, p6, p7, nrow = 2)
## `geom_smooth()` using formula = 'y ~ x'
```



#Fitting a model
lmod <- lm(ogtt ~ .-weight-height, df)
summary(lmod)</pre>

##

```
## Call:
## lm(formula = ogtt ~ . - weight - height, data = df)
##
## Residuals:
      Min
              1Q Median
                            3Q
##
  -92.58 -28.78 -7.93
                        16.32 408.00
##
## Coefficients:
               Estimate Std. Error t value
                                                        Pr(>|t|)
##
## (Intercept) 40.47390
                          11.29572
                                     3.583
                                                        0.000349 ***
## hi
               10.36588
                           3.28110
                                     3.159
                                                        0.001610 **
## age
                0.93173
                           0.07835
                                    11.892 < 0.0000000000000000 ***
               -0.46345
                           2.45359
                                    -0.189
                                                        0.850205
## gender
## edu
               -3.21982
                           1.03195
                                    -3.120
                                                        0.001839 **
               -0.48940
                           0.68555
                                    -0.714
                                                        0.475405
## marry
## race
                1.04505
                           0.82420
                                     1.268
                                                        0.204994
## bmi
                1.25956
                           0.19532
                                     6.449
                                                  0.00000000147 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 49.94 on 1662 degrees of freedom
## Multiple R-squared: 0.1227, Adjusted R-squared: 0.119
```

```
## F-statistic: 33.2 on 7 and 1662 DF, p-value: < 0.00000000000000022
#Producing diagnistics plots:
par (mfrow = c(2,2))
plot (lmod)
                                                      Standardized residuals
                                                                           Normal Q-Q
                  Residuals vs Fitted
                                                                                              $556690
Residuals
      300
                                                            9
                                                            ^{\circ}
      -100
                                                            7
            80
                  100
                        120
                               140
                                     160
                                            180
                                                                        -2
                                                                                  0
                                                                                                 3
                                                                   -3
                                                                            -1
                       Fitted values
                                                                        Theoretical Quantiles
(Standardized residuals)
                                                      Standardized residuals
                    Scale-Location
                                                                     Residuals vs Leverage
                                                            \infty
      1.5
      0.0
                                                               0.000 0.005
            80
                  100
                        120
                               140
                                     160
                                            180
                                                                               0.010 0.015 0.020
                       Fitted values
                                                                              Leverage
par (mfrow = c(1,1))
cook <- cooks.distance(lmod)</pre>
cook[which(cook>0.5)]
## named numeric(0)
library("faraway")
##
## Attaching package: 'faraway'
## The following objects are masked from 'package:car':
##
##
        logit, vif
```

halfnorm(cook, 18, ylab = "Cook's Distance")

```
10 971
310 990
823
      0.04
      0.03
                                                                           396
                                                                          1418
Cook's Distance
                                                                          142
                                                                         847
      0.01
      8
             0.0
                        0.5
                                   1.0
                                              1.5
                                                         2.0
                                                                    2.5
                                                                               3.0
                                                                                         3.5
                                        Half-normal quantiles
df [971,]
##
        height weight hi ogtt age gender edu marry race
                   150 2 532 55
                                          2
                                               1
                                                      3
#oberservation 10 and 971 has the high distance
# Detect outliers using studentized residuals. Use the Bonferroni correction.
stud <- rstudent(lmod)</pre>
stud[which.max(abs(stud))]
```

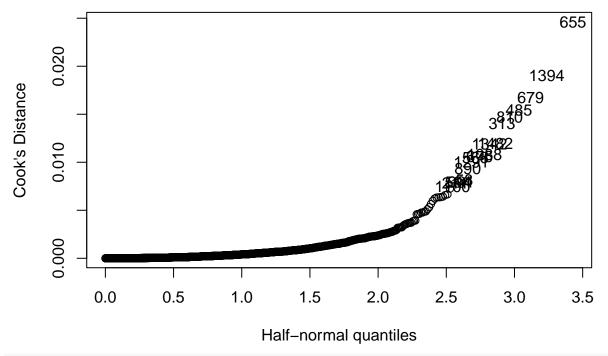
```
##
        19
## 8.357222
p <- 5
n <- nrow (model.matrix(lmod))</pre>
qt(1-.05/(n*2),n-p-1)
## [1] 4.185502
which(abs(stud) > qt(1-.05/(n*2), n-p-1))
    19 221 488 573 583 1205 1242 1423 1556 2108 2222
##
    10 142 331 396 403 823 847 971 1059 1418 1490
subsetdf \leftarrow df[c(-10, -19, -221, -488, -573, -583, -1205, -1242, -1423, -1556, -142, -331, -396, -403,
corMatrix <- cor(subsetdf)</pre>
corMatrix <- round (corMatrix, 2)</pre>
corMatrix
         height weight
                        hi ogtt
                                 age gender edu marry race
## height
          1.00
                0.46 -0.04 -0.05 -0.05 -0.68 0.13 0.03 0.10 -0.04
                 1.00 -0.05 0.14 -0.01 -0.29 0.01 -0.04 -0.09 0.86
## weight
          0.46
          ## hi
                0.14 -0.03 1.00 0.34 0.01 -0.10 -0.14 -0.06 0.18
## ogtt
          -0.05
```

0.02 -0.07 -0.35 -0.08 0.03

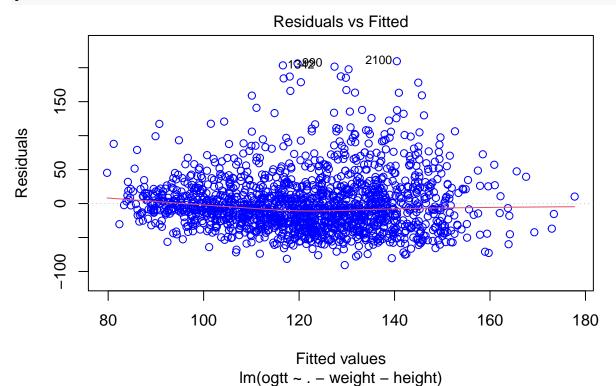
-0.05 -0.01 -0.27 0.34 1.00

age

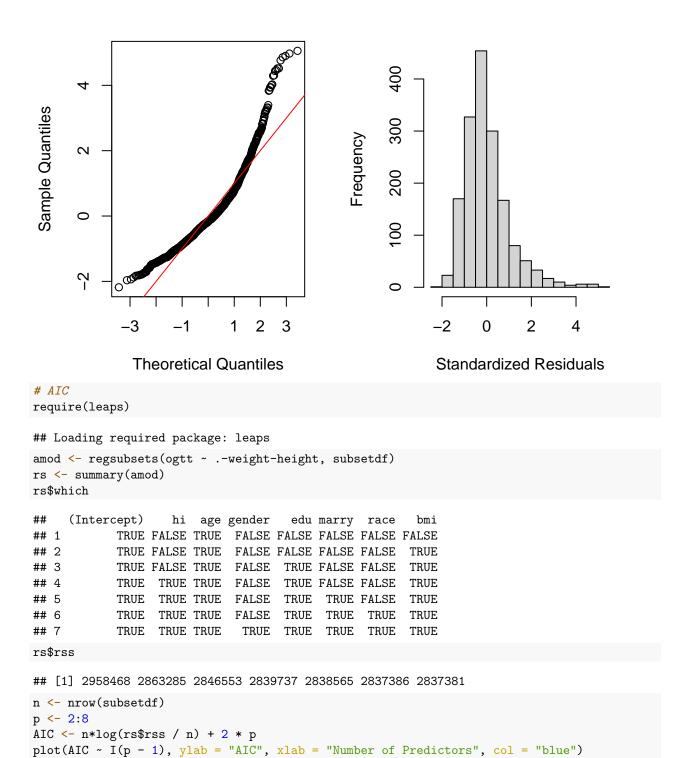
```
## gender -0.68 -0.29 -0.04 0.01 0.02 1.00 0.02 0.01 -0.05 0.05
## edu
           0.13
                ## marry
           0.03 -0.04 0.17 -0.14 -0.35 0.01 -0.04 1.00 0.01 -0.06
          0.10 -0.09 -0.13 -0.06 -0.08 -0.05 0.30 0.01 1.00 -0.16
## race
                0.86 -0.03 0.18 0.03
## bmi
          -0.04
                                       0.05 -0.06 -0.06 -0.16 1.00
vif(lmod)
                     gender
        hi
                age
                                edu
                                       marry
                                                race
## 1.179021 1.231975 1.007441 1.170569 1.153592 1.144568 1.033909
#new model without outliers
model2 <- lm(ogtt ~ .-weight-height, subsetdf)</pre>
summary(model2)
##
## Call:
## lm(formula = ogtt ~ . - weight - height, data = subsetdf)
##
## Residuals:
      Min
              1Q Median
                              3Q
                                    Max
## -90.567 -27.070 -6.727 17.465 209.521
##
## Coefficients:
             Estimate Std. Error t value
                                                   Pr(>|t|)
                                 4.893
## (Intercept) 46.44578
                        9.49170
                                           0.000001088828390 ***
              5.81777
                         2.76504
                                 2.104
                                                    0.03553 *
## age
              0.86729
                       0.06543 13.255 < 0.0000000000000000 ***
                         2.05443 -0.055
## gender
             -0.11337
                                                    0.95600
                         0.86579 -2.630
## edu
              -2.27663
                                                    0.00863 **
## marry
             -0.47407
                         0.57457 -0.825
                                                    0.40945
                         0.69143 0.822
## race
             0.56829
                                                    0.41125
## bmi
             1.19310
                         0.16364 7.291 0.00000000000476 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 41.57 on 1642 degrees of freedom
## Multiple R-squared: 0.1486, Adjusted R-squared: 0.145
## F-statistic: 40.95 on 7 and 1642 DF, p-value: < 0.000000000000000022
cook2 <- cooks.distance(model2)</pre>
halfnorm(cook2, 18, ylab = "Cook's Distance")
```

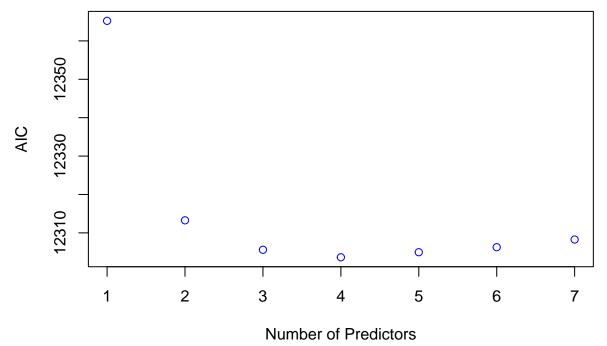


plot(model2, which = 1, col = "blue")



```
#Quantile Quantile plot and a histogram based on the standardized residuals
par (mfrow = c (1,2))
qqnorm(rstandard(model2), main = "")
abline(0,1, col = "red")
hist (rstandard(model2), main = "", xlab = "Standardized Residuals")
```





```
# Fourth has the loest AIC
model1 <- lm(ogtt ~ hi+age + edu +bmi, subsetdf)</pre>
summary(model1)
##
## Call:
## lm(formula = ogtt ~ hi + age + edu + bmi, data = subsetdf)
##
## Residuals:
##
      Min
                1Q Median
                                3Q
                                       Max
## -90.038 -27.130 -6.739 17.671 208.116
##
## Coefficients:
                                                       Pr(>|t|)
##
              Estimate Std. Error t value
                                     5.686
                                               0.0000001535162 ***
## (Intercept) 46.28888
                           8.14072
## hi
                5.44893
                           2.74227
                                     1.987
                                                         0.0471 *
               0.88013
                           0.06178
                                    14.247 < 0.000000000000000 ***
## age
## edu
              -2.05719
                           0.83375
                                    -2.467
                                                         0.0137 *
               1.17932
                           0.16114
                                     7.319
                                               0.0000000000039 ***
## bmi
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 41.55 on 1645 degrees of freedom
## Multiple R-squared: 0.1479, Adjusted R-squared: 0.1458
## F-statistic: 71.39 on 4 and 1645 DF, p-value: < 0.00000000000000022
require(MASS)
## Loading required package: MASS
```

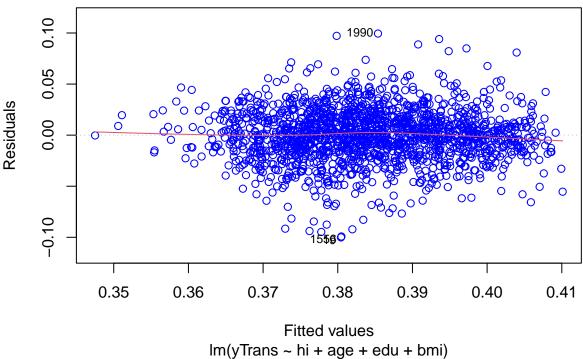
Attaching package: 'MASS'

```
## The following object is masked from 'package:dplyr':
##
##
         select
obj <- boxcox(model1, plotit = TRUE)</pre>
log-Likelihood
      -4600
       -5000
               -2
                                   -1
                                                         0
                                                                              1
                                                                                                  2
                                                         λ
obj <- boxcox(model1, plotit = TRUE, lambda=seq(-0.8,0,by=0.1))</pre>
                95%
       -4190
       -4210
log-Likelihood
       -4230
       -4250
             -0.8
                                  -0.6
                                                       -0.4
                                                                           -0.2
                                                                                                 0.0
                                                         λ
mlLambda <- obj$x[which.max(obj$y)]</pre>
mlLambda
```

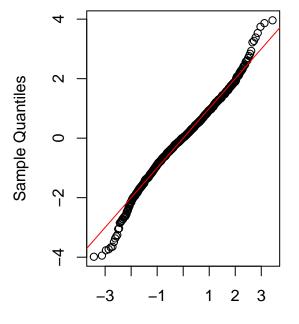
[1] -0.2020202

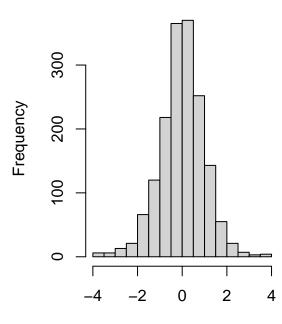
```
yTrans <- df$ogtt^(mlLambda) # Transform the response
lmodTrans <- lm(yTrans ~ hi + age + edu + bmi, data=df) # Fit a new regression</pre>
summary(lmodTrans)
##
## Call:
## lm(formula = yTrans ~ hi + age + edu + bmi, data = df)
## Residuals:
##
       Min
                 1Q
                       Median
                                   3Q
                                           Max
  ##
## Coefficients:
##
                Estimate Std. Error t value
                                                    Pr(>|t|)
## (Intercept) 0.43568807
                        0.00488297 89.226 < 0.000000000000000 ***
## hi
             -0.00490832
                        0.00164002 -2.993
                                                      0.0028 **
             -0.00055486
                        0.00003727 -14.886 < 0.000000000000000 ***
##
  age
              0.00112185 0.00050058
                                    2.241
                                                      0.0252 *
## edu
             ## bmi
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.02516 on 1665 degrees of freedom
## Multiple R-squared: 0.1598, Adjusted R-squared: 0.1578
## F-statistic: 79.19 on 4 and 1665 DF, p-value: < 0.00000000000000000022
plot(lmodTrans, which = 1, col = "blue")
```

Residuals vs Fitted



```
par (mfrow = c (1,2))
qqnorm(rstandard(lmodTrans), main = "")
abline(0,1, col = "red")
hist (rstandard(lmodTrans), main = "", xlab = "Standardized Residuals")
```





Theoretical Quantiles

Standardized Residuals