Zhang_Yue_FinalProject

Yue Zhang

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[1] "/Users/yuezhang/Documents/Biostat/Biostatistics/PHL_1700/Code"

getwd()

library(tidyverse)
library(lubridate)

1 0 0 0

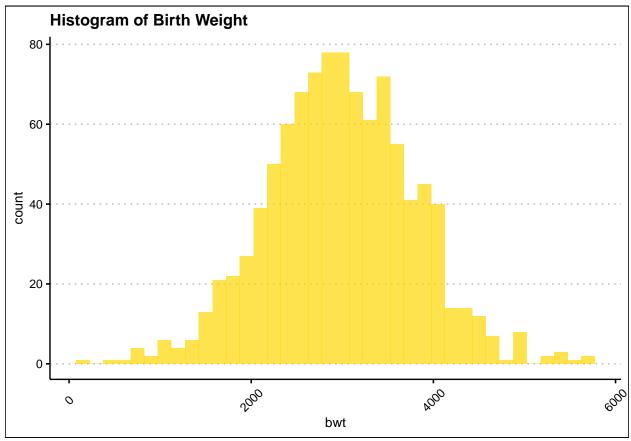
0

```
library(dplyr)
library(ggthemes)
library(ggplot2)
library(readxl)
library(lmtest)
library(mfx)
library(pROC)
library(haven)
library(car)
library(PMCMRplus)
library(VGAM)
library(describedata)
library(olsrr)
mytheme = theme_clean(base_size = 12) +
    axis.text = element_text(color = "black"),
    legend.position = "right",
    axis.text.x = element_text(
      angle = 45,
      vjust = 0.5,
      hjust = 0.5
    ),
    plot.title = element_text(size = 12)
theme_set(mytheme)
#Load Data
bwt = read.csv("/Users/yuezhang/Documents/Biostat/Biostatistics/PHL_1700/Data/Raw/Birthweight data Chen
#Data Exploration
#Check Missing Data
bwt %>% summarize(across(everything(), ~sum(is.na(.))))
## id low age lwt smoke ptl ht ui bwt race
```

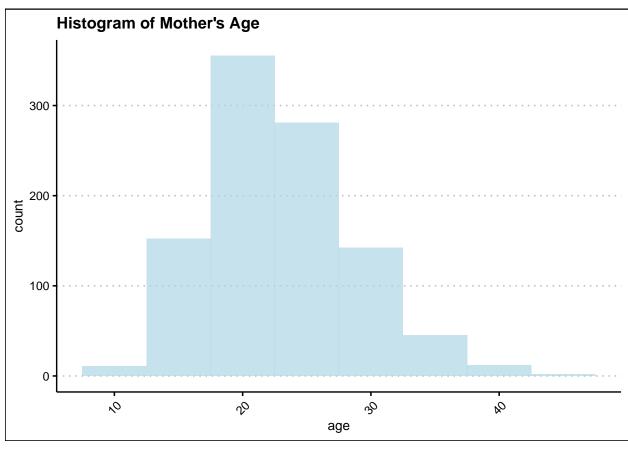
0 0 0 0

```
#Descriptive Statistics for Low Birth Weight
summary_low = bwt %>% filter(low == 1) %>% summarize(
  age mean = mean(age, na.rm = TRUE),
  age_sd = sd(age, na.rm = TRUE),
  lwt_mean = mean(lwt, na.rm = TRUE),
  lwt_sd = sd(lwt, na.rm = TRUE),
  race_prop = list(prop.table(table(race))),
  smoke_prop = mean(smoke == 1, na.rm = TRUE),
  ptl_prop = mean(ptl == 1, na.rm = TRUE),
  ht_prop = mean(ht == 1, na.rm = TRUE),
  ui_prop = mean(ui == 1, na.rm = TRUE)
print(summary_low)
##
     age_mean
                age_sd lwt_mean lwt_sd
                                                                race_prop
## 1 22.36839 5.805274 143.3286 28.67558 0.4944238, 0.1078067, 0.3977695
   smoke_prop ptl_prop ht_prop ui_prop
## 1 0.4832714 0.197026 0.0929368 0.2416357
print(summary_low$race_prop)
## [[1]]
## race
           1
## 0.4944238 0.1078067 0.3977695
low = bwt %>% filter(low == 1)
cau = low %>% filter(race == 1)
aa = low %>% filter(race == 2)
other = low %>% filter(race ==3)
smoke = low %>% filter(smoke ==1)
pl = low %>% filter(ptl == 1)
hy = low \%\% filter(ht == 1)
u = low %>% filter(ui == 1)
high = bwt %>% filter(low == 0)
cau2 = high %>% filter(race == 1)
aa2 = high %>% filter(race == 2)
other2 = high %>% filter(race ==3)
smoke2 = high %>% filter(smoke ==1)
pl2 = high %>% filter(ptl == 1)
hy2 = high \%\% filter(ht == 1)
u2 = high %>% filter(ui == 1)
#Descriptive Statistics for High Birth Weight
summary_high = bwt %>% filter(low == 0) %>% summarize(
  age_mean = mean(age, na.rm = TRUE),
  age_sd = sd(age, na.rm = TRUE),
  lwt_mean = mean(lwt, na.rm = TRUE),
  lwt sd = sd(lwt, na.rm = TRUE),
  race_prop = list(prop.table(table(race))),
  smoke_prop = mean(smoke == 1, na.rm = TRUE),
  ptl_prop = mean(ptl == 1, na.rm = TRUE),
  ht_prop = mean(ht == 1, na.rm = TRUE),
```

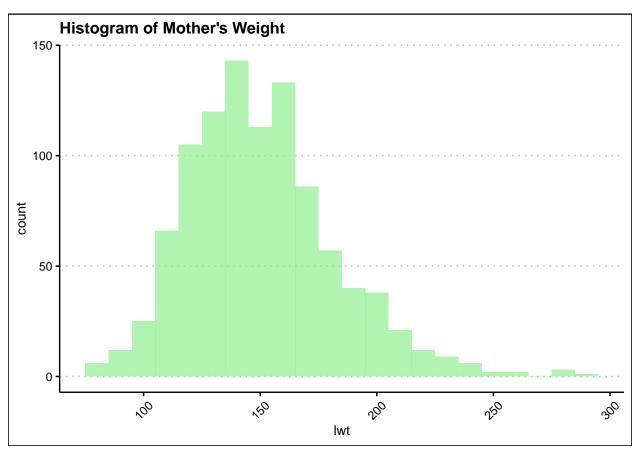
```
ui_prop = mean(ui == 1, na.rm = TRUE)
print(summary_high)
     age_mean age_sd lwt_mean
                                lwt_sd
                                                              race_prop
## 1 23.04547 5.497755 153.1519 32.25093 0.5745554, 0.1244870, 0.3009576
    smoke_prop ptl_prop
                            ht_prop ui_prop
## 1 0.3584131 0.1600547 0.05471956 0.1025992
print(summary_high$race_prop)
## [[1]]
## race
##
          1
                    2
## 0.5745554 0.1244870 0.3009576
#Histogram of Birth Weight
ggplot(bwt, aes(x = bwt)) +
  geom_histogram(binwidth = 150, fill = "gold", alpha = 0.7) +
  ggtitle("Histogram of Birth Weight")
```



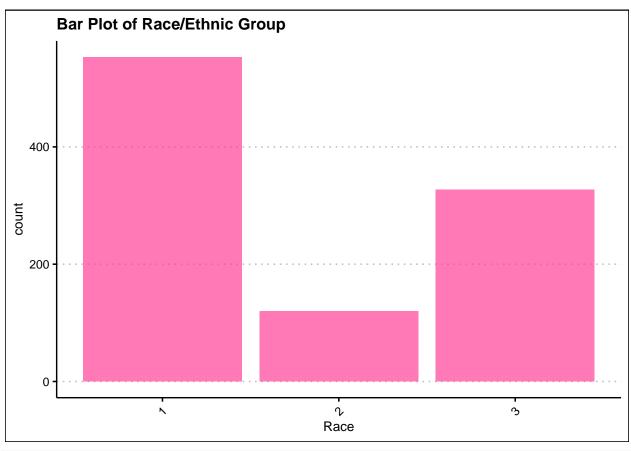
```
#Histogram of Mother's Age
ggplot(bwt, aes(x = age)) +
  geom_histogram(binwidth = 5, fill = "lightblue", alpha = 0.7) +
  ggtitle("Histogram of Mother's Age")
```



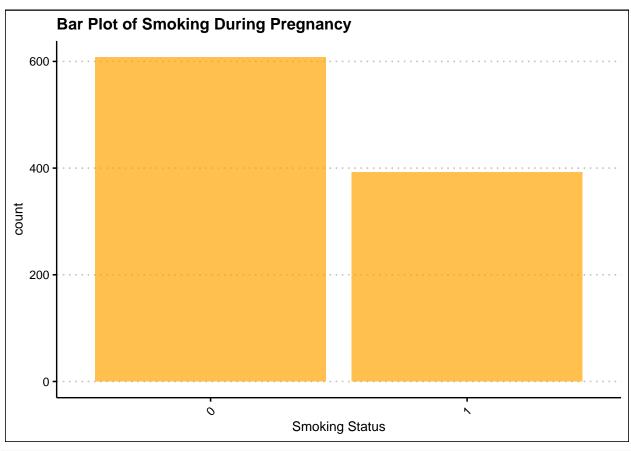
```
#Histogram of Mother's Weight at Last Menstrual Period
ggplot(bwt, aes(x = lwt)) +
  geom_histogram(binwidth = 10, fill = "lightgreen", alpha = 0.7) +
  ggtitle("Histogram of Mother's Weight")
```



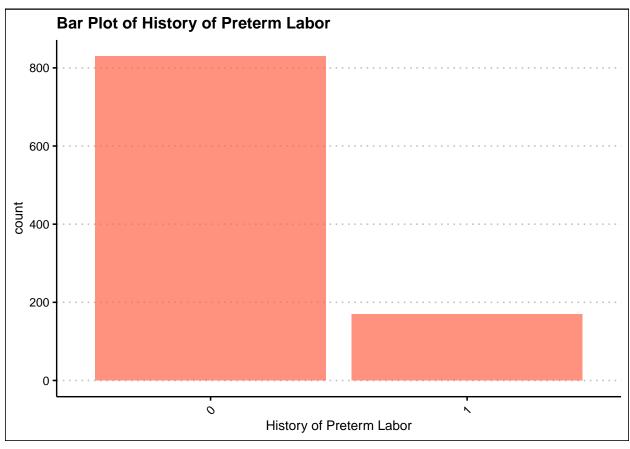
```
#Bar Plot of Race
ggplot(bwt, aes(x = factor(race))) +
  geom_bar(fill = "violetred1", alpha = 0.7) +
  xlab("Race") +
  ggtitle("Bar Plot of Race/Ethnic Group")
```



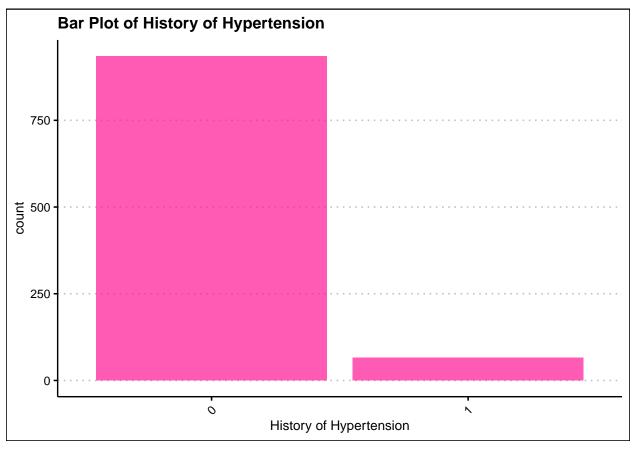
```
#Bar Plot of Smoking Status
ggplot(bwt, aes(x = factor(smoke))) +
  geom_bar(fill = "orange", alpha = 0.7) +
  xlab("Smoking Status") +
  ggtitle("Bar Plot of Smoking During Pregnancy")
```



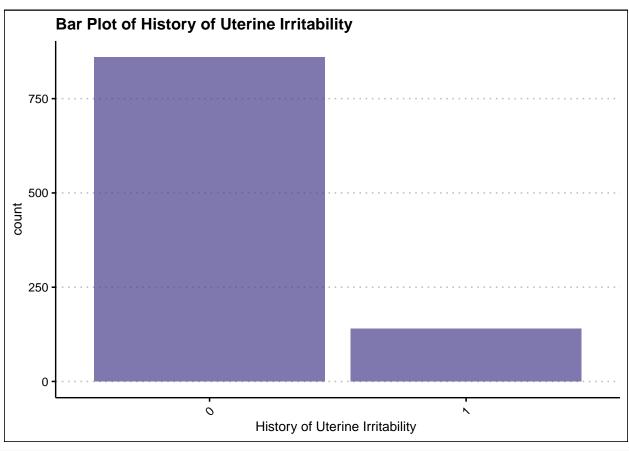
```
#Bar Plot of History of Preterm Labor
ggplot(bwt, aes(x = factor(ptl))) +
  geom_bar(fill = "tomato1", alpha = 0.7) +
  xlab("History of Preterm Labor") +
  ggtitle("Bar Plot of History of Preterm Labor")
```



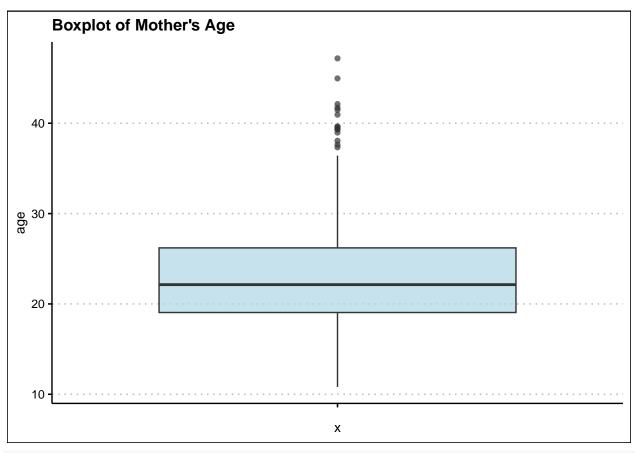
```
#Bar Plot of History of Hypertension
ggplot(bwt, aes(x = factor(ht))) +
  geom_bar(fill = "deeppink", alpha = 0.7) +
  xlab("History of Hypertension") +
  ggtitle("Bar Plot of History of Hypertension")
```



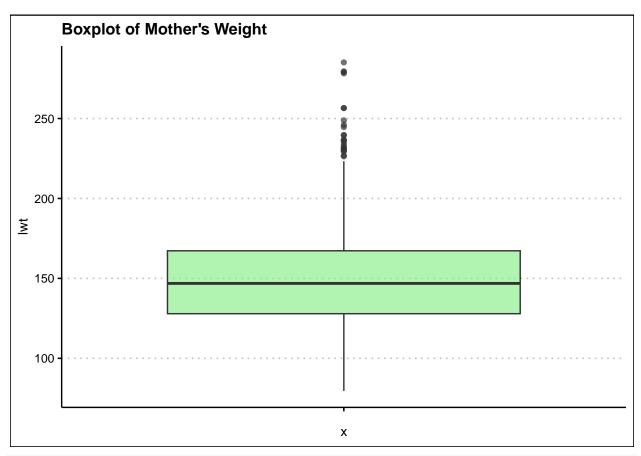
```
#Bar Plot of History of Uterine Irritability
ggplot(bwt, aes(x = factor(ui))) +
  geom_bar(fill = "darkslateblue", alpha = 0.7) +
  xlab("History of Uterine Irritability ") +
  ggtitle("Bar Plot of History of Uterine Irritability ")
```



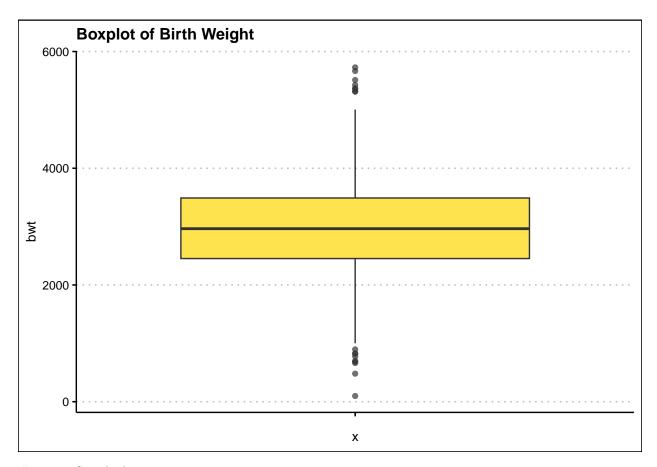
```
#Identify Outliers
ggplot(bwt, aes(x = "", y = age)) +
  geom_boxplot(fill = "lightblue", alpha = 0.7) +
  ggtitle("Boxplot of Mother's Age")
```



```
ggplot(bwt, aes(x = "", y = lwt)) +
geom_boxplot(fill = "lightgreen", alpha = 0.7) +
ggtitle("Boxplot of Mother's Weight")
```



```
ggplot(bwt, aes(x = "", y = bwt)) +
geom_boxplot(fill = "gold", alpha = 0.7) +
ggtitle("Boxplot of Birth Weight")
```



```
#Testing Simple Association
```

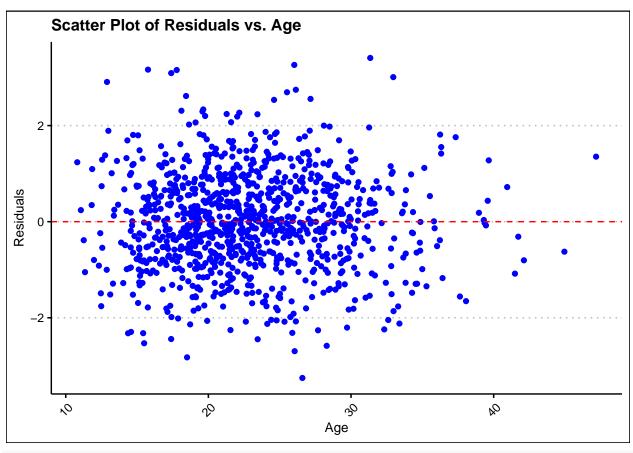
5.665601 13.981135

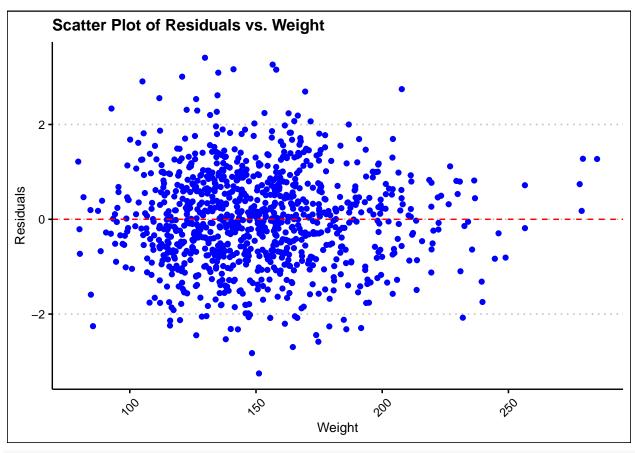
```
#Perform a t-test comparing age between low birth weight (low = 1) and normal birth weight (low = 0)
t.test(age ~ low, data = bwt)
##
   Welch Two Sample t-test
##
##
## data: age by low
## t = 1.6587, df = 455.86, p-value = 0.09787
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
## -0.1251195 1.4792737
## sample estimates:
## mean in group 0 mean in group 1
                         22.36839
         23.04547
#Perform a t-test comparing weight between low birth weight (low = 1) and normal birth weight (low = 0)
t.test(lwt ~ low, data = bwt)
   Welch Two Sample t-test
##
##
## data: lwt by low
## t = 4.6413, df = 533.15, p-value = 4.364e-06
## alternative hypothesis: true difference in means between group 0 and group 1 is not equal to 0
## 95 percent confidence interval:
```

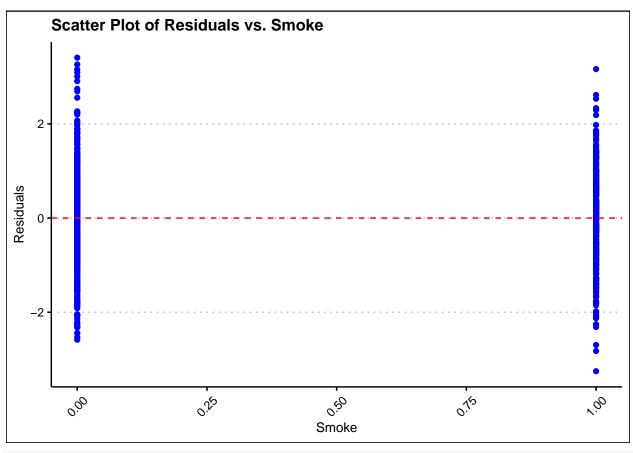
```
## sample estimates:
## mean in group 0 mean in group 1
          153.1519
                          143.3286
#Chi-square test for association between smoking status and birth weight
chisq.test(table(bwt$smoke, bwt$low))
##
##
  Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(bwt$smoke, bwt$low)
## X-squared = 12.344, df = 1, p-value = 0.0004425
#Chi-square test for association between race and birth weight
chisq.test(table(bwt$race, bwt$low))
##
##
  Pearson's Chi-squared test
## data: table(bwt$race, bwt$low)
## X-squared = 8.3753, df = 2, p-value = 0.01518
bwt$is_caucasian = ifelse(bwt$race == 1, 1, 0)
caucasian_low_table = table(bwt$is_caucasian, bwt$low)
print(caucasian_low_table)
##
##
         0
             1
     0 311 136
##
     1 420 133
chisq_result = chisq.test(caucasian_low_table)
chisq_result
##
## Pearson's Chi-squared test with Yates' continuity correction
## data: caucasian_low_table
## X-squared = 4.7889, df = 1, p-value = 0.02864
bwt$is_aa = ifelse(bwt$race == 2, 1, 0)
aa_low_table = table(bwt$is_aa, bwt$low)
print(aa_low_table)
##
##
         0
             1
##
     0 640 240
##
     1 91 29
chisq_result2 = chisq.test(aa_low_table)
chisq_result2
##
## Pearson's Chi-squared test with Yates' continuity correction
```

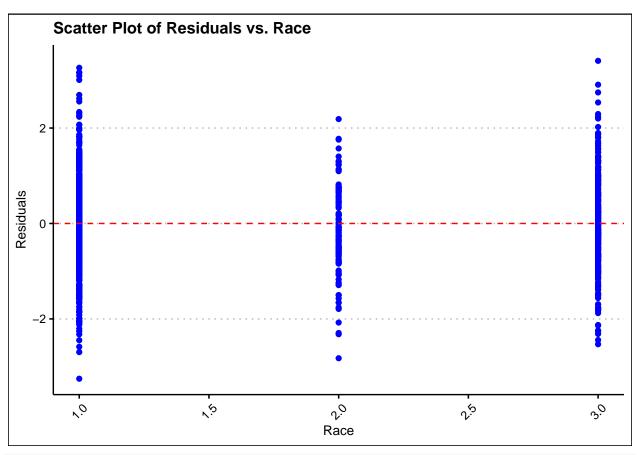
```
##
## data: aa_low_table
## X-squared = 0.37218, df = 1, p-value = 0.5418
bwt$is_o = ifelse(bwt$race == 3, 1, 0)
o_low_table = table(bwt$is_o, bwt$low)
print(o_low_table)
##
##
         Ω
             1
##
     0 511 162
     1 220 107
##
chisq_result3 = chisq.test(o_low_table)
chisq_result3
##
##
   Pearson's Chi-squared test with Yates' continuity correction
##
## data: o_low_table
## X-squared = 7.9405, df = 1, p-value = 0.004834
#Chi-square test for association between preterm labor and birth weight
chisq.test(table(bwt$ptl, bwt$low))
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(bwt$ptl, bwt$low)
## X-squared = 1.6519, df = 1, p-value = 0.1987
#Chi-square test for association between history of hypertension and birth weight
chisq.test(table(bwt$ht, bwt$low))
##
   Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(bwt$ht, bwt$low)
## X-squared = 4.1178, df = 1, p-value = 0.04244
#Chi-square test for association between history of uterine irritability and birth weight
chisq.test(table(bwt$ui, bwt$low))
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: table(bwt$ui, bwt$low)
## X-squared = 30.428, df = 1, p-value = 3.465e-08
#Linear Regression
bwt_model = lm(bwt ~ age + lwt + race + smoke + ptl + ht + ui, data = bwt)
summary(bwt_model)
## Call:
## lm(formula = bwt ~ age + lwt + race + smoke + ptl + ht + ui,
```

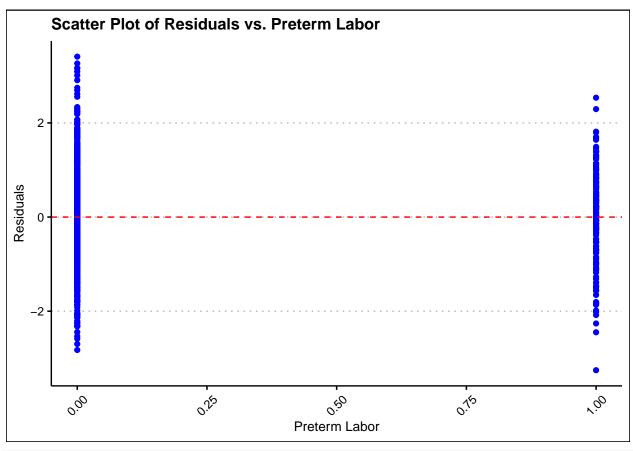
```
data = bwt)
##
##
## Residuals:
               1Q Median
##
       Min
                                  ЗQ
                                          Max
## -2404.64 -498.20
                     -5.41 495.22 2518.72
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2747.2149 163.8710 16.765 < 2e-16 ***
                         4.3744 -0.286 0.774821
## age
              -1.2518
## lwt
                 4.6710
                           0.7806 5.984 3.04e-09 ***
                        27.0320 -5.012 6.38e-07 ***
## race
              -135.4816
              -275.5430 50.2206 -5.487 5.20e-08 ***
## smoke
## ptl
               57.7048 64.3120 0.897 0.369797
## ht
              -330.3174 97.3277 -3.394 0.000716 ***
## ui
              -560.3181
                        68.6249 -8.165 9.71e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 745.7 on 992 degrees of freedom
## Multiple R-squared: 0.1538, Adjusted R-squared: 0.1478
## F-statistic: 25.76 on 7 and 992 DF, p-value: < 2.2e-16
sr = rstudent(bwt_model)
#Scatter Plot of Studentized Residuals vs. Age
ggplot(bwt, aes(x = age, y = sr)) +
 geom point(color = "blue") +
 geom_hline(yintercept = 0, color = "red", linetype = "dashed") +
 labs(title = "Scatter Plot of Residuals vs. Age",
      x = "Age",
      y = "Residuals")
```

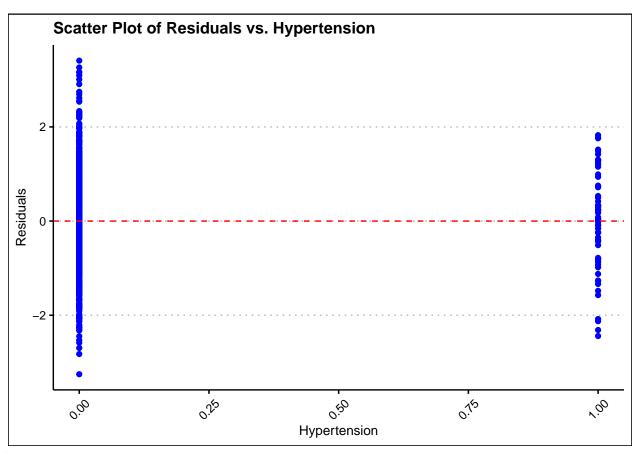


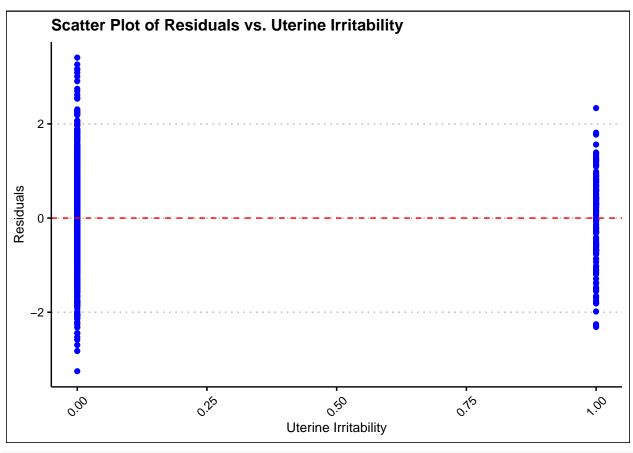


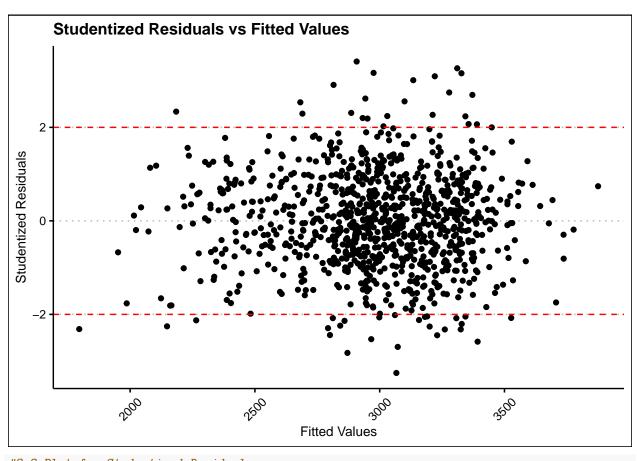






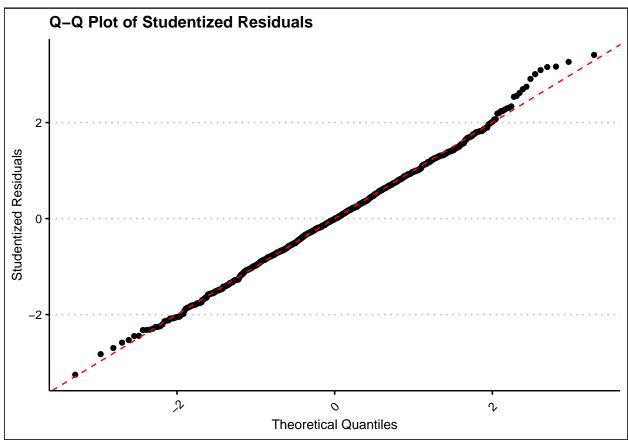


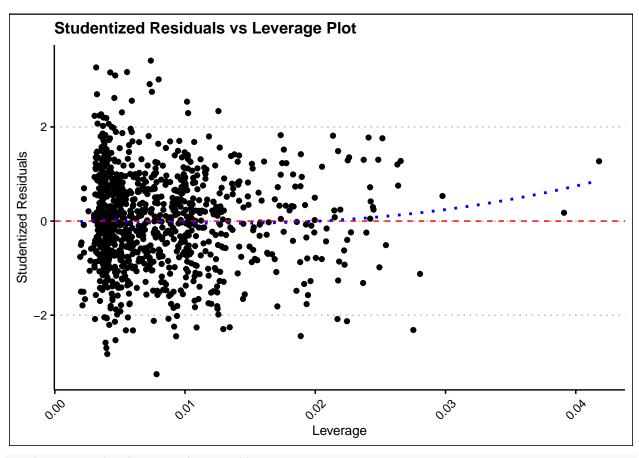




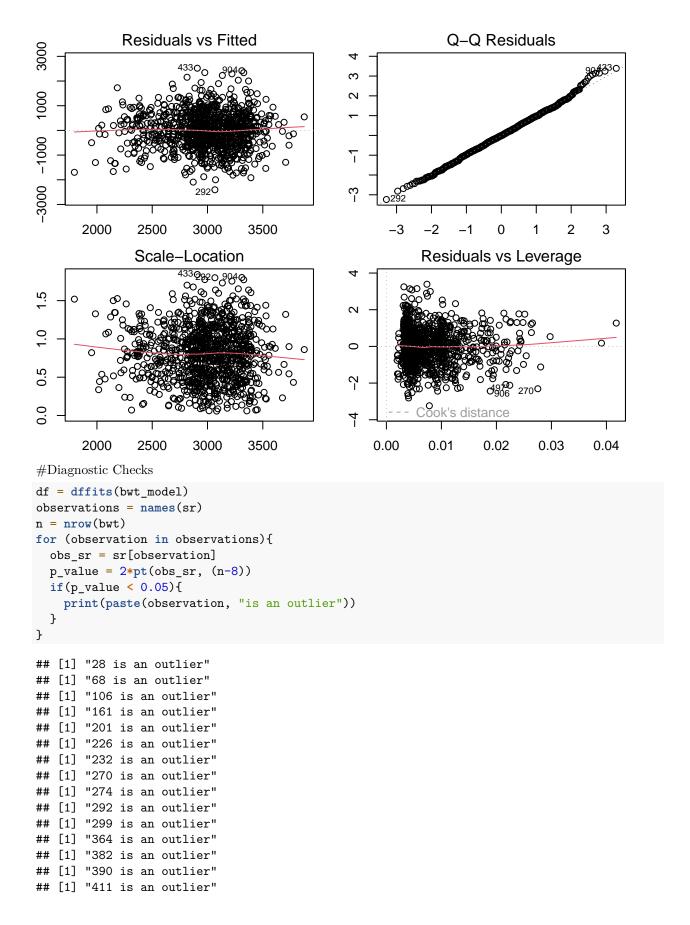
```
#Q-Q Plot for Studentized Residuals
qq_data = data.frame(
   Theoretical = qqnorm(sr, plot.it = FALSE)$x,
   Sample = qqnorm(sr, plot.it = FALSE)$y
)

ggplot(qq_data, aes(x = Theoretical, y = Sample)) +
   geom_point() +
   geom_abline(
   intercept = 0,
   slope = 1,
   color = "red",
   linetype = "dashed"
) +
   labs(title = "Q-Q Plot of Studentized Residuals", x = "Theoretical Quantiles", y = "Studentized Residuals"
```





```
par(mfrow = c(2,2), mar=c(2,2,2,2))
plot(bwt_model, cex.axis = 1, cex.lab = 1)
```



```
## [1] "436 is an outlier"
## [1] "457 is an outlier"
## [1] "497 is an outlier"
## [1] "499 is an outlier"
## [1] "513 is an outlier"
## [1] "519 is an outlier"
## [1] "549 is an outlier"
## [1] "608 is an outlier"
## [1] "710 is an outlier"
## [1] "887 is an outlier"
## [1] "892 is an outlier"
## [1] "906 is an outlier"
## [1] "946 is an outlier"
p = length(coef(bwt_model))
avgLeverage = 2*p/n
highLeverage = which(h > avgLeverage)
influential = which(df > 2*sqrt(p/n))
print(paste("High Leverage Point:", toString(highLeverage)))
## [1] "High Leverage Point: 3, 9, 13, 29, 33, 48, 60, 64, 69, 78, 82, 116, 121, 146, 148, 163, 178, 18
print(paste("Influential:", toString(influential)))
## [1] "Influential: 48, 60, 62, 69, 148, 183, 316, 383, 433, 536, 550, 556, 632, 693, 704, 706, 709, 8
As results shown, we will remove all the outliers and influential points. Then do the linear regression again to
see whether the model improved.
#Refit the Model
bwt_new = bwt[-c(28, 48, 60, 62, 68, 69, 106, 148, 161, 183, 201, 226, 232, 270, 274, 292, 299, 316, 36
bwt_model2 = lm(bwt ~ age + lwt + race + smoke + ptl + ht + ui, data = bwt_new)
summary(bwt_model2)
##
## lm(formula = bwt ~ age + lwt + race + smoke + ptl + ht + ui,
##
       data = bwt_new)
##
## Residuals:
##
        Min
                  1Q
                       Median
                                     3Q
                                             Max
## -1419.32 -480.62
                       -13.09
                                457.75
                                        1981.77
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2749.3465
                           149.6874 18.367
                                             < 2e-16 ***
## age
                 -3.0358
                             4.0035
                                     -0.758
                                                0.448
## lwt
                             0.7088
                                      7.150 1.75e-12 ***
                  5.0675
## race
               -141.4258
                            24.2364
                                     -5.835 7.39e-09 ***
                            45.0514
                                     -6.118 1.39e-09 ***
               -275.6111
## smoke
## ptl
                 73.6399
                            58.0798
                                      1.268
                                                0.205
## ht
               -457.4925
                            95.8887
                                     -4.771 2.12e-06 ***
               -566.2879
                            61.9126
                                     -9.147 < 2e-16 ***
## ui
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
## Residual standard error: 650.1 on 937 degrees of freedom
## Multiple R-squared: 0.2022, Adjusted R-squared: 0.1962
## F-statistic: 33.93 on 7 and 937 DF, p-value: < 2.2e-16
par(mfrow = c(2,2), mar=c(2,2,2,2))
plot(bwt_model2, cex.axis = 1, cex.lab = 1)
                                                             Q-Q Residuals
             Residuals vs Fitted
                                                                                  656810
1000
-1000
    2000
             2500
                     3000
                                       4000
                                                                     0
                                                                          1
                                                                               2
                                                                                    3
                              3500
                                                    -3
                                                         Residuals vs Leverage
              Scale-Location
0
0.5
0.0
    2000
             2500
                     3000
                              3500
                                       4000
                                                  0.00
                                                         0.01
                                                                 0.02
                                                                        0.03
                                                                                0.04
olsrr::ols_step_best_subset(bwt_model2)
             Best Subsets Regression
##
   Model Index
                  Predictors
##
##
        1
                  ui
                  lwt ui
##
##
                  lwt smoke ui
##
                  lwt race smoke ui
                  lwt race smoke ht ui
##
                  lwt race smoke ptl ht ui
                  age lwt race smoke ptl ht ui
##
##
##
                                                                    Subsets Regression Summary
##
                           Adj.
                                       Pred
## Model
                        R-Square
                                     R-Square
                                                    C(p)
                                                                 AIC
                                                                                SBIC
                                                                                              SBC
            R-Square
```

##

```
##
             0.0830
                         0.0820
                                     0.0791
                                               136.0585
                                                           15053.2647
                                                                         12370.9504
                                                                                       15067.8183
    1
##
    2
             0.1310
                         0.1292
                                     0.1257
                                                81.6224
                                                           15004.4039
                                                                         12322.1545
                                                                                       15023.8086
##
    3
             0.1529
                         0.1502
                                     0.146
                                                57.8495
                                                          14982.2341
                                                                         12300.0344
                                                                                       15006.4900
##
    4
             0.1813
                                     0.1729
                                                26.5286
                                                           14952.0407
                         0.1778
                                                                         12270.0743
                                                                                       14981.1478
##
    5
             0.2005
                         0.1963
                                     0.1911
                                                 5.9890
                                                           14931.6247
                                                                         12249.9076
                                                                                       14965.5830
##
    6
             0.2017
                         0.1966
                                     0.1906
                                                 6.5750
                                                           14932.2006
                                                                         12250.5175
                                                                                       14971.0101
##
             0.2022
                         0.1962
                                     0.1894
                                                 8.0000
                                                           14933.6209
                                                                         12251.9635
                                                                                       14977.2815
## ----
## AIC: Akaike Information Criteria
## SBIC: Sawa's Bayesian Information Criteria
## SBC: Schwarz Bayesian Criteria
## MSEP: Estimated error of prediction, assuming multivariate normality
## FPE: Final Prediction Error
## HSP: Hocking's Sp
## APC: Amemiya Prediction Criteria
#Reduced model: bwt ~ lwt + race + smoke + ht + ui
bwt_model3 = lm(bwt ~ lwt + race + smoke + ht + ui, data = bwt_new)
summary(bwt_model3)
##
## Call:
## lm(formula = bwt ~ lwt + race + smoke + ht + ui, data = bwt_new)
##
## Residuals:
##
                 1Q
                     Median
                                   3Q
                                           Max
       Min
## -1439.62 -476.60
                      -12.95
                               451.58 1970.75
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                          124.4408 21.743 < 2e-16 ***
## (Intercept) 2705.7584
## lwt
                           0.6982
                                    7.041 3.67e-12 ***
                 4.9160
                           23.9583 -5.785 9.87e-09 ***
## race
              -138.5982
## smoke
              -268.0582
                           44.6746 -6.000 2.81e-09 ***
                           95.8639 -4.748 2.38e-06 ***
## ht
              -455.1259
## ui
              -556.5304
                           61.5186 -9.047 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 650.1 on 939 degrees of freedom
## Multiple R-squared: 0.2005, Adjusted R-squared: 0.1963
## F-statistic: 47.1 on 5 and 939 DF, p-value: < 2.2e-16
par(mfrow = c(2,2), mar=c(2,2,2,2))
plot(bwt_model3, cex.axis = 1, cex.lab = 1)
```

4

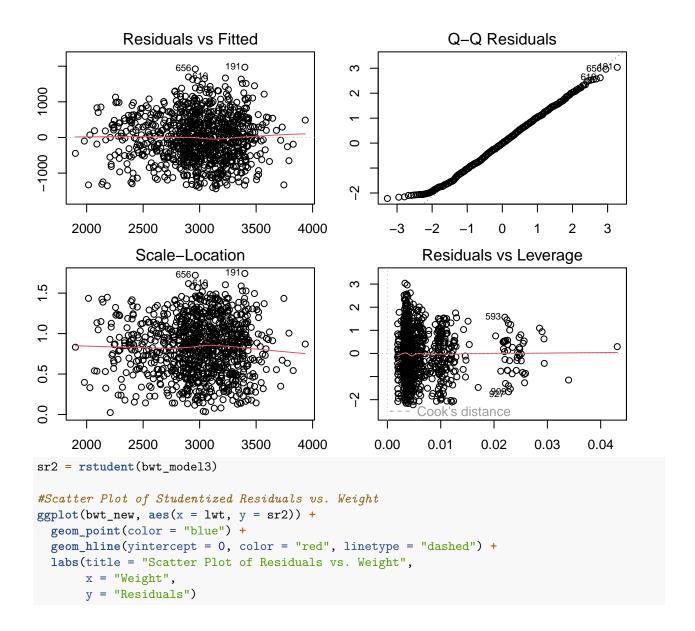
4

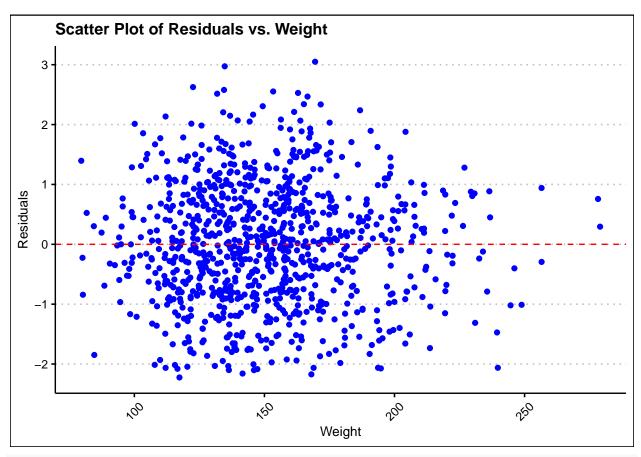
4

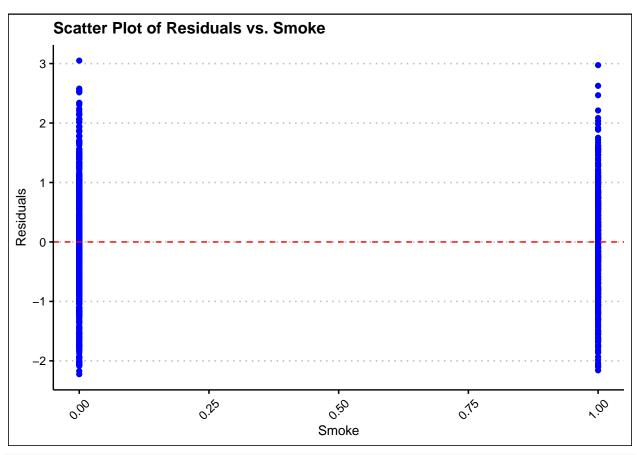
4

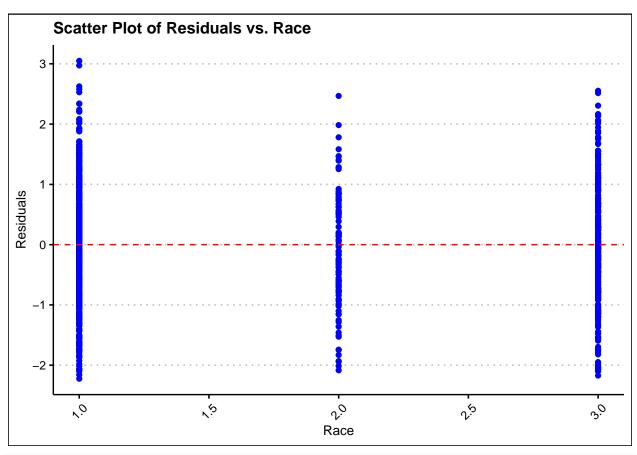
3

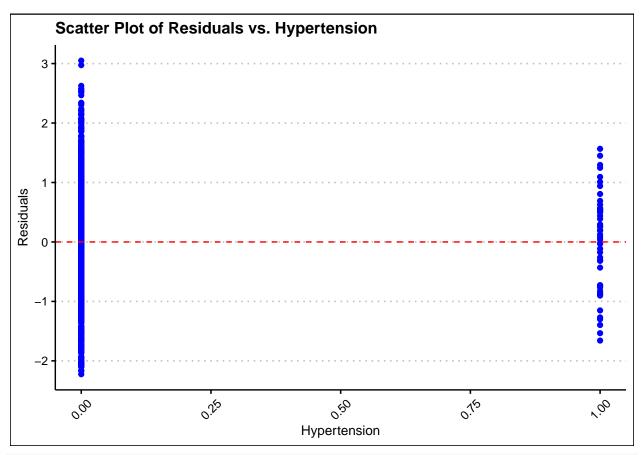
3

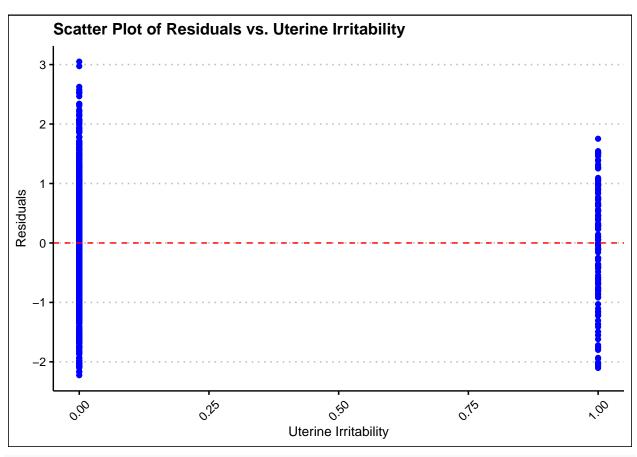


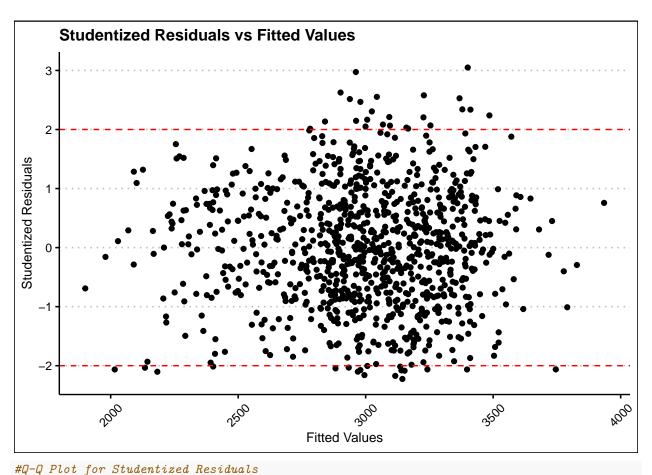






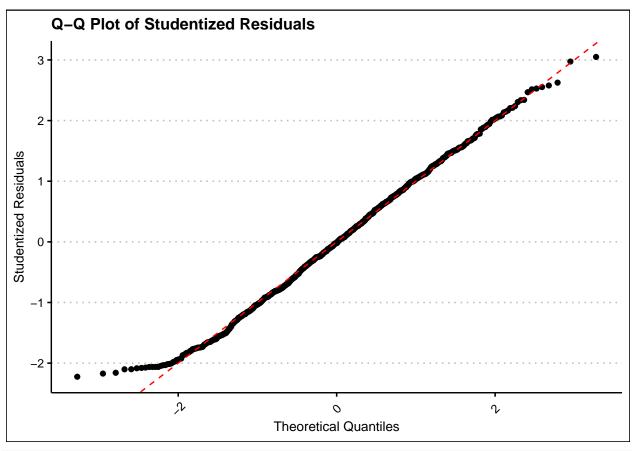


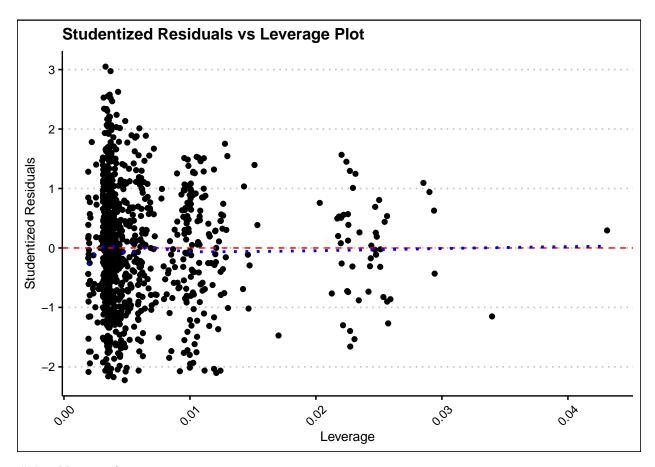




```
qq_data2 = data.frame(
  Theoretical = qqnorm(sr2, plot.it = FALSE)$x,
  Sample = qqnorm(sr2, plot.it = FALSE)$y
)

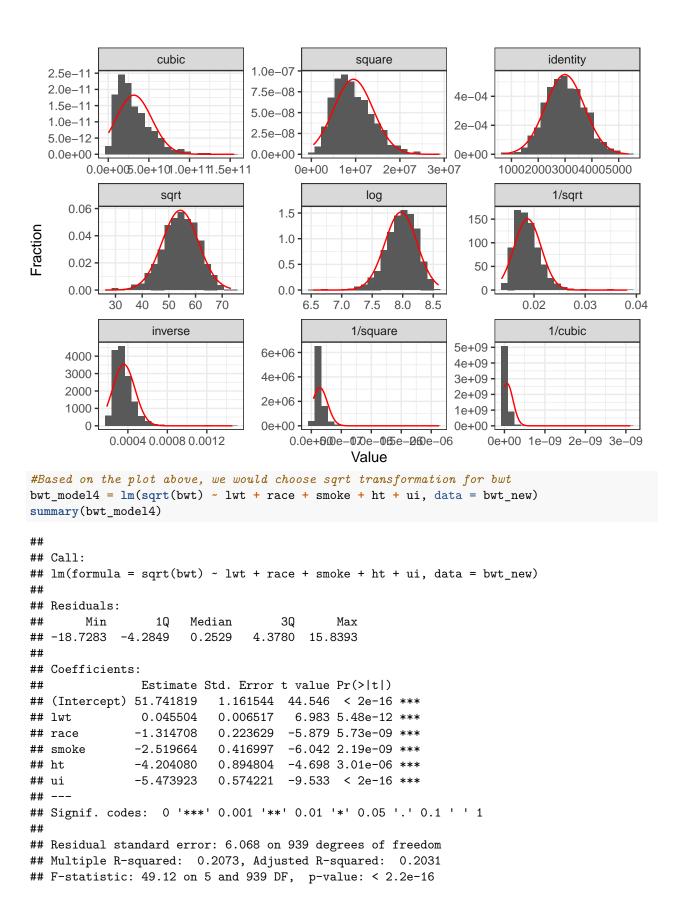
ggplot(qq_data2, aes(x = Theoretical, y = Sample)) +
  geom_point() +
  geom_abline(
   intercept = 0,
   slope = 1,
   color = "red",
   linetype = "dashed"
  ) +
  labs(title = "Q-Q Plot of Studentized Residuals", x = "Theoretical Quantiles", y = "Studentized Residuals"
```



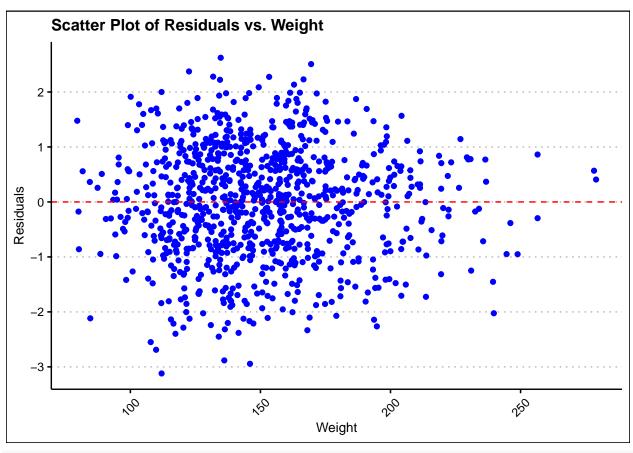


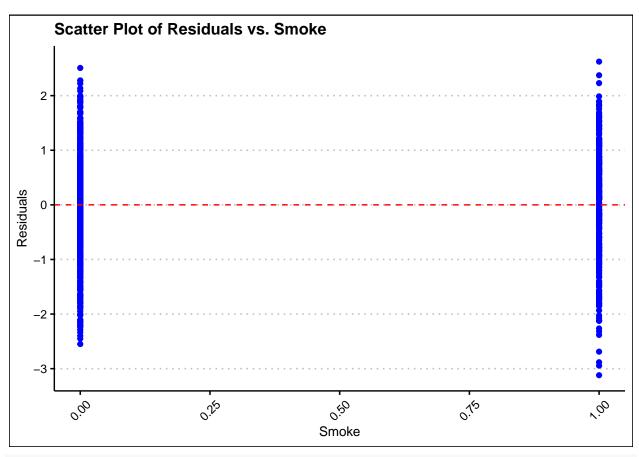
 $\# Variable\ Transformation$

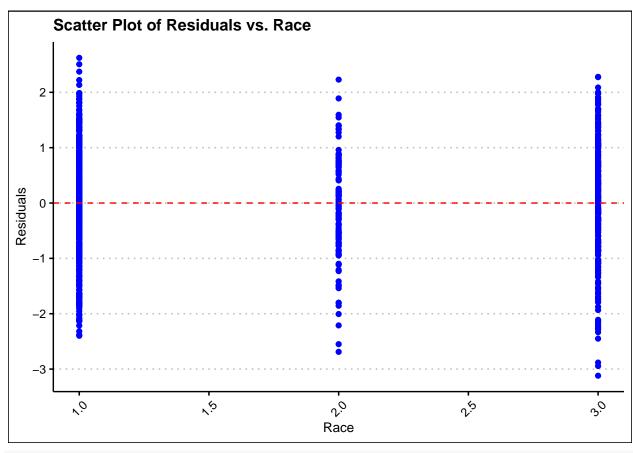
#Define whether the independent variable needs transformation
gladder(bwt_new\$bwt)

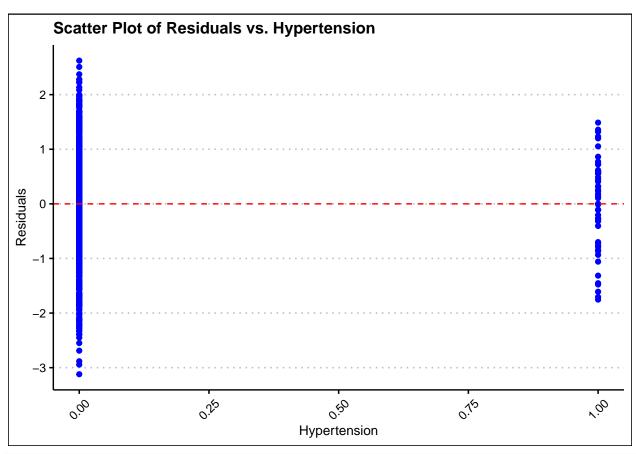


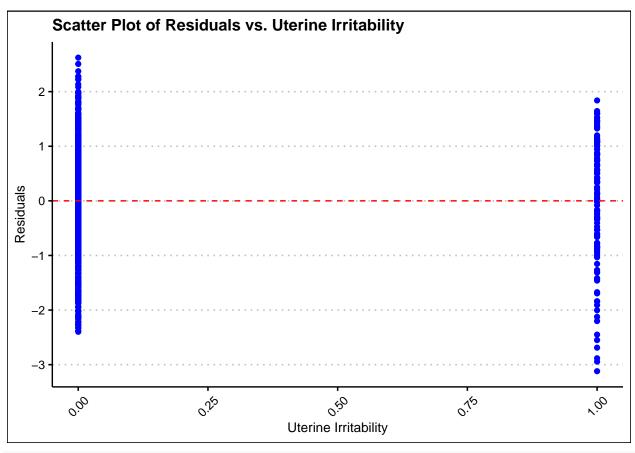
```
par(mfrow = c(2,2), mar=c(2,2,2,2))
plot(bwt_model4, cex.axis = 1, cex.lab = 1)
                                                            Q-Q Residuals
            Residuals vs Fitted
20
                                              က
10
0
                                              0
-10
               50
      45
                        55
                                 60
                                                    -3
                                                         -2
                                                              -1
                                                                    0
                                                                         1
                                                                              2
                                                                                   3
              Scale-Location
                                                         Residuals vs Leverage
1.5
1.0
                                                                                     0
0.5
                                              က
      45
               50
                        55
                                 60
                                                 0.00
                                                         0.01
                                                                 0.02
                                                                         0.03
                                                                                 0.04
sr3 = rstudent(bwt_model4)
#Scatter Plot of Studentized Residuals vs. Weight
ggplot(bwt_new, aes(x = lwt, y = sr3)) +
  geom_point(color = "blue") +
  geom_hline(yintercept = 0, color = "red", linetype = "dashed") +
  labs(title = "Scatter Plot of Residuals vs. Weight",
       x = "Weight",
       y = "Residuals")
```



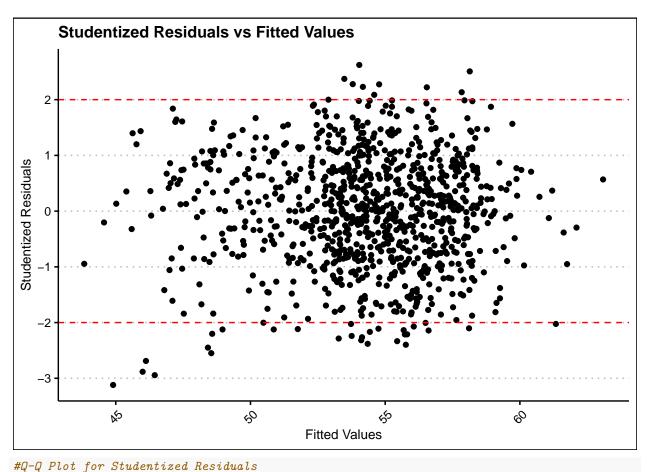








```
#Studentized Residuals vs. Fitted Value
ggplot(data = data.frame(Fitted = fitted(bwt_model4), Residuals = sr3),
          aes(x = Fitted, y = Residuals)) +
geom_point() +
geom_hline(yintercept = c(-2, 2), linetype = "dashed", color = "red") +
labs(title = "Studentized Residuals vs Fitted Values",
          x = "Fitted Values",
          y = "Studentized Residuals")
```



```
qq_data3 = data.frame(
   Theoretical = qqnorm(sr3, plot.it = FALSE)$x,
   Sample = qqnorm(sr3, plot.it = FALSE)$y
)

ggplot(qq_data3, aes(x = Theoretical, y = Sample)) +
   geom_point() +
   geom_abline(
   intercept = 0,
   slope = 1,
   color = "red",
   linetype = "dashed"
) +
   labs(title = "Q-Q Plot of Studentized Residuals", x = "Theoretical Quantiles", y = "Studentized Residuals"
```

