

Life Expectancy Prediction

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Abstract

We have implemented a Linear Regression model to predict the life expectancy of the human population with an adjusted R-squared value of , MSE of , and MAE of .

Introduction

The goal of this project is to build a Linear Regression model that can predict the life expectancy of the human population based on several factors such as the amount of alcohol consumption, average Body Mass Index (BMI), immunization of various vaccines among 1-year-olds such as Hepatitis B, Polio, and Diphtheria vaccines, and more, and also derive insights into what factors are significant in determining a higher or lower life expectancy of the human population.

Materials and Methods

Dataset

The data related to life expectancy and health factors for 193 countries is taken from the Global Health Observatory (GHO) data repository under the World Health Organization (WHO). Its corresponding economic data was collected from the United Nations website for a period of 16 years (2000-2015).

The dataset is available at <https://www.kaggle.com/datasets/kumarajarshi/life-expectancy-who>.

```
life = read.csv('Life Expectancy Data.csv') # Load Dataset
```

Life Expectancy Data.csv contains the following fields:

- **Country** - Country Observed.
- **Year** - Year Observed.
- **Status** - Developed or Developing status.
- **Life.expectancy** - Life Expectancy in age.
- **Adult.Mortality** - Adult Mortality Rates on both sexes (probability of dying between 15-60 years/1000 population).
- **infant.deaths** - Number of Infant Deaths per 1000 population.
- **Alcohol** - Alcohol recorded per capita (15+) consumption (in litres of pure alcohol).
- **percentage.expenditure** - Expenditure on health as a percentage of Gross Domestic Product per capita (%).
- **Hepatitis.B** - Hepatitis B (HepB) immunization coverage among 1-year-olds (%).
- **Measles** - Number of reported Measles cases per 1000 population.
- **BMI** - Average Body Mass Index of entire population.
- **under.five.deaths** - Number of under-five deaths per 1000 population.
- **Polio** - Polio (Pol3) immunization coverage among 1-year-olds (%).
- **Total.expenditure** - General government expenditure on health as a percentage of total government expenditure (%).
- **Diphtheria** - Diphtheria tetanus toxoid and pertussis (DTP3) immunization coverage among 1-year-olds (%).
- **HIV.AIDS** - Deaths per 1000 live births due to HIV/AIDS (0-4 years).
- **GDP** - Gross Domestic Product per capita (in USD).
- **Population** - Population of the country.
- **thinness..1.19.years** - Prevalence of thinness among children and adolescents for Age 10 to 19 (%).
- **thinness.5.9.years** - Prevalence of thinness among children for Age 5 to 9 (%).
- **Income.composition.of.resources** - Human Development Index in terms of income composition of resources (index ranging from 0 to 1).
- **Schooling** - Number of years of Schooling (years).

In total, there are 2938 observations of 22 variables with 20 of them being numerical and 2 categorical (Country and Status).

We will be using **Life.expectancy** to predict the life expectancy of the human population with the given dependent variables in the dataset.

Clean Data

We will drop any observation that does not contain any value in any of its columns.

```
life = na.omit(life)
```

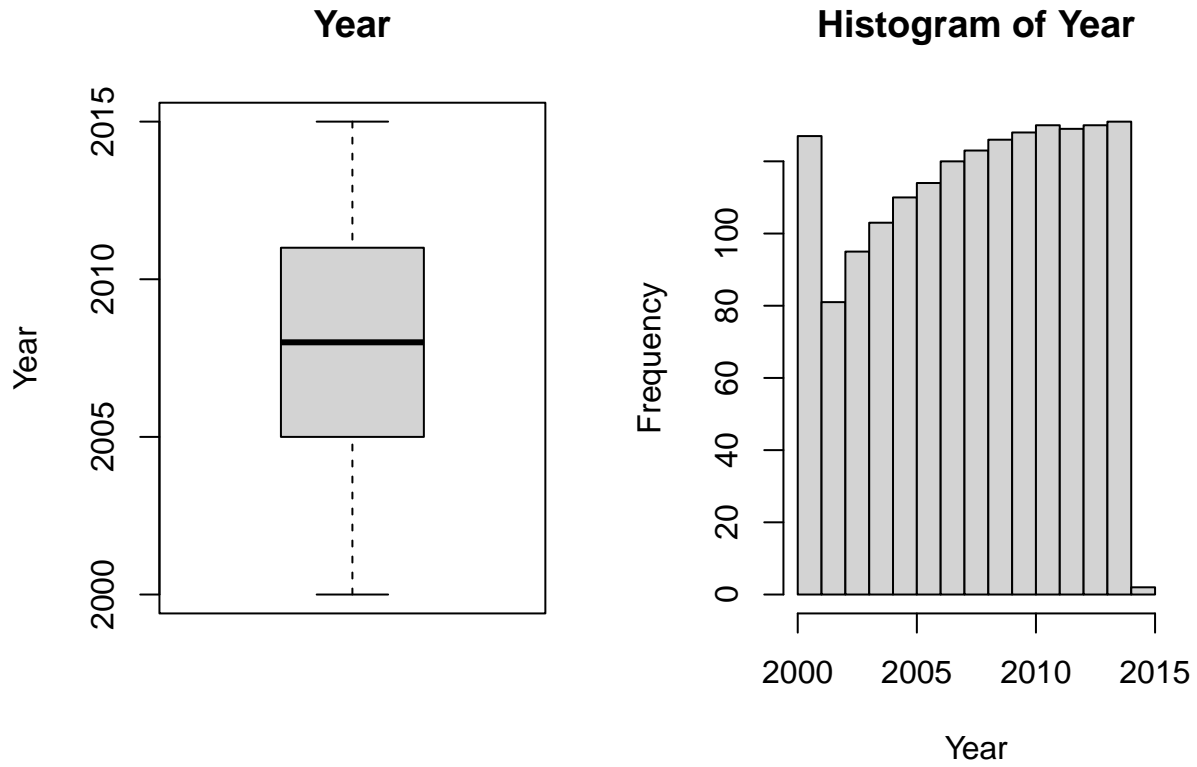
This shrinks our dataset to 1649 observations.

Data Exploration

```
summary(life)
```

```
##      Country          Year      Status      Life.expectancy
## Length:1649      Min.    :2000      Length:1649      Min.    :44.0
## Class :character  1st Qu.:2005      Class :character  1st Qu.:64.4
## Mode  :character  Median  :2008      Mode  :character  Median :71.7
##                               Mean    :2008      Mean    :69.3
##                               3rd Qu.:2011      3rd Qu.:75.0
##                               Max.    :2015      Max.    :89.0
## Adult.Mortality infant.deaths      Alcohol      percentage.expenditure
## Min.    : 1.0      Min.    : 0.00      Min.    : 0.010      Min.    : 0.00
## 1st Qu.: 77.0      1st Qu.: 1.00      1st Qu.: 0.810      1st Qu.: 37.44
## Median :148.0      Median : 3.00      Median : 3.790      Median : 145.10
## Mean    :168.2      Mean    : 32.55      Mean    : 4.533      Mean    : 698.97
## 3rd Qu.:227.0      3rd Qu.: 22.00      3rd Qu.: 7.340      3rd Qu.: 509.39
## Max.    :723.0      Max.    :1600.00      Max.    :17.870      Max.    :18961.35
## Hepatitis.B      Measles      BMI      under.five.deaths
## Min.    : 2.00      Min.    : 0      Min.    : 2.00      Min.    : 0.00
## 1st Qu.:74.00      1st Qu.: 0      1st Qu.:19.50      1st Qu.: 1.00
## Median :89.00      Median : 15      Median :43.70      Median : 4.00
## Mean    :79.22      Mean    : 2224      Mean    :38.13      Mean    : 44.22
## 3rd Qu.:96.00      3rd Qu.: 373      3rd Qu.:55.80      3rd Qu.: 29.00
## Max.    :99.00      Max.    :131441      Max.    :77.10      Max.    :2100.00
## Polio      Total.expenditure      Diphtheria      HIV.AIDS
## Min.    : 3.00      Min.    : 0.740      Min.    : 2.00      Min.    : 0.100
## 1st Qu.:81.00      1st Qu.: 4.410      1st Qu.:82.00      1st Qu.: 0.100
## Median :93.00      Median : 5.840      Median :92.00      Median : 0.100
## Mean    :83.56      Mean    : 5.956      Mean    :84.16      Mean    : 1.984
## 3rd Qu.:97.00      3rd Qu.: 7.470      3rd Qu.:97.00      3rd Qu.: 0.700
## Max.    :99.00      Max.    :14.390      Max.    :99.00      Max.    :50.600
## GDP      Population      thinness..1.19.years
## Min.    : 1.68      Min.    :3.400e+01      Min.    : 0.100
## 1st Qu.: 462.15      1st Qu.:1.919e+05      1st Qu.: 1.600
## Median : 1592.57      Median :1.420e+06      Median : 3.000
## Mean    : 5566.03      Mean    :1.465e+07      Mean    : 4.851
## 3rd Qu.: 4718.51      3rd Qu.:7.659e+06      3rd Qu.: 7.100
## Max.    :119172.74      Max.    :1.294e+09      Max.    :27.200
## thinness.5.9.years      Income.composition.of.resources      Schooling
## Min.    : 0.100      Min.    :0.0000      Min.    : 4.20
## 1st Qu.: 1.700      1st Qu.:0.5090      1st Qu.:10.30
## Median : 3.200      Median :0.6730      Median :12.30
## Mean    : 4.908      Mean    :0.6316      Mean    :12.12
## 3rd Qu.: 7.100      3rd Qu.:0.7510      3rd Qu.:14.00
## Max.    :28.200      Max.    :0.9360      Max.    :20.70
```

```
par(mfrow = c(1, 2))
boxplot(life$Year, main = 'Year', ylab = 'Year')
hist(life$Year, main = 'Histogram of Year', xlab = 'Year')
```



```

par(mfrow = c(1, 2))
boxplot(life$Life.expectancy, main = 'Life Expectancy', ylab = 'Age')
hist(life$Life.expectancy, main = 'Histogram of Life Expectancy', xlab = 'Age')

```

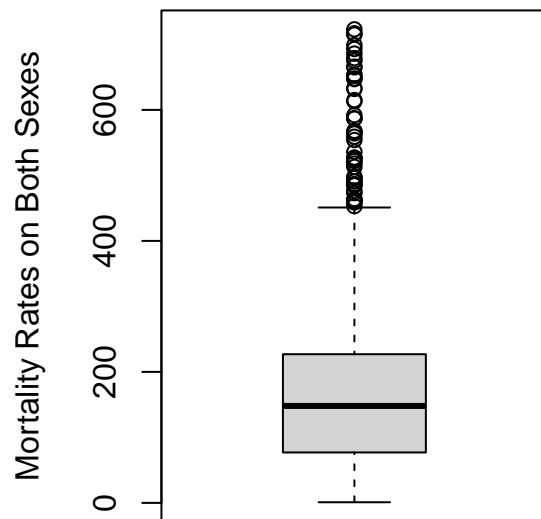


```

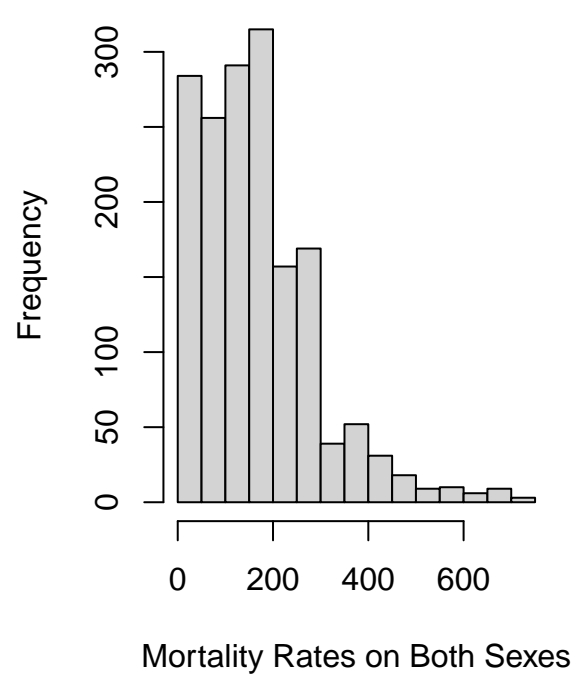
par(mfrow = c(1, 2))
boxplot(life$Adult.Mortality, main = 'Adult Mortality Rate', ylab = 'Mortality Rates on Both Sexes')
hist(
  life$Adult.Mortality,
  main = 'Histogram of Adult Mortality Rate',
  xlab = 'Mortality Rates on Both Sexes',
  cex.main = 0.9
)

```

Adult Mortality Rate



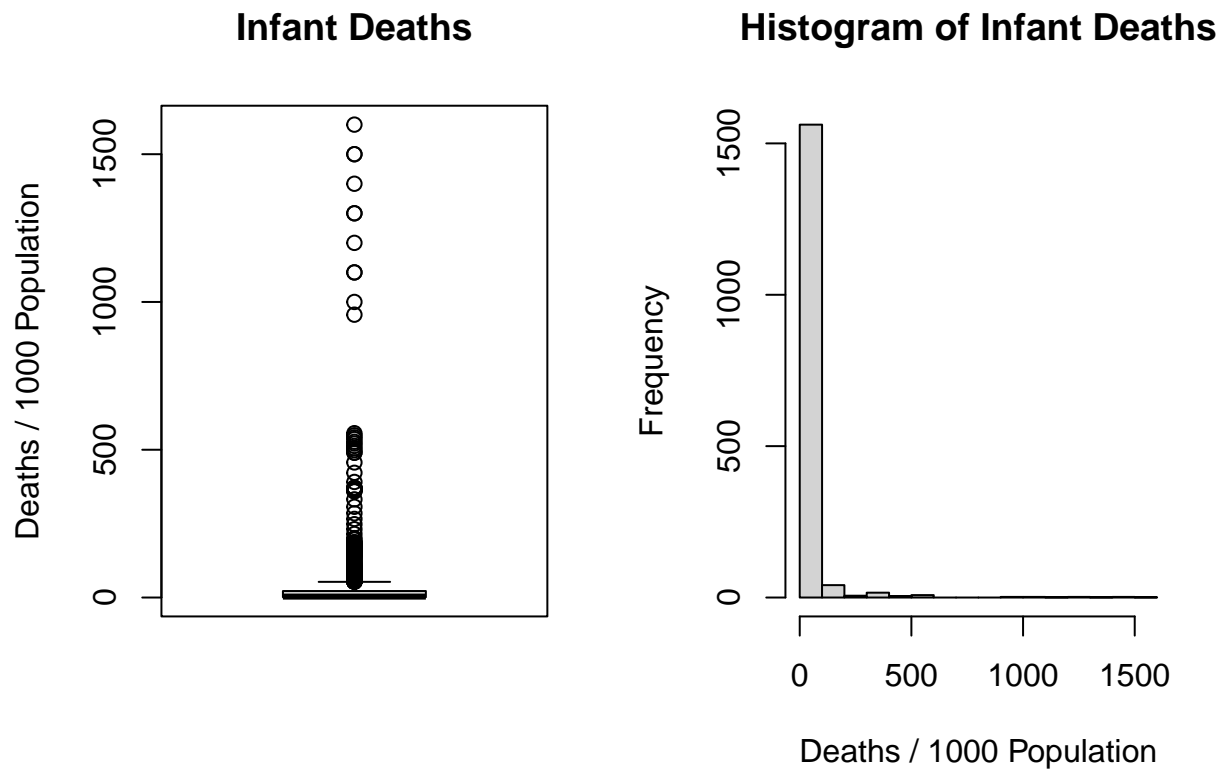
Histogram of Adult Mortality Rate



```

par(mfrow = c(1, 2))
boxplot(life$infant.deaths, main = 'Infant Deaths', ylab = 'Deaths / 1000 Population')
hist(life$infant.deaths, main = 'Histogram of Infant Deaths', xlab = 'Deaths / 1000 Population')

```

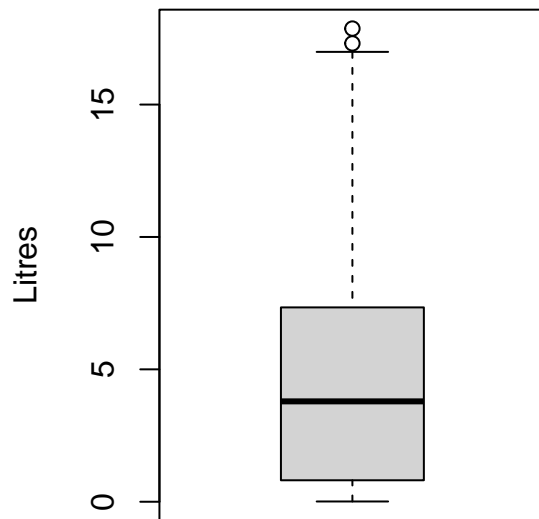


```

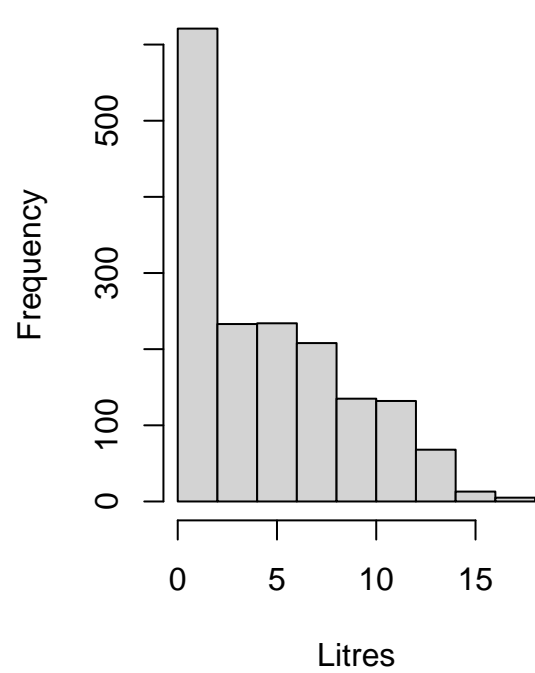
par(mfrow = c(1, 2))
boxplot(life$Alcohol, main = 'Alcohol Consumption', ylab = 'Litres')
hist(life$Alcohol,
     main = 'Histogram of Alcohol Consumption',
     xlab = 'Litres',
     cex.main = 0.9)

```

Alcohol Consumption



Histogram of Alcohol Consumption

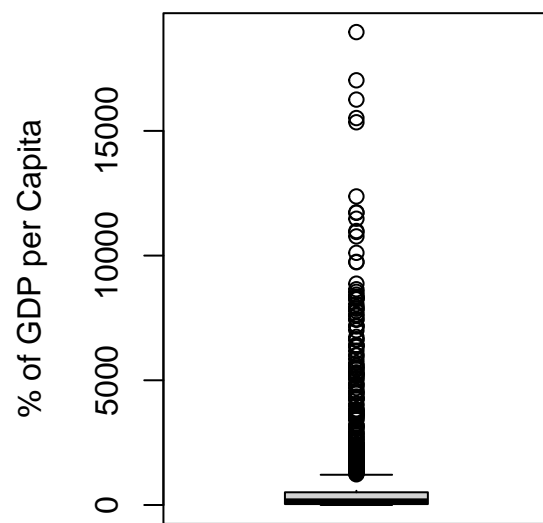



```

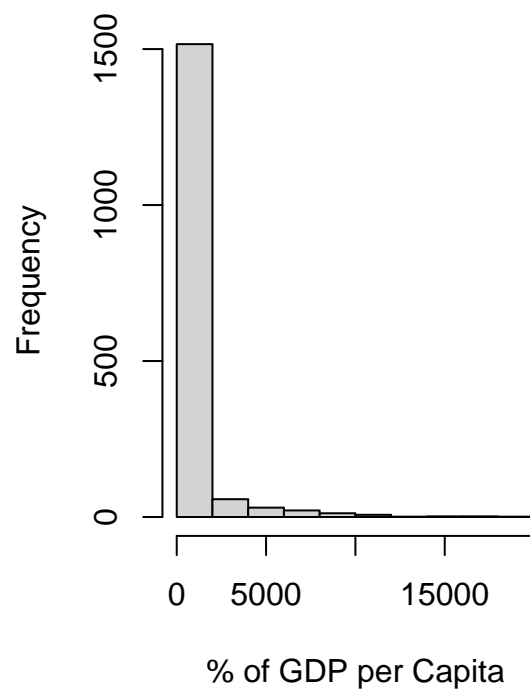
par(mfrow = c(1, 2))
boxplot(life$percentage.expenditure,
        main = 'Health Expenditure',
        ylab = '% of GDP per Capita')
hist(
  life$percentage.expenditure,
  main = 'Histogram of Health Expenditure',
  xlab = '% of GDP per Capita',
  cex.main = 0.9
)

```

Health Expenditure



Histogram of Health Expenditure

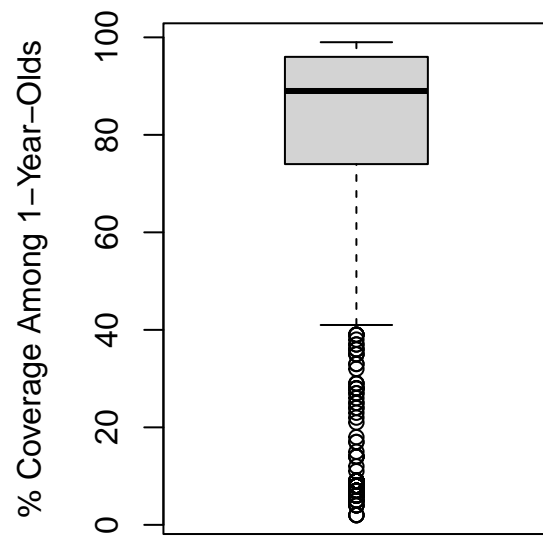


```

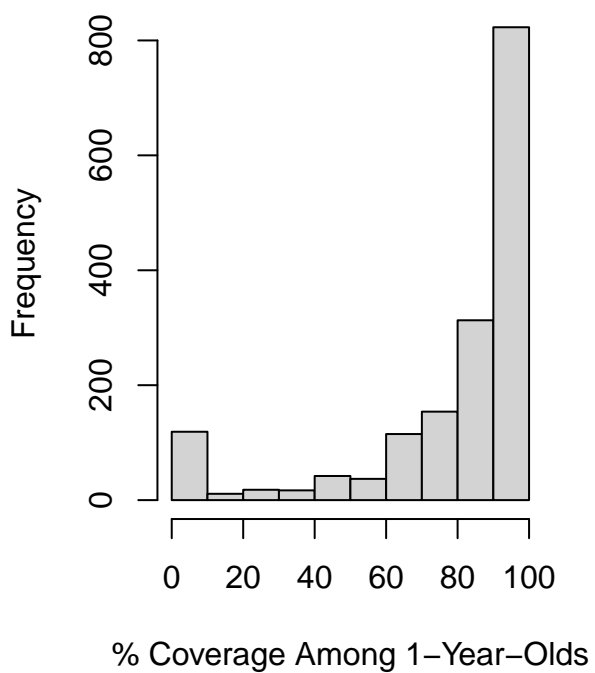
par(mfrow = c(1, 2))
boxplot(life$Hepatitis.B,
        main = 'Hepatitis B (HepB) Immunization',
        ylab = '% Coverage Among 1-Year-Olds',
        cex.main = 0.9)
hist(life$Hepatitis.B,
     main = 'Histogram of HepB Immunization',
     xlab = '% Coverage Among 1-Year-Olds',
     cex.main = 0.9)

```

Hepatitis B (HepB) Immunization



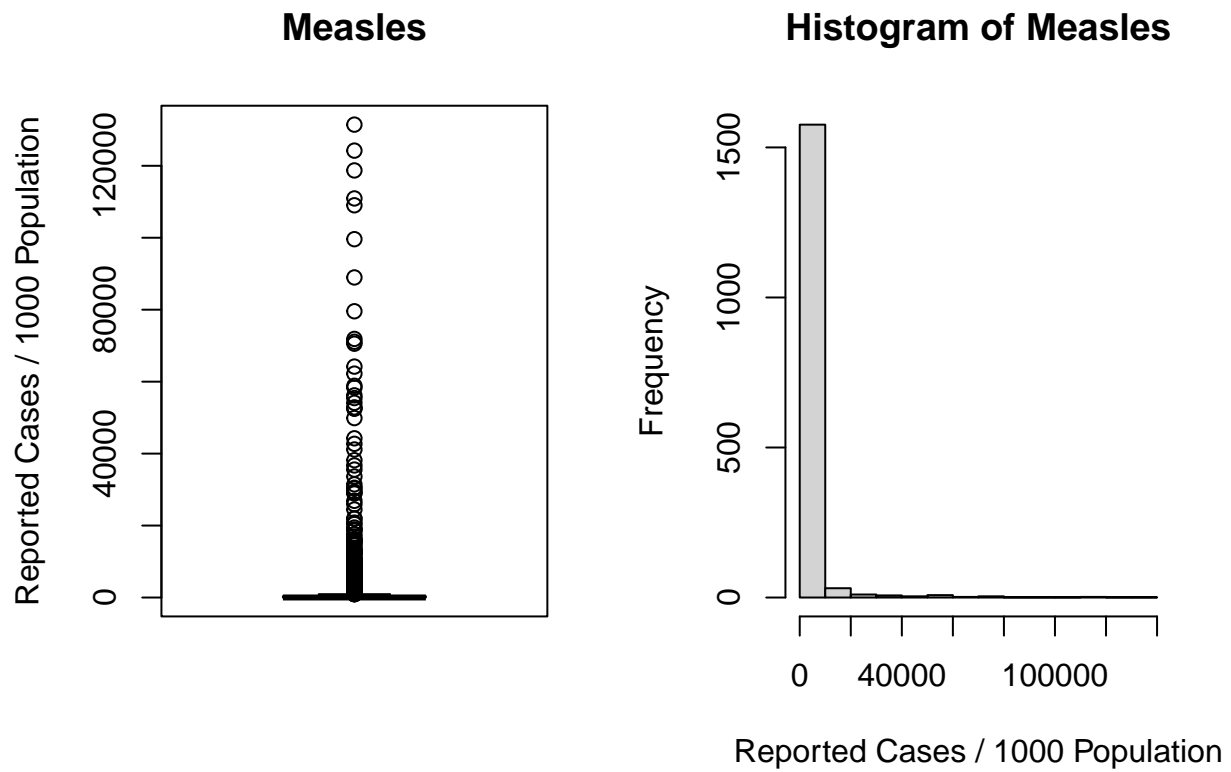
Histogram of HepB Immunization



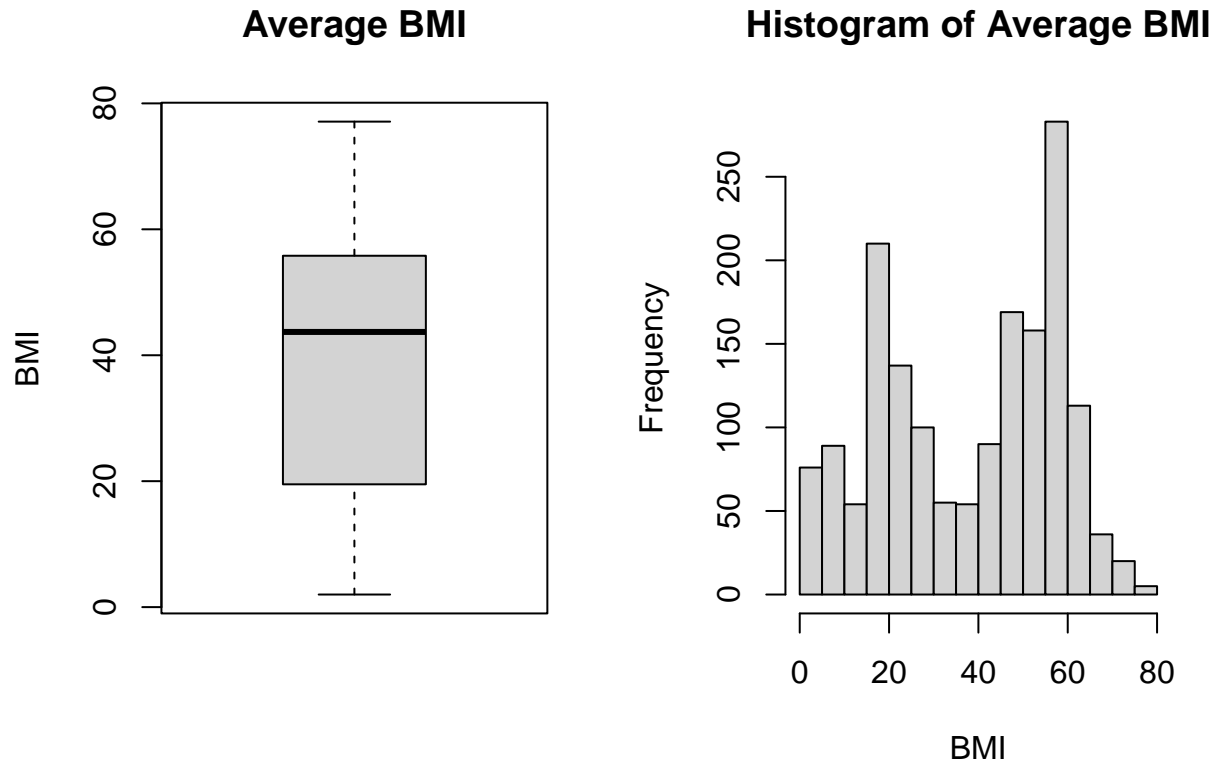
```

par(mfrow = c(1, 2))
boxplot(life$Measles, main = 'Measles', ylab = 'Reported Cases / 1000 Population')
hist(life$Measles, main = 'Histogram of Measles', xlab = 'Reported Cases / 1000 Population')

```



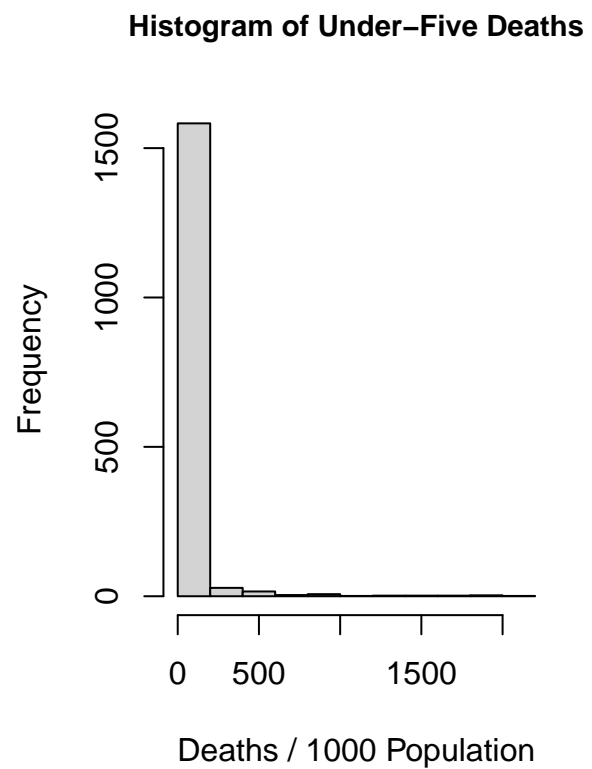
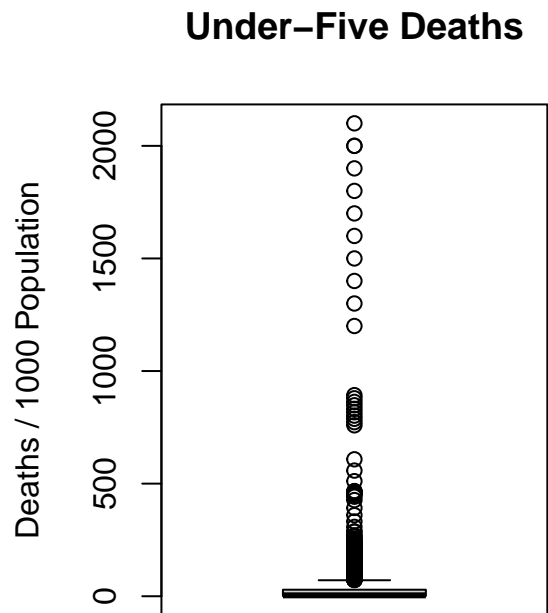
```
par(mfrow = c(1, 2))
boxplot(life$BMI, main = 'Average BMI', ylab = 'BMI')
hist(life$BMI, main = 'Histogram of Average BMI', xlab = 'BMI')
```



```

par(mfrow = c(1, 2))
boxplot(life$under.five.deaths, main = 'Under-Five Deaths', ylab = 'Deaths / 1000 Population')
hist(
  life$under.five.deaths,
  main = 'Histogram of Under-Five Deaths',
  xlab = 'Deaths / 1000 Population',
  cex.main = 0.9
)

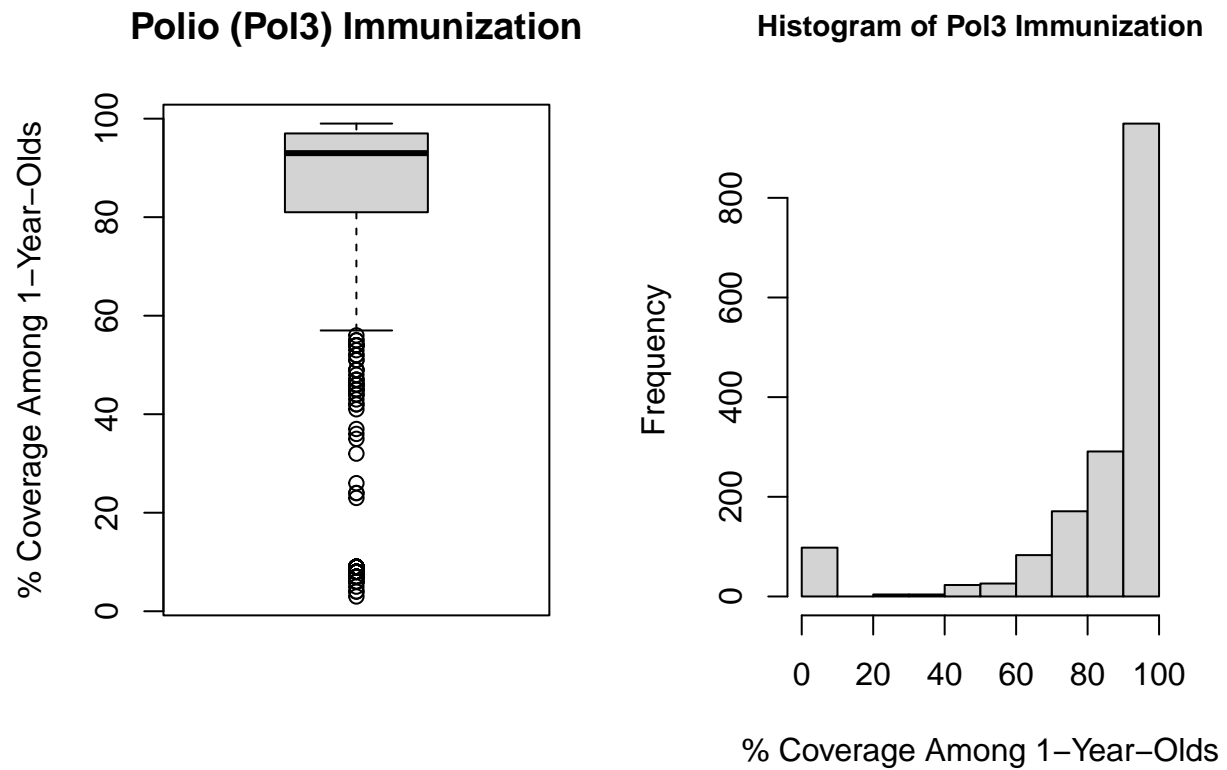
```



```

par(mfrow = c(1, 2))
boxplot(life$Polio, main = 'Polio (Pol3) Immunization', ylab = '% Coverage Among 1-Year-Olds')
hist(life$Polio,
     main = 'Histogram of Pol3 Immunization',
     xlab = '% Coverage Among 1-Year-Olds',
     cex.main = 0.9)

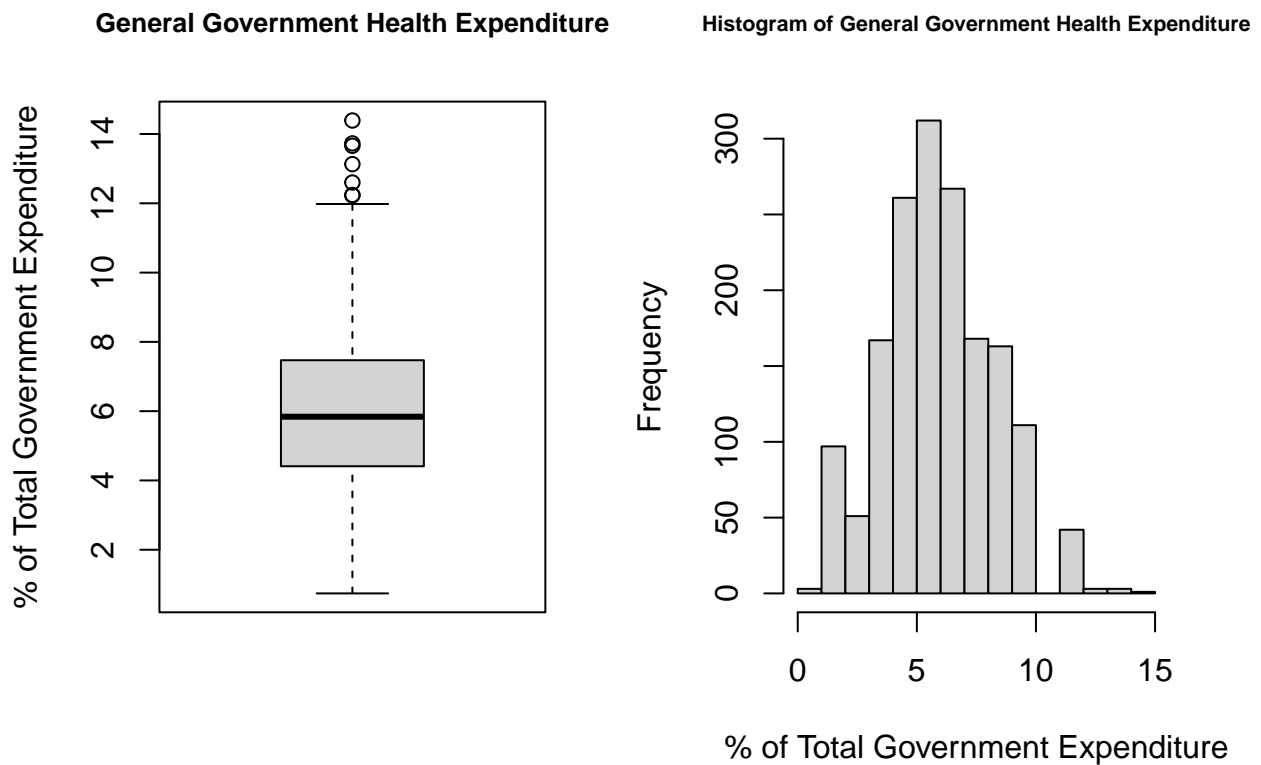
```



```

par(mfrow = c(1, 2))
boxplot(
  life$Total.expenditure,
  main = 'General Government Health Expenditure',
  ylab =
    '% of Total Government Expenditure',
  cex.main = 0.8
)
hist(
  life$Total.expenditure,
  main = 'Histogram of General Government Health Expenditure',
  xlab =
    '% of Total Government Expenditure',
  cex.main = 0.7
)

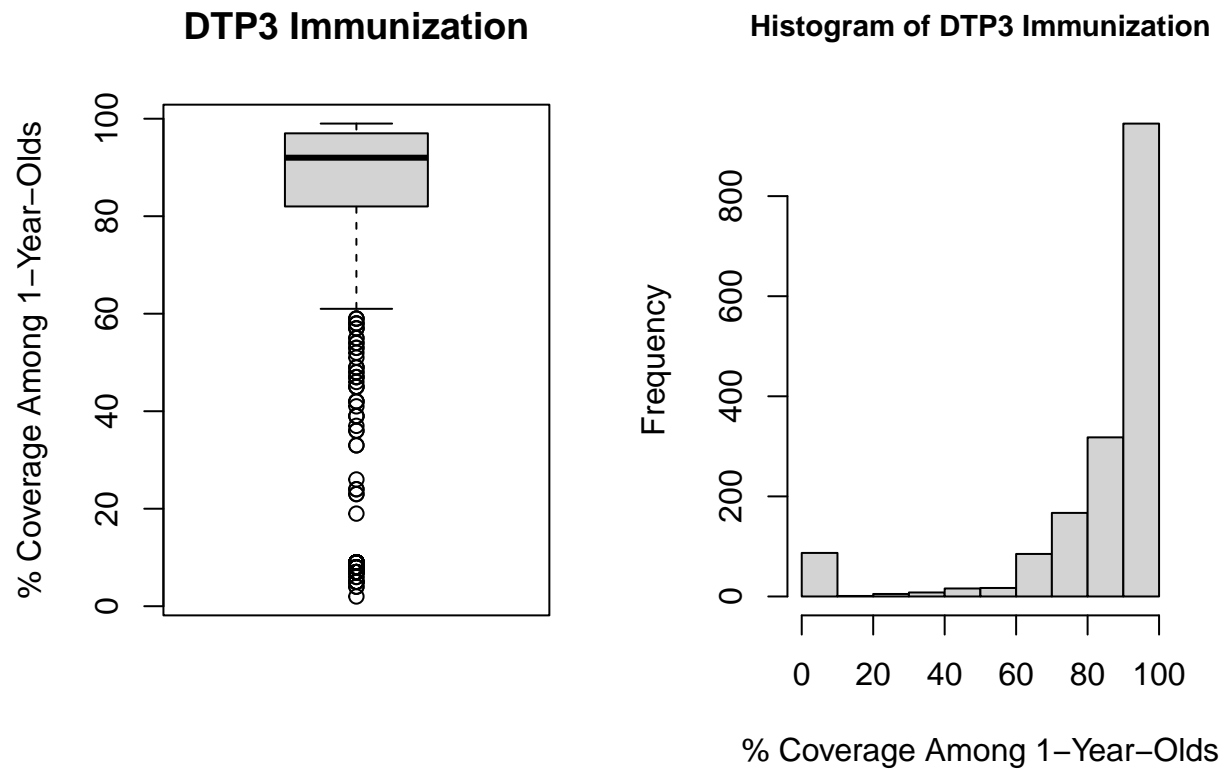
```



```

par(mfrow = c(1, 2))
boxplot(life$Diphtheria, main = 'DTP3 Immunization', ylab = '% Coverage Among 1-Year-Olds')
hist(life$Diphtheria,
     main = 'Histogram of DTP3 Immunization',
     xlab = '% Coverage Among 1-Year-Olds',
     cex.main = 0.9)

```

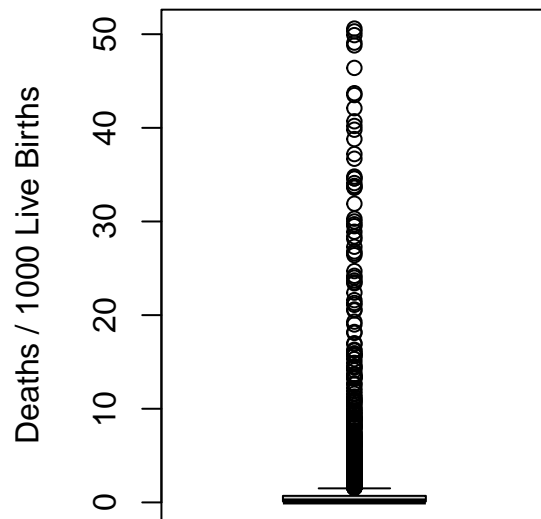



```

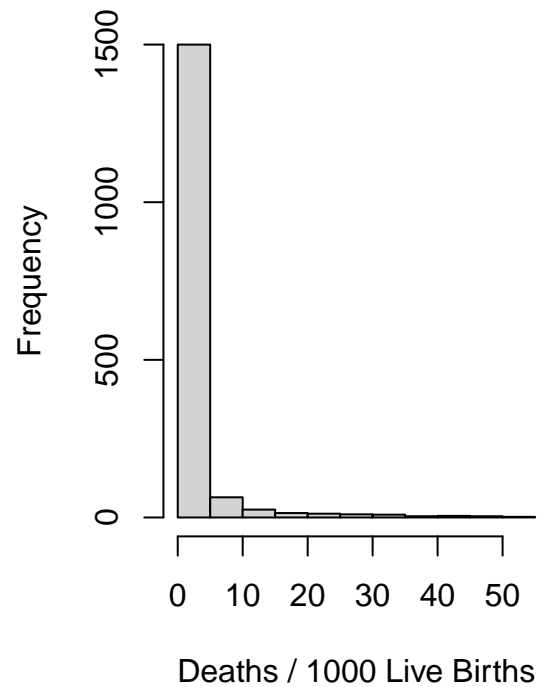
par(mfrow = c(1, 2))
boxplot(life$HIV.AIDS, main = 'HIV/AIDS (0-4 Years)', ylab = 'Deaths / 1000 Live Births')
hist(life$HIV.AIDS,
     main = 'Histogram of HIV/AIDS (0-4 Years)',
     xlab = 'Deaths / 1000 Live Births',
     cex.main = 0.9)

```

HIV/AIDS (0-4 Years)



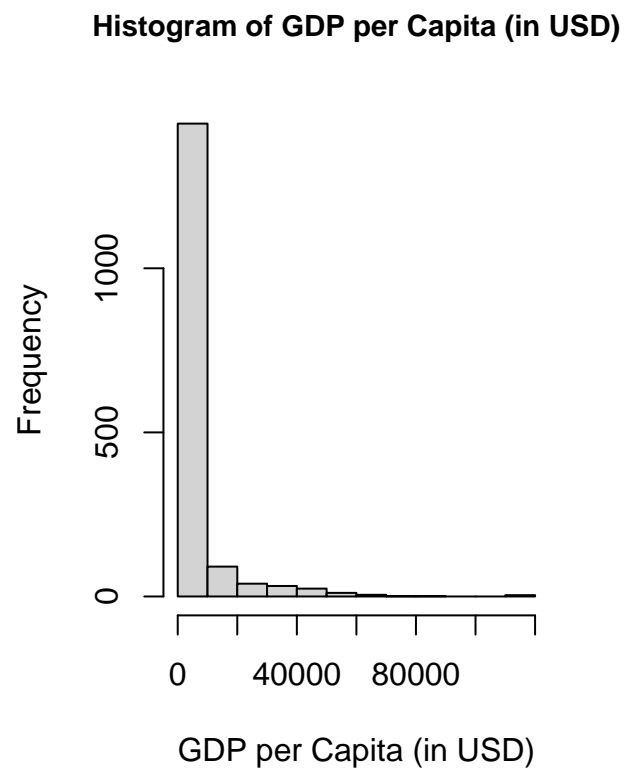
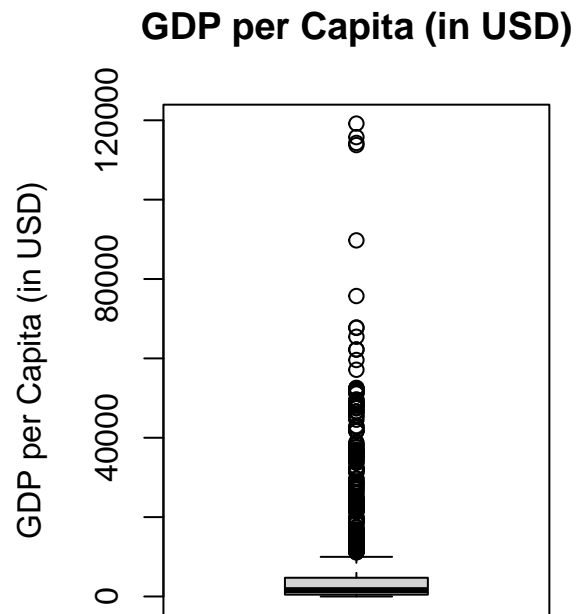
Histogram of HIV/AIDS (0-4 Years)



```

par(mfrow = c(1, 2))
boxplot(life$GDP, main = 'GDP per Capita (in USD)', ylab = 'GDP per Capita (in USD)')
hist(life$GDP,
     main = 'Histogram of GDP per Capita (in USD)',
     xlab = 'GDP per Capita (in USD)',
     cex.main = 0.9)

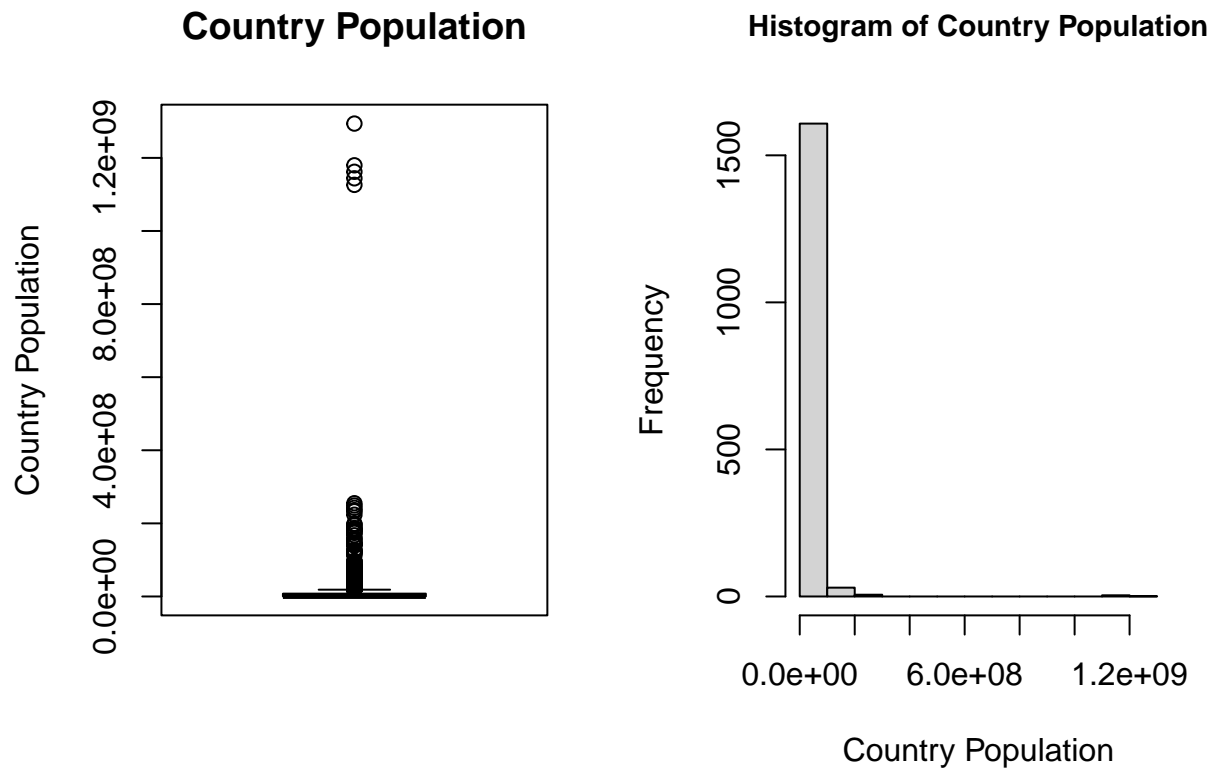
```



```

par(mfrow = c(1, 2))
boxplot(life$Population, main = 'Country Population', ylab = 'Country Population')
hist(life$Population,
     main = 'Histogram of Country Population',
     xlab = 'Country Population',
     cex.main = 0.9)

```

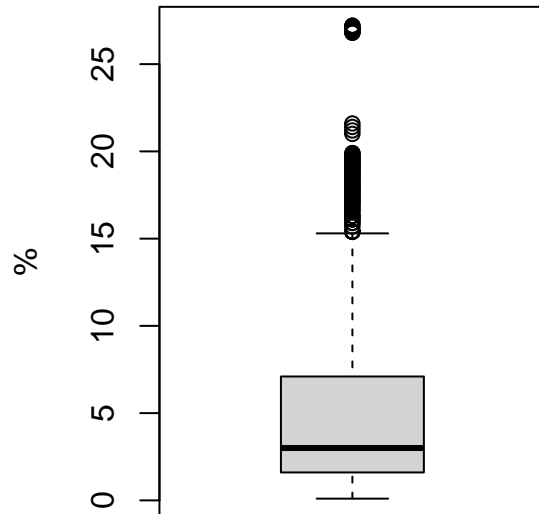


```

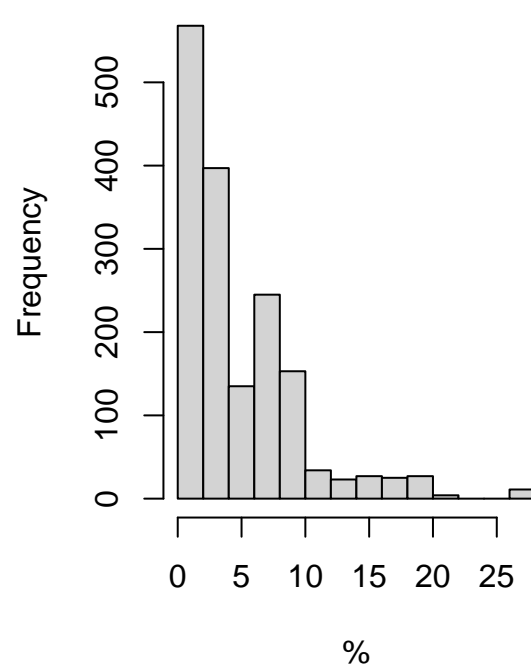
par(mfrow = c(1, 2))
boxplot(
  life$thinness..1.19.years,
  main = 'Prevalence of Thinness (10-19 Years)',
  ylab = '%',
  cex.main = 0.9
)
hist(
  life$thinness..1.19.years,
  main = 'Histogram of Prevalence of Thinness (10-19 Years)',
  xlab = '%',
  cex.main = 0.7
)

```

Prevalence of Thinness (10–19 Years)



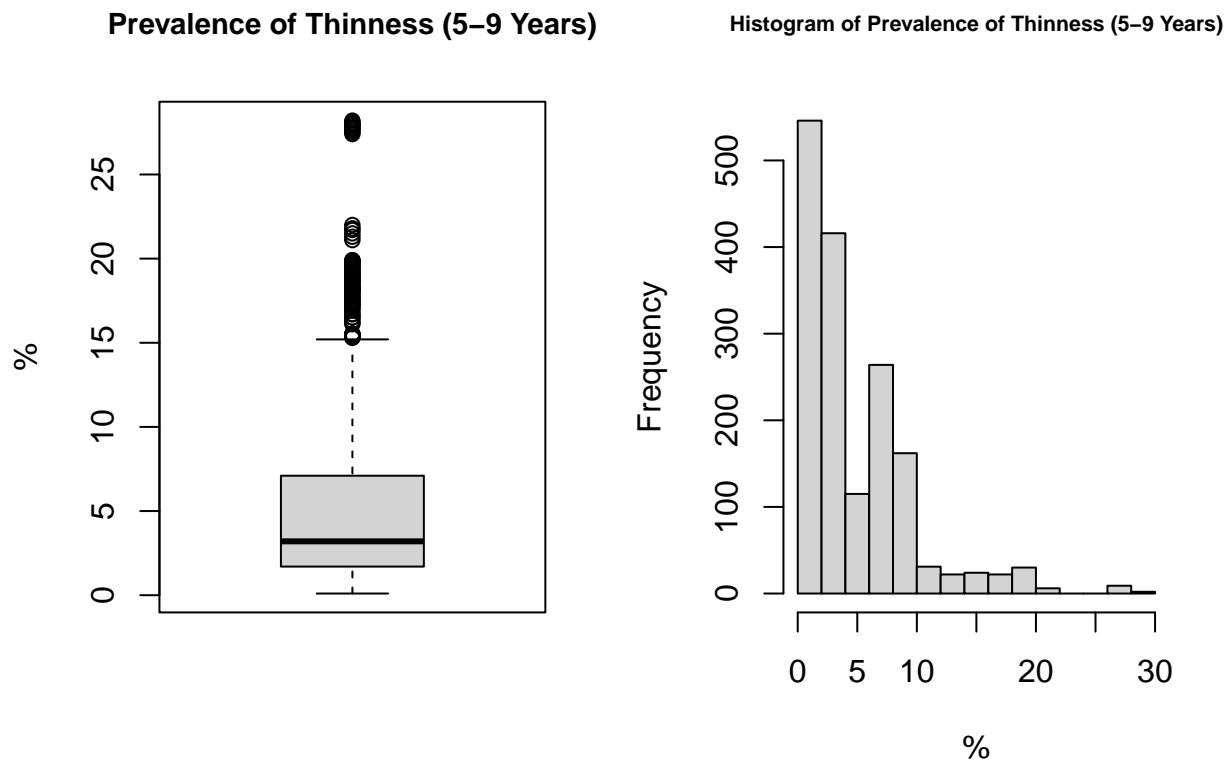
Histogram of Prevalence of Thinness (10–19 Years)



```

par(mfrow = c(1, 2))
boxplot(
  life$thinness.5.9.years,
  main = 'Prevalence of Thinness (5-9 Years)',
  ylab = '%',
  cex.main = 0.9
)
hist(
  life$thinness.5.9.years,
  main = 'Histogram of Prevalence of Thinness (5-9 Years)',
  xlab =
    '%',
  cex.main = 0.7
)

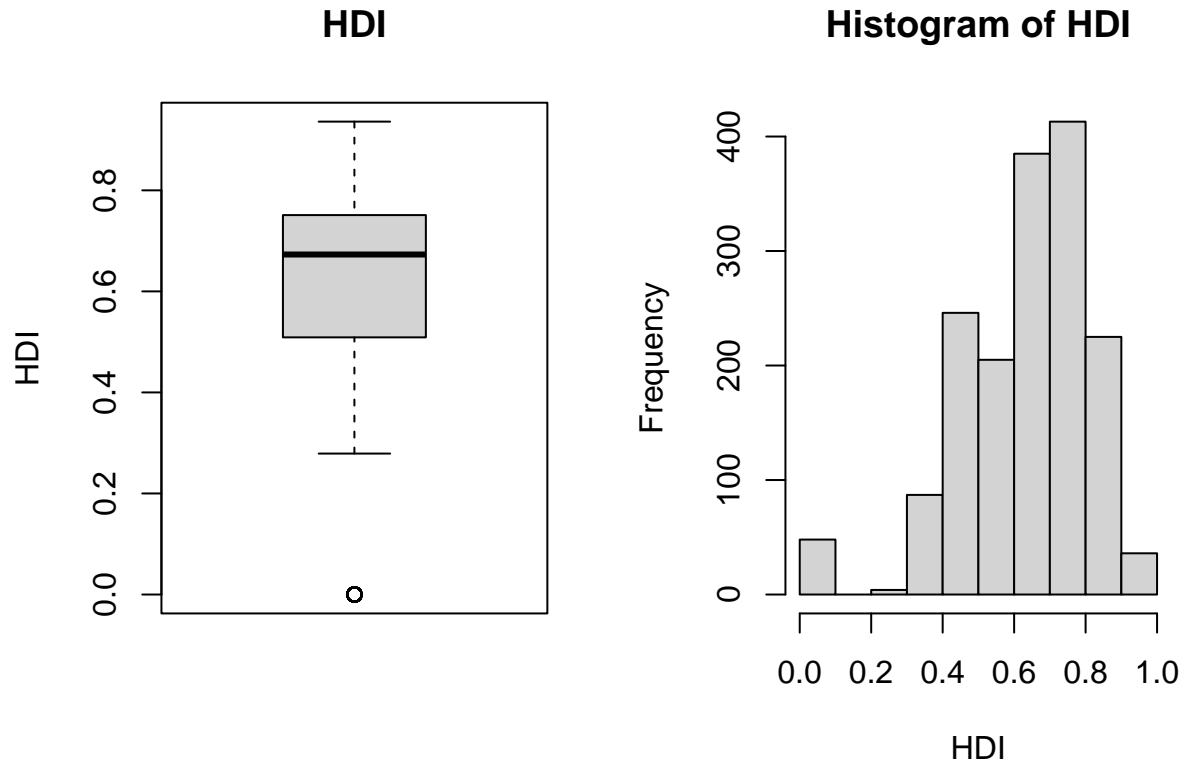
```



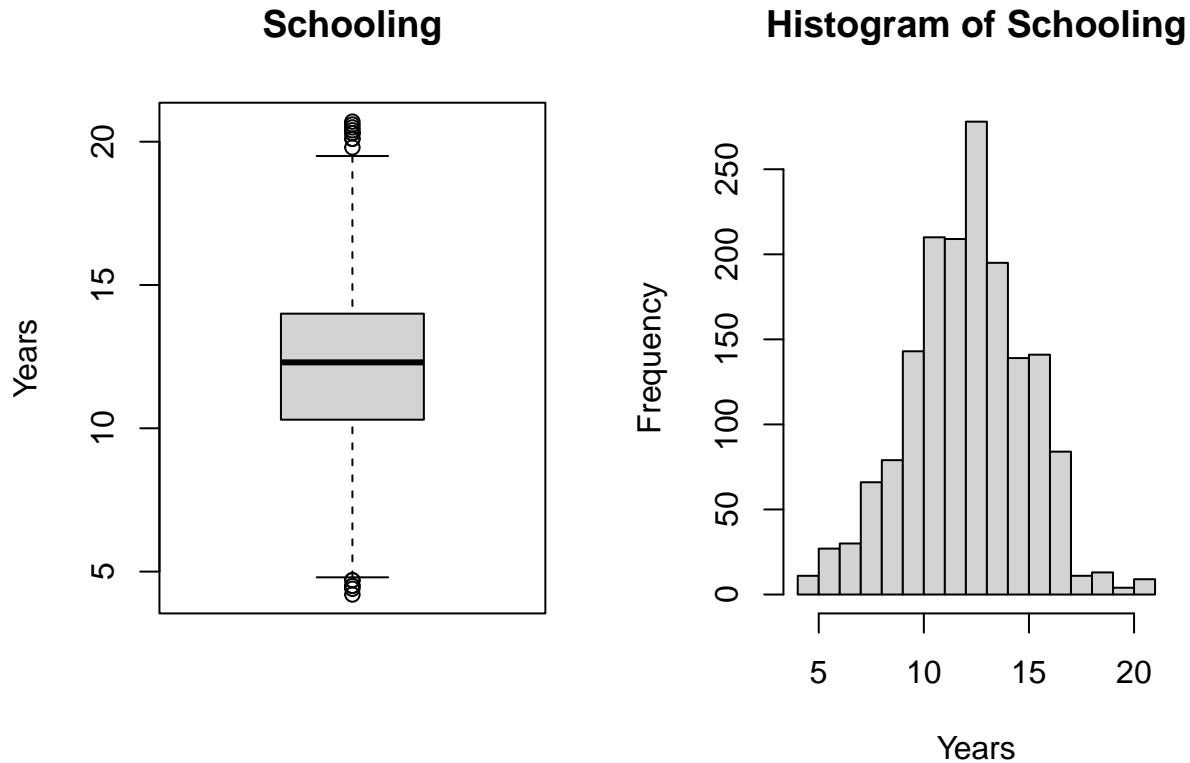
```

par(mfrow = c(1, 2))
boxplot(life$Income.composition.of.resources, main = 'HDI', ylab='HDI')
hist(life$Income.composition.of.resources, main = 'Histogram of HDI', xlab='HDI')

```



```
par(mfrow = c(1, 2))
boxplot(life$Schooling, main = 'Schooling', ylab = 'Years')
hist(life$Schooling, main = 'Histogram of Schooling', xlab = 'Years')
```



Feature Selection

We will be removing some of the variables for building the model due to the reasons mentioned below:

- Country** - Contains too many levels with no additional information to predict **Life expectancy**.
- Year** - Contains time series data with no additional information to predict **Life expectancy**.

```
life = life[, !(names(life) %in% c('Country', 'Year'))]
```

We will be mutating **Hepatitis.B**, **Polio** and **Diphtheria** for building the model since their range between the minimum value and the 1st Quartile is too wide. We will be mutating their values into 2 categorical values: '<90% Covered' and '>=90% Covered'.

```
life$Hepatitis.B = ifelse(life$Hepatitis.B < 90, '<90% Covered', '>=90% Covered')
life$Polio = ifelse(life$Polio < 90, '<90% Covered', '>=90% Covered')
life$Diphtheria = ifelse(life$Diphtheria < 90, '<90% Covered', '>=90% Covered')
```

This leaves us with 1649 observations of 20 variables with 16 of them being numerical and 4 categorical (**Status**, **Hepatitis.B**, **Polio** and **Diphtheria**).

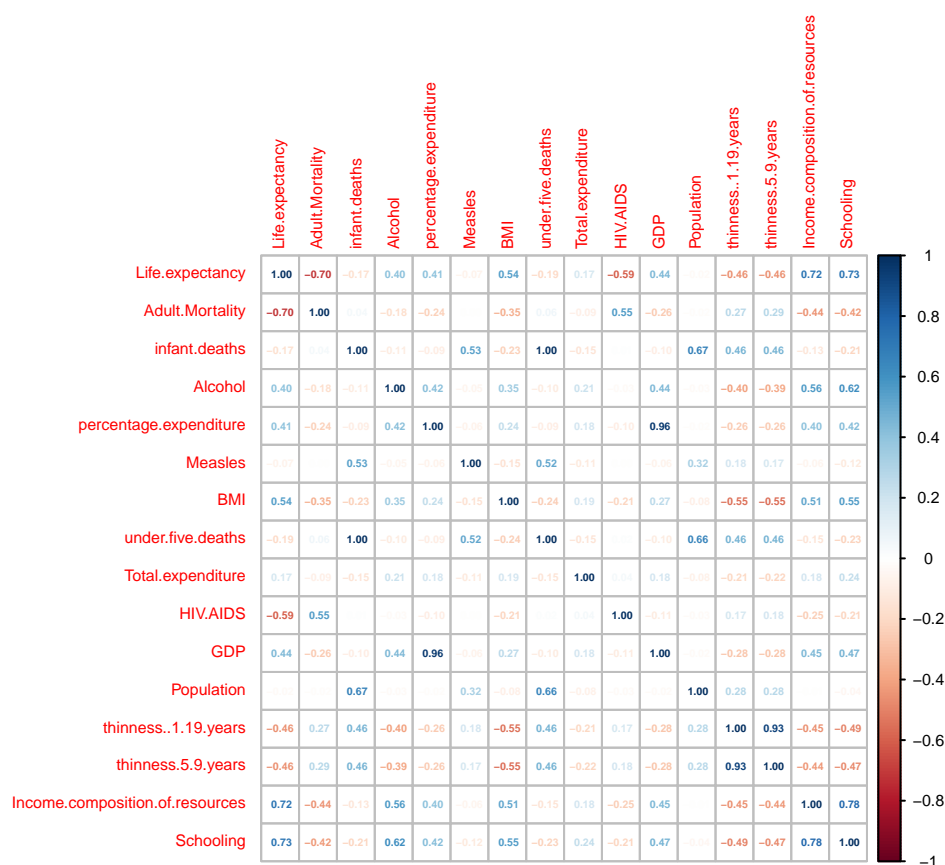
```
summary(life)
```

```
##      Status      Life.expectancy Adult.Mortality infant.deaths
## Length:1649      Min.   :44.0      Min.   : 1.0      Min.   : 0.00
## Class :character  1st Qu.:64.4      1st Qu.: 77.0      1st Qu.: 1.00
## Mode  :character  Median :71.7      Median :148.0      Median : 3.00
##                               Mean  :69.3      Mean  :168.2      Mean   : 32.55
##                               3rd Qu.:75.0      3rd Qu.:227.0      3rd Qu.: 22.00
##                               Max.   :89.0      Max.   :723.0      Max.   :1600.00
##      Alcohol      percentage.expenditure Hepatitis.B      Measles
## Min.   : 0.010      Min.   : 0.00      Length:1649      Min.   : 0
## 1st Qu.: 0.810      1st Qu.: 37.44      Class :character  1st Qu.: 0
## Median : 3.790      Median : 145.10      Mode  :character  Median : 15
## Mean   : 4.533      Mean   : 698.97                               Mean   : 2224
## 3rd Qu.: 7.340      3rd Qu.: 509.39                               3rd Qu.: 373
## Max.   :17.870      Max.   :18961.35                               Max.   :131441
##      BMI      under.five.deaths      Polio      Total.expenditure
## Min.   : 2.00      Min.   : 0.00      Length:1649      Min.   : 0.740
## 1st Qu.:19.50      1st Qu.: 1.00      Class :character  1st Qu.: 4.410
## Median :43.70      Median : 4.00      Mode  :character  Median : 5.840
## Mean   :38.13      Mean   : 44.22                               Mean   : 5.956
## 3rd Qu.:55.80      3rd Qu.: 29.00                               3rd Qu.: 7.470
## Max.   :77.10      Max.   :2100.00                               Max.   :14.390
##      Diphtheria      HIV.AIDS      GDP      Population
## Length:1649      Min.   : 0.100      Min.   : 1.68      Min.   :3.400e+01
## Class :character  1st Qu.: 0.100      1st Qu.: 462.15      1st Qu.:1.919e+05
## Mode  :character  Median : 0.100      Median : 1592.57      Median :1.420e+06
##                               Mean   : 1.984      Mean   : 5566.03      Mean   :1.465e+07
##                               3rd Qu.: 0.700      3rd Qu.: 4718.51      3rd Qu.:7.659e+06
##                               Max.   :50.600      Max.   :119172.74      Max.   :1.294e+09
## thinness..1.19.years thinness.5.9.years Income.composition.of.resources
## Min.   : 0.100      Min.   : 0.100      Min.   :0.0000
## 1st Qu.: 1.600      1st Qu.: 1.700      1st Qu.:0.5090
## Median : 3.000      Median : 3.200      Median :0.6730
## Mean   : 4.851      Mean   : 4.908      Mean   :0.6316
## 3rd Qu.: 7.100      3rd Qu.: 7.100      3rd Qu.:0.7510
## Max.   :27.200      Max.   :28.200      Max.   :0.9360
##      Schooling
## Min.   : 4.20
## 1st Qu.:10.30
## Median :12.30
## Mean   :12.12
## 3rd Qu.:14.00
## Max.   :20.70
```


Correlations

Since the number of variables is moderately large, we will plot the correlation plot of the dataset rather than looking at the correlation matrix by itself. The color and its shade easily guide us which 2 variables are correlated.

```
life_nums = unlist(lapply(life, is.numeric), use.names = FALSE)
corrplot(
  cor(life[, life_nums]),
  method = 'number',
  tl.cex = 0.5,
  number.cex = 0.33,
  cl.cex = 0.5
)
```



There are a few takeaways from this correlation plot:

- Life expectancy has a strong positive correlation with Income.composition.of.resources and Schooling.
- Life expectancy has a negative correlation with Adult Mortality, which makes sense since if the mortality rate of adult is high, then obviously the life expectancy will be low.
- Life expectancy has a very weak correlation with Measles and Population.
- There is a very strong correlation between infant.deaths and under.five.deaths, indicating multicollinearity between them. Therefore, we will remove under.five.deaths for building the model.

```
life = life[, !(names(life) %in% c('under.five.deaths'))]
```

Model Building

We will now build a Linear Regression Model using all the remaining variables to predict the life expectancy of the human population.

```
lmod = lm(Life.expectancy ~ ., data = life)
summary(lmod)
```

```
##
## Call:
## lm(formula = Life.expectancy ~ ., data = life)
##
## Residuals:
```

| | Min | 1Q | Median | 3Q | Max |
|--|----------|---------|--------|--------|---------|
| | -17.0291 | -2.1529 | 0.0557 | 2.3893 | 11.5018 |

```
##
## Coefficients:
```

| | Estimate | Std. Error | t value | Pr(> t) |
|---------------------------------|------------|------------|---------|--------------|
| (Intercept) | 5.500e+01 | 8.108e-01 | 67.833 | < 2e-16 *** |
| StatusDeveloping | -9.815e-01 | 3.464e-01 | -2.834 | 0.00466 ** |
| Adult.Mortality | -1.780e-02 | 9.674e-04 | -18.399 | < 2e-16 *** |
| infant.deaths | -3.007e-03 | 1.266e-03 | -2.376 | 0.01762 * |
| Alcohol | -1.552e-01 | 3.380e-02 | -4.590 | 4.77e-06 *** |
| percentage.expenditure | 3.491e-04 | 1.862e-04 | 1.875 | 0.06094 . |
| Hepatitis.B>=90% Covered | -6.372e-01 | 3.192e-01 | -1.996 | 0.04611 * |
| Measles | 1.683e-05 | 1.079e-05 | 1.560 | 0.11906 |
| BMI | 3.585e-02 | 6.161e-03 | 5.819 | 7.13e-09 *** |
| Polio>=90% Covered | 5.680e-01 | 4.439e-01 | 1.280 | 0.20087 |
| Total.expenditure | 6.994e-02 | 4.179e-02 | 1.674 | 0.09439 . |
| Diphtheria>=90% Covered | 9.097e-01 | 4.899e-01 | 1.857 | 0.06352 . |
| HIV.AIDS | -4.279e-01 | 1.849e-02 | -23.142 | < 2e-16 *** |
| GDP | 9.181e-06 | 2.925e-05 | 0.314 | 0.75368 |
| Population | 2.496e-09 | 1.766e-09 | 1.414 | 0.15769 |
| thinness..1.19.years | -5.018e-02 | 5.469e-02 | -0.918 | 0.35899 |
| thinness.5.9.years | 1.519e-03 | 5.374e-02 | 0.028 | 0.97745 |
| Income.composition.of.resources | 1.048e+01 | 8.507e-01 | 12.316 | < 2e-16 *** |
| Schooling | 8.843e-01 | 6.172e-02 | 14.328 | < 2e-16 *** |

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.686 on 1630 degrees of freedom
## Multiple R-squared:  0.8263, Adjusted R-squared:  0.8244
## F-statistic: 430.9 on 18 and 1630 DF,  p-value: < 2.2e-16
```

There are a few takeaways from this model:

- The p-value of the model is $2.2e-16 < 0.05$, indicating that it is significant.
- The Adj R-squared value of the model is 0.8244, indicating that about 82.44% of the observed variation can be explained by the variables in the model, which is quite a good result and can possibly be improved even further with model selection.
- Adult.Mortality, Alcohol, BMI, HIV.AIDS, Income.composition.of.resources and Schooling are the most significant variables with p-value < 0.5 .
- From the model we can interpret that StatusDeveloping, Adult.Mortality, infant.deaths, Alcohol,

HIV.AIDS, and thinness..1.19.years may have a negative effect on life expectancy.

- From the model we can interpret that Income.composition.of.resources has a strong positive effect on life expectancy.
- A peculiar result we can interpret from the model is that Hepatitis.B90% Covered also has a negative effect on life expectancy.

Model Selection

We will now generate models by using different techniques like Forward Selection Method, Backward Elimination Method and Stepwise Selection Method.

Build Model using Forward Selection Method.

```
ols_step_forward_p(lmod)
```

```
##
##                               Selection Summary
## -----
```

| ## Step | Variable Entered | R-Square | Adj. R-Square | C(p) | AIC | RMSE |
|---------|---------------------------------|----------|---------------|-----------|------------|--------|
| ## 1 | Schooling | 0.5294 | 0.5292 | 2771.7513 | 10612.7157 | 6.0362 |
| ## 2 | HIV.AIDS | 0.7304 | 0.7301 | 887.6286 | 9696.3271 | 4.5704 |
| ## 3 | Adult.Mortality | 0.7871 | 0.7867 | 357.3801 | 9308.9473 | 4.0627 |
| ## 4 | Income.composition.of.resources | 0.8092 | 0.8087 | 152.1307 | 9130.3986 | 3.8474 |
| ## 5 | percentage.expenditure | 0.8147 | 0.8141 | 102.1617 | 9083.8457 | 3.7924 |
| ## 6 | BMI | 0.8201 | 0.8194 | 54.0203 | 9037.6049 | 3.7384 |
| ## 7 | Diphtheria | 0.8218 | 0.8211 | 39.2920 | 9023.1915 | 3.7210 |
| ## 8 | Alcohol | 0.8231 | 0.8222 | 29.5343 | 9013.5567 | 3.7090 |
| ## 9 | thinness..1.19.years | 0.8240 | 0.8230 | 22.9694 | 9007.0292 | 3.7006 |
| ## 10 | Status | 0.8249 | 0.8238 | 16.6366 | 9000.6904 | 3.6924 |
| ## 11 | Hepatitis.B | 0.8252 | 0.8240 | 15.5038 | 8999.5443 | 3.6900 |
| ## 12 | Total.expenditure | 0.8255 | 0.8242 | 14.8813 | 8998.9062 | 3.6881 |
| ## 13 | infant.deaths | 0.8257 | 0.8243 | 14.8516 | 8998.8614 | 3.6870 |
| ## 14 | Measles | 0.8259 | 0.8244 | 14.7734 | 8998.7652 | 3.6858 |
| ## 15 | Population | 0.8262 | 0.8246 | 14.7661 | 8998.7380 | 3.6846 |
| ## 16 | Polio | 0.8263 | 0.8246 | 15.0990 | 8999.0524 | 3.6839 |

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

```
lmod_forward = lm(
  Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.composition.of.resources + percent
    BMI + Diphtheria + Alcohol + thinness..1.19.years + Status + Hepatitis.B +
    Total.expenditure + infant.deaths + Measles + Population + Polio,
  data = life
)
summary(lmod_forward)
```

```
##
## Call:
## lm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality +
##     Income.composition.of.resources + percentage.expenditure +
##     BMI + Diphtheria + Alcohol + thinness..1.19.years + Status +
##     Hepatitis.B + Total.expenditure + infant.deaths + Measles +
##     Population + Polio, data = life)
##
## Residuals:
```

```
##      Min      1Q   Median      3Q      Max
## -17.0291 -2.1512  0.0485   2.3846  11.4744
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5.499e+01  8.094e-01  67.942 < 2e-16 ***
## Schooling         8.858e-01  6.141e-02  14.426 < 2e-16 ***
## HIV.AIDS        -4.279e-01  1.848e-02 -23.157 < 2e-16 ***
## Adult.Mortality -1.779e-02  9.656e-04 -18.428 < 2e-16 ***
## Income.composition.of.resources 1.050e+01  8.481e-01  12.378 < 2e-16 ***
## percentage.expenditure 4.043e-04  6.128e-05   6.597 5.64e-11 ***
## BMI              3.579e-02  6.096e-03   5.871 5.24e-09 ***
## Diphtheria>=90% Covered 9.024e-01  4.888e-01   1.846  0.06505 .
## Alcohol         -1.551e-01  3.378e-02 -4.591 4.75e-06 ***
## thinness..1.19.years -4.903e-02  2.788e-02 -1.758  0.07885 .
## StatusDeveloping -9.882e-01  3.454e-01 -2.861  0.00428 **
## Hepatitis.B>=90% Covered -6.299e-01  3.180e-01 -1.981  0.04780 *
## Total.expenditure 6.940e-02  4.169e-02   1.664  0.09621 .
## infant.deaths    -2.996e-03  1.259e-03 -2.379  0.01746 *
## Measles          1.682e-05  1.077e-05   1.561  0.11869
## Population       2.486e-09  1.764e-09   1.409  0.15892
## Polio>=90% Covered 5.728e-01  4.433e-01   1.292  0.19657
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.684 on 1632 degrees of freedom
## Multiple R-squared:  0.8263, Adjusted R-squared:  0.8246
## F-statistic: 485.3 on 16 and 1632 DF, p-value: < 2.2e-16
```

Build Model using Backward Elimination Method.

```
ols_step_backward_p(lmod)
```

```
##
##
##              Elimination Summary
## -----
##      Variable              Adj.
## Step      Removed      R-Square R-Square    C(p)      AIC      RMSE
## -----
##      1  thinness.5.9.years  0.8263    0.8245  17.0008   9000.9530   3.6849
##      2      GDP            0.8263    0.8246  15.0990   8999.0524   3.6839
## -----
```

```
lmod_backward = lm(
  Life.expectancy ~ Status + Adult.Mortality + infant.deaths + Alcohol +
  percentage.expenditure + Hepatitis.B + Measles + BMI + Polio + Total.expenditure +
  Diphtheria + HIV.AIDS + Population + thinness..1.19.years + Income.composition.of.resources +
  Schooling,
  data = life
)
summary(lmod_backward)
```

```
##
## Call:
## lm(formula = Life.expectancy ~ Status + Adult.Mortality + infant.deaths +
```

```

##      Alcohol + percentage.expenditure + Hepatitis.B + Measles +
##      BMI + Polio + Total.expenditure + Diphtheria + HIV.AIDS +
##      Population + thinness..1.19.years + Income.composition.of.resources +
##      Schooling, data = life)
##
## Residuals:
##      Min        1Q    Median        3Q        Max
## -17.0291  -2.1512   0.0485   2.3846  11.4744
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5.499e+01  8.094e-01  67.942 < 2e-16 ***
## StatusDeveloping -9.882e-01  3.454e-01  -2.861  0.00428 **
## Adult.Mortality  -1.779e-02  9.656e-04 -18.428 < 2e-16 ***
## infant.deaths    -2.996e-03  1.259e-03  -2.379  0.01746 *
## Alcohol          -1.551e-01  3.378e-02  -4.591  4.75e-06 ***
## percentage.expenditure 4.043e-04  6.128e-05   6.597  5.64e-11 ***
## Hepatitis.B>=90% Covered -6.299e-01  3.180e-01  -1.981  0.04780 *
## Measles           1.682e-05  1.077e-05   1.561  0.11869
## BMI               3.579e-02  6.096e-03   5.871  5.24e-09 ***
## Polio>=90% Covered  5.728e-01  4.433e-01   1.292  0.19657
## Total.expenditure  6.940e-02  4.169e-02   1.664  0.09621 .
## Diphtheria>=90% Covered 9.024e-01  4.888e-01   1.846  0.06505 .
## HIV.AIDS          -4.279e-01  1.848e-02 -23.157 < 2e-16 ***
## Population        2.486e-09  1.764e-09   1.409  0.15892
## thinness..1.19.years -4.903e-02  2.788e-02  -1.758  0.07885 .
## Income.composition.of.resources 1.050e+01  8.481e-01  12.378 < 2e-16 ***
## Schooling          8.858e-01  6.141e-02  14.426 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.684 on 1632 degrees of freedom
## Multiple R-squared:  0.8263, Adjusted R-squared:  0.8246
## F-statistic: 485.3 on 16 and 1632 DF, p-value: < 2.2e-16

```

Build Model using Stepwise Selection Method.

```
ols_step_both_p(lmod)
```

```
##
##                                     Stepwise Selection Summary
## -----
##                                     Added/
## Step          Variable              Removed    R-Square    Adj.
##                                     R-Square    C(p)          AIC
## -----
##      1          Schooling            addition    0.529        0.529
##      2          HIV.AIDS              addition    0.730        0.730
##      3          Adult.Mortality        addition    0.787        0.787
##      4      Income.composition.of.resources addition    0.809        0.809
##      5          percentage.expenditure addition    0.815        0.814
##      6          BMI                   addition    0.820        0.819
##      7          Diphtheria             addition    0.822        0.821
##      8          Alcohol                 addition    0.823        0.822
##      9          thinness..1.19.years    addition    0.824        0.823
##     10          Status                  addition    0.825        0.824
##     11          Hepatitis.B             addition    0.825        0.824
## -----
```

```
lmod_stepwise = lm(
  Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.composition.of.resources +
  percentage.expenditure + BMI + Diphtheria + Alcohol + thinness..1.19.years +
  Status + Hepatitis.B,
  data = life
)
summary(lmod_stepwise)
```

```
##
## Call:
## lm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality +
##      Income.composition.of.resources + percentage.expenditure +
##      BMI + Diphtheria + Alcohol + thinness..1.19.years + Status +
##      Hepatitis.B, data = life)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.2593  -2.1481   0.0745   2.4046  11.5838
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.534e+01  7.750e-01  71.401  < 2e-16 ***
## Schooling       9.061e-01  6.102e-02  14.848  < 2e-16 ***
## HIV.AIDS       -4.239e-01  1.833e-02 -23.122  < 2e-16 ***
## Adult.Mortality -1.779e-02  9.636e-04 -18.464  < 2e-16 ***
## Income.composition.of.resources 1.037e+01  8.444e-01  12.280  < 2e-16 ***
## percentage.expenditure  4.098e-04  6.119e-05   6.698 2.90e-11 ***
## BMI             3.610e-02  6.071e-03   5.946 3.36e-09 ***
## Diphtheria>=90% Covered  1.439e+00  3.443e-01   4.181 3.05e-05 ***
## Alcohol        -1.605e-01  3.353e-02  -4.788 1.84e-06 ***
## thinness..1.19.years -7.223e-02  2.491e-02  -2.900 0.00378 **
## StatusDeveloping -1.014e+00  3.454e-01  -2.934 0.00339 **
```

```
## Hepatitis.B>=90% Covered      -5.567e-01  3.149e-01  -1.768  0.07723 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.69 on 1637 degrees of freedom
## Multiple R-squared:  0.8252, Adjusted R-squared:  0.824
## F-statistic: 702.7 on 11 and 1637 DF,  p-value: < 2.2e-16
```

Build Model using All Possible Regressions Method.

```
# ols_step_all_possible(lmod, sbc = TRUE)
```

In summary, variables chosen by the methods (x denotes the variable was chosen by the method):

| Model Selection Method | Status | Adult.Mortality | infant.deaths | Alcohol |
|------------------------|--------|-----------------|---------------|---------|
| Forward Selection | x | x | x | x |
| Backward Elimination | x | x | x | x |
| Stepwise Selection | x | x | | x |

| Model Selection Method | percentage.expenditure | Hepatitis.B | Measles | BMI | Polio |
|------------------------|------------------------|-------------|---------|-----|-------|
| Forward Selection | x | x | x | x | x |
| Backward Elimination | x | x | x | x | x |
| Stepwise Selection | x | x | | x | |

| Model Selection Method | Total.expenditure | Diphtheria | HIV.AIDS | GDP | Population |
|------------------------|-------------------|------------|----------|-----|------------|
| Forward Selection | x | x | x | | x |
| Backward Elimination | x | x | x | | x |
| Stepwise Selection | | x | x | | |

| Model Selection Method | thinness..1.19.years | thinness.5.9.years |
|------------------------|----------------------|--------------------|
| Forward Selection | x | |
| Backward Elimination | x | |
| Stepwise Selection | x | |

| Model Selection Method | Income.compostition.of.resources | Schooling |
|------------------------|----------------------------------|-----------|
| Forward Selection | x | x |
| Backward Elimination | x | x |
| Stepwise Selection | x | x |

Both the Forward Selection method and Backward Elimination method have chosen the same set of variables.

Adj. R-squared values of the above models:

```
data.frame(
  model = c('lmod', 'lmod_forward', 'lmod_backward', 'lmod_stepwise'),
  AdjRsquare = c(
    summary(lmod)$adj.r.square,
    summary(lmod_forward)$adj.r.square,
    summary(lmod_backward)$adj.r.square,
    summary(lmod_stepwise)$adj.r.square
  )
)
```

```
##           model AdjRsquare
## 1           lmod  0.8244244
## 2 lmod_forward  0.8246289
## 3 lmod_backward 0.8246289
## 4 lmod_stepwise 0.8240486
```


We will be choosing the model chosen by Forward Selection method `lmod_forward` as it has the highest Adj. R-squared value.

```
lmod_final = lmod_forward
summary(lmod_final)
```

```
##
## Call:
## lm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality +
##     Income.composition.of.resources + percentage.expenditure +
##     BMI + Diphtheria + Alcohol + thinness..1.19.years + Status +
##     Hepatitis.B + Total.expenditure + infant.deaths + Measles +
##     Population + Polio, data = life)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -17.0291  -2.1512   0.0485   2.3846  11.4744
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5.499e+01  8.094e-01  67.942 < 2e-16 ***
## Schooling         8.858e-01  6.141e-02  14.426 < 2e-16 ***
## HIV.AIDS        -4.279e-01  1.848e-02 -23.157 < 2e-16 ***
## Adult.Mortality  -1.779e-02  9.656e-04 -18.428 < 2e-16 ***
## Income.composition.of.resources  1.050e+01  8.481e-01  12.378 < 2e-16 ***
## percentage.expenditure  4.043e-04  6.128e-05   6.597 5.64e-11 ***
## BMI              3.579e-02  6.096e-03   5.871 5.24e-09 ***
## Diphtheria>=90% Covered  9.024e-01  4.888e-01   1.846  0.06505 .
## Alcohol         -1.551e-01  3.378e-02  -4.591 4.75e-06 ***
## thinness..1.19.years  -4.903e-02  2.788e-02  -1.758  0.07885 .
## StatusDeveloping  -9.882e-01  3.454e-01  -2.861  0.00428 **
## Hepatitis.B>=90% Covered  -6.299e-01  3.180e-01  -1.981  0.04780 *
## Total.expenditure  6.940e-02  4.169e-02   1.664  0.09621 .
## infant.deaths    -2.996e-03  1.259e-03  -2.379  0.01746 *
## Measles          1.682e-05  1.077e-05   1.561  0.11869
## Population       2.486e-09  1.764e-09   1.409  0.15892
## Polio>=90% Covered  5.728e-01  4.433e-01   1.292  0.19657
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.684 on 1632 degrees of freedom
## Multiple R-squared:  0.8263, Adjusted R-squared:  0.8246
## F-statistic: 485.3 on 16 and 1632 DF,  p-value: < 2.2e-16
```

Results

Model Error Estimation

We will now use our final model to see how well it performs in predicting the life expectancy of the human population.

```
result = predict(lmod_final, life)
```

Mean Squared Error:

```
mse = mean((life$Life.expectancy - result) ^ 2)
mse
```

```
## [1] 13.43106
```

Root Mean Squared Error:

```
rmse = sqrt(mse)
rmse
```

```
## [1] 3.664841
```

Mean Absolute Error:

```
n = length(result)
sum = 0

for (i in 1:n) {
  sum = sum + abs(life$Life.expectancy[i] - result[i])
}

mae = sum / n
mae
```

```
##          1
## 2.817618
```

In summary,

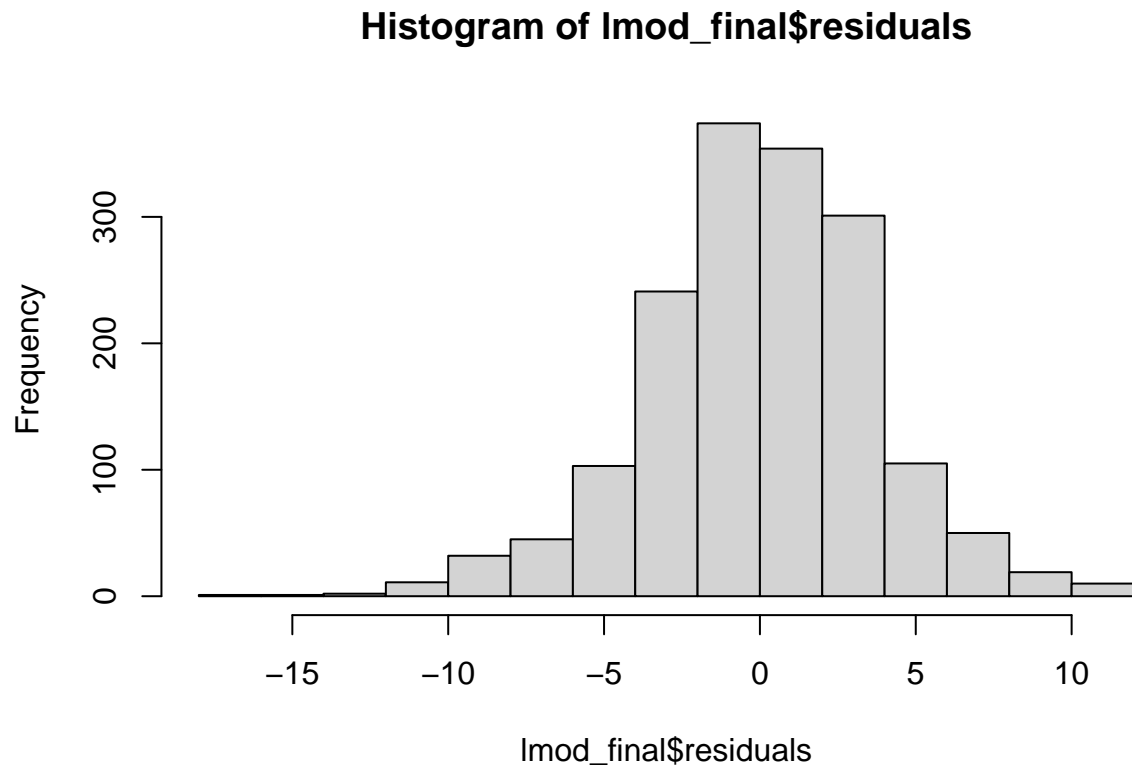
```
data.frame(
  Method = c('MSE', 'RMSE', 'MAE'),
  Result = c(mse, rmse, mae)
)
```

```
##   Method   Result
## 1    MSE 13.431062
## 2   RMSE  3.664841
## 3    MAE  2.817618
```

Model Adequacy Checking

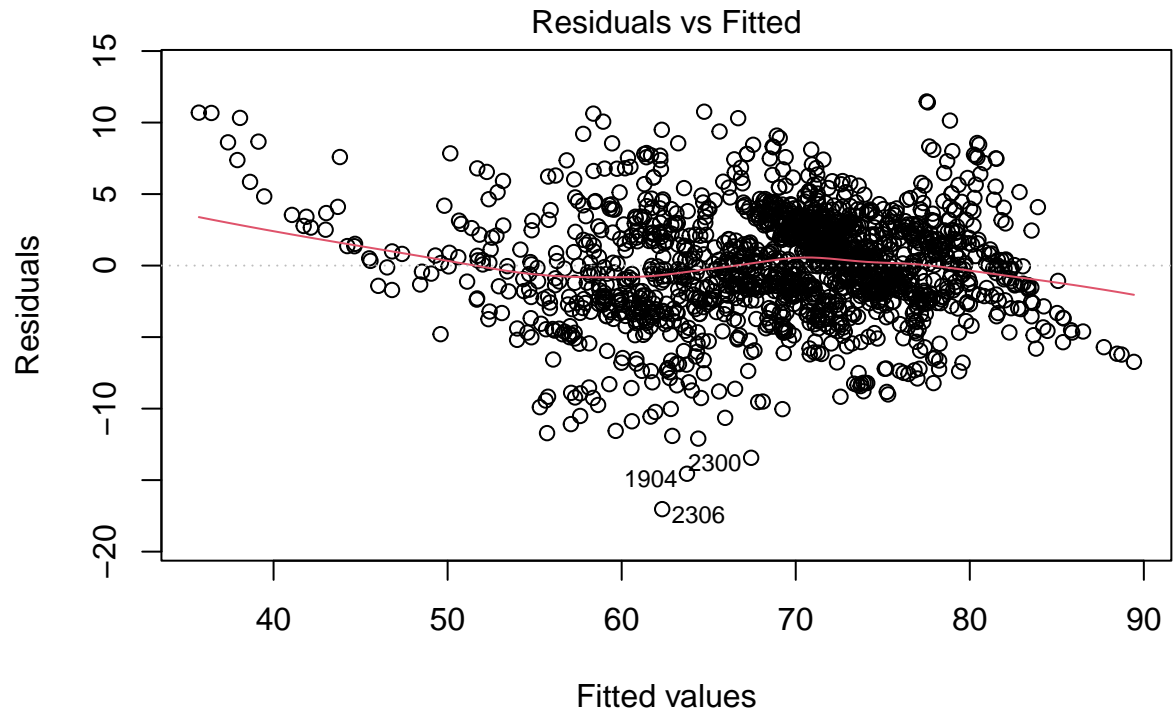
Normality Testing:

```
hist(lmod_final$residuals, breaks = 20)
```

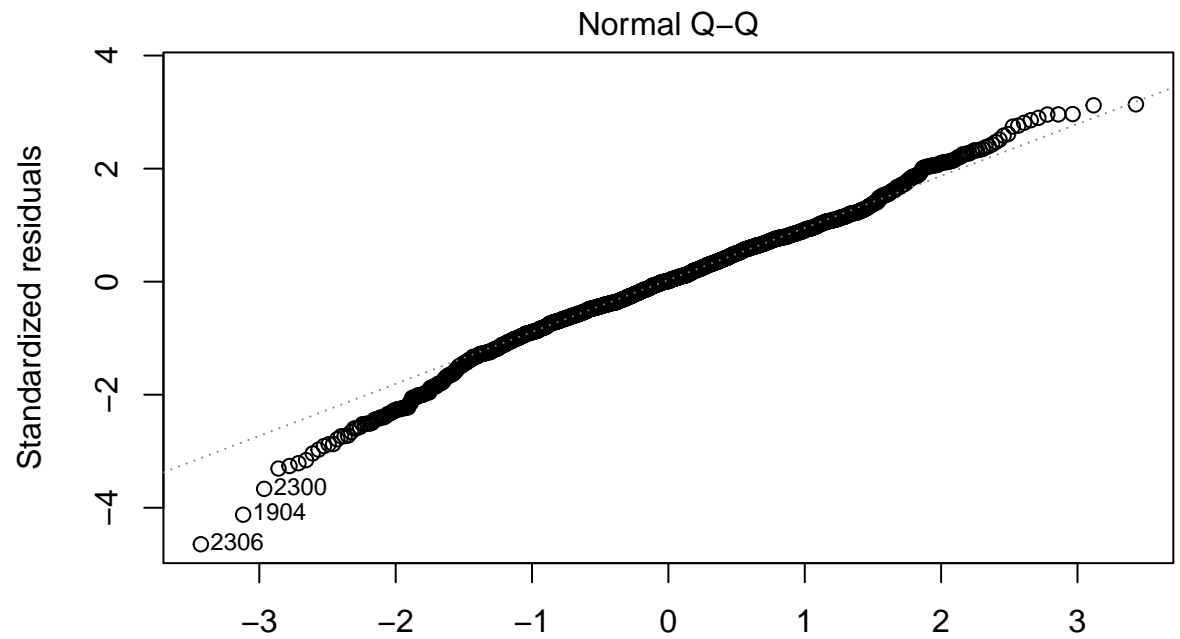


Most of the residuals seem to be distributed in the center, indicating that they are distributed normally.

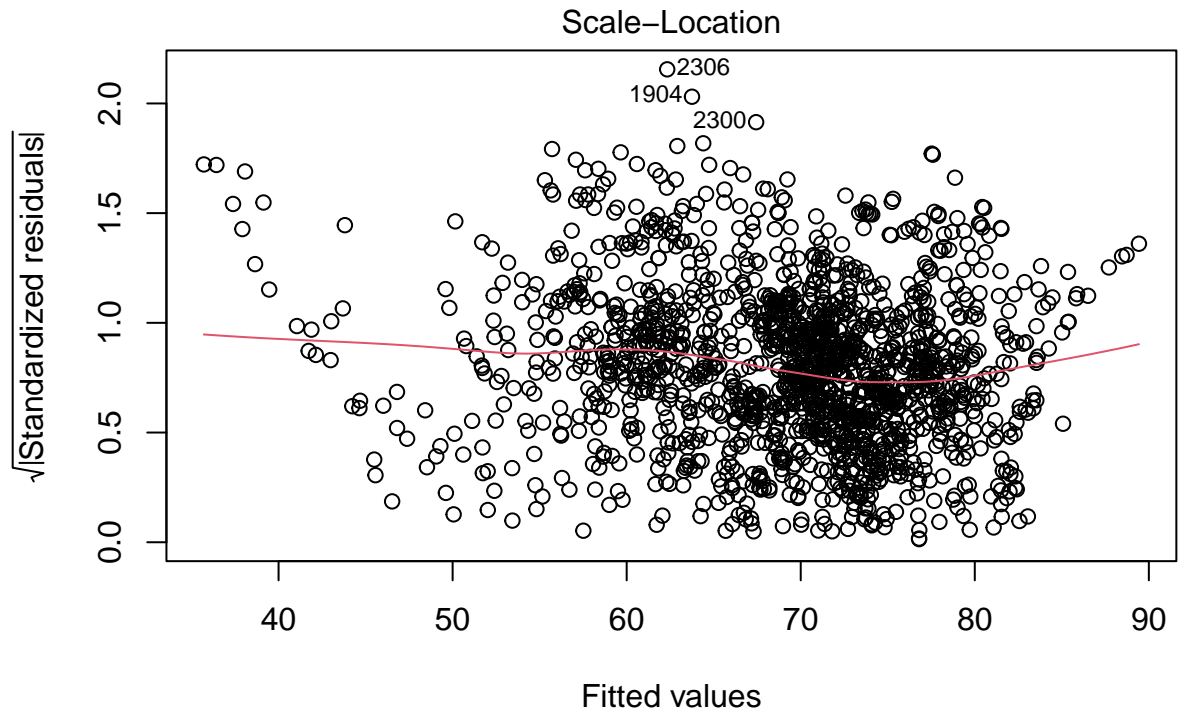
```
plot(lmod_final, which = c(1:6))
```



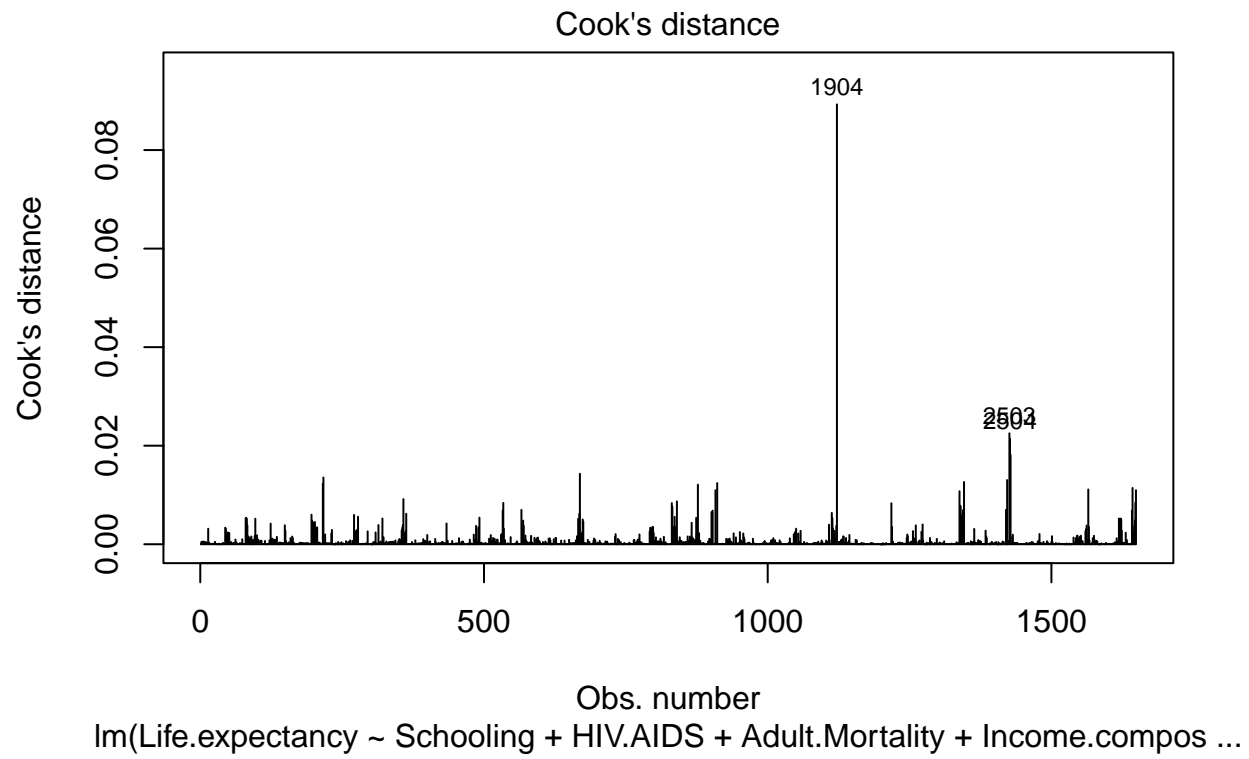
lm(Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.compos ...

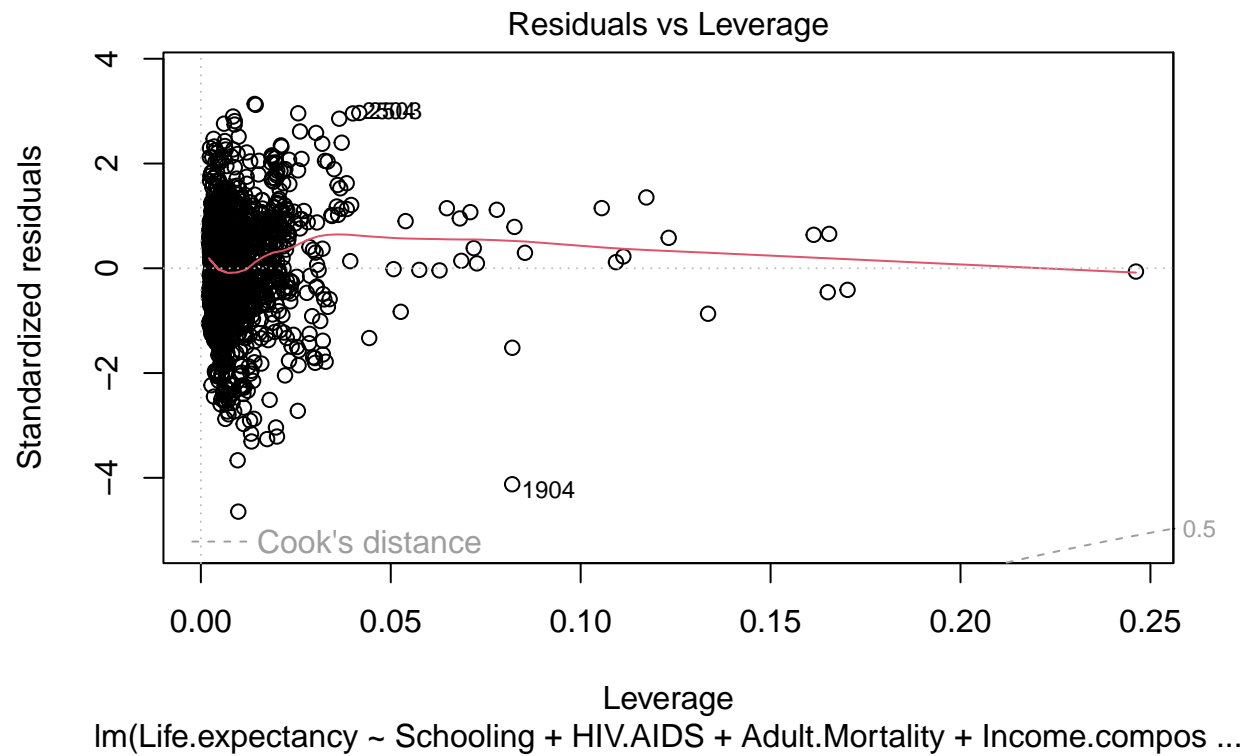


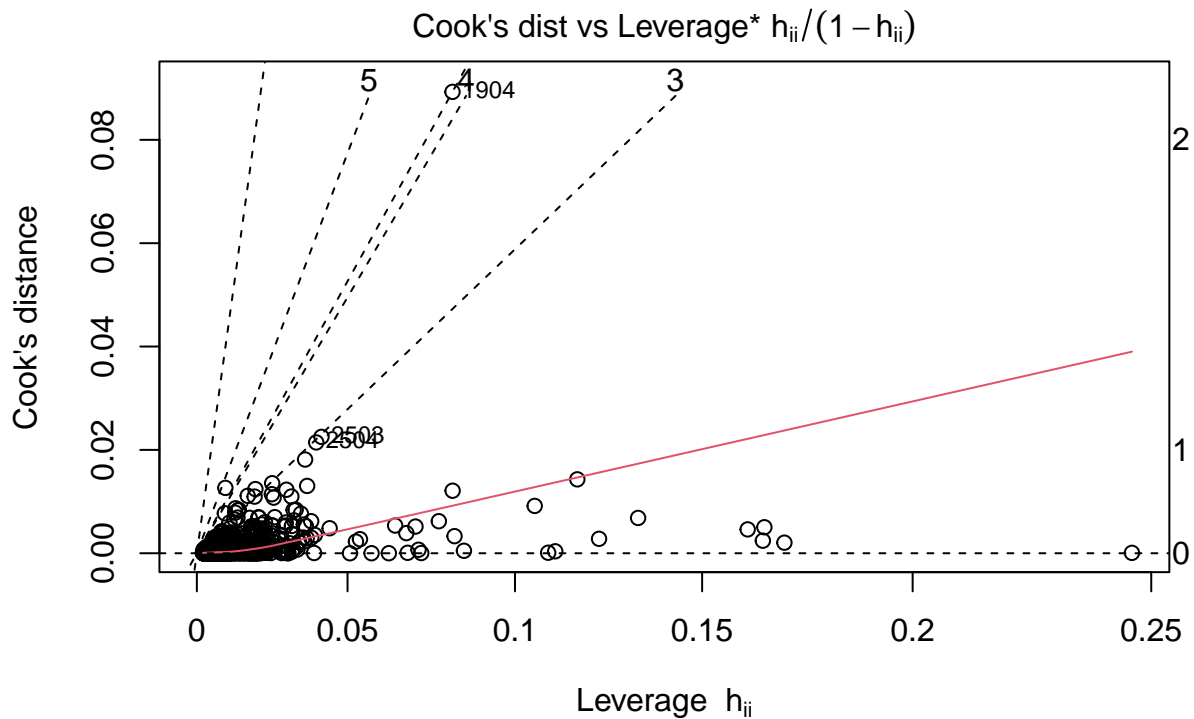
Im(Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.compos ...



lm(Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.compos ...







lm(Life.expectancy ~ Schooling + HIV.AIDS + Adult.Mortality + Income.compos ...

There is no obvious observable pattern in the above plots, indicating that the model is appropriate.

Multicollinearity Test:

```
vif(lmod_final)
```

| | | |
|----|------------------------|---------------------------------|
| ## | Schooling | HIV.AIDS |
| ## | 3.578091 | 1.509013 |
| ## | Adult.Mortality | Income.composition.of.resources |
| ## | 1.778090 | 2.927679 |
| ## | percentage.expenditure | BMI |
| ## | 1.411270 | 1.761017 |
| ## | Diphtheria | Alcohol |
| ## | 7.102613 | 2.249650 |
| ## | thinness..1.19.years | Status |
| ## | 1.996547 | 1.815140 |
| ## | Hepatitis.B | Total.expenditure |
| ## | 3.072344 | 1.116175 |
| ## | infant.deaths | Measles |
| ## | 2.811727 | 1.433389 |
| ## | Population | Polio |
| ## | 1.876386 | 5.834447 |

A VIF > 10 implies serious problems with multicollinearity.

Since the VIF for all of the predictors is less than 10, there seems to be no issue with multicollinearity.

Discussion