

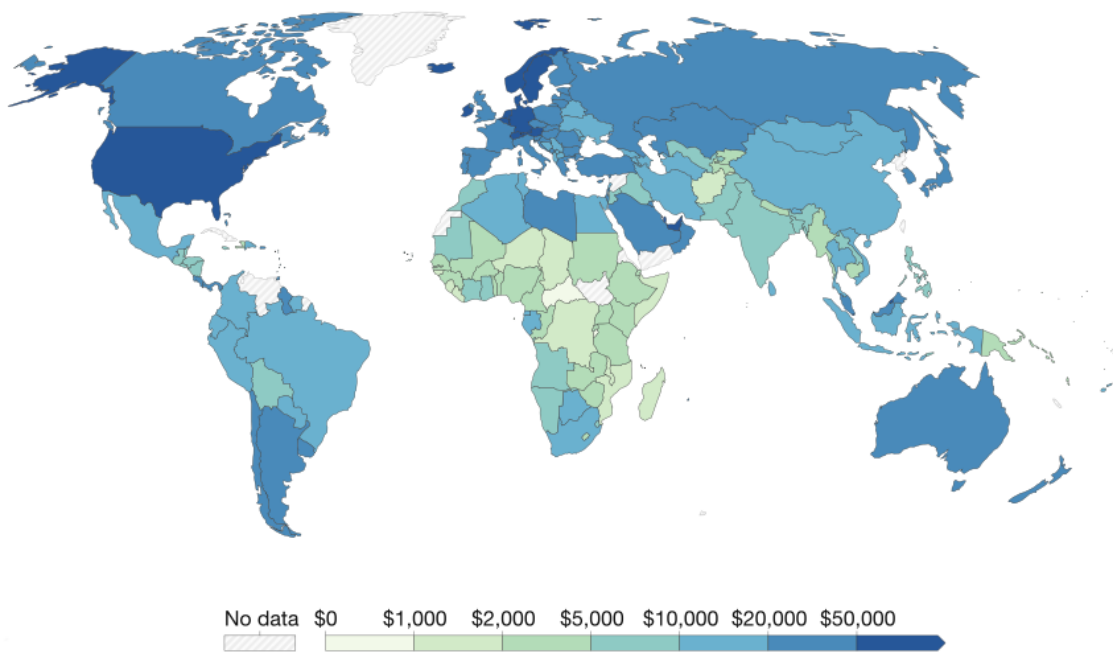
The Puzzle of Wealth: Understanding Per Capita GDP Pieces

NARB

GDP per capita, 2021

This data is adjusted for inflation and for differences in the cost of living between countries.

Our World
in Data



(GDP per capita, 2021)

Introduction

GDP is a crucial indicator of a nation's financial well-being as it reflecting average incomes, levels of consumption, job availability, and other wealth indicators. In our model GDP per capita is measured in US dollars and represents the average economic output per person for each represented country. Understanding the factors influencing GDP helps countries know what key issues to focus on successfully promote economic growth. Our research questions are as followed:

- Does a high fertility rate correlate with lower average GDP?
- Is there a relationship between average population age and GDP?
- What factor most significantly predicts GDP?

Low fertility rates may signal higher parental incomes, allowing greater investment in quality education, potentially indicating a higher average GDP (Doepke, 2022). A higher average age might be a sign of increased labor productivity or, conversely, a higher dependency rate. If neither fertility rates nor average age emerges as the primary GDP predictor, our dataset, combining 2023 US Census Bureau and Focus Economics data, enables exploration of other influential factors. The Census Bureau gathers information through censuses and surveys, while Focus Economics compiles monthly data from diverse international forecasts (International Database, 2023; About Us, 2023). The data used for this project contained country-level information across 165 nations on quantitative and categorical factors like life expectancy, dependency, growth rates, income, etc. The target variable for prediction modeling was the log-transformed version of GDP per capita calculated in USD.

Methods and Analysis

First, we cleaned the data by filtering out N/A values, converting the structures of the objects, and reformatting the position and ordering for easier access. Then, we conducted an initial exploratory analysis by plotting all quantitative variables with GDP on scatterplots to understand the relationships between these variables and GDP. We analyzed qualitative factors like country size, income level, and growth rate with boxplots and interaction plots. Multicollinearity and variable screening through stepwise regression helped identify highly correlated predictor variables. The removal and addition of quantitative predictors, like fertility and dependency, through variable screens improved model accuracy.

The remaining significant quantitative predictors were average age, life expectancy, and their interaction. After testing combinations of qualitative factors through comparison of model fit

statistics, we chose to include income because it was significant. Our final model contained average age, life expectancy, their interaction, and income level classified into four levels and three dummy variables: high income (the control), low income, lower middle income, upper middle income. The best model was chosen through a Global F test, Nested F test on income levels, t-tests on important predictors, R^2 values, and 2s values. The linear regression model was tested for violations of assumptions through residual analysis split into residual plots, influence analysis, and outlier detection. From the residual plots, we confirmed that the zero mean and constant variance assumptions were not violated. From the Q-Q plot and histogram, we confirmed that the normality assumption was not violated. However, after careful analysis of the influential observations, we noticed five influential points that can be categorized as outliers due to their influence on the data: Angola, San Marino, Seychelles, Sierra Leone, and Ukraine. With cook's distance, studentized residuals, and leverage plots, we used a subsetting technique to confirm that the removal of these data points sharpened the model by increasing the R^2 and F Statistics, and reducing the residual standard error and P-value. Lastly, we applied a box-cox transformation to check sensitivity to transformations and learned that the log transformation applied to the GDP was significant and contributed to the accurate linearization of the data.

Results

We concluded that our model was adequate with an R^2 of 0.8733, meaning 87.33% of the variability observed in GDP per capita is explained by the regression model. Comparing this to the first model with a 0.6393 R^2 and the pre-residual analysis model with a 0.7396 R^2 , our latest and final model accounts for the best parameters and fits our data most accurately. From this, we used our model for estimation and prediction and calculated the confidence and prediction intervals for $E(y)$ and y , respectively. Using data within our frame, we accurately predicted multiple parameters such as the GDP, Average Age, and more of a country given values using 95% confidence intervals. Thus, our final regression model stands as a statistically sound and reliable tool for explaining and predicting GDP per capita variability between countries.

Conclusions

From our data analysis, we found that fertility rate has a moderate relationship with GDP. As the average number of children the women in a country will have increases, GDP per capita decreases by a rate of 0.67. However, fertility rate displays extreme multicollinearity

with growth rate (0.86), average age (-0.85), and dependency (0.859). These relationships suggest that these explanatory variables provide similar information about GDP per capita, so it can't be determined that fertility rate alone is a significant predictor. Alternatively, while average age presents some concerns of multicollinearity, we confirmed through a t-test and tests for violations of assumptions that it is significant for predicting GDP. We concluded that for each one year increase in the log of average age, the log of GDP per capita decreases by 5.2341. Finally, we found that income level is also a significant predictor. If a country is classified as a low income country, the log of GDP per capita decreases by 3.0819, in lower middle income countries the log of GDP per capita decreases by 0.9872, in upper middle income countries the log of GDP per capita decreases by 0.9782, and in high income countries, the log of GDP per capita doesn't increase or decrease. Ultimately, an increase in income is associated with an expected decrease in the log of GDP per capita.

Our dataset has limitations, including the exclusion of some countries present in only one data source due to resource constraints, leading to a bias against smaller or less developed nations. Additionally, variations in census collection policies among countries pose challenges, with some, like the United States, updating data every 10 years, while others, such as Mauritania, may have outdated information since 2008. Moreover, concerns about source accuracy arise, The International Dataset relies partially on estimations and is subject to sample errors (Jacobsen, 2023). Additionally, the data collected in 2020 may be inaccurate or unreliable due to COVID-19 pandemic.

Future steps for our research include comparisons between other years' census data and considering other variables in our analysis. Because many countries collect census data every 10 years, we could revisit this model in 2030 to observe what trends might have continued or changed over time. Another benefit of this would be that many countries may be more resourced and able to collect data with ample time to recover from the effects of the pandemic. Similarly, we could look at data from 2010 to observe which countries have improved or worsened GDP growth rates. By viewing trends across a 20 year time span, we can make more accurate predictions about GDP. We could also try adding a country's capital as an explanatory variable to predict the trajectory of GDP. Capital in this case refers to the supply of things like buildings, machinery, software, etc (The Conference Board, 2022). An understanding of capital in combination with labor and total factor productivity, which refers to efficiency) will give a more comprehensive picture of GDP.

Appendix A: Data Dictionary

Variable Name	Measured	Units/Levels	Description	Data Source
GDP	GDP per capita by country	USD	Gross Domestic Product (GDP) per capita is the total economic output (the sum of all goods and services produced within a country's borders) divided by the population of the country. It represents the average income per person in the country.	World Bank
Life_Expectancy	Life expectancy by country	Years	Life expectancy is the average number of years a person is expected to live if current mortality rates remain the same throughout their lifetime.	US Census Bureau

Variable Name	Measured	Units/Levels	Description	Data Source
Density	Population density by country	People per square kilometer	Population density is the average number of people living per square kilometer of land. Population density provides a data point on how crowded or urban a country is.	US Census Bureau
Dependency	Dependency ratio by country	Percentage	The dependency ratio is the ratio of the dependent population to the working-age population. The dependent population is people aged 0-14 and 65+. The working-age population is people ages 15-64.	US Census Bureau

Variable Name	Measured	Units/Levels	Description	Data Source
Fertility	Total fertility ratio by country	Children per woman	The total fertility rate or TFR is the average number of children that a woman has in the country.	US Census Bureau
Sex_Ratio	Gender ratio by country	Ratio of males to females	The gender ratio is the number of males to the number of females in the country.	US Census Bureau
Debt	Debt as a percentage of GDP by country	Percentage of GDP	Debt as a percentage of GDP is the ratio of a country's total debt to its Gross Domestic Product/GDP.	Focus Economics

Variable Name	Measured	Units/Levels	Description	Data Source
Income	Income group of countries	High income, low income, middle income	Income group classification is determined by a country's Gross National Income (GNI) per capita. Lower Income: \$1,135 or less of GNI per capita. Middle Income: between \$1,136 and \$13,845. High Income: \$113,846 or more of GNI per capita.	World Bank
Growth	Population growth rate by country	Negative, Average, or Rapid	Population growth rate is the percentage change in a country's population from one year to the next. For reference, the World's population growth rate in 2022 was .9%.	US Census Bureau

Variable Name	Measured	Units/Levels	Description	Data Source
Size	Size of the country by population	Small, medium, or large	The size of a country by population is the total number of people in the country, including all land within its borders.	US Census Bureau

Appendix B: Data Rows

	Name	Size	Growth	Density	Fertility	Life_Expectancy	Sex_Ratio
1	Albania	M	A	113.2	1.55	79.7	0.97
2	Algeria	L	R	19.4	2.97	77.8	1.03
3	Andorra	S	N	182.6	1.46	83.6	1.05
4	Angola	L	R	28.9	5.76	62.5	0.96
5	Antigua and Barbuda	S	R	229.1	1.94	78.0	0.89
6	Argentina	L	A	17	2.17	78.6	0.98
7	Armenia	M	N	106	1.65	76.4	0.96
8	Australia	L	R	3.4	1.73	83.3	0.99
9	Austria	L	A	108.4	1.51	82.5	0.96
10	Azerbaijan	L	A	128.3	1.69	74.9	1.00

	Dependency	Income	Debt	GDP
1	48.1	Upper middle income	73.32	6802.805
2	60.5	Lower middle income	9.06	4273.922
3	46.7	High income	40.90	41992.793
4	98.1	Lower middle income	64.24	2998.501
5	47.2	High income	99.10	18745.174
6	56.7	Upper middle income	52.13	13686.009
7	48.1	Upper middle income	46.89	7014.207
8	54.1	High income	37.64	64491.430
9	53.6	High income	86.19	52131.447
10	45.4	Upper middle income	28.32	7736.701

Appendix C: Final Model Output and Plots

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

## corplot 0.92 loaded

##
## Attaching package: 'Hmisc'

## The following objects are masked from 'package:dplyr':
##
##   src, summarize

## The following objects are masked from 'package:base':
##
##   format.pval, units

## Loading required package: carData

##
## Attaching package: 'car'

## The following object is masked from 'package:dplyr':
##
##   recode

##
## Attaching package: 'olsrr'

## The following object is masked from 'package:datasets':
##
##   rivers

##
## Attaching package: 'MASS'

## The following object is masked from 'package:olsrr':
##
##   cement

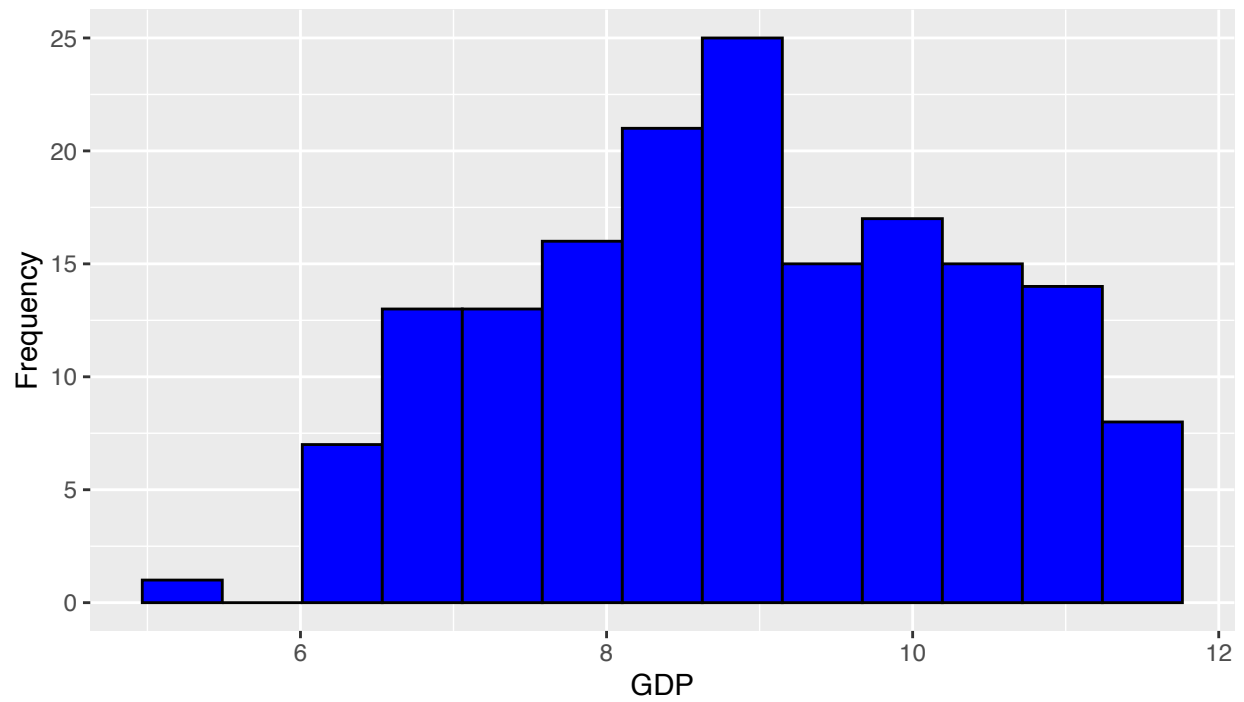
## The following object is masked from 'package:dplyr':
##
##   select
```

EDA(Plots)

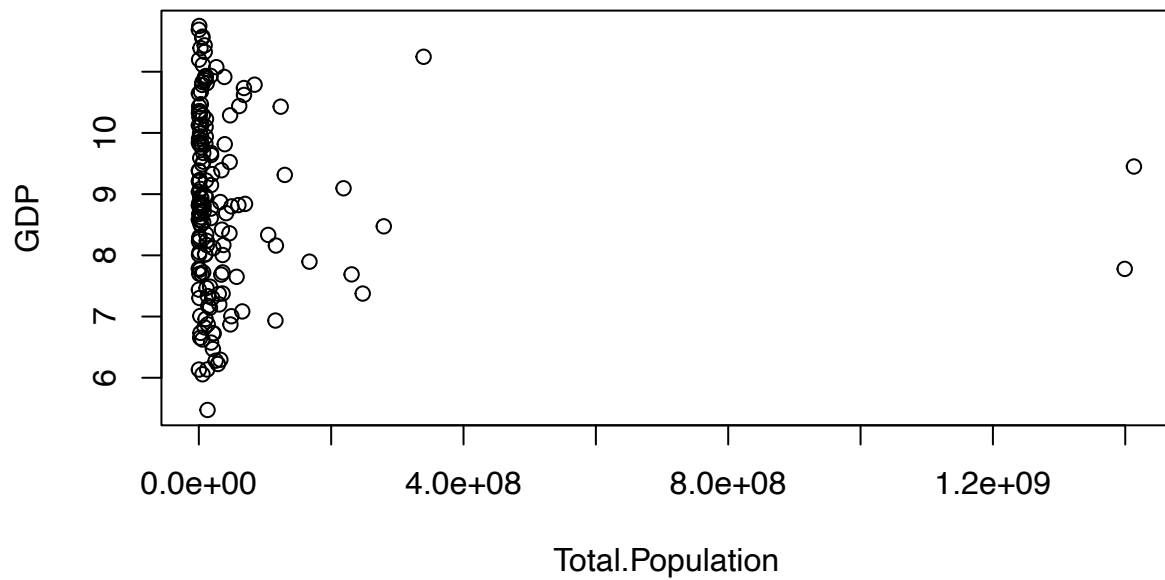
```
##           Name Total.Population Growth.Rate..Value. Density Fertility
## 1      Albania      3101621          0.19    113.2      1.55
## 2      Algeria      46286076         1.62     19.4      2.97
## 3      Andorra       85468        -0.11    182.6      1.46
## 4      Angola      35981281         3.34     28.9      5.76
## 5 Antigua and Barbuda    101489         1.13    229.1      1.94
## 6      Argentina    46621847         0.80     17.0      2.17
## Life_Expectancy Under.5.Mortality.Rate Sex_Ratio Dependency Average.Age Debt
## 1      79.7          12.3      0.97      48.1    3.577948 73.32
## 2      77.8          21.7      1.03      60.5    3.363842 9.06
## 3      83.6           4.1      1.05      46.7    3.873282 40.90
## 4      62.5          87.1      0.96      98.1    2.785011 64.24
## 5      78.0          16.8      0.89      47.2    3.514526 99.10
## 6      78.6          10.8      0.98      56.7    3.496508 52.13
## Size Growth      Income      GDP
## 1    M    A Upper middle income 8.825090
## 2    L    R Lower middle income 8.360287
## 3    S    N      High income 10.645253
## 4    L    R Lower middle income 8.005868
## 5    S    R      High income 9.838692
## 6    L    A Upper middle income 9.524129

## 'data.frame': 165 obs. of 15 variables:
## $ Name : chr "Albania" "Algeria" "Andorra" "Angola" ...
## $ Total.Population : num 3101621 46286076 85468 35981281 101489 ...
## $ Growth.Rate..Value. : num 0.19 1.62 -0.11 3.34 1.13 0.8 -0.4 1.19 0.31 0.43 ...
## $ Density : num 113.2 19.4 182.6 28.9 229.1 ...
## $ Fertility : num 1.55 2.97 1.46 5.76 1.94 2.17 1.65 1.73 1.51 1.69 ...
## $ Life_Expectancy : num 79.7 77.8 83.6 62.5 78 78.6 76.4 83.3 82.5 74.9 ...
## $ Under.5.Mortality.Rate: num 12.3 21.7 4.1 87.1 16.8 10.8 13.5 3.6 3.8 13.4 ...
## $ Sex_Ratio : num 0.97 1.03 1.05 0.96 0.89 0.98 0.96 0.99 0.96 1 ...
## $ Dependency : num 48.1 60.5 46.7 98.1 47.2 56.7 48.1 54.1 53.6 45.4 ...
## $ Average.Age : num 3.58 3.36 3.87 2.79 3.51 ...
## $ Debt : num 73.32 9.06 40.9 64.24 99.1 ...
## $ Size : Factor w/ 3 levels "L","M","S": 2 1 3 1 3 1 2 1 1 1 ...
## $ Growth : Factor w/ 3 levels "A","N","R": 1 3 2 3 3 1 2 3 1 1 ...
## $ Income : Factor w/ 4 levels "High income",...: 4 3 1 3 1 4 4 1 1 4 ...
## $ GDP : num 8.83 8.36 10.65 8.01 9.84 ...
```

Histogram of Response Variable – GDP

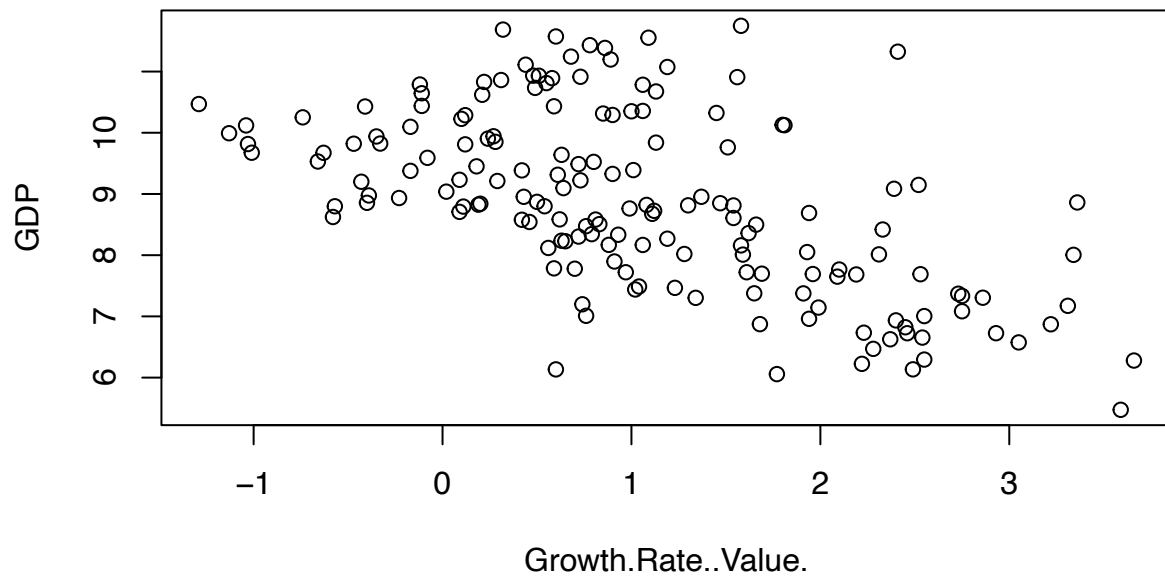


The Relationship between GDP and Total.Population



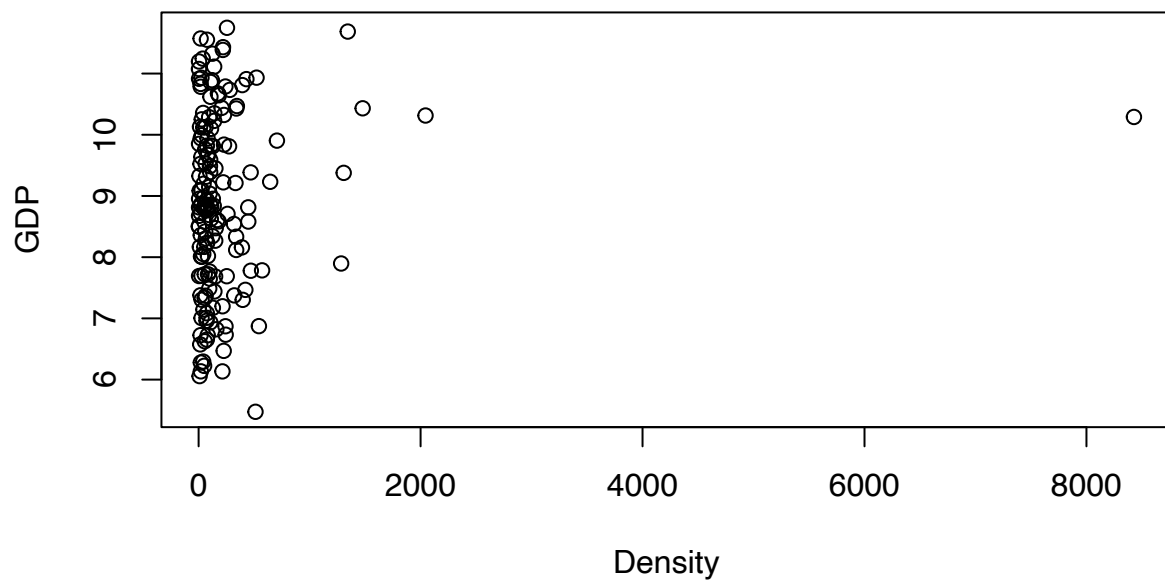
```
## [1] -0.03447192
```

The Relationship between GDP and Growth.Rate..Value.



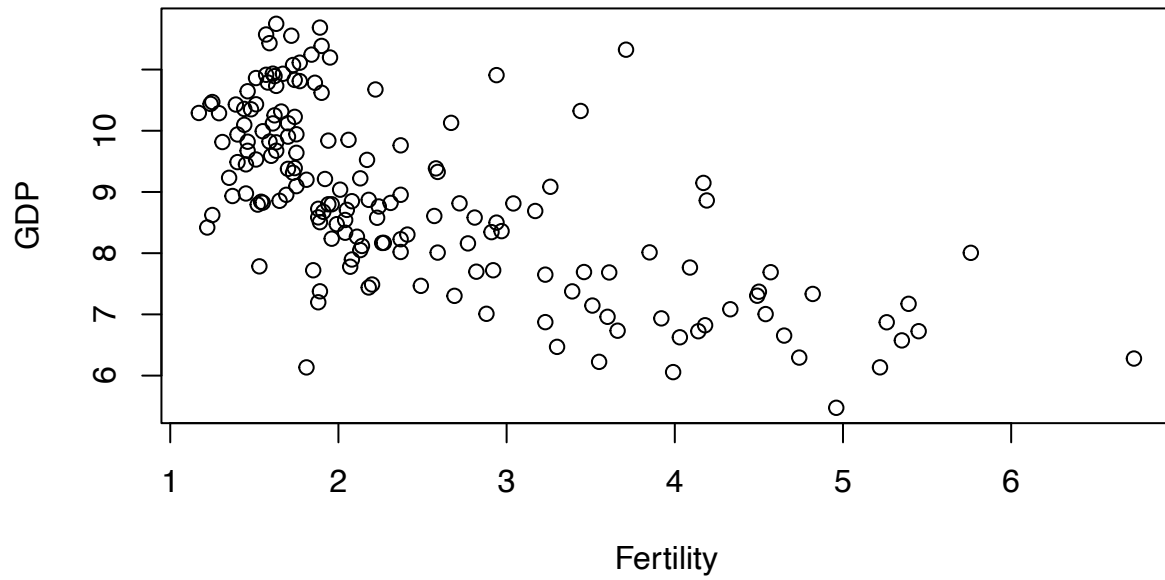
```
## [1] -0.580973
```

The Relationship between GDP and Density



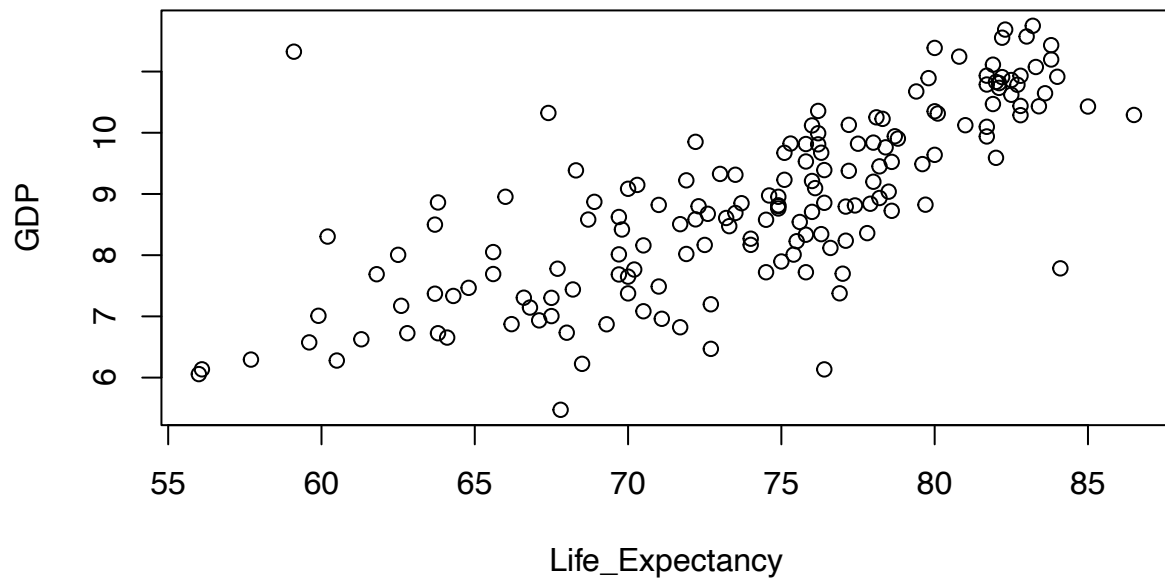
```
## [1] 0.119799
```

The Relationship between GDP and Fertility



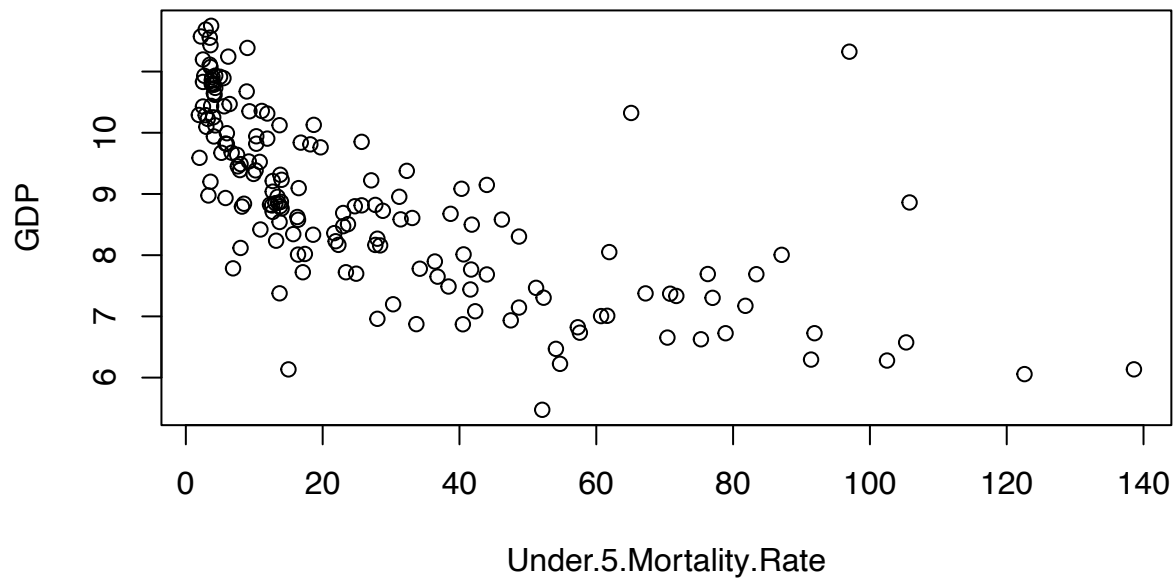
```
## [1] -0.6707076
```

The Relationship between GDP and Life_Expectancy



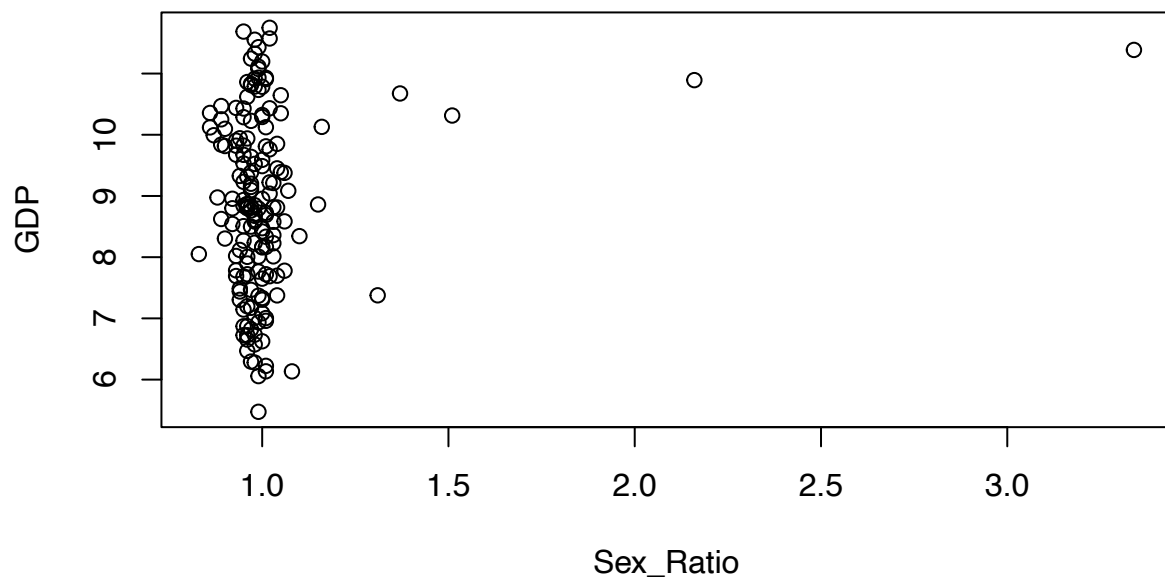
```
## [1] 0.7482856
```

The Relationship between GDP and Under.5.Mortality.Rate



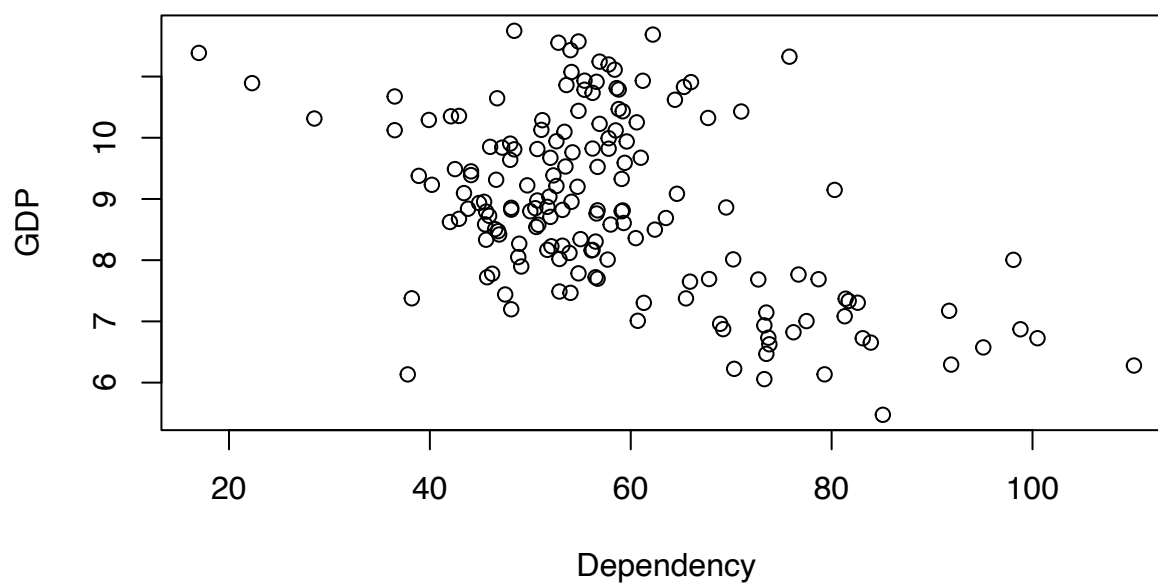
```
## [1] -0.6833988
```

The Relationship between GDP and Sex_Ratio



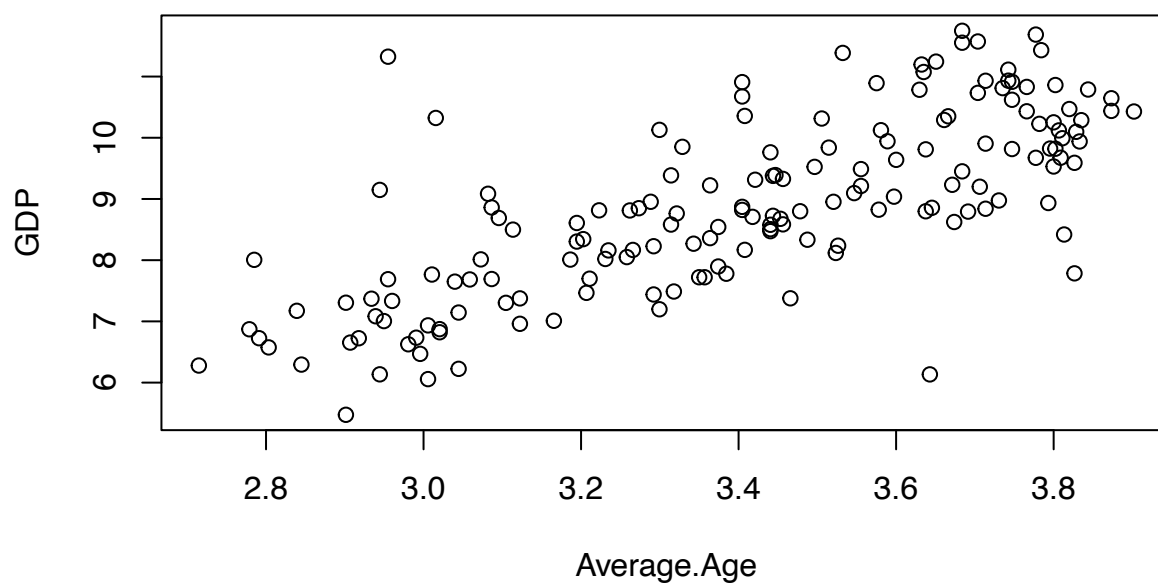
```
## [1] 0.168265
```


The Relationship between GDP and Dependency



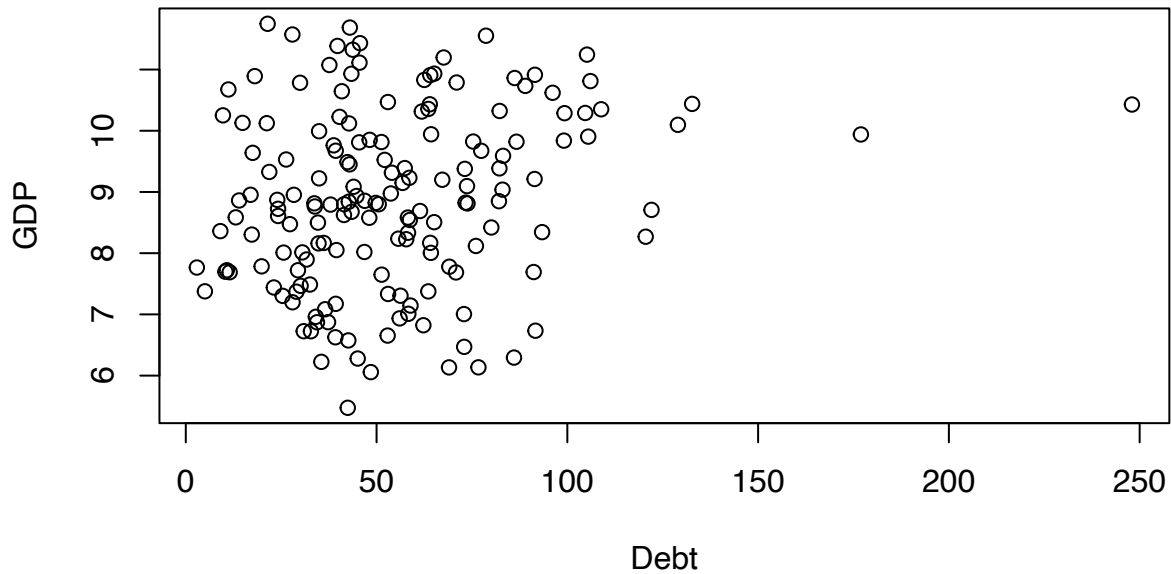
```
## [1] -0.4901131
```

The Relationship between GDP and Average.Age



```
## [1] 0.7420579
```

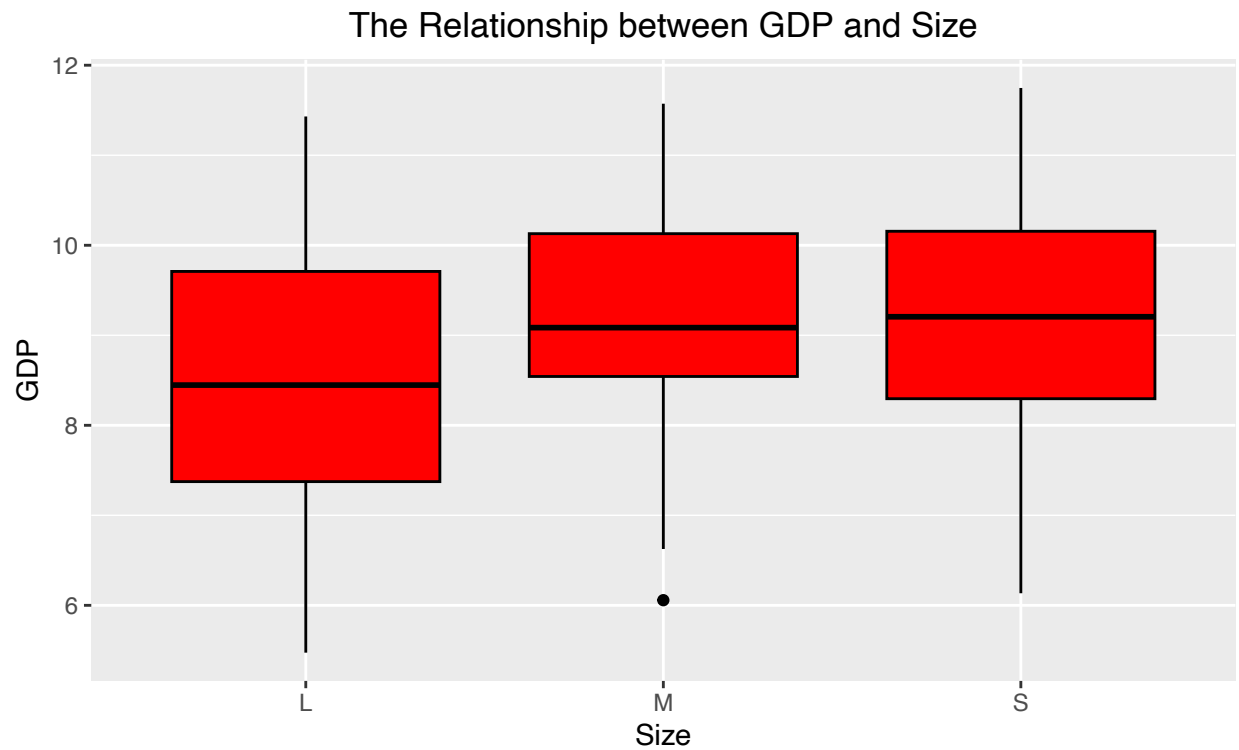
The Relationship between GDP and Debt



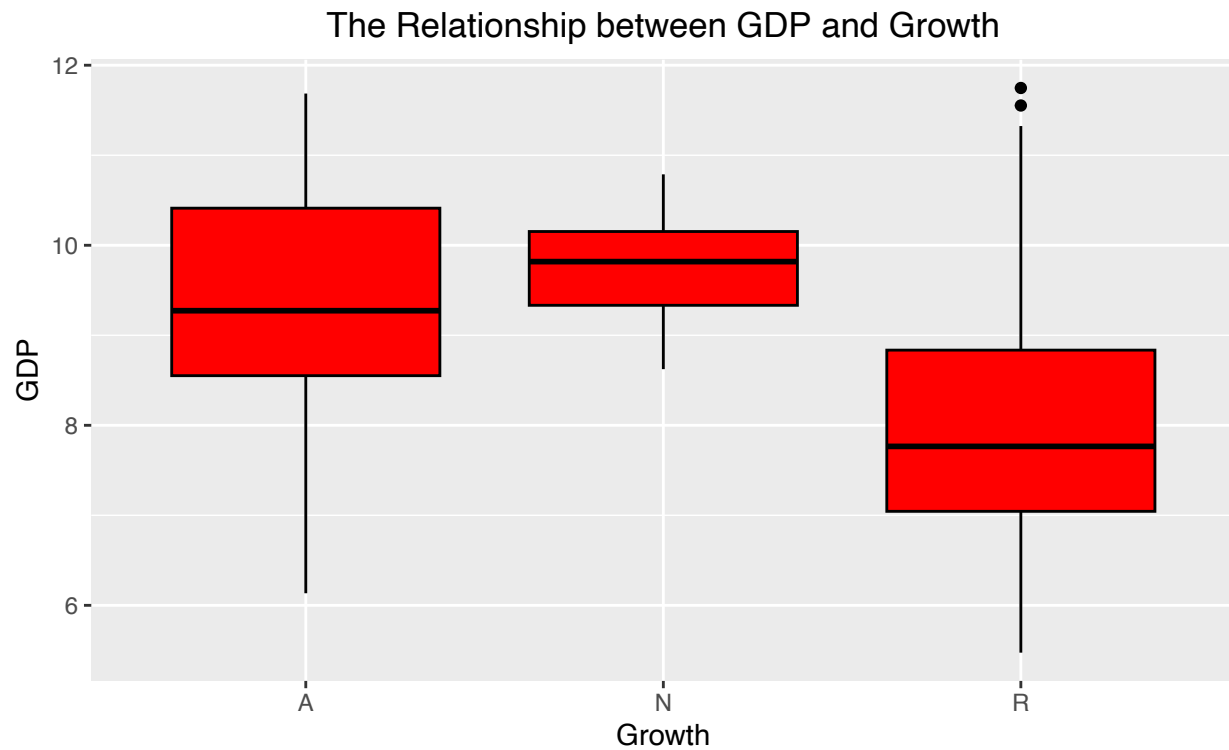
```
## [1] 0.2105487
```

```
## Warning: 'aes_string()' was deprecated in ggplot2 3.0.0.
## i Please use tidy evaluation idioms with 'aes()'.
## i See also 'vignette("ggplot2-in-packages")' for more information.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was
## generated.
```

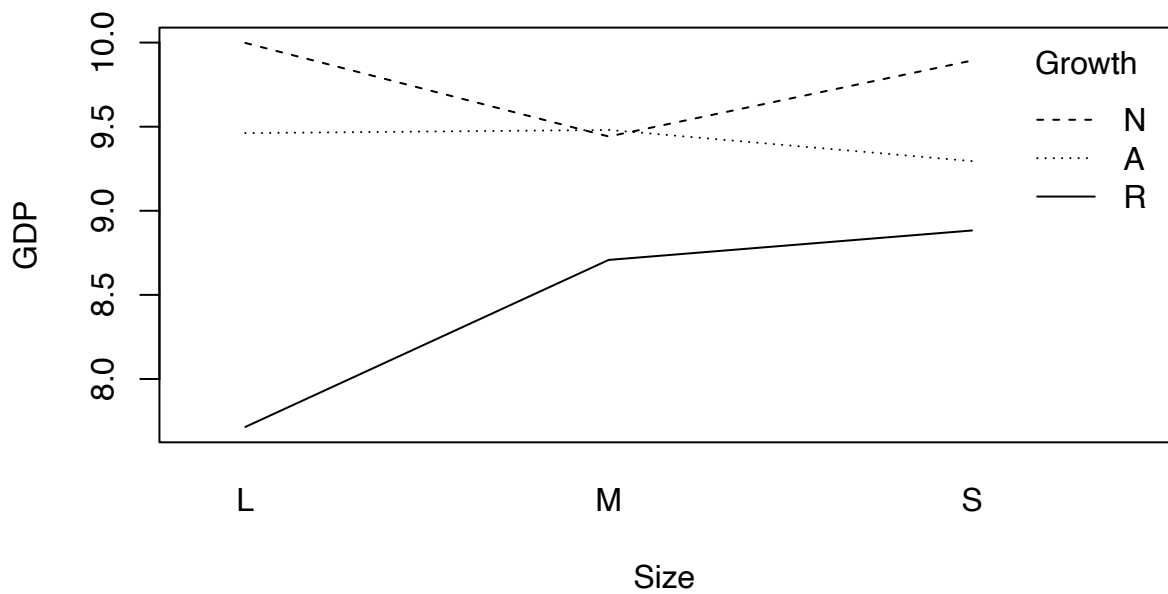
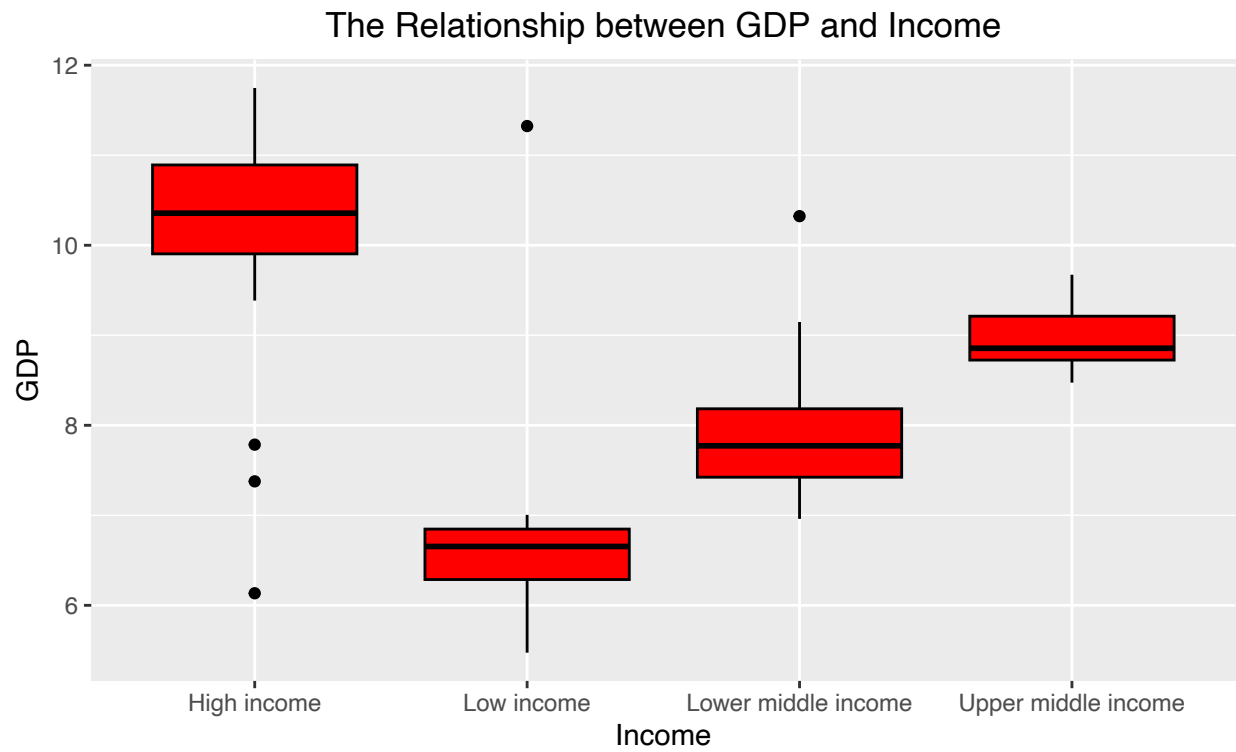
```
## $L
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  5.474  7.374   8.447   8.603  9.709  11.431
##
## $M
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  6.057  8.542   9.085   9.199 10.129  11.573
##
## $S
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  6.134  8.295   9.206   9.205 10.155  11.747
```



```
## $A
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  6.134  8.551   9.273   9.418 10.412  11.686
##
## $N
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  8.624  9.333   9.818   9.744 10.153  10.788
##
## $R
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  5.474  7.044   7.766   8.136  8.835  11.747
```

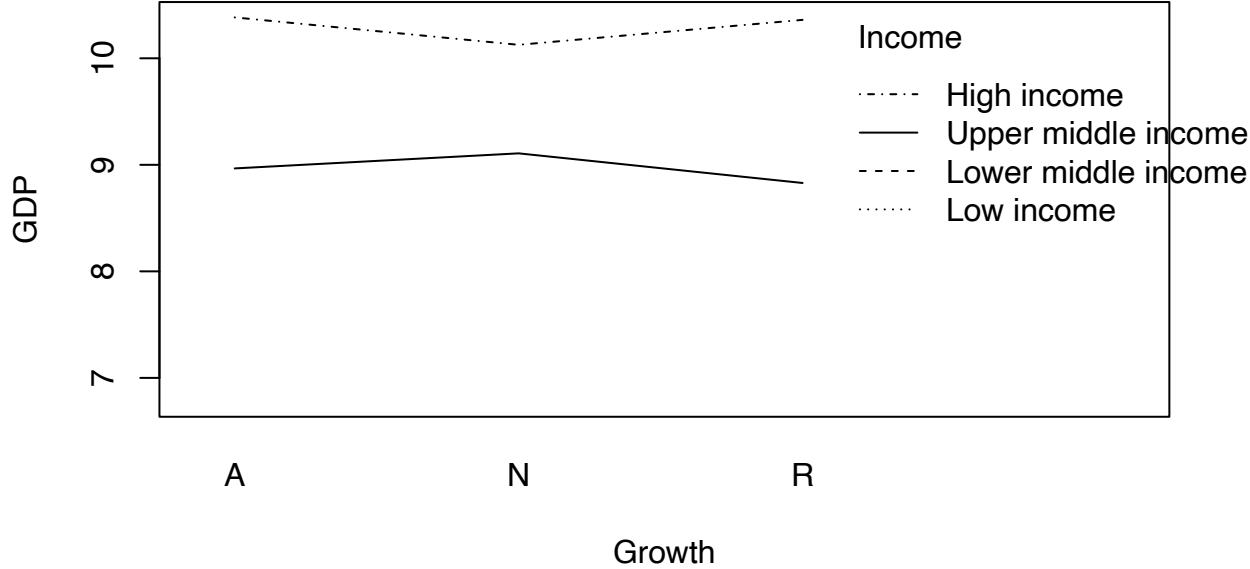


```
## $'High income'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  6.134  9.904  10.356  10.311  10.892  11.747
##
## $'Low income'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  5.474  6.286  6.654  6.779  6.847  11.324
##
## $'Lower middle income'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  6.961  7.424  7.772  7.867  8.183  10.323
##
## $'Upper middle income'
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  8.474  8.725  8.856  8.955  9.212  9.672
```

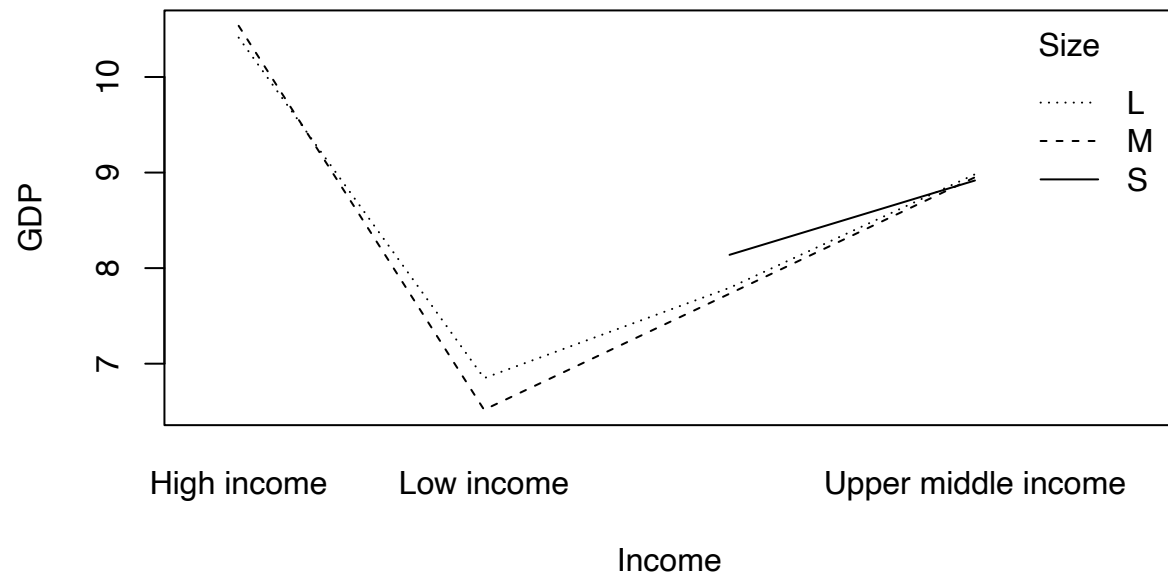


```
##
##      A  N  R
## L 33  9 46
```

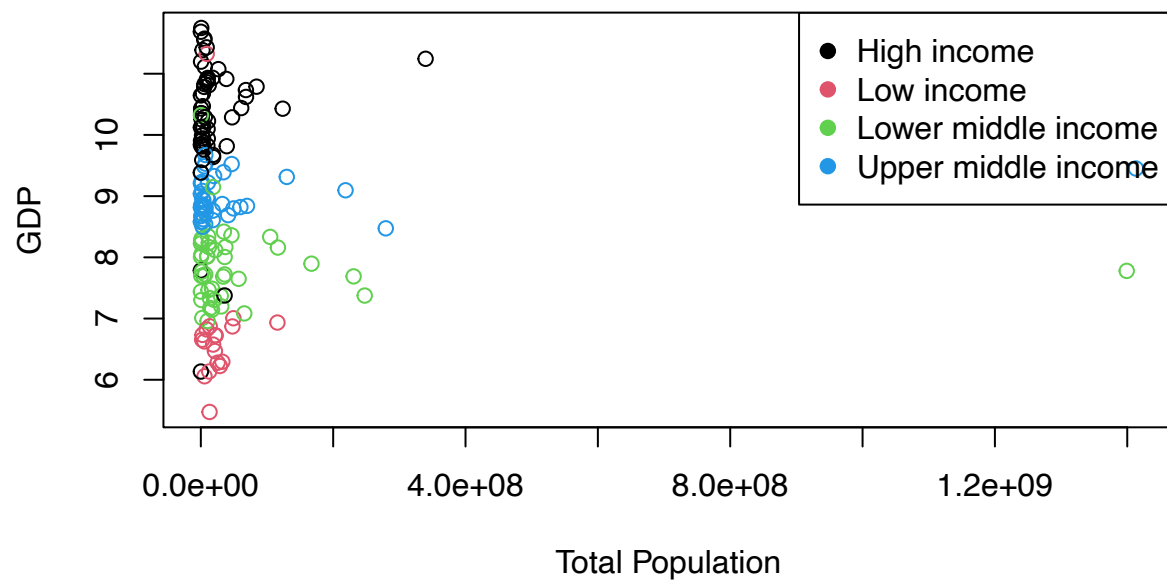
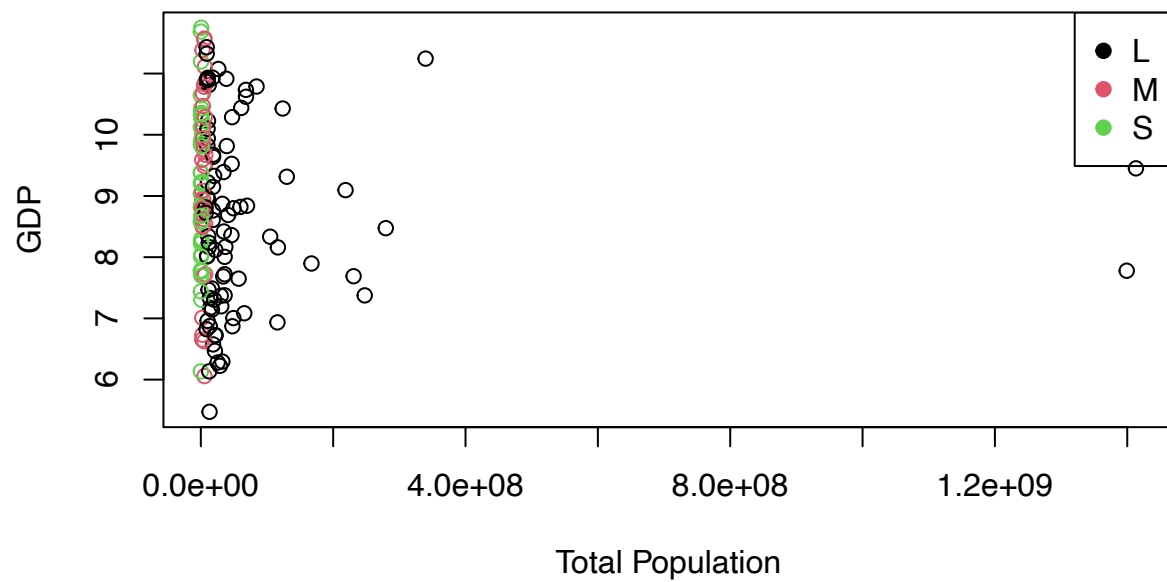
M 14 10 13
S 19 5 16

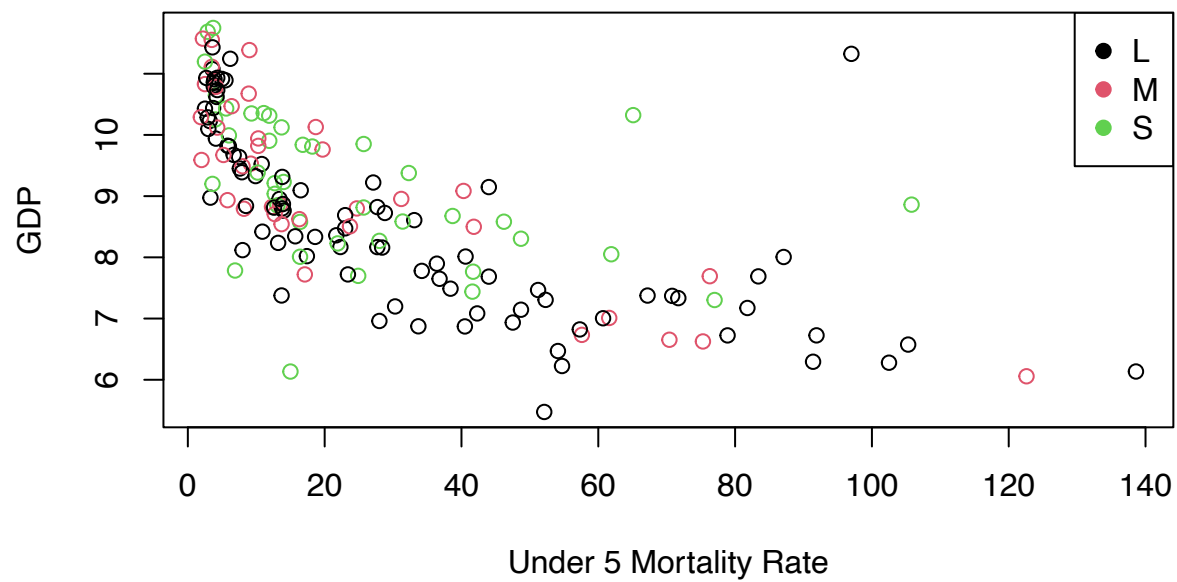
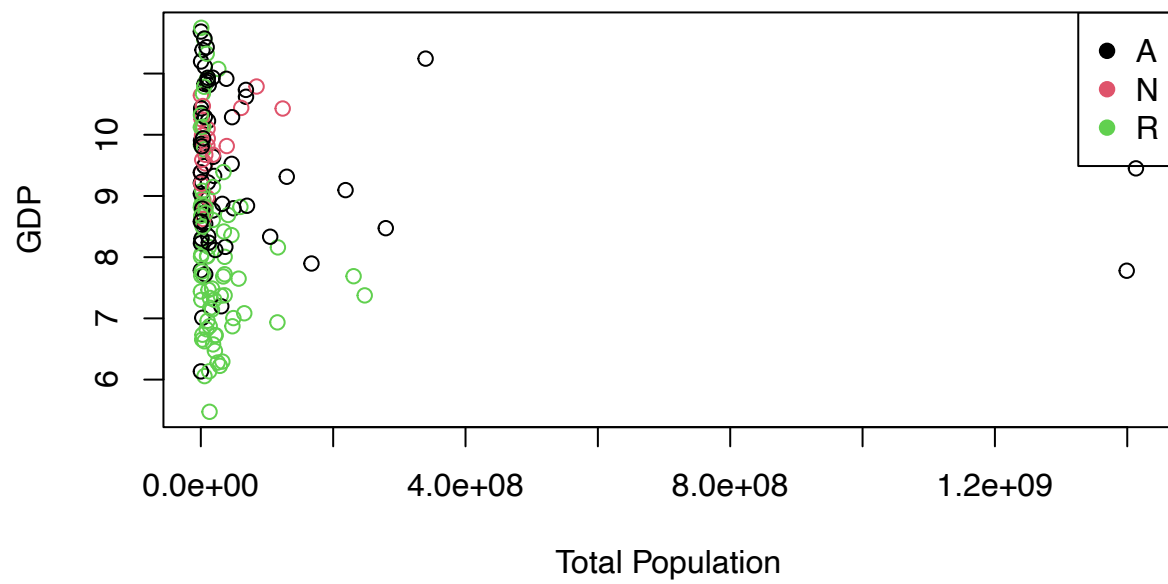


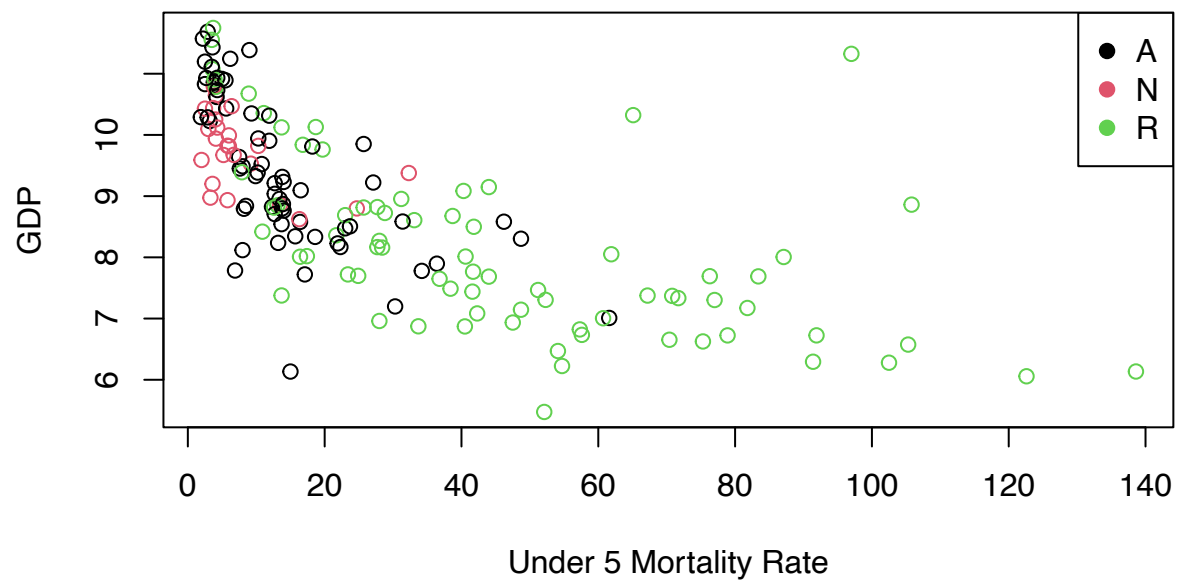
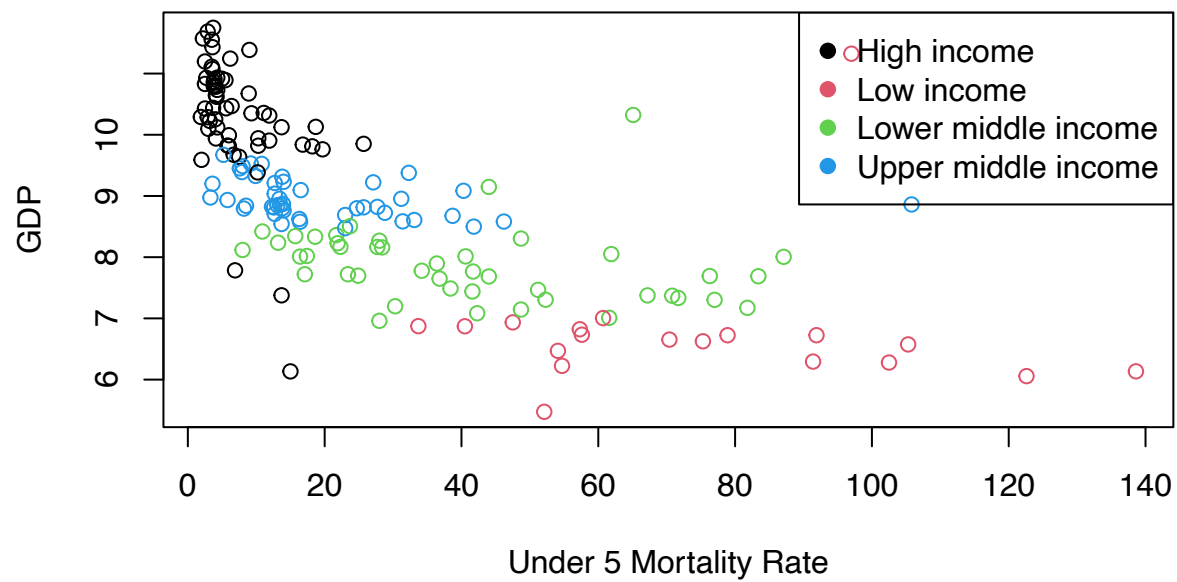
##		High income	Low income	Lower middle income	Upper middle income
##	A	30	0	13	23
##	N	15	0	0	9
##	R	12	19	31	13

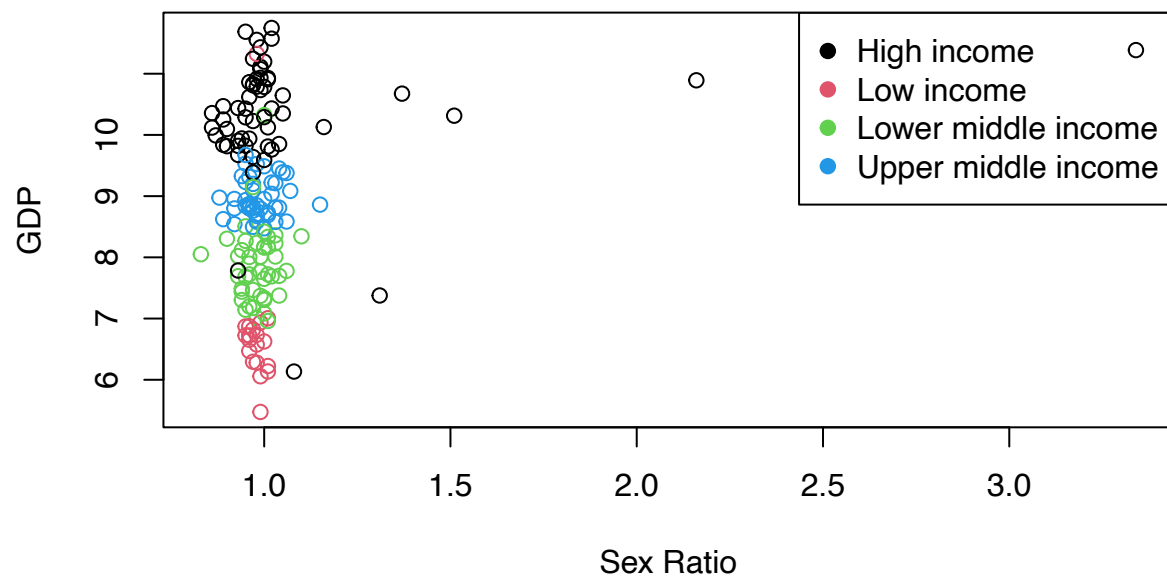
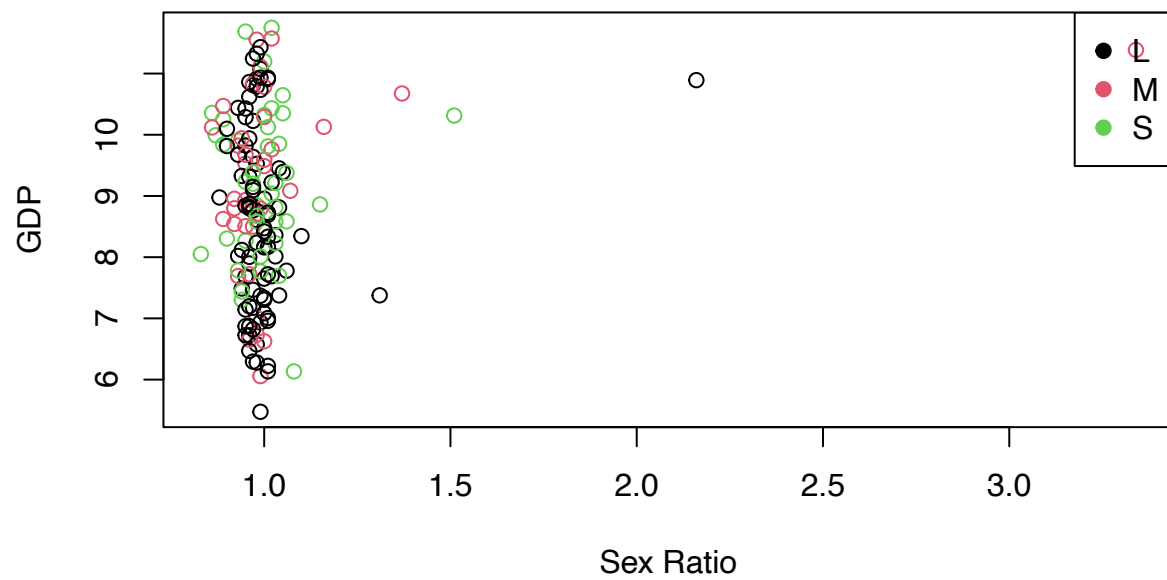


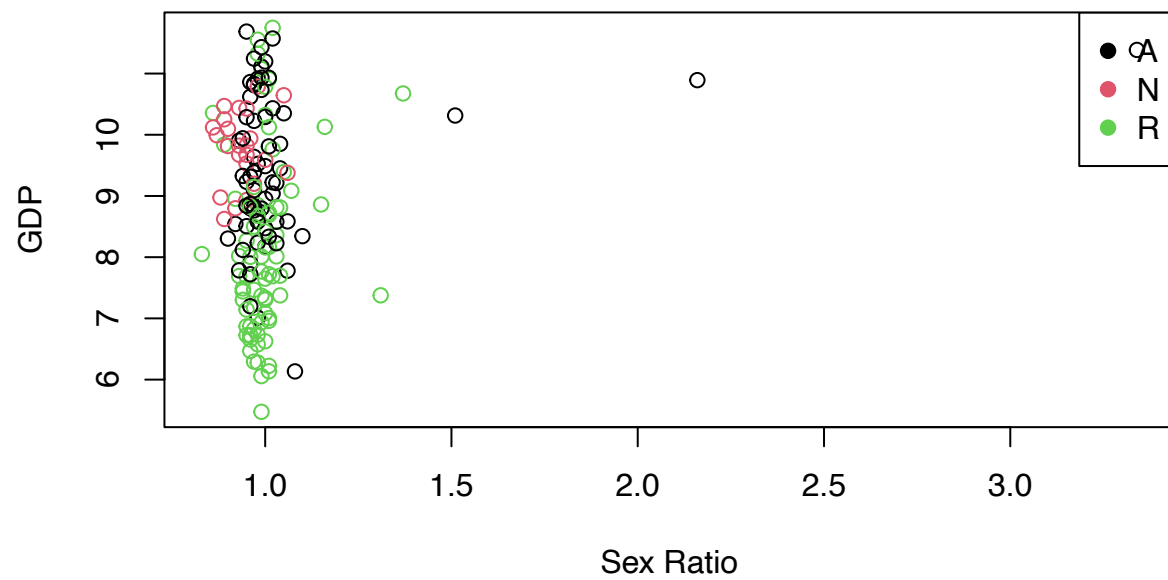
##		High income Low income Lower middle income Upper middle income			
##	L	24	15	30	19
##	M	15	4	4	14
##	S	18	0	10	12



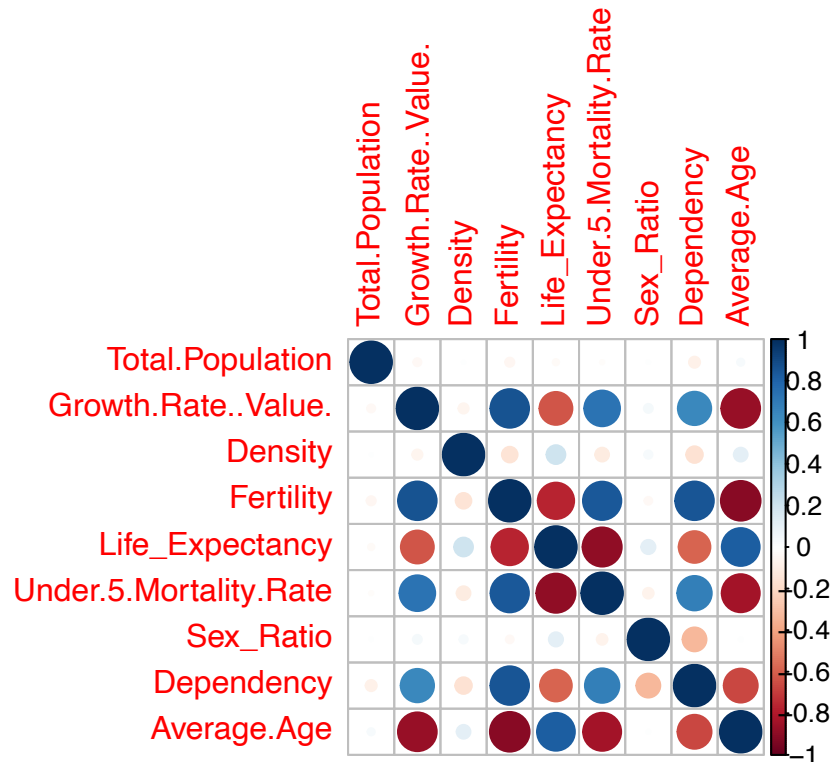








Multicollinearity/Variable Screening



```
##
## Call:
## lm(formula = GDP ~ . - Name - Size - Growth - Income, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.9879 -0.4764 -0.0382  0.4332  4.0862
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.721e+00  2.668e+00  -1.769  0.078797 .
## Total.Population -2.836e-10  4.413e-10  -0.643  0.521464
## Growth.Rate..Value.  1.761e-01  1.586e-01   1.110  0.268556
## Density        -5.045e-06  1.059e-04  -0.048  0.962078
## Fertility      -7.422e-01  3.337e-01  -2.225  0.027569 *
## Life_Expectancy  7.808e-02  2.811e-02   2.778  0.006154 **
## Under.5.Mortality.Rate  3.649e-03  6.826e-03   0.535  0.593681
## Sex_Ratio       1.537e+00  4.388e-01   3.503  0.000602 ***
## Dependency       3.991e-02  1.552e-02   2.572  0.011067 *
## Average.Age     1.635e+00  8.069e-01   2.026  0.044451 *
## Debt          -9.179e-04  2.423e-03  -0.379  0.705278
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8941 on 154 degrees of freedom
```

```
## Multiple R-squared:  0.642, Adjusted R-squared:  0.6187
## F-statistic: 27.62 on 10 and 154 DF,  p-value: < 2.2e-16
```

```
##      Total.Population  Growth.Rate..Value.      Density
##      1.020             5.747                1.127
##      Fertility         Life_Expectancy Under.5.Mortality.Rate
##      29.577            7.475                7.463
##      Sex_Ratio         Dependency          Average.Age
##      1.866             10.413              12.885
##      Debt
##      1.249
```

```
## [1] 7.8822
```

```
##
##                               Stepwise Selection Summary
## -----
##      Step      Variable      Added/      R-Square      Adj.      C(p)      AIC      RMSE
##      Step      Variable      Removed      R-Square      R-Square      C(p)      AIC      RMSE
## -----
##      1      Life_Expectancy      addition      0.560      0.557      28.2990      459.9667      0.9635
##      2      Average.Age      addition      0.608      0.603      9.5690      442.8290      0.9120
##      3      Sex_Ratio      addition      0.622      0.615      5.5970      438.8774      0.8985
##      4      Dependency      addition      0.629      0.619      4.7300      437.9425      0.8933
##      5      Fertility      addition      0.636      0.625      3.3950      436.4608      0.8867
## -----
```

```
##
##                               Selection Summary
## -----
##      Step      Variable      R-Square      Adj.      C(p)      AIC      RMSE
##      Step      Entered      R-Square      R-Square      C(p)      AIC      RMSE
## -----
##      1      Life_Expectancy      0.5599      0.5572      28.2994      459.9667      0.9635
##      2      Average.Age      0.6081      0.6033      9.5688      442.8290      0.9120
##      3      Sex_Ratio      0.6220      0.6150      5.5968      438.8774      0.8985
##      4      Dependency      0.6287      0.6194      4.7303      437.9425      0.8933
##      5      Fertility      0.6364      0.6250      3.3950      436.4608      0.8867
## -----
```

```
##
##                               Elimination Summary
## -----
##      Step      Variable      R-Square      Adj.      C(p)      AIC      RMSE
##      Step      Removed      R-Square      R-Square      C(p)      AIC      RMSE
## -----
##      1      Density      0.642      0.6212      9.0023      441.9169      0.8912
##      2      Debt      0.6416      0.6233      7.1533      440.0786      0.8888
##      3      Under.5.Mortality.Rate      0.6411      0.6251      5.3859      438.3274      0.8866
##      4      Total.Population      0.640      0.6264      3.8475      436.8200      0.8851
##      5      Growth.Rate..Value.      0.6364      0.625      3.3950      436.4608      0.8867
## -----
```

```
##
## Call:
## lm(formula = GDP ~ . - Name - Size - Growth - Income - Fertility -
##      Density - Growth.Rate..Value. - Total.Population - Under.5.Mortality.Rate -
##      Debt - Dependency, data = data)
##
## Coefficients:
##      (Intercept)  Life_Expectancy      Sex_Ratio    Average.Age
##      -4.71826      0.08205      0.80012      1.97195

##
## Call:
## lm(formula = GDP ~ . - Name - Size - Growth - Income - Fertility -
##      Density - Growth.Rate..Value. - Total.Population - Under.5.Mortality.Rate -
##      Debt - Dependency, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.4640 -0.4363 -0.0852  0.4354  4.5822
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -4.71826    0.85327  -5.530 1.27e-07 ***
## Life_Expectancy  0.08205    0.01871   4.386 2.08e-05 ***
## Sex_Ratio      0.80012    0.32903   2.432  0.0161 *
## Average.Age     1.97195    0.40628   4.854 2.84e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Residual standard error: 0.8985 on 161 degrees of freedom
## Multiple R-squared:  0.622, Adjusted R-squared:  0.615
## F-statistic: 88.31 on 3 and 161 DF, p-value: < 2.2e-16

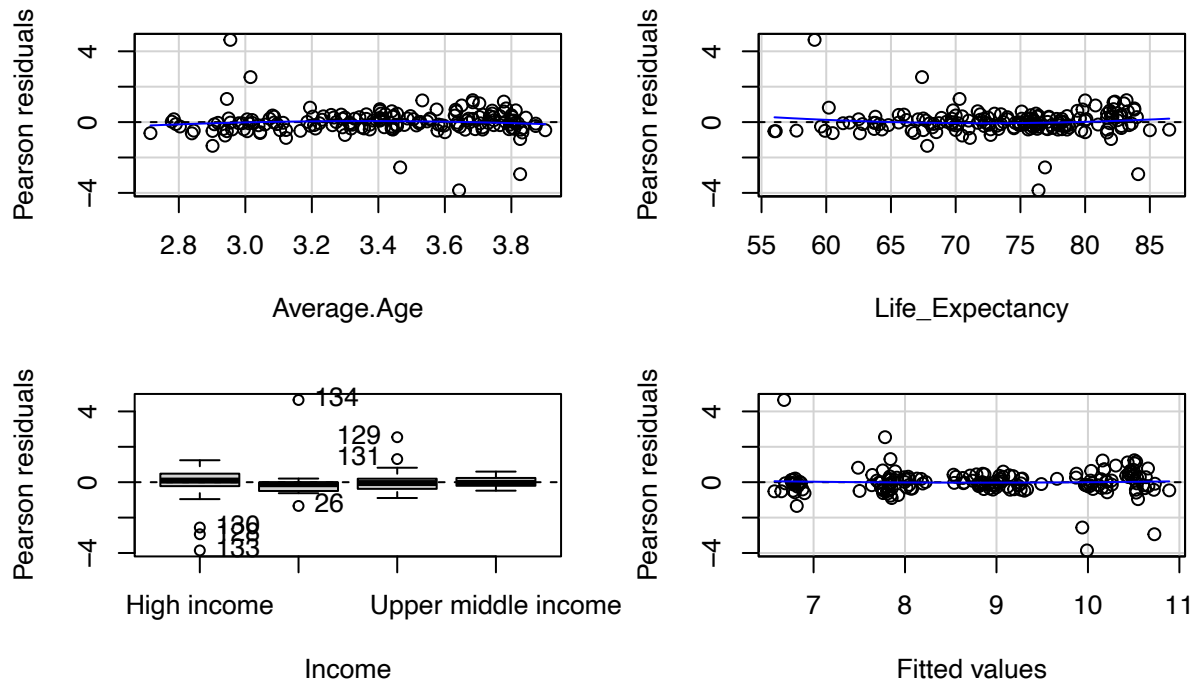
## Life_Expectancy      Sex_Ratio    Average.Age
##           3.278           1.039           3.235

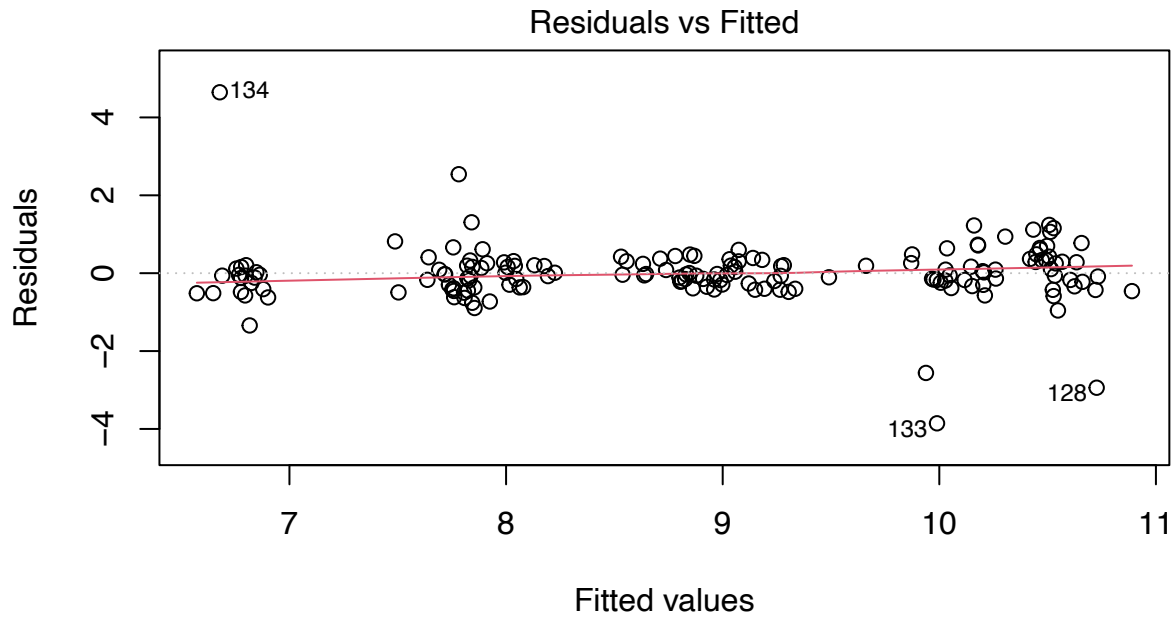
## [1] 2.517333
```

Pre-Residual Removal Model

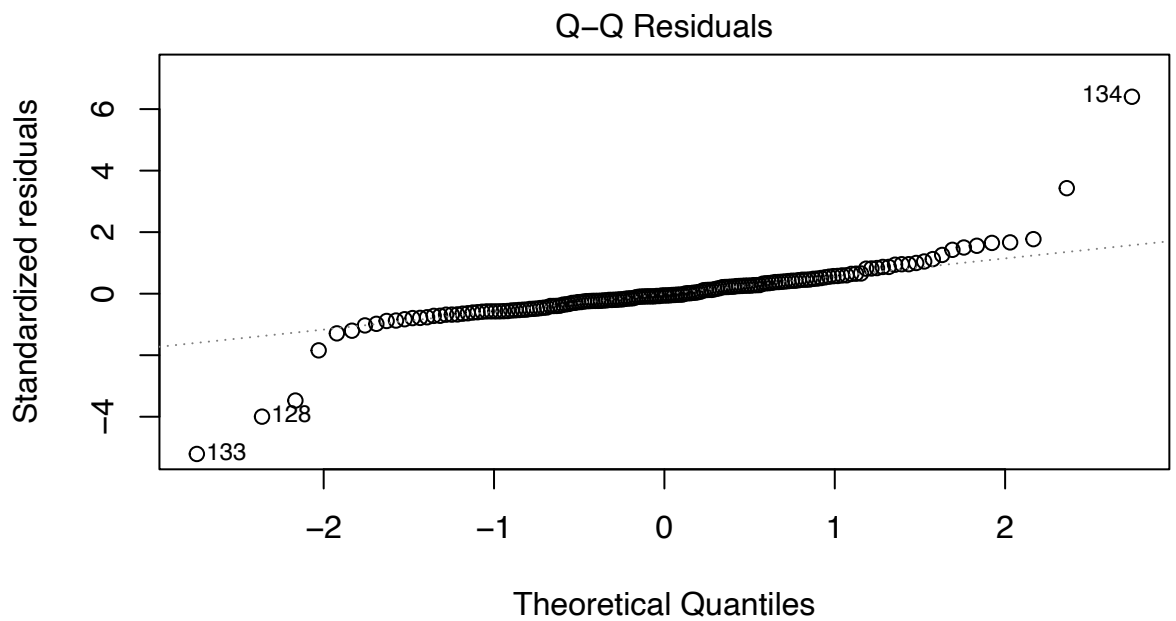
```
##
## Call:
## lm(formula = GDP ~ Average.Age + Life_Expectancy + Average.Age *
##     Life_Expectancy + Income, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.8552 -0.2987 -0.0421  0.2809  4.6460
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25.59939    9.27481   2.760  0.00646 **
## Average.Age       -5.75669    2.87551  -2.002  0.04700 *
## Life_Expectancy   -0.22559    0.12645  -1.784  0.07634 .
## IncomeLow income   -2.75660    0.34318  -8.033 2.07e-13 ***
## IncomeLower middle income -1.75484    0.24629  -7.125 3.48e-11 ***
## IncomeUpper middle income -0.87003    0.19900  -4.372 2.22e-05 ***
## Average.Age:Life_Expectancy  0.08119    0.03847   2.110  0.03640 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7528 on 158 degrees of freedom
## Multiple R-squared:  0.7396, Adjusted R-squared:  0.7297
## F-statistic: 74.79 on 6 and 158 DF, p-value: < 2.2e-16
```


Assumptions/Residual Analysis/Influential Observation Analysis



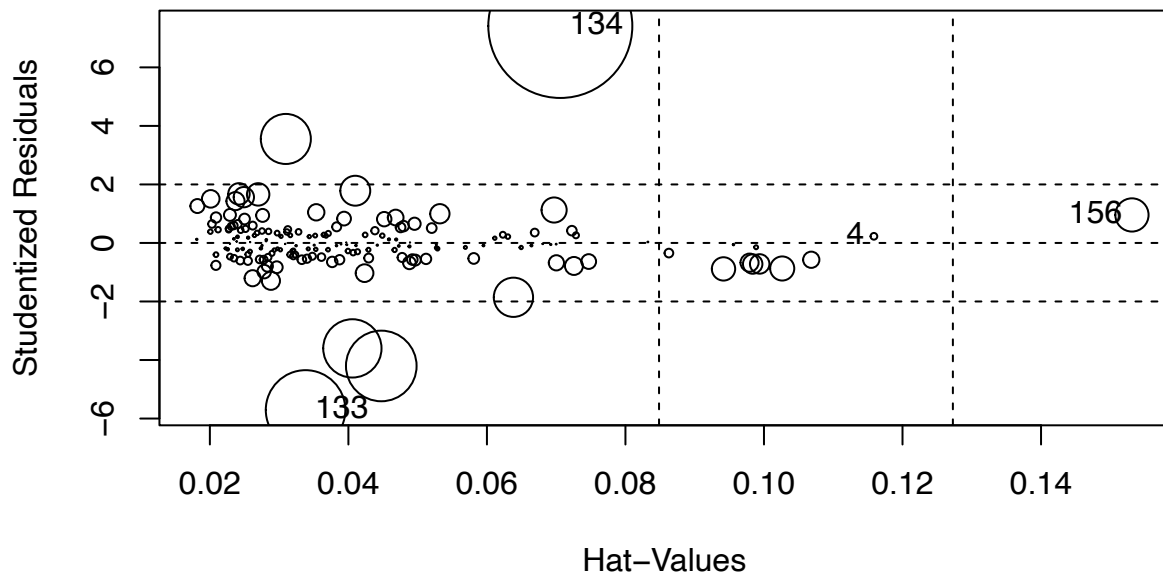
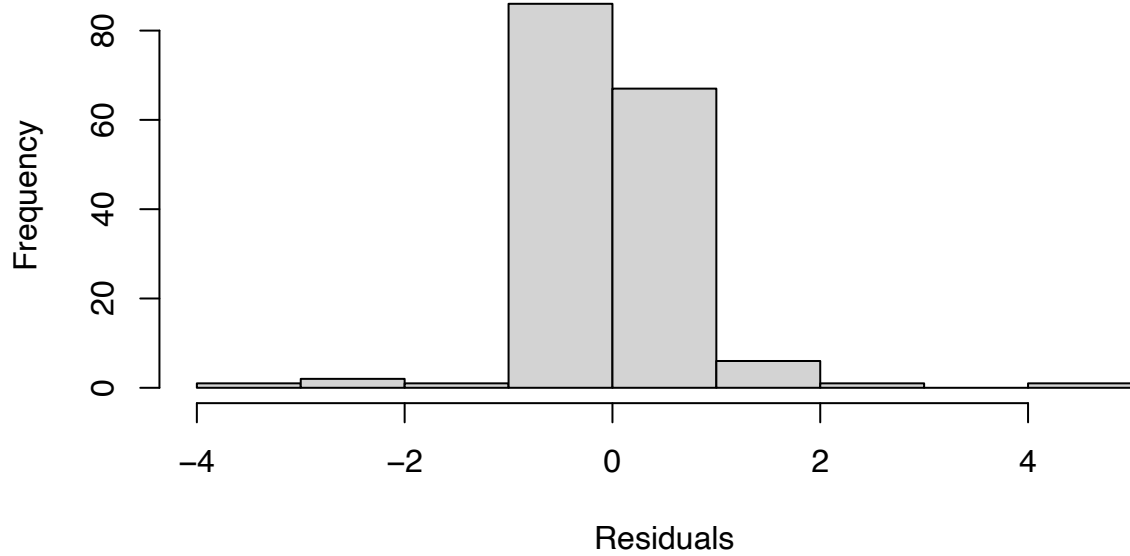


lm(GDP ~ Average.Age + Life_Expectancy + Average.Age * Life_Expectancy + ln ...)



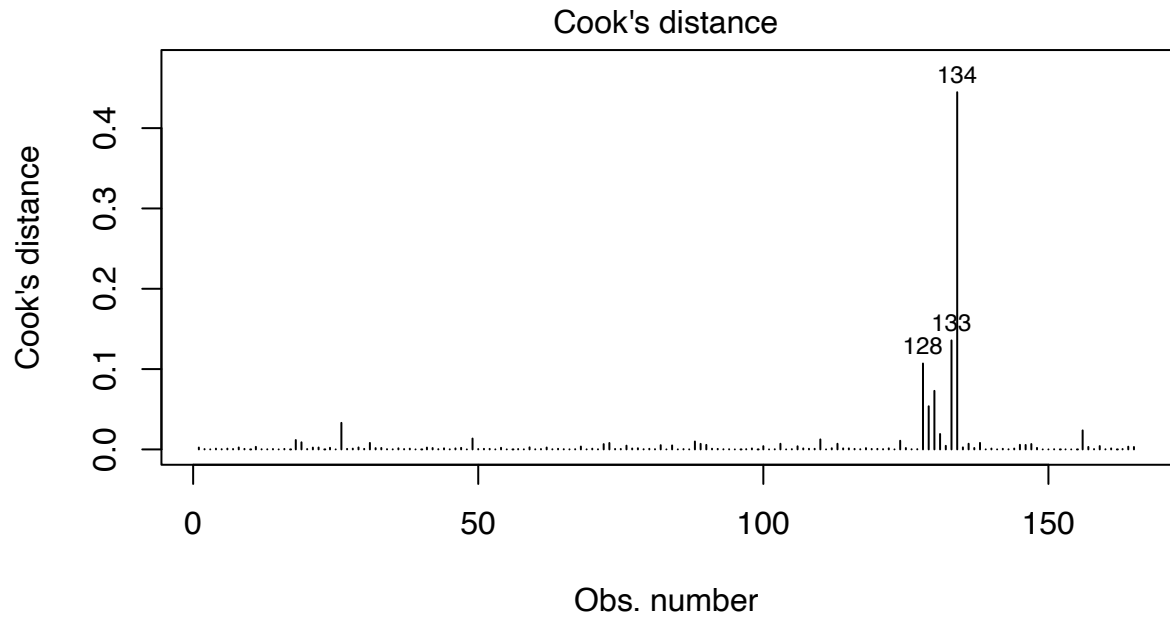
lm(GDP ~ Average.Age + Life_Expectancy + Average.Age * Life_Expectancy + ln ...)

The Histogram of Residuals of our Quantitative Model



```
##      StudRes      Hat      CookD
## 4      0.2270134 0.11586985 0.0009706756
## 133 -5.7069476 0.03381706 0.1357298668
```

```
## 134 7.4152335 0.07059705 0.4447179793
## 156 0.9570428 0.15312495 0.0236713177
```



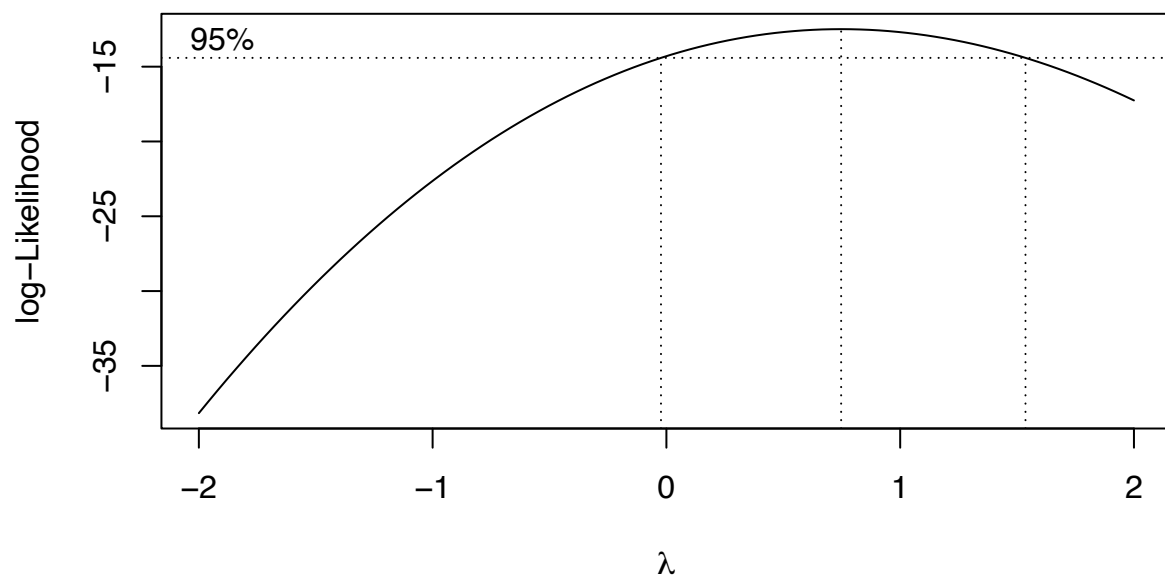
lm(GDP ~ Average.Age + Life_Expectancy + Average.Age * Life_Expectancy + ln ...

```
##          4          128          133          134          156
## 0.08218236 -0.90978337 -1.06768253 2.04369504 0.40695339

## [1] 0.06342613 -0.57164490 0.52941454 0.47810477 -0.19417639

## [1] -0.00269222
```

Box Cox Plot



```
## [1] 0.3943296
```

Confidence and Prediction Intervals

```
# We can therefore, employ this new and final model for estimation and
## prediction using the metrics below.

# E(y) confidence interval
conf_interval <- predict(finalModel, interval = "confidence", level = 0.95,
                          newdata = test_data)

conf_interval
```

	fit	lwr	upr
2	8.223635	8.002709	8.444562
3	10.833050	10.599054	11.067046
10	8.974368	8.818028	9.130709
15	8.972765	8.759713	9.185816
18	7.727090	7.385143	8.069037
20	7.913944	7.741419	8.086469
28	7.826610	7.644841	8.008378
29	7.684491	7.419507	7.949475
33	10.352846	10.201828	10.503864
45	8.951398	8.771367	9.131429
55	6.585890	6.314379	6.857402
56	8.781440	8.569424	8.993456
57	10.631667	10.449865	10.813469
58	7.821522	7.647344	7.995700
59	10.625109	10.447719	10.802499
66	7.886086	7.718331	8.053841
70	8.860997	8.702888	9.019107
73	10.395311	10.152000	10.638622
80	7.678842	7.485353	7.872331
85	6.377456	6.116422	6.638490
88	10.658407	10.493546	10.823268
96	7.652781	7.464847	7.840716
98	8.873747	8.715846	9.031648
100	7.851593	7.632793	8.070393
105	9.559761	9.230475	9.889048

```

108 10.580063 10.420816 10.739309
116 10.171166 9.960258 10.382074
119 8.586129 8.385028 8.787229
124 10.319898 10.148162 10.491634
126 6.512186 6.249504 6.774868
130 10.080140 9.869506 10.290774
134 6.334268 6.049178 6.619358
145 10.784487 10.584814 10.984159

```

```
confint(finalModel)
```

	2.5 %	97.5 %
(Intercept)	9.70065800	37.0707114302
Average.Age	-9.57727572	-1.0708742289
Life_Expectancy	-0.36891085	-0.0006319617
IncomeLow income	-3.56869768	-2.5950085314
IncomeLower middle income	-2.22729666	-1.5397335200
IncomeUpper middle income	-1.25379182	-0.7026959425
Average.Age:Life_Expectancy	0.01638531	0.1288613515

Final Model and Summary

```
##
## Call:
## lm(formula = GDP ~ Average.Age + Life_Expectancy + Average.Age *
##     Life_Expectancy + Income, data = subdata)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.70317 -0.30829 -0.02124  0.24556  2.56927
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      23.38568     6.92706   3.376 0.000932 ***
## Average.Age       -5.32407     2.15288  -2.473 0.014492 *
## Life_Expectancy   -0.18477     0.09321  -1.982 0.049229 *
## IncomeLow income   -3.08185     0.24643 -12.506 < 2e-16 ***
## IncomeLower middle income -1.88352     0.17401 -10.824 < 2e-16 ***
## IncomeUpper middle income -0.97824     0.13948  -7.014 7.01e-11 ***
## Average.Age:Life_Expectancy  0.07262     0.02847   2.551 0.011717 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5212 on 153 degrees of freedom
## Multiple R-squared:  0.8733, Adjusted R-squared:  0.8684
## F-statistic: 175.8 on 6 and 153 DF,  p-value: < 2.2e-16

## [1] 0.2716175

““
```


Appendix D: References

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