

INDIAN INSTITUTE OF TECHNOLOGY, KANPUR

INDIAN INSTITUTE OF TECHNOLOGY, KANPUR

DEPARTMENT  
OF  
INDUSTRIAL AND MANAGEMENT ENGINEERING



MBA652A

STATISTICAL MODELLING FOR BUSINESS ANALYSIS

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## PANEL DATA ANALYSIS

### STATISTICAL ANALYSIS ON FACTORS AFFECTING LIFE EXPECTENCY

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## *The Research*

Although there have been lot of studies undertaken in the past on factors affecting life expectancy considering demographic variables, income composition and mortality rates. It was found that affect of immunization and human development index was not taken into account in the past. Also, some of the past research was done considering multiple linear regression based on data set of one year for all the countries. Hence, this gives motivation to resolve both the factors stated previously by formulating a regression model based on mixed effects model and multiple linear regression while considering data from a period of 2000 to 2015 for all the countries. Important immunization like Hepatitis B, Polio and Diphtheria will also be considered. In a nutshell, this study will focus on immunization factors, mortality factors, economic factors, social factors and other health related factors as well. Since the observations this dataset are based on different countries, it will be easier for a country to determine the predicting factor which is contributing to lower value of life expectancy. This will help in suggesting a country which area should be given importance in order to efficiently improve the life expectancy of its population.

## *Executive Summary*

The project relies on accuracy of data. The Global Health Observatory (GHO) data repository under World Health Organization (WHO) keeps track of the health status as well as many other related factors for all countries. The data-sets are made available to public for the purpose of health data analysis. The data-set related to life expectancy, health factors for 193 countries has been collected from the same WHO data repository website and its corresponding economic data was collected from United Nation website. Among all categories of health-related factors only those critical factors were chosen which are more representative. It has been observed that in the past 15 years, there has been a huge development in health sector resulting in improvement of human mortality rates especially in the developing nations in comparison to the past 30 years. Therefore, in this project we have considered data from year 2000-2015 for 193 countries for further analysis. The individual data files have been merged together into a single data-set. On initial visual inspection of the data showed some missing values. As the data-sets were from WHO, we found no evident errors. Missing data was handled in R software by using Missmap command. The result indicated that most of the missing data was for population, Hepatitis B and GDP. The missing data were from less known countries like Vanuatu, Tonga, Togo, Cabo Verde etc. Finding all data for these countries was difficult and hence, it was decided that we exclude these countries from the final model data-set. The final merged file(final dataset) consists of 22 Columns and 2938 rows which meant 20 predicting variables. All predicting variables was then divided into several broad categories: Immunization related factors, Mortality factors, Economical factors and Social factors.

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# The Data

## Source

<https://www.kaggle.com/kumarajarshi/life-expectancy-who?select=Life+Expectancy+Data.csv>

## Description

Our data set is a Panel Data set. A panel data set, also called longitudinal data set, is one that studies the same parameters at different points in time. It has a total of 2398 observations studied over 16 years for 193 Countries. The balance in the dataset is established as there are no missing values.

It has statistics about the spot exchange rates (the logarithmic function of the exchange rate), the long and short-term interest rates in the US, the long and the short-term interest rates in the respective country, the price level (the logarithmic function of price level) of a basket of goods in the respective country.

## Variables

Country	Year
Status	Life.expectancy
Adult.Mortality	infant.deaths
Alcohol	percentage.expenditure
Hepatitis.B	Measles
BMI	under.five.deaths
Polio	Total.expenditure
Diphtheria	HIV.AIDS
GDP	Population
thinness..1.19.years	thinness.5.9.years
Income.composition.of.resources	Schooling
avg	

## Dependent Variables and Independent Variables

### Dependent Variable: Life Expectancy

The dependent variable is the one being tested and measured and whose value may be affected by the change of other independent variables. We will examine the effect of other variables on the life expectancy

### Independent Variables:

An independent variable is a variable that is changed or varied in an experiment to examine the effect on dependent variable. In our analysis, we are taking the other above variables as the independent ones. We believe that comparing these variables with dependent variable might suggest the effect that they have on the dependent variable.

## Descriptive Statistics

Country		Year		Status
Afghanistan	: 16	2013	: 193	Developed: 508
Albania	: 16	2000	: 183	Developing: 2417
Algeria	: 16	2002	: 183	
Angola	: 16	2004	: 183	
Antigua and Barbuda	: 16	2005	: 183	
Argentina	: 16	2006	: 183	
(Other)	:2829	(Other)	:1817	
Life.expectancy		Adult.Mortality		infant.deaths
Min.	:36.30	Min.	: 1.0	Min. : 0.00
1st Qu.:	63.00	1st Qu.:	74.0	1st Qu.: 0.00
Median	:72.00	Median	:144.0	Median : 3.00
Mean	:69.19	Mean	:165.2	Mean : 30.39
3rd Qu.:	75.60	3rd Qu.:	228.0	3rd Qu.: 22.00
Max.	:89.00	Max.	:723.0	Max. :1800.00
percentage.expenditure		Measles		BMI
				Total.expenditure

Min. : 0.000	Min. : 0	Min. : 1.00	Min. : 0.370
1st Qu.: 5.235	1st Qu.: 0	1st Qu.:19.10	1st Qu.: 4.260
Median : 66.378	Median : 17	Median :43.50	Median : 5.710
Mean : 741.532	Mean : 2427	Mean :38.24	Mean : 5.913
3rd Qu.: 443.475	3rd Qu.: 362	3rd Qu.:56.10	3rd Qu.: 7.470
Max. :19479.912	Max. :212183	Max. :87.30	Max. :17.600

Diphtheria	HIV.AIDS	thinness..1.19.years
Min. : 2.00	Min. : 0.100	Min. : 0.100
1st Qu.:78.00	1st Qu.: 0.100	1st Qu.: 1.600
Median :93.00	Median : 0.100	Median : 3.300
Mean :82.22	Mean : 1.746	Mean : 4.892
3rd Qu.:97.00	3rd Qu.: 0.800	3rd Qu.: 7.200
Max. :99.00	Max. :50.600	Max. :27.700

Income.composition.of.resources	Schooling
Min. :0.0000	Min. : 0.00
1st Qu.:0.4940	1st Qu.:10.10
Median :0.6810	Median :12.30
Mean :0.6302	Mean :12.02
3rd Qu.:0.7810	3rd Qu.:14.30
Max. :0.9480	Max. :20.70

## Heterogeneity

The data is checked for heterogeneity, the functions “ggplot”, “plotmeans” were used. Relationship between Y and X variables for the cross- section of the 17 entities quarterly across 26 years. Scatter plots have been used to identify any existence of correlation between the quantitative variables.

Figure 1: Variation of Life Expectancy across countries

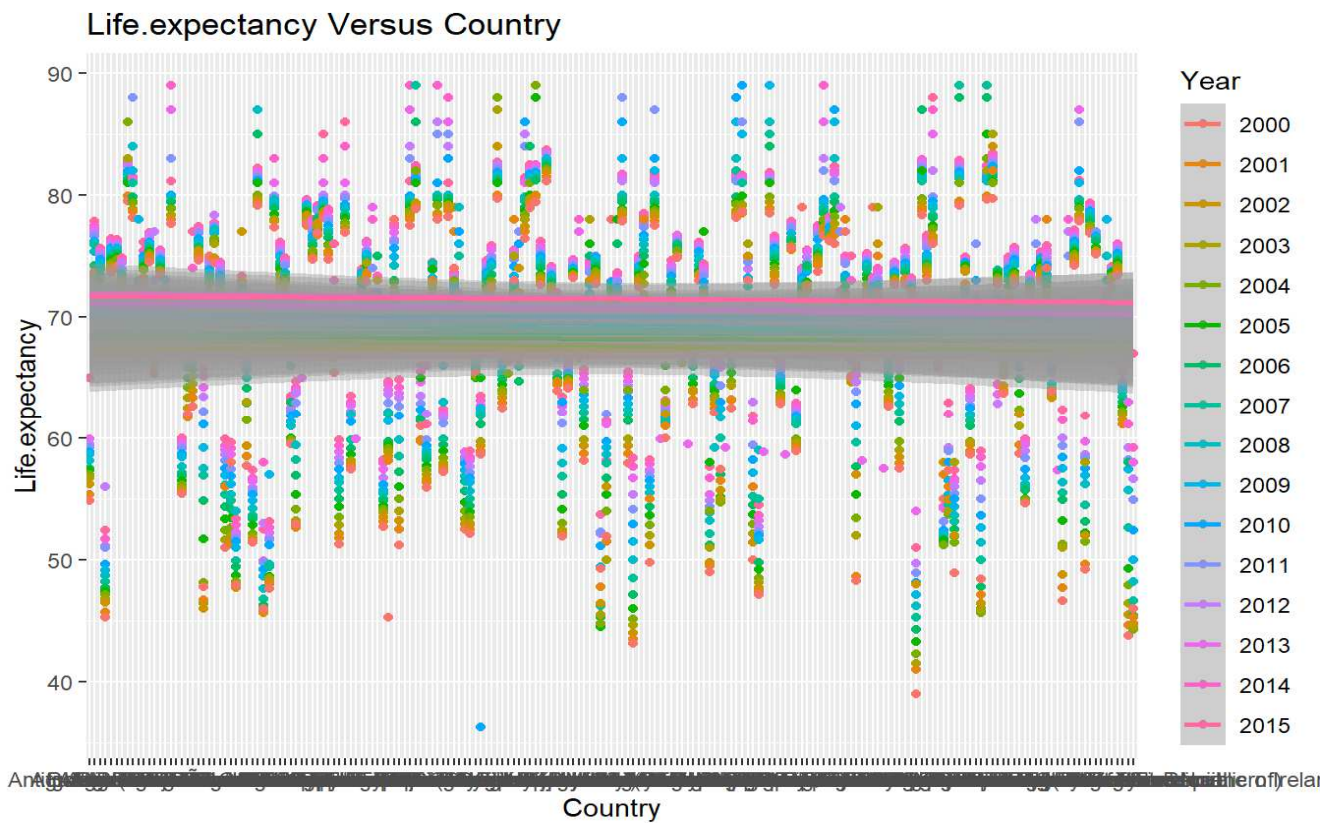


Figure 2: Heterogeneity across Years

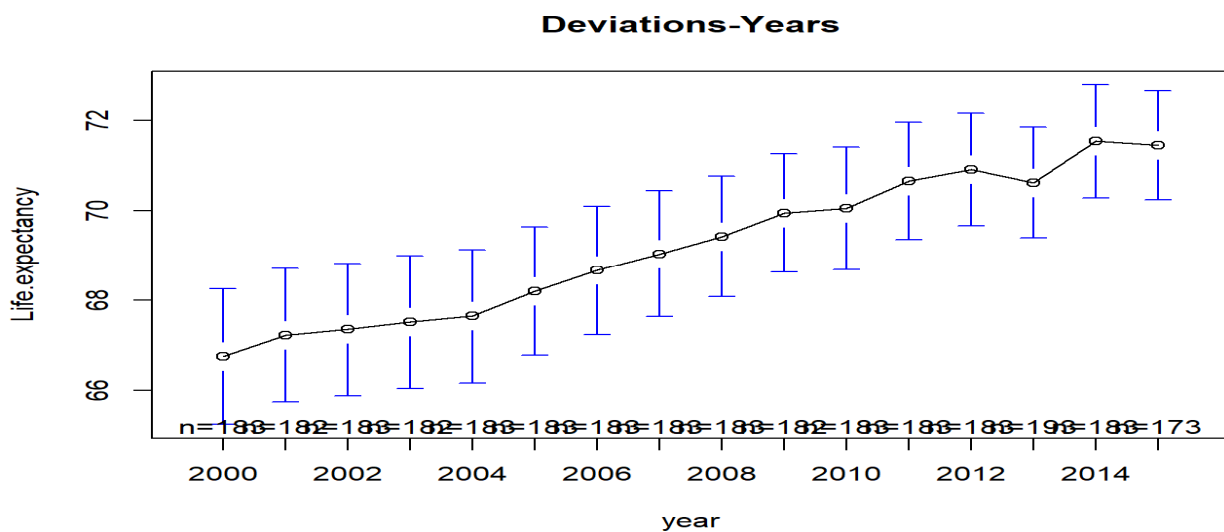
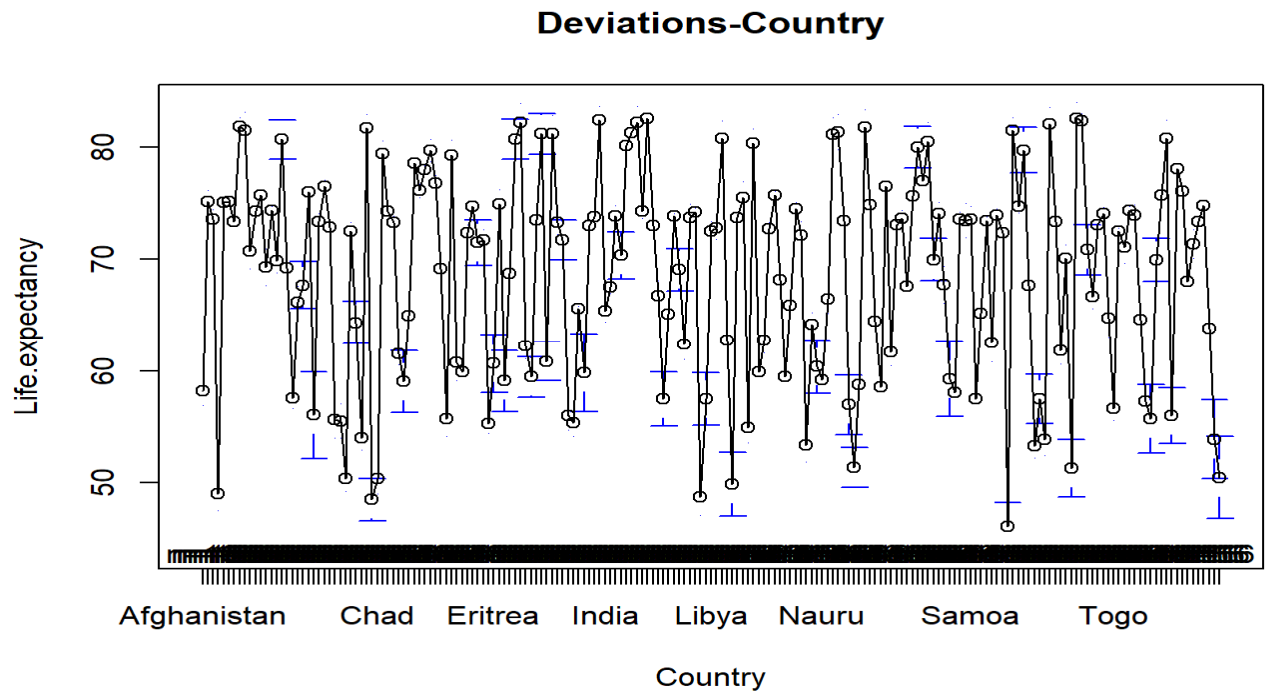
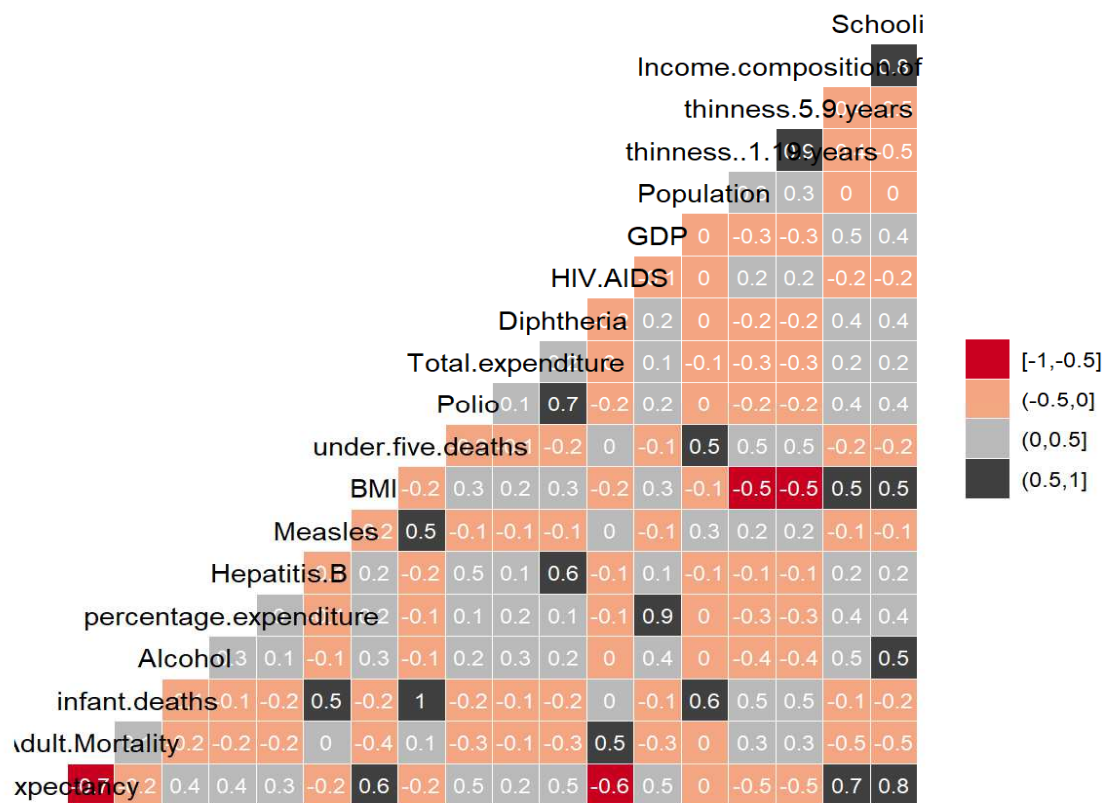


Figure 3: Heterogeneity across Countries



## Correlation Matrix :



With the help of Correlation matrix , we removed Hepatitis , under. five, Polio, GDP, Population, thinness 5 years because of high correlation .



## Panel Data Analysis

Panel data sets, or a longitudinal data set, is a set of cross-sectional data collected for the same parameters across various years. For the analysis of the panel dataset in hand, we have employed the following regression techniques:

1. Pooled Regression
2. Fixed Effects Regression
  - I. Entity Fixed Effects
    - i. Entity Fixed Effects (Entity demeaned)
    - ii. Entity Fixed Effects with significant Variables
  - II. Entity Fixed Effects using Binary Variables
  - III. Time Fixed Effects
  - IV. Entity and Time Fixed Effects
3. Random Effect model

The statistical modelling has been done in the R software using the PLM function. This function is used for linear modelling of panel data. or first model we regress Life expectancy against schooling.

### Pooled Regression Models

The pooled regression model ignores any differences over entities or time and treats each data as a separate entity. We use the “plm” function to carry out pooled regression on the data in R. We made 7 models by forward regression starting from schooling because of high significance. We are showing the final pooled model.

### Final Model

```
Pooling Model

## plm(formula = Life.expectancy ~ Schooling + Income.composition.of.resources +
##      BMI + percentage.expenditure + HIV.AIDS + thinness..1.19.years +
##      Diphtheria, data = Pooled_Data, model = "pooling")
## Unbalanced Panel: n = 193, T = 1-16, N = 2925
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -27.77658  -2.78594   0.12939   2.65059   27.05782
##
## Coefficients:
##
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept)    4.8030e+01  4.9899e-01  96.2540 < 2.2e-16
## Schooling       8.1468e-01  4.4181e-02  18.4393 < 2.2e-16
```

```

## Income.composition.of.resources  7.6556e+00  7.0037e-01  10.9308 < 2.2e-16
## BMI                             6.1836e-02  5.6500e-03  10.9444 < 2.2e-16
## percentage.expenditure          5.1861e-04  4.8212e-05  10.7569 < 2.2e-16
## HIV.AIDS                       -6.8896e-01  1.8136e-02 -37.9883 < 2.2e-16
## thinness..1.19.years            -1.2304e-01  2.4047e-02  -5.1167 3.311e-07
## Diphtheria                     6.8059e-02  4.1005e-03  16.5980 < 2.2e-16
##
## (Intercept)                    ***
## Schooling                      ***
## Income.composition.of.resources ***
## BMI                            ***
## percentage.expenditure         ***
## HIV.AIDS                       ***
## thinness..1.19.years           ***
## Diphtheria                     ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    264660
## Residual Sum of Squares: 66239
## R-Squared:              0.74972
## Adj. R-Squared: 0.74912
## F-statistic: 1248.3 on 7 and 2917 DF, p-value: < 2.22e-16

```

## Analysis

This model has a decent adjusted  $R^2$ , but we'll try to build some more models which will handle the time or entity specific effects, that we might have missed in this model.

## Fixed Regression Models

### Entity Fixed Effects (Entity demeaned)

In entity fixed models, each country is considered as a separate entity. In entity demeaned method, all the values across the years, for each country are averaged and the average is then subtracted from their respective observations. This way, we obtain the entity demeaned observations which are then regressed. Here, we have regressed the life expectancy over all the independent variables. We started with one variable and by forward regression our final model results are described below

### Final Model

```
## Oneway (individual) effect Within Model

## plm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Diphtheria +
##      infant.deaths + Adult.Mortality + Alcohol + Measles +
Income.composition.of.resources,
##      data= Pooled_Data, effect = "individual", model = "within")

## Unbalanced Panel: n = 193, T = 1-16, N = 2925

## Residuals:

##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -22.95693  -0.99549  -0.16796   0.66764  10.44901

## Coefficients:

##                                Estimate Std. Error t-value Pr(>|t|)
## Schooling                        4.7687e-01  3.4734e-02  13.7295 < 2.2e-16
## HIV.AIDS                         -4.2978e-01  1.6335e-02 -26.3108 < 2.2e-16
## Diphtheria                       1.4793e-02  2.3369e-03   6.3300 2.858e-10
## infant.deaths                    -8.5933e-03  1.6838e-03  -5.1035 3.564e-07
## Adult.Mortality                  -2.4614e-03  5.0977e-04  -4.8286 1.451e-06
## Alcohol                         -8.7181e-02  2.1725e-02  -4.0129 6.161e-05
## Measles                         -2.1507e-05  4.8608e-06  -4.4246 1.004e-05
## Income.composition.of.resources  2.0060e+00  4.4190e-01   4.5394 5.887e-06

## Schooling                        ***
## HIV.AIDS                         ***
## Diphtheria                       ***
## infant.deaths                    ***
## Adult.Mortality                  ***
## Alcohol                         ***
## Measles                         ***
## Income.composition.of.resources ***
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Total Sum of Squares:    19571
## Residual Sum of Squares: 12352
## R-Squared:              0.36886
## Adj. R-Squared: 0.32252
## F-statistic: 199.002 on 8 and 2724 DF, p-value: < 2.22e-16
```

### Analysis

This model has a decent adjusted  $R^2$ , but we'll try to build some more models which will handle the time or entity specific effects, that we might have missed in this model.

## Entity Fixed Effects (Binary Regressor)

### Model 1(Binary Regressor for each country)

In this model, we have a binary variable for each of the 195 countries. Each countries table is too large so we are showing some results only.

```
## Residual standard error: 2.165 on 2727 degrees of freedom
## Multiple R-squared:  0.9517, Adjusted R-squared:  0.9482
## F-statistic: 272.8 on 197 and 2727 DF, p-value: < 2.2e-16
```

### Analysis

This model has a high adjusted  $R^2$ , but only the entity fixed effects have been handled. In next model we'll try to handle the time fixed effects.

## Time Fixed Effects (Time demeaned)

### Model 1

In time fixed models, variations across the time are taken into consideration but not across entities. For each time period the values are averaged, and the average is then subtracted from the observations of the entity. This way, we obtain the time demeaned observations which are then regressed. Here, we have regressed the life expectancy over the independent variables.

```
## Oneway (time) effect Within Model
## plm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Income.composition.of.resources +
+
##      BMI + Diphtheria + Measles + percentage.expenditure + Adult.Mortality +
##      Alcohol, data = Pooled_Data, effect = "time", model = "within")
## Unbalanced Panel: n = 193, T = 1-16, N = 2925
## Residuals:
```

```

##          Min.      1st Qu.      Median      3rd Qu.      Max.
## -22.984437  -2.202750    0.010137    2.389799    23.090898

## Coefficients:
##
##              Estimate Std. Error t-value Pr(>|t|)
## Schooling          6.9727e-01  4.1268e-02  16.8962 < 2.2e-16
## HIV.AIDS           -4.8279e-01  1.8578e-02 -25.9870 < 2.2e-16
## Income.composition.of.resources  6.3657e+00  6.4095e-01   9.9316 < 2.2e-16
## BMI                5.1696e-02  4.8405e-03  10.6798 < 2.2e-16
## Diphtheria         5.9902e-02  3.7417e-03  16.0092 < 2.2e-16
## Measles            -3.4211e-05  7.1048e-06  -4.8152 1.546e-06
## percentage.expenditure  4.1298e-04  4.4739e-05   9.2308 < 2.2e-16
## Adult.Mortality     -2.1591e-02  8.3882e-04 -25.7400 < 2.2e-16
## Alcohol            9.1950e-02  2.3545e-02   3.9054 9.623e-05
##
## Schooling          ***
## HIV.AIDS           ***
## Income.composition.of.resources ***
## BMI                ***
## Diphtheria         ***
## Measles            ***
## percentage.expenditure ***
## Adult.Mortality     ***
## Alcohol            ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    257610
## Residual Sum of Squares: 53648
## R-Squared:              0.79175
## Adj. R-Squared: 0.79002
## F-statistic: 1225.03 on 9 and 2900 DF, p-value: < 2.22e-16

```

## Analysis

This model has a high adjusted  $R^2$ , but only the time fixed effects have been handled. In next models we'll try to handle the time fixed effects by using the binary regressor for the time periods.

### Time Fixed Effects (Binary Regressor)

#### Model 1 (Binary Regressor for each time period)

In this model, we have a binary variable for each year(time period).

```
## Call:
## lm(formula = Life.expectancy ~ Year + Schooling + HIV.AIDS +
##      Income.composition.of.resources + BMI + Diphtheria + Measles +
##      percentage.expenditure + Adult.Mortality + Alcohol, data = Pooled_Data)
## Residuals:
##      Min        1Q    Median        3Q        Max
## -22.9844  -2.2028   0.0101   2.3898  23.0909
## Coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      5.443e+01  5.367e-01 101.420  < 2e-16 ***
## Year2001        -5.294e-01  4.509e-01  -1.174  0.24045
## Year2002        -7.472e-01  4.504e-01  -1.659  0.09727 .
## Year2003        -7.036e-01  4.511e-01  -1.560  0.11893
## Year2004        -6.592e-01  4.511e-01  -1.461  0.14410
## Year2005        -7.711e-01  4.513e-01  -1.709  0.08760 .
## Year2006        -9.089e-01  4.525e-01  -2.009  0.04467 *
## Year2007        -1.155e+00  4.531e-01  -2.548  0.01089 *
## Year2008        -8.208e-01  4.540e-01  -1.808  0.07070 .
## Year2009        -7.786e-01  4.544e-01  -1.713  0.08674 .
## Year2010        -7.813e-01  4.539e-01  -1.722  0.08526 .
## Year2011        -1.070e+00  4.557e-01  -2.347  0.01899 *
## Year2012        -1.065e+00  4.571e-01  -2.331  0.01982 *
## Year2013        -1.382e+00  4.525e-01  -3.053  0.00228 **
## Year2014        -5.912e-01  4.600e-01  -1.285  0.19889
```

```
## Year2015                -4.562e-01  4.643e-01  -0.983  0.32593
## Schooling                6.973e-01  4.127e-02  16.896  < 2e-16 ***
## HIV.AIDS                -4.828e-01  1.858e-02 -25.987  < 2e-16 ***
## Income.composition.of.resources  6.366e+00  6.410e-01   9.932  < 2e-16 ***
## BMI                     5.170e-02  4.841e-03  10.680  < 2e-16 ***
## Diphtheria              5.990e-02  3.742e-03  16.009  < 2e-16 ***
## Measles                 -3.421e-05  7.105e-06  -4.815  1.55e-06 ***
## percentage.expenditure   4.130e-04  4.474e-05   9.231  < 2e-16 ***
## Adult.Mortality         -2.159e-02  8.388e-04 -25.740  < 2e-16 ***
## Alcohol                 9.195e-02  2.354e-02   3.905  9.62e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 4.301 on 2900 degrees of freedom
## Multiple R-squared:  0.7973, Adjusted R-squared:  0.7956
## F-statistic: 475.3 on 24 and 2900 DF,  p-value: < 2.2e-16
```

## Analysis

This model has a high adjusted  $R^2$ , but because of the introduction of several binary variable, a lot of degrees of freedom have been lost. In further models we'll try to handle both the time fixed and entity fixed effects and see if the new models is better able to explain the relationship between the variables.

## Entity & Time Fixed Effects

### Model 1

In this model, variations across both the time and countries are considered. Through this both the effects that vary across the countries but remain same over time and the effects that vary across the countries and remain same across the countries are handled.

```
## Twoways effects Within Model
## plm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Diphtheria +
##      Measles + Adult.Mortality + Alcohol, data = Pooled_Data,
##      effect = "twoways", model = "within", index = c("Country",
##      "Year"))
```

```

## Unbalanced Panel: n = 193, T = 1-16, N = 2925

## Residuals:

##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -23.67796  -0.85333   -0.26622    0.49311    9.81653

##

## Coefficients:

##              Estimate Std. Error t-value Pr(>|t|)
## Schooling      1.5347e-01  3.0397e-02   5.0487 4.743e-07 ***
## HIV.AIDS       -3.3097e-01  1.5021e-02 -22.0336 < 2.2e-16 ***
## Diphtheria      8.2777e-03  2.1006e-03   3.9406 8.332e-05 ***
## Measles        -1.4051e-05  4.3363e-06  -3.2403 0.001208 **
## Adult.Mortality -1.8053e-03  4.5584e-04  -3.9604 7.675e-05 ***
## Alcohol        -5.4393e-02  2.0015e-02  -2.7176 0.006618 **
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Total Sum of Squares:    11922

## Residual Sum of Squares: 9758.8

## R-Squared:      0.18145

## Adj. R-Squared: 0.11713

## F-statistic: 100.156 on 6 and 2711 DF, p-value: < 2.22e-16

bptest(ETF_Modelfinal)

## studentized Breusch-Pagan test

## data:  ETF_Modelfinal

## BP = 462.23, df = 6, p-value < 2.2e-16

```

## Analysis

This model does not have good adjusted  $R^2$ .



## Random Effect Regression Models

In this the entities are chosen at random, hence the effect of not including the entity would not be correlated with the dependent variable.

```
## Oneway (individual) effect Random Effect Model

##      (Swamy-Arora's transformation)

## plm(formula = Life.expectancy ~ Schooling + HIV.AIDS + Income.composition.of.resources
+
##      BMI + thinness..1.19.years + Diphtheria + Measles + percentage.expenditure +
##      Adult.Mortality, data = Pooled_Data, model = "random", index = c("Country",
##      "Year"))

## Unbalanced Panel: n = 193, T = 1-16, N = 2925

##

## Effects:

##              var std.dev share

## idiosyncratic 4.588    2.142 0.416

## individual    6.449    2.540 0.584

## theta:

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.

##  0.3553  0.7937  0.7937  0.7917  0.7937  0.7937

## Residuals:

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.

## -23.0063 -1.2080  0.0792  0.0449  1.0373  12.2079

## Coefficients:

##              Estimate Std. Error z-value Pr(>|z|)

## (Intercept)      5.9507e+01  5.0236e-01 118.4552 < 2.2e-16

## Schooling         6.4412e-01  3.6188e-02  17.7994 < 2.2e-16

## HIV.AIDS          -4.5636e-01  1.7597e-02 -25.9335 < 2.2e-16
```

```

## Income.composition.of.resources  2.4650e+00  4.8711e-01  5.0604 4.184e-07
## BMI                             1.5484e-02  3.6188e-03  4.2786 1.881e-05
## thinness..1.19.years            -1.0803e-01  2.3334e-02  -4.6296 3.664e-06
## Diphtheria                       1.9710e-02  2.6039e-03  7.5695 3.746e-14
## Measles                          -2.3947e-05  5.3944e-06  -4.4392 9.030e-06
## percentage.expenditure           1.3908e-04  3.2958e-05  4.2198 2.445e-05
## Adult.Mortality                  -5.0552e-03  5.6662e-04  -8.9217 < 2.2e-16
##
## (Intercept)                      ***
## Schooling                        ***
## HIV.AIDS                         ***
## Income.composition.of.resources ***
## BMI                              ***
## thinness..1.19.years             ***
## Diphtheria                       ***
## Measles                          ***
## percentage.expenditure           ***
## Adult.Mortality                  ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    35789
## Residual Sum of Squares: 16949
## R-Squared:              0.53095
## Adj. R-Squared: 0.52951
## Chisq: 3240.08 on 9 DF, p-value: < 2.22e-16

```

## Analysis

Although this model tries to handle the correlation among dependent variable but the adjusted  $R^2$  is not as good as the other models.

### First Difference Model:

```
## Oneway (individual) effect First-Difference Model

## plm(formula = Life.expectancy ~ Schooling + HIV.AIDS + +Total.expenditure +
##      Adult.Mortality, data = Pooled_Data, model = "fd", index = c("Country",
##      "Year"))

## Unbalanced Panel: n = 193, T = 1-16, N = 2925

## Observations used in estimation: 2732

##      Min.      1st Qu.      Median      3rd Qu.      Max.

## -25.896840  -0.194116  -0.021581   0.240465   24.699545

## Coefficients:      Estimate Std. Error t-value Pr(>|t|)

## (Intercept)      0.2725984  0.0362711   7.5156 7.638e-14 ***

## Schooling        0.0963220  0.0427546   2.2529 0.024345 *

## HIV.AIDS         -0.3103088  0.0555702  -5.5841 2.581e-08 ***

## Total.expenditure -0.0544359  0.0186296  -2.9220 0.003506 **

## Adult.Mortality  -0.0011498  0.0003492  -3.2928 0.001005 **

## Total Sum of Squares:      9286.8

## Residual Sum of Squares: 9103

## R-Squared:      0.019787

## Adj. R-Squared: 0.018349

## F-statistic: 13.7618 on 4 and 2727 DF, p-value: 4.0976e-11
```

## Tests

### Testing for heteroskedasticity

#### Breusch-Pagan test for heteroskedasticity

The null hypothesis for the Breusch-Pagan test is homoscedasticity.

data: BinCountry_Model6
BP = 427.26, df = 197, p-value = 2.2e-13

Since, p-value is less than 0.05, we reject the null hypothesis and establish the fact that there is presence of heteroskedasticity in the data.

So, the predicted dependent variables and the residuals of these variables are expected to show heteroskedasticity. To correct this heteroskedasticity-robust standard error can be used.

### Testing for serial collinearity

#### Breusch-Godfrey/ Wooldridge test for serial correlation

The null hypothesis for the Breusch-Godfrey/ Wooldridge test is that there is no serial correlation.

Serial correlation is usually present in panel data spanning over a long time period, which is true in our case (about 26 years).

```
## Breusch-Godfrey/Wooldridge test for serial correlation in panel
## models

## data: Life.expectancy ~ Schooling + HIV.AIDS + Diphtheria + infant.deaths +
Adult.Mortality + Alcohol + Measles + Income.composition.of.resources

## chisq = 689.96, df = 1, p-value < 2.2e-16

## alternative hypothesis: serial correlation in idiosyncratic errors
```

Since, p-value is less than 0.05, we reject the null hypothesis and establish the fact that there is presence of serial correlation in the data.

### Testing for Panel Effect

#### LM Test

To decide whether the OLS Pooled model is better or the Fixed Effects Model, we use the LM Test.

The null hypothesis is that OLS is better than the fixed effects model.

```
## Lagrange Multiplier Test - (Breusch-Pagan) for unbalanced panels
## data: Life.expectancy ~ Schooling + Income.composition.of.resources + ...
## chisq = 7814.9, df = 1, p-value < 2.2e-16
```

```
## alternative hypothesis: significant effects
```

Since p-value is less than 0.05, we reject null hypothesis, thus we can say that the fixed effects model should be chosen over the OLS pooling model.

### **Hausman Test**

#### **Test for choosing between Fixed Effect Model or Random Effect Model**

To decide whether the Fixed effect model is better or the Random effect model, we use the Hausman Test.

The null hypothesis is that Fixed Effect model is better than Random Effect model.

```
## Hausman Test

## data: Life.expectancy ~ Schooling + HIV.AIDS + Income.composition.of.resources + ...

## chisq = 1985.8, df = 5, p-value < 2.2e-16

## alternative hypothesis: one model is inconsistent
```

Since the p-value is larger than 0.05, we fail to reject the null hypothesis, thus we can say that fixed effect model is better than the random effects model.

## **Analysis**

Through the above PF Test, we found that fixed entity models will be better able to explain the relationship, so we will be choosing the fixed entity model with the higher adjusted  $R^2$  value.

Also, as per the results of Hausman test, Fixed Effects model is better than the Random effects test. This is understandable because Random effects model presumes that there is no fixed effects & that all variations are caused due to random effects, which clearly is not the case here as each country has its own entity specific effects.

After going through several models, we have found that the Entity Fixed Effects Regression Model with Binary regressor for country is found to be better as it is able to explain the dependable variable much better compared to other models.

Also as mentioned above, the data had heteroskedasticity & serial correlation, so we will be using clustered standard errors & robust standard errors to give more correct values.

Below are the best model result .

```
## lm(formula = Life.expectancy ~ Country + Schooling + HIV.AIDS +
##      Adult.Mortality + Income.composition.of.resources + Alcohol,
##      data = Pooled_Data)

## Residuals:

##      Min      1Q    Median      3Q      Max
```

```
## -22.8099 -0.9721 -0.1509 0.6582 10.9420
```

```
##
```

```
## Coefficients:
```

```
## Estimate
```

```
## (Intercept) 5.367e+01
```

```
## CountryAlbania 1.412e+01
```

```
## CountryAlgeria 1.210e+01
```

```
## CountryAngola -7.609e+00
```

```
## CountryAntigua and Barbuda 1.666e+01
```

```
## CountryArgentina 1.204e+01
```

```
## CountryArmenia 1.263e+01
```

```
## CountryAustralia 1.666e+01
```

```
## CountryAustria 1.901e+01
```

```
## CountryAzerbaijan 9.954e+00
```

```
## CountryBahamas 1.369e+01
```

```
## CountryBahrain 1.324e+01
```

```
## CountryBangladesh 1.031e+01
```

```
## CountryBarbados 1.215e+01
```

```
## CountryBelarus 8.510e+00
```

```
## CountryBelgium 1.738e+01
```

```
## CountryBelize 8.597e+00
```

```
## CountryBenin -2.221e-01
```

```
## CountryBhutan 7.564e+00
```

```
## CountryBolivia (Plurinational State of) 6.127e+00
```

```
## CountryBosnia and Herzegovina 1.552e+01
```

```
## CountryBotswana 3.364e+00
```

```
## CountryBrazil 1.174e+01
```

```
## CountryBrunei Darussalam 1.384e+01
```

```
## CountryBulgaria 1.156e+01
```

```
## CountryBurkina Faso 4.148e-01
```

```
## CountryBurundi -3.580e-01
```

```
## CountryCôte d'Ivoire -5.193e+00
```

## CountryCabo Verde	1.204e+01
## CountryCambodia	5.489e+00
## CountryCameroon	-1.570e+00
## CountryCanada	1.859e+01
## CountryCentral African Republic	-4.307e+00
## CountryChad	-4.667e+00
## CountryChile	1.668e+01
## CountryChina	1.369e+01
## CountryColombia	1.240e+01
## CountryComoros	2.741e+00
## CountryCongo	2.065e+00
## CountryCook Islands	2.895e+00
## CountryCosta Rica	1.710e+01
## CountryCroatia	1.467e+01
## CountryCuba	1.551e+01
## CountryCyprus	1.805e+01
## CountryCzechia	1.640e+01
## CountryDemocratic People's Republic of Korea	8.168e+00
## CountryDemocratic Republic of the Congo	-1.783e+00
## CountryDenmark	1.566e+01
## CountryDjibouti	5.797e+00
## CountryDominica	-1.320e+00
## CountryDominican Republic	1.194e+01
## CountryEcuador	1.341e+01
## CountryEgypt	1.065e+01
## CountryEl Salvador	1.065e+01
## CountryEquatorial Guinea	-3.792e-01
## CountryEritrea	5.231e+00
## CountryEstonia	1.232e+01
## CountryEthiopia	3.124e+00
## CountryFiji	6.585e+00
## CountryFinland	1.708e+01

## CountryFrance	1.953e+01
## CountryGabon	5.327e+00
## CountryGambia	2.433e+00
## CountryGeorgia	1.249e+01
## CountryGermany	1.799e+01
## CountryGhana	2.602e+00
## CountryGreece	1.828e+01
## CountryGrenada	1.239e+01
## CountryGuatemala	1.226e+01
## CountryGuinea	-7.076e-01
## CountryGuinea-Bissau	-4.424e-01
## CountryGuyana	6.521e+00
## CountryHaiti	2.829e+00
## CountryHonduras	1.330e+01
## CountryHungary	1.180e+01
## CountryIceland	1.814e+01
## CountryIndia	5.972e+00
## CountryIndonesia	6.932e+00
## CountryIran (Islamic Republic of)	1.233e+01
## CountryIraq	1.051e+01
## CountryIreland	1.654e+01
## CountryIsrael	1.790e+01
## CountryItaly	1.914e+01
## CountryJamaica	1.383e+01
## CountryJapan	1.982e+01
## CountryJordan	1.118e+01
## CountryKazakhstan	5.274e+00
## CountryKenya	2.355e+00
## CountryKiribati	5.047e+00
## CountryKuwait	1.153e+01
## CountryKyrgyzstan	8.279e+00
## CountryLao People's Democratic Republic	3.614e+00



## CountryLatvia	1.135e+01
## CountryLebanon	1.274e+01
## CountryLesotho	1.554e-01
## CountryLiberia	-2.543e-01
## CountryLibya	9.879e+00
## CountryLithuania	1.030e+01
## CountryLuxembourg	1.914e+01
## CountryMadagascar	4.073e+00
## CountryMalawi	-1.594e+00
## CountryMalaysia	1.224e+01
## CountryMaldives	1.456e+01
## CountryMali	-1.571e+00
## CountryMalta	1.818e+01
## CountryMarshall Islands	5.017e+00
## CountryMauritania	5.166e+00
## CountryMauritius	1.106e+01
## CountryMexico	1.471e+01
## CountryMicronesia (Federated States of)	8.390e+00
## CountryMonaco	-2.476e+00
## CountryMongolia	4.913e+00
## CountryMontenegro	1.443e+01
## CountryMorocco	1.214e+01
## CountryMozambique	4.199e-01
## CountryMyanmar	5.708e+00
## CountryNamibia	6.533e+00
## CountryNauru	-5.638e-01
## CountryNepal	6.714e+00
## CountryNetherlands	1.750e+01
## CountryNew Zealand	1.684e+01
## CountryNicaragua	1.345e+01
## CountryNiger	1.653e+00
## CountryNigeria	-4.368e+00

## CountryNiue	-3.162e+00
## CountryNorway	1.766e+01
## CountryOman	1.350e+01
## CountryPakistan	6.648e+00
## CountryPalau	-2.611e+00
## CountryPanama	1.535e+01
## CountryPapua New Guinea	3.757e+00
## CountryParaguay	1.251e+01
## CountryPeru	1.229e+01
## CountryPhilippines	7.392e+00
## CountryPoland	1.335e+01
## CountryPortugal	1.738e+01
## CountryQatar	1.515e+01
## CountryRepublic of Korea	2.167e+01
## CountryRepublic of Moldova	7.808e+00
## CountryRomania	1.259e+01
## CountryRussian Federation	6.799e+00
## CountryRwanda	2.912e+00
## CountrySaint Kitts and Nevis	-2.800e+00
## CountrySaint Lucia	1.320e+01
## CountrySaint Vincent and the Grenadines	1.249e+01
## CountrySamoa	1.248e+01
## CountrySan Marino	-4.888e+00
## CountrySao Tome and Principe	6.178e+00
## CountrySaudi Arabia	1.135e+01
## CountrySenegal	4.942e+00
## CountrySerbia	1.267e+01
## CountrySeychelles	1.141e+01
## CountrySierra Leone	-1.082e+01
## CountrySingapore	1.885e+01
## CountrySlovakia	1.300e+01
## CountrySlovenia	1.669e+01

## CountrySolomon Islands	8.902e+00
## CountrySomalia	-7.765e+00
## CountrySouth Africa	5.424e+00
## CountrySouth Sudan	1.891e+00
## CountrySpain	1.901e+01
## CountrySri Lanka	1.182e+01
## CountrySudan	4.667e+00
## CountrySuriname	1.037e+01
## CountrySwaziland	6.799e+00
## CountrySweden	1.936e+01
## CountrySwitzerland	1.963e+01
## CountrySyrian Arab Republic	1.056e+01
## CountryTajikistan	6.724e+00
## CountryThailand	1.240e+01
## CountryThe former Yugoslav republic of Macedonia	1.340e+01
## CountryTimor-Leste	4.889e+00
## CountryTogo	-9.616e-01
## CountryTonga	1.034e+01
## CountryTrinidad and Tobago	1.048e+01
## CountryTunisia	1.203e+01
## CountryTurkey	1.246e+01
## CountryTurkmenistan	6.175e+00
## CountryTuvalu	2.136e+00
## CountryUganda	2.508e-01
## CountryUkraine	8.374e+00
## CountryUnited Arab Emirates	1.385e+01
## CountryUnited Kingdom of Great Britain and Northern Ireland	1.628e+01
## CountryUnited Republic of Tanzania	-2.919e+00
## CountryUnited States of America	1.767e+01
## CountryUruguay	1.359e+01
## CountryUzbekistan	7.614e+00
## CountryVanuatu	1.187e+01

## CountryVenezuela (Bolivarian Republic of)	1.244e+01
## CountryViet Nam	1.432e+01
## CountryYemen	5.252e+00
## CountryZambia	-6.241e-01
## CountryZimbabwe	2.399e+00
## Schooling	5.112e-01
## HIV.AIDS	-4.396e-01
## Adult.Mortality	-2.360e-03
## Income.composition.of.resources	2.411e+00
## Alcohol	-8.298e-02
##	Std. Error
## (Intercept)	6.139e-01
## CountryAlbania	7.908e-01
## CountryAlgeria	7.815e-01
## CountryAngola	7.756e-01
## CountryAntigua and Barbuda	7.866e-01
## CountryArgentina	8.226e-01
## CountryArmenia	7.817e-01
## CountryAustralia	8.701e-01
## CountryAustria	8.399e-01
## CountryAzerbaijan	7.777e-01
## CountryBahamas	8.076e-01
## CountryBahrain	7.941e-01
## CountryBangladesh	7.691e-01
## CountryBarbados	8.132e-01
## CountryBelarus	8.364e-01
## CountryBelgium	8.425e-01
## CountryBelize	7.889e-01
## CountryBenin	7.664e-01
## CountryBhutan	7.809e-01
## CountryBolivia (Plurinational State of)	7.881e-01
## CountryBosnia and Herzegovina	7.892e-01

## CountryBotswana	8.295e-01
## CountryBrazil	8.020e-01
## CountryBrunei Darussalam	7.944e-01
## CountryBulgaria	8.194e-01
## CountryBurkina Faso	7.767e-01
## CountryBurundi	7.726e-01
## CountryCôte d'Ivoire	7.739e-01
## CountryCabo Verde	7.818e-01
## CountryCambodia	7.689e-01
## CountryCameroon	7.769e-01
## CountryCanada	8.236e-01
## CountryCentral African Republic	7.802e-01
## CountryChad	7.714e-01
## CountryChile	8.229e-01
## CountryChina	7.832e-01
## CountryColombia	7.838e-01
## CountryComoros	7.708e-01
## CountryCongo	7.735e-01
## CountryCook Islands	2.239e+00
## CountryCosta Rica	7.892e-01
## CountryCroatia	8.303e-01
## CountryCuba	7.998e-01
## CountryCyprus	8.198e-01
## CountryCzechia	8.266e-01
## CountryDemocratic People's Republic of Korea	7.930e-01
## CountryDemocratic Republic of the Congo	7.669e-01
## CountryDenmark	8.447e-01
## CountryDjibouti	7.763e-01
## CountryDominica	2.236e+00
## CountryDominican Republic	7.902e-01
## CountryEcuador	7.855e-01
## CountryEgypt	7.751e-01

## CountryEl Salvador	7.930e-01
## CountryEquatorial Guinea	7.873e-01
## CountryEritrea	7.720e-01
## CountryEstonia	8.234e-01
## CountryEthiopia	7.681e-01
## CountryFiji	7.891e-01
## CountryFinland	8.412e-01
## CountryFrance	8.466e-01
## CountryGabon	8.055e-01
## CountryGambia	7.679e-01
## CountryGeorgia	7.897e-01
## CountryGermany	8.461e-01
## CountryGhana	7.698e-01
## CountryGreece	8.249e-01
## CountryGrenada	8.424e-01
## CountryGuatemala	7.727e-01
## CountryGuinea	7.796e-01
## CountryGuinea-Bissau	7.777e-01
## CountryGuyana	7.880e-01
## CountryHaiti	7.785e-01
## CountryHonduras	7.760e-01
## CountryHungary	8.313e-01
## CountryIceland	8.413e-01
## CountryIndia	7.717e-01
## CountryIndonesia	7.744e-01
## CountryIran (Islamic Republic of)	7.819e-01
## CountryIraq	7.725e-01
## CountryIreland	8.621e-01
## CountryIsrael	8.070e-01
## CountryItaly	8.373e-01
## CountryJamaica	7.839e-01
## CountryJapan	8.133e-01

## CountryJordan	7.844e-01
## CountryKazakhstan	7.987e-01
## CountryKenya	7.820e-01
## CountryKiribati	7.806e-01
## CountryKuwait	7.898e-01
## CountryKyrgyzstan	7.783e-01
## CountryLao People's Democratic Republic	7.717e-01
## CountryLatvia	8.198e-01
## CountryLebanon	7.941e-01
## CountryLesotho	8.626e-01
## CountryLiberia	7.729e-01
## CountryLibya	7.920e-01
## CountryLithuania	8.466e-01
## CountryLuxembourg	8.473e-01
## CountryMadagascar	7.669e-01
## CountryMalawi	8.176e-01
## CountryMalaysia	7.822e-01
## CountryMaldives	7.798e-01
## CountryMali	7.689e-01
## CountryMalta	8.086e-01
## CountryMarshall Islands	2.267e+00
## CountryMauritania	7.695e-01
## CountryMauritius	7.860e-01
## CountryMexico	7.885e-01
## CountryMicronesia (Federated States of)	7.724e-01
## CountryMonaco	2.239e+00
## CountryMongolia	7.790e-01
## CountryMontenegro	7.766e-01
## CountryMorocco	7.736e-01
## CountryMozambique	7.891e-01
## CountryMyanmar	7.684e-01
## CountryNamibia	8.257e-01

## CountryNauru	2.236e+00
## CountryNepal	7.705e-01
## CountryNetherlands	8.377e-01
## CountryNew Zealand	8.549e-01
## CountryNicaragua	7.772e-01
## CountryNiger	7.764e-01
## CountryNigeria	7.931e-01
## CountryNiue	2.239e+00
## CountryNorway	8.302e-01
## CountryOman	7.830e-01
## CountryPakistan	7.720e-01
## CountryPalau	2.251e+00
## CountryPanama	7.954e-01
## CountryPapua New Guinea	7.663e-01
## CountryParaguay	7.875e-01
## CountryPeru	7.899e-01
## CountryPhilippines	7.802e-01
## CountryPoland	8.264e-01
## CountryPortugal	8.587e-01
## CountryQatar	7.895e-01
## CountryRepublic of Korea	7.940e-01
## CountryRepublic of Moldova	8.238e-01
## CountryRomania	8.113e-01
## CountryRussian Federation	8.213e-01
## CountryRwanda	7.804e-01
## CountrySaint Kitts and Nevis	2.245e+00
## CountrySaint Lucia	8.190e-01
## CountrySaint Vincent and the Grenadines	7.936e-01
## CountrySamoa	7.829e-01
## CountrySan Marino	2.241e+00
## CountrySao Tome and Principe	7.766e-01
## CountrySaudi Arabia	7.897e-01



## CountrySenegal	7.687e-01
## CountrySerbia	8.086e-01
## CountrySeychelles	7.929e-01
## CountrySierra Leone	7.713e-01
## CountrySingapore	8.077e-01
## CountrySlovakia	8.243e-01
## CountrySlovenia	8.428e-01
## CountrySolomon Islands	7.687e-01
## CountrySomalia	7.960e-01
## CountrySouth Africa	8.501e-01
## CountrySouth Sudan	8.034e-01
## CountrySpain	8.370e-01
## CountrySri Lanka	7.846e-01
## CountrySudan	7.701e-01
## CountrySuriname	7.838e-01
## CountrySwaziland	9.435e-01
## CountrySweden	8.212e-01
## CountrySwitzerland	8.356e-01
## CountrySyrian Arab Republic	7.737e-01
## CountryTajikistan	7.709e-01
## CountryThailand	7.911e-01
## CountryThe former Yugoslav republic of Macedonia	7.860e-01
## CountryTimor-Leste	7.706e-01
## CountryTogo	7.732e-01
## CountryTonga	7.885e-01
## CountryTrinidad and Tobago	8.028e-01
## CountryTunisia	7.942e-01
## CountryTurkey	7.826e-01
## CountryTurkmenistan	7.790e-01
## CountryTuvalu	2.277e+00
## CountryUganda	7.958e-01
## CountryUkraine	8.033e-01

## CountryUnited Arab Emirates	7.895e-01
## CountryUnited Kingdom of Great Britain and Northern Ireland	8.693e-01
## CountryUnited Republic of Tanzania	8.053e-01
## CountryUnited States of America	8.019e-01
## CountryUruguay	8.060e-01
## CountryUzbekistan	7.740e-01
## CountryVanuatu	7.744e-01
## CountryVenezuela (Bolivarian Republic of)	7.953e-01
## CountryViet Nam	7.779e-01
## CountryYemen	7.663e-01
## CountryZambia	8.118e-01
## CountryZimbabwe	8.601e-01
## Schooling	3.506e-02
## HIV.AIDS	1.657e-02
## Adult.Mortality	5.171e-04
## Income.composition.of.resources	4.463e-01
## Alcohol	2.204e-02
##	t value
## (Intercept)	87.426
## CountryAlbania	17.854
## CountryAlgeria	15.485
## CountryAngola	-9.811
## CountryAntigua and Barbuda	21.176
## CountryArgentina	14.640
## CountryArmenia	16.151
## CountryAustralia	19.153
## CountryAustria	22.636
## CountryAzerbaijan	12.799
## CountryBahamas	16.954
## CountryBahrain	16.671
## CountryBangladesh	13.407
## CountryBarbados	14.936

## CountryBelarus	10.175
## CountryBelgium	20.626
## CountryBelize	10.897
## CountryBenin	-0.290
## CountryBhutan	9.685
## CountryBolivia (Plurinational State of)	7.775
## CountryBosnia and Herzegovina	19.671
## CountryBotswana	4.055
## CountryBrazil	14.643
## CountryBrunei Darussalam	17.420
## CountryBulgaria	14.102
## CountryBurkina Faso	0.534
## CountryBurundi	-0.463
## CountryCôte d'Ivoire	-6.709
## CountryCabo Verde	15.396
## CountryCambodia	7.138
## CountryCameroon	-2.021
## CountryCanada	22.572
## CountryCentral African Republic	-5.520
## CountryChad	-6.050
## CountryChile	20.272
## CountryChina	17.483
## CountryColombia	15.820
## CountryComoros	3.556
## CountryCongo	2.670
## CountryCook Islands	1.293
## CountryCosta Rica	21.661
## CountryCroatia	17.666
## CountryCuba	19.389
## CountryCyprus	22.022
## CountryCzechia	19.845
## CountryDemocratic People's Republic of Korea	10.300

## CountryDemocratic Republic of the Congo	-2.325
## CountryDenmark	18.536
## CountryDjibouti	7.467
## CountryDominica	-0.591
## CountryDominican Republic	15.106
## CountryEcuador	17.070
## CountryEgypt	13.735
## CountryEl Salvador	13.433
## CountryEquatorial Guinea	-0.482
## CountryEritrea	6.776
## CountryEstonia	14.968
## CountryEthiopia	4.067
## CountryFiji	8.345
## CountryFinland	20.303
## CountryFrance	23.072
## CountryGabon	6.613
## CountryGambia	3.169
## CountryGeorgia	15.816
## CountryGermany	21.265
## CountryGhana	3.379
## CountryGreece	22.162
## CountryGrenada	14.705
## CountryGuatemala	15.864
## CountryGuinea	-0.908
## CountryGuinea-Bissau	-0.569
## CountryGuyana	8.275
## CountryHaiti	3.634
## CountryHonduras	17.143
## CountryHungary	14.199
## CountryIceland	21.558
## CountryIndia	7.738
## CountryIndonesia	8.952

## CountryIran (Islamic Republic of)	15.768
## CountryIraq	13.599
## CountryIreland	19.183
## CountryIsrael	22.187
## CountryItaly	22.860
## CountryJamaica	17.644
## CountryJapan	24.372
## CountryJordan	14.258
## CountryKazakhstan	6.603
## CountryKenya	3.012
## CountryKiribati	6.466
## CountryKuwait	14.605
## CountryKyrgyzstan	10.637
## CountryLao People's Democratic Republic	4.684
## CountryLatvia	13.840
## CountryLebanon	16.050
## CountryLesotho	0.180
## CountryLiberia	-0.329
## CountryLibya	12.473
## CountryLithuania	12.168
## CountryLuxembourg	22.591
## CountryMadagascar	5.311
## CountryMalawi	-1.950
## CountryMalaysia	15.644
## CountryMaldives	18.669
## CountryMali	-2.044
## CountryMalta	22.486
## CountryMarshall Islands	2.213
## CountryMauritania	6.714
## CountryMauritius	14.068
## CountryMexico	18.656
## CountryMicronesia (Federated States of)	10.861

## CountryMonaco	-1.106
## CountryMongolia	6.307
## CountryMontenegro	18.582
## CountryMorocco	15.688
## CountryMozambique	0.532
## CountryMyanmar	7.428
## CountryNamibia	7.912
## CountryNauru	-0.252
## CountryNepal	8.713
## CountryNetherlands	20.887
## CountryNew Zealand	19.701
## CountryNicaragua	17.310
## CountryNiger	2.128
## CountryNigeria	-5.508
## CountryNiue	-1.413
## CountryNorway	21.274
## CountryOman	17.236
## CountryPakistan	8.611
## CountryPalau	-1.160
## CountryPanama	19.293
## CountryPapua New Guinea	4.903
## CountryParaguay	15.886
## CountryPeru	15.562
## CountryPhilippines	9.475
## CountryPoland	16.149
## CountryPortugal	20.240
## CountryQatar	19.190
## CountryRepublic of Korea	27.293
## CountryRepublic of Moldova	9.477
## CountryRomania	15.518
## CountryRussian Federation	8.279
## CountryRwanda	3.731

## CountrySaint Kitts and Nevis	-1.247
## CountrySaint Lucia	16.113
## CountrySaint Vincent and the Grenadines	15.745
## CountrySamoa	15.938
## CountrySan Marino	-2.181
## CountrySao Tome and Principe	7.955
## CountrySaudi Arabia	14.375
## CountrySenegal	6.430
## CountrySerbia	15.666
## CountrySeychelles	14.386
## CountrySierra Leone	-14.027
## CountrySingapore	23.337
## CountrySlovakia	15.777
## CountrySlovenia	19.803
## CountrySolomon Islands	11.580
## CountrySomalia	-9.754
## CountrySouth Africa	6.380
## CountrySouth Sudan	2.353
## CountrySpain	22.707
## CountrySri Lanka	15.061
## CountrySudan	6.060
## CountrySuriname	13.224
## CountrySwaziland	7.206
## CountrySweden	23.571
## CountrySwitzerland	23.489
## CountrySyrian Arab Republic	13.652
## CountryTajikistan	8.723
## CountryThailand	15.671
## CountryThe former Yugoslav republic of Macedonia	17.047
## CountryTimor-Leste	6.344
## CountryTogo	-1.244
## CountryTonga	13.119

## CountryTrinidad and Tobago	13.051
## CountryTunisia	15.142
## CountryTurkey	15.920
## CountryTurkmenistan	7.927
## CountryTuvalu	0.938
## CountryUganda	0.315
## CountryUkraine	10.425
## CountryUnited Arab Emirates	17.541
## CountryUnited Kingdom of Great Britain and Northern Ireland	18.730
## CountryUnited Republic of Tanzania	-3.624
## CountryUnited States of America	22.036
## CountryUruguay	16.856
## CountryUzbekistan	9.837
## CountryVanuatu	15.326
## CountryVenezuela (Bolivarian Republic of)	15.641
## CountryViet Nam	18.403
## CountryYemen	6.854
## CountryZambia	-0.769
## CountryZimbabwe	2.789
## Schooling	14.580
## HIV.AIDS	-26.529
## Adult.Mortality	-4.564
## Income.composition.of.resources	5.403
## Alcohol	-3.765
##	Pr(> t )
## (Intercept)	< 2e-16 ***
## CountryAlbania	< 2e-16 ***
## CountryAlgeria	< 2e-16 ***
## CountryAngola	< 2e-16 ***
## CountryAntigua and Barbuda	< 2e-16 ***
## CountryArgentina	< 2e-16 ***
## CountryArmenia	< 2e-16 ***



## CountryAustralia	< 2e-16 ***
## CountryAustria	< 2e-16 ***
## CountryAzerbaijan	< 2e-16 ***
## CountryBahamas	< 2e-16 ***
## CountryBahrain	< 2e-16 ***
## CountryBangladesh	< 2e-16 ***
## CountryBarbados	< 2e-16 ***
## CountryBelarus	< 2e-16 ***
## CountryBelgium	< 2e-16 ***
## CountryBelize	< 2e-16 ***
## CountryBenin	0.772029
## CountryBhutan	< 2e-16 ***
## CountryBolivia (Plurinational State of)	1.06e-14 ***
## CountryBosnia and Herzegovina	< 2e-16 ***
## CountryBotswana	5.15e-05 ***
## CountryBrazil	< 2e-16 ***
## CountryBrunei Darussalam	< 2e-16 ***
## CountryBulgaria	< 2e-16 ***
## CountryBurkina Faso	0.593288
## CountryBurundi	0.643077
## CountryCôte d'Ivoire	2.37e-11 ***
## CountryCabo Verde	< 2e-16 ***
## CountryCambodia	1.21e-12 ***
## CountryCameroon	0.043399 *
## CountryCanada	< 2e-16 ***
## CountryCentral African Republic	3.70e-08 ***
## CountryChad	1.65e-09 ***
## CountryChile	< 2e-16 ***
## CountryChina	< 2e-16 ***
## CountryColombia	< 2e-16 ***
## CountryComoros	0.000383 ***
## CountryCongo	0.007626 **

## CountryCook Islands	0.196098
## CountryCosta Rica	< 2e-16 ***
## CountryCroatia	< 2e-16 ***
## CountryCuba	< 2e-16 ***
## CountryCyprus	< 2e-16 ***
## CountryCzechia	< 2e-16 ***
## CountryDemocratic People's Republic of Korea	< 2e-16 ***
## CountryDemocratic Republic of the Congo	0.020158 *
## CountryDenmark	< 2e-16 ***
## CountryDjibouti	1.10e-13 ***
## CountryDominica	0.554895
## CountryDominican Republic	< 2e-16 ***
## CountryEcuador	< 2e-16 ***
## CountryEgypt	< 2e-16 ***
## CountryEl Salvador	< 2e-16 ***
## CountryEquatorial Guinea	0.630104
## CountryEritrea	1.50e-11 ***
## CountryEstonia	< 2e-16 ***
## CountryEthiopia	4.90e-05 ***
## CountryFiji	< 2e-16 ***
## CountryFinland	< 2e-16 ***
## CountryFrance	< 2e-16 ***
## CountryGabon	4.50e-11 ***
## CountryGambia	0.001548 **
## CountryGeorgia	< 2e-16 ***
## CountryGermany	< 2e-16 ***
## CountryGhana	0.000737 ***
## CountryGreece	< 2e-16 ***
## CountryGrenada	< 2e-16 ***
## CountryGuatemala	< 2e-16 ***
## CountryGuinea	0.364116
## CountryGuinea-Bissau	0.569446

## CountryGuyana	< 2e-16 ***
## CountryHaiti	0.000285 ***
## CountryHonduras	< 2e-16 ***
## CountryHungary	< 2e-16 ***
## CountryIceland	< 2e-16 ***
## CountryIndia	1.41e-14 ***
## CountryIndonesia	< 2e-16 ***
## CountryIran (Islamic Republic of)	< 2e-16 ***
## CountryIraq	< 2e-16 ***
## CountryIreland	< 2e-16 ***
## CountryIsrael	< 2e-16 ***
## CountryItaly	< 2e-16 ***
## CountryJamaica	< 2e-16 ***
## CountryJapan	< 2e-16 ***
## CountryJordan	< 2e-16 ***
## CountryKazakhstan	4.83e-11 ***
## CountryKenya	0.002622 **
## CountryKiribati	1.19e-10 ***
## CountryKuwait	< 2e-16 ***
## CountryKyrgyzstan	< 2e-16 ***
## CountryLao People's Democratic Republic	2.96e-06 ***
## CountryLatvia	< 2e-16 ***
## CountryLebanon	< 2e-16 ***
## CountryLesotho	0.857082
## CountryLiberia	0.742134
## CountryLibya	< 2e-16 ***
## CountryLithuania	< 2e-16 ***
## CountryLuxembourg	< 2e-16 ***
## CountryMadagascar	1.18e-07 ***
## CountryMalawi	0.051332 .
## CountryMalaysia	< 2e-16 ***
## CountryMaldives	< 2e-16 ***

## CountryMali	0.041087 *
## CountryMalta	< 2e-16 ***
## CountryMarshall Islands	0.026967 *
## CountryMauritania	2.30e-11 ***
## CountryMauritius	< 2e-16 ***
## CountryMexico	< 2e-16 ***
## CountryMicronesia (Federated States of)	< 2e-16 ***
## CountryMonaco	0.268766
## CountryMongolia	3.31e-10 ***
## CountryMontenegro	< 2e-16 ***
## CountryMorocco	< 2e-16 ***
## CountryMozambique	0.594668
## CountryMyanmar	1.46e-13 ***
## CountryNamibia	3.66e-15 ***
## CountryNauru	0.800955
## CountryNepal	< 2e-16 ***
## CountryNetherlands	< 2e-16 ***
## CountryNew Zealand	< 2e-16 ***
## CountryNicaragua	< 2e-16 ***
## CountryNiger	0.033393 *
## CountryNigeria	3.98e-08 ***
## CountryNiue	0.157914
## CountryNorway	< 2e-16 ***
## CountryOman	< 2e-16 ***
## CountryPakistan	< 2e-16 ***
## CountryPalau	0.246137
## CountryPanama	< 2e-16 ***
## CountryPapua New Guinea	9.99e-07 ***
## CountryParaguay	< 2e-16 ***
## CountryPeru	< 2e-16 ***
## CountryPhilippines	< 2e-16 ***
## CountryPoland	< 2e-16 ***

## CountryPortugal	< 2e-16 ***
## CountryQatar	< 2e-16 ***
## CountryRepublic of Korea	< 2e-16 ***
## CountryRepublic of Moldova	< 2e-16 ***
## CountryRomania	< 2e-16 ***
## CountryRussian Federation	< 2e-16 ***
## CountryRwanda	0.000194 ***
## CountrySaint Kitts and Nevis	0.212391
## CountrySaint Lucia	< 2e-16 ***
## CountrySaint Vincent and the Grenadines	< 2e-16 ***
## CountrySamoa	< 2e-16 ***
## CountrySan Marino	0.029244 *
## CountrySao Tome and Principe	2.61e-15 ***
## CountrySaudi Arabia	< 2e-16 ***
## CountrySenegal	1.50e-10 ***
## CountrySerbia	< 2e-16 ***
## CountrySeychelles	< 2e-16 ***
## CountrySierra Leone	< 2e-16 ***
## CountrySingapore	< 2e-16 ***
## CountrySlovakia	< 2e-16 ***
## CountrySlovenia	< 2e-16 ***
## CountrySolomon Islands	< 2e-16 ***
## CountrySomalia	< 2e-16 ***
## CountrySouth Africa	2.08e-10 ***
## CountrySouth Sudan	0.018669 *
## CountrySpain	< 2e-16 ***
## CountrySri Lanka	< 2e-16 ***
## CountrySudan	1.55e-09 ***
## CountrySuriname	< 2e-16 ***
## CountrySwaziland	7.40e-13 ***
## CountrySweden	< 2e-16 ***
## CountrySwitzerland	< 2e-16 ***

## CountrySyrian Arab Republic	< 2e-16 ***
## CountryTajikistan	< 2e-16 ***
## CountryThailand	< 2e-16 ***
## CountryThe former Yugoslav republic of Macedonia	< 2e-16 ***
## CountryTimor-Leste	2.60e-10 ***
## CountryTogo	0.213684
## CountryTonga	< 2e-16 ***
## CountryTrinidad and Tobago	< 2e-16 ***
## CountryTunisia	< 2e-16 ***
## CountryTurkey	< 2e-16 ***
## CountryTurkmenistan	3.25e-15 ***
## CountryTuvalu	0.348196
## CountryUganda	0.752714
## CountryUkraine	< 2e-16 ***
## CountryUnited Arab Emirates	< 2e-16 ***
## CountryUnited Kingdom of Great Britain and Northern Ireland	< 2e-16 ***
## CountryUnited Republic of Tanzania	0.000295 ***
## CountryUnited States of America	< 2e-16 ***
## CountryUruguay	< 2e-16 ***
## CountryUzbekistan	< 2e-16 ***
## CountryVanuatu	< 2e-16 ***
## CountryVenezuela (Bolivarian Republic of)	< 2e-16 ***
## CountryViet Nam	< 2e-16 ***
## CountryYemen	8.84e-12 ***
## CountryZambia	0.442106
## CountryZimbabwe	0.005324 **
## Schooling	< 2e-16 ***
## HIV.AIDS	< 2e-16 ***
## Adult.Mortality	5.25e-06 ***
## Income.composition.of.resources	7.12e-08 ***
## Alcohol	0.000170 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1	

```
## Residual standard error: 2.165 on 2727 degrees of freedom
## Multiple R-squared:  0.9517, Adjusted R-squared:  0.9482
## F-statistic: 272.8 on 197 and 2727 DF,  p-value: < 2.2e-16
```

## *Conclusion*

After making several models and carrying out various tests, we have found that Fixed Effects Regression model is better than Pooled OLS or Random Effect Model, and that the Entity Fixed Effect model with Binary regressor has a higher adjusted  $R^2$  value and is better able to explain the dependable variable.

Also, the heterogeneity graphs over individual and over years suggest that the data has heterogeneity. Binary regressor with countries is our best model. Coefficients are shown in previous page. Table is large because of 195 countries.

Even after all these regressions, we can't guarantee that the models in here will be able to perfectly capture the relation of the dependable variable as there might be omitted variable bias.

## *R Code*

```
data<-read.csv("datasets_12603_17232_Life Expectancy Data.csv")

colSums(is.na(data))

data1<-data[-c(0:3)]

library(ggplot2)

library(GGally)

library(corrplot)

ggcorr(data1, nbreaks = 4, palette = "RdGy", label = TRUE, label_size = 3, label_color = "white")

library(dplyr)

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Adult.Mortality))

data$Adult.Mortality[is.na(data$Adult.Mortality)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Life.expectancy))

data$Life.expectancy[is.na(data$Life.expectancy)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Polio))

data$Polio[is.na(data$Polio)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Diphtheria))

data$Diphtheria[is.na(data$Diphtheria)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(thinness..1.19.years))

data$thinness..1.19.years[is.na(data$thinness..1.19.years)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Schooling))

data$Schooling[is.na(data$Schooling)] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(BMI ))

data$BMI [is.na(data$BMI )] <- data$avg

data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(thinness.5.9.years))
```



```

data$thinness.5.9.years[is.na(data$thinness.5.9.years)] <- data$avg
data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Alcohol))
data$Alcohol[is.na(data$Alcohol)] <- data$avg
data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(avg))
data$avg[is.na(data$avg)] <- data$avg
data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Total.expenditure))
data$Total.expenditure[is.na(data$Total.expenditure)] <- data$avg
data<- data %>% group_by(Country,Year) %>% mutate(avg =
mean(Income.composition.of.resources))
data$Income.composition.of.resources[is.na(data$Income.composition.of.resources)] <- data$avg
data<- data %>% group_by(Country,Year) %>% mutate(avg = mean(Alcohol))
data$Alcohol[is.na(data$Alcohol)] <- data$avg
colSums(is.na(data))
data$Year<-as.factor(data$Year)

```

```

data<-select(data,-c(Hepatitis.B,under.five.deaths,GDP,Polio,Population, thinness.5.9.years,avg))
datafinal<-na.omit(data)
summary(datafinal)

```

```

ggplot(data = datafinal,aes(y=Life.expectancy,x=Adult.Mortality))+geom_point()
ggplot(data = datafinal,aes(y=Life.expectancy,x=Adult.Mortality,col=Status))+geom_point()
ggplot(data = datafinal,aes(y=Life.expectancy,x=infant.deaths))+geom_point()
ggplot(data = datafinal,aes(y=Life.expectancy,x=Alcohol,col=Status))+geom_point()
ggplot(data = datafinal,aes(y=Life.expectancy,x=HIV.AIDS,col=Status))+geom_point()

```

## HISTOGRAM

```

ggplot(data = datafinal,aes(x=Life.expectancy))+geom_histogram(bins = 50,fill="palegreen4")

```

```
ggplot(data = datafinal,aes(x=Life.expectancy,fill=Status))+geom_histogram(bins = 50,position = "fill")
```

```
ggplot(data = datafinal,aes(x=Schooling,fill=Status))+geom_histogram(bins = 50,position = "fill")
```

```
ggplot(data = datafinal,aes(x=HIV.AIDS,fill=Status))+geom_histogram(bins = 50,position = "fill")
```

```
ggplot(data = datafinal,aes(x=BMI,fill=Status))+geom_histogram(bins = 50,position = "fill")
```

```
ggplot(data=datafinal,aes(x=Life.expectancy,col=Year))+geom_freqpoly(bins=30)
```

```
ggplot(data=datafinal,aes(x=Life.expectancy,col=Status))+geom_freqpoly(bins=30)
```

```
ggplot(data=datafinal,aes(x=factor(Year),y=Life.expectancy))+geom_boxplot()
```

```
ggplot(data=datafinal,aes(x=factor(Year),y=Life.expectancy,fill=Status))+geom_boxplot()
```

```
library(olsrr)
```

```
library(gvlma)
```

```
library(car)
```

```
library(gplots)
```

```
library(plm)
```

```
library(sandwich)
```

```
library(lmtest)
```

```
ggplot(data= datafinal, aes(x=Country, y=Life.expectancy, group = Year, colour = Year))+
```

```
geom_point() +
```

```
geom_smooth(method =lm, se= TRUE) +
```

```
labs(x= "Country", y= "Life.expectancy", title = "Life.expectancy Versus Country")
```

```
plotmeans(Life.expectancy~Year, error.bars="sd", main="Deviations-Years",  
data=datafinal,xlab="year",ylab="Life.expectancy")
```

```
plotmeans(Life.expectancy ~ Country, main="Deviations-Country", data=data, xlab="Country",  
ylab= "Life.expectancy" )
```

## MODELS

### #POOLING MODELS

#model1

```
Pooled_Data<-pdata.frame(datafinal,index=c("Country","Year"))
```

```
Pooled_model1 <- plm(Life.expectancy~ Schooling,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model1)
```

#model2

```
Pooled_model2 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model2)
```

#model3

```
Pooled_model3 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources+BMI,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model3)
```

#model4

```
Pooled_model4 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources+BMI+Alcohol,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model4)
```

#model5

```
Pooled_model5 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources+BMI+Alcohol+percentage.expenditure,data=Pooled  
_Data, model = 'pooling')
```

```
summary(Pooled_model5)
```

#model6

```
Pooled_model6 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources+BMI+Alcohol+percentage.expenditure+HIV.AIDS,  
data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model6)
```

```
#model7
```

```
Pooled_model7 <- plm(Life.expectancy~  
Schooling+Income.composition.of.resources+BMI+Alcohol+percentage.expenditure+HIV.AIDS+  
thinness..1.19.years,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_model7)
```

```
#model8
```

```
Pooled_model8<-  
plm(Life.expectancy~Schooling+Income.composition.of.resources+BMI+percentage.expenditure  
+HIV.AIDS+thinness..1.19.years+Diphtheria+Total.expenditure,data=Pooled_Data, model =  
'pooling')
```

```
summary(Pooled_model8)
```

```
#final pooled model(selected)
```

```
Pooled_modelfinal <-  
plm(Life.expectancy~Schooling+Income.composition.of.resources+BMI+percentage.expenditure  
+HIV.AIDS+thinness..1.19.years+Diphtheria,data=Pooled_Data, model = 'pooling')
```

```
summary(Pooled_modelfinal)
```

Entity Demeaned Models

```
#model1
```

```
EF_Model1 <- plm(Life.expectancy~ Schooling,data=Pooled_Data, model = 'within', effect =  
'individual')
```

```
summary(EF_Model1)
```

```
#model2
```

```
EF_Model2 <- plm(Life.expectancy~ Schooling+HIV.AIDS,data=Pooled_Data, model = 'within',  
effect = 'individual')
```

```
summary(EF_Model2)
```

```
#MODEL3
```

```
EF_Model4 <- plm(Life.expectancy~ Schooling+HIV.AIDS+BMI,data=Pooled_Data, model =  
'within', effect = 'individual')
```

```
summary(EF_Model4)
```

```
#model4
```

```
EF_Model5 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure,data=Pooled_Data, model = 'within', effect =  
'individual')
```

```
summary(EF_Model5)
```

```
#model5
```

```
EF_Model6 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria,data=Pooled_Data, model = 'within',  
effect = 'individual')
```

```
summary(EF_Model6)
```

```
#model6
```

```
EF_Model7 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Diphtheria+Total.expenditure+thinness..1.19.years,data=Pooled_Dat  
a, model = 'within', effect = 'individual')
```

```
summary(EF_Model7)
```

```
#model7
```

```
EF_Model8 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure,data=Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Model8)
```

```
#model8
```

```
EF_Model9 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure+infant.deaths,data=Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Model9)
```

```
#model9
```

```
EF_Model8 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure+infant.deaths+Adult.Mortality,data=Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Model8)
```

```
#model10
```

```
EF_Model11<- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure+infant.deaths+Adult.Mortality+Alcohol,data=Pooled_Data, model = 'within', effect =  
'individual')
```

```
summary(EF_Model11)
```

```
#model11
```

```
EF_Model12 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure+i
```

```
nfant.deaths+Adult.Mortality+Alcohol+Measles,data=Pooled_Data, model = 'within', effect =  
'individual')
```

```
summary(EF_Model12)
```

```
#model12
```

```
EF_Model8 <- plm(Life.expectancy~  
Schooling+HIV.AIDS+BMI+Total.expenditure+Diphtheria+thinness..1.19.years+percentage.expe  
nditure+infant.deaths+Adult.Mortality+Alcohol+Measles+Income.composition.of.resources,data=  
Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Model8)
```

```
#final model(selected)
```

```
EF_Modelfinal <- plm(Life.expectancy~  
Schooling+HIV.AIDS+Diphtheria+infant.deaths+Adult.Mortality+Alcohol+Measles+Income.com  
position.of.resources,data=Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Modelfinal)
```

```
#Binary Variables for Countries
```

```
#model1
```

```
BinCountry_Model1 <- lm(Life.expectancy~Country+Schooling,data=Pooled_Data)
```

```
summary(BinCountry_Model1)
```

```
#model2
```

```
BinCountry_Model2 <- lm(Life.expectancy~Country+Schooling+HIV.AIDS,data=Pooled_Data)
```

```
summary(BinCountry_Model2)
```

```
#model3
```

```
BinCountry_Model3 <-
```

```
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality,data=Pooled_Data)
```

```
summary(BinCountry_Model3)
```

```
#model4
```

```
BinCountry_Model4 <-
```

```
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Total.expenditure,data=Pooled_Data)
```

```
summary(BinCountry_Model4)
```

```
#model5
```

```
BinCountry_Model5 <-
```

```
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Total.expenditure+BMI,data=Pooled_Data)
```

```
summary(BinCountry_Model5)
```

```
#model6
```

```
BinCountry_Model6<-
```

```
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Total.expenditure+BMI+Income.composition.of.resources,data=Pooled_Data)
```

```
summary(BinCountry_Model6)
```

```
#model7
```

```
BinCountry_Model7 <-
```

```
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Total.expenditure+BMI+Income.composition.of.resources+thinness..1.19.years+Alcohol+Measles+infant.deaths+percentage.expenditure,data=Pooled_Data)
```

```
summary(BinCountry_Model7)
```

```
#final model(selected)
```

```

BinCountry_finalmodel<-
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Income.composition.of.res
ources+Alcohol,data=Pooled_Data)

summary(BinCountry_finalmodel)

#Binary Variables for time

#model1

BinTime_Model1 <- lm(Life.expectancy~Year+Schooling,data=Pooled_Data)

summary(BinTime_Model1)

#model2

BinTime_Model2<- lm(Life.expectancy~Year+Schooling+HIV.AIDS,data=Pooled_Data)

summary(BinTime_Model2)

#model3

BinTime_Model3 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources,data=Pooled_
Data)

summary(BinTime_Model3)

#model4

BinTime_Model4 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI,data=P
ooled_Data)

summary(BinTime_Model4)

#model5

BinTime_Model5 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years,data=Pooled_Data)

summary(BinTime_Model5)

#model6

```



```

BinTime_Model6<-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria,data=Pooled_Data)

summary(BinTime_Model6)

#model7

BinTime_Model7<-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria+Total.expenditure,data=Pooled_Data)

summary(BinTime_Model7)

#model8

BinTime_Model8 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria+Total.expenditure+Measles,data=Pooled_Data)

summary(BinTime_Model8)

#model9

BinTime_Model9 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure,data=Pooled_Data)

summary(BinTime_Model9)

#model10

BinTime_Model10 <-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths,data=
Pooled_Data)

summary(BinTime_Model10)

#model11

BinTime_Model11<-
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes
s..1.19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths+Adul
t.Mortality,data=Pooled_Data)

summary(BinTime_Model11)

```

```
#model12
```

```
BinTime_Model12 <-  
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinnes  
s..1.19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths+Adul  
t.Mortality+Alcohol,data=Pooled_Data)
```

```
summary(BinTime_Model12)
```

```
#final model(selected)
```

```
BinTime_finalmodel <-  
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+Diphth  
eria+Measles+percentage.expenditure+Adult.Mortality+Alcohol,data=Pooled_Data)
```

```
summary(BinTime_finalmodel)
```

```
#Time Demeaned Models
```

```
#model1
```

```
TF_Model1 <-plm(Life.expectancy~Schooling,data=Pooled_Data, model = 'within', effect =  
'time')
```

```
summary(TF_Model1)
```

```
#model2
```

```
TF_Model2 <-plm(Life.expectancy~Schooling+HIV.AIDS,data=Pooled_Data, model = 'within',  
effect = 'time')
```

```
summary(TF_Model2)
```

```
#model3
```

```
TF_Model3<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources,data=Pooled_Data,  
model = 'within', effect = 'time')
```

```
summary(TF_Model3)
```

```
#model4
```

```
TF_Model4<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI,data=Pooled  
_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model4)
```

```
#model5
```

```
TF_Model5<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model5)
```

```
#model6
```

```
TF_Model6<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model6)
```

```
#model7
```

```
TF_Model7 <-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model7)
```

```
#model8
```

```
TF_Model8<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure+Measles,data=Pooled_Data, model = 'within', effect =  
'time')
```

```
summary(TF_Model8)
```

```
#model9
```

```
TF_Model9<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure,data=Pooled_Data,  
model = 'within', effect = 'time')
```

```
summary(TF_Model9)
```

```
#model10
```

```
TF_Model10 <-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths,data=Pool  
ed_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model10)
```

```
#model11
```

```
TF_Model11 <-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths+Adult.M  
ortality,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model11)
```

```
#model12
```

```
TF_Model12<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths+Adult.M  
ortality+Alcohol,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Model12)
```

```
#final model
```

```
TF_Modelfinalmodel<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+Diphtheria+  
Measles+percentage.expenditure+Adult.Mortality+Alcohol,data=Pooled_Data, model = 'within',  
effect = 'time')
```

```
summary(TF_Modelfinalmodel)
```

```
#Entity and Time Fixed Models
```

```
#model1
```

```
ETF_Model1 <- plm(Life.expectancy~Schooling,data=Pooled_Data, model =  
'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model1)
```

```
#model2
```

```
ETF_Model2 <- plm(Life.expectancy~Schooling+HIV.AIDS,data=Pooled_Data, model =  
'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model2)
```

```
#model3
```

```
ETF_Model3 <-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources,data=Pooled_Data,  
model = 'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model3)
```

```
#model4
```

```
ETF_Model4 <-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI,data=Pooled  
_Data, model = 'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model4)
```

```
#model5
```

```
ETF_Model5<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years,data=Pooled_Data, model = 'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model5)
```

```
#model6
```

```
ETF_Model6 <-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria,data=Pooled_Data, model = 'within',index=c("Country","Year"), effect  
='twoways')
```

```
summary(ETF_Model6)
```

```
#model7
```

```
ETF_Model7<-
```

```
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria++Total.expenditure,data=Pooled_Data, model =  
'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Model7)
```

```
#model8
```

```

ETF_Model8 <-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria++Total.expenditure+Measles,data=Pooled_Data, model =
'within',index=c("Country","Year"), effect ='twoways')

summary(ETF_Model8)

#model9

ETF_Model9<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria++Total.expenditure+Measles+percentage.expenditure,data=Pooled_Data,
model = 'within',index=c("Country","Year"), effect ='twoways')

summary(ETF_Model9)

#model10

ETF_Model10 <-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths,data=Pool
ed_Data, model = 'within',index=c("Country","Year"), effect ='twoways')

summary(ETF_Model10)

#model11

ETF_Model11<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+infant.deaths+Adult.M
ortality,data=Pooled_Data, model = 'within',index=c("Country","Year"), effect ='twoways')

summary(ETF_Model11)

#finalmodel

ETF_Modelfinalmodel <-
plm(Life.expectancy~Schooling+HIV.AIDS+Diphtheria+Measles+Adult.Mortality+Alcohol,data
=Pooled_Data, model = 'within',index=c("Country","Year"), effect ='twoways')

summary(ETF_Modelfinalmodel)

#First Difference Model

#model1

```

```
FD_Model1 <- plm(Life.expectancy~Schooling,data=Pooled_Data,  
index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model1)
```

```
#model2
```

```
FD_Model2<- plm(Life.expectancy~Schooling+HIV.AIDS,data=Pooled_Data,  
index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model2)
```

```
#model3
```

```
FD_Model3<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources,data=Pooled_Data,  
index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model3)
```

```
#model4
```

```
FD_Model4<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI,data=Pooled  
_Data, index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model4)
```

```
#model5
```

```
FD_Model5<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years,data=Pooled_Data, index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model5)
```

```
#model6
```

```
FD_Model6<-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.  
19.years+Diphtheria,data=Pooled_Data, index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Model6)
```

```
#model7
```

```

FD_Model7<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles,data=Pooled_Data,
index=c("Country","Year"),model = 'fd')

summary(FD_Model7)

#model8

FD_Model8<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure,data=Pooled_Data,
index=c("Country","Year"),model = 'fd')

summary(FD_Model8)

#model9

FD_Model9<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+Adult.Mortality,data=P
ooled_Data, index=c("Country","Year"),model = 'fd')

summary(FD_Model9)

#model10

FD_Model10<-
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.
19.years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+Adult.Mortality+Alcoh
ol,data=Pooled_Data, index=c("Country","Year"),model = 'fd')

summary(FD_Model10)

#final model

FD_Modelfinalmodel<-
plm(Life.expectancy~Schooling+HIV.AIDS++Total.expenditure+Adult.Mortality,data=Pooled_D
ata, index=c("Country","Year"),model = 'fd')

summary(FD_Modelfinalmodel)

#random effect model1

#model

```



```

RF_Model1 <- plm(formula
=Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.19.
years+Diphtheria+Total.expenditure+Measles+percentage.expenditure+Adult.Mortality+Alcohol ,
data = Pooled_Data, model = "random", index = c("Country", "Year"))

summary(RF_Model1)

#final random effect

RF_Modelfinal<- plm(formula
=Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.19.
years+Diphtheria+Measles+percentage.expenditure+Adult.Mortality , data = Pooled_Data, model
= "random", index = c("Country", "Year"))

summary(RF_Modelfinal)

#final choices in all method

#pooling model

Pooled_Data<-pdata.frame(datafinal,index=c("Country","Year"))

Pooled_modelfinal <-
plm(Life.expectancy~Schooling+Income.composition.of.resources+BMI+percentage.expenditure
+HIV.AIDS+thinness..1.19.years+Diphtheria,data=Pooled_Data, model = 'pooling')

summary(Pooled_modelfinal)

bptest(Pooled_modelfinal)

pbgttest(Pooled_modelfinal)

#random effect model

RF_Modelfinal <- plm(formula
=Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+thinness..1.19.
years+Diphtheria+Measles+percentage.expenditure+Adult.Mortality , data = Pooled_Data, model
= "random", index = c("Country", "Year"))

summary(RF_Modelfinal)

bptest(RF_Modelfinal)

pbgttest(RF_Modelfinal)

#first difference

```

```
FD_Modelfinal <-  
plm(Life.expectancy~Schooling+HIV.AIDS++Total.expenditure+Adult.Mortality,data=Pooled_Data, index=c("Country","Year"),model = 'fd')
```

```
summary(FD_Modelfinal)
```

```
#fixed effect model(within)
```

```
ETF_Modelfinal <-  
plm(Life.expectancy~Schooling+HIV.AIDS+Diphtheria+Measles+Adult.Mortality+Alcohol,data=Pooled_Data, model = 'within',index=c("Country","Year"), effect = 'twoways')
```

```
summary(ETF_Modelfinal)
```

```
bptest(ETF_Modelfinal)
```

```
pbgttest(ETF_Modelfinal)
```

```
#time demeaned model
```

```
TF_Modelfinal <-  
plm(Life.expectancy~Schooling+HIV.AIDS+Income.composition.of.resources+BMI+Diphtheria+Measles+percentage.expenditure+Adult.Mortality+Alcohol,data=Pooled_Data, model = 'within', effect = 'time')
```

```
summary(TF_Modelfinal)
```

```
bptest(TF_Modelfinal, studentize = 'FALSE')
```

```
pbgttest(TF_Modelfinal)
```

```
#binary variable for time
```

```
BinTime_Modelfinal <-  
lm(Life.expectancy~Year+Schooling+HIV.AIDS+Income.composition.of.resources+BMI+Diphtheria+Measles+percentage.expenditure+Adult.Mortality+Alcohol,data=Pooled_Data)
```

```
summary(BinTime_Modelfinal)
```

```
bptest(BinTime_Modelfinal)
```

```
#binary variable for Country
```

```
BinCountry_Modelfinal <-  
lm(Life.expectancy~Country+Schooling+HIV.AIDS+Adult.Mortality+Income.composition.of.resources+Alcohol,data=Pooled_Data)
```

```
summary(BinCountry_Modelfinal)
```

```
bptest(BinCountry_Modelfinal)
```

```
#ENTITY demeaned
```

```
EF_Modelfinal<- plm(Life.expectancy~  
Schooling+HIV.AIDS+Diphtheria+infant.deaths+Adult.Mortality+Alcohol+Measles+Income.com  
position.of.resources,data=Pooled_Data, model = 'within', effect = 'individual')
```

```
summary(EF_Modelfinal)
```

```
bptest(EF_Modelfinal)
```

```
pbgtest(EF_Modelfinal)
```

```
#hausman test
```

```
phptest(RF_Modelfinal,ETF_Modelfinal)
```

```
#TESTING FOR TIME FIXED EFFECT
```

```
plmtest(TF_Modelfinal,c("time"),type=("bp"))
```

```
pFtest(TF_Modelfinal,EF_Modelfinal)
```

```
#testing for LM test
```

```
plmtest(Pooled_modelfinal,type=("bp"))
```