

```

library(readxl)

library(dplyr)

library(lubridate)

library(ggplot2)

library(reshape2)

library(corrplot)

library(Hmisc)

library(caTools)

library(car)

library(gridExtra)

#WorldHappinessData_2016 <- read_excel("C:/Users/Tanmay/Desktop/HBU/ANLY 502
- Analytical Methods I/Project/WorldHappinessData_2016.xlsx")
#View(WorldHappinessData_2016)

setwd("C:/Users/HCL/Documents/Nilay Back
Up/Quicky/Master/harrisburg/Sem1/ANLY 502/Project/")
WorldHappinessData_2016 <- read_excel("data_2016.xlsx", sheet = 1)

str(WorldHappinessData_2016)

## Classes 'tbl_df', 'tbl' and 'data.frame': 157 obs. of 13 variables:
## $ Country : chr "Denmark" "Switzerland" "Iceland"
"Norway" ...
## $ Region : chr "Western Europe" "Western Europe"
"Western Europe" "Western Europe" ...
## $ Happiness_Rank : num 1 2 3 4 5 6 7 8 9 10 ...
## $ Happiness_Score : num 7.53 7.51 7.5 7.5 7.41 ...
## $ Lower Confidence Interval : num 7.46 7.43 7.33 7.42 7.35 ...
## $ Upper Confidence Interval : num 7.59 7.59 7.67 7.58 7.47 ...
## $ Economy : num 1.44 1.53 1.43 1.58 1.41 ...
## $ Family : num 1.16 1.15 1.18 1.13 1.13 ...
## $ Health_Life_Expectancy : num 0.795 0.863 0.867 0.796 0.811 ...
## $ Freedom : num 0.579 0.586 0.566 0.596 0.571 ...
## $ Trust_Government_Corruption: num 0.445 0.412 0.15 0.358 0.41 ...
## $ Generosity : num 0.362 0.281 0.477 0.379 0.255 ...
## $ Dystopia_Residual : num 2.74 2.69 2.83 2.66 2.83 ...

# Data clean-up and preparation :
# Remove columns 5 and 6 as they are just CI values
Rev_WorldHappinessData_2016<- WorldHappinessData_2016[, -c(5,6)]
#View(Rev_WorldHappinessData_2016)

```

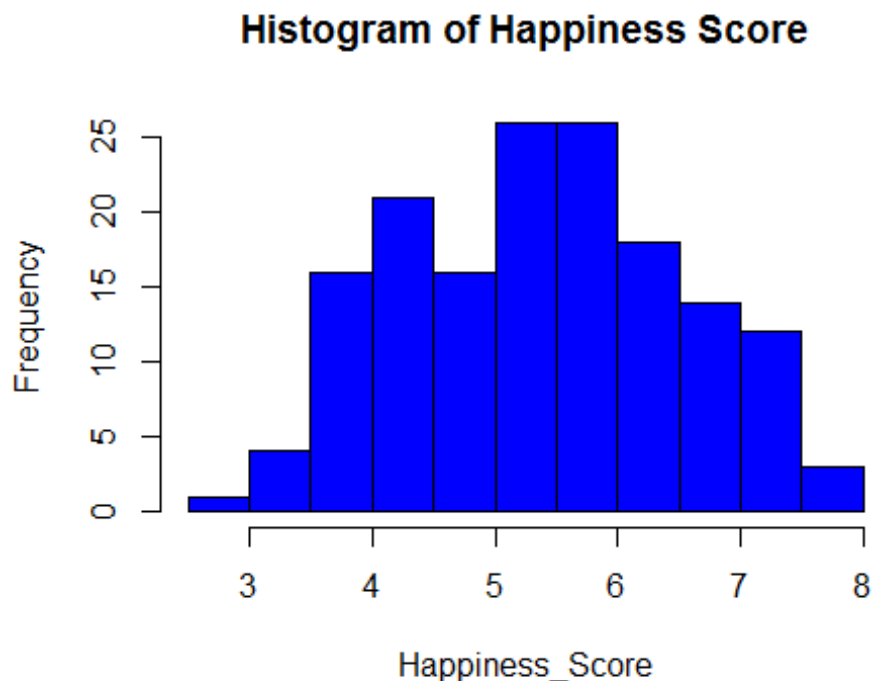
```
str(Rev_WorldHappinessData_2016)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':  157 obs. of  11 variables:
## $ Country          : chr  "Denmark" "Switzerland" "Iceland"
## "Norway" ...
## $ Region           : chr  "Western Europe" "Western Europe"
## "Western Europe" "Western Europe" ...
## $ Happiness_Rank    : num  1 2 3 4 5 6 7 8 9 10 ...
## $ Happiness_Score   : num  7.53 7.51 7.5 7.5 7.41 ...
## $ Economy           : num  1.44 1.53 1.43 1.58 1.41 ...
## $ Family            : num  1.16 1.15 1.18 1.13 1.13 ...
## $ Health_Life_Expectancy : num  0.795 0.863 0.867 0.796 0.811 ...
## $ Freedom           : num  0.579 0.586 0.566 0.596 0.571 ...
## $ Trust_Government_Corruption: num  0.445 0.412 0.15 0.358 0.41 ...
## $ Generosity        : num  0.362 0.281 0.477 0.379 0.255 ...
## $ Dystopia_Residual  : num  2.74 2.69 2.83 2.66 2.83 ...
```

```
par(mfrow = c(1,1))
```

```
### Creating a Histogram for the Happiness_score to depict the Normality of
distribution
```

```
hist(Rev_WorldHappinessData_2016$Happiness_Score,xlab = "Happiness_Score",col
= "blue",main = "Histogram of Happiness Score")
```

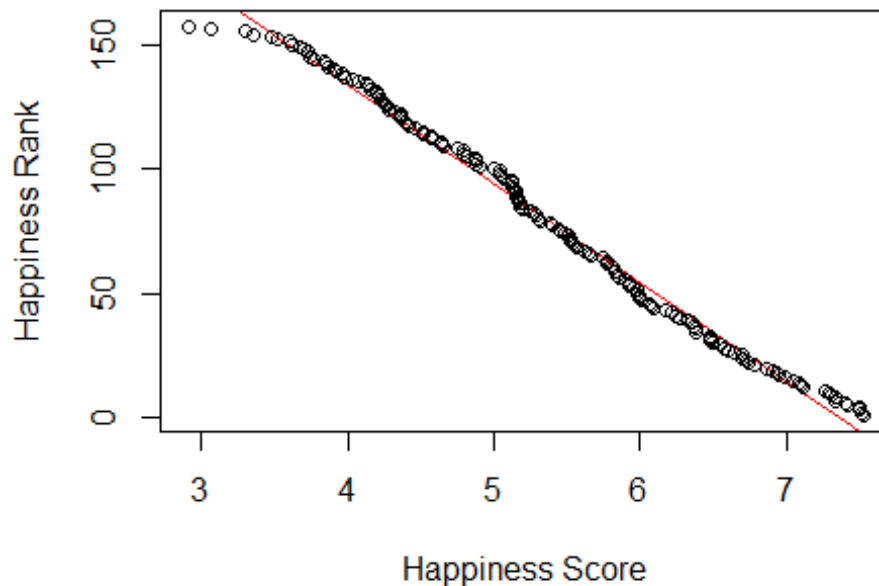


```
### Scatter plot to determine the relation between Happiness Rank and
Happiness Score
```

```
plot(Rev_WorldHappinessData_2016$Happiness_Score,Rev_WorldHappinessData_2016$
```

```
Happiness_Rank,xlab = "Happiness Score",
  ylab = "Happiness Rank",main = "Scatterplot to determine Happiness Rank
by Happiness Score",
  abline(lm(Rev_WorldHappinessData_2016$Happiness_Rank ~
Rev_WorldHappinessData_2016$Happiness_Score),col = "Red"))
```

atterplot to determine Happiness Rank by Happiness



```
### Correlation matrix for the Happiness score and other variables
```

```
corrdata <- Rev_WorldHappinessData_2016[, -1:-2]
corrdata
```

```
## # A tibble: 157 x 9
```

```
##   Happiness_Rank Happiness_Score Economy Family Health_Life_Expe~ Freedom
##   <dbl>          <dbl>    <dbl>  <dbl>      <dbl>      <dbl>
## 1             1.         7.53    1.44    1.16      0.795    0.579
## 2             2.         7.51    1.53    1.15      0.863    0.586
## 3             3.         7.50    1.43    1.18      0.867    0.566
## 4             4.         7.50    1.58    1.13      0.796    0.596
## 5             5.         7.41    1.41    1.13      0.811    0.571
## 6             6.         7.40    1.44    1.10      0.828    0.574
## 7             7.         7.34    1.46    1.03      0.812    0.552
## 8             8.         7.33    1.36    1.17      0.831    0.581
## 9             9.         7.31    1.44    1.10      0.851    0.568
## 10            10.         7.29    1.45    1.09      0.831    0.582
```

```
## # ... with 147 more rows, and 3 more variables:
```

```
## #   Trust_Government_Corruption <dbl>, Generosity <dbl>,
```

```
## #   Dystopia_Residual <dbl>
```

```
res <- rcorr(as.matrix(corrdata))
res
```

```
##              Happiness_Rank Happiness_Score Economy Family
## Happiness_Rank              1.00          -1.00  -0.79  -0.73
## Happiness_Score            -1.00              1.00   0.79   0.74
## Economy                   -0.79              0.79   1.00   0.67
## Family                    -0.73              0.74   0.67   1.00
## Health_Life_Expectancy    -0.77              0.77   0.84   0.59
## Freedom                   -0.56              0.57   0.36   0.45
## Trust_Government_Corruption -0.39              0.40   0.29   0.21
## Generosity                 -0.15              0.16  -0.03   0.09
## Dystopia_Residual         -0.54              0.54   0.07   0.12
##
##              Health_Life_Expectancy Freedom
## Happiness_Rank              -0.77  -0.56
## Happiness_Score              0.77   0.57
## Economy                     0.84   0.36
## Family                      0.59   0.45
## Health_Life_Expectancy      1.00   0.34
## Freedom                     0.34   1.00
## Trust_Government_Corruption  0.25   0.50
## Generosity                   0.08   0.36
## Dystopia_Residual           0.10   0.09
##
##              Trust_Government_Corruption Generosity
## Happiness_Rank              -0.39  -0.15
## Happiness_Score              0.40   0.16
## Economy                     0.29  -0.03
## Family                      0.21   0.09
## Health_Life_Expectancy      0.25   0.08
## Freedom                     0.50   0.36
## Trust_Government_Corruption  1.00   0.31
## Generosity                   0.31   1.00
## Dystopia_Residual           0.00  -0.13
##
##              Dystopia_Residual
## Happiness_Rank             -0.54
## Happiness_Score              0.54
## Economy                     0.07
## Family                      0.12
## Health_Life_Expectancy      0.10
## Freedom                     0.09
## Trust_Government_Corruption  0.00
## Generosity                  -0.13
## Dystopia_Residual           1.00
##
## n= 157
##
##
## P
##              Happiness_Rank Happiness_Score Economy Family
## Happiness_Rank              0.0000          0.0000  0.0000
```

```
## Happiness_Score      0.0000      0.0000      0.0000      0.0000
## Economy              0.0000      0.0000      0.0000      0.0000
## Family               0.0000      0.0000      0.0000      0.0000
## Health_Life_Expectancy 0.0000      0.0000      0.0000      0.0000
## Freedom              0.0000      0.0000      0.0000      0.0000
## Trust_Government_Corruption 0.0000      0.0000      0.0002      0.0072
## Generosity           0.0693      0.0498      0.7509      0.2643
## Dystopia_Residual    0.0000      0.0000      0.3931      0.1355
##
## Health_Life_Expectancy Freedom
## Happiness_Rank      0.0000      0.0000
## Happiness_Score     0.0000      0.0000
## Economy             0.0000      0.0000
## Family              0.0000      0.0000
## Health_Life_Expectancy 0.0000
## Freedom            0.0000
## Trust_Government_Corruption 0.0016      0.0000
## Generosity          0.3442      0.0000
## Dystopia_Residual   0.2088      0.2537
##
## Trust_Government_Corruption Generosity
## Happiness_Rank      0.0000      0.0693
## Happiness_Score     0.0000      0.0498
## Economy             0.0002      0.7509
## Family              0.0072      0.2643
## Health_Life_Expectancy 0.0016      0.3442
## Freedom            0.0000      0.0000
## Trust_Government_Corruption 0.0000
## Generosity          0.0000
## Dystopia_Residual   0.9712      0.0968
##
## Dystopia_Residual
## Happiness_Rank      0.0000
## Happiness_Score     0.0000
## Economy             0.3931
## Family              0.1355
## Health_Life_Expectancy 0.2088
## Freedom            0.2537
## Trust_Government_Corruption 0.9712
## Generosity          0.0968
## Dystopia_Residual
```

Extract the correlation coefficients

res\$r

```
## Happiness_Rank Happiness_Score Economy
## Happiness_Rank      1.0000000 -0.9957434 -0.79357712
## Happiness_Score     -0.9957434      1.0000000  0.79032202
## Economy             -0.7935771      0.7903220  1.00000000
## Family              -0.7332763      0.7392516  0.66953969
## Health_Life_Expectancy -0.7679908      0.7653843  0.83706723
## Freedom            -0.5571687      0.5668267  0.36228285
## Trust_Government_Corruption -0.3871016      0.4020322  0.29418478
```

```

## Generosity -0.1453688 0.1568478 -0.02553066
## Dystopia_Residual -0.5426158 0.5437376 0.06862402
## Family Health_Life_Expectancy Freedom
## Happiness_Rank -0.73327635 -0.76799078 -0.55716871
## Happiness_Score 0.73925158 0.76538433 0.56682667
## Economy 0.66953969 0.83706723 0.36228285
## Family 1.00000000 0.58837678 0.45020820
## Health_Life_Expectancy 0.58837678 1.00000000 0.34119929
## Freedom 0.45020820 0.34119929 1.00000000
## Trust_Government_Corruption 0.21356094 0.24958329 0.50205397
## Generosity 0.08962885 0.07598731 0.36175133
## Dystopia_Residual 0.11967231 0.10085705 0.09163094
## Trust_Government_Corruption Generosity
## Happiness_Rank -0.387101640 -0.14536878
## Happiness_Score 0.402032245 0.15684780
## Economy 0.294184775 -0.02553066
## Family 0.213560938 0.08962885
## Health_Life_Expectancy 0.249583285 0.07598731
## Freedom 0.502053974 0.36175133
## Trust_Government_Corruption 1.000000000 0.30592986
## Generosity 0.305929861 1.000000000
## Dystopia_Residual -0.002909346 -0.13300141
## Dystopia_Residual
## Happiness_Rank -0.542615800
## Happiness_Score 0.543737614
## Economy 0.068624025
## Family 0.119672314
## Health_Life_Expectancy 0.100857046
## Freedom 0.091630941
## Trust_Government_Corruption -0.002909346
## Generosity -0.133001407
## Dystopia_Residual 1.000000000

```

Extract p-values
res\$P

```

## Happiness_Rank Happiness_Score Economy
## Happiness_Rank NA 0.000000e+00 0.000000e+00
## Happiness_Score 0.000000e+00 NA 0.000000e+00
## Economy 0.000000e+00 0.000000e+00 NA
## Family 0.000000e+00 0.000000e+00 0.000000e+00
## Health_Life_Expectancy 0.000000e+00 0.000000e+00 0.000000e+00
## Freedom 3.508305e-14 1.021405e-14 3.123973e-06
## Trust_Government_Corruption 5.492284e-07 1.797843e-07 1.841886e-04
## Generosity 6.928267e-02 4.979137e-02 7.509449e-01
## Dystopia_Residual 2.122746e-13 1.851852e-13 3.931090e-01
## Family Health_Life_Expectancy
## Happiness_Rank 0.000000e+00 0.000000e+00
## Happiness_Score 0.000000e+00 0.000000e+00
## Economy 0.000000e+00 0.000000e+00

```

```

## Family NA 4.440892e-16
## Health_Life_Expectancy 4.440892e-16 NA
## Freedom 3.300642e-09 1.225460e-05
## Trust_Government_Corruption 7.240343e-03 1.619787e-03
## Generosity 2.642850e-01 3.442107e-01
## Dystopia_Residual 1.354692e-01 2.088101e-01
## Freedom Trust_Government_Corruption
## Happiness_Rank 3.508305e-14 5.492284e-07
## Happiness_Score 1.021405e-14 1.797843e-07
## Economy 3.123973e-06 1.841886e-04
## Family 3.300642e-09 7.240343e-03
## Health_Life_Expectancy 1.225460e-05 1.619787e-03
## Freedom NA 2.100009e-11
## Trust_Government_Corruption 2.100009e-11 NA
## Generosity 3.237423e-06 9.759224e-05
## Dystopia_Residual 2.537199e-01 9.711526e-01
## Generosity Dystopia_Residual
## Happiness_Rank 6.928267e-02 2.122746e-13
## Happiness_Score 4.979137e-02 1.851852e-13
## Economy 7.509449e-01 3.931090e-01
## Family 2.642850e-01 1.354692e-01
## Health_Life_Expectancy 3.442107e-01 2.088101e-01
## Freedom 3.237423e-06 2.537199e-01
## Trust_Government_Corruption 9.759224e-05 9.711526e-01
## Generosity NA 9.679955e-02
## Dystopia_Residual 9.679955e-02 NA

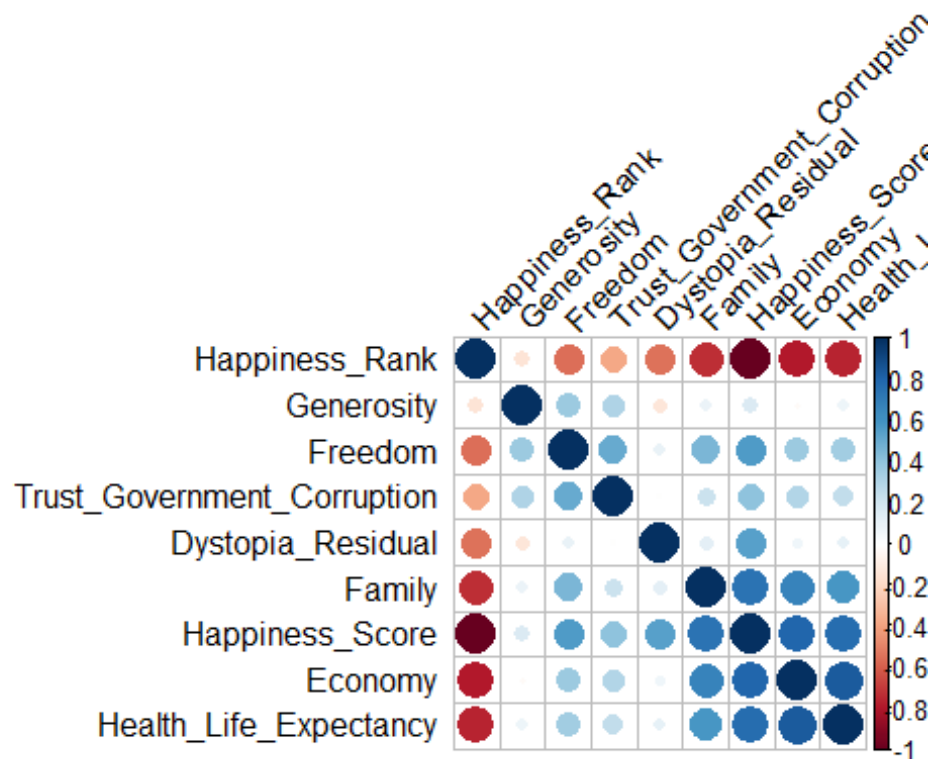
## Plot Correlation plot
res2<-rcorr(as.matrix(corrddata[,1:8]))
res2

## Happiness_Rank Happiness_Score Economy Family
## Happiness_Rank 1.00 -1.00 -0.79 -0.73
## Happiness_Score -1.00 1.00 0.79 0.74
## Economy -0.79 0.79 1.00 0.67
## Family -0.73 0.74 0.67 1.00
## Health_Life_Expectancy -0.77 0.77 0.84 0.59
## Freedom -0.56 0.57 0.36 0.45
## Trust_Government_Corruption -0.39 0.40 0.29 0.21
## Generosity -0.15 0.16 -0.03 0.09
## Health_Life_Expectancy Freedom
## Happiness_Rank -0.77 -0.56
## Happiness_Score 0.77 0.57
## Economy 0.84 0.36
## Family 0.59 0.45
## Health_Life_Expectancy 1.00 0.34
## Freedom 0.34 1.00
## Trust_Government_Corruption 0.25 0.50
## Generosity 0.08 0.36
## Trust_Government_Corruption Generosity

```

```
## Happiness_Rank -0.39 -0.15
## Happiness_Score 0.40 0.16
## Economy 0.29 -0.03
## Family 0.21 0.09
## Health_Life_Expectancy 0.25 0.08
## Freedom 0.50 0.36
## Trust_Government_Corruption 1.00 0.31
## Generosity 0.31 1.00
##
## n= 157
##
##
## P
##
## Happiness_Rank Happiness_Score Economy Family
## Happiness_Rank 0.0000 0.0000 0.0000 0.0000
## Happiness_Score 0.0000 0.0000 0.0000 0.0000
## Economy 0.0000 0.0000 0.0000 0.0000
## Family 0.0000 0.0000 0.0000 0.0000
## Health_Life_Expectancy 0.0000 0.0000 0.0000 0.0000
## Freedom 0.0000 0.0000 0.0000 0.0000
## Trust_Government_Corruption 0.0000 0.0000 0.0002 0.0072
## Generosity 0.0693 0.0498 0.7509 0.2643
##
## Health_Life_Expectancy Freedom
## Happiness_Rank 0.0000 0.0000
## Happiness_Score 0.0000 0.0000
## Economy 0.0000 0.0000
## Family 0.0000 0.0000
## Health_Life_Expectancy 0.0000
## Freedom 0.0000
## Trust_Government_Corruption 0.0016 0.0000
## Generosity 0.3442 0.0000
##
## Trust_Government_Corruption Generosity
## Happiness_Rank 0.0000 0.0693
## Happiness_Score 0.0000 0.0498
## Economy 0.0002 0.7509
## Family 0.0072 0.2643
## Health_Life_Expectancy 0.0016 0.3442
## Freedom 0.0000 0.0000
## Trust_Government_Corruption 0.0000
## Generosity 0.0000
```

```
corrplot(res$r,type = "full",order = "hclust",tl.col = "black", tl.srt = 45)
```

```
## Distribution of Happiness Score by Region
p<- Rev_WorldHappinessData_2016 %>%
  dplyr::mutate(Region=as.factor(Region)) %>%
  dplyr::select(Happiness_Score, Region) %>%
  ggplot2::ggplot(ggplot2::aes(x=Happiness_Score)) +
  ggplot2::geom_density(lty=3) +
  ggplot2::geom_density(ggplot2::aes(fill=Region, colour=Region), alpha=0.55)
+
  ggplot2::xlab("Happiness_Score") +
  ggplot2::ggtitle("Distribution of Happiness_Score by Region") +
  ggplot2::labs(fill="Region", y="Density")

## Warning: package 'bindrcpp' was built under R version 3.4.4

# Display the plot

## Distribution of Economy GDP by Region
q <- Rev_WorldHappinessData_2016 %>%
  dplyr::mutate(Region=as.factor(Region)) %>%
  dplyr::select(Economy, Region) %>%
  ggplot2::ggplot(ggplot2::aes(x=Economy)) +
  ggplot2::geom_density(lty=3) +
  ggplot2::geom_density(ggplot2::aes(fill=Region, colour=Region), alpha=0.55)
+
  ggplot2::xlab("Economy_GDP") +
  ggplot2::ggtitle("Distribution of Economy_GDP by Region") +
```

```

ggplot2::labs(fill="Region", y="Density")
# Display the plot

### Distribution of Family by Region
r <- Rev_WorldHappinessData_2016 %>%
  dplyr::mutate(Region=as.factor(Region)) %>%
  dplyr::select(Family, Region) %>%
  ggplot2::ggplot(ggplot2::aes(x=Family)) +
  ggplot2::geom_density(lty=3) +
  ggplot2::geom_density(ggplot2::aes(fill=Region, colour=Region), alpha=0.55)
+
  ggplot2::xlab("Family") +
  ggplot2::ggtitle("Distribution of Family by Region") +
  ggplot2::labs(fill="Region", y="Density")

# Display the plot

### Distribution of Health_Life_expectancy by Region
s<- Rev_WorldHappinessData_2016 %>%
  dplyr::mutate(Region=as.factor(Region)) %>%
  dplyr::select(Health_Life_Expectancy, Region) %>%
  ggplot2::ggplot(ggplot2::aes(x=Health_Life_Expectancy)) +
  ggplot2::geom_density(lty=3) +
  ggplot2::geom_density(ggplot2::aes(fill=Region, colour=Region), alpha=0.55)
+
  ggplot2::xlab("Health_Life_Expectancy") +
  ggplot2::ggtitle("Distribution of Health_Life_Expectancy by Region") +
  ggplot2::labs(fill="Region", y="Density")

# Display the plot

#####
# train

# Remove Country and Region from data as they are Categorical values
df = Rev_WorldHappinessData_2016[, -c(1,2)]
str(df)

## Classes 'tbl_df', 'tbl' and 'data.frame':   157 obs. of  9 variables:
## $ Happiness_Rank      : num  1 2 3 4 5 6 7 8 9 10 ...
## $ Happiness_Score     : num  7.53 7.51 7.5 7.5 7.41 ...
## $ Economy             : num  1.44 1.53 1.43 1.58 1.41 ...
## $ Family              : num  1.16 1.15 1.18 1.13 1.13 ...
## $ Health_Life_Expectancy : num  0.795 0.863 0.867 0.796 0.811 ...
## $ Freedom             : num  0.579 0.586 0.566 0.596 0.571 ...
## $ Trust_Government_Corruption: num  0.445 0.412 0.15 0.358 0.41 ...

```

```

## $ Generosity                : num  0.362 0.281 0.477 0.379 0.255 ...
## $ Dystopia_Residual          : num  2.74 2.69 2.83 2.66 2.83 ...

## Training and Testing the processed data-set:
set.seed(101)

# Randomly Splitting the sample, basically randomly assigns a booleans to a
# new column "sample"
sample <- sample.split(df$Happiness_Score, SplitRatio = 0.70) # SplitRatio =
# percent of sample==TRUE

# Creating a train data
train = subset(df, sample == TRUE)

# Testing the Data
test = subset(df, sample == FALSE)

#####
# Model selection using Multiple Linear Regression:

# 1. Consider 'Happiness_score' against 'Economy GDP per Capita' for
# Multiple Linear Regression model; given by model1
## Forward selection model
model1 <- lm(Happiness_Score ~ Economy ,train)
summary(model1)

##
## Call:
## lm(formula = Happiness_Score ~ Economy, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.85783 -0.46877 -0.03564  0.56025  2.17835
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.2617     0.1610   20.26  <2e-16 ***
## Economy       2.2286     0.1562   14.27  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.7093 on 107 degrees of freedom
## Multiple R-squared:  0.6554, Adjusted R-squared:  0.6522
## F-statistic: 203.5 on 1 and 107 DF,  p-value: < 2.2e-16

#65%

model2 <- lm(Happiness_Score ~ Economy+Family ,train)
summary(model2)

```

```
##
## Call:
## lm(formula = Happiness_Score ~ Economy + Family, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.76628 -0.43484  0.04531  0.42332  2.26964
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   2.6228     0.1967  13.337 < 2e-16 ***
## Economy       1.5213     0.2033   7.484 2.22e-11 ***
## Family        1.6289     0.3351   4.861 4.07e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6444 on 106 degrees of freedom
## Multiple R-squared:  0.7182, Adjusted R-squared:  0.7129
## F-statistic: 135.1 on 2 and 106 DF,  p-value: < 2.2e-16
```

#71%

```
model3 <- lm(Happiness_Score ~ Economy+Family+ Health_Life_Expectancy ,train)
summary(model3)
```

```
##
## Call:
## lm(formula = Happiness_Score ~ Economy + Family + Health_Life_Expectancy,
##      data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.67279 -0.37257  0.03052  0.40805  2.25427
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   2.5077     0.1940  12.927 < 2e-16 ***
## Economy       0.8671     0.2972   2.918  0.00432 **
## Family        1.4993     0.3267   4.589 1.24e-05 ***
## Health_Life_Expectancy 1.5183     0.5178   2.932  0.00413 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.6225 on 105 degrees of freedom
## Multiple R-squared:  0.7395, Adjusted R-squared:  0.7321
## F-statistic: 99.38 on 3 and 105 DF,  p-value: < 2.2e-16
```

#73%

```
model4 <- lm(Happiness_Score ~ Economy+Family+ Health_Life_Expectancy +
```

```

Freedom,train)
summary(model4)

##
## Call:
## lm(formula = Happiness_Score ~ Economy + Family + Health_Life_Expectancy +
##     Freedom, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.43885 -0.22300 -0.01371  0.32842  1.51214
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.0850     0.1823  11.438 < 2e-16 ***
## Economy           0.7234     0.2586   2.797 0.006145 **
## Family            1.0977     0.2909   3.773 0.000268 ***
## Health_Life_Expectancy 1.5970     0.4488   3.558 0.000564 ***
## Freedom           2.2734     0.3795   5.990 3.04e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5393 on 104 degrees of freedom
## Multiple R-squared:  0.8063, Adjusted R-squared:  0.7989
## F-statistic: 108.3 on 4 and 104 DF,  p-value: < 2.2e-16

```

#79

```

model5 <- lm(Happiness_Score ~ Economy+Family+ Health_Life_Expectancy +
Freedom + Trust_Government_Corruption,train)
summary(model5)

##
## Call:
## lm(formula = Happiness_Score ~ Economy + Family + Health_Life_Expectancy +
##     Freedom + Trust_Government_Corruption, data = train)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.63274 -0.21810  0.01872  0.29009  1.36488
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.1256     0.1791  11.869 < 2e-16 ***
## Economy           0.6800     0.2536   2.681 0.008542 **
## Family            1.1409     0.2851   4.002 0.000119 ***
## Health_Life_Expectancy 1.5696     0.4391   3.574 0.000536 ***
## Freedom           1.7247     0.4364   3.953 0.000142 ***
## Trust_Government_Corruption 1.3047     0.5455   2.392 0.018577 *
## ---

```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5275 on 103 degrees of freedom
## Multiple R-squared:  0.8165, Adjusted R-squared:  0.8076
## F-statistic: 91.69 on 5 and 103 DF,  p-value: < 2.2e-16
```

#80

Adding geneoristy does not result in higher adjusted R2, so this is our best model

Grab residuals

```
res <- residuals(model5)
# Convert to DataFrame for ggplot
res <- as.data.frame(res)
head(res)
```

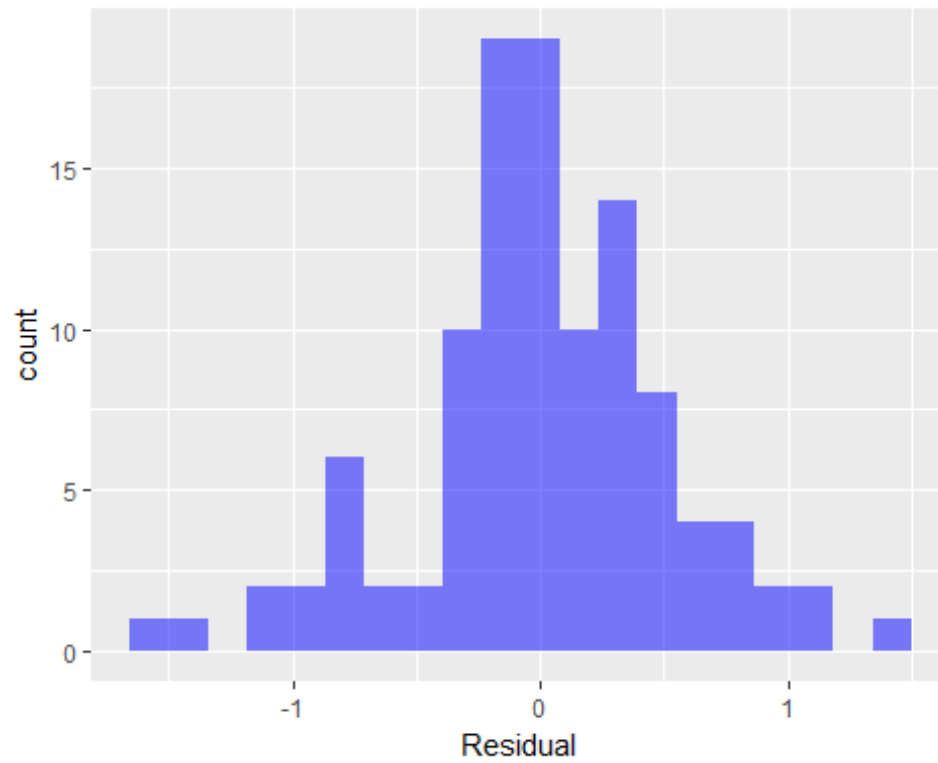
```
##           res
## 1 0.2651225
## 2 0.1361174
## 3 0.2701727
## 4 0.2441902
## 5 0.3513707
## 6 0.4256380
```

```
##           res
## 1 0.2651225
## 2 0.1361174
## 3 0.2701727
## 4 0.2441902
## 5 0.3513707
## 6 0.4256380
```

```
Residual = res
```

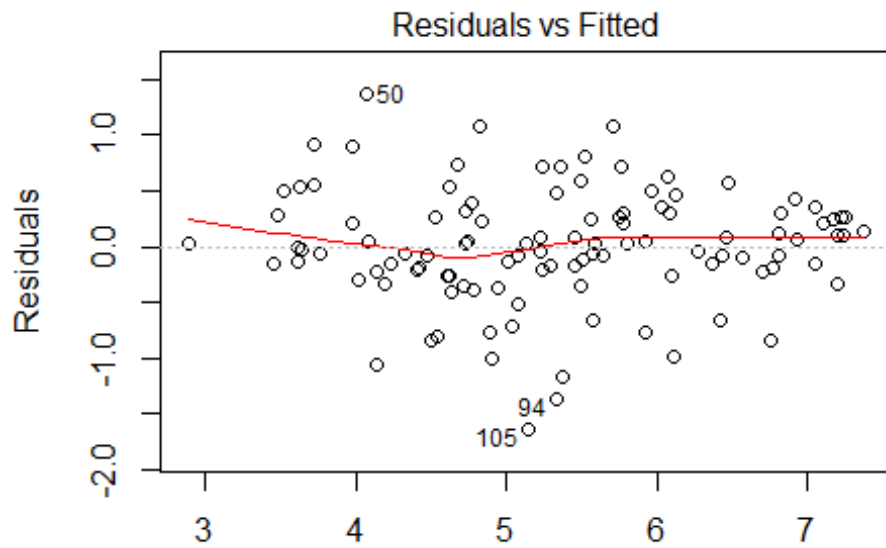
```
ggplot(Residual,aes(Residual)) + geom_histogram(fill='blue',alpha=0.5, bins
= 20)
```

```
## Don't know how to automatically pick scale for object of type data.frame.
Defaulting to continuous.
```

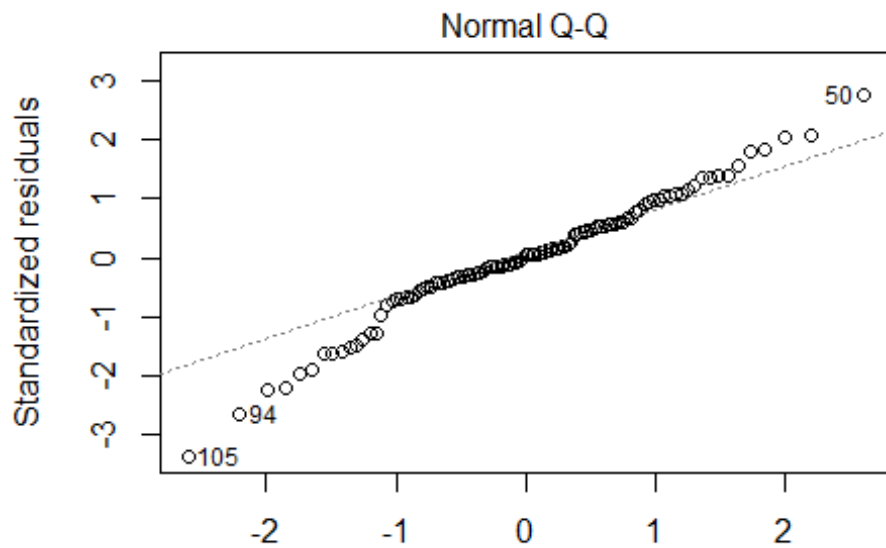


```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

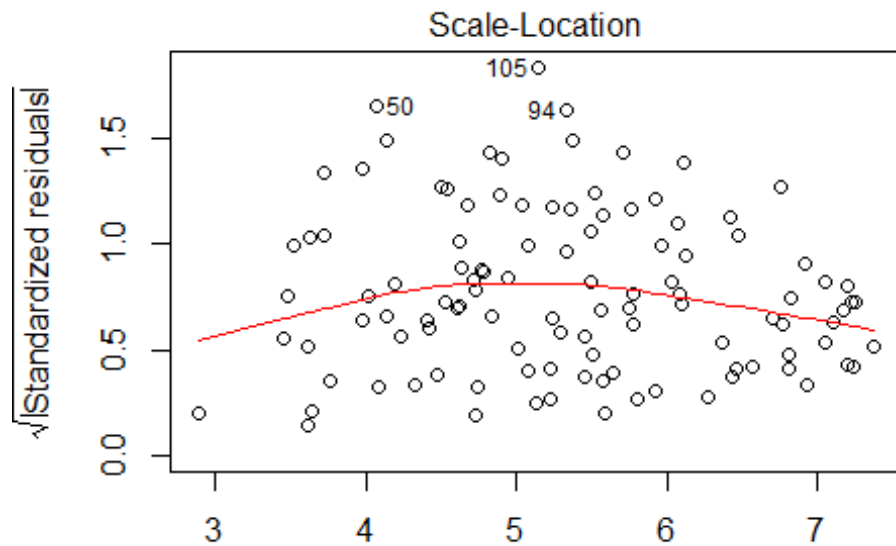
```
plot(model5)
```



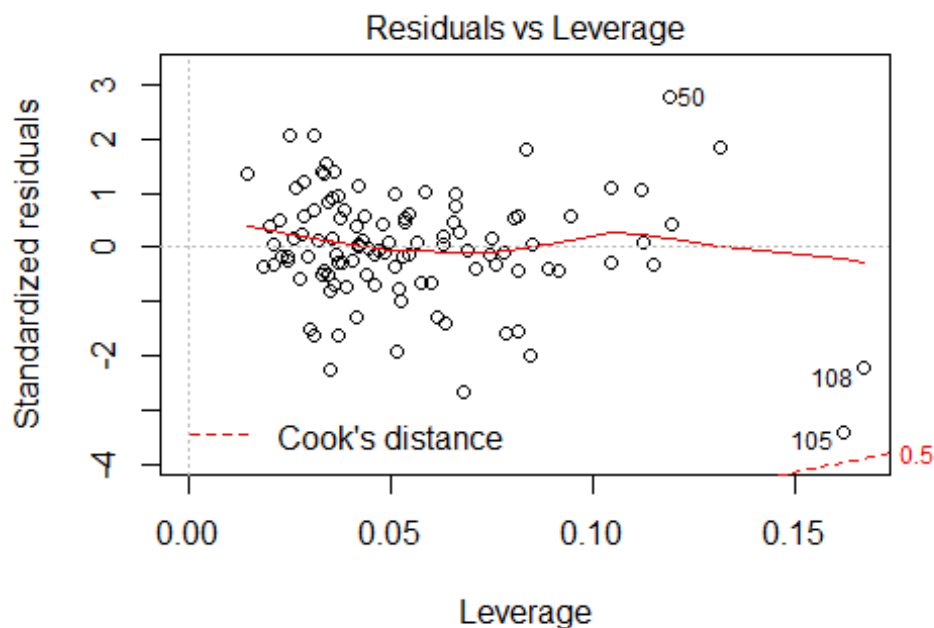
Fitted values
 $\text{appiness_Score} \sim \text{Economy} + \text{Family} + \text{Health_Life_Expectancy} + \text{Fre}$



Theoretical Quantiles
 $\text{appiness_Score} \sim \text{Economy} + \text{Family} + \text{Health_Life_Expectancy} + \text{Fre}$



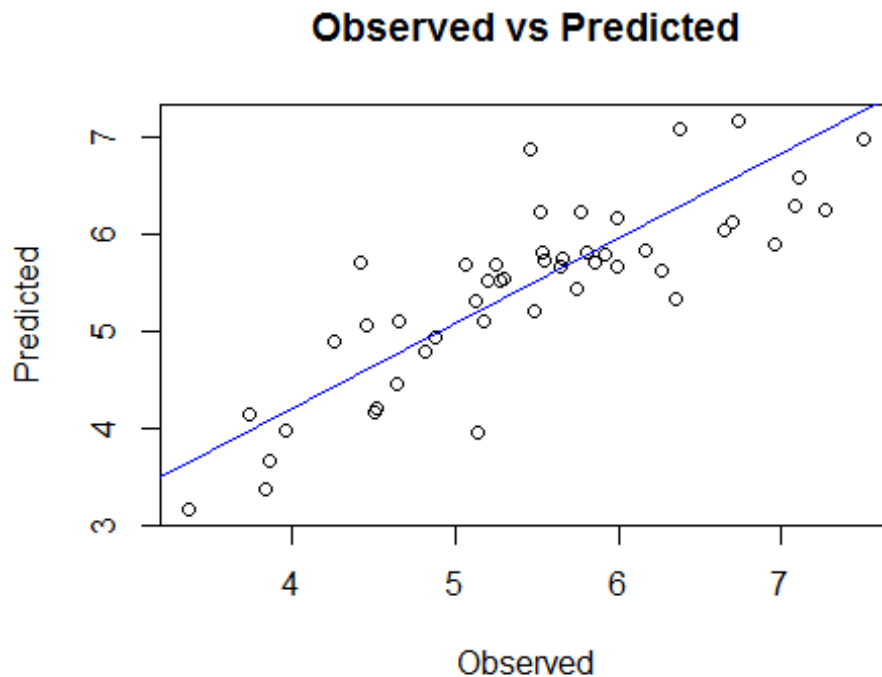
Happiness_Score ~ Economy + Family + Health_Life_Expectancy + Fre



Happiness_Score ~ Economy + Family + Health_Life_Expectancy + Fre

```
G3.predictions <- predict(model15,test)
results <- cbind(G3.predictions,test$Happiness_Score)
colnames(results) <- c('Predicted','Observed')
results <- as.data.frame(results)
```

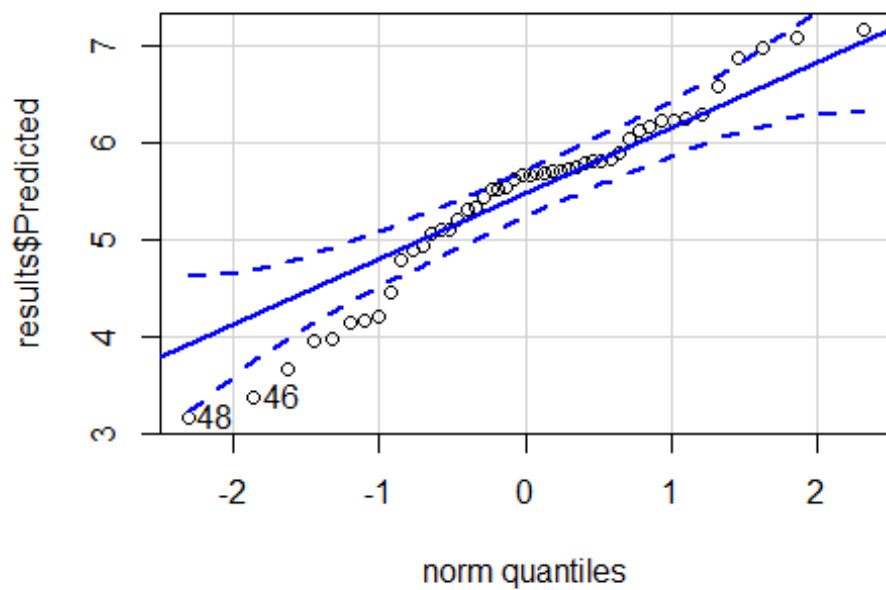
```
plot(results$Observed, results$Predicted, main = "Observed vs Predicted",
xlab = "Observed", ylab = "Predicted")
abline(lm(results$Observed ~ results$Predicted ), col = "blue")
```



```
## Alternate way of showing the Actuals vs Predicted
##plot(results$Predicted)
##lines(results$Observed)

## Another Option
#plot_internal <- ggplot(results)+aes(x = Observed, y =
Predicted)+geom_point()+
# + geom_smooth(method = "lm")
# plot_internal

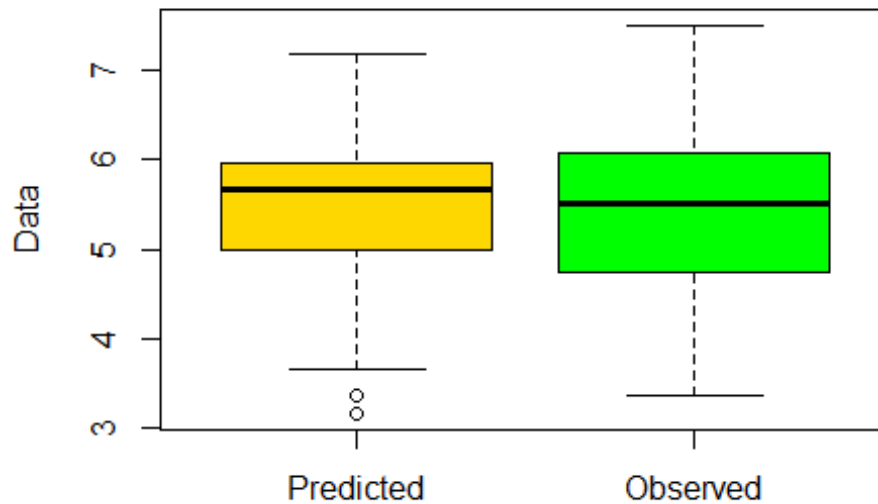
## if data follows normal distribution or not
qqPlot(results$Predicted, distribution = "norm")
```



```
## [1] 48 46
```

```
boxplot(results, main="Happiness Score- Predicted vs Observed",
col=c("gold", "green"), ylab="Data", names=c("Predicted", "Observed"))
```

Happiness Score- Predicted vs Observed



```
mse <- mean((results$Observed-results$Predicted)^2)
print(mse)

## [1] 0.3145682

#the result is ok model and there is no overfitting
## Multiple linear regression model for happiness score is - 2.1256 +0.6800
Economy + 1.14 Family + 1.56969 Health_Life_expectancy + 1.7247 Freedom +
1.3047 Trust Government Corruption

#### Predictor vs Residual

plot1 = ggplot(train, aes(Economy, residuals(model5))) + geom_point() +
geom_smooth()
plot2=ggplot(train, aes(Family, residuals(model5))) + geom_point() +
geom_smooth()
plot3=ggplot(train, aes(Health_Life_Expectancy, residuals(model5))) +
geom_point() + geom_smooth()
plot4=ggplot(train, aes(Freedom, residuals(model5))) + geom_point() +
geom_smooth()
plot5=ggplot(train, aes(Trust_Government_Corruption, residuals(model5))) +
geom_point() + geom_smooth()

grid.arrange(plot1,plot2,plot3,plot4,plot5,ncol=2,nrow=3)

## `geom_smooth()` using method = 'loess'
```

```
## `geom_smooth()` using method = 'loess'  
## `geom_smooth()` using method = 'loess'  
## `geom_smooth()` using method = 'loess'  
## `geom_smooth()` using method = 'loess'
```

