World_Happiness_report

Raghavendran Shankar 19 March 2017

\$ Generosity

\$ Dystopia.Residual

Loading dataset and libraries

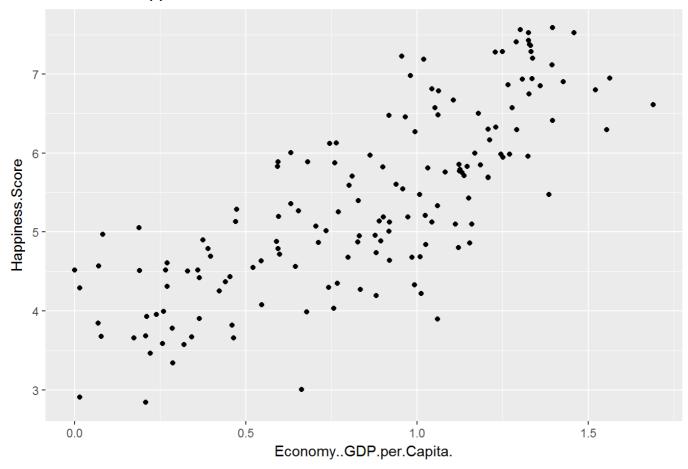
```
library (ggplot2)
## Warning: package 'ggplot2' was built under R version 3.3.2
library (ggrepel)
## Warning: package 'ggrepel' was built under R version 3.3.2
library (reshape2)
## Warning: package 'reshape2' was built under R version 3.3.2
library (GGally)
## Warning: package 'GGally' was built under R version 3.3.2
happy 2015 <- read.csv('2015.csv')</pre>
str(head(happy 2015))
## 'data.frame': 6 obs. of 12 variables:
                                 : Factor w/ 158 levels "Afghanistan",..: 136 59
## $ Country
38 106 25 46
## $ Region
                                  : Factor w/ 10 levels "Australia and New Zealan
d",...: 10 10 10 10 6 10
## $ Happiness.Rank
                                 : int 1 2 3 4 5 6
## $ Happiness.Score
                                 : num 7.59 7.56 7.53 7.52 7.43 ...
## $ Standard.Error
                                 : num 0.0341 0.0488 0.0333 0.0388 0.0355 ...
## $ Economy..GDP.per.Capita. : num 1.4 1.3 1.33 1.46 1.33 ...
## $ Family
                                 : num 1.35 1.4 1.36 1.33 1.32 ...
## $ Health..Life.Expectancy. : num 0.941 0.948 0.875 0.885 0.906 ...
## $ Freedom
                                 : num 0.666 0.629 0.649 0.67 0.633 ...
## $ Trust..Government.Corruption.: num 0.42 0.141 0.484 0.365 0.33 ...
```

Data Visualization with ggplots and geom points

: num 0.297 0.436 0.341 0.347 0.458 ... : num 2.52 2.7 2.49 2.47 2.45 ...

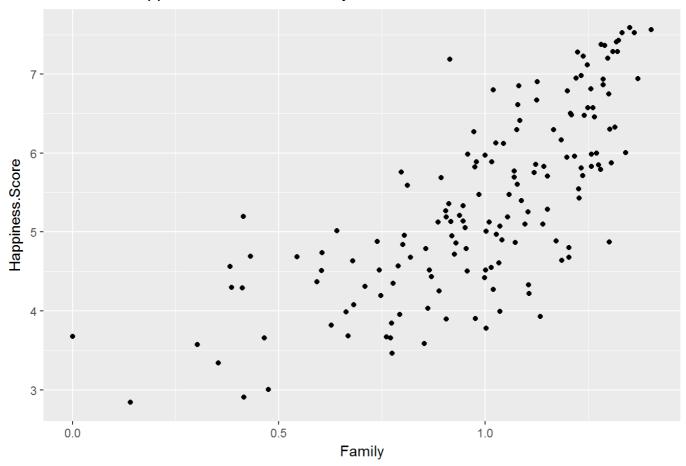
```
# ggplot for various features of Country across Happiness score
ggplot(data = happy_2015,aes(y = Happiness.Score,x = Economy..GDP.per.Capita.))+geo
m_point()+labs(title = "GGPlot of Happiness Score with GDP")
```

GGPlot of Happiness Score with GDP



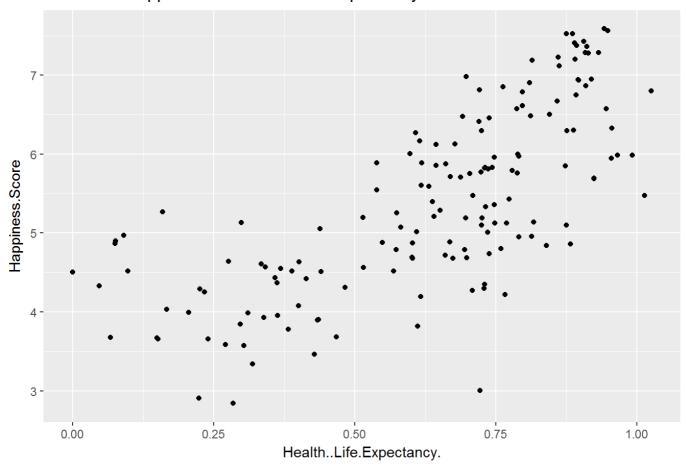
 $ggplot(data = happy_2015, aes(y = Happiness.Score, x = Family)) + geom_point() + labs(title = "GGPlot of Happiness Score with Family")$

GGPlot of Happiness Score with Family



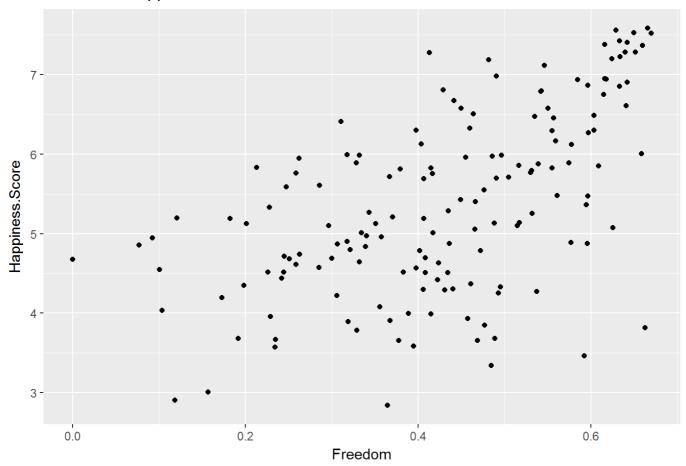
 $\label{eq:ggplot} $$ \gcd(\text{data} = \text{happy}_2015, \text{aes}(\text{y} = \text{Happiness}.\text{Score}, \text{x} = \text{Health..Life}.\text{Expectancy.})) + \text{geo} \\ \text{m_point}() + \text{labs}(\text{title} = \text{"GGPlot of Happiness Score with Life expectancy"}) \\$

GGPlot of Happiness Score with Life expectancy



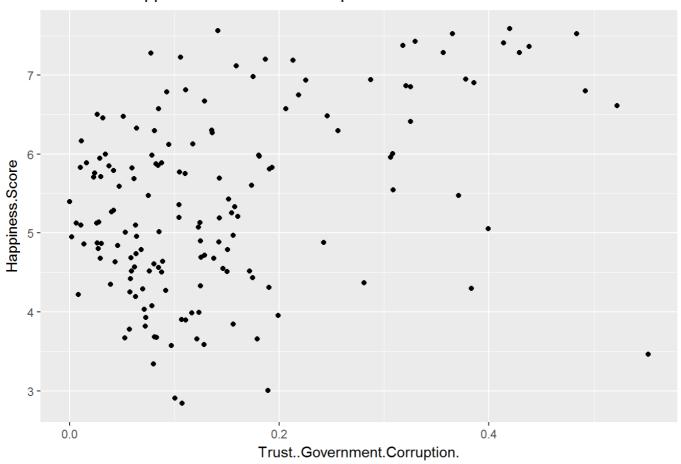
ggplot(data = happy_2015,aes(y = Happiness.Score,x = Freedom))+geom_point()+labs(t
itle = "GGPlot of Happiness Score with Freedom")

GGPlot of Happiness Score with Freedom



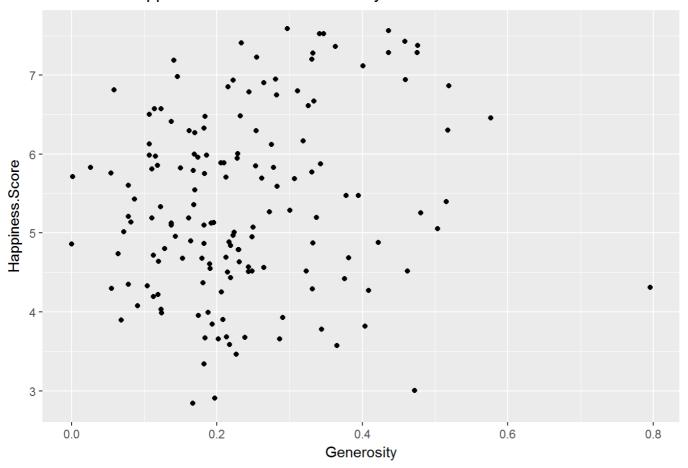
ggplot(data = happy_2015, aes(y = Happiness.Score, x = Trust..Government.Corruption.)
) +geom_point() +labs(title = "GGPlot of Happiness Score with Corruption")

GGPlot of Happiness Score with Corruption



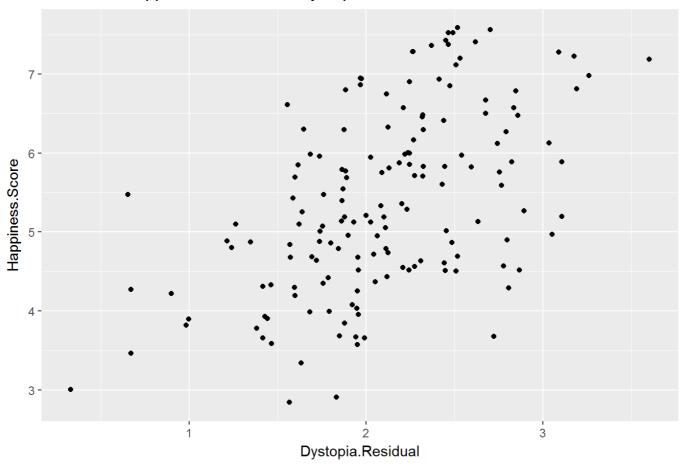
ggplot(data = happy_2015,aes(y = Happiness.Score,x = Generosity))+geom_point()+lab
s(title = "GGPlot of Happiness Score with Generosity")

GGPlot of Happiness Score with Generosity



 $\label{eq:continuous_score} $$ ggplot(data = happy_2015, aes(y = Happiness.Score, x = Dystopia.Residual)) + geom_point() + labs(title = "GGPlot of Happiness Score with Dystopia") \\$

GGPlot of Happiness Score with Dystopia

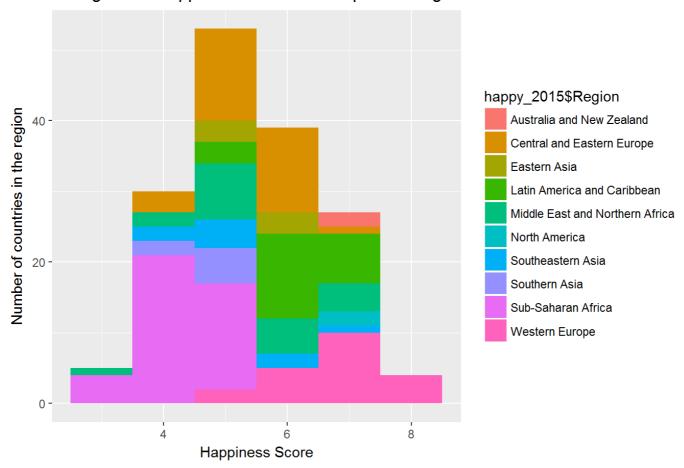


Inference: There is a strong positive increase in trend between features like GDP growth, Life expectancy, Freedom and Family with Happiness score.

Data Visualization with ggplots and geom histogram

ggplot(melt(as.numeric(happy_2015\$Happiness.Score)),aes(x=as.numeric(happy_2015\$Hap
piness.Score),fill= happy_2015\$Region)) + geom_histogram(position="stack",binwidth
= 1) + labs(title = "Histogram of Happiness score with respect to Region") + xlab("
Happiness Score") +ylab("Number of countries in the region")

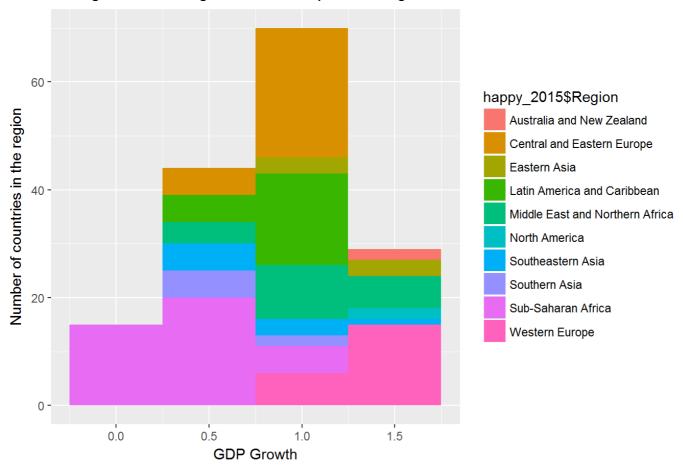
Histogram of Happiness score with respect to Region



Happiness score is more in Western Europe, Australia and North America and less in Middle East and Sub Saharan Africa

ggplot(melt(as.numeric(happy_2015\$Economy..GDP.per.Capita.)), aes(x=as.numeric(happy_2015\$Economy..GDP.per.Capita.), fill= happy_2015\$Region)) + geom_histogram(position = "stack", binwidth = 0.5) + labs(title = "Histogram of GDP growth with respect to Re gion") + xlab("GDP Growth") +ylab("Number of countries in the region")

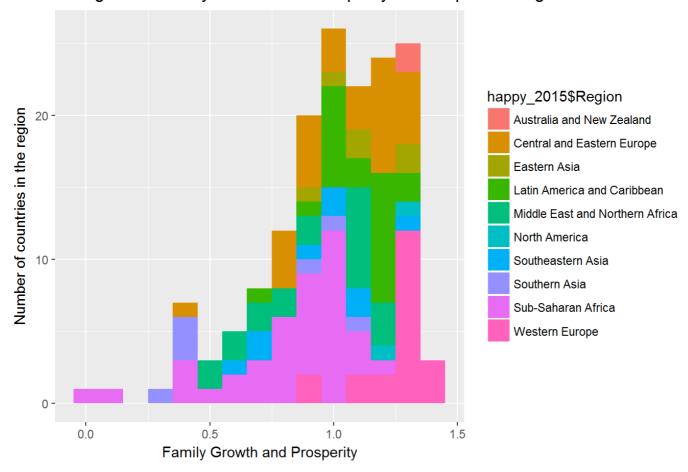
Histogram of GDP growth with respect to Region



GDP Growth is more in Western Europe and North America, Australia and NewZealand and less in some parts of Latin America, southern Asia and very less in Sub-Saharan Africa

 $\label{eq:continuous_series} $$ $\operatorname{ggplot}\left(\operatorname{melt}\left(\operatorname{as.numeric}\left(\operatorname{happy}_{2015\$Family}\right)\right), \operatorname{aes}\left(\operatorname{x=as.numeric}\left(\operatorname{happy}_{2015\$Family}\right), \operatorname{fill} \right) \\ = \operatorname{happy}_{2015\$Region}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right) + \operatorname{labs}\left(\operatorname{titl} \right) \\ = \operatorname{"Histogram}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right) + \operatorname{labs}\left(\operatorname{titl} \right) \\ = \operatorname{"Histogram}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right) + \operatorname{labs}\left(\operatorname{Tamlor}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right)\right) \\ = \operatorname{"Histogram}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right) + \operatorname{labs}\left(\operatorname{Tamlor}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right)\right) \\ = \operatorname{"Histogram}\left(\operatorname{position="stack"}, \operatorname{binwidth} = 0.1\right)\right)$

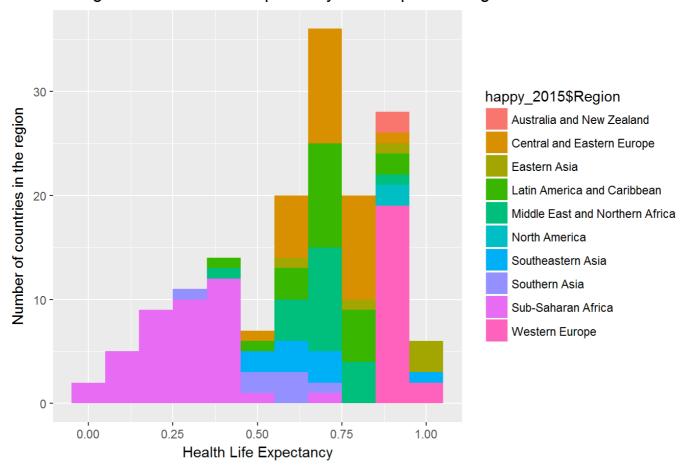
Histogram of Family Growth and Prosperity with respect to Region



Family growth and prosperity is seen most in Western Europe, North America, some parts of SouthEastern Asia and Australia and NewZealand followed by major parts in Latin America, Middle East and Central Europe.

ggplot(melt(as.numeric(happy_2015\$Health..Life.Expectancy.)),aes(x=as.numeric(happy_2015\$Health..Life.Expectancy.),fill= happy_2015\$Region)) + geom_histogram(position = "stack",binwidth = 0.1) + labs(title = "Histogram of Health Life Expectancy with r espect to Region") + xlab("Health Life Expectancy") +ylab("Number of countries in t he region")

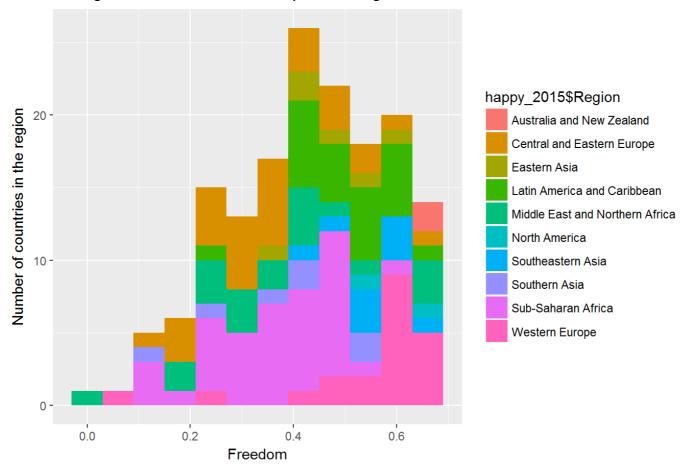
Histogram of Health Life Expectancy with respect to Region



Health Life Expectancy is the best in Western Europe, some parts of Eastern Asia, Australia and North America followed by Central and Eastern Europe. The least healt h expectancy rate lies in Sub-Sharan Africa region

 $\label{eq:continuous_section} $$ $\operatorname{ggplot}(\operatorname{melt}(\operatorname{as.numeric}(\operatorname{happy}_2015\$\operatorname{Freedom})), \operatorname{aes}(x=\operatorname{as.numeric}(\operatorname{happy}_2015\$\operatorname{Freedom}), fill= \operatorname{happy}_2015\$\operatorname{Region})) + \operatorname{geom_histogram}(\operatorname{position="stack"}, \operatorname{binwidth} = 0.06) + \operatorname{labs}(\operatorname{title} = "\operatorname{Histogram} \text{ of Freedom with respect to Region"}) + \operatorname{xlab}("\operatorname{Freedom"}) + \operatorname{ylab}("\operatorname{Number of countries in the region"})$

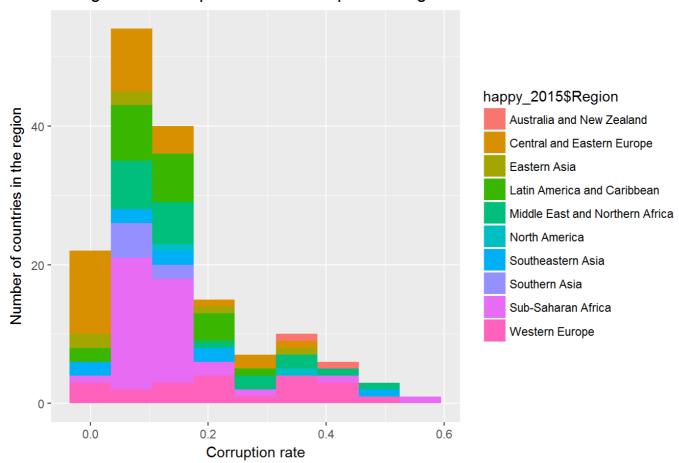
Histogram of Freedom with respect to Region



#Freedom is more prevalent in Western Europe, North America, South Eastern Asia and Australia followed by major parts of Latin America. The countries which does not contribute towards freedom are Iraq and Greece.

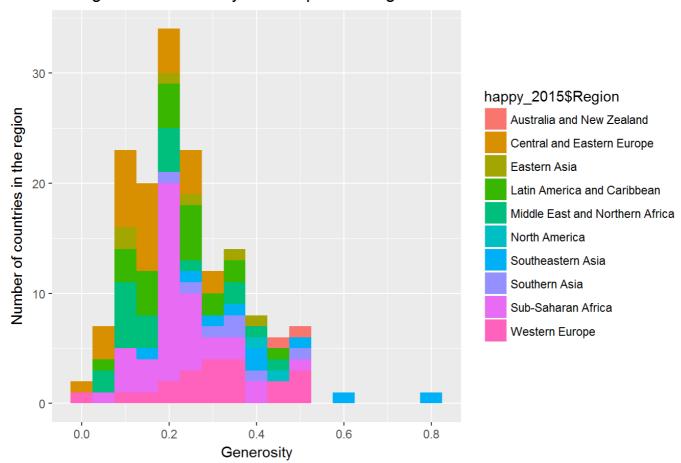
 $\label{lem:corruption} $$\operatorname{ggplot}(\operatorname{melt}(\operatorname{as.numeric}(\operatorname{happy}_2015\$\operatorname{Trust..Government}.\operatorname{Corruption.}))$, aes(x=as.numeric(\operatorname{happy}_2015\$\operatorname{Trust..Government}.\operatorname{Corruption.}), fill= \operatorname{happy}_2015\$\operatorname{Region})) + \operatorname{geom}_\operatorname{histogram}(\operatorname{position}="\operatorname{stack}",\operatorname{binwidth}=0.07) + \operatorname{labs}(\operatorname{title}="\operatorname{Histogram}(\operatorname{corruption}(\operatorname{rate}))) + \operatorname{modes}(\operatorname{corruption}(\operatorname{cor$

Histogram of Corruption rate with respect to Region



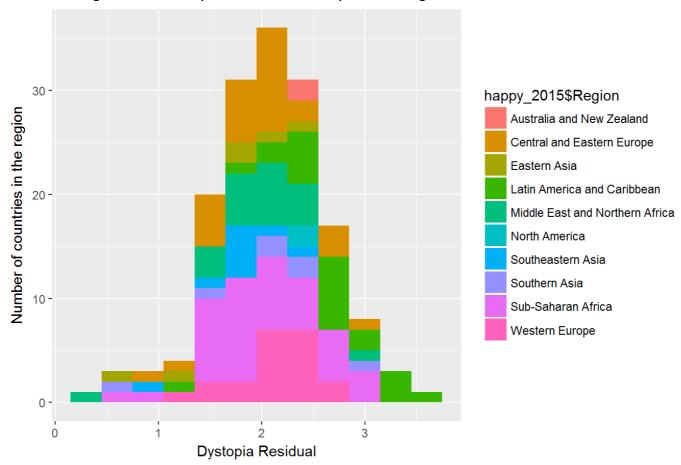
 $\label{lem:ggplot} $$ ggplot (melt(as.numeric(happy_2015\$Generosity)), aes(x=as.numeric(happy_2015\$Generosity), fill= happy_2015\$Region)) + geom_histogram(position="stack", binwidth = 0.05) + labs(title = "Histogram of Generosity with respect to Region") + xlab("Generosity") + ylab("Number of countries in the region") \\$

Histogram of Generosity with respect to Region



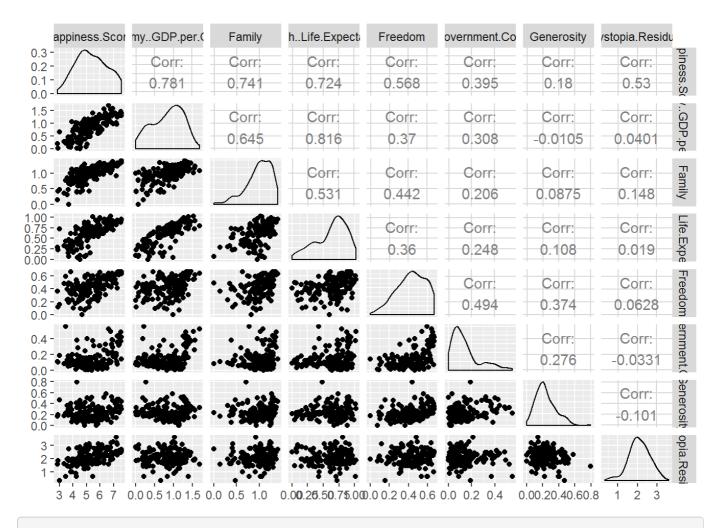
 $\label{eq:continuous_problem} $$ ggplot (melt(as.numeric(happy_2015$Dystopia.Residual)), aes(x=as.numeric(happy_2015$Dystopia.Residual)), fill= happy_2015$Region)) + geom_histogram(position="stack", binwidth= 0.3) + labs(title= "Histogram of Corruption rate with respect to Region") + xlab("Dystopia Residual") + ylab("Number of countries in the region") \\$

Histogram of Corruption rate with respect to Region



Data Visualization and Correlation

```
ggpairs(data = happy_2015, columns = c(4,6:12))
```



From the correlation, it is seen that Happiness score strongly depends on GDP, Fa mily and Life expectancy of which all the three are directly peopotional to the hap piness score. The weak correlations are Government corruption and Generosity.

Linear Regression Model for Happiness Score on 2015 dataset

```
happy_part_2015 <- happy_2015[c(-1,-2,-3,-5)]
lm_model <- lm(formula = Happiness.Score ~.,data = happy_part_2015)
summary(lm_model)</pre>
```

```
##
## Call:
## lm(formula = Happiness.Score ~ ., data = happy part 2015)
## Residuals:
       Min
                   1Q Median 3Q
## -5.701e-04 -2.224e-04 -2.580e-06 2.471e-04 5.054e-04
##
## Coefficients:
                           Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                             6.405e-05 1.245e-04 0.514 0.608
## Economy..GDP.per.Capita. 1.000e+00 1.129e-04 8855.750 <2e-16 ***
                             1.000e+00 1.153e-04 8675.863 <2e-16 ***
## Family
## Health..Life.Expectancy.
                            9.999e-01 1.619e-04 6175.103 <2e-16 ***
                              9.997e-01 1.976e-04 5059.468 <2e-16 ***
## Freedom
## Trust..Government.Corruption. 9.999e-01 2.236e-04 4470.866 <2e-16 ***
                              1.000e+00 2.018e-04 4956.272 <2e-16 ***
## Generosity
## Dystopia.Residual
                             1.000e+00 4.166e-05 24003.904 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.0002821 on 150 degrees of freedom
## Multiple R-squared: 1, Adjusted R-squared:
## F-statistic: 3.695e+08 on 7 and 150 DF, p-value: < 2.2e-16
```

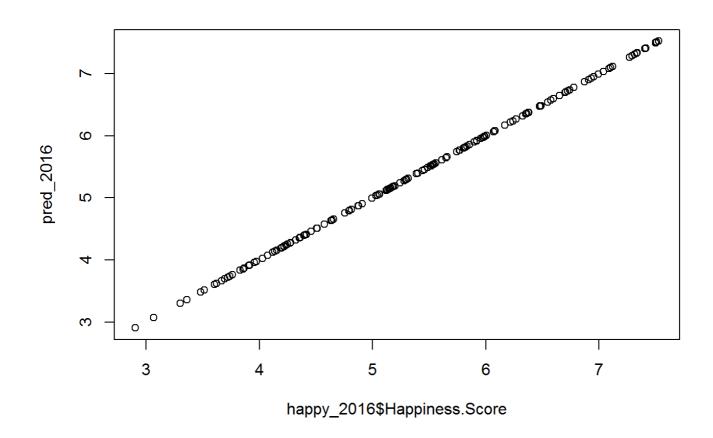
The summary model has Adjusted r-squared value of 1 which is 100% accurate model. All the predictors are significant with response variable (Happiness score).

Prediction of Linear model on 2016 dataset

```
happy_2016 <- read.csv('2016.csv')
pred_2016 <- predict(object = lm_model,newdata = happy_2016[7:13]) # Predict Happin
ess score for 2016 data
pred_2016</pre>
```

```
3
                                      5
## 7.525575 7.508631 7.501392 7.497571 7.413462 7.404020 7.339146 7.334414
             10
                   11
                          12 13 14
## 7.312625 7.291362 7.267043 7.118973 7.103724 7.086692 7.038771 6.994472
       17
              18
                      19 20 21
                                             22
                                                    23
## 6.952123 6.929438 6.906649 6.871171 6.778218 6.738688 6.725360 6.704849
              26
                   27
                            28 29
                                             30
## 6.701442 6.650174 6.596324 6.573326 6.544578 6.487913 6.480944 6.477614
               34
                      35
                              36
                                     37
                                             38
       33
## 6.473668 6.379070 6.379506 6.374535 6.361068 6.355004 6.323644 6.269306
       41
              42
                     43 44 45
                                            46
## 6.239316 6.218179 6.167775 6.084522 6.078573 6.068420 6.005092 5.991910
              50
                              52
                                     53
                                             54
##
       49
                      51
                                                    55
## 5.987019 5.977583 5.976410 5.955672 5.921028 5.918656 5.897355 5.856056
              58
                  59
                          60
                                61
                                            62
## 5.834972 5.835176 5.822115 5.813147 5.802362 5.770523 5.767677 5.742930
              66
                      67
                             68
                                    69
                                            70
## 5.657745 5.647805 5.615434 5.559620 5.545782 5.537958 5.528442 5.516632
                 75
                             76 77 78
              74
                                                   79
## 5.509925 5.487965 5.458010 5.440316 5.401526 5.389520 5.313593 5.303273
      81
              82
                     83 84 85
                                           86
                                                   87
## 5.291542 5.278960 5.245241 5.195607 5.184684 5.176889 5.163341 5.161080
             90
                 91 92 93
                                           94
                                                   95
       89
## 5.154916 5.151255 5.145375 5.132439 5.129505 5.122898 5.120786 5.060693
              98
                     99 100 101 102
                                                   103
## 5.057305 5.045466 5.032928 4.995744 4.906892 4.876084 4.875440 4.870639
             106 107 108 109 110
                                                   111
## 4.813262 4.795360 4.792725 4.753781 4.655258 4.643129 4.634739 4.575006
             114
                    115
                            116
                                   117
                                           118
## 4.573994 4.512629 4.508035 4.458650 4.415440 4.404439 4.395321 4.361737
             122
                    123 124 125
                                           126
                                                   127
## 4.359637 4.355713 4.324521 4.276001 4.272030 4.252344 4.235733 4.218886
                                  133 134
             130
                  131
                            132
                                                   135
## 4.216628 4.201317 4.192591 4.155947 4.138806 4.121544 4.073173 4.027851
              138
                    139
                            140
                                   141 142
      137
                                                   143
## 3.973882 3.955697 3.915714 3.906973 3.866597 3.856156 3.832092 3.762912
    145
             146
                  147
                            148 149 150
                                                  151
## 3.739223 3.739354 3.723644 3.694896 3.665787 3.621743 3.607285 3.514969
##
      153
              154
                     155
                             156
                                    157
## 3.483686 3.359868 3.302584 3.069181 2.904631
```

plot(happy_2016\$Happiness.Score,pred_2016) # Compares the actual 2016 data with the
predicted data



#Conclusion: The Multiple linear regression model works perfect on 2015 data and is able to predict the happiness score of 2016 data.

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