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R Markdown

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see http://rmarkdown.rstudio.com.

When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
df1 = read.csv('C:\\Users\\guesm\\Documents\\Methods of Data Science\\2020.csv')
df2 = read.csv('C:\\Users\\guesm\\Documents\\Methods of Data Science\\2021.csv')

##Data Cleaning
renaming the data columns for both data sets

names(df1)<- c('Country.name', 'Regional.indicator', 'Ladder.score', 'Standard.error.of.ladder.score',
names(df1)</pre>
```

```
[1] "Country.name"
##
   [2] "Regional.indicator"
   [3] "Ladder.score"
##
    [4] "Standard.error.of.ladder.score"
   [5] "upperwhisker"
##
##
   [6] "lowerwhisker"
   [7] "Logged.GDP.Per.capita"
##
##
   [8] "Social.support"
##
   [9] "Healthy.life.expectancy"
## [10] "Freedom.to.make.life.choices"
## [11] "Generosity"
## [12] "Perceptions.of.corruption"
## [13] "Ladder.score.in.dystopia"
## [14] "Explained.by.log.GDP.Per.Capita"
## [15] "Explained.by.social.support"
## [16] "Explained.by.healthy.life.expectancy"
## [17] "Explained.by.freedom.to.make.choices"
## [18] "Explained.by.Generosity"
## [19] "Explained.by.perceptions.of.corruptions"
## [20] "Dystopia.residual"
names(df2)<- c('Country.name', 'Regional.indicator', 'Ladder.score', 'Standard.error.of.ladder.score',</pre>
names(df2)
```

```
## [1] "Country.name"
## [2] "Regional.indicator"
## [3] "Ladder.score"
```

```
[4] "Standard.error.of.ladder.score"
## [5] "upperwhisker"
## [6] "lowerwhisker"
## [7] "Logged.GDP.Per.capita"
   [8] "Social.support"
## [9] "Healthy.life.expectancy"
## [10] "Freedom.to.make.life.choices"
## [11] "Generosity"
## [12] "Perceptions.of.corruption"
## [13] "Ladder.score.in.dystopia"
## [14] "Explained.by.log.GDP.Per.Capita"
## [15] "Explained.by.social.support"
## [16] "Explained.by.healthy.life.expectancy"
## [17] "Explained.by.freedom.to.make.choices"
## [18] "Explained.by.Generosity"
## [19] "Explained.by.perceptions.of.corruptions"
## [20] "Dystopia.residual"
```

deleting unnesscary columns for both data sets. removing the same columns from each dataset. This is for data base number 1.

```
df1<-subset(df1, select = -c(Standard.error.of.ladder.score))
df1 <- subset(df1, select = -c(upperwhisker))
df1 <- subset(df1, select = -c(lowerwhisker))
df1 <- subset(df1, select = -c(Ladder.score.in.dystopia))
df1 <- subset(df1, select = -c(Explained.by.log.GDP.Per.Capita))
df1 <- subset(df1, select = -c(Explained.by.social.support))
df1 <- subset(df1, select = -c(Explained.by.healthy.life.expectancy))
df1 <- subset(df1, select = -c(Explained.by.freedom.to.make.choices))
df1 <- subset(df1, select = -c(Explained.by.Generosity))
df1 <- subset(df1, select = -c(Explained.by.perceptions.of.corruptions))
df1 <- subset(df1, select = -c(Dystopia.residual))
df1 <- subset(df1, select = -c(Country.name))
df1 <- subset(df1, select = -c(Regional.indicator))</pre>
```

This code block deletes columns we do not need from column from dataset 2.

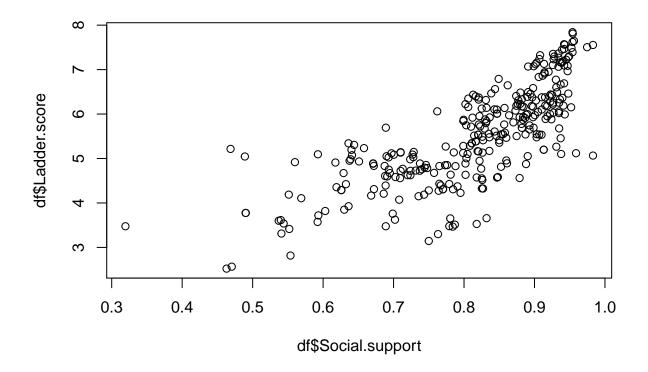
```
df2<-subset(df2, select = -c(Standard.error.of.ladder.score))
df2 <- subset(df2, select = -c(upperwhisker))</pre>
```

```
df2 <- subset(df2, select = -c(lowerwhisker))
df2 <- subset(df2, select = -c(Ladder.score.in.dystopia))
df2 <- subset(df2, select = -c(Explained.by.log.GDP.Per.Capita))
df2 <- subset(df2, select = -c(Explained.by.social.support))
df2 <- subset(df2, select = -c(Explained.by.healthy.life.expectancy))
df2 <- subset(df2, select = -c(Explained.by.freedom.to.make.choices))
df2 <- subset(df2, select = -c(Explained.by.Generosity))
df2 <- subset(df2, select = -c(Explained.by.perceptions.of.corruptions))
df2 <- subset(df2, select = -c(Explained.by.perceptions.of.corruptions))
df2 <- subset(df2, select = -c(Dystopia.residual))
df2 <- subset(df2, select = -c(Country.name))</pre>
```

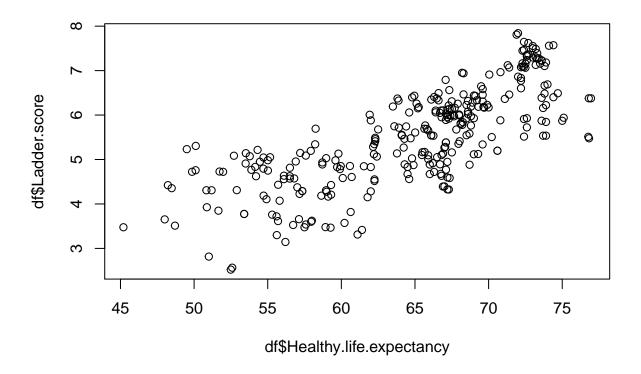
df = rbind(df1,df2)

EDA

```
plot(x = df$Social.support, y = df$Ladder.score)
```



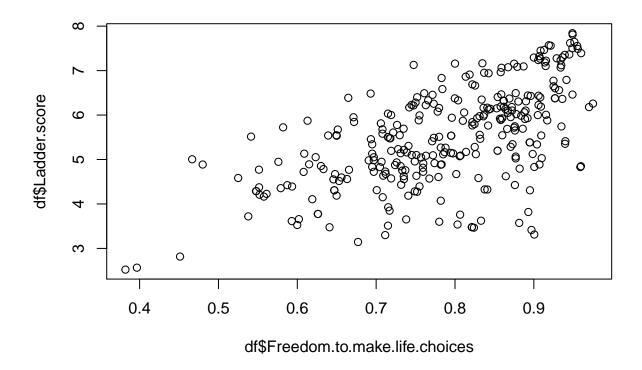
plot(x = df\$Healthy.life.expectancy, y = df\$Ladder.score)



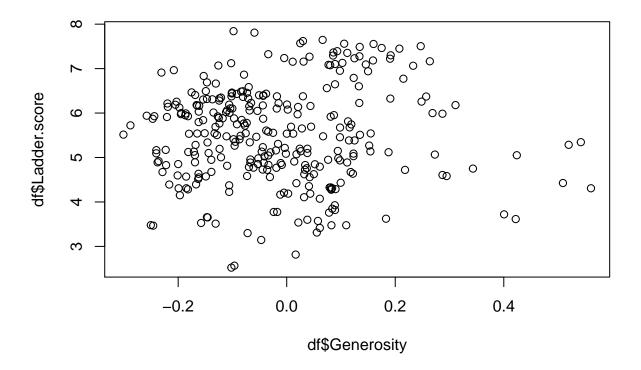
plot(x = df\$Logged.GDP.Per.capita, y = df\$Ladder.score)



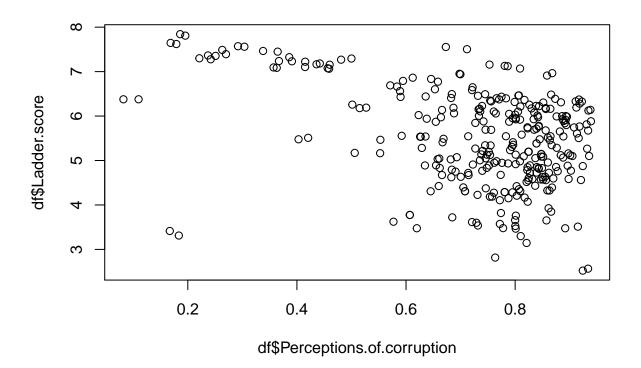
plot(x = df\$Freedom.to.make.life.choices, y = df\$Ladder.score)



plot(x = df\$Generosity, y = df\$Ladder.score)



plot(x = df\$Perceptions.of.corruption, y = df\$Ladder.score)



Multiple Linear Regression

```
df$Ladder.score = as.numeric(df$Ladder.score)
str(df)
  'data.frame':
                                 7 variables:
##
                    302 obs. of
   $ Ladder.score
                                         7.81 7.65 7.56 7.5 7.49 ...
                                  : num
   $ Logged.GDP.Per.capita
                                         10.6 10.8 11 10.8 11.1 ...
                                  : num
   $ Social.support
                                         0.954 0.956 0.943 0.975 0.952 ...
##
                                  : num
##
  $ Healthy.life.expectancy
                                         71.9 72.4 74.1 73 73.2 ...
                                  : num
   $ Freedom.to.make.life.choices: num
                                        0.949 0.951 0.921 0.949 0.956 ...
                                         -0.0595 0.0662 0.1059 0.2469 0.1345 ...
   $ Generosity
                                  : num
   $ Perceptions.of.corruption
                                  : num 0.195 0.168 0.304 0.712 0.263 ...
library(caTools)
set.seed(123)
split = sample.split(df$Ladder.score, SplitRatio = 0.8)
training_set = subset(df, split == TRUE)
test_set = subset(df, split == FALSE)
```

making the multiple regression model

```
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support + Healthy.life.expectance data = training_set)
```

summary(regressor)

```
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support +
##
      Healthy.life.expectancy + Freedom.to.make.life.choices +
##
      Generosity + Perceptions.of.corruption, data = training_set)
##
## Residuals:
##
       Min
                1Q
                    Median
                                3Q
                                        Max
## -1.73710 -0.29520 0.08116 0.34831 1.43395
##
## Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                             ## Logged.GDP.Per.capita
                             0.20939
                                        0.06747
                                                 3.103 0.002150 **
## Social.support
                              2.51410
                                       0.51985 4.836 2.40e-06 ***
## Healthy.life.expectancy
                              0.04009
                                        0.01049
                                                 3.821 0.000170 ***
## Freedom.to.make.life.choices 1.63880
                                        0.40631
                                                4.033 7.45e-05 ***
## Generosity
                              0.42141
                                        0.25873
                                                 1.629 0.104711
                             ## Perceptions.of.corruption
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5561 on 234 degrees of freedom
## Multiple R-squared: 0.7471, Adjusted R-squared: 0.7406
## F-statistic: 115.2 on 6 and 234 DF, p-value: < 2.2e-16
```

Beginning backwards elimination. this step removing Freedom.to.make.life.choices because it has the highest p value, and is larger than a sig level of .5

```
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support + Healthy.life.expectance data = training_set)
```

summary(regressor)

```
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support +
##
       Healthy.life.expectancy + Generosity + Perceptions.of.corruption,
       data = training_set)
##
##
## Residuals:
##
       Min
                 1Q
                      Median
                                   3Q
                                           Max
  -1.55038 -0.32440 0.06997 0.35345 1.58615
##
## Coefficients:
##
                            Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                            -1.00690 0.47718 -2.110 0.03591 *
## Logged.GDP.Per.capita
                             0.21011
                                        0.06963
                                                  3.018 0.00283 **
## Social.support
                                        0.51818 5.899 1.27e-08 ***
                             3.05684
                                        0.01076 4.164 4.40e-05 ***
## Healthy.life.expectancy
                             0.04480
```

```
## Generosity
                              0.71258
                                        0.25640 2.779 0.00589 **
                                      0.24319 -4.659 5.33e-06 ***
## Perceptions.of.corruption -1.13298
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.5739 on 235 degrees of freedom
## Multiple R-squared: 0.7295, Adjusted R-squared: 0.7238
## F-statistic: 126.8 on 5 and 235 DF, p-value: < 2.2e-16
more elimination. Taking away the variable perceptions of corruption, because it is now has the highest p
value, and it is above a .5 sig level.
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support + Healthy.life.expectanc
               data = training_set)
summary(regressor)
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support +
       Healthy.life.expectancy + Generosity, data = training_set)
##
##
## Residuals:
                 1Q
                     Median
## -1.52552 -0.37844 0.03287 0.42871 1.63594
##
## Coefficients:
                          Estimate Std. Error t value Pr(>|t|)
                                      0.36332 -6.953 3.48e-11 ***
## (Intercept)
                          -2.52613
## Logged.GDP.Per.capita
                           0.25220
                                      0.07201 3.502 0.000552 ***
## Social.support
                                       0.53647 5.154 5.40e-07 ***
                           2.76481
## Healthy.life.expectancy 0.05300
                                       0.01107 4.788 2.98e-06 ***
                            1.04602
                                       0.25678 4.074 6.32e-05 ***
## Generosity
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.5985 on 236 degrees of freedom
## Multiple R-squared: 0.7045, Adjusted R-squared: 0.6995
## F-statistic: 140.7 on 4 and 236 DF, p-value: < 2.2e-16
more elimination. taking away generosity because it now has the highest p value and is greater than sig
value of .5
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support + Healthy.life.expectance
               data = training_set)
summary(regressor)
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support +
```

```
##
       Healthy.life.expectancy + Perceptions.of.corruption, data = training_set)
##
## Residuals:
       Min
                      Median
                                    3Q
##
                  1Q
                                            Max
## -1.69451 -0.36611 0.08114 0.34306 1.58812
##
## Coefficients:
##
                             Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                         0.46428 -1.363
                             -0.63288
                                                           0.1741
## Logged.GDP.Per.capita
                                                           0.0106 *
                              0.17952
                                         0.06973
                                                   2.575
## Social.support
                              3.23877
                                         0.52130
                                                   6.213 2.33e-09 ***
## Healthy.life.expectancy
                                         0.01089
                                                   3.961 9.89e-05 ***
                              0.04315
## Perceptions.of.corruption -1.32165
                                         0.23683 -5.581 6.56e-08 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 0.582 on 236 degrees of freedom
## Multiple R-squared: 0.7206, Adjusted R-squared: 0.7159
## F-statistic: 152.2 on 4 and 236 DF, p-value: < 2.2e-16
more elimination, getting rid of Healthy, life, expectancy for the same reasons as before
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support + Perceptions.of.corrupt
               data = training_set)
summary(regressor)
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support +
       Perceptions.of.corruption, data = training_set)
##
## Residuals:
##
        Min
                  1Q
                      Median
                                    3Q
                                            Max
## -1.97951 -0.41013 0.09092 0.39426 1.59484
## Coefficients:
                             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                              0.23507
                                         0.42182
                                                   0.557
                                                             0.578
## Logged.GDP.Per.capita
                                         0.05476
                                                   6.543 3.68e-10 ***
                              0.35833
## Social.support
                              3.67484
                                         0.52510
                                                   6.998 2.64e-11 ***
## Perceptions.of.corruption -1.46675
                                         0.24112 -6.083 4.69e-09 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 0.5997 on 237 degrees of freedom
## Multiple R-squared: 0.7021, Adjusted R-squared: 0.6983
## F-statistic: 186.1 on 3 and 237 DF, p-value: < 2.2e-16
```

more elimination. getting rid of perceptions. of.corruption

```
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support,
              data = training_set)
summary(regressor)
##
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita + Social.support,
      data = training_set)
##
##
## Residuals:
##
       Min
                 1Q
                     Median
                                   3Q
## -2.11136 -0.46455 0.01637 0.48901 1.66947
##
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                        -1.48340
                                    0.33610 -4.414 1.54e-05 ***
                                              7.785 2.13e-13 ***
## Logged.GDP.Per.capita 0.44258
                                     0.05685
                         3.48528
                                    0.56242 6.197 2.52e-09 ***
## Social.support
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
\#\# Residual standard error: 0.6435 on 238 degrees of freedom
## Multiple R-squared: 0.6555, Adjusted R-squared: 0.6526
## F-statistic: 226.5 on 2 and 238 DF, \, p-value: < 2.2e-16
more elimination. getting rid of Social.support
regressor = lm(formula = Ladder.score ~ Logged.GDP.Per.capita,
              data = training_set)
summary(regressor)
##
## Call:
## lm(formula = Ladder.score ~ Logged.GDP.Per.capita, data = training_set)
##
## Residuals:
       Min
                 1Q Median
                                   3Q
## -2.32256 -0.49248 0.05879 0.55666 1.42085
##
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
                        -1.24108
                                    0.35899 -3.457 0.000646 ***
## (Intercept)
                                    0.03796 18.932 < 2e-16 ***
## Logged.GDP.Per.capita 0.71873
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.692 on 239 degrees of freedom
```

Multiple R-squared: 0.6, Adjusted R-squared: 0.5983
F-statistic: 358.4 on 1 and 239 DF, p-value: < 2.2e-16</pre>

predicting test set results.

```
y_pred = predict(regressor, newdata = test_set)
```