

MSDS 6306 401 Case Study 1: Exploration of Countries' GDPs Vs Income Groups

Yao Yao

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Introduction:

The following is a data exploration of GDPs and income groups for countries around the world. GDP data of countries is taken from <http://data.worldbank.org/data-catalog/GDP-ranking-table>, which was last updated on 01-Feb-2017. Income group of countries is taken from <http://data.worldbank.org/data-catalog/ed-stats>, which was last updated on 22-Feb-2017.

This case study is an exercise of gathering, cleaning, and analyzing data using R markdown to source .R files from various directories, and thus creating the paper file.

- Both data sets are stored in .csv format where headers are imported directly from the original

data set, where the columns that are not used to answer the questions in the analysis are then eliminated.

- Column name description:

1. CountryCode – The 3 letter country shortcode
2. Ranking – Country ranking by GDP with 1 being the highest
3. Economy – Country name
4. US Dollars (millions) – Gross Domestic Product of a certain country, in U.S. Dollars
5. Income.Group – The income group of a country

- Problems with the data:

1. The countries that have missing values for those columns listed above are not included in the analysis
2. The download file is updated regularly and may create different results later on

Folder Description:

Directions to run the code:

Install and load required packages as needed for the gathering and analysis procedures

```
InstallLoadMultiPackage <- function(pkg){  
  new.pkg <- pkg[!(pkg %in% installed.packages()[, "Package"])]  
  if (length(new.pkg))  
    install.packages(new.pkg, dependencies = TRUE)  
  sapply(pkg, require, character.only = TRUE)  
}  
InstallLoadMultiPackage(c("downloader", "ggplot2", "reshape2"))
```

```
## Loading required package: downloader
```

```
## Loading required package: ggplot2
```

```
## Loading required package: reshape2
```

```
## downloader    ggplot2    reshape2  
##             TRUE      TRUE      TRUE
```

==Gather GDP and Educational data==

Directory set up and confirmation of files and location

```
#setwd(".\\Data") # set as data directory  
getwd()
```

```
## [1] "C:/Users/Yao/Documents/GitHub/DDS-HW8/Data"
```

```
list.files()
```

```
## [1] "EducationalWeb.csv"  
## [2] "EducationRaw.csv"  
## [3] "GDPData.csv"  
## [4] "GDPRaw.csv"  
## [5] "GDPWeb.csv"  
## [6] "HINonOECD.csv"
```

```
## [7] "HIOECD.csv"
## [8] "LowerMiddleTop38.csv"
## [9] "MergeData1.csv"
## [10] "MergeData2.csv"
## [11] "NegGDP.csv"
## [12] "Quantiles.csv"
## [13] "Yao Yao MSDS 6306 401 Case Study1 Makefile.txt"
## [14] "Yao Yao MSDS 6306 401 Case Study1.R"
## [15] "Yao Yao MSDS 6306 401 Case Study1.Rmd"
## [16] "Yao_Yao_MSDS_6306_401_Case_Study1.pdf"
## [17] "Yao_Yao_MSDS_6306_401_Case_Study1.Rmd"
```

Download files via internet, read files into csv into respective destination files as needed

```
if (!file.exists("GDPWeb.csv")) {
  download('https://d396qusza40orc.cloudfront.net/getdata%2Fdata%2FGDP.csv',
    destfile='GDPWeb.csv')
}
if (!file.exists("EducationalWeb.csv")) {
  download('https://d396qusza40orc.cloudfront.net/getdata%2Fdata%2FEDSTATS_Country.csv',
    destfile='EducationalWeb.csv')
}
list.files() #make sure the files are there
```

```
## [1] "EducationalWeb.csv"
## [2] "EducationRaw.csv"
## [3] "GDPData.csv"
## [4] "GDPRaw.csv"
## [5] "GDPWeb.csv"
## [6] "HINonOECD.csv"
## [7] "HIOECD.csv"
## [8] "LowerMiddleTop38.csv"
## [9] "MergeData1.csv"
## [10] "MergeData2.csv"
## [11] "NegGDP.csv"
## [12] "Quantiles.csv"
## [13] "Yao Yao MSDS 6306 401 Case Study1 Makefile.txt"
## [14] "Yao Yao MSDS 6306 401 Case Study1.R"
## [15] "Yao Yao MSDS 6306 401 Case Study1.Rmd"
## [16] "Yao_Yao_MSDS_6306_401_Case_Study1.pdf"
## [17] "Yao_Yao_MSDS_6306_401_Case_Study1.Rmd"
```

==Clean GDP Data==

Import csv into raw GDP file and examine attributes

```
GDPRaw <- read.csv('GDPWeb.csv',stringsAsFactors = FALSE, header = TRUE)
str(GDPRaw)

## 'data.frame':    330 obs. of  10 variables:
## $ X : chr  "" "" "" "" ...
## $ Gross.domestic.product.2012: chr  "" "" "Ranking" "" ...
## $ X.1 : logi  NA NA NA NA NA NA ...
## $ X.2 : chr  "" "" "Economy" "" ...
```

```
## $ X.3 : chr "" "(millions of" "US dollars)" "" ...
## $ X.4 : chr "" "" "" "" ...
## $ X.5 : logi NA NA NA NA NA NA ...
## $ X.6 : logi NA NA NA NA NA NA ...
## $ X.7 : logi NA NA NA NA NA NA ...
## $ X.8 : logi NA NA NA NA NA NA ...
```

Eliminate header rows, rows without GDP, and unused columns. Reset column count and examine data

```
GDPData <- GDPRaw[5:194,1:5]
rownames(GDPData) <- seq(length=nrow(GDPData))
head(GDPData)
```

```
##      X Gross.domestic.product.2012 X.1      X.2      X.3
## 1 USA                          1 NA United States 16,244,600
## 2 CHN                          2 NA      China 8,227,103
## 3 JPN                          3 NA      Japan 5,959,718
## 4 DEU                          4 NA      Germany 3,428,131
## 5 FRA                          5 NA      France 2,612,878
## 6 GBR                          6 NA United Kingdom 2,471,784
```

```
tail(GDPData)
```

```
##      X Gross.domestic.product.2012 X.1      X.2      X.3
## 185 FSM                          185 NA Micronesia, Fed. Sts. 326
## 186 STP                          186 NA São Tomé and Príncipe 263
## 187 PLW                          187 NA      Palau 228
## 188 MHL                          188 NA      Marshall Islands 182
## 189 KIR                          189 NA      Kiribati 175
## 190 TUV                          190 NA      Tuvalu 40
```

Retitle columns and remove unused columns

```
colnames(GDPData) <- c("CountryCode", "Ranking", "x", "Economy", "US Dollars (millions)")
head(GDPData)
```

```
## CountryCode Ranking x      Economy US Dollars (millions)
## 1      USA      1 NA United States 16,244,600
## 2      CHN      2 NA      China 8,227,103
## 3      JPN      3 NA      Japan 5,959,718
## 4      DEU      4 NA      Germany 3,428,131
## 5      FRA      5 NA      France 2,612,878
## 6      GBR      6 NA United Kingdom 2,471,784
```

```
GDPData <- GDPData[,c("CountryCode", "Ranking", "Economy", "US Dollars (millions)")]
head(GDPData)
```

```
## CountryCode Ranking      Economy US Dollars (millions)
## 1      USA      1 United States 16,244,600
## 2      CHN      2      China 8,227,103
## 3      JPN      3      Japan 5,959,718
## 4      DEU      4      Germany 3,428,131
## 5      FRA      5      France 2,612,878
## 6      GBR      6 United Kingdom 2,471,784
```

Set the ranking as integer and GDP as numeric for later analysis, recheck attributes

```
GDPData$Ranking <- as.integer(GDPData$Ranking)
GDPData$`US Dollars (millions)` <- as.numeric(gsub(",", "", GDPData$`US Dollars (millions)`))
```

```
str(GDPData)
```

```
## 'data.frame':   190 obs. of  4 variables:
## $ CountryCode      : chr  "USA" "CHN" "JPN" "DEU" ...
## $ Ranking          : int   1 2 3 4 5 6 7 8 9 10 ...
## $ Economy           : chr  "United States" "China" "Japan" "Germany" ...
## $ US Dollars (millions): num  16244600 8227103 5959718 3428131 2612878 ...
```

```
dim(GDPData)
```

```
## [1] 190  4
```

==Clean Educational Data==

Import raw education data from csv and examine attributes

```
EducationRaw <- read.csv('EducationalWeb.csv',stringsAsFactors = FALSE, header = TRUE)
str(EducationRaw)
```

```
## 'data.frame':   234 obs. of  31 variables:
## $ CountryCode      : chr  "ABW" "ADO" "AFG" "AGO" ...
## $ Long.Name        : chr  "Aruba" "Principality of Andorra" "Islamic Republic of Afghanistan" ...
## $ Income.Group      : chr  "High income: nonOECD" "High income: nonOECD" ...
## $ Region           : chr  "Latin America & Caribbean" "Europe & Central Asia" ...
## $ Lending.category  : chr  "" "" "IDA" "IDA" ...
## $ Other.groups      : chr  "" "" "HIPC" "" ...
## $ Currency.Unit     : chr  "Aruban florin" "Euro" "Afghan afghani" ...
## $ Latest.population.census : chr  "2000" "Register based" "1979" "1970" ...
## $ Latest.household.survey : chr  "" "" "MICS, 2003" "MICS, 2001, MIS, 2006" ...
## $ Special.Notes     : chr  "" "" "Fiscal year end: March 20; reporting period: April 1-2006" ...
## $ National.accounts.base.year : chr  "1995" "" "2002/2003" "1997" ...
## $ National.accounts.reference.year : int   NA NA NA NA 1996 NA NA 1996 NA NA ...
## $ System.of.National.Accounts : int   NA NA NA NA 1993 NA 1993 1993 NA NA ...
## $ SNA.price.valuation : chr  "" "" "VAB" "VAP" ...
## $ Alternative.conversion.factor : chr  "" "" "" "1991-96" ...
## $ PPP.survey.year    : int   NA NA NA 2005 2005 NA 2005 2005 NA NA ...
## $ Balance.of.Payments.Manual.in.use : chr  "" "" "" "BPM5" ...
## $ External.debt.Reporting.status : chr  "" "" "Actual" "Actual" ...
## $ System.of.trade    : chr  "Special" "General" "General" "Special" ...
## $ Government.Accounting.concept : chr  "" "" "Consolidated" "" ...
## $ IMF.data.dissemination.standard : chr  "" "" "GDDS" "GDDS" ...
## $ Source.of.most.recent.Income.and.expenditure.data : chr  "" "" "" "IHS, 2000" ...
## $ Vital.registration.complete : chr  "" "Yes" "" "" ...
## $ Latest.agricultural.census : chr  "" "" "" "1964-65" ...
## $ Latest.industrial.data : int   NA NA NA NA 2005 NA 2001 NA NA NA ...
## $ Latest.trade.data : int   2008 2006 2008 1991 2008 2008 2008 2008 NA ...
## $ Latest.water.withdrawal.data : int   NA NA 2000 2000 2000 2005 2000 2000 NA 1991 ...
## $ X2.alpha.code     : chr  "AW" "AD" "AF" "AO" ...
## $ WB.2.code         : chr  "AW" "AD" "AF" "AO" ...
## $ Table.Name        : chr  "Aruba" "Andorra" "Afghanistan" "Angola" ...
## $ Short.Name        : chr  "Aruba" "Andorra" "Afghanistan" "Angola"
```

```
dim(EducationRaw)
```

```
## [1] 234 31
```

==Merge Education and GDP data==

Merge all columns from raw GDP data and Education based by country code. Save into raw merge file and examine attributes.

```
MergeData1 <- merge(x = GDPData, y = EducationRaw, by = 'CountryCode', all=TRUE)
str(MergeData1)
```

```
## 'data.frame': 235 obs. of 34 variables:
## $ CountryCode : chr "ABW" "ADO" "AFG" "AGO" ...
## $ Ranking : int 161 NA 105 60 125 32 26 133 NA 172 ...
## $ Economy : chr "Aruba" NA "Afghanistan" "Angola" ...
## $ US Dollars (millions) : num 2584 NA 20497 114147 12648 ...
## $ Long.Name : chr "Aruba" "Principality of Andorra" "Islamic Republic of Afghanistan" ...
## $ Income.Group : chr "High income: nonOECD" "High income: nonOECD" ...
## $ Region : chr "Latin America & Caribbean" "Europe & Central Asia" ...
## $ Lending.category : chr "" "" "IDA" "IDA" ...
## $ Other.groups : chr "" "" "HIPC" "" ...
## $ Currency.Unit : chr "Aruban florin" "Euro" "Afghan afghani" ...
## $ Latest.population.census : chr "2000" "Register based" "1979" "1970" ...
## $ Latest.household.survey : chr "" "" "MICS, 2003" "MICS, 2001, MIS, 2006" ...
## $ Special.Notes : chr "" "" "Fiscal year end: March 20; reporting period: April 1-2006" ...
## $ National.accounts.base.year : chr "1995" "" "2002/2003" "1997" ...
## $ National.accounts.reference.year : int NA NA NA NA 1996 NA NA 1996 NA NA ...
## $ System.of.National.Accounts : int NA NA NA NA 1993 NA 1993 1993 NA NA ...
## $ SNA.price.valuation : chr "" "" "VAB" "VAP" ...
## $ Alternative.conversion.factor : chr "" "" "" "1991-96" ...
## $ PPP.survey.year : int NA NA NA 2005 2005 NA 2005 2005 NA NA ...
## $ Balance.of.Payments.Manual.in.use : chr "" "" "" "BPM5" ...
## $ External.debt.Reporting.status : chr "" "" "Actual" "Actual" ...
## $ System.of.trade : chr "Special" "General" "General" "Special" ...
## $ Government.Accounting.concept : chr "" "" "Consolidated" "" ...
## $ IMF.data.dissemination.standard : chr "" "" "GDDS" "GDDS" ...
## $ Source.of.most.recent.Income.and.expenditure.data : chr "" "" "" "IHS, 2000" ...
## $ Vital.registration.complete : chr "" "Yes" "" "" ...
## $ Latest.agricultural.census : chr "" "" "" "1964-65" ...
## $ Latest.industrial.data : int NA NA NA NA 2005 NA 2001 NA NA NA ...
## $ Latest.trade.data : int 2008 2006 2008 1991 2008 2008 2008 2008 NA ...
## $ Latest.water.withdrawal.data : int NA NA 2000 2000 2000 2005 2000 2000 NA 1995 ...
## $ X2.alpha.code : chr "AW" "AD" "AF" "AO" ...
## $ WB.2.code : chr "AW" "AD" "AF" "AO" ...
## $ Table.Name : chr "Aruba" "Andorra" "Afghanistan" "Angola" ...
## $ Short.Name : chr "Aruba" "Andorra" "Afghanistan" "Angola"
```

```
dim(MergeData1)
```

```
## [1] 235 34
```

Remove merged columns based on lack of value in GDP, country, GDP ranking, or income group. Examine attributes and reset column count based on country code

```
MergeData2<-MergeData1[rowSums(is.na(MergeData1[,2:5]))==FALSE,]
MergeData2<-MergeData2[,c("CountryCode", "Ranking", "Economy", "US Dollars (millions)",
                           "Income.Group")]
head(MergeData2)
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 1      ABW      161      Aruba      2584
## 3      AFG      105      Afghanistan      20497
## 4      AGO      60      Angola      114147
## 5      ALB      125      Albania      12648
## 6      ARE      32      United Arab Emirates      348595
## 7      ARG      26      Argentina      475502
##      Income.Group
## 1 High income: nonOECD
## 3      Low income
## 4 Lower middle income
## 5 Upper middle income
## 6 High income: nonOECD
## 7 Upper middle income
```

```
tail(MergeData2)
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 230      WSM      181      Samoa      684
## 231      YEM      90      Yemen, Rep.      35646
## 232      ZAF      28      South Africa      384313
## 233      ZAR      112      Congo, Dem. Rep.      17204
## 234      ZMB      104      Zambia      20678
## 235      ZWE      134      Zimbabwe      9802
##      Income.Group
## 230 Lower middle income
## 231 Lower middle income
## 232 Upper middle income
## 233      Low income
## 234      Low income
## 235      Low income
```

```
dim(MergeData2)
```

```
## [1] 189  5
```

```
rownames(MergeData2) <- seq(length=nrow(MergeData2))
```

Export data frames into csv file in the data directory to keep track of edits and merges

```
write.csv(MergeData1, "MergeData1.csv")
write.csv(MergeData2, "MergeData2.csv")
write.csv(GDPRaw, "GDPRaw.csv")
write.csv(GDPData, "GDPData.csv")
write.csv(EducationRaw, "EducationRaw.csv")
write.csv(GDPData, "GDPData.csv")
```

==Analysis to answer questions==

0) Include code to count the number of missing values for each variable used in the analysis:

Extract the number of rows from original merged raw data, Track the number of cumulative matched rows, number of N/A values in Rankings, Economies, GDP, Income groups, and cumulative unmatched rows

```

print(paste0("Total Number of Rows in Merged Data: ", nrow(MergeData1)))

## [1] "Total Number of Rows in Merged Data: 235"
NARanking<-sum(is.na(MergeData1$Ranking) == TRUE)
print(paste0("Number of N/A Rankings: ", NARanking))

## [1] "Number of N/A Rankings: 45"
NAEconomy<-sum(is.na(MergeData1$Economy) == TRUE)
print(paste0("Number of N/A Economies: ", NAEconomy))

## [1] "Number of N/A Economies: 45"
NAGDP<-sum(is.na(MergeData1$`US Dollars (millions)` == TRUE)
print(paste0("Number of N/A GDPs: ", NAGDP))

## [1] "Number of N/A GDPs: 45"
NAIncomeGroup<-sum(is.na(MergeData1$Income.Group) == TRUE)
print(paste0("Number of N/A Income Groups: ", NAIncomeGroup))

## [1] "Number of N/A Income Groups: 1"
NATotal<-sum(is.na(MergeData1$Ranking) == TRUE| is.na(MergeData1$Economy) == TRUE |
             MergeData1$`US Dollars (millions)` == TRUE |
             is.na(MergeData1$Income.Group) == TRUE)
print(paste0("Total Number of Unmatched Rows: ", NATotal))

## [1] "Total Number of Unmatched Rows: 46"

```

For each of the variables utilized in the analysis, there are 45 N/A values for rankings, economies, and GDP. There is one country with a N/A value for income groups and is also eliminated, which results in a total of 46 unmatched rows

1) Merge the data based on the country shortcode. How many of the IDs match?

```

print(paste0("Number of Rows in Merged Data without N/A values: ", nrow(MergeData2)))

## [1] "Number of Rows in Merged Data without N/A values: 189"

```

After merging the data by country shortcode and eliminating pertinent rows with N/A values, 189 countries has GDP, educational, and ranking values matching with income group.

2) Sort the data frame in ascending order by GDP (so United States is last). What is the 13th country in the resulting data frame?

Rank the merged data by ascending GDP ranking, examine attributes and export dataset

```

NegGDP <- MergeData2[order(MergeData2$`US Dollars (millions)`),]
head(NegGDP)

##      CountryCode Ranking      Economy US Dollars (millions)
## 173          TUV    190          Tuvalu                40
##  92          KIR    189          Kiribati               175
## 113          MHL    188 Marshall Islands              182
## 137          PLW    187          Palau                 228

```



```
## 155      STP      186 São Tomé and Príncipe      263
## 59       FSM      185 Micronesia, Fed. Sts.      326
##           Income.Group
## 173 Lower middle income
## 92  Lower middle income
## 113 Lower middle income
## 137 Upper middle income
## 155 Lower middle income
## 59  Lower middle income

str(NegGDP)

## 'data.frame':  189 obs. of  5 variables:
##  $ CountryCode      : chr  "TUV" "KIR" "MHL" "PLW" ...
##  $ Ranking           : int   190 189 188 187 186 185 184 183 182 181 ...
##  $ Economy           : chr  "Tuvalu" "Kiribati" "Marshall Islands" "Palau" ...
##  $ US Dollars (millions): num  40 175 182 228 263 326 472 480 596 684 ...
##  $ Income.Group      : chr  "Lower middle income" "Lower middle income" "Lower middle income" "Up

write.csv(NegGDP, "NegGDP.csv")
```

Code to find 13th country with the ascending GDP. More code to show that there is a tie between St. Kitts and Grenada at 12th place, which results the alphabetical order to dictate St. Kitts at 13th place in ranking.

```
country13NegGDP<-NegGDP[13,3]
print(paste0("The 13th country in ascending order by GDP is: ", country13NegGDP))
```

```
## [1] "The 13th country in ascending order by GDP is: St. Kitts and Nevis"
```

```
NegGDP[12:13,]
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 69      GRD      178      Grenada      767
## 93      KNA      178 St. Kitts and Nevis      767
##           Income.Group
## 69 Upper middle income
## 93 Upper middle income
```

From ascending GDP, country #13 is St. Kitts and Nevis in the resulting data frame NegGDP. Technically, St. Kitts and Grenada are tied at 12th in ascending GDP and further ascending alphabetical sorting makes St. Kitts appear at 13th and Grenada at 12th place.

3) What are the average GDP rankings for the “High income: OECD” and “High income: nonOECD” groups?

Assign a subset of High Income OECD countries from income group and find the mean of their GDP rank

```
HIOECD <- MergeData2[ which(MergeData2$Income.Group=='High income: OECD'), ]
head(HIOECD)
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 9      AUS      12      Australia      1532408
## 10     AUT      27      Austria      394708
## 13     BEL      25      Belgium      483262
## 31     CAN      11      Canada      1821424
## 32     CHE      20      Switzerland      631173
## 44     CZE      51      Czech Republic      196446
```

```
##           Income.Group
## 9  High income: OECD
## 10 High income: OECD
## 13 High income: OECD
## 31 High income: OECD
## 32 High income: OECD
## 44 High income: OECD

write.csv(HIOECD, "HIOECD.csv")
HAvgGDPRank<- mean(HIOECD$Ranking)
print(paste0("The average GDP ranking of high income, OECD countries is: ",
             round(HAvgGDPRank, digits = 2)))
```

```
## [1] "The average GDP ranking of high income, OECD countries is: 32.97"
```

Assign a subset of High Income nonOECD countries from income group and find the mean of their GDP rank

```
HINonOECD <- MergeData2[ which(MergeData2$Income.Group=='High income: nonOECD'), ]
head(HINonOECD)
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 1          ABW      161          Aruba             2584
## 5          ARE       32 United Arab Emirates         348595
## 18         BHR       93          Bahrain             29044
## 19         BHS      138      Bahamas, The             8149
## 23         BMU      149          Bermuda             5474
## 26         BRB      153          Barbados             4225
##           Income.Group
## 1  High income: nonOECD
## 5  High income: nonOECD
## 18 High income: nonOECD
## 19 High income: nonOECD
## 23 High income: nonOECD
## 26 High income: nonOECD
```

```
write.csv(HINonOECD, "HINonOECD.csv")
NAvgGDPRank<- mean(HINonOECD$Ranking)
print(paste0("The average GDP ranking of high income, nonOECD countries is: ",
             round(NAvgGDPRank, digits = 2)))
```

```
## [1] "The average GDP ranking of high income, nonOECD countries is: 91.91"
```

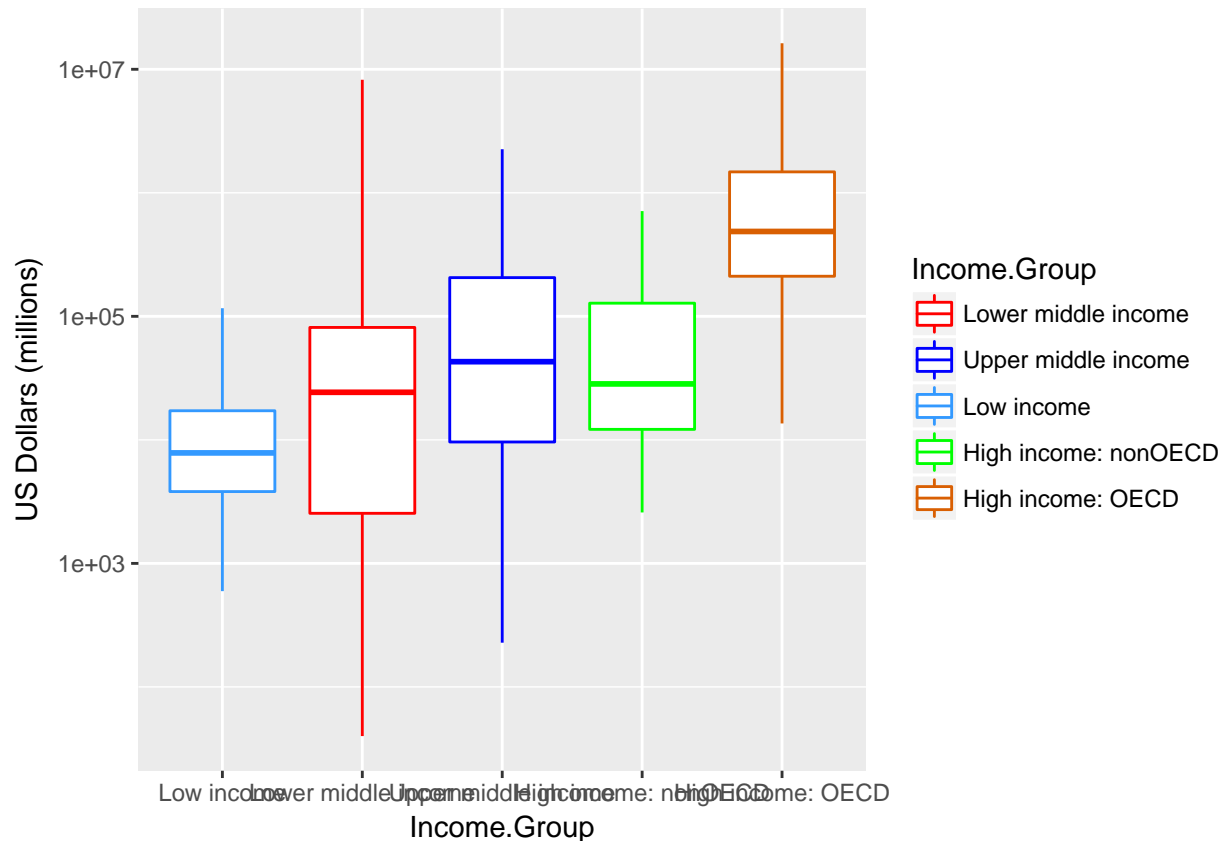
By income group, the average GDP rankings for High income: OECD countries is 32.97 and for High income: nonOECD countries is 91.91. High income OECD countries have higher GDP than that of High income nonOECD countries. Higher GDP ranking suggests that high income countries that are open to free world trade and development are more prosperous.

4) Show the distribution of GDP value for all the countries and color plots by income group. Use ggplot2 to create your plot.

Using ggplot2, individual countries with matching rows in GDP are logarithmically plotted by separately colored income group box plots to show quantile distribution.

```
NegGDP$Income.Group <- factor(NegGDP$Income.Group, levels=c("Low income",
                  "Lower middle income", "Upper middle income", "High income: nonOECD",
                  "High income: OECD"))
color.codes<-as.character(c("#3399FF", "#FF0000", "#0000FF", "#00FF00", "#D95F02"))
```

```
ggplot(data = NegGDP, aes(y = `US Dollars (millions)`, x = Income.Group,
  colour = Income.Group)) + geom_boxplot() + scale_y_log10() +
  scale_colour_manual(breaks = NegGDP$Income.Group, values =
    unique(as.character(color.codes)))
```



Graphically by boxplot log distribution, it was expected that the median GDP of countries grouped by income group rose from low income to lower middle income to upper middle income. For high income countries, there is a discrepancy between OECD and nonOECD countries. If the country is high income but does not allow free global trade and development, they have an median GDP lower than that of upper middle income countries and about equivalent to that of lower middle income countries. Otherwise, if the high income country is an OECD member, they continue the trend of GDP prosperity. In addition, the giant range of counties that fall into the lower middle income category suggests that the distinction of countries by income groups is not solely based on GDP qualities alone.

5) Provide summary statistics of GDP by income groups.

GDP summary statistics of countries based on income groups

```
tapply(NegGDP$`US Dollars (millions)`, NegGDP$Income.Group, summary)
```

```
## $`Low income`
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    596   3814   7843   14410   17200   116400
##
## $`Lower middle income`
##   Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
```

```
##      40      2549      24270      256700      81450      8227000
##
## $`Upper middle income`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      228      9613      42940      231800      205800      2253000
##
## $`High income: nonOECD`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      2584      12840      28370      104300      131200      711000
##
## $`High income: OECD`
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      13580      211100      486500      1484000      1480000      16240000
```

From the boxplot log distribution of countries' GDP separated by income groups, the quantile distributions were plotted by range, interquantile range, and medians. The summary statistics show that the mean GDP per income group is very different than that of the median, with the mean being 0.8x, 9.5x, 4.3x, 2.6x, and 2x greater than that of the median for their respective income groups by ascending classification.

The range overlap in country GDP further suggests that countries separated by income group was not solely based on GDP. The order of mean GDP by income group is low income, high income: nonOECD, upper middle income, lower middle income, and high income: OECD, which means that there are more factors that dictate how a country is classified into income groups than GDP alone.

6) Cut the GDP ranking into 5 separate quantile groups. Make a table versus Income.Group. How many countries are Lower middle income but among the 38 nations with highest GDP?

Breaks the GDP rankings into 5 separate quantile groups, with increment of 20%, and writes the quantiles into csv. Negdata is used because factors and levels are defined previously

```
Quantiles<-cut(NegGDP$Ranking, breaks=quantile(NegGDP$Ranking,seq(0, 1, 0.2)))
head(Quantiles)
```

```
## [1] (152,190] (152,190] (152,190] (152,190] (152,190] (152,190]
## Levels: (1,38.6] (38.6,76.2] (76.2,114] (114,152] (152,190]
```

```
write.csv(Quantiles, "Quantiles.csv")
```

Using reshape2, a table shows the number of contries per income group that falls inside their respective 20% quantile groups based on individual GDP ranking

```
table(MergeData2$Income.Group, Quantiles)
```

```
##              Quantiles
##              (1,38.6] (38.6,76.2] (76.2,114] (114,152] (152,190]
## High income: nonOECD      1          6          4          5          7
## High income: OECD        4          5          7          9          5
## Low income               8          8          6          8          6
## Lower middle income      16          9         12          8          9
## Upper middle income       8         10          8          8         11
```

It was expected that low income has more countries that fall inside the higher quantile GDP rankings while that of the higher income: OECD has more that fall inside the lower GDP quantile rankings. Lower middle income has a concentration of counties that fall inside ther higher GDP rankings with some of its countries in the lower quantile GDP rankings. Upper middle income countries has an even distribution of countries in each quantile category while that of high income: nonOECD countries have countries falling in the middle

GDP quantile rankings. GDP ranking 1 to 38 is the top 20% quantile of all the nations. There are 5 lower middle income countries among the 38 nations with the highest GDP.

Document which countries from lower middle income group has the top 38 GDP rankings

```
LowerMiddleTop38 <- NegGDP[which(NegGDP$Ranking <= 38 &
                                NegGDP$Income.Group == "Lower middle income"),]
```

```
LowerMiddleTop38
```

```
##      CountryCode Ranking      Economy US Dollars (millions)
## 51          EGY      38 Egypt, Arab Rep.          262832
## 165         THA      31      Thailand          365966
## 77          IDN      16      Indonesia          878043
## 78          IND      10        India          1841710
## 34          CHN       2        China          8227103
##              Income.Group
## 51 Lower middle income
## 165 Lower middle income
## 77 Lower middle income
## 78 Lower middle income
## 34 Lower middle income
```

```
write.csv(LowerMiddleTop38, "LowerMiddleTop38.csv")
```

Conclusion:

0 and 1) As the online data set updates to include more GDP and income groups, more of the world's countries would be included to do a full-world analysis. For now, the analysis is for 189 of the 235 available countries, with 46 countries with missing data.

- 2) If there is a tie in GDP rankings at #12 for Grenada and St. Kitts, further alphabetical sorting is used to distinguish St. Kitts as the 13th country in ascending GDP ranking.
- 3) The rankings gap between the average GDP ranking of high income, OECD countries (32.97) and that of high income, nonOECD countries (91.91) is quite significant, given that the range of GDP rankings is from 1 to 189. High income OECD countries that are open to free trade and development have a higher average GDP ranking than those nonOECD countries that do not.
- 4) When boxplot distributions are plotted for GDPs by income group, there is some upwards trend when comparing median GDPs for low income to lower middle income to upper middle income to high income OECD countries. As stated for number 3, high income nonOECD countries cripple their GDP by not having open trade to all countries for development and its median GDP fall close to that for lower middle income. There are non-GDP factors when categorizing certain countries by income group because of the wide GDP range that the lower middle income group constitutes.
- 5) The summary statistics show that the mean GDP per income group is very different than that of the median, with the mean being 0.8x, 9.5x, 4.3x, 2.6x, and 2x greater than that of the median for their respective income groups by ascending classification. As stated in number 4, there is quite a bit of GDP overlap when classifying certain countries to income groups and classification of income group is not solely based on GDP.
- 6) There are 5 lower middle income countries among the 38 nations with the highest GDP, which constitutes the top 20% quantile of all the nations analyzed. As stated in number 4) there are factors outside of GDP that qualify certain countries to certain income classifications.
 - The world data sets are observational and no causal effect could be inferenced. The country data sampled are not randomized for population inference and does not reflect data from all the nations in

the world.

- Writing functions in R makes the work reproducible for future analysis and R markdown is good for documenting all the steps.

Further Work:

Future work would be to analyze country GDP per capita or per land size to see if the GDP distributions per income group would change based on those incremental factors. It would also be good to know what constitutes a country to be categorized to a certain income group and see if any of the other columns imported from world data sets could indicate more trends based on column data from other factors.