

Assignment 2

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Assignment 2

Creating Dataframe and using it to plot the data

1. Create data frame with these two column vectors in R Studio $x = 1:30$ $y = x^3$

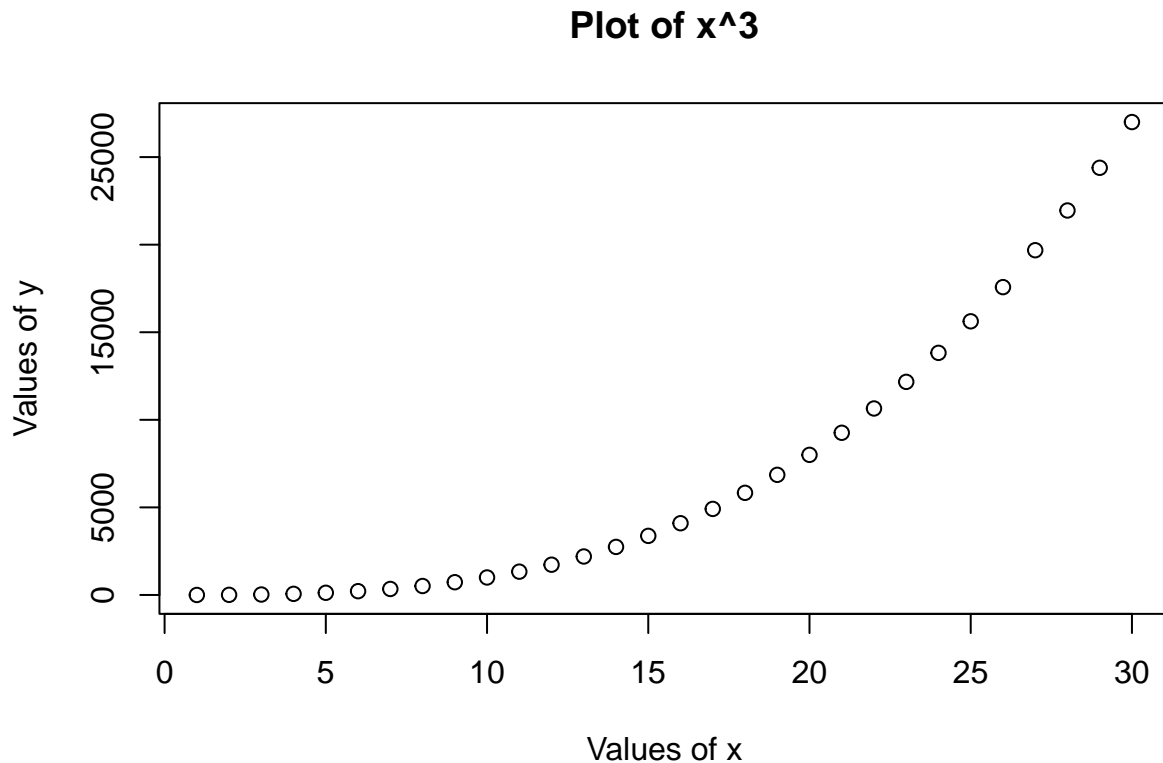
```
# Creating a dataframe
df<-data.frame(x<-c(1:30), y<-x^3)
# Giving names to the columns
colnames(df)<-c('x','y')
print(df)
```

```
##      x      y
## 1    1      1
## 2    2     8
## 3    3    27
## 4    4   64
## 5    5  125
## 6    6  216
## 7    7  343
## 8    8  512
## 9    9  729
## 10   10 1000
## 11   11 1331
## 12   12 1728
## 13   13 2197
## 14   14 2744
## 15   15 3375
## 16   16 4096
## 17   17 4913
## 18   18 5832
## 19   19 6859
## 20   20 8000
## 21   21 9261
## 22   22 10648
## 23   23 12167
## 24   24 13824
## 25   25 15625
## 26   26 17576
## 27   27 19683
```

```
## 28 28 21952
## 29 29 24389
## 30 30 27000
```

2. Create plot of x and y variables in R Studio and interpret it carefully

```
# Plotting the values of x and y
plot(df$x,df$y, main = "Plot of x^3", xlab = 'Values of x', ylab = 'Values of y')
```



In the plot above we can see that the value of y increases exponentially. From the graph, we can see that, as value of x grows a small change in value of x increases the value of y drastically.

3. Get appropriate correlation coefficient of this data in R Studio and interpret it carefully

Since the relationship between the variables is not linear we should not be using pearson's correlation coefficient rather we use spearman's correlation.

```
cor_val<-cor(df$x,df$y,method = "spearman")
print(cor_val)
```

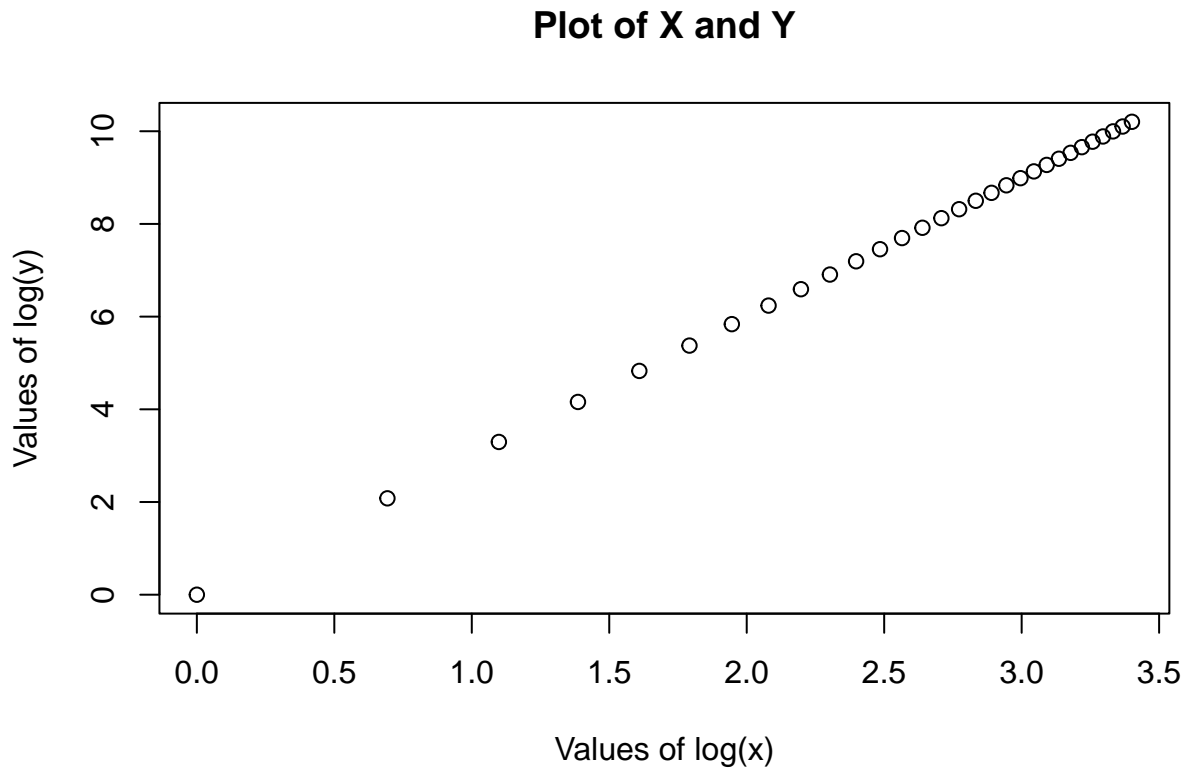
```
## [1] 1
```

This shows that there is perfect correlation between the variables.

Converting Non-Linear to Linear using Log

4. Transform the plot to linear using appropriate mathematical function in R Studio

```
# Using log function to transform the plot to linear  
df$a<-log(x)  
df$b<-log(y)  
plot(df$a,df$b,main = "Plot of X and Y", xlab = 'Values of log(x)', ylab = 'Values of log(y)')
```



5. Get appropriate correlation coefficient now in R Studio and interpret it carefully too

Since we have converted the values into linear we can use the pearson correlation coefficient.

```
cor_val_lin <-cor(df$a,df$b,method = "pearson")  
print(cor_val_lin)
```

```
## [1] 1
```

This shows that there is a perfect correlation between the two variables.

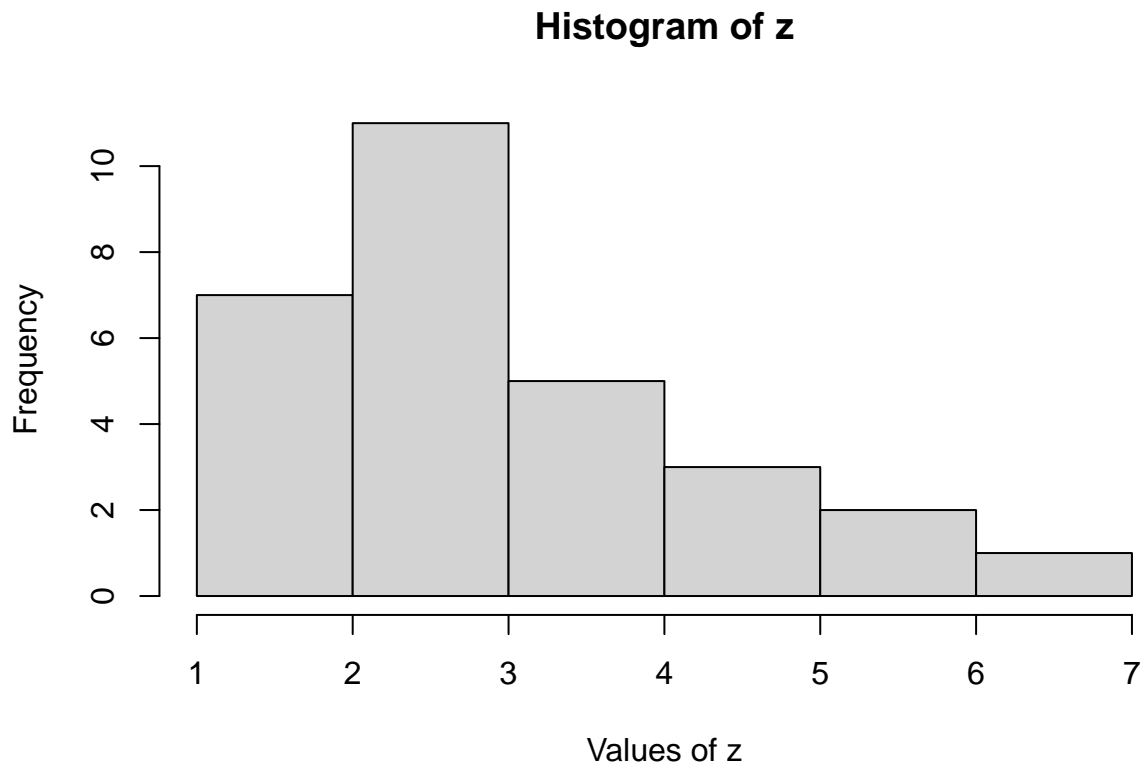
6. Create a new column vector z defined in the slide 18 of session two slide deck in R Studio

```
z<-c(1,1,2,2,2,2,2,3,3,3,3,3,3,3,3,3,3,3,4,4,4,4,4,5,5,5,6,6,7)
z
```

```
## [1] 1 1 2 2 2 2 2 3 3 3 3 3 3 3 3 3 3 3 4 4 4 4 4 5 5 5 6 6 7
```

7. Create a histogram of z variable in R Studio and interpret it carefully

```
hist(z,main="Histogram of z",xlab = "Values of z")
```



The histogram shows that the value 3 has highest frequency. It also shows a right skewed distribution. In case of skewed data median is the appropriate measure of central tendency.

8. Get summary statistics of z variable in R Studio and interpret it carefully

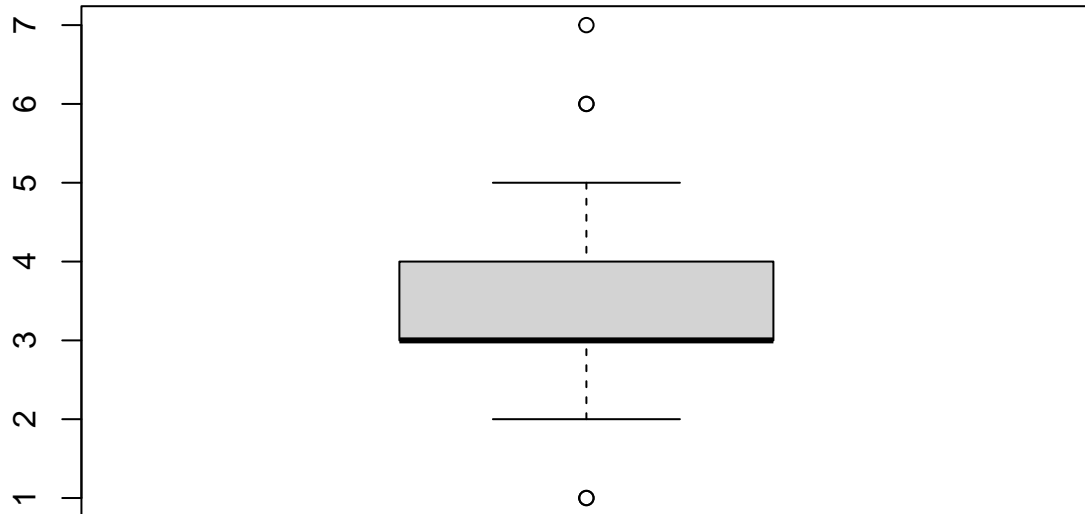
```
summary(z)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      1.000   3.000   3.000   3.414   4.000   7.000
```

The summary provides a quick glimpse at the data. It gives us mean, median, minimum and maximum values alongside q1 and q3.

9. Get box-plot of z variable in R Studio and interpret the result carefully.

```
boxplot(z)
```



The box plot above shows that the median of the data is 3. There are 3 outlier points. If a data point is greater than $Q3 + (1.5 * IQR)$ or less than $Q1 - (1.5 * IQR)$ then they are considered as outliers. In our case, $IQR = Q3 - Q1 = 4 - 3 = 1$. So, for data points less than 1.5 and greater than 5.5 are shown as outliers indicated by 'o' symbol.

10. Import "covnep_252days.csv" data in R Studio and describe the variables in it

```
file_path = "covnep_252days.csv"
data_csv = read.csv(file = file_path)
print(head(data_csv))
```

```
##      date totalCases newCases totalRecoveries newRecoveries totalDeaths
## 1 1/23/2020         1         1              0              0           0
## 2 1/24/2020         0         0              0              0           0
## 3 1/25/2020         0         0              0              0           0
## 4 1/26/2020         0         0              0              0           0
## 5 1/27/2020         0         0              0              0           0
## 6 1/28/2020         0         0              0              0           0
## newDeaths
## 1         0
## 2         0
## 3         0
## 4         0
```

```
## 5      0
## 6      0
```

```
names(data_csv)
```

```
## [1] "date"          "totalCases"    "newCases"      "totalRecoveries"
## [5] "newRecoveries" "totalDeaths"   "newDeaths"
```

```
summary(data_csv)
```

```
##      date          totalCases      newCases      totalRecoveries
## Length:252      Min.   :    0      Min.   :  0.0      Min.   :    0
## Class :character 1st Qu.:    2      1st Qu.:  0.0      1st Qu.:    2
## Mode  :character Median :  963      Median :  82.5     Median :   182
##                Mean  :13376      Mean  :  308.8     Mean   :  8380
##                3rd Qu.:19340      3rd Qu.: 463.2     3rd Qu.:13932
##                Max.   :77816      Max.   :2020.0     Max.   :56282
## newRecoveries    totalDeaths      newDeaths
## Min.   :    0.0      Min.   :  0.00      Min.   : 0.000
## 1st Qu.:    0.0      1st Qu.:  0.00      1st Qu.: 0.000
## Median :    3.5      Median :  6.00      Median : 0.000
## Mean   :   223.3      Mean   : 66.67      Mean   : 1.976
## 3rd Qu.:   197.2      3rd Qu.: 53.75      3rd Qu.: 2.000
## Max.   :  2287.0      Max.   :498.00      Max.   :16.000
```

There are seven variables in the csv file and summary of each of them is shown above.

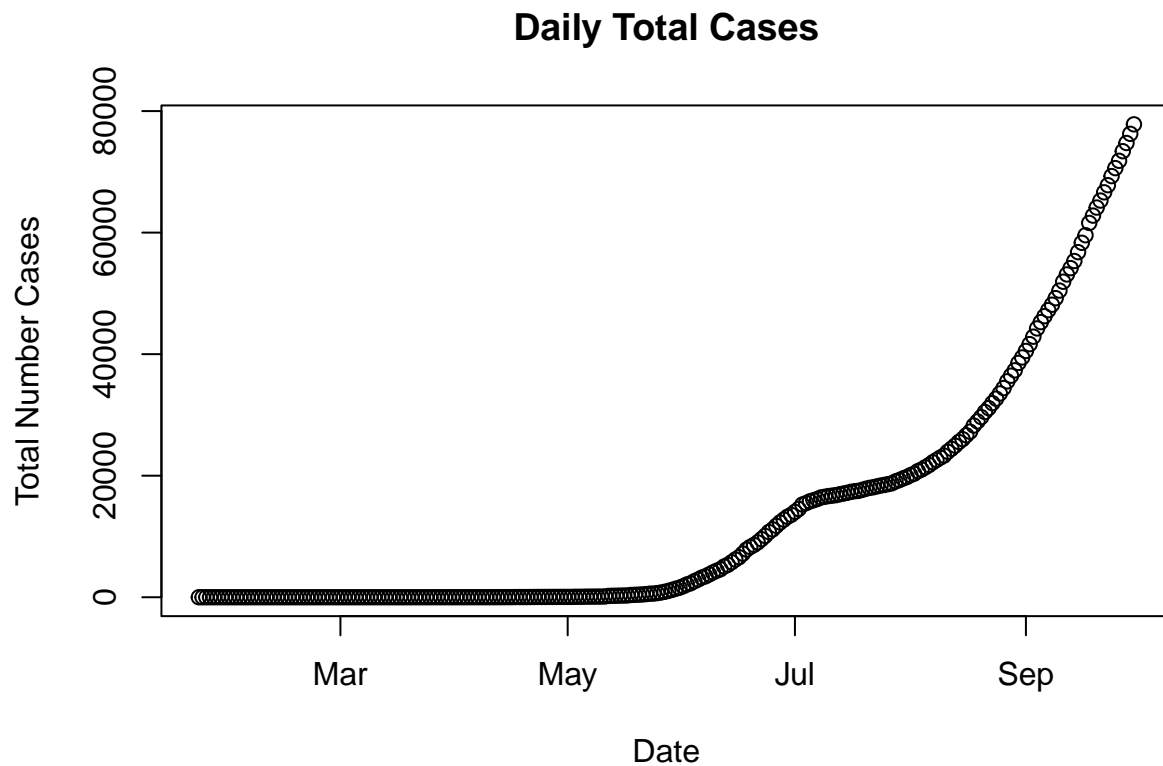
11. Create a chart with “totalCases” variable in y-axis and “date” variable in the x-axis in R Studio, describe the process leading to the creation of this chart

```
# Setting the data type of the date variable as date
data_csv$date<-as.Date(data_csv$date,format="%m/%d/%y")
```

```
head(data_csv)
```

```
##      date totalCases newCases totalRecoveries newRecoveries totalDeaths
## 1 2020-01-23         1         1             0             0             0
## 2 2020-01-24         0         0             0             0             0
## 3 2020-01-25         0         0             0             0             0
## 4 2020-01-26         0         0             0             0             0
## 5 2020-01-27         0         0             0             0             0
## 6 2020-01-28         0         0             0             0             0
## newDeaths
## 1         0
## 2         0
## 3         0
## 4         0
## 5         0
## 6         0
```

```
# The totalCases column is a cumulative value column. In the first row, the value is 1 and second row t
data_csv['totalCases'][data_csv['totalCases']==0]<-1
plot(data_csv$date,data_csv$totalCases,main='Daily Total Cases',xlab='Date',ylab ='Total Number Cases')
```

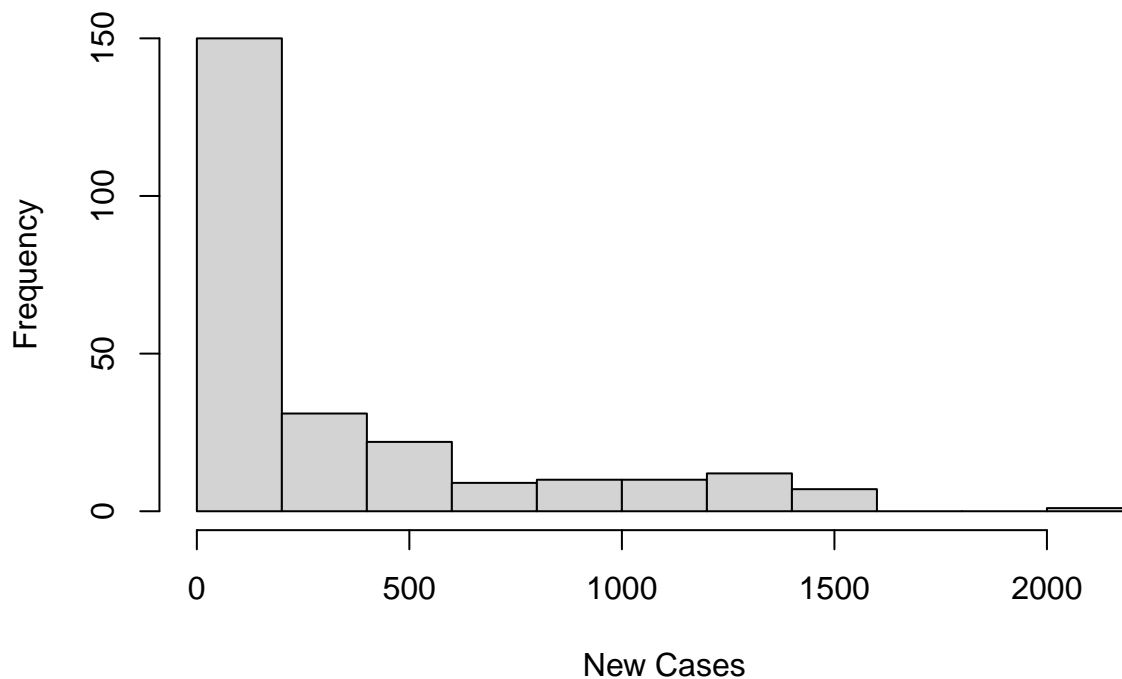


The steps leading upto the plot above is described below. 1. Reading the data into a dataframe using `read_csv` function. 2. Converted the `date` variable into appropriate date data type. 3. Replaced 0 with 1 in `totalCases` column since in the first row, the value is 1 and second row the value is 0 which is not mistake

12. Create histogram of “newCases” variable in R Studio and interpret it carefully

```
hist(data_csv$newCases, main = 'Histogram of newCases',xlab = 'New Cases')
```

Histogram of newCases



The histogram above shows highly skewed data. The distribution of data is right skewed.

13. Get summary statistics of “newCases” variable in R Studio and interpret it carefully

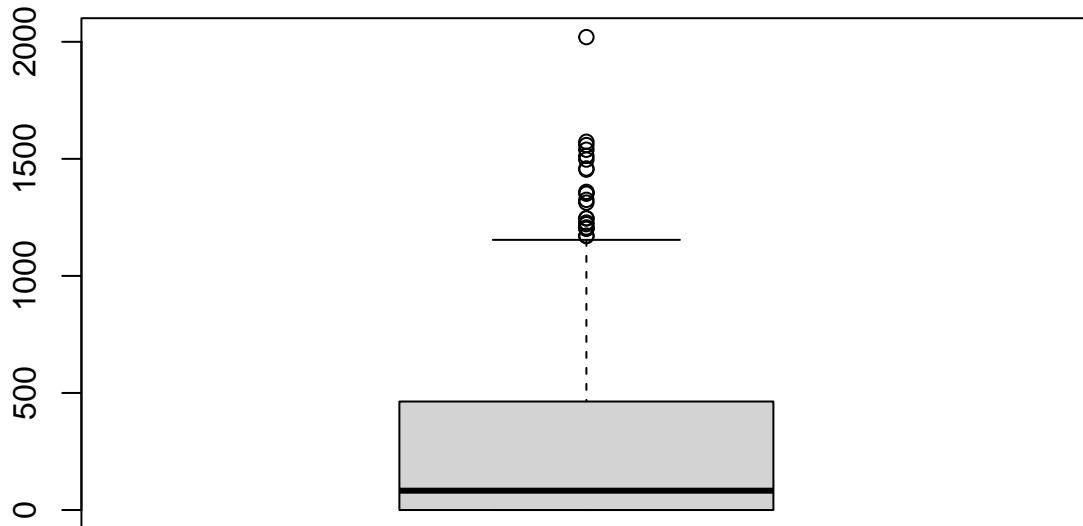
```
summary(data_csv$newCases)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##       0.0     0.0    82.5   308.8   463.2   2020.0
```

The summary shows that the median is 82.5 which means that for half of the dates we took the new cases per day was less than 82.5 and the new cases per day picked at 2020.

14. Get “box and whisker” plot of “newCases” variable in R Studio and interpret it carefully


```
boxplot(data_csv$newCases)
```



The plot shows that the very high new cases per day was exception and not the norm. The median of the data is very low compared to the max value which means that the for up-to mid point the number of cases till the mid point of the data was less and it later on increased exponentially.

15. Import “SAQ8.sav” data in R Studio and get frequency distribution (number and percentage of the attributes) of q01, q03, q06 and q08 variables on R Studio and interpret them carefully

```
# Reading the SPSS file
library(haven)
file_path1="SAQ8.sav"
savdf<-read_sav(file = file_path1)
head(savdf)
```

```
## # A tibble: 6 x 8
##       q01      q02      q03      q04      q05      q06      q07      q08
##   <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl+lbl> <dbl+lbl>
## 1 2 [Agree]      1 [Stro~ 4 [Disa~ 2 [Agre~ 2 [Agre~ 2 [Agr~ 3 [Nei~ 1 [Str~
## 2 1 [Strongly agree] 1 [Stro~ 4 [Disa~ 3 [Neit~ 2 [Agre~ 2 [Agr~ 2 [Agr~ 2 [Agr~
## 3 2 [Agree]      3 [Neit~ 2 [Agre~ 2 [Agre~ 4 [Disa~ 1 [Str~ 2 [Agr~ 2 [Agr~
## 4 3 [Neither]    1 [Stro~ 1 [Stro~ 4 [Disa~ 3 [Neit~ 3 [Nei~ 4 [Dis~ 2 [Agr~
## 5 2 [Agree]      1 [Stro~ 3 [Neit~ 2 [Agre~ 2 [Agre~ 3 [Nei~ 3 [Nei~ 2 [Agr~
## 6 2 [Agree]      1 [Stro~ 3 [Neit~ 2 [Agre~ 4 [Disa~ 4 [Dis~ 4 [Dis~ 2 [Agr~
```

```
summary(savdf)
```

```
##          q01          q02          q03          q04
## Min.    :1.000   Min.    :1.000   Min.    :1.000   Min.    :1.000
## 1st Qu.:2.000   1st Qu.:1.000   1st Qu.:2.000   1st Qu.:2.000
## Median :2.000   Median :1.000   Median :3.000   Median :3.000
## Mean    :2.374   Mean    :1.623   Mean    :2.585   Mean    :2.786
## 3rd Qu.:3.000   3rd Qu.:2.000   3rd Qu.:3.000   3rd Qu.:3.000
## Max.    :5.000   Max.    :5.000   Max.    :5.000   Max.    :5.000
##          q05          q06          q07          q08
## Min.    :1.000   Min.    :1.000   Min.    :1.000   Min.    :1.000
## 1st Qu.:2.000   1st Qu.:1.000   1st Qu.:2.000   1st Qu.:2.000
## Median :3.000   Median :2.000   Median :3.000   Median :2.000
## Mean    :2.722   Mean    :2.227   Mean    :2.924   Mean    :2.237
## 3rd Qu.:3.000   3rd Qu.:3.000   3rd Qu.:4.000   3rd Qu.:3.000
## Max.    :5.000   Max.    :5.000   Max.    :5.000   Max.    :5.000
```

```
# install.packages('plyr')
library(plyr)
col_list<-c('q01','q03','q06','q08')
for (i in 1:length(col_list)){
  cat("Frequency and Percentage For",col_list[i],"\n")
  df_count<-count(savdf[col_list[i]])
  df_count$Percentage <- round(100*df_count$freq/sum(df_count$freq),3)
  print(df_count)
}
```

```
## Frequency and Percentage For q01
##   q01 freq Percentage
## 1   1  270    10.502
## 2   2 1338    52.042
## 3   3  735    28.588
## 4   4  187     7.273
## 5   5   41     1.595
## Frequency and Percentage For q03
##   q03 freq Percentage
## 1   1  497    19.331
## 2   2  672    26.138
## 3   3  878    34.150
## 4   4  448    17.425
## 5   5   76     2.956
## Frequency and Percentage For q06
##   q06 freq Percentage
## 1   1  702    27.305
## 2   2 1127    43.835
## 3   3  344    13.380
## 4   4  252     9.802
## 5   5  146     5.679
## Frequency and Percentage For q08
##   q08 freq Percentage
## 1   1  383    14.897
## 2   2 1487    57.837
## 3   3  482    18.748
```

```
## 4 4 147 5.718
## 5 5 72 2.800
```

For the given columns we calculated the Frequency and Percentage of each factor

16. Import “MR_drugs.xls” data in R Studio and replicate multiple response frequency distribution as shown in the slide 35 of the session 2 slide deck

```
library(readxl)
file_path_xl = "MR_Drugs.xls"
drug_df<-readxl::read_xls(file_path_xl)
head(drug_df)
```

```
## # A tibble: 6 x 27
##   id sex city inco1 inco2 inco3 inco4 inco5 inco6 inco7 pinco1 pinco2
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 1001 2 1 0 0 0 0 0 1 0 6 -1
## 2 1002 2 1 0 1 0 0 0 0 0 2 -1
## 3 1003 2 1 0 0 0 0 0 1 0 6 -1
## 4 1004 2 1 0 1 0 0 0 0 0 2 -1
## 5 1005 2 1 0 0 0 0 0 0 1 7 -1
## 6 1006 2 1 1 1 0 0 0 0 0 2 1
## # ... with 15 more variables: pinco3 <dbl>, pinco4 <dbl>, pinco5 <dbl>,
## # pinco6 <dbl>, sinco1 <chr>, sinco2 <chr>, sinco3 <chr>, sinco4 <chr>,
## # sinco5 <chr>, sinco6 <chr>, crime1 <dbl>, crime2 <dbl>, crime3 <dbl>,
## # crime4 <dbl>, crime5 <dbl>
```

```
summary(drug_df)
```

```
##           id           sex           city           inco1
## Min.      :1001   Min.    :1.000   Min.    :1.000   Min.    :0.0000
## 1st Qu.:1254   1st Qu.:1.000   1st Qu.:1.000   1st Qu.:0.0000
## Median :3148   Median :2.000   Median :2.000   Median :0.0000
## Mean     :2803   Mean     :1.736   Mean     :1.988   Mean     :0.2325
## 3rd Qu.:4098   3rd Qu.:2.000   3rd Qu.:3.000   3rd Qu.:0.0000
## Max.     :4365   Max.      :2.000   Max.      :3.000   Max.      :1.0000
##
##           NA's      :1
##           inco2           inco3           inco4           inco5
## Min.      :0.0000   Min.    :0.0000   Min.    :0.00000   Min.    :0.00000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.:0.00000   1st Qu.:0.00000
## Median :1.0000   Median :0.0000   Median :0.00000   Median :0.00000
## Mean     :0.6245   Mean     :0.3014   Mean     :0.05144   Mean     :0.08436
## 3rd Qu.:1.0000   3rd Qu.:1.0000   3rd Qu.:0.00000   3rd Qu.:0.00000
## Max.     :1.0000   Max.      :1.0000   Max.      :1.00000   Max.      :1.00000
##
##           inco6           inco7           pinco1           pinco2
## Min.      :0.0000   Min.    :0.0000   Min.     :-1.000   Min.     :-1.000
## 1st Qu.:0.0000   1st Qu.:0.0000   1st Qu.: 2.000   1st Qu.: -1.000
## Median :0.0000   Median :0.0000   Median : 3.000   Median : 1.000
## Mean     :0.1553   Mean     :0.3621   Mean      :3.628   Mean      :1.297
## 3rd Qu.:0.0000   3rd Qu.:1.0000   3rd Qu.: 6.000   3rd Qu.: 3.000
## Max.     :1.0000   Max.      :1.0000   Max.       :7.000   Max.       :7.000
```

```
##
##      pinco3      pinco4      pinco5      pinco6
## Min.   :-1.00000 Min.   :-1.0000 Min.   :-1.0000 Min.   :-1.0000
## 1st Qu.: -1.00000 1st Qu.: -1.0000 1st Qu.: -1.0000 1st Qu.: -1.0000
## Median :-1.00000 Median :-1.0000 Median :-1.0000 Median :-1.0000
## Mean   :-0.01646 Mean    :-0.7274 Mean    :-0.9095 Mean    :-0.9794
## 3rd Qu.: -1.00000 3rd Qu.: -1.0000 3rd Qu.: -1.0000 3rd Qu.: -1.0000
## Max.    : 7.00000 Max.     : 7.0000 Max.     : 7.0000 Max.     : 6.0000
##
##      sinco1      sinco2      sinco3      sinco4
## Length:972      Length:972      Length:972      Length:972
## Class :character Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character Mode  :character
##
##
##
##      sinco5      sinco6      crime1      crime2
## Length:972      Length:972      Min.    :0.0000 Min.    :0.00000
## Class :character Class :character 1st Qu.:0.0000 1st Qu.:0.00000
## Mode  :character Mode  :character Median :0.0000 Median :0.00000
##                                     Mean   :0.3881 Mean   :0.08159
##                                     3rd Qu.:0.0000 3rd Qu.:0.00000
##                                     Max.    :3.0000 Max.    :3.00000
##                                     NA's    :65   NA's    :65
##
##      crime3      crime4      crime5
## Min.    :0.00000 Min.    :0.0000 Min.    :0.00000
## 1st Qu.:0.00000 1st Qu.:0.0000 1st Qu.:0.00000
## Median :0.00000 Median :0.0000 Median :0.00000
## Mean    :0.06946 Mean    :0.2745 Mean    :0.07056
## 3rd Qu.:0.00000 3rd Qu.:0.0000 3rd Qu.:0.00000
## Max.    :2.00000 Max.    :3.0000 Max.    :3.00000
## NA's    :65     NA's    :65     NA's    :65
```

```
drug_data_inc<-data.frame(N=colSums(drug_df[4:10]),
                           Percent=round((colSums(drug_df[4:10])/sum(drug_df[4:10]))*100,3),
                           PercentOfCases=round((colSums(drug_df[4:10])/nrow(drug_df[4:10]))*100,3)
                           )
drug_data_inc
```

```
##      N Percent PercentOfCases
## inco1 226  12.834      23.251
## inco2 607  34.469      62.449
## inco3 293  16.638      30.144
## inco4  50   2.839       5.144
## inco5  82   4.656       8.436
## inco6 151   8.575      15.535
## inco7 352  19.989      36.214
```