Report on US crimes for last 10 years

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1 Introduction

RStudio is a free and open source integrated development environment (IDE) for R. An R is a programming language for statistical computing and graphics.

1.1 Introduction to Quandl

Quandl is data platform covering millions of time-series datasets from hundreds of sources. Quandl covers all the numerical data in world in different formats. In this paper the data is collected in the csv file format from the following Quandl data source:

 $https: //www.quandl.com/FBI_{U}CR/USCRIME_{S}TATE_{I}LLINOIS$

1.2 Introduction to data

In this tutorial we are taking raw data of all the crimes reported from the year 1965 to 2010. Though we will only be using last 10 years records by scrubing the data from raw data. Also we will only take three columns including the population, violent crimes total and murders and nonnegligent menslaughters.

2 Reading in the data

Reading data means we will read data from csv file in R. This can be done by uploading csv file to R using Upload option in the right hand corner box. Also following are the libraries used in R. For example ggplot2 is used for plotting graph from the clean data.

- > library(ggplot2)
- > library(XML)
- > library(reshape2)
- > # Upload csv file to R using upload option

```
> # Code for reading the csv file
> UScrimes <- read.csv(file="crimes.csv", header=TRUE, sep=",")</pre>
```

Here the **read.csv** command reads the csv file in R **file** = ""- In double quotes the name of the file is written which is to be read.

1) header=TRUE: You can choose whether or not to have a header by specifying the optional argument header=TRUE or header=FALSE to the read.csv function.

This is important if you have a header but lack row names, because R's guess is based on the fact that the header line has one less entry than the next row

2) sep: the field separator character. Values on each line of the file are separated by this character. If sep = "" (the default) the separator is $\mathring{a} \check{A} \ddot{Y}$ white $space \mathring{a} \check{A} \acute{Z}$, that is one or more spaces, tabs, newlines or carriage returns. But here it is separated by comma(,).

Use of head command Now since we have read data in R, we want to see how the data looks like in R. So we use *head* command to observe first few rows and all the columns of data.

> head (UScrimes)

	Year	Population	Violent.Crimes.To	tal	
1	2010-12-31	12830632	55	835	
2	2009-12-31	12910409	64	185	
3	2008-12-31	12842954	67	840	
4	2007-12-31	12852548	68	528	
5	2006-12-31	12831970	69	498	
6	2005-12-31	12765427	70	496	
	Murders.and	d.Nonneglige	ent.Manslaughters	Forible.Rapes	Robberies
1			706	3033	20054
2			773	3901	22923
3			790	4104	24067
4			752	4103	23100
5			780	4078	23782
6			770	4313	23255
	Aggravated	.Assaults Pr	coperty.Crimes.Tot	al Burglaries	Larceny.Thefts
1		32042	3439	75399	239794
2		36588	3533	77850	248821
3		38879	3812	47 78968	269553
4		40573	3773	75524	267911
5		40858	3874	77259	272578
6		42158	3946	77635	277662
	Motor.Vehic	cle.Thefts V	$^\prime$ iolent.Crime.Rate	•	
1		28796	435.2	!	
2		26676	497.2	!	
3		32726	528.2	!	
4		33887	533.2	<u>}</u>	

```
5
                                       541.6
                  37641
6
                  39373
                                       552.0
  Murder.and.Nonnegligent.Manslaughter.Rate Forible.Rape.Rate Robbery.Rate
1
                                           5.5
                                                              23.6
                                                                           156.3
2
                                           6.0
                                                              30.2
                                                                           177.6
3
                                                              32.0
                                           6.2
                                                                           187.4
4
                                           5.9
                                                              31.9
                                                                           179.7
5
                                           6.1
                                                              31.8
                                                                           185.3
6
                                           6.0
                                                              33.8
                                                                           182.2
  Aggravated. Assault. Rate Property. Crime. Rate Burglary. Rate Larceny. Theft. Rate
1
                     249.7
                                          2681.0
                                                          587.6
                                                                              1868.9
2
                     283.4
                                          2736.9
                                                          603.0
                                                                              1927.3
3
                                                                              2098.8
                     302.7
                                          2968.5
                                                          614.9
4
                     315.7
                                          2935.8
                                                          587.6
                                                                              2084.5
5
                     318.4
                                          3019.6
                                                          602.1
                                                                              2124.2
6
                     330.3
                                          3092.0
                                                          608.2
                                                                              2175.0
  Motor.Vehicle.Theft.Rate
1
                      224.4
2
                      206.6
3
                      254.8
4
                      263.7
5
                      293.3
6
                      308.0
```

Above is the raw data from which we have to obtain clean data

3 Cleaning the data

Now the next step is to extract the required data from the raw data. The process includes scrubing and cleaning of data.

```
> # We use the following command to clean the data from csv file > part1<-UScrimes[1:10 ,1:4]
```

Here in part1 which is name of the file we are storing 1 to 10 rows and 1 to 4 columns

As we can see in above tables we have inconvenient long column names. The command below allows to change the column names.

```
> colnames(part1) <- c("Year", "Population", "Violent crimes", "Murders")
> head(part1,10)
```

	Year	Population	Violent	crimes	Murders
1	2010-12-31	12830632		55835	706
2	2009-12-31	12910409		64185	773
3	2008-12-31	12842954		67840	790

4	2007-12-31	12852548	68528	752
5	2006-12-31	12831970	69498	780
6	2005-12-31	12765427	70496	770
7	2004-12-31	12712016	69365	780
8	2003-12-31	12649087	70376	895
9	2002-12-31	12586447	75759	961
10	2001-12-31	12520227	79270	982

The colnames helps to rename columns

4 Write new csv file using clean data

As we have clean data now we can write that clean data in a new csv file format by using the write.csv command

```
> # writing new csv file with cleaned data
> write.csv(part1, file = "part1.csv")
```

Here write.csv is the command. part1 is the table containing the data and part1.csv is the file name for the csv file format

5 Data

Different commands are used to see the data in different ways. For example: 1) Class command:- To find out an object's type use the class command

```
> class(part1)
```

[1] "data.frame"

After running the *class* command we got the result as *data.frame* which means that *part1* is a *dataframe*

Also we can find the object type of each column. For example:

```
> class(part1$Year)
```

[1] "factor"

The result says it is a factor. Similarly for others such as population:

```
> class(part1$Population)
```

[1] "numeric"

The object type is numeric

> class(part1\$Murders)

[1] "numeric"

The object type is numeri:

2) Structure command: It provides information about the structure of some oject

> str(part1)

```
'data.frame': 10 obs. of 4 variables:

$ Year : Factor w/ 51 levels "1960-12-31","1961-12-31",...: 51 50 49 48 47 46 45 44

$ Population : num 12830632 12910409 12842954 12852548 12831970 ...

$ Violent crimes: num 55835 64185 67840 68528 69498 ...

$ Murders : num 706 773 790 752 780 770 780 895 961 982
```

In this structure the line says 10 observations of 4 variables which means we have four variables in the data frame which are: 1) Year 2) Population 3) Violent crimes 4) Murders

3) Summary command: summary is a generic function used to produce result summaries of the results of various model fitting functions

> summary(part1)

Year	Population	Violent crimes	Murders
2001-12-31:1	Min. :12520227	Min. :55835	Min. :706.0
2002-12-31:1	1st Qu.:12664819	1st Qu.:68012	1st Qu.:770.8
2003-12-31:1	Median :12798030	Median :69432	Median :780.0
2004-12-31:1	Mean :12750172	Mean :69115	Mean :818.9
2005-12-31:1	3rd Qu.:12840208	3rd Qu.:70466	3rd Qu.:868.8
2006-12-31:1	Max. :12910409	Max. :79270	Max. :982.0
(Other) :4			

The **summary** function gives the diffrent values for different columns from the new table such as:

Population: *Minimum* 12520227 *Maximun* 12910409 *Median* 2798030 *Mean* 12750172

Violent crimes: Minimum 55835 Maximun 79270 Median 69432 Mean 69115 Murders: Minimum 706 Maximun 982 Median 780 Mean 818.9

5.1 Result

After cleaning the data and performing various commands we get the following results:

1.) Table— Now we have four columns in our table with new column names namely *Year* which is a *factor*. *Population* with the object type as *nueric*. *Violent crimes* again with the object type as *numeric* and last column as *Murders* with the objet type *numeric*. All the columns have last ten years record that is from 2001 to 2010.

According to the Population numbers the population has increased every year **except** from the year 2009 to 2010 the population has decreased from 12910409 to 12830632.

Figures of Violent crimes show that they have decreased from 79270 to 69365 in year 2004 but again increased to 70496 in 2005 and then gradually decreased to 55835 in 2010.

In case of Murders the figures are decreasing from 982 to 706 in 2010 though their was a little increase in between 2007 to 2008.

6 Reshaping the data

The melt function comes from the library reshape 2. The melt function takes data in wide format and stacks a set of columns into a single column of data.

```
> # Using the melt command for the
> part1.m <- melt(part1)
> part1.m
```

	Year	va	riable	value
1	2010-12-31	Popu	lation	12830632
2	2009-12-31	-	lation	12910409
3	2008-12-31	-	lation	12842954
4	2007-12-31	-	lation	12852548
5	2006-12-31	-	lation	12831970
6	2005-12-31	-	lation	12765427
7	2004-12-31	-	lation	12712016
8	2003-12-31	_	lation	12649087
9	2002-12-31	-	lation	12586447
10	2001-12-31	-	lation	12520227
11	2010-12-31	Violent		55835
12	2009-12-31	Violent	crimes	64185
13	2008-12-31	Violent	crimes	67840
14	2007-12-31	Violent	crimes	68528
15	2006-12-31	Violent	crimes	69498
16	2005-12-31	Violent	crimes	70496
17	2004-12-31	Violent	crimes	69365
18	2003-12-31	Violent	crimes	70376
19	2002-12-31	Violent	crimes	75759
20	2001-12-31	Violent	crimes	79270
21	2010-12-31	M	urders	706
22	2009-12-31	M	urders	773
23	2008-12-31	M	urders	790
24	2007-12-31	M	urders	752
25	2006-12-31	M	urders	780
26	2005-12-31	M	urders	770
27	2004-12-31	M	urders	780

28	2003-12-31	Murders	895
29	2002-12-31	Murders	961
30	2001-12-31	Murders	982

Here melt command is melting the part1 table into part1.m which will be our new table with new table name

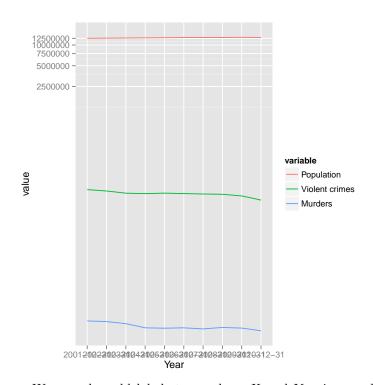
Now the part1.m table contains three columns namely Year, Variable and Values. The variable column has three names grouped which are opulation, Violent crimes and Murders and all the figures from these three columns are grouped into value column. We do this step to create the graph or plot.

7 Creating a Plot

Since we have already imported library ggplot2, we can now use the ggplot command to plot our graph from the melted data.

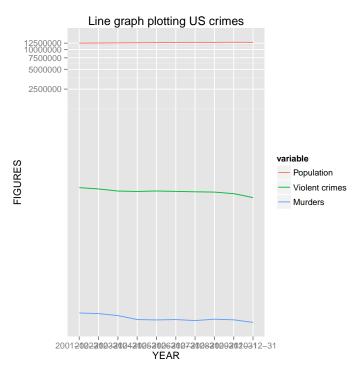
```
> # Using the ggplot 2 we make a line graph for three variables
> plot<-ggplot(part1.m, aes(x=Year, y=value, group=variable, colour=variable)) +geom_line()-</pre>
```

> print(plot)



We can also add labels to graph on X and Y axis as well as give title to graph.

- > # Now, we want to add a title to the graph and custom legends
- $> plot1 < -ggplot(part1.m, aes(x=Year, y=value, group=variable, colour=variable)) + geom_line() +$
- > print(plot1)



In order to increase the width of the plot to 10 we can simply type width = 10 in the (\blacksquare fig=true, echo= false, width=10 \blacksquare =) and it will increase the size of plot horizontly

7.1 Result

In the first Graph

The graph shows three coloured lines each for **Population**, **Violent crimes** and **Murders** which were grouped under the cloumn named *variable* in the reshaped table. On the X axis shows Year from 2001 to 2010. and on Y axis is the scale for the figures grouped under the value column of the reshaped table. The population line looks constant due to the large scale used with little variation in values. The violent crimes line in green color shows the value decreasing and the line in blue color named Murders shows small increase and decrease in middle years with final decrease in 2010.

The changes from year 2001 to 2010 explains the population increase within those years while gradual decrease in violent crimes and murders. In conclusion the crime rate has decreased from 2001 to 2010.

In second graph

The second graph merely shows the labeling of X and Y axis with inclusion of title in the graph