

# Question 1

Use the “USArrests” built-in dataset to plot beautiful graphs and find meaningful insight about the dataset. You may draft questions yourself and summarize the results. You will be marked for [10]

- a. Creativity
- b. Presentation
- c. Originality
- d. Summarization of results

## Getting to know the dataset

Let us start by viewing and understanding the dataset.

```
data(USArrests)
help(USArrests)
```

USArrests {datasets}

R Documentation

## Violent Crime Rates by US State

### Description

This data set contains statistics, in arrests per 100,000 residents for assault, murder, and rape in each of the 50 US states in 1973. Also given is the percent of the population living in urban areas.

### Usage

USArrests

### Format

A data frame with 50 observations on 4 variables.

- [,1] Murder     numeric Murder arrests (per 100,000)
- [,2] Assault    numeric Assault arrests (per 100,000)
- [,3] UrbanPop   numeric Percent urban population
- [,4] Rape        numeric Rape arrests (per 100,000)

### Note

USArrests contains the data as in McNeil's monograph. For the UrbanPop percentages, a review of the table (No. 21) in the Statistical Abstracts 1975 reveals a transcription error for Maryland (and that McNeil used the same “round to even” rule that R's [round\(\)](#) uses), as found by Daniel S Coven (Arizona).

See the example below on how to correct the error and improve accuracy for the '<n>.5' percentages.

### Source

World Almanac and Book of facts 1975. (Crime rates).

Statistical Abstracts of the United States 1975, p. 20. (Urban rates) is available online at <https://books.google.ch/books?id=1QeAAAMAAJ18&pg=PA20>

```
names(USArrests)
```

```
[1] "Murder" "Assault" "UrbanPop" "Rape"
```

```
dim(USArrests)
```

```
[1] 50 4
```

```
View(USArrests)
```

	Murder	Assault	UrbanPop	Rape
Alabama	13.2	236	58	21.2
Alaska	10.0	263	48	44.5
Arizona	8.1	294	80	31.0
Arkansas	8.8	190	50	19.5
California	9.0	276	91	40.6
Colorado	7.9	204	78	38.7
Connecticut	3.3	110	77	11.1
Delaware	5.9	238	72	15.8
Florida	15.4	335	80	31.9
Georgia	17.4	211	60	25.8
Hawaii	5.3	46	83	20.2
Idaho	2.6	120	54	14.2
Illinois	10.4	249	83	24.0
Indiana	7.2	113	65	21.0
Iowa	2.2	56	57	11.3
Kansas	6.0	115	66	18.0
Kentucky	9.7	109	52	16.3
Louisiana	15.4	249	66	22.2
Maine	2.1	83	51	7.8
Maryland	11.3	300	67	27.8

Showing 1 to 21 of 50 entries, 4 total columns

So this is a dataset about the arrest reports in USA. There are four attributes and we can try to find the relations between them. We can also try to find the least crime filled states or the most crime filled states. The dataset provides us with a list of 50 rows (1 for each state) and its statistics for 4 types of crimes - Murder, Assault, UrbanPop and Rape. Let us try to have some closer look at the dataset by obtaining its summary.

```
summary(USArrests)
```

```
> summary(USArrests)
```

Murder	Assault	UrbanPop	Rape
Min. : 0.800	Min. : 45.0	Min. : 32.00	Min. : 7.30
1st Qu.: 4.075	1st Qu.:109.0	1st Qu.:54.50	1st Qu.:15.07
Median : 7.250	Median :159.0	Median :66.00	Median :20.10
Mean : 7.788	Mean :170.8	Mean :65.54	Mean :21.23
3rd Qu.:11.250	3rd Qu.:249.0	3rd Qu.:77.75	3rd Qu.:26.18
Max. :17.400	Max. :337.0	Max. :91.00	Max. :46.00

```
> |
```

Let us also try to see if the four crimes have any correlation to each other.

```
cor(USArrests)
```

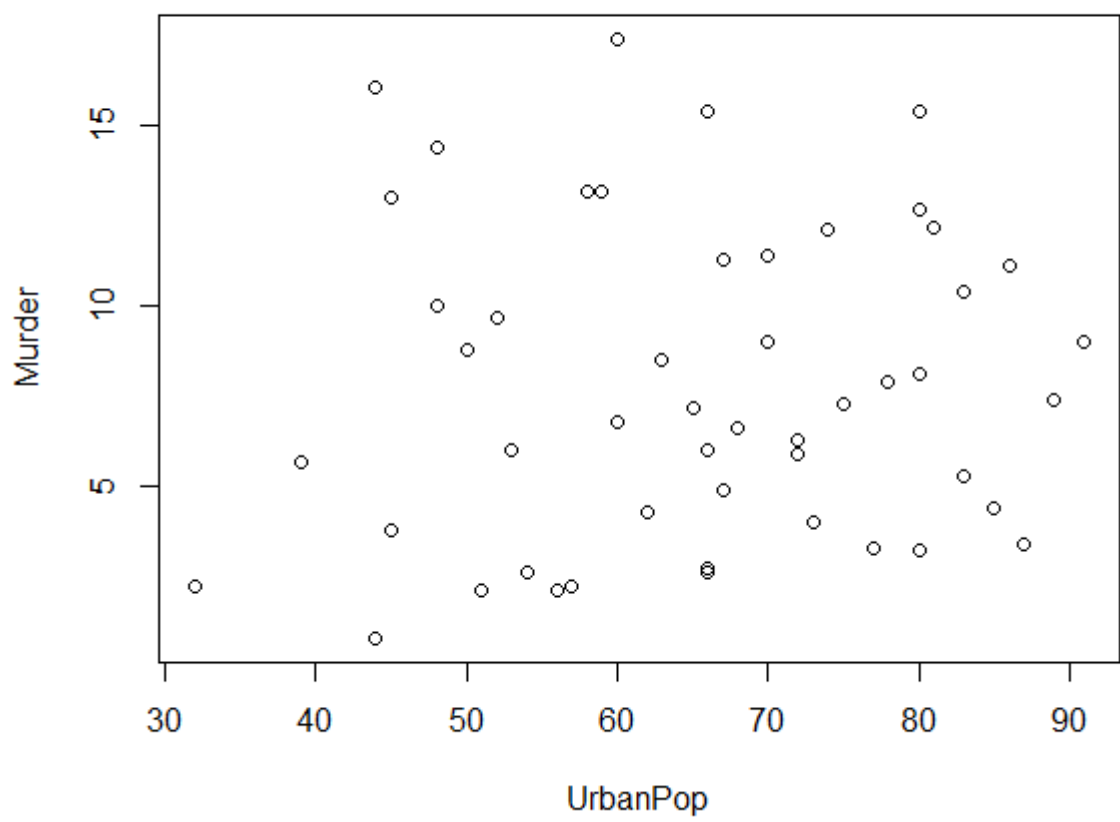
```
> cor(USArrests)
      Murder  Assault  UrbanPop  Rape
Murder  1.00000000  0.8018733  0.06957262  0.5635788
Assault  0.80187331  1.0000000  0.25887170  0.6652412
UrbanPop 0.06957262  0.2588717  1.00000000  0.4113412
Rape     0.56357883  0.6652412  0.41134124  1.0000000
> |
```

Through the above data we infer that **Assault** are the most frequent crimes that are happening as it has the highest averages. Further **Rape** crimes are the least likely to happen (which is a good thing).

### Visualising data

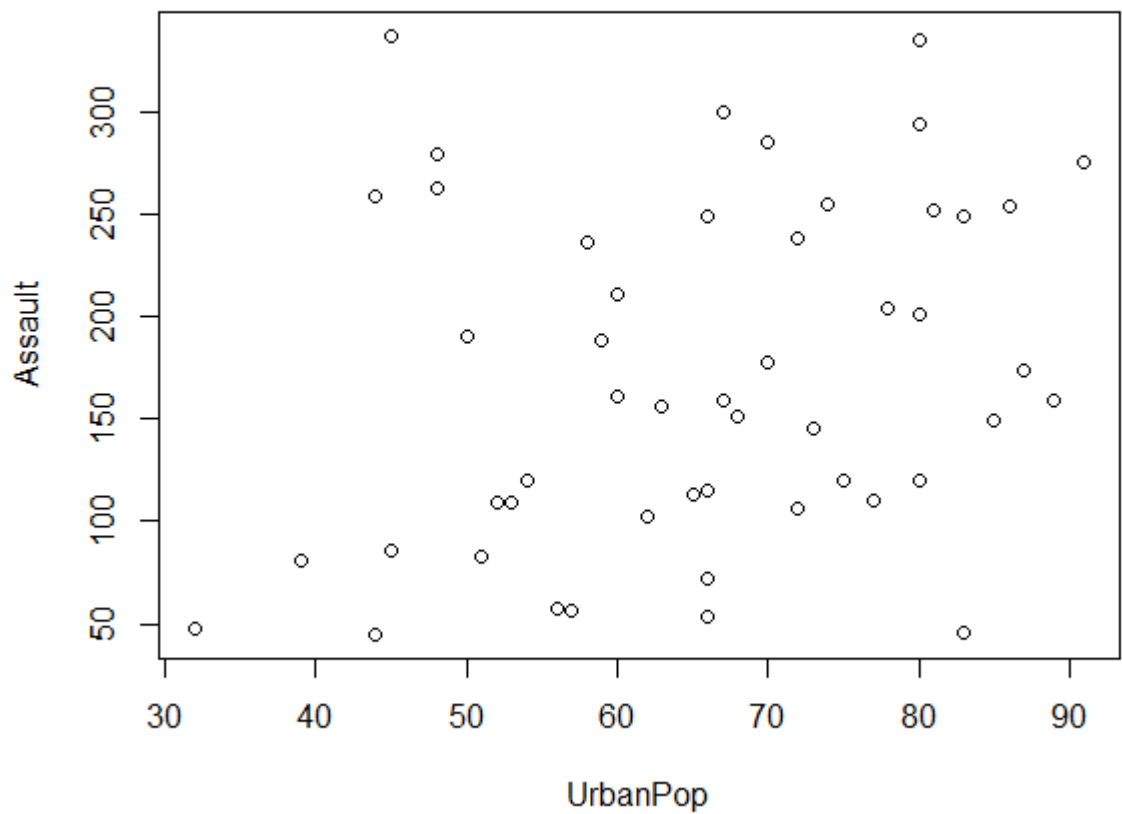
Running plot for these combinations, Murder and Assault do not appear to have a relation to UrbanPop. The distribution of plot points are scattered to the point that they do not appear to correlate to UrbanPop.

```
with(USArrests, plot(UrbanPop, Murder))
```



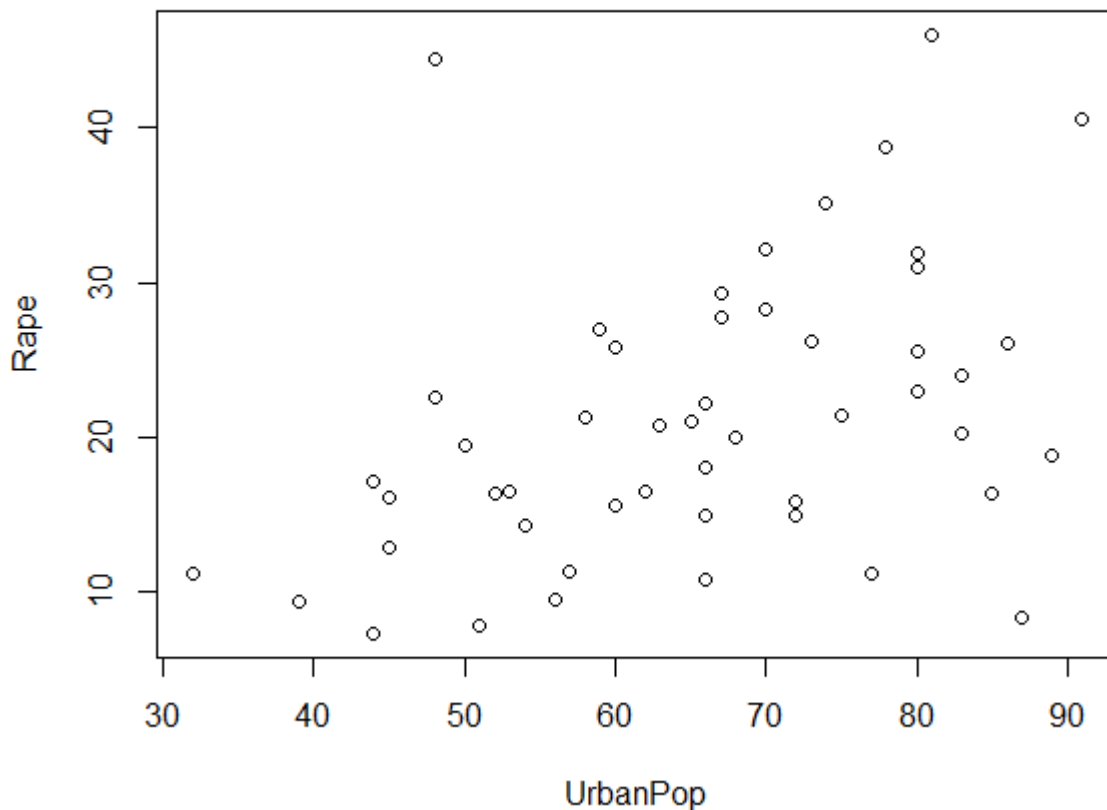
We see that most murders are likely to occur with higher Urban population. Although this isn't a fair measure since the plot seems to be very scattered. Let us try another one.

```
with(USArrests, plot(UrbanPop, Assault))
```



It is quite evident that more Assaults are likely to occur with more Urban Population.

```
with(USArrests, plot(UrbanPop, Rape))
```



Rape crimes are more likely to occur in states with average urban population. This goes to contrary to standard belief that more Urban population leads to more rapes.

### Which states has most and least assault, murder, and rape arrests?

Let us try to figure out which states are more safer than the others so that if we ever plan a trip to US, we know where to steer clear off.

#### Most and Least assault

```
x <- which(USArrests$Assault == max(USArrests$Assault))
rownames(USArrests)[x]
```

```
[1] "North Carolina"
```

```
x <- which(USArrests$Assault == min(USArrests$Assault))
rownames(USArrests)[x]
```

[1] "North Dakota"

#### Most and Least murder

```
x <- which(USArrests$Murder == max(USArrests$Murder))
rownames(USArrests)[x]
```

```
[1] "Georgia"
```

```
x <- which(USArrests$Murder == min(USArrests$Murder))
rownames(USArrests)[x]
```

```
[1] "North Dakota"
```

### Most and least rape

```
x <- which(USArrests$Rape == max(USArrests$Rape))
rownames(USArrests)[x]
```

```
[1] "Nevada"
```

```
x <- which(USArrests$Rape == min(USArrests$Rape))
rownames(USArrests)[x]
```

```
[1] "North Dakota"
```

### States which have assault arrests more than median of the country.

```
assault.median = median(USArrests$Assault)
assault.median
```

```
[1] 159
```

```
subset(USArrests, Assault > assault.median, select= c(UrbanPop, Assault))
```

```
> subset(USArrests, Assault > assault.median, select= c(UrbanPop, Assault))
```

	UrbanPop	Assault
Alabama	58	236
Alaska	48	263
Arizona	80	294
Arkansas	50	190
California	91	276
Colorado	78	204
Delaware	72	238
Florida	80	335
Georgia	60	211
Illinois	83	249
Louisiana	66	249
Maryland	67	300
Michigan	74	255
Mississippi	44	259
Missouri	70	178
Nevada	81	252
New Mexico	70	285
New York	86	254
North Carolina	45	337
Rhode Island	87	174
South Carolina	48	279
Tennessee	59	188
Texas	80	201
Wyoming	60	161

### States that are in the bottom 25% of murder

These are the safer states that I would prefer to go to.

```
bottomQuartileMurderRate <- quantile(USArrests$Murder)[2]
bottomQuartileMurderRate
```

25%  
4.075

```
subset(USArrests, Murder < bottomQuartileMurderRate, select= c(UrbanPop, Murder))
```

```
> subset(USArrests, Murder < bottomQuartileMurderRate, select= c(UrbanPop, Murder))
      UrbanPop Murder
Connecticut      77   3.3
Idaho            54   2.6
Iowa             57   2.2
Maine            51   2.1
Minnesota        66   2.7
New Hampshire    56   2.1
North Dakota     44   0.8
Rhode Island     87   3.4
South Dakota     45   3.8
Utah             80   3.2
Vermont          32   2.2
Washington       73   4.0
Wisconsin        66   2.6
> |
```

### States which are in the top 25% of the murder.

Better stay away from these states for our own safety.

```
topQuartileMurderRate <- quantile(USArrests$Murder)[4]
topQuartileMurderRate
```

```
##      75%
## 11.25
```

```
subset(USArrests, Murder > topQuartileMurderRate, select= c(UrbanPop, Murder))
```

```
----
> subset(USArrests, Murder > topQuartileMurderRate, select= c(UrbanPop, Murder))
      UrbanPop Murder
Alabama        58  13.2
Florida        80  15.4
Georgia        60  17.4
Louisiana      66  15.4
Maryland       67  11.3
Michigan       74  12.1
Mississippi    44  16.1
Nevada         81  12.2
New Mexico     70  11.4
North Carolina 45  13.0
South Carolina 48  14.4
Tennessee     59  13.2
Texas         80  12.7
> |
```