

ECOM20001 Econometrics 1

Tutorial 2 Semester 1, 2022

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Part 1: Visualising and Describing Data in R

- `tute2_crime.csv`

```
## Set the working directory for the tutorial file  
setwd("your working directory")
```

Dataset

Dataset `tute2_crime.csv` has the following 5 variables:

- `stateid`: identifier for a US state
- `vio`: violent crime rate: incidents per 100,000 people
- `rob`: robbery rate: incidents per 100,000 people
- `density`: population per square mile of land
- `avginc`: real per capita personal income in the state

Reading in data

```
## Load the dataset from a comma separate value  
data=read.csv("tute2_crime.csv")
```

```
## List the variables in the dataset named data  
names(data)
```

```
## [1] "stateid" "vio"      "rob"      "dens"     "avginc"
```

```
## Dimension of the dataset: 45 observations (states), 5 variables  
dim(data)
```

```
## [1] 45 5
```

Describing data

- 1 Discuss the sample means, standard deviations, minimums and maximums for each of the four main variables in the dataset: vio, rob, density, avginc.

- What does a “typical” state look like in the dataset? Focus on sample means in describing a typical state.

Be sure to state the units of a variable to accurately describe what a typical state looks like.

- Discuss the minimum and maximum of each variable, highlighting the range of values that each variable takes on.
- How varied is the degree of violent crimes and robbery rates, population densities, and per capita incomes in the sample? How violent and robbery-filled is the worst state compared to the best state?

```
## Using the summary() function
```

```
summary(data)      # Mean, Min, Max, Median, 25th/75th percentile
```

##	stateid	vio	rob	dens	avginc
##	Min. : 1	Min. : 66.9	Min. : 8.8	Min. : 1.086	Min. :12.37
##	1st Qu.:12	1st Qu.:275.5	1st Qu.: 75.3	1st Qu.: 34.542	1st Qu.:13.92
##	Median :23	Median :382.8	Median :100.9	Median : 76.529	Median :15.80
##	Mean :23	Mean :431.5	Mean :106.7	Mean :105.656	Mean :15.82
##	3rd Qu.:34	3rd Qu.:570.0	3rd Qu.:152.5	3rd Qu.:157.042	3rd Qu.:17.11
##	Max. :45	Max. :854.0	Max. :240.8	Max. :385.441	Max. :20.27

Alternative R commands for descriptive statistics

```
sapply(data, mean)    # Means
```

```
##   stateid      vio      rob      dens    avginc  
## 23.00000 431.48444 106.65556 105.65617 15.81649
```

```
sapply(data, median) # Median
```

```
##   stateid      vio      rob      dens    avginc  
## 23.00000 382.80000 100.90000  76.52950 15.79737
```

```
sapply(data, sd)      # Standard Deviation
```

```
##   stateid      vio      rob      dens    avginc  
## 13.13393 209.54125  64.19275  97.66395  1.93695
```

Alternative R commands for descriptive statistics

```
sapply(data, min)      # Min
```

```
## stateid      vio      rob      dens  avginc  
## 1.00000 66.90000  8.80000  1.08610 12.37023
```

```
sapply(data, max)      # Max
```

```
## stateid      vio      rob      dens  avginc  
## 45.0000 854.0000 240.8000 385.4414 20.2728
```

```
sapply(data, quantile)
```

```
##      stateid  vio  rob    dens  avginc  
## 0%           1 66.9   8.8   1.0861 12.37023  
## 25%          12 275.5 75.3  34.5422 13.91905  
## 50%          23 382.8 100.9 76.5295 15.79737  
## 75%          34 570.0 152.5 157.0423 17.11416  
## 100%         45 854.0 240.8 385.4414 20.27280
```

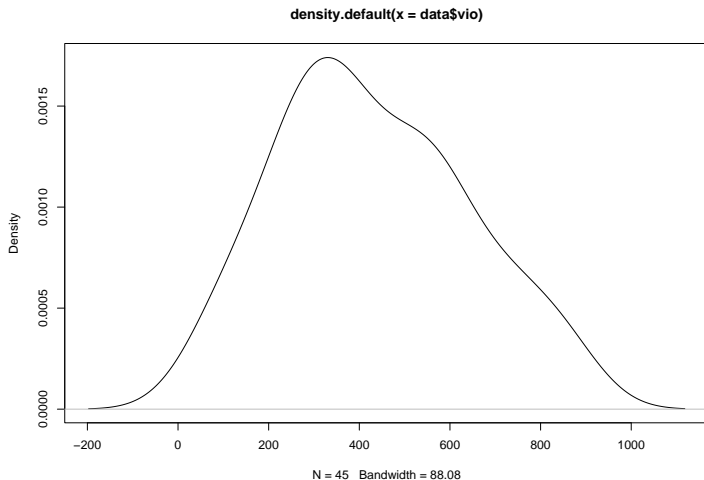
```
## Descriptive Statistics: stargazer()
stargazer(data,
           summary.stat = c("n", "mean", "sd", "median", "min", "max"),
           type = "text",
           title = "Descriptive Statistics")
```

```
##
## Descriptive Statistics
## =====
## Statistic N      Mean      St. Dev. Median    Min      Max
## -----
## stateid    45 23.000    13.134     23         1      45
## vio        45 431.484  209.541   382.800   66.900  854.000
## rob        45 106.656   64.193   100.900    8.800  240.800
## dens       45 105.656   97.664    76.530    1.086  385.441
## avginc     45 15.816    1.937    15.797   12.370  20.273
## -----
```


- ② How do the respective probability densities of `vio`, `rob`, `density`, `avginc` look?
 - Focus on their means, and skewness

A simple plot of probability density

```
plot(density(data$vio))
```



Fancy probability densities

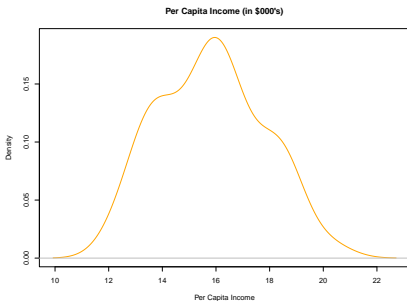
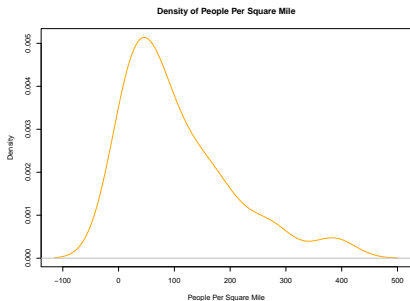
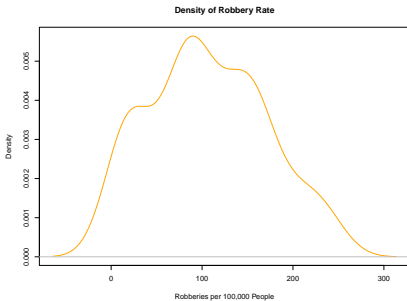
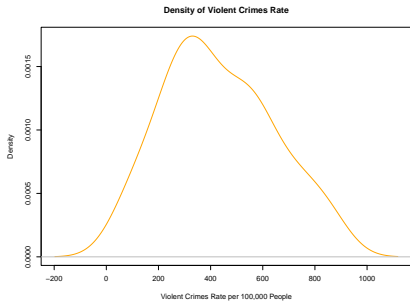
```
## Create probability densities for all relevant variables
```

```
plot(density(data$vio),  
     main="Density of Violent Crimes Rate",  
     xlab="Violent Crimes Rate per 100,000 People",  
     ylab="Density", col="orange")
```

```
plot(density(data$rob),  
     main="Density of Robbery Rate",  
     xlab="Robberies per 100,000 People",  
     ylab="Density", col="orange")
```

```
plot(density(data$dens),  
     main="Density of People Per Square Mile",  
     xlab="People Per Square Mile",  
     ylab="Density", col="orange")
```

```
plot(density(data$avginc),  
     main="Per Capita Income (in $000's)",  
     xlab="Per Capita Income",  
     ylab="Density", col="orange")
```



Describing relationship between two variables

- 3 Comment on the 3 scatter plots below

Visually, does a relationship appear exist in each graph? If so, offer an **economic explanation** for why the relationship might exist.

There may be multiple explanations, so you may offer various explanations if you wish. But just one explanation is fine.

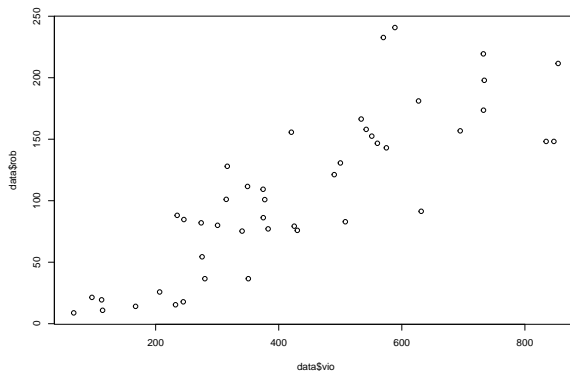
- Robbery vs Violence
- Robbery vs Per Capita Income
- Robbery vs People per Square Mile

Economic explanations

Economic explanations focus on the costs and benefits of a particular behaviour for explaining empirical patterns.

A simple scatter plot

```
## Create scatter plot for vio (x-variable) vs rob (y-variable)  
plot(data$vio,data$rob)
```



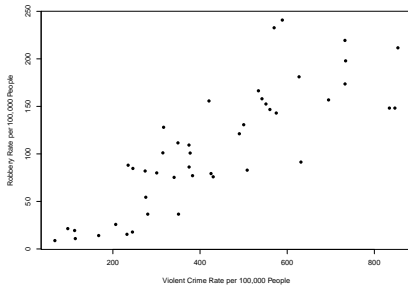
Fancy scatter plots

```
plot(data$vio,data$rob,  
     main="Relationship Between Robbery Rate and Violent Crime Rate",  
     xlab="Violent Crime Rate per 100,000 People",  
     ylab="Robbery Rate per 100,000 People",  
     pch=16)
```

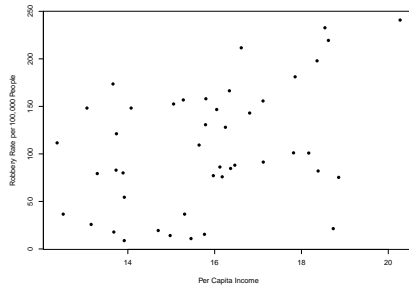
```
plot(data$avginc,data$rob,  
     main="Relationship Between Robbery Rate and Per Capita Income",  
     xlab="Per Capita Income",  
     ylab="Robbery Rate per 100,000 People",  
     pch=16)
```

```
plot(data$dens,data$rob,  
     main="Relationship Between Robbery Rate and Population Density",  
     xlab="Population per Square Mile of Land",  
     ylab="Robbery Rate per 100,000 People",  
     col="red",  
     pch=16)
```

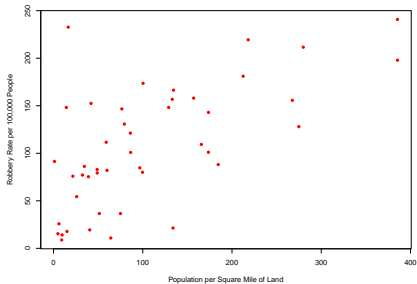
Relationship Between Robbery Rate and Violent Crime Rate



Relationship Between Robbery Rate and Per Capita Income



Relationship Between Robbery Rate and Population Density



To be clear: All “explanations” are just hypotheses and none of them are proven from a simple scatter plot.

- There are potentially many other hypotheses.
- Later in ECOM20001, and throughout ECOM30002: Econometrics 2, we will develop empirical approaches to unpack these various explanations for correlations found in scatter plots.

Part 2: Summation Practice Problems

- 1 Show the following equality is true

$$\sum_{i=1}^n (x_i - \bar{x}) = 0$$

- 2 Show the following equality is true:

$$n\bar{x} = \sum_{i=1}^n x_i$$

- 3 Show the following equality is true

$$\sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - n\bar{x}^2$$

- 4 Show the following equality is true

$$\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) = \sum_{i=1}^n x_i y_i - n\bar{x}\bar{y}$$

Workings to Part 2

1

$$\begin{aligned}\sum_{i=1}^n (x_i - \bar{x}) &= \sum_{i=1}^n x_i - \sum \bar{x} \\ &= \sum_{i=1}^n x_i - n\bar{x} \\ &= \sum_{i=1}^n x_i - n \frac{\sum_{i=1}^n x_i}{n} \\ &= \sum_{i=1}^n x_i - \sum_{i=1}^n x_i = 0\end{aligned}$$

$$\begin{aligned} n\bar{x} &= \sum_{i=1}^n x_i \\ &= n \frac{\sum_{i=1}^n x_i}{n} \\ &= \sum_{i=1}^n x_i \end{aligned}$$

Notice how you can manipulate summations $\sum_{i=1}^n x_i$ and multiply them by $\frac{n}{n}$ to get means and sample sizes e.g. :

$$\sum_{i=1}^n x_i = \frac{n}{n} \sum_{i=1}^n x_i = n\bar{x}$$

$$\begin{aligned}\sum (x_i - \bar{x})^2 &= \sum (x_i^2 - 2\bar{x}x_i + \bar{x}^2) \\ &= \sum x_i^2 - \sum (2\bar{x}x_i) + \sum (\bar{x}^2) \\ &= \sum x_i^2 - 2\bar{x} \sum x_i + n\bar{x}^2 \\ &= \sum x_i^2 - 2\bar{x}n\bar{x} + n\bar{x}^2 \\ &= \sum x_i^2 - n\bar{x}^2\end{aligned}$$

In line 3, you could also multiply the term $2\bar{x} \sum x_i$ by $\frac{n}{n}$ which would give the result is the same as the above.

$$\begin{aligned}
\sum (x_i - \bar{x})(y_i - \bar{y}) &= \sum (x_i y_i - \bar{x} y_i - \bar{y} x_i + \bar{x} \bar{y}) \\
&= \sum (x_i y_i) - \sum (\bar{x} y_i) \\
&\quad - \sum (\bar{y} x_i) + \sum (\bar{x} \bar{y}) \\
&= \sum x_i y_i - \bar{x} \sum y_i - \bar{y} \sum x_i + n \bar{x} \bar{y} \\
&= \sum x_i y_i - n \bar{x} \bar{y} - n \bar{x} \bar{y} + n \bar{x} \bar{y} \\
&= \sum x_i y_i - n \bar{x} \bar{y}
\end{aligned}$$

the terms $\bar{x} \sum y_i$ and $\bar{y} \sum x_i$ in the 3rd line could also be divided by $\frac{n}{n}$ yielding the same result