## YSU ASDS, Statistics, Fall 2019 Lecture 01

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02 September 2019

Welcome

# Welcome to the ASDS Program

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And Happy New Year Semester! =

### Contents

- Syllabus highlights
- ► Intro to the Course
- ▶ Intro to the Descriptive Statistics

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- ► Teaching Assistant: Maybe

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- Other: I have prepared some R Intro Slides

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- Advice: Run over the Probability Topics, especially, about RVs and Distributions

**QA** Session

# Questions?

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Why I need to learn Statistics?

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Simple - to pass this course  $\ddot{\ }$ 

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#### **About Statistics**

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Well, maybe we will talk a little bit about this at the end of the course.

▶ Why I need to learn Statistics?

And finally, to understand the everyday usage of Statistical language, graphs and estimates, say, about polls and salaries —

The structure of our course will be the following:

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The rest will use these topics intensively.

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► Models, Statistical Inference and Learning:

Here we will talk about Parametric and Non-Parametric Statistics, and the main problems of the Parametric Statistics: Parameter Estimation, Confidence Intervals and Hypothesis Testing

Then we will run over these three problems:

Parameter Point Estimates

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- Parameter Point Estimates
- Confidence Intervals
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Then we will focus on the simplest Statistical Model for the relationship between different Variables: we will learn

Linear Regression

And at the end of the course we will return back to Testing and cover:

Goodness of fit tests

Descriptive Statistics is to get the first, basic information about the Data, either in the Visual or Numerical form.

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This dataset contains a subset of the fuel economy data that the EPA makes available on http://fueleconomy.gov. It contains only models which had a new release every year between 1999 and 2008.

head(ggplot2::mpg, 3)

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Lets look at the first 3 rows of our dataset:

```
## # A tibble: 3 x 11
    manufacturer model displ year
##
                                      cyl trans drv
                                                         ctv
##
     <chr>>
                  <chr> <dbl> <int> <int> <chr> <chr> <int> <int> <in</pre>
## 1 audi
                         1.8 1999 4 auto(~ f
                                                          18
                  a4
                 а4
                         1.8 1999 4 manua~ f
                                                          21
## 2 audi
## 3 audi
                  a4
                          2
                               2008
                                        4 manua~ f
                                                          20
```

The variable cty is the *city miles per gallon*, and the variable cyl is the *number of cylinders*. Let's separate that Variables:

```
cty <- ggplot2::mpg$cty
cyl <- ggplot2::mpg$cyl</pre>
```

#### Let's see the results:

cyl

Let's see the results:

cyl

Can you describe this data? What can be said about the No. of Cylinders of these cars?

cty

Let's see the results for cty:

```
[1] 18 21 20 21 16 18 18 18 16 20 19 15 17 17 15 15 17 16 1
##
##
    [24] 16 15 16 15 15 14 11 11 14 19 22 18 18 17 18 17 16 16 1
         16 16 15 14 13 14 14 14 9 11 11 13 13
##
                                                9 13 11 13 11 1
           11 11 13 11 11 11 12 14 15 14 13 13 13 14 14 13 13 1
##
##
    [93] 17 16 15 15 15 15 14 28 24 25 23 24 26 25 24 21 18 18 2
   [116] 19 19 20 20 17 16 17 17 15 15 14 9 14 13 11 11 12 12 1
   [139] 13 13 13 21 19 23 23 19 19 18 19 19 14 15 14 12 18 16 1
   [162] 20 19 20 18 21 19 19 19 20 20 19 20 15 16 15 15 16 14 2
   [185] 18 19 21 21 21 22 18 18 18 24 24 26 28 26 11 13 15 16 1
   [208] 21 19 21 22 17 33 21 19 22 21 21 21 16 17 35 29 21 19 2
   [231] 21 16 18 17
```

cty

Let's see the results for cty:

```
[1] 18 21 20 21 16 18 18 18 16 20 19 15 17 17 15 15 17 16 1
##
##
    [24] 16 15 16 15 15 14 11 11 14 19 22 18 18 17 18 17 16 16 1
    [47] 16 16 15 14 13 14 14 14 9 11 11 13 13 9 13 11 13 11 1
##
          9 11 11 13 11 11 11 12 14 15 14 13 13 13 14 14 13 13 1
##
##
    [93] 17 16 15 15 15 15 14 28 24 25 23 24 26 25 24 21 18 18 2
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   [231] 21 16 18 17
```

Again, can you describe this data? What can be said about the City Miles per Gallon values of these cars?

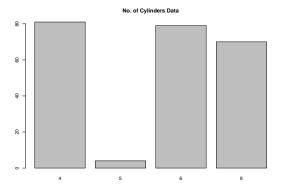
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For example, let us draw the BarPlot for the frequencies of the cyl variable:

barplot(table(cyl), main = "No. of Cylinders Data")



Now, let us give some numerical summaries for cty: calculate the average Miles per Gallon for a City, and its max and min.

```
cat("mean = ", mean(cty))

## mean = 16.85897

cat("Max = ", max(cty))

## Max = 35

cat("Min = ", min(cty))

## Min = 9
```

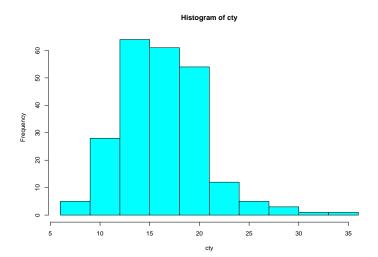
summary(cty)

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## Max = 35
cat("Min = ", min(cty))
## Min = 9
And we can use the summary command to get some numerical info:
```

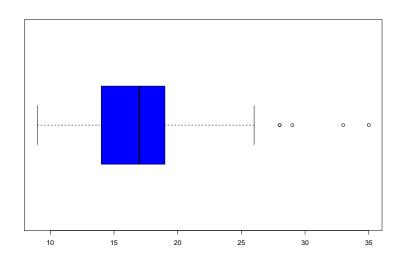
## Min. 1st Qu. Median Mean 3rd Qu. Max. ## 9.00 14.00 17.00 16.86 19.00 35.00

To get some visual information about the Variable cty, its distribution, we can draw the Histogram:



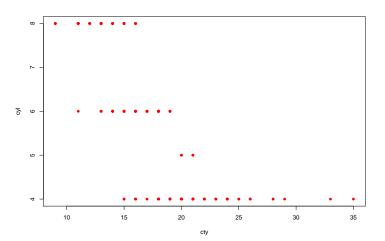
Now, we can draw the BoxPlot of the cty data:

```
boxplot(cty, horizontal = T, col = "blue")
```



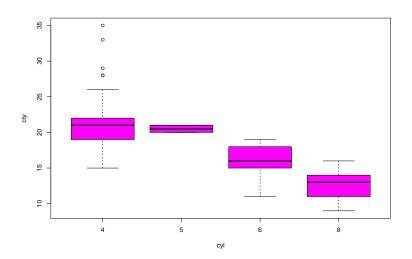
Now, instead of just getting information about cyl and cty separately, let us give visually the relationship between them:

```
plot(cty, cyl, pch=16, col = "red")
```



... or draw a BoxPlot of cty for each type of the cylinder:

```
boxplot(cty~cyl, col="magenta")
```



Moral: our brain cannot get an insight from the list of numbers, but Descriptive Statistics can help  $\ddot{-}$ 

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## Some Important Notions and Definitions

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- ▶ **Statistics** is a numerical characteristic of the *Sample*

Here is one of the standard Datasets in  $\mathbf{R}$ :

#### head(cars)

##		speed	dist
##	1	4	2
##	2	4	10
##	3	7	4
##	4	7	22
##	5	8	16
##	6	9	10

Here is one of the standard Datasets in R:

#### head(cars)

```
## speed dist
## 1 4 2
## 2 4 10
## 3 7 4
## 4 7 22
## 5 8 16
## 6 9 10
```

Which are the Variables ?

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```

- ► Which are the **Variables**?
- Give two Observations.

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