

INTRODUCTION TO DATA SCIENCE

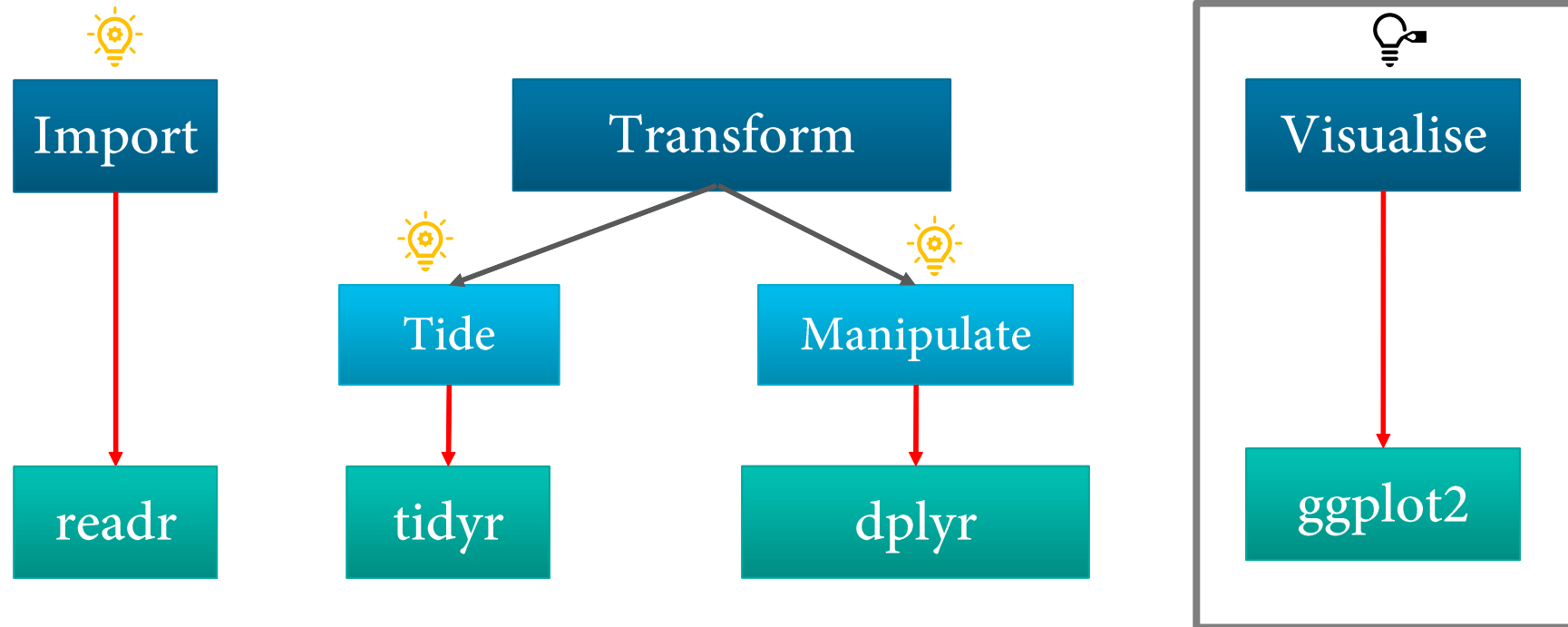
Lecture 10

Dr. Ibrahim Radwan

- Data Wrangling, a recap
- Exploratory Data Analysis, a recap
- EDA; Univariate Analysis
- EDA; Bivariate Analysis
 - Continuous
 - Continuous + discrete
 - Discrete

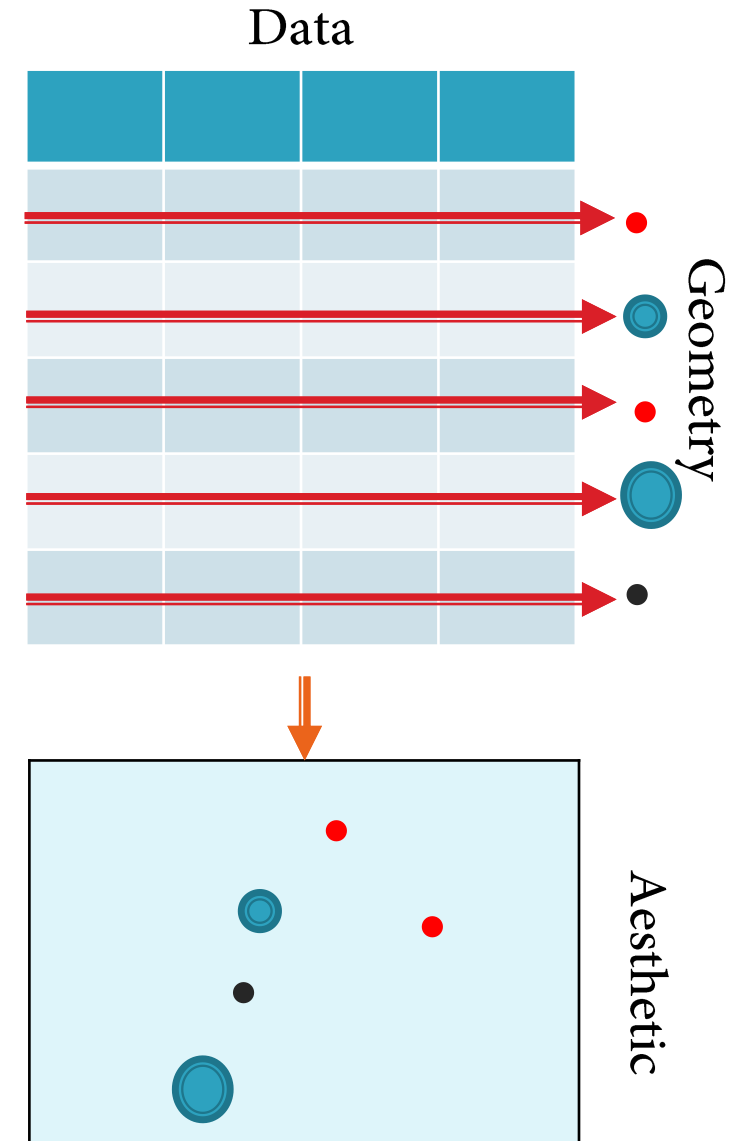
DATA WRANGLING

- Practically, we have three main processes to wrangle the data



GRAPH COMPONENTS

- To build a graph in R , you will need to specify three components:
 1. **Data:** the set of records/variables that we need to represent with a graph
 2. **Geometry:** the type of the plot, which will be generated, usually it is a function such as (scatterplot, boxplot, barplot, histogram, smooth density, etc.)
 3. **Aesthetic mapping:** the coordinate map and the other visual cues, such as size, scale and color.



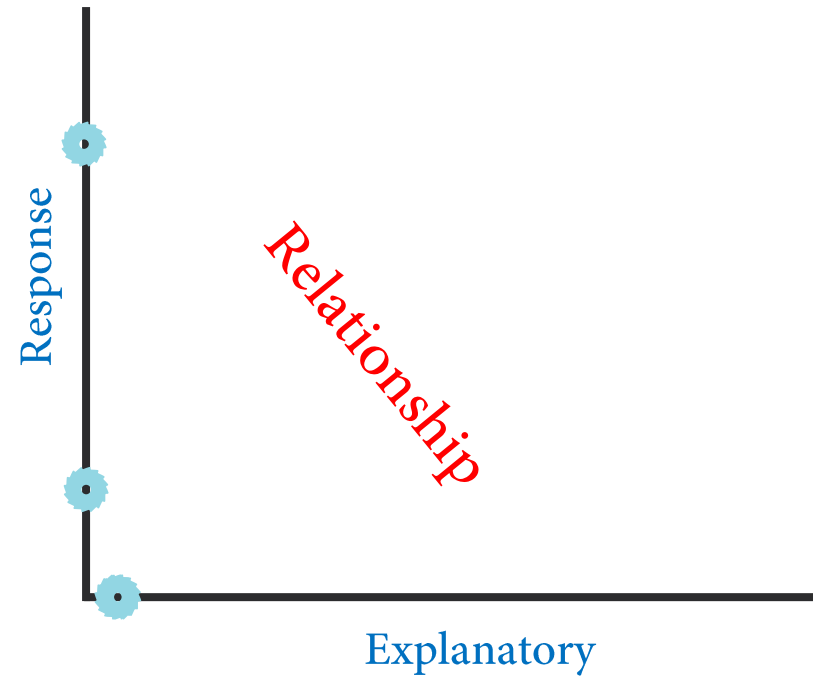
EXPLORATORY DATA ANALYSIS (EDA)

- To extract the relationships between variables or to discover the patterns/distributions of the variables, we need to check on the types of these variables.
- To conduct the data analysis on variables for sake of understanding their relationships, this analysis can be either:
 - Uni-variate analysis
 - Discover the variations of the data into **one** variable
 - Multi-variate analysis
 - Discover the co-variation of **multiple** variables
 - Bi-variate analysis is a special type of this analysis with only **two** variables.

EDA – UNIVARIATE (RECAP)

- There are two types of visualization-based univariate analysis:
 - Variation of a continuous variable
 - Variation of a discrete variable
- Examples of univariate continuous :
 - Histograms, etc.
- Examples of univariate discrete:
 - bar plots, etc.

EDA – BI-VARIATE ANALYSIS



EDA – BI-VARIATE ANALYSIS (2)

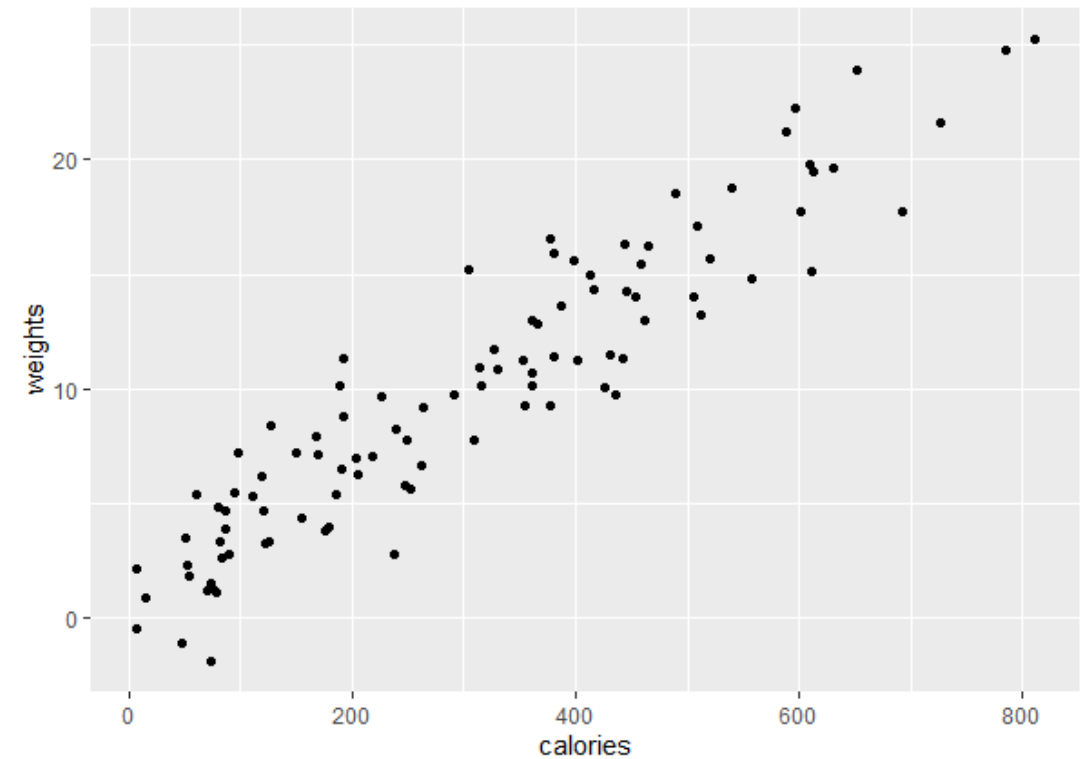
- The bivariate analysis will be done for each of the following pairs:
 - Bivariate analysis of a continuous variable with respect to another continuous variable
 - Bivariate analysis of a continuous variable with respect to a discrete variable
 - Bivariate analysis of a discrete variable with respect to another discrete variable

EDA – BIVARIATE, CONTINUOUS

- The analysis is done to inspect the relationship between two variables, where both of them are ‘numerical’, continuous variables.
- ggplot2 provides many functions to plot the variation between these two variables such as:
 - *geom_point()*, for a scatter plot
 - *geom_jitter()*, for scatter plot with displacing the overlapped points a bit away from each other
 - Etc.

SCATTER PLOTS

- Summarise the relationship between two variables as scatter dots in the cartesian plane.
- A scatter plot can be used to check properties, such as:
 - Correlation between variables;
 - Presence of outliers



BUILDING A SCATTERPLOT

1- a) First layer contains the data specification
b) Also specifies the two variables

2- a) Second layer specifies the type of plot;
the geometric function

3- a) Third and following layers specify axis
labels and further visual cues

```
ggplot(data,aes(x= explanatory, y= response))+
```

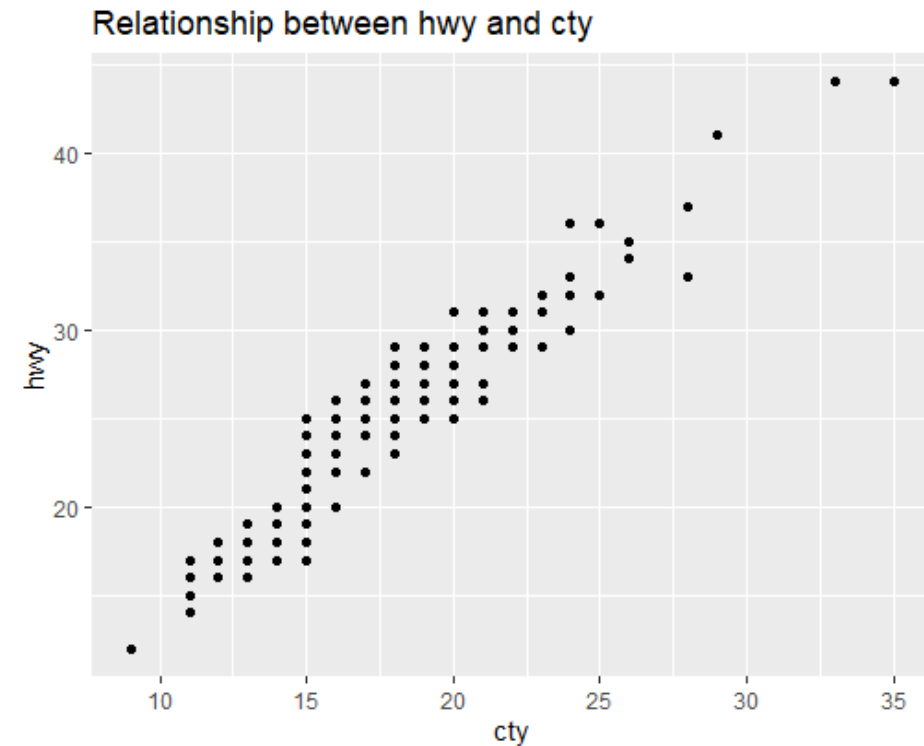
```
geom_point() +
```

```
xlab("x axis label") +  
ylab("y axis label")+  
ggtitle("plot title")
```

EXAMPLE

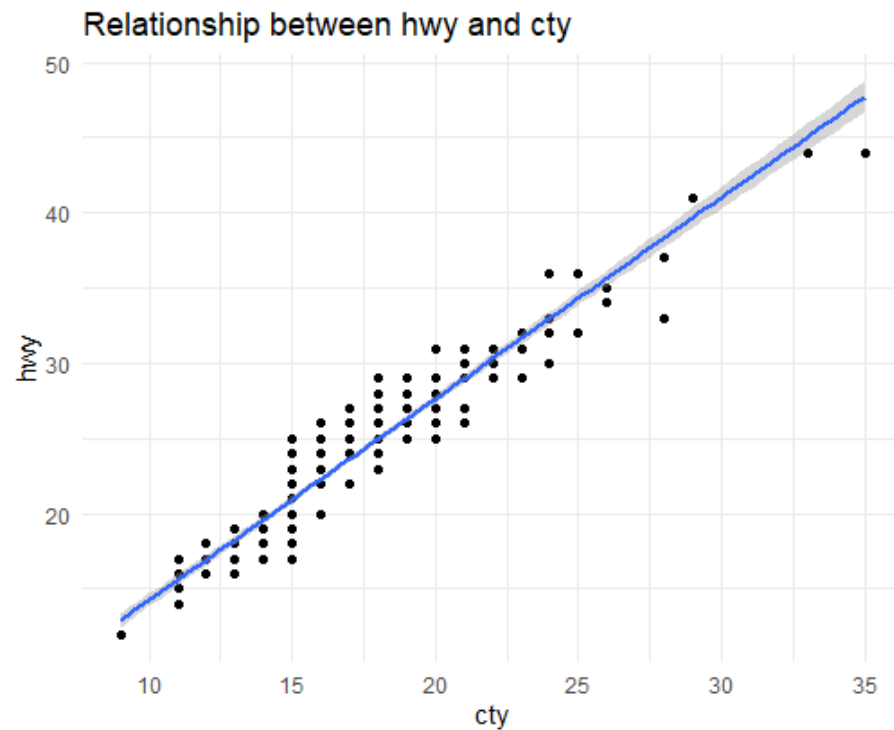
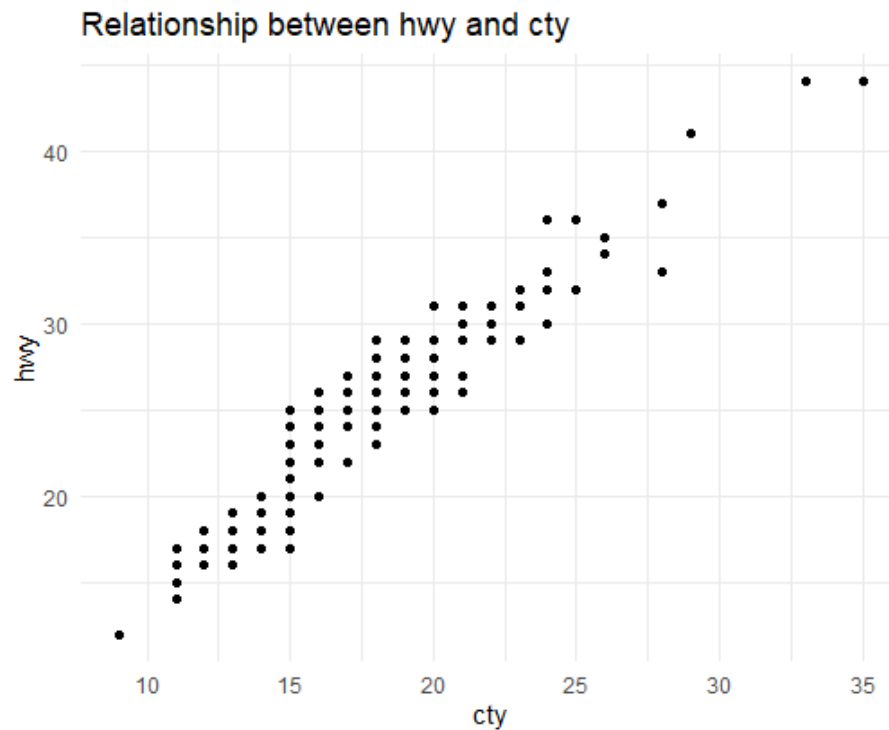
Basic scatterplot visualization

```
ggplot(data= mpg, aes(x= cty, y= hwy)) +  
  geom_point() +  
  theme_minimal() +  
  ggtitle("Relationship between hwy and cty")
```



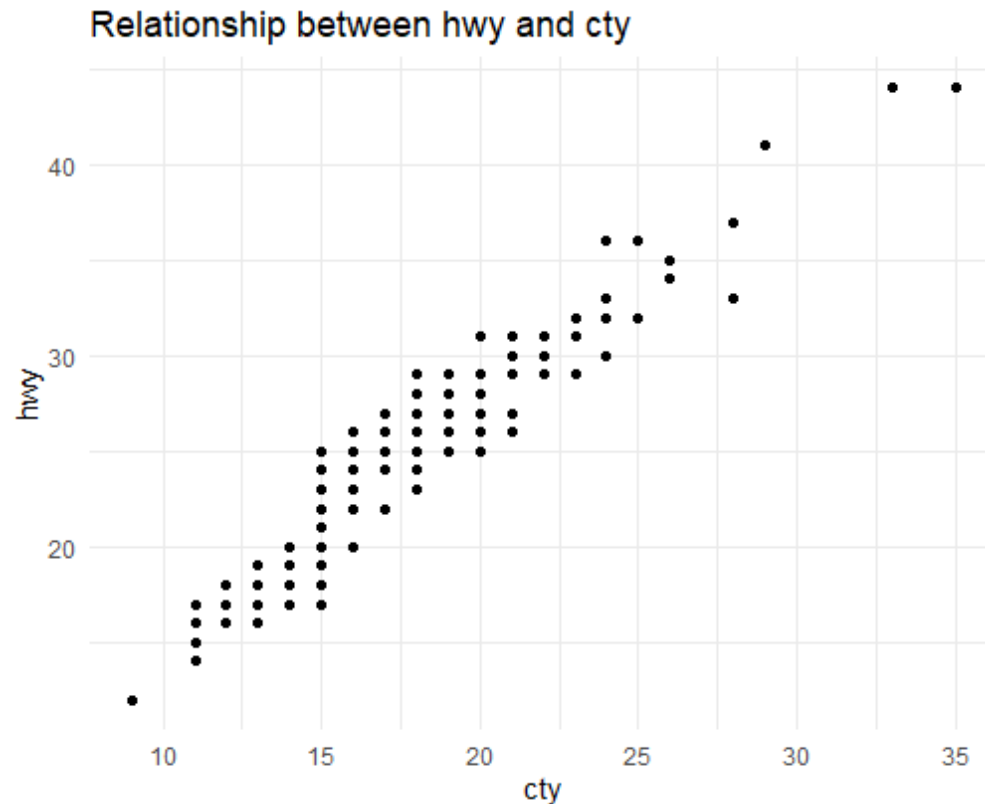
EXAMPLE (2)

Adding extra visual cues and more options



The full code is shared under week 10 on Canvas

SCATTER PLOT INTERPRETATION



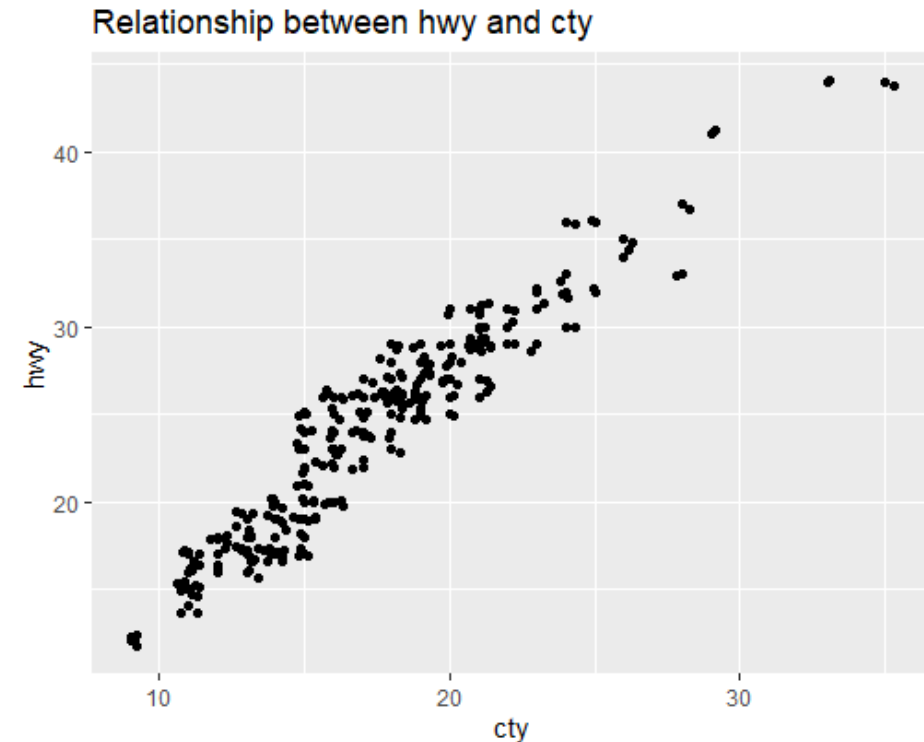
Scatter plot can be used to answer these questions:

- Are the variables related? [**Form**]
- Are they positively or negatively related? [**Direction**]
- What is the strength of the correlation? [**Strength**]
- Are there outliers?

EXAMPLE OF JITTER PLOT

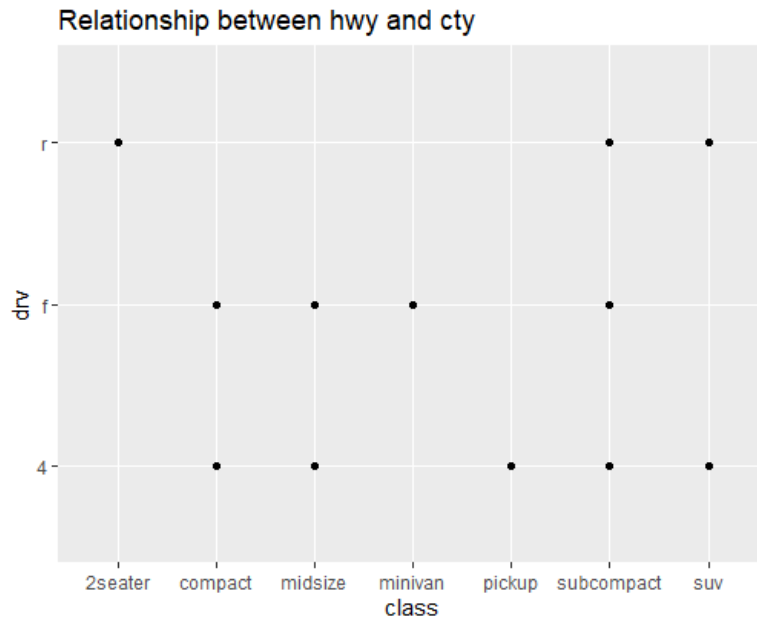
Basic jitterplot visualization

```
ggplot(data= mpg, aes(x= cty, y= hwy)) +  
  geom_jitter() +  
  theme_minimal() +  
  ggtitle("Relationship between hwy and cty")
```

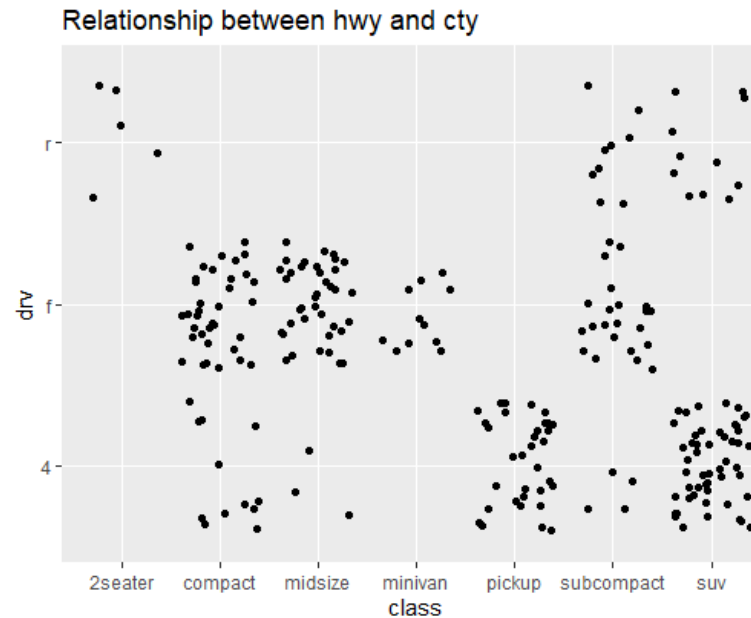


EXAMPLE OF JITTER PLOT (2)

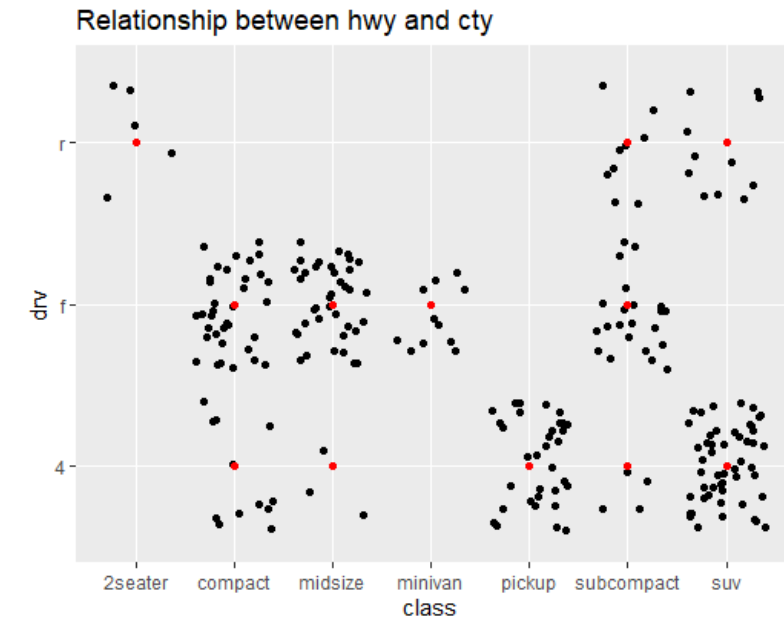
Adding extra visual cues and more options



```
mpg %>%  
  ggplot(aes(class, drv)) +  
  geom_point() +  
  labs(title = "Relationship between class and drv")
```



```
mpg %>%  
  ggplot(aes(class, drv)) +  
  geom_jitter() +  
  labs(title = "Relationship between class and drv")
```



```
mpg %>%  
  ggplot(aes(class, drv)) +  
  geom_point(colour="red") +  
  geom_jitter() +  
  labs(title = "Relationship between class and drv")
```


BIVARIATE, CONTINUOUS & DISCRETE

- The analysis is done to inspect the relationship between two variables, where the explanatory variable is discrete and the response variable is continuous.
- ggplot2 provides some functions to plot this variation such as:
 - *geom_bar()*, for a bar plot
 - *geom_boxplot()*, for visualising the quantiles of response variable
 - Etc.

BAR PLOTS

- Bar plot is usually used to visualize the counts or the frequencies of the categories of the explanatory variable
- In this case, the visualisation is represented by bars
- The response variable is the counts or the frequencies



- Boxplot usually allows us to compare the distribution of a continuous variable across categories of a discrete variable
- It shows the distributions by marking the following values:
 - Minimum
 - Maximum
 - Median
 - IQR: inter Quantile Range
 - 25% percentile
 - 27% percentile

BUILDING A BOXPLOT

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the geometric function

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labels and further visual cues

```
ggplot(data,aes(x= explanatory, y= response))+
```

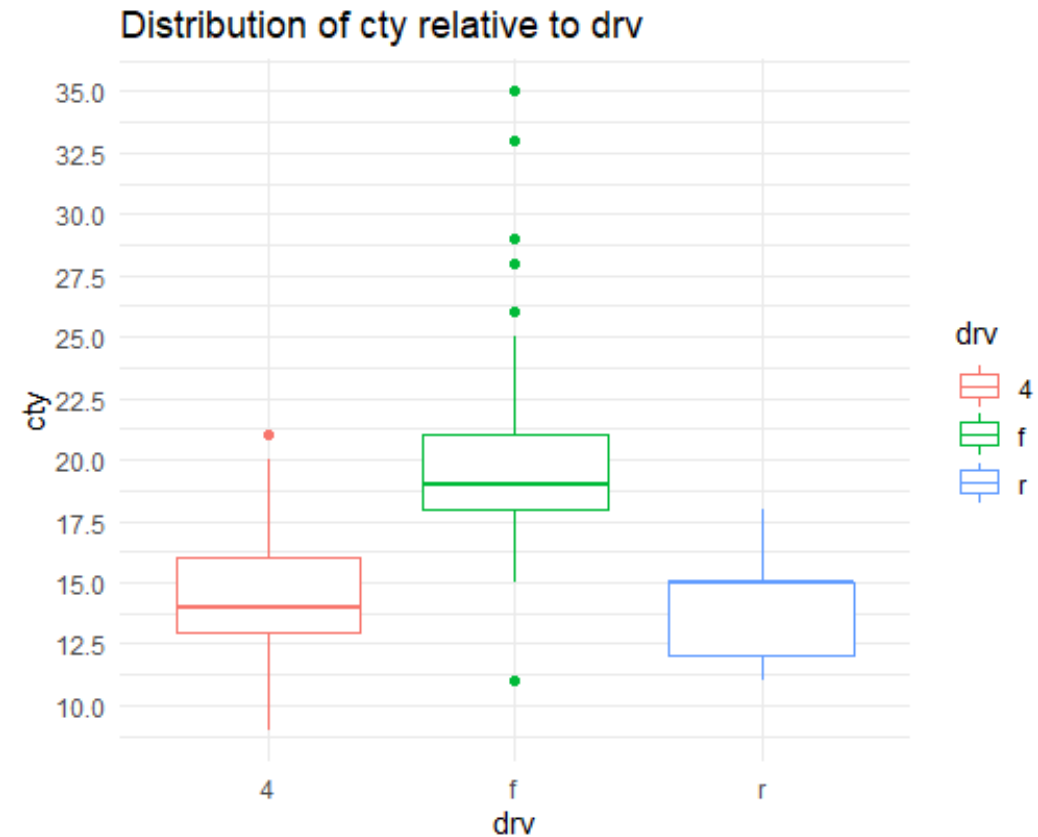
```
geom_boxplot(colour=classification) +
```

```
xlab("x axis label") +  
ylab("y axis label")+  
ggtitle("plot title")
```

EXAMPLE

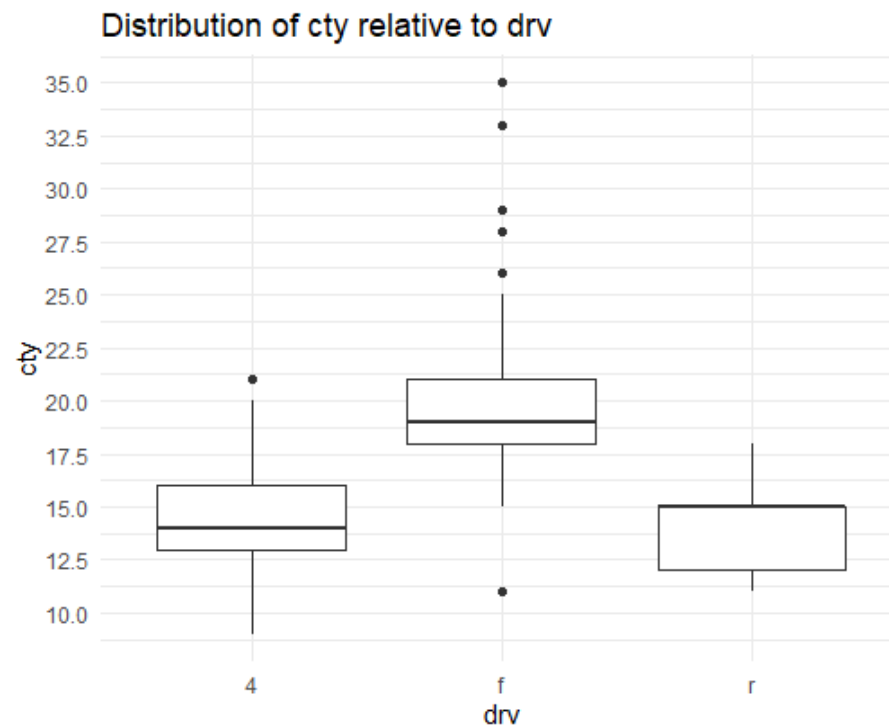
Basic boxplot visualization

```
ggplot(data= mpg, aes(x= drv, y= cty)) +  
  geom_boxplot(colour=) +  
  theme_minimal() +  
  ggtitle("Relationship between drv and cty")
```

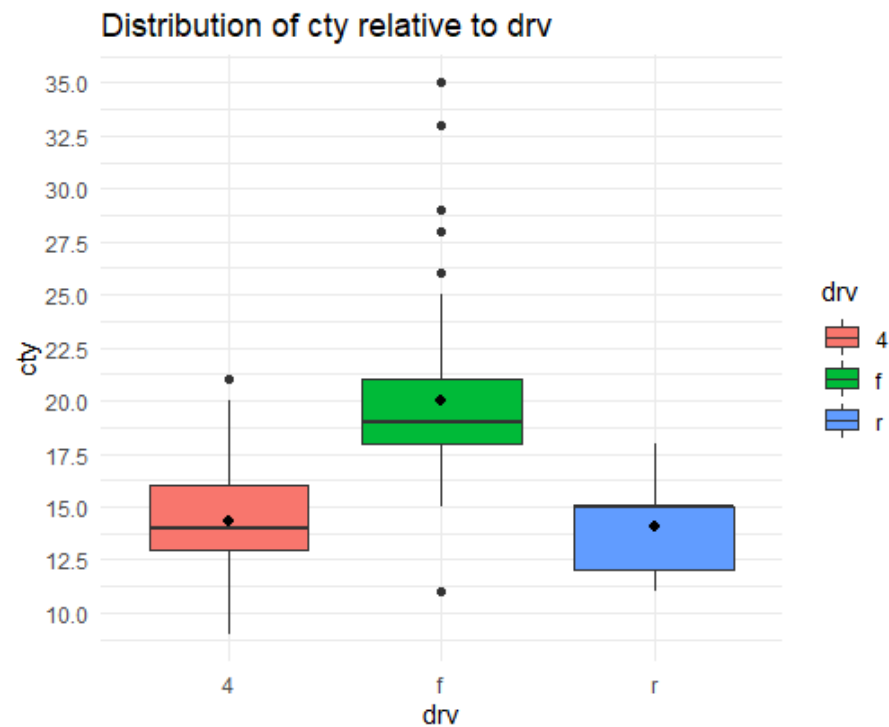


EXAMPLE (2)

Adding extra visual cues and more options



Basic plot

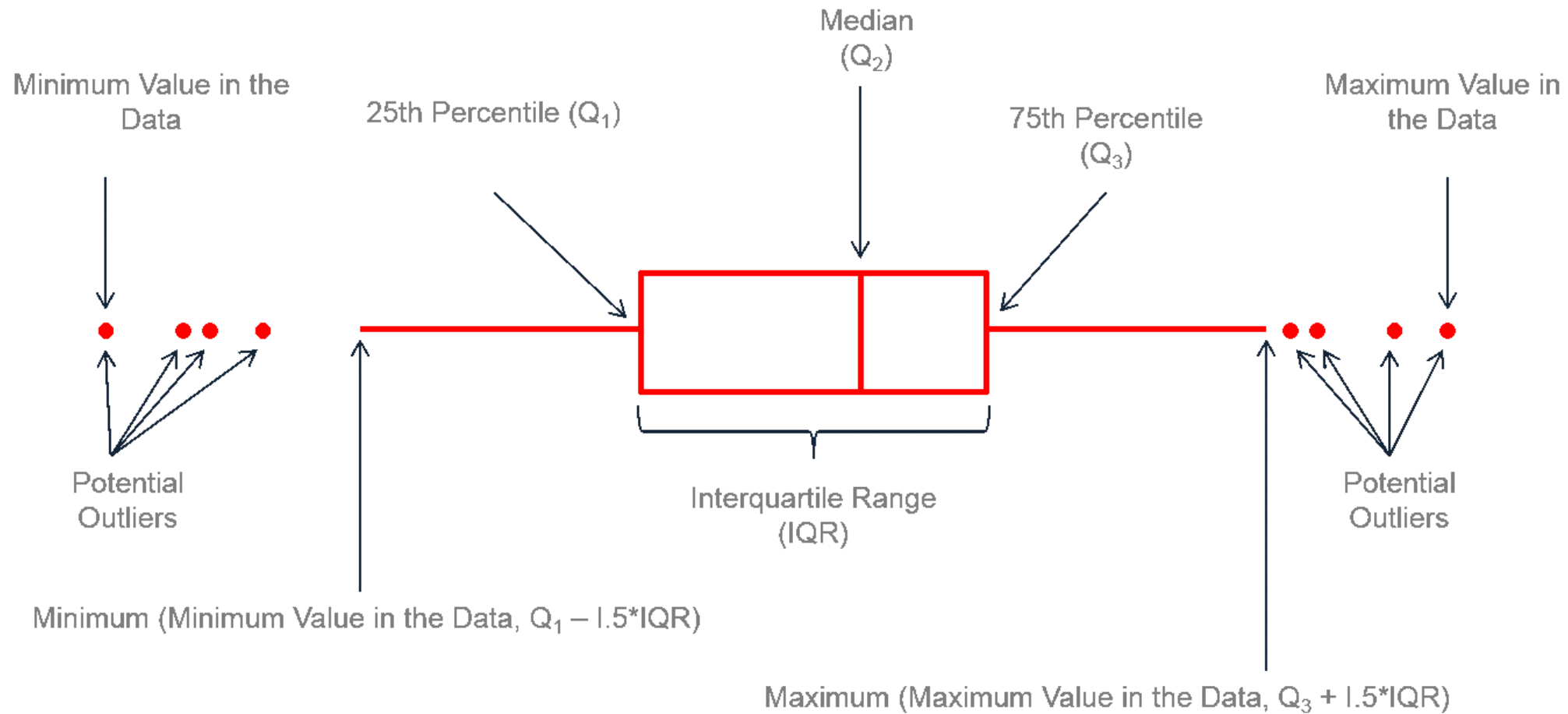


Combined with stat_summary

BOXPLOTS INTERPRETATION

- Bivariate box plots combined with `stat_summary()` can help to understand the distribution of the variables.
- They can also show the distribution of several levels, allowing you to efficiently identify outliers, find patterns and spread of the observations
- Excellent for conveying the variation and particularly detecting and illustrating the variation changes between different levels of data
- Boxplots can help answer the following questions:
 - Is a category significant over the others?
 - Does the center differ between categories?
 - Does the variation differ between categories?
 - Are there any outliers?

BOXPLOTS INTERPRETATION (2)



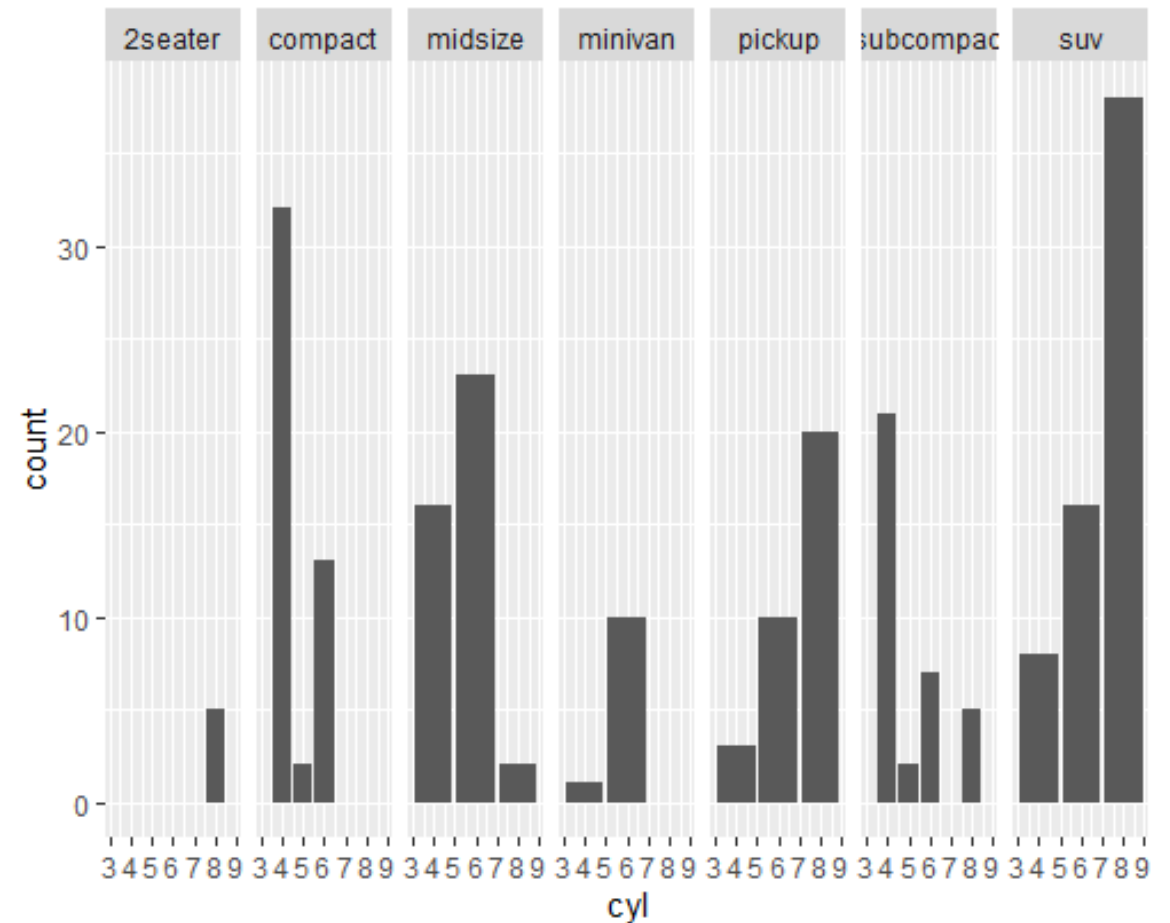
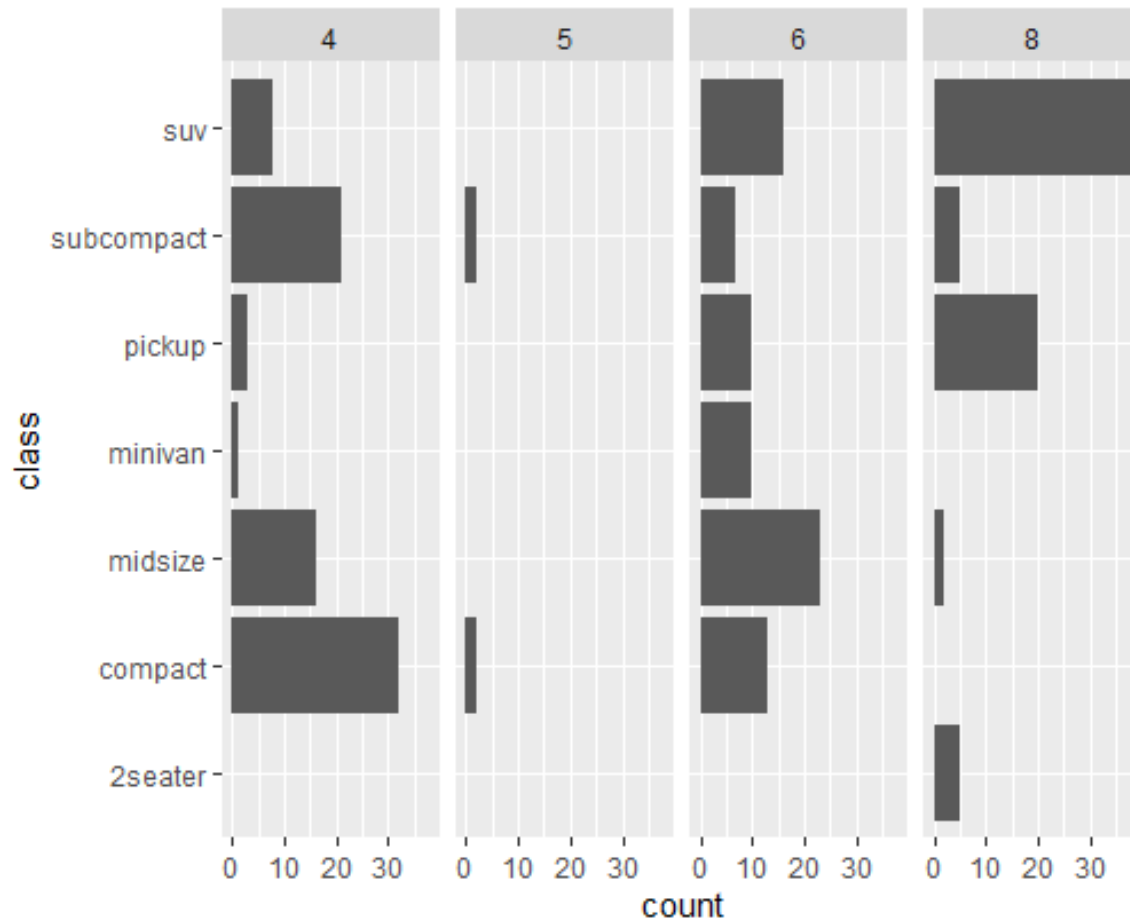
Credit: <https://www.leansigmacorporation.com/box-plot-with-minitab>

BIVARIATE, DISCRETE

- The analysis is done to inspect the relationship between two variables, where both of them are discrete.
- ggplot2 provides some functions to plot this variation such as:
 - *facet + geom_bar()*, for a bar chart for different categories as subplots
 - *geom_bin2d()*

EXAMPLE OF DISCRETE BIVARIATE

Facet + barplot



RECOMMENDED READING

- You are recommended to read chapters 7 from the “*R for Data Science*” book:
 - <https://r4ds.had.co.nz/exploratory-data-analysis.html>
 - https://beta.rstudioconnect.com/content/3350/dplyr_tutorial.html