

# DATA MANIPULATION

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# LEARNING OBJECTIVES

- © Review about cleaning the Data
- © Review the value of Exploratory Data Analysis

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**OPENING**

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# **CLEANING DATA**

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# DATA FORMAT

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## DATA FORMAT

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- ◉ As a result of experiments or observations data is **commonly** presented as:
  - ◉ Rows, one for each case, observation, subject
  - ◉ Columns
    - ◉ Identifiers, subject id; date, time or timestamp
    - ◉ Explanatory variables
    - ◉ Outcome variable (sometimes absent)
- ◉ Stored on paper, spread sheets, databases or text files

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# MISSING VALUES

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## MISSING VALUES

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- ◉ MCAR, Missing Completely at Random
  - ◉ does not depend on the variable of interest or any other variable in the dataset
  - ◉ very rarely found and the best method is to ignore such cases
- ◉ MAR, Missing at Random
  - ◉  $X_i$  is missing at random if missingness does not depend on the value of  $X_i$  after controlling for another variable

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# MISSING VALUES

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## MISSING VALUES

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- ◉ NAMR, Not missing at Random
  - ◉ the missingness mechanism depends on the actual value of missing data
  - ◉ modelling such a condition is a very difficult task to achieve
  - ◉ the only way to attain an estimate of parameters is to model the missingness, meaning to write a model for missing data and then integrate it back
  - ◉ easier said than done

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**OPENING**

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**EXPLORATORY**

**DATA**

**ANALYSIS**

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# ROOT CAUSE FOR EDA

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## ROOT CAUSE FOR EDA

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- ◎ Column of numbers are difficult to read, especially in large volumes, and so determining important characteristics of the data
- ◎ Exploratory Data Analysis techniques have been devised as an aid
- ◎ The techniques work in part by hiding certain aspects of the data while making other aspects more clear

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# MAIN REASONS FOR EDA

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## MAIN REASONS FOR EDA

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- ◉ Detection of mistakes
  - ◉ Checking of assumptions
  - ◉ Preliminary selection of appropriate models
  - ◉ Determining relationships among the explanatory variables
  - ◉ Assessing the direction and rough size of relationships between explanatory and outcome variables
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- ◉ Most of data handling that is not formal statistical modelling and inference can be considered exploratory data analysis

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# EDA CLASSIFICATION

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## EDA CLASSIFICATION

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- ◉ Presentation
  - ◉ Non-Graphical, computation of statistics
  - ◉ Graphical, uses charts, diagrams and visual resources
- ◉ Scope
  - ◉ Univariate, each variable by itself, one at a time
  - ◉ Multivariate (usually Bivariate), look for relationships amongst the variables
- ◉ There are further divisions based on the variable's role (outcome or explanatory) and type (categorical or quantitative)



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## **UNIVARIATE NON-GRAPHICAL EDA**

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## UNIVARIATE NON-GRAPHICAL EDA

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- ◉ The observations or measurements make a sample distribution
- ◉ Useful to understand the population
- ◉ The usual goals of univariate non-graphical EDA
  - ◉ to better appreciate the “sample distribution”
  - ◉ to make some tentative conclusions about what population distribution(s) is/are compatible with the sample distribution
  - ◉ to detect outliers

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## **UNIVARIATE NON-GRAPHICAL EDA - CATEGORICAL DATA**

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## UNIVARIATE NON-GRAPHICAL EDA - CATEGORICAL DATA

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- ◉ The characteristics of interest for a categorical variable are
  - ◉ the range of values
  - ◉ the frequency (or relative frequency) of occurrence for each value
- ◉ For ordinal variables it is sometimes appropriate to treat them as quantitative variables
- ◉ The only useful techniques is some form of tabulation of the frequencies, usually along with calculation of the fraction (or percent) of data that falls in each category

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## **UNIVARIATE NON-GRAPHICAL EDA - CATEGORICAL DATA**

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## UNIVARIATE NON-GRAPHICAL EDA - CATEGORICAL DATA

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- ◉ Losing data is a common mistake and EDA is very helpful for finding mistakes
- ◉ Expect that the proportions add up to 1.00 (or 100%) if the calculations are correct (count/total)

	A	B	C	Other	Total
Count	5	6	4	5	20
Proportion	0.25	0.30	0.20	0.25	1.00
Percent	25%	30%	20%	25%	100%

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## **UNIVARIATE NON-GRAPHICAL EDA - QUANTITATIVE DATA**

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## UNIVARIATE NON-GRAPHICAL EDA - QUANTITATIVE DATA

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- ◉ Characteristics of a quantitative variable of a population distribution
  - ◉ centre
  - ◉ spread
  - ◉ modality (number of peaks in the PDF (Probability Density Function))
  - ◉ shape (including “heaviness of the tails”)
  - ◉ outliers
- ◉ Observed data represent just one sample out of an infinite number of possible samples
  - ◉ The characteristics of a randomly observed sample are not inherently interesting, except to the degree that they represent the population that it came from



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## **BIVARIATE NON-GRAPHICAL EDA**

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## BIVARIATE NON-GRAPHICAL EDA

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- ◉ Categorical data (and quantitative data with only a few different values)
- ◉ Cross-tabulation for two variables
  - ◉ a two-way table with column headings that match the levels of one variable
  - ◉ row headings that match the levels of the other variable
  - ◉ the counts of all subjects that share a pair of levels
- ◉ The two variables might be both explanatory, both outcome or one of each
- ◉ Row percentages (which add to 100% for each row), column percentages (which add to 100% for each column) and cell percentages (which add to 100% over all cells) are also useful

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## **BIVARIATE NON-GRAPHICAL EDA**

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## BIVARIATE NON-GRAPHICAL EDA

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### ◎ Cross-tabulation

	Female	Male	Total
Young	2	3	5
Middle	2	1	3
Old	3	0	3
Total	7	4	11

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# HOW TO MAKE A BAD GRAPH

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## HOW TO MAKE A BAD GRAPH

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- ◉ The aim of good data graphics
  - ◉ Display data accurately and clearly
- ◉ Some rules for displaying data badly
  - ◉ Display as little information as possible
  - ◉ Obscure what you do show (with chart junk)
    - ◉ Use pseudo-3D and colour gratuitously
  - ◉ Make a pie chart (preferably in colour and 3D)
    - ◉ Use a poorly chosen scale

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## EDA RECAP

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## EDA RECAP

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- ◉ In a nutshell
  - ◉ You should always perform appropriate EDA before further analysis of your data
  - ◉ Perform whatever steps are necessary to become more familiar with your data
  - ◉ check for obvious mistakes
  - ◉ learn about variable distributions and learn about relationships between variables
- ◉ EDA is not an exact science - it is a very important art!



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**CONCLUSION**

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# TOPIC REVIEW

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## TOPIC REVIEW

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- ◉ Data Manipulation to fix the data
- ◉ Exploratory Data Analysis to understand the data

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## DATA MANIPULATION

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# Q & A