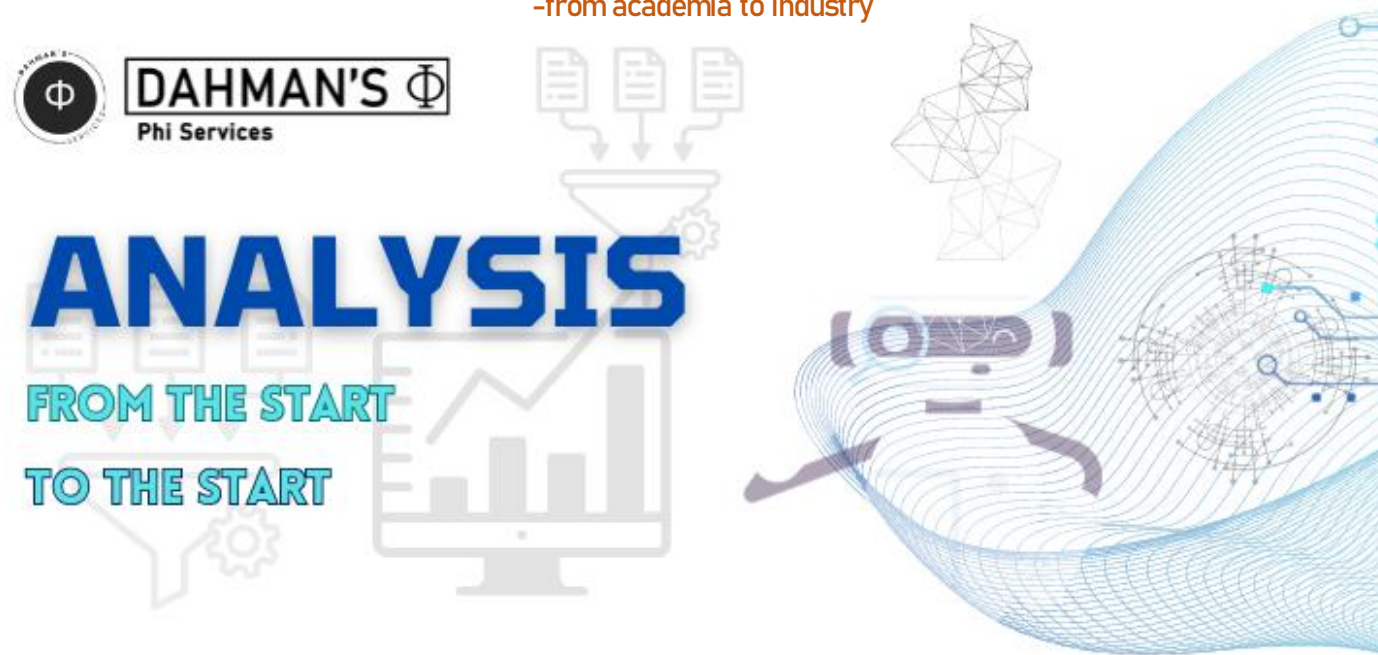




## THE BIG BANG OF DATA SCIENCE

-from academia to industry

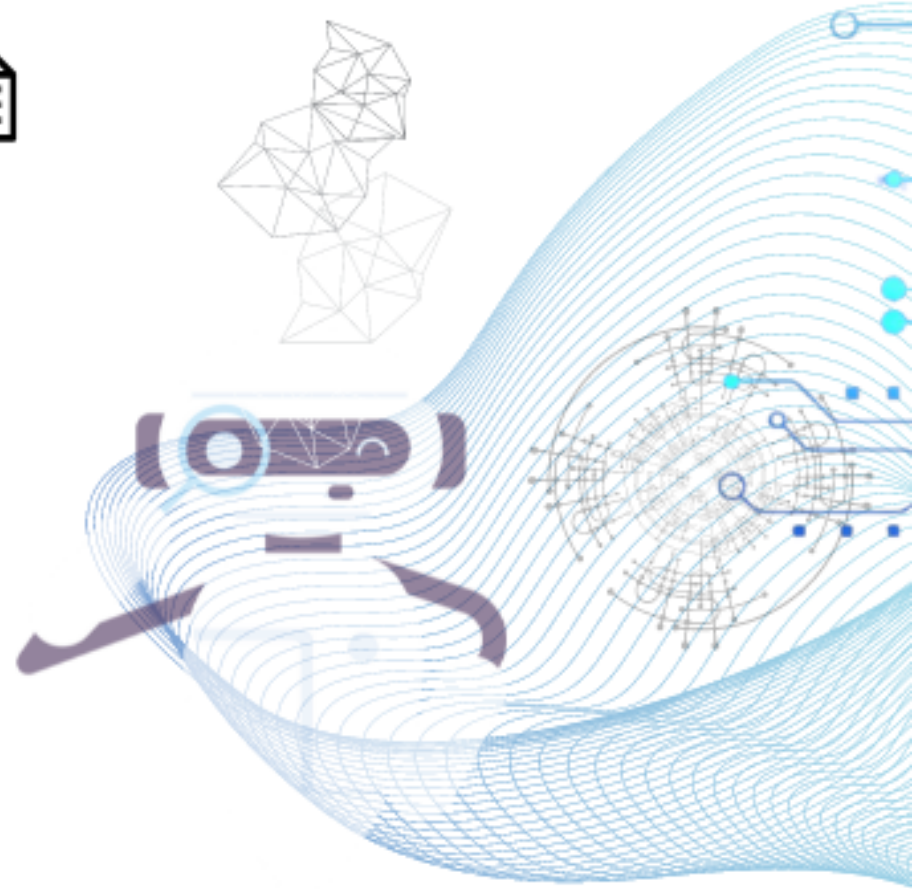
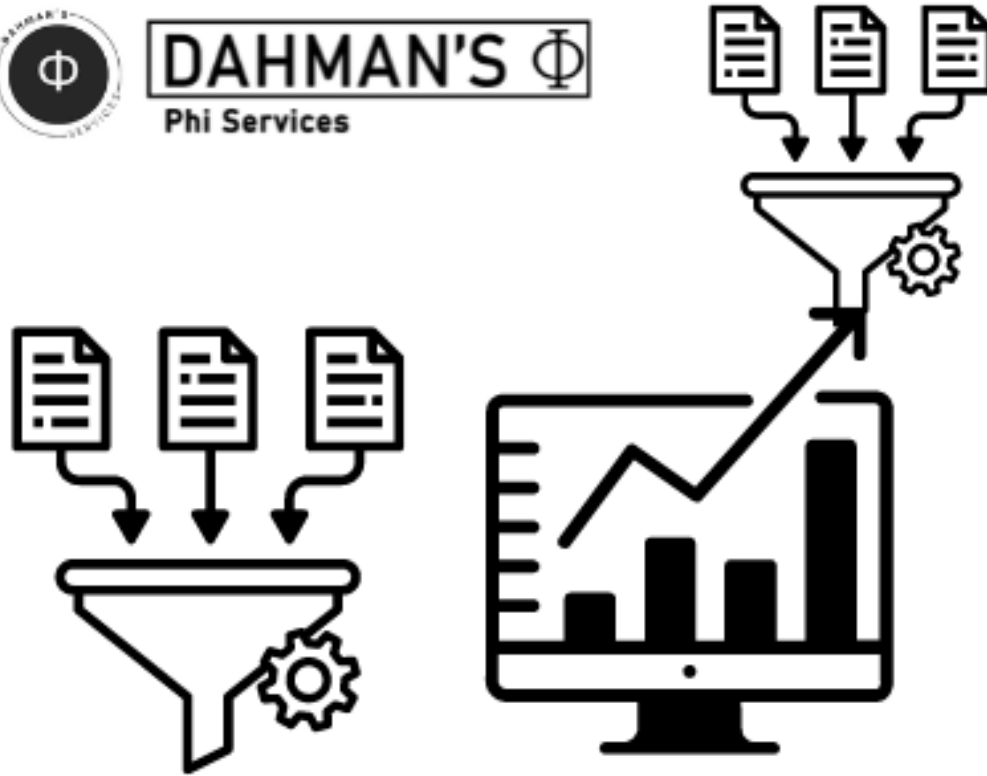


by: Dr. Deniz Dahman's

BOOK TWO



**DAHMAN'S**  $\Phi$   
Phi Services





Chapter One



Chapter Two



Chapter Three



Chapter Four



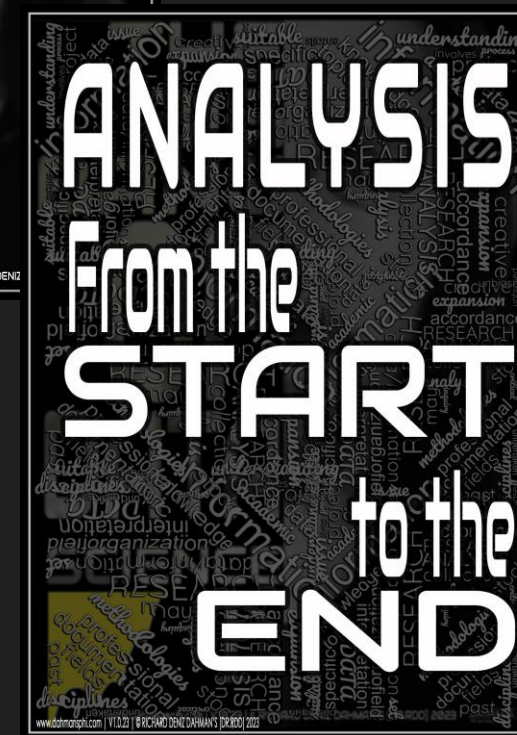
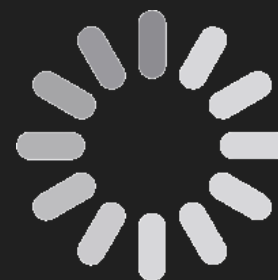
Chapter Five



Chapter Six



Chapter Seven





# Chapter One



## Introduction

Chapter Two



Chapter Three



- ✓ RESEARCH MAP
- ✓ THREATS TO CONCLUSION VALIDITY
- ✓ STATISTICAL POWER
- ✓ IMPOROVE CONCLUSION VALIDITY
- ✓ ANALYSIS

Chapter Four



Chapter Five



Chapter Six



Chapter Seven





Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



Chapter Six



Chapter Seven



## DATA PREPARATION

- ✓ LOGGING THE DATA
- ✓ DATA ACCURACY CONTROL
- ✓ DATABASE STRUCTURE
- ✓ ENTERING DATA TO THE COMPUTER
- ✓ DATA TRANSFORMATION



Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



Chapter Six



Chapter Seven



## Descriptive Statistics

- ✓ Introduction to EDA
- ✓ Distribution
- ✓ Central Tendency
- ✓ Dispersion
- ✓ Bivariate descriptive
- ✓ Multivariate descriptive



Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



Chapter Six



Chapter Seven



## Inferential Statistics

- ✓ Introduction
- ✓ Estimating parameters
- ✓ Hypothesis testing



## STATISTICAL SOFTWARE

- ✓ Introduction
- ✓ Statistical software
- ✓ LAB implementation-  
by JAMOVl

Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



Chapter Six



Chapter Seven







Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



Chapter Six



Chapter Seven



## LAB-SECTION 03

- ✓ Review
- ✓ EDA analysis
- ✓ Inferential Analysis



Chapter One



Chapter Two



Chapter Three



Chapter Four



Chapter Five



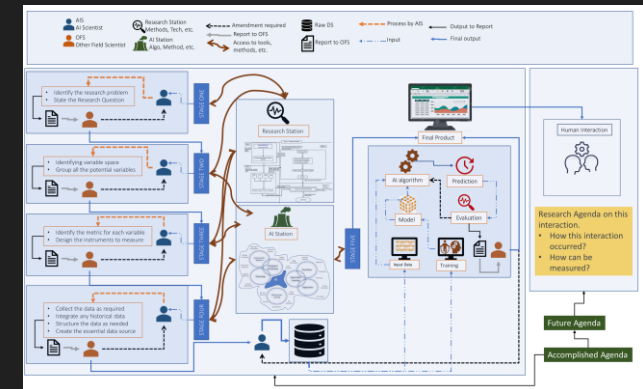
Chapter Six



Chapter Seven



## THE MAP





# Chapter One

## INTRODUCTION

✓ RESEARCH MAP



✓ THREATS TO CONCLUSION VALIDITY



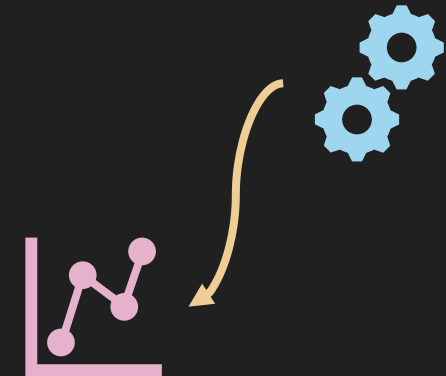
✓ STATISTICAL POWER



✓ IMPOROVE CONCLUSION VALIDITY



✓ ANALYSIS



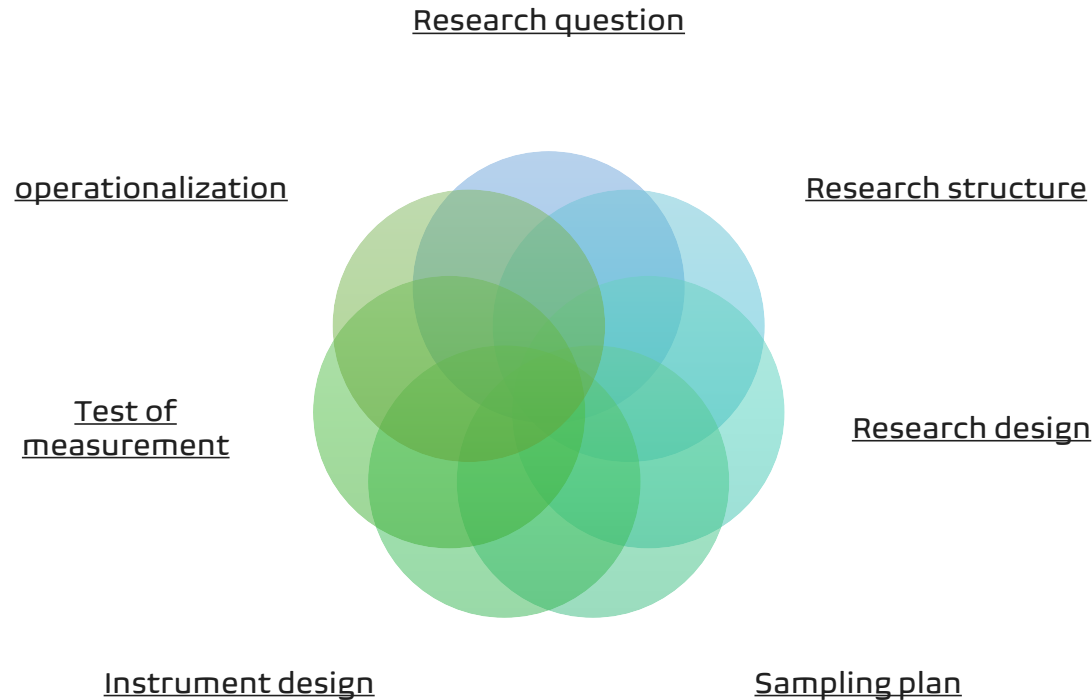


# RESEARCH MAP

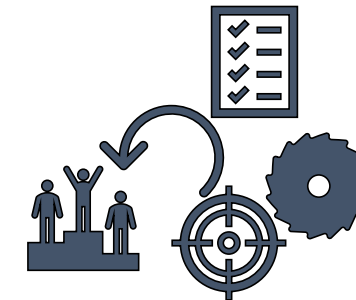


- ✓ The research from the start to the end

- Research map
- Threats to conclusion validity
- Statistical power
- Improve conclusion validity
- analysis



1	3	alfa-romero	gas	std	two	convertible	rwd	front	88.6	168.8	64.1	48.8	2548
2	3	alfa-romero	gas	std	two	convertible	rwd	front	88.6	168.8	64.1	48.8	2548
3	1	alfa-romero	gas	std	two	hatchback	rwd	front	94.5	171.2	65.5	52.4	2823
4	2	audi 100 ls	gas	std	four	sedan	fwd	front	99.8	176.6	66.2	54.3	2337
5	2	audi 100ls	gas	std	four	sedan	4wd	front	99.4	176.6	66.4	54.3	2824
6	2	audi fox	gas	std	two	sedan	fwd	front	99.8	177.3	66.3	53.1	2507
7	1	audi 100ls	gas	std	four	sedan	fwd	front	105.8	192.7	71.4	55.7	2844
8	1	audi 5000	gas	std	four	wagon	fwd	front	105.8	192.7	71.4	55.7	2954
9	1	audi 4000	gas	turbo	four	sedan	fwd	front	105.8	192.7	71.4	55.9	3086
10	0	audi 5000s (	gas	turbo	two	hatchback	4wd	front	99.5	178.2	67.9	52	3053
11	2	bmw 320i	gas	std	two	sedan	rwd	front	101.2	176.8	64.8	54.3	2395
12	0	bmw 320i	gas	std	four	sedan	rwd	front	101.2	176.8	64.8	54.3	2395
13	0	bmw x1	gas	std	two	sedan	rwd	front	101.2	176.8	64.8	54.3	2710
14	0	bmw x3	gas	std	four	sedan	rwd	front	101.2	176.8	64.8	54.3	2765
15	1	bmw z4	gas	std	four	sedan	rwd	front	103.5	189	66.9	55.7	3055
16	0	bmw x4	gas	std	four	sedan	rwd	front	103.5	189	66.9	55.7	3230
17	0	bmw x5	gas	std	two	sedan	rwd	front	103.5	193.8	67.9	53.7	3380
18	0	bmw x3	gas	std	four	sedan	rwd	front	110	197	70.9	56.3	3505





# Threats to conclusion validity

- ✓ Finding **no relationship** when there is one (or, “missing the needle in the haystack”)
- ✓ Finding a **relationship** when there is not one (or “seeing things that aren’t there”)
- ✓ Problems that can lead to either conclusion error



- Research map
- Threats to conclusion validity
- Statistical power
- Improve conclusion validity
- analysis

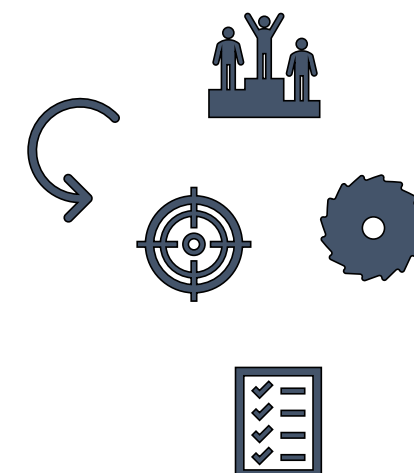
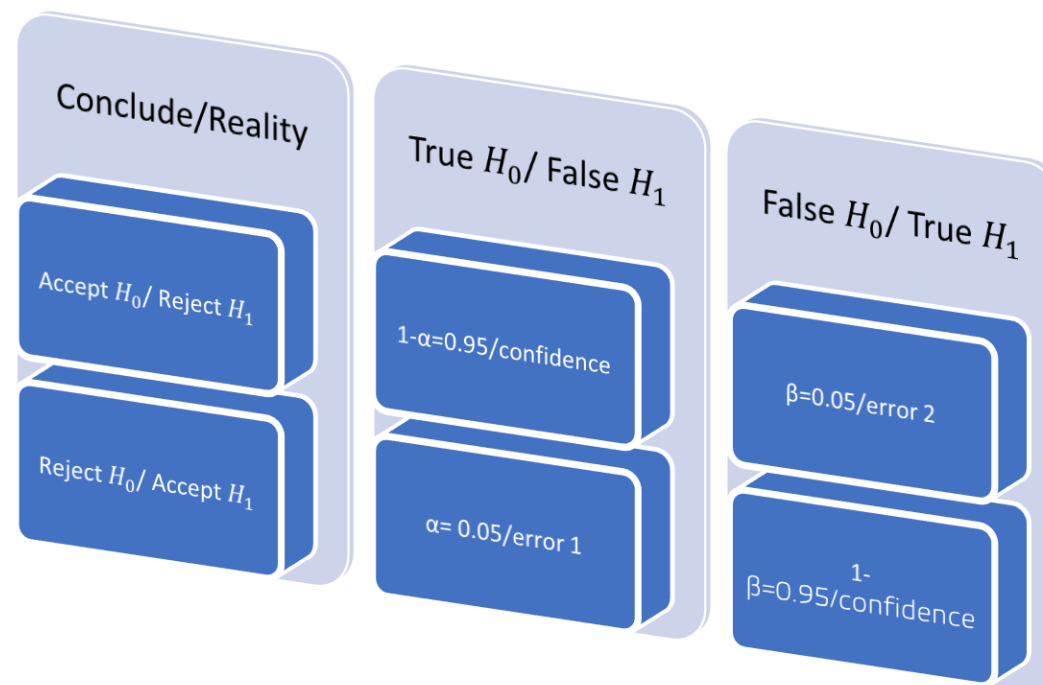




# Statistical power

- Research map
- Threats to conclusion validity
- Statistical power
- Improve conclusion validity
- analysis

- ✓ The four components (sample size, statistical significance, confidence interval, effect size, alpha level  $\alpha$  AND beta level  $\beta$ )
- ✓ The decision matrix:

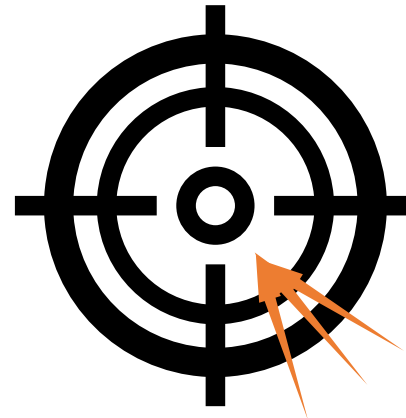
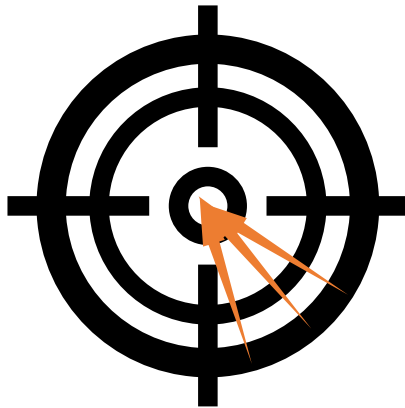




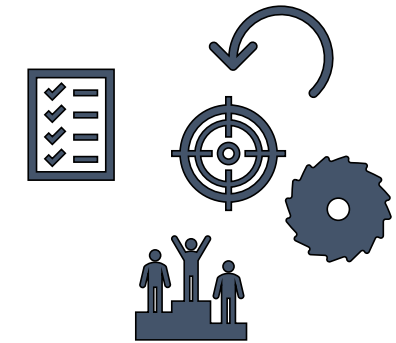
# Improve conclusion validity



- ✓ Good statistical power
- ✓ Good reliability
- ✓ Good implementation



- Research map
- Threats to conclusion validity
- Statistical power
- Improve conclusion validity
- analysis



# analysis



- ✓ Data preparation
- ✓ Descriptive analysis
- ✓ Inferential analysis

- Research map
- Threats to conclusion validity
- Statistical power
- Improve conclusion validity analysis



PREDICTION  
probability regression










## Chapter Two

# DATA PREPARATION

---

- ✓ LOGGING THE DATA 
- ✓ DATA ACCURACY CONTROL 
- ✓ DATABASE STRUCTURE 
- ✓ ENTERING DATA TO THE COMPUTER 
- ✓ DATA TRANSFORMATION 





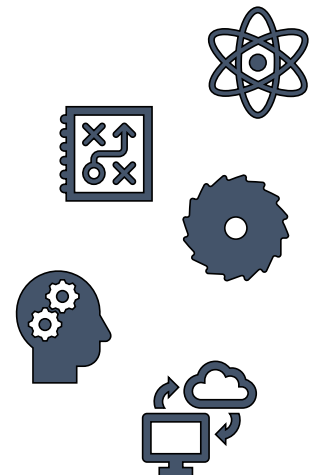
# LOGGING THE DATA



- ✓ In any research project you may have data coming from a number of different sources at different times:
  - mail surveys returns
  - coded interview data
  - pretest or posttest data
  - observational data
- ✓ you need to set up a procedure for logging the information and keeping track of it until you are ready to do a comprehensive data analysis
- ✓ You could do this with any standard computerized database program (e.g. MS Access), or standard statistical programs (e.g. SPSS)
- ✓ Most professional researchers will retain such records for at least 5-7 years
- ✓ For important or expensive studies, the original data might be stored in a data archive



- Logging the data
- Data Accuracy Control
- Database Structure
- Entering Data to Computer
- Data transformation



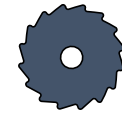
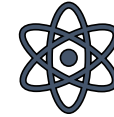


# Data Accuracy Control

- ✓ Questions you may consider while you screen the data for accuracy:
  - Are the response readable/legible?
  - Are all important questions answered?
  - Are the responses complete?
  - Is all relevant contextual information included (e.g. data, time, place, researcher)?
- ✓ Benefits for such approach:
  - doing this right away will allow you to go back to the sample to clarify any problems or errors
  - As most research stress on the fact that quality of measurement is major issue, Assuring that the data collection process does not contribute inaccuracies will help assure the overall quality of subsequent analyses



- Logging the data
- Data Accuracy Control
- Database Structure
- Entering Data to Computer
- Data transformation





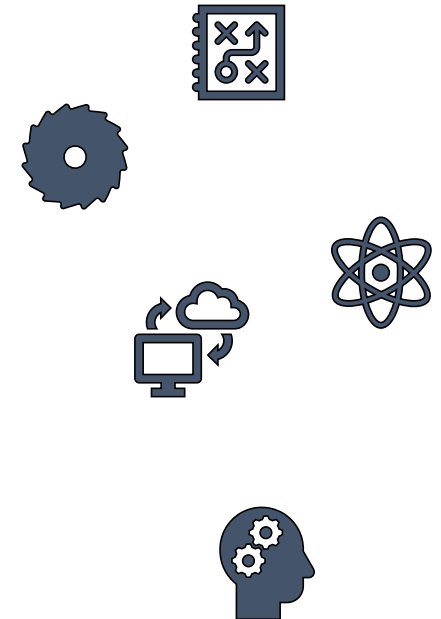
# Database Structure



- ✓ manner in which you intend to store the data for the study
- ✓ in large complex studies, you might have one structure for logging data and another for storing it
- ✓ you should generate a printed **codebook** that describes the data and indicates where and how it can be accessed
- ✓ codebook should include the following items for each variable:
  - variable name
  - variable description
  - variable format (number, data, text)
  - instrument/method of collection
  - date collected
  - respondent or group
  - variable location (in database)
  - Notes
- ✓ codebook together with the database, provide comprehensive documentation for further analysis you need no additional information



- Logging the data
- Data Accuracy Control
- Database Structure
- Entering Data to Computer
- Data transformation



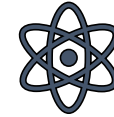


# Entering Data to Computer

- ✓ Simply to enter your data to the computer, available approaches to control the accuracy could be
- ✓ Double entry approach: enter the data once, and then second time, automated app will discover if any discrepancy
- ✓ Procedural entry approach: enter the data once, and then set up procedure for checking the data for accuracy, e.g. available software controls that by summary of data, e.g. age is 606, or maybe 7 is entered when the expected value is only 1 to 5



- Logging the data
- Data Accuracy Control
- Database Structure
- Entering Data to Computer
- Data transformation





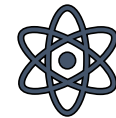
# Data transformation



- ✓ it is almost always necessary to transform the raw data into variables that are usable in the analyses
- ✓ Some of the more common are:
  - Item reversal
  - Scale total
  - Categories



- Logging the data
- Data Accuracy Control
- Database Structure
- Entering Data to Computer
- Data transformation





## Chapter Three

# Descriptive Analysis

✓ Introduction to EDA



✓ Distribution



✓ Central Tendency



✓ Dispersion



✓ Bivariate descriptive



✓ Multivariate descriptive





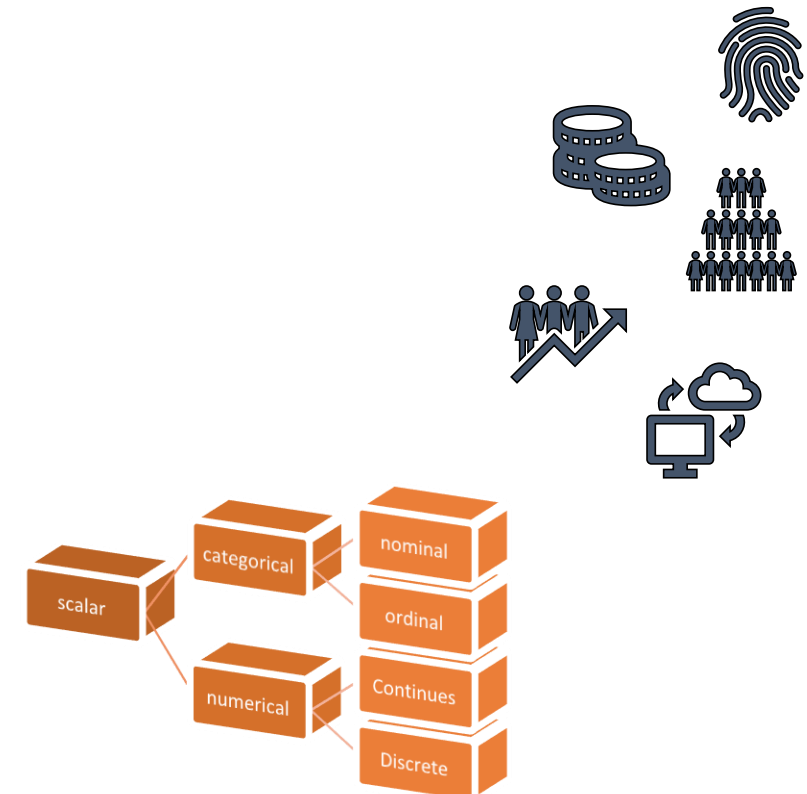
# Introduction to EDA

## DESCRIPTIVE=STATISTICS

PROBABILITY HAS NOTHING TO DO IN HERE

- ✓ To answer research question:
  - The variable
- ✓ To quantify the variable:
  - The value (scalar)/ outcome
- ✓ many subjects of ONE variable on that scalar:
  - The univariate (vector): to analyze it:
    - (1) distribution, (2) central tendency, (3) dispersion
- ✓ To analyze two variable scalars/vectors:
  - The bivariate (column)
- ✓ To analyze more than two variables:
  - The multivariate (matrix)

- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive





# Distribution

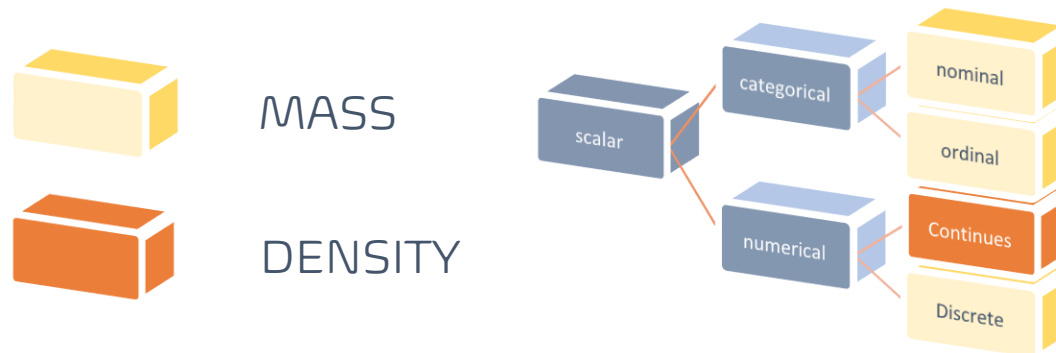
- ✓ The distribution is a summary of the frequency of individual values or ranges of values for a variable

- ✓ Format of distribution:

Table distribution (frequency distribution)

Graph distribution (counting distribution)

- ✓ Explore distribution in terms of the scalar:



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

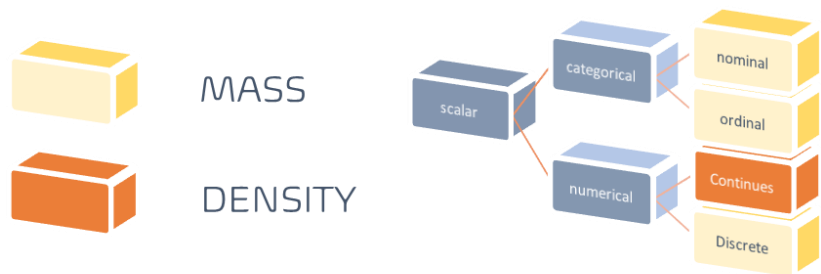
- Introductory
- Mass distribution
- Density distribution





# Distribution

- ✓ Illustration of the (variable outcome in form of scalar or value where it's mass)
- ✓ illustrate the number of fruits in the fridge that might only have (apple, banana, orange)
- ✓ Common mass distribution graph:
  - Bar
  - Pie
  - customized
- ✓ Distribution Table



- Introduction to EDA
  - Distribution
  - Central Tendency
  - Dispersion
  - Bivariate descriptive
  - Multivariate descriptive
- Introductory
  - Mass distribution
  - Density distribution

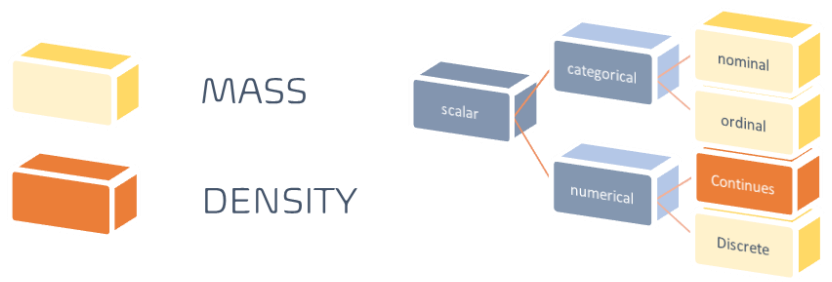




# Distribution

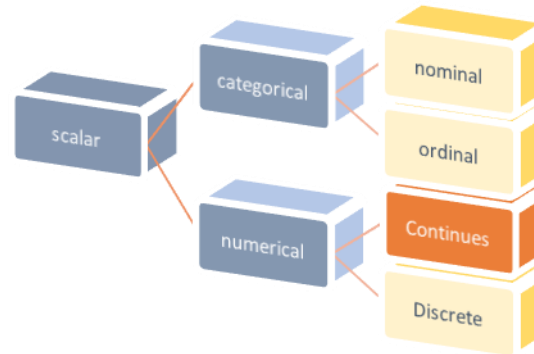
- ✓ Illustration of the (variable outcome in form of scalar or value when it's density)
- ✓ illustrate the element weight in gram
- ✓ Common mass distribution graph:
  - Histogram
  - customized
- ✓ Distribution table

- Introduction to EDA
  - Distribution
  - Central Tendency
  - Dispersion
  - Bivariate descriptive
  - Multivariate descriptive
- Introductory
  - Mass distribution
  - Density distribution



# Central Tendency

- ✓ Illustration of the (average/expectation)
- ✓ Illustration on the center point
- ✓ Common calculations on the center tendency:
  - Mean
  - Median
  - Mood
  - Proportion
- ✓ Story of outliers



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Mean
- Median
- Mood
- Proportion





# Central Tendency

- ✓ Illustration of the (mean)
- ✓ Illustration on the difference of symbols ( $\mu$ ;  $\lambda$ )
- ✓ To find the mean, simply add up all response values and divide the sum by the total number of responses. The total number of responses or observations is called N.
- ✓ Graphical illustration



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Mean
- Median
- Mood
- Proportion





# Central Tendency

- ✓ Illustration of the (median)
- ✓ The median is the value that's exactly in the middle
- ✓ To find the median, order each response value from the smallest to the biggest. Then, the median is the number in the middle. If there are two numbers in the middle, find their mean.
- ✓ Graphical illustration



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Mean
- Median
- Mood
- Proportion





# Central Tendency

- ✓ Illustration of the (mode)
- ✓ the most popular or most frequent response value
- ✓ To find the mode, order your data set from lowest to highest and find the response that occurs most frequently.
- ✓ Graphical illustration



## Chapter Three DISRIPTIVE ANALYSIS

- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Mean
- Median
- Mode
- Proportion





# Central Tendency

- ✓ Illustration of the **proportion**
- ✓ A population proportion is a fraction of the population that has a certain characteristic
- ✓ Its relation to ordinal data.
- ✓ Graphical illustration



## Chapter Three DISRIPTIVE ANALYSIS

- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Mean
- Median
- Mode
- Proportion

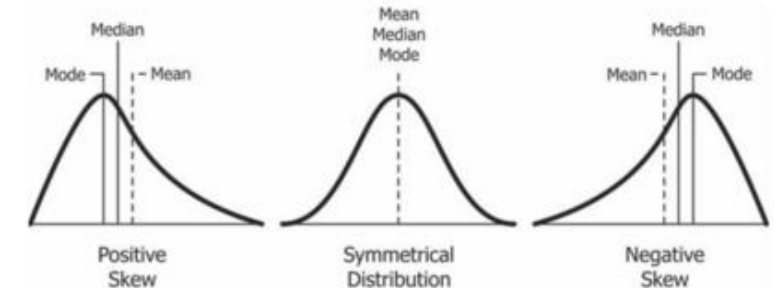
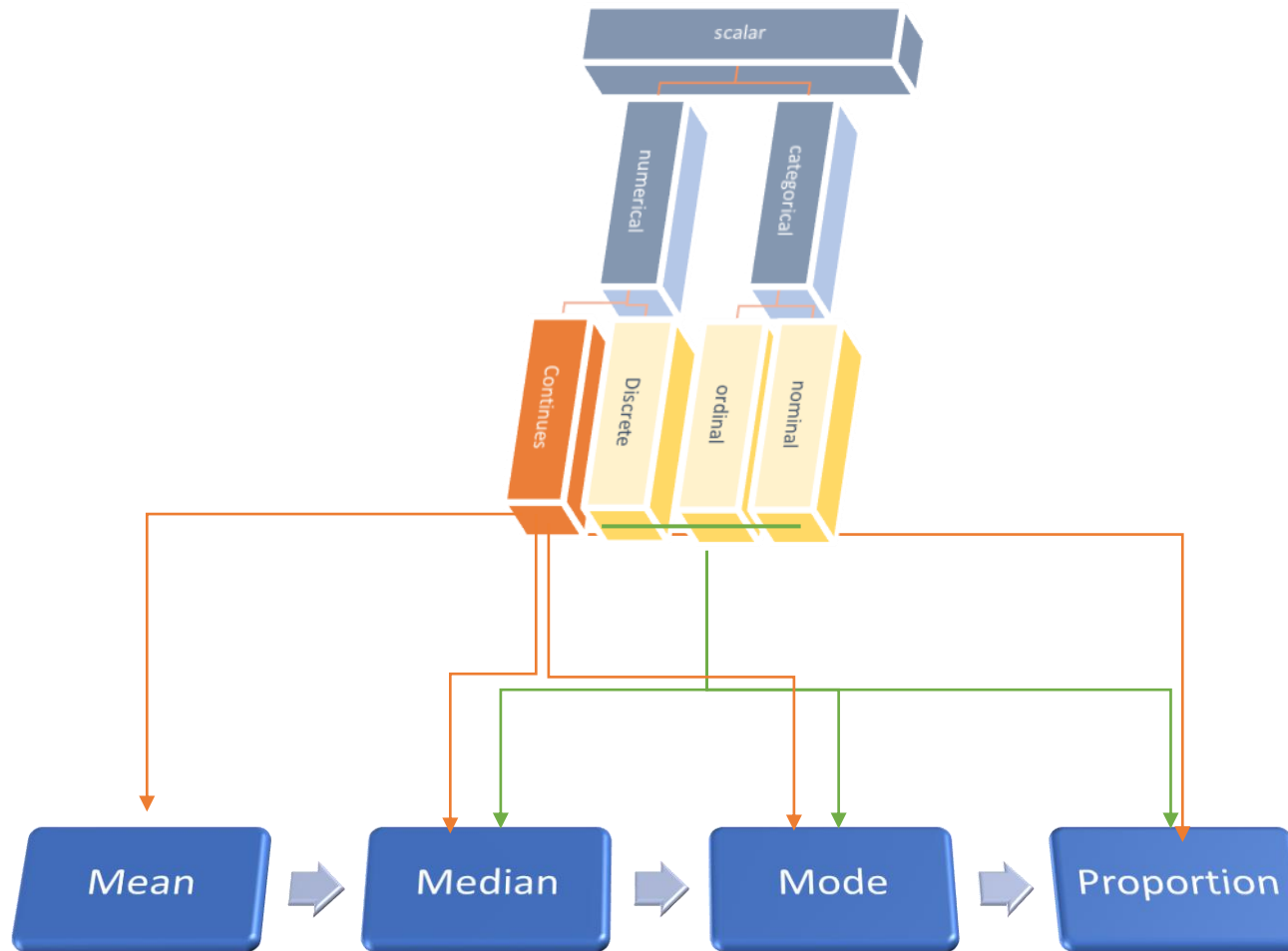




# Central Tendency



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive





# Dispersion



- ✓ Illustration of the (dispersion)
- ✓ Variability describes how far apart data points lie from each other and from the center of a distribution
- ✓ Variability is also referred to as spread, scatter or dispersion
- ✓ Tools:
  - Range
  - Interquartile
  - Variance
  - Standard deviation



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Range
- Interquartile
- Variance
- Standard deviation





# Dispersion

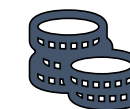


- ✓ Illustration of the **range**
- ✓ The range tells you the spread of your data from the lowest to the highest value
- ✓ To find the range, simply **subtract** the lowest value from the highest value in the data set.
- ✓ Because only 2 numbers are used, the range is influenced by **outliers** and **doesn't give you any information** about the distribution of values.
- ✓ It's best used in combination with other measures



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Range
- Interquartile
- Variance
- Standard deviation





# Dispersion



- ✓ Illustration of the (interquartile range)
- ✓ The interquartile range gives you the spread of the middle of your distribution
- ✓ The interquartile range is the third quartile (Q3) minus the first quartile (Q1). This gives us the range of the middle half of a data set
- ✓ the interquartile range uses only 2 values in its calculation. But the IQR is less affected by outliers.
- ✓ The Five-number summary:
  - Lowest value
  - Q1: 25th percentile
  - Q2: the median
  - Q3: 75th percentile
  - Highest value



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Range
- Interquartile
- Variance
- Standard deviation





# Dispersion

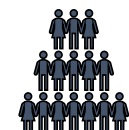
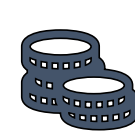


- ✓ Illustration of the (variance)
- ✓ A deviation from the mean is how far a score lies from the mean
- ✓ Variance reflects the degree of spread in the data set. The more spread the data, the larger the variance is in relation to the mean
- ✓ it's harder to interpret the variance number intuitively, it's important to calculate variance for comparing different data sets in statistical tests like ANOVAs



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Range
- Interquartile
- Variance
- Standard deviation





# Dispersion



- ✓ Illustration of the (standard deviation)
- ✓ The standard deviation is the average amount of variability in your dataset
- ✓ It tells you, on average, how far each score lies from the mean. The larger the standard deviation, the more variable the data set is
- ✓ The difference between biased and conservative estimates of standard deviation gets much smaller when you have a large sample size



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive

- Introductory
- Range
- Interquartile
- Variance
- Standard deviation





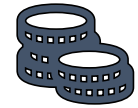
# Bivariate descriptive

- ✓ Illustration of the (bivariate descriptive)
- ✓ study the relationship that exists between two variables

**Correlation Coefficients** – The coefficient lets you know if the data in question are related. When the correlation coefficient is **zero**, the variables are **not related**. If the correlation coefficient is a **positive or a negative 1**, then this means that the variables are **perfectly correlated**

- ✓ Tools:
  - **Scatter plots** – This gives an idea of the patterns that can be formed using the two variables
  - **contingency table** displays frequencies for combinations of two categorical variables.

- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive



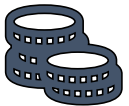


# Multivariate descriptive

- ✓ Illustration of the (multivariate descriptive)
- ✓ involves analyzing relationships between more than two variables
- ✓ Main Tools:
  - All what you know about uni-bi variate
  - covariance and correlation matrices.



- Introduction to EDA
- Distribution
- Central Tendency
- Dispersion
- Bivariate descriptive
- Multivariate descriptive







## Chapter Four

# Inferential Analysis

---

✓ Introduction



✓ Estimating parameters



✓ Hypothesis testing





# Introduction

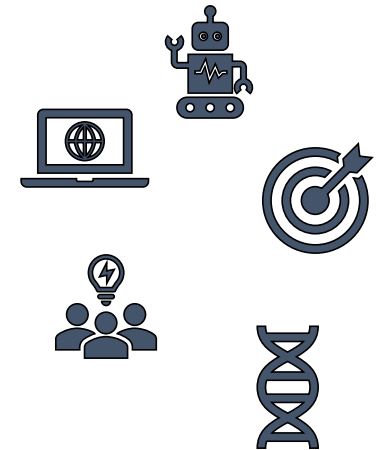


- Introduction
- Estimating parameters
- Hypothesis testing

- ✓ Illustration of the (inferential descriptive)
- ✓ While descriptive statistics can only summarize a sample's characteristics, inferential statistics use your sample to make reasonable guesses about the larger population
- ✓ With inferential statistics, it's important to use random and unbiased sampling methods

characteristics of samples and populations ( $\mu$  &  $\sigma$ ):

- A statistic is a measure that describes the sample (e.g., sample mean).
- A parameter is a measure that describes the whole population (e.g., population mean)





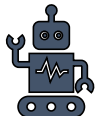
# Estimating parameters

- ✓ This means taking a statistic from your sample data ( $n\hat{y}_{sp}$ ) and using it to say something about a population parameter ( $N\mu\sigma P$ )
- ✓ Steps:
  - Select random and representative sample
  - Collect info from the sample
  - Calculate the sample statistics of interest
  - Assign value to the population parameter



- Introduction
- Estimating parameters
- Hypothesis testing

- Point estimate
- Interval estimate

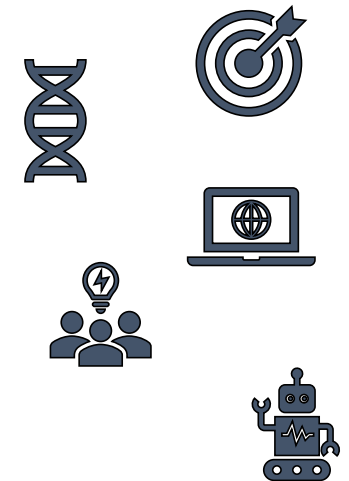
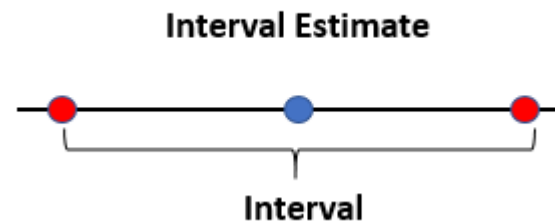


# Estimating parameters

- ✓ A confidence interval estimate is a range of values constructed from sample data so that the population parameter is likely to occur within the range at a specified probability.
- ✓ factors that determine the width of a confidence level:
  - Sample size
  - Population variability
  - Desired level of confidence
- ✓ Confidence Interval

- Introduction
- Estimating parameters
- Hypothesis testing

- Point estimate
- Interval estimate



# Hypothesis testing

- ✓ The main purpose of statistics is **to test a hypothesis**
- ✓ A hypothesis is an **educated guess** about something in the world around you.
- ✓ It should be testable, either **by experiment or observation**. For example:
  - A new medicine you think might work.
  - A way of teaching you think might be better.
  - A possible location of new species.
  - A fairer way to administer standardized tests.
- ✓ Visual illustration on the right test as if it is the right tool



color



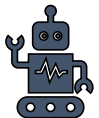
weight



size

- Introduction
- Estimating parameters
- Hypothesis testing

- Introductory
- The factors
- The road map





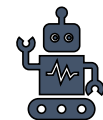
# Hypothesis testing

- ✓ Type of the variable based on the idea of (numerical or categorical)
- ✓ Number of samples
  - One, you test against the hypothesized value
  - One sample and each subject is measured on different variable
  - Two or more, and you compare to each other
- ✓ Type of analysis- purpose
  - Test against the hypothesized values/points or i.e. the claim
  - Comparison of two statistics
  - Relationship
- ✓ Distribution
  - Normal
  - Non-normal
  - Dichotomous



- Introduction
- Estimating parameters
- Hypothesis testing

- Introductory
- The factors
- The road map



Outcome	continues											discrete		
# samples	One – one Var		One – Two Var			Not One						One – one Var	One – Two Var	
						Two		> two						
purpose	Hypothesized value		Comparison		Relationship	Comparison						Hypothesized value	Comparison	Relationship
	mean	variance	mean	variance		mean	variance	mean			variance			
								One factor	Two factor	< two factor				
test	Test of mean T-test	Chi square for variance	Paired sample t-test	F-test	<ul style="list-style-type: none"><li>Pearson’s test</li><li>Spearman’s test</li></ul>	Indepen dent sample t-test	<ul style="list-style-type: none"><li>F-test</li><li>Bartlett test</li><li>Leven test</li></ul>	ANOVA- One way	<ul style="list-style-type: none"><li>ANOVA- Two way</li><li>ANCOVA</li></ul>	<ul style="list-style-type: none"><li>MANOVA</li><li>MANCOVA</li></ul>	<ul style="list-style-type: none"><li>F-test</li><li>Bartlett test</li><li>Leven test</li></ul>	test for proportion	Z-test for proportion	Chi square for independence

		Nominal		Ordinal	Continues		
One Sample		Hypothesized value			Hypothesized value		Hypothesized value
		Proportion		Median	Mean		Variance
		Binomial Test test for proportion One Sample Chi-Square Test		One-sample median test	T-test one sample		Chi square for variance
Two sample							
Independent		Proportion		Median	Mean		Variance
		z-test for independent proportions		Median test Mann-Whitney U Test	t-test for homogeneous variances t test for heterogeneous variances		F-test
Dependent		Proportion		Median	Mean		Variance
		Comparison	Relationship		Comparison	Relationship	
		Paired Samples Z-Test McNemar-Test	Chi square test	Wilcoxon-Test	Paired t test	Pearson's test Spearman's test	F-test
> Two samples							
Independent		Proportion		Median	Mean		Variance
		Chi-Square Test		H-Test Kruskal & Wallis	One factor	ANOVA one way/wo repeat	Bartlett test Leven test
					Two factors	ANOVA two way/wo repeat ANCOVA/wo repeat	
					> Two factors	MANOVA/ wo repeat MNCOVA/ wo repeat	
Dependent		Cochran-Test		Friedman-Test	One factor	ANOVA one way/w repeat	Bartlett test Leven test
					Two factors	ANOVA two way/w repeat ANCOVA/wo repeat	
					> Two factors	MANOVA/ w repeat MNCOVA/ w repeat	



# Hypothesis testing

- Introduction
- Estimating parameters
- Hypothesis testing

	<u>Comparison</u>				<u>Relationship</u>
	2 dataset		> 2 dataset		TWO VARIABLES
	Paired	Unpaired	Paired	Unpaired	
<u>Normal distribution</u>	P/ t test	Up/ t test	Repeated measures of ANOVA	One way ANOVA	Person correlation
<u>Non-Normal distribution</u>	Wilcoxon signed rank	<ul style="list-style-type: none"> <li>• Wilcoxon rank sum</li> <li>• Mann Whitney U test</li> </ul>	Friedman test	Kruskal Wallis test	Spearman's rank correlation
<u>Dichotomous distribution</u>	McNemar's test	<ul style="list-style-type: none"> <li>• Chi square test</li> <li>• Fisher exact test</li> </ul>	Cochran's Q test	Chi square test	Contingency coefficient

- Introductory
- The factors
- The road map





## Chapter Five

# STATISTICAL SOFTWARE

✓ Introduction



✓ Statistical software



✓ LAB implementation





# Introduction

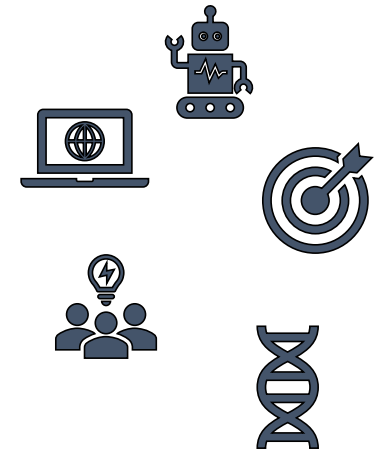
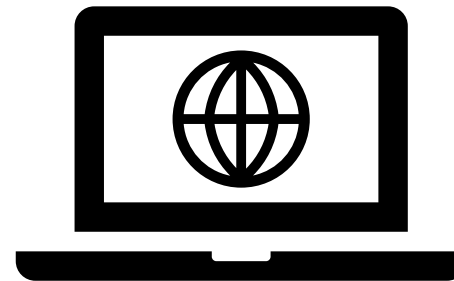


## Chapter Five

### STATISTICAL SOFTWARE

- Introduction
- Statistical Software
- Lab implementation

- ✓ Illustration of the (statistical software)
- ✓ specialized computer programs for analysis in statistics and econometrics
- ✓ Code or software ?





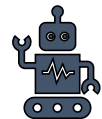
# Statistical Software



- ✓ Illustration of the (available options)
- ✓ Open source
  - ADaMSoft
  - Jamovi
  - R
- ✓ Public domain
  - CPro
  - Dataplot
- ✓ Freeware
  - MaxStat Lite
  - MINUIT
- ✓ Proprietary
  - SAS (software)
  - SPSS Statistics
  - S-PLUS



- Introduction
- Statistical Software
- Lab implementation



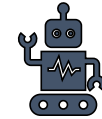
# Lab implementation



- ✓ Illustration of the (Jamovi)
- ✓ Software illustration
- ✓ BUG DATASET ANALYSIS



- Introduction
- Statistical Software
- Lab implementation





# Chapter Six

## LAB-SECTION\_03

---

✓ Review



✓ EDA analysis



✓ Inferential Analysis





# Review

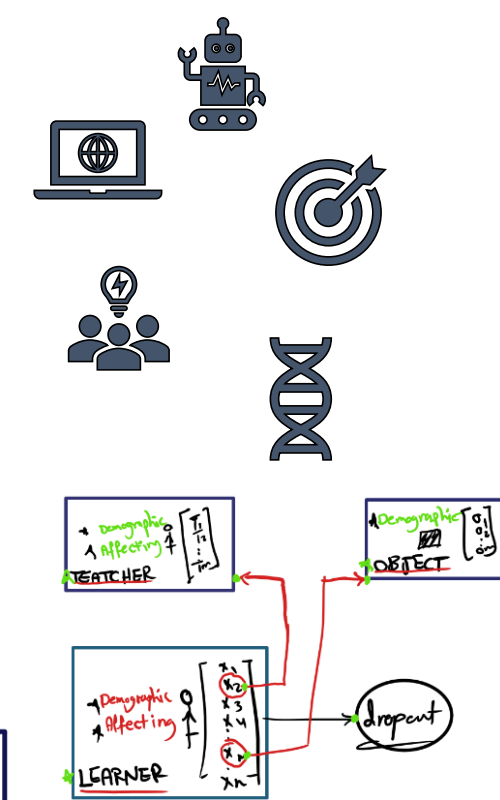
## ✓ Research Question:

- do the demographic & the affecting variables affect the decision of a life long learner to drop out a life-long training course?
- can those variables, collectively, subgroup or individually be predictive to the decision of dropping out the course?

## ✓ Research Hypothesis:

- H<sub>0</sub>: the demographic & affecting variables, collectively or subgroup **have NO affect** on the decision of a life-long learner to drop-out a life-long training course.
- H<sub>1</sub>: the demographic & affecting variables, collectively or subgroup **have affect** on the decision of a life-long learner to drop-out a life-long training course

- Review
- EDA Analysis
- Inferential Analysis



	44 (24-29)	42 (30-39)	43 (40-49)	46 (50-59)
Age	44	42	43	46
Gender	M	F	M	F
M.S	M1	M2	M3	M4
Children	0	1	2	3
Edu	E1	E2	E3	E4
Rank/educ	R1	R2	R3	R4
Job	J1	J2	J3	J4
V.S	V1	V2	V3	V4
P.I	P1	P2	P3	P4

INSTRUMENT A

Motivation	60	1. I like to work for an employer who is going to do better than I expect in doing this course, in expectation.
Aptitude	60	1. I like most of the work that I am doing in this course.
Personality	60	1. I am a person who is not very happy in this course.
Self-Confidence	60	1. I am a person who is not very happy in this course.

	44 (24-29)	42 (30-39)	43 (40-49)	46 (50-59)
Age	44	42	43	46
Gender	M	F	M	F
M.S	M1	M2	M3	M4
Children	0	1	2	3
Edu	E1	E2	E3	E4
Residence	R1	R2	R3	R4
year of Exp	Y1	Y2	Y3	Y4
T.Ry	T1	T2	T3	T4

INSTRUMENT B

Motivation	60	1. I like to work for an employer who is going to do better than I expect in doing this course, in expectation.
Aptitude	60	1. I like most of the work that I am doing in this course.
Personality	60	1. I am a person who is not very happy in this course.
Self-Confidence	60	1. I am a person who is not very happy in this course.

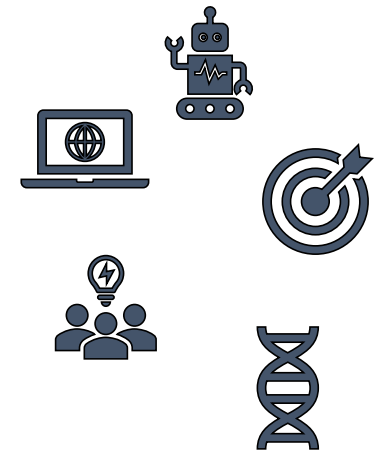
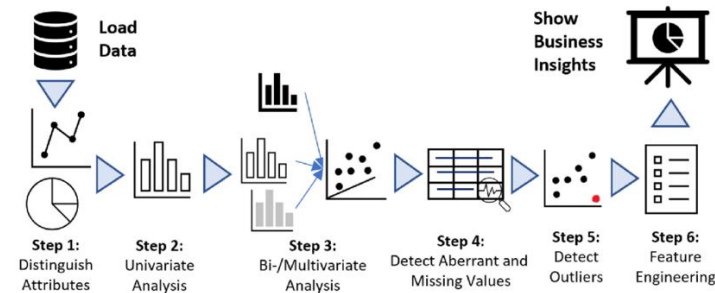
Subject Domain	S1, S2, S3
Previous Know	Yes, No
Class Requ	Yes, No
Nature of Sub	theory, application

INSTRUMENT C

# EDA Analysis

- Review
- EDA Analysis
- Inferential Analysis

- ✓ Organize
- ✓ Clean data
- ✓ Missing
- ✓ Descriptive model

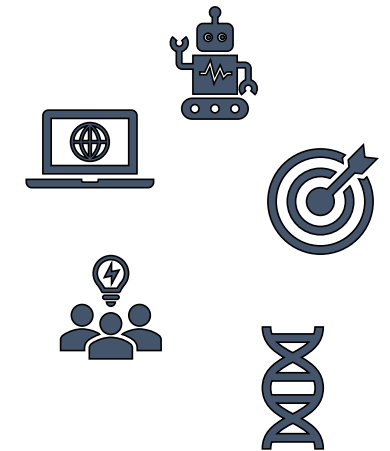
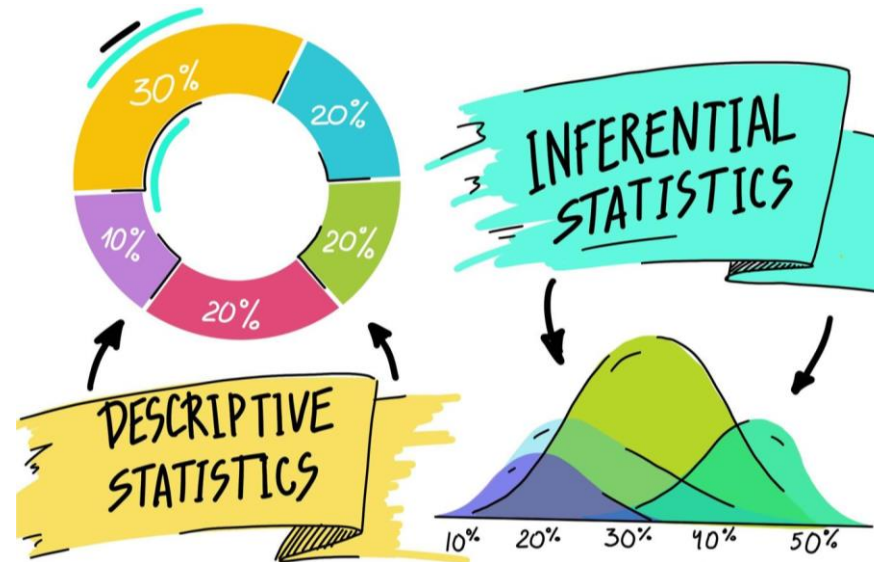




# Inferential Analysis

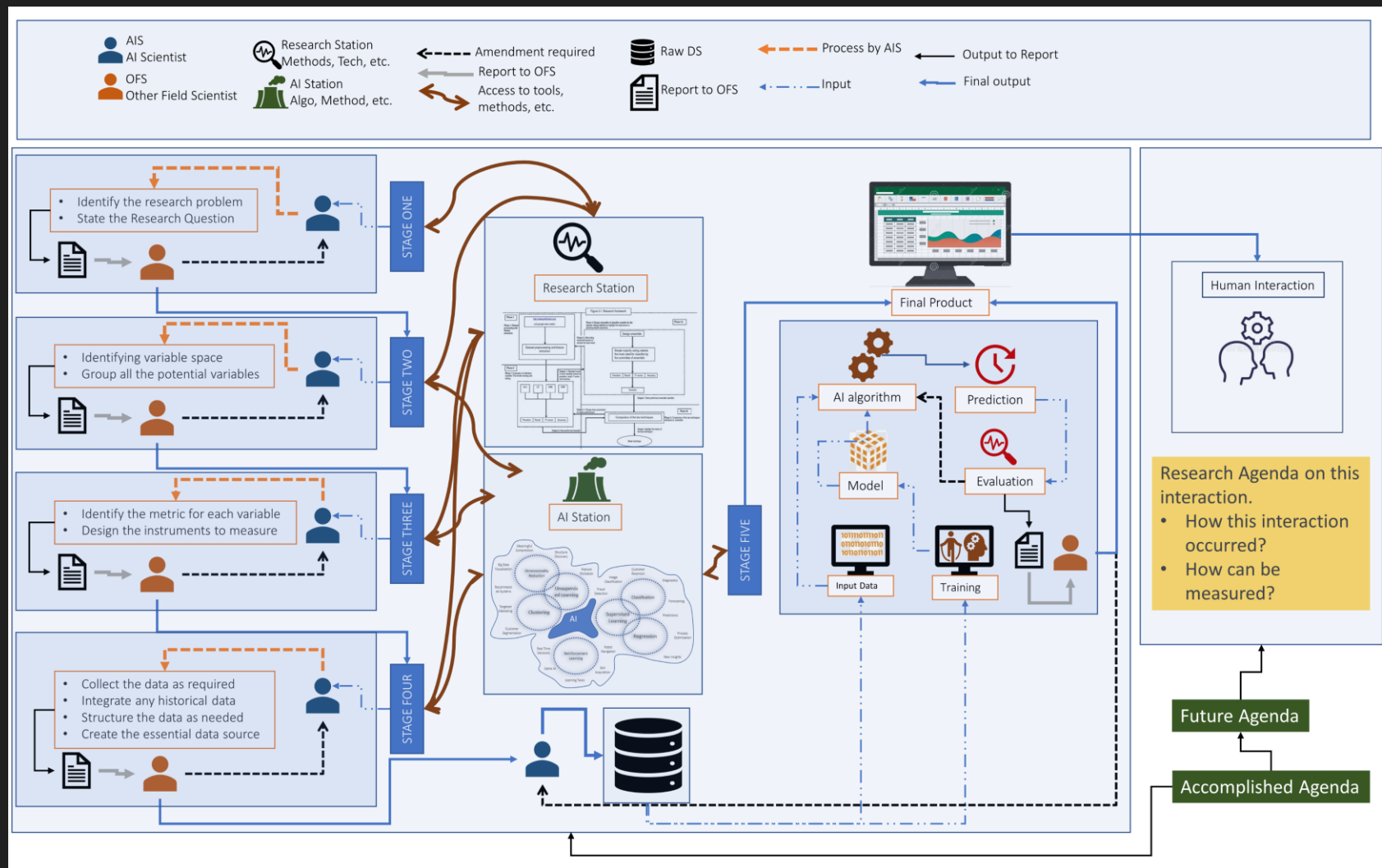
- Review
- EDA Analysis
- Inferential Analysis

- ✓ Explore the sample
- ✓ Determine the estimates
- ✓ Test the hypothesis





# THE MAP



## AISS V2.1.0

