



OWEBINAR

THE T-TEST

with Statistics for Data Analysis

June 13th 🗹







SPSS

WEBINAR AGENDA

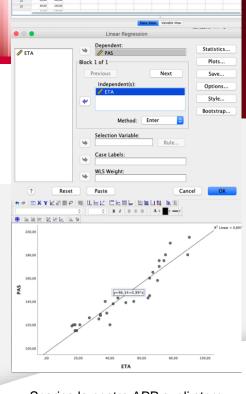
About Us Statistics for D

Statistics for Data Analysis T-Test with Statistics for Data Analysis

25+ years of history... in brief

Services included in the solution (LaunchBox, Add-On, APP, webinar, Support, ...)

- Notes on the Theory of Estimation and Statistical Inference
- The main differences between parametric and non-parametric tests
- T-test prerequisites
- Practical example
- Q&A



Scarica la nostra APP sugli store











Statistics for Data Analysis

Theory of Estimation and Statistical Inference







Descriptive statistics

Descriptive statistics deals with the comprehensive collection of manifestations of the phenomenon under consideration and provides a kind of description of the characteristics of the phenomenon itself.

For example, ... (an organisation wants to know some information about its territory)





Inferential Statistics

Inductive or inferential statistics aims to describe not so much what is apparent from the observed manifestations (partial surveys), but what would emerge if the survey were extended to the set of all manifestations of the phenomenon.

For example, ... (For example, an organization wants to conduct a survey on the impact of pollution on its territory)



Inferential Statistics

The uncertainty that arises from the partiality of the survey is dominated by the **Theory of Probability**.

With the methods of inferential statistics, you can move from the knowledge of the sample to that of the universe.

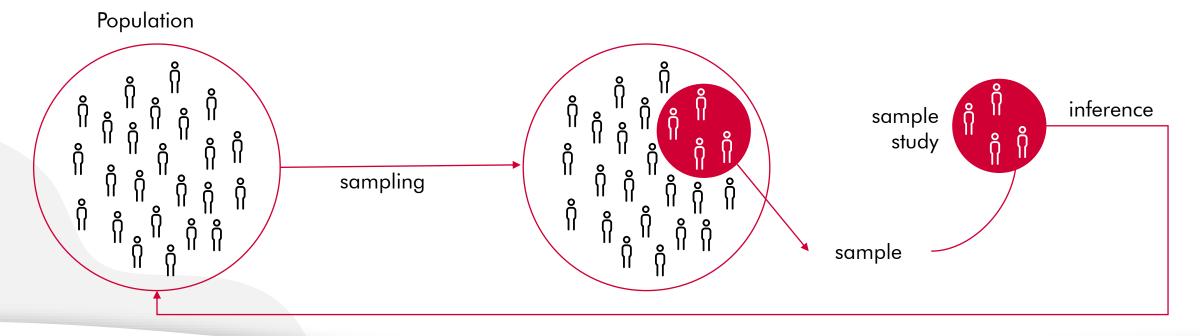


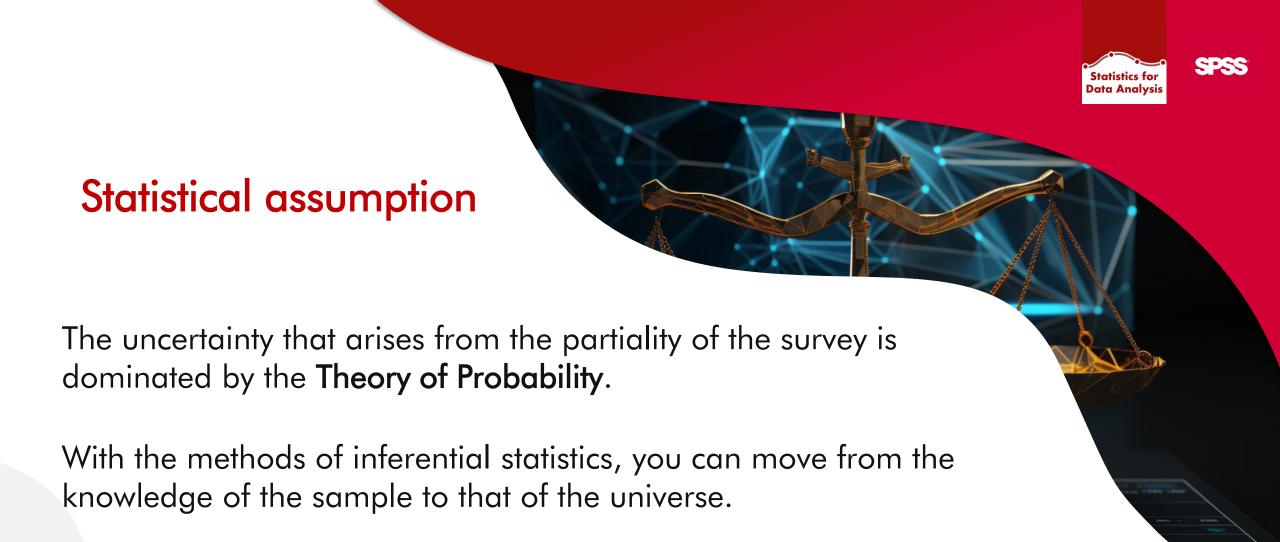


Statistics for Data Analysis

Inferential Statistics

- Rarely ALL the units that make up a population can be studied.
- Therefore, often only a part (SAMPLE) of the population is studied; in order to generalize the results (this process of generalization is called 'inference').









Logic

Premise

Logical arguments

Contradiction (reductio ad absurdum)

Conclusions
Falsification of the premise

Statistics

null hypothesis

Application of a test

Unlikely outcome (p<0,05)

Conclusions

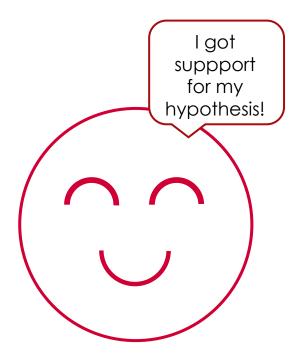
- The hypothesis is not rejected (a not highly improbable result occurred)
- The hypothesis is rejected (the observed result is inconsistent with what is specified in the hypothesis)



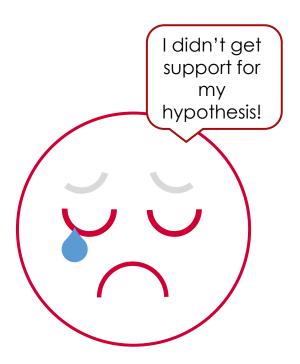




Verification of a Statistical Hypothesis



- Correct
- Incorrect (Data Reflects Chance)



- Correct
- Incorrect (Didn't Detect Effects)



Statistics for Data Analysis

Verification of a Statistical Hypothesis

What Researcher Does

Rejects

the Null

NOT Reject

e Null the Null

NOT Reject the Null





REALITY

(What researcher should do)

Rejects the Null









Assumptions for parametric tests

- Sample observations must be independent, meaning the sample must be random.
- The observations must belong to populations that are normally distributed.
- The populations must have the same variance (homoscedasticity)
- The observed variables must be measurable on an interval scale

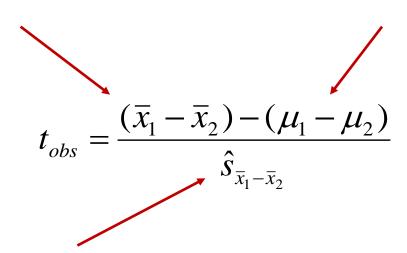
Many parametric tests are robust enough to withstand slight deviations from some of these postulates, especially when the sample size is sufficiently large.



Statistics for Data Analysis

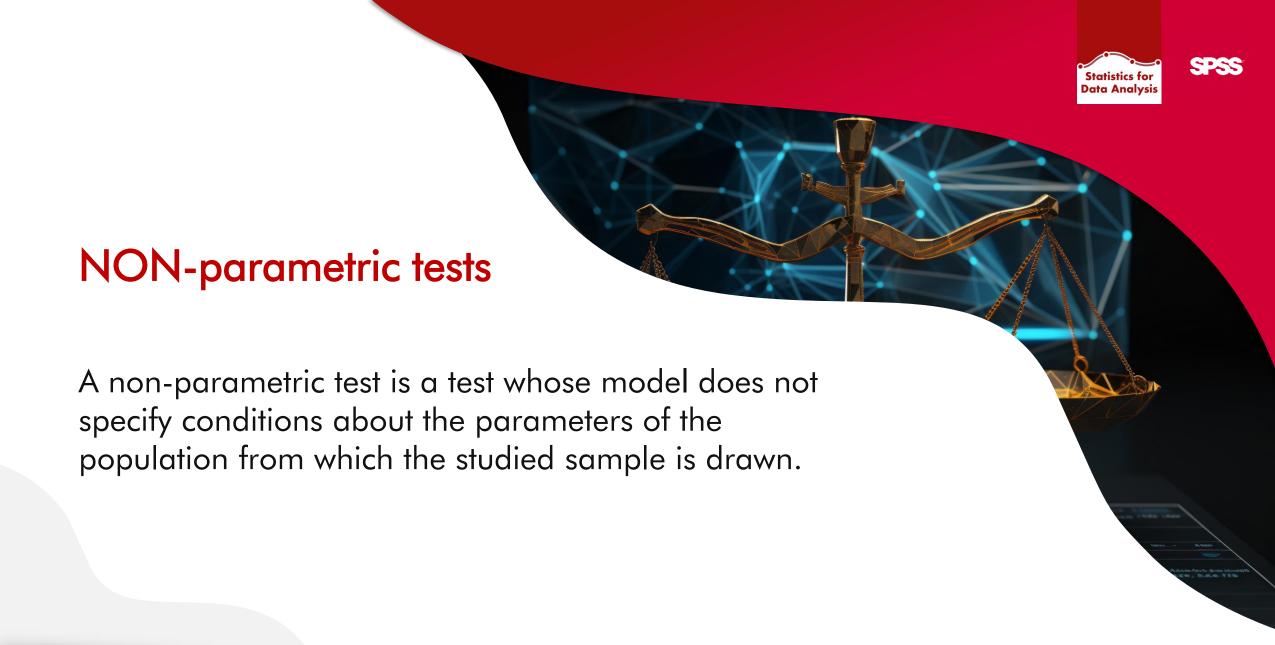
Assumptions for parametric tests

Difference between sample averages



 $(H_0 \text{ takes on this difference} = 0)$

Standard Error of the Difference between Sample Averages (IF both group variances are equal)







NON-parametric tests

Advantages of non-parametric tests:

- Less restrictive assumptions
- Feasibility of use even with small samples
- Analysis of different populations
- Analysis of ranks or nominal data

Disadvantages of non-parametric tests:

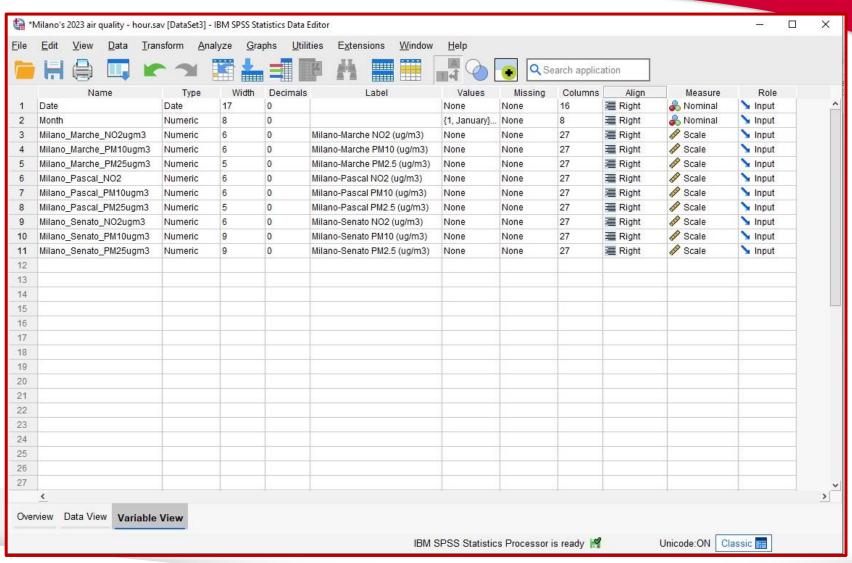
 Less precise with the same amount of information (if the assumptions were verified)







Examples





SPSS

Examples

