

Experimental Methods in Computer Science

Departamento de Engenharia Informática, FCTUC, 2023/2024

Experimental Methods in Computer Science (Metodologias Experimentais em Informática)

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Why do we need experiments?



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Why do we need experiments?

Researchers:

- Collect evidence facts about the world (or system)
- Validate hypothesis
- Support the definition, validation, parameterization of models
- Validate models
- Confirm theories
- Etc, etc...

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Why do we need experiments?

Engineers (including informatics engineers):

- Tune up systems
- Compare and select among different project choices
- **Verify** that requirements or specifications are met
- **Validate** mechanisms and/or solutions
- Measure/evaluate features, e.g., to access efficiency of mechanisms.
- Assess the effectiveness of processes, e.g., software development processes
- Etc, etc...

“Experimentation as the feedback step in the engineering loop”

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Key properties

- **Relevance**
Are the goals of the experiment and the expected results important for the progress (do they have impact in science, technology, market, etc.)?
- **Representativeness**
Is the experiment realistic and representative of real-world scenarios?
- **Repeatability**
Is it possible to repeat the experiment and achieve same or statistically similar results?
- **Reproducibility**
Is there enough information to allow others to reproduce the experiment?
- **Results analysis and generalization**
Is the analysis of results sound? Is the generalization of conclusions credible?
- **Cost**
Is the cost of the experiments compatible with the expected benefits?

Many types of experiments...

- Controlled experiments
- Field studies
- Case studies
- Pilot studies
- Benchmarks
- Simulations
- Surveys
- Rational reconstructions
- Artifact/archive analysis
- Ethnographies
- Quasi-experiments

Design of experiments (a first look)

Laboratory experiments, controlled experiments

1. Problem statement (or research question)
 2. Identify variables
 3. Generate hypothesis
 4. Define the experimental setup/scenario
 5. Develop tools and procedures for the experiment
 6. Run experiments and collect the data/measurements
 7. Perform data analysis and test hypothesis
 8. Draw conclusions (often go back to the beginning and reformulate the problem statement or test a different hypothesis)
- Design of the experiment
- Measurements
- Analysis
- Conclusions

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- A good (i.e., relevant) problem statement should be focused enough to allow the clear identification of the variables of the problem but, at the same time, should be sufficiently open to allow different hypothesis to answer the problem/question.
- Possible generic formulation:
How does X affect Y under conditions Z?

Design of experiments (a first look)

Laboratory experiments, controlled experiments

1. Problem statement (or research question)
2. Identify variables
3. Generate hypotheses
4. Define the experiment
5. Develop test cases
6. Run experiments
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8. Draw conclusions (often go back to the beginning and reformulate the problem statement or test a different hypothesis)

To formulate good problem statements:

- Must know the subject area: process, system, technique, product, product market, etc.
- Must be precise and clear
- Must be sure that the problem/question is relevant

Design of experiments (a first look)

Laboratory experiments, controlled experiments

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• **Dependent variable** (response variable)

Measured output (e.g., response time, throughput, no. bugs, downtime, latency, error detection coverage, etc., etc.)

• **Independent variables** (factors)

Input variables that can be changed in the experiment (e.g., memory size, clock rate, file size, channel bandwidth, etc., etc.)

• **Levels**

Values taken by the variables. Can be (nearly) continuous (e.g., ~time, size in bytes) or discrete (type of system, type of algorithm, etc.)

Design of experiments (a first look)

Laboratory experiments, controlled experiments

1. Problem statement (or research question)
2. Identify variables
3. Generate hypotheses
4. Define factors and levels
5. Develop experimental design
6. Run experiments
7. Perform statistical analysis
8. Draw conclusions and reformulate hypotheses

- **Change one factor at the time**

Simple scenario. The analysis is simple, as it is easy to understand the effect of a given factor on the independent variable.

- **Full factorial**

Change two or more factors simultaneously. Much more complex (to be seen in detail later on in the course). Has two important advantages:

- More efficient experiments (save time and effort)
- Allow the study of possible interactions among factors

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- **Terminology:**

- **Baseline (often called golden run)**

Set of factor values (i.e., independent variables) that represent a baseline scenario for the experiments

- **Repetition of golden run**

Used to estimate the experimental error (noise) in the system and identify small effects that may cause variations in the results.

- Allow the study of possible interactions among factors

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Design of experiments (a first look)

Laboratory experiment

1. Problem statement
2. Identify variables
3. Generate hypothesis
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6. Run experiments and collect data
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8. Draw conclusions (often reformulate the problem statement or test a different hypothesis)

Terminology:

- **Randomization**

Minimize potential uncontrollable biases in the experiments by randomly assigning factors to “average out” the effects of possible extraneous factors.

- **Blocking**

The experiment is divided in homogeneous segments (blocks such as sets of machines, users, loads, etc.) to improve precision. The goal is to control the variability block to block.

- **Confounding variable**

Extraneous variable that influences the relationship between the dependent and independent variables (i.e., correlates with both the dependent and independent variables).

Design of experiments (a first look)

Laboratory experiment

1. Problem statement (or objective)
2. Identify variables
3. Generate hypothesis
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8. Draw conclusions (often reformulate the problem statement or test a different hypothesis)

- Hypothesis describe provisional relationships between factors (independent variables) and the response variable (dependent). It is a interim answer to the problem statement.

- Can be directional or non-directional

- May lead to a model allowing prediction of what is going to happen in future cases.

- Quite often (in computers) the goal of the experiments is to quantify the relationship (not just confirm that exists)

Design of experiments (a first look)

Laboratory experiments

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- Experiment complexity
 - Experiment cost
 - Availability of tools and frameworks that may help
 - Degree of automation

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Design of experiments (a first look)

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- Continuous and/or discrete measurements
 - Accuracy, precision, and resolution
 - Basic measurements in computers...
 - Count
 - Duration
 - Size
 - Any value derived from the combination of basic measurements

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Design of experiments (a first look)

Laboratory

1. Problem statement
2. Identification of variables
3. Generalization
4. Definition of hypotheses
5. Development of experimental design
6. Run experiments and collect the data/measurements
7. Perform data analysis and test hypothesis
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- Exploratory data analysis
- Statistical data analysis
 - Tables, charts, etc., average, standard deviation
 - Coping with measurement errors
 - Confidence intervals
 - Statistical comparison of alternatives
 - Tests to check if measured data fit known distributions (chi-square, K-S tests,...)

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- The written **report** of the experiments is quite often the **single outcome** of months or years of work
- **Quality of writing** is essential. Some relevant attributes of the report:
 - Clear (in the goals, approach, setup, steps, analysis, discussion, conclusions)
 - Credible (in the data reported, conclusion, etc.)
 - Self-contained

Example

Consider the following problem/question: Is the number of software bugs found in the tests of program units developed by programmers dependent on the average number of sleeping hours of the programmers?

Assume that you have the detailed specifications of a set of program units to be developed and consider that the program units include units of high, medium and low complexity. Additionally, you have comprehensive unit test suits to test each program unit.

In these circumstances, describe how you would organize an experiment to answer the proposed question (problem statement). Your answer should be as complete as possible, focusing on the experiment design steps (obviously, it does not make sense to speculate about the experiment results and conclusions), and indicate at least the following:

- a) Dependent and independent variables.
- b) Levels you would consider for the independent variables.
- c) Hypothesis under evaluation.
- d) Hypothesis testing technique you would use.
- e) Brief description of the experimental setup, taking into account in your answers to the previous points and the fact that the experiment deals with people (the programmers).