Experiment planning

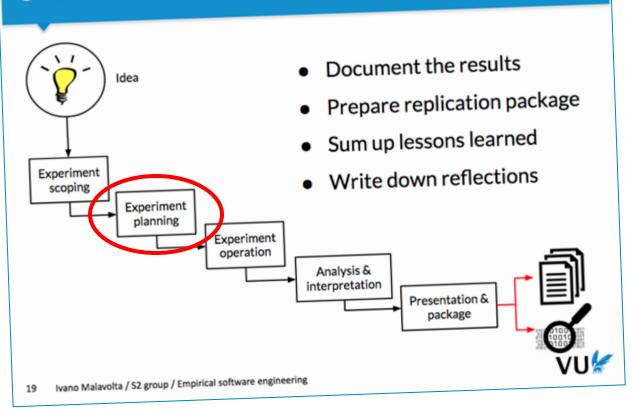
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Recall

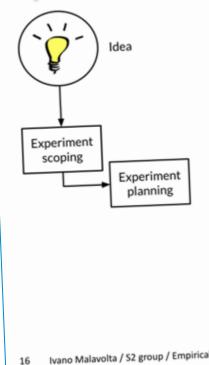
5 - Presentation & package





Recall

2 - Experiment planning



- Define context
- Formulate hypotheses
- Identify input and output variables
- Design the study
- Instrumentation
- Analyze validity threats ٠







Experiment planning

- Context selection
- Research questions and hypotheses formulation
- Variables selection
- Subjects selection

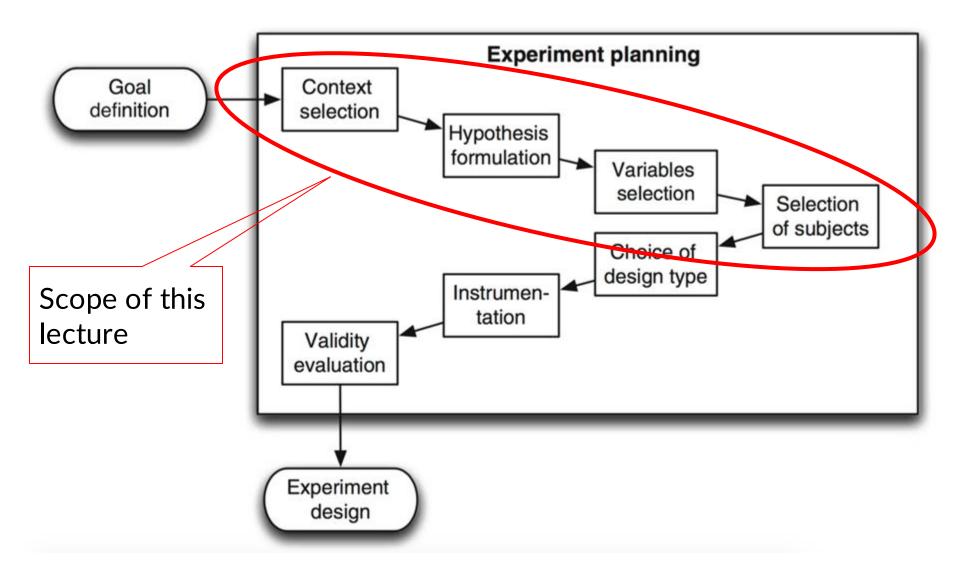


Scoping VS planning

- Experiment scoping describes WHY we run an experiment
 - With a hint about the "how"
- The planning determines **HOW** the experiment will be executed
 - Be careful here \rightarrow the result of the experiment can be disturbed (or even destroyed) if not planned properly



Planning phases



Context selection



We already heard about context...

Image encoding	example - goal
----------------	----------------

Analyze	Encoding algorithms
for the purpose of	Evaluation
with respect to their	Energy Efficiency
from the point of view of	Software Developer
in the context of	Mobile Software Applications
(Empirical software engineering	

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Context selection

CONTEXT: the broad perspective of the experiment

Our **goal** here is to achieve the most general results \rightarrow optimum: large real software projects, with practitioners

Many risks involved....



Context selection dimensions

Lab

• Off-line

- Students
- Toy objects

• Specific

Reality

• On-line

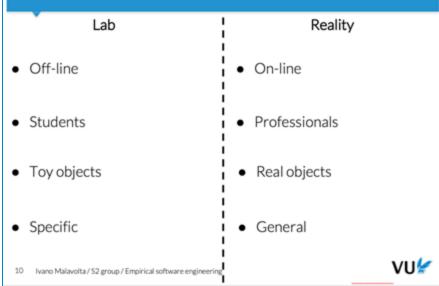
- Professionals
- Real objects

• General



Quick exercise

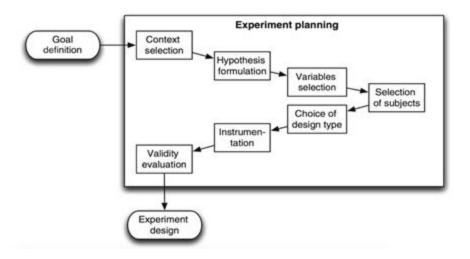
Context selection dimensions



Think about the Image encoding example and formulate a potential context for an experiment



Research questions and hypotheses formulation





From research questions to hypotheses

- When a research question is going to be addressed by applying a <u>statistical test</u>, it is necessary to formulate one or more hypotheses
- Very useful to select what kind of statistical procedure you need to use



Hypotheses formulation

- Conjecture (P)
 - Administration of treatment <u>has influence</u> on some features
- Consequence (Q)
 - We <u>observe</u> a significant difference in terms of some features

$P \rightarrow Q$



Hypotheses formulation

Hypothesis: a formal statement about a phenomenon

- Null hypothesis H₀: no real trends or patterns in the experiment setting (aka ~Q)
- Alternative hypothesis H_a: there are real trends or patterns in the experiment setting (aka Q)

There must be at least a pair of null and alternative hypotheses **for each research question** in your GQM



Falsification (modus tollens)

• We aim at rejecting the absence of trends (~Q is false)...

• we test the null hypothesis H_0

If we can reject the null hypothesis $\sim Q \rightarrow$ we can draw conclusions

This comes from Popper (1959): any statement in a scientific field is true until anybody can contradict it

• Aiming at verifying Q is <u>WRONG</u>

- "Affirming the consequent"
- Provides no insight on the conjecture

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This is like a "guilty" verdict in a criminal trial: the evidence is sufficient to reject innocence



Example

• Question:

• What is the impact of image encoding algorithms on the energy *consumption of mobile apps*?

• <u>Conjecture</u> (P):

- using different algorithms leads to different levels of energy consumption
- <u>Consequence</u> (Q):
 - (when applying different algorithms) we observe a different energy consumption



Example

• Null hypothesis (¬Q): there is no change in terms of energy consumption

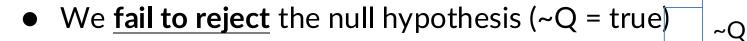
• Alternative hypothesis (Q): the energy consumption when using PNG images is different than the one consumed when using JPG images

$$H_a: mean(E_{png}) != mean(E_{jpg})$$



What can happen now?

- We <u>reject</u> the null hypothesis (~Q = false)
 - Q = true (with a certain level of confidence)
 - our conjecture P has been corroborated → we are confident that different algorithms impact energy consumption (P)



 $\circ \sim P = true^{-1}$

• our conjecture P has been falsified \rightarrow no conclusions can be made





Modus tollens:

 $P \rightarrow Q$

~P

Example

Native vs Web Apps: Comparing the Energy Consumption and Performance of Android Apps and their Web Counterparts

Ruben Horn, Abdellah Lahnaoui, Edgardo Reinoso, Sicheng Peng, Vadim Isakov, Tanjina Islam, Ivano Malavolta Vrije Universiteit Amsterdam, The Netherlands

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0:

Abstract-Context. Many Internet content platforms, such as Even though those apps provide similar features to the end Spotify and YouTube, provide their services via both native and Web apps. Even though those apps provide similar features to the end user, using their native version or Web counterpart might lead to different levels of energy consumption and performance. Goal. The goal of this study is to empirically assess the energy consumption and performance of native and Web apps in the context of Internet content platforms on Android.

Method. We select 10 Internet content platforms across 5 categories. Then, we measure them based on the energy consumption, network traffic volume, CPU load, memory load, and frame time of their native and Web versions; then, we statistically analyze the collected measures and report our results.

Results. We confirm that native apps consume significantly less energy than their Web counterparts, with large effect size. Web apps use more CPU and memory, with statistically significant difference and large effect size. Therefore, we conclude that native apps tend to require fewer hardware resources than their corresponding Web versions. The network traffic volume exhibits statistically significant difference in favour of native apps, with small effect size. Our results do not allow us to draw any

user, using their native version or Web counterpart might lead to different levels of energy consumption and performance.



TABLE III: Dependent variables

Variable	Description	RQ
Energy consumption (e)	Energy consumption is measured in Joules (J) as the energy consumed by the mobile device during the experi- ment run	RQ1
Network traffic (n)	Amount of data in Bytes (B) sent and received by the mobile device during the experiment run	RQ2
CPU load (c)	Mean relative (%) device CPU uti- lization across all cores	RQ2
Memory load (m)	Mean (kB) device memory utilization	RQ2
Frame time (f)	Median time in nanoseconds (ns) be- tween two successive frames (We use median as an aggregation measure, since we expect extreme outliers due to apps blocking on the main thread during certain operations)	RQ2

RQ1 How does energy consumption vary between native and Web versions of the same app?

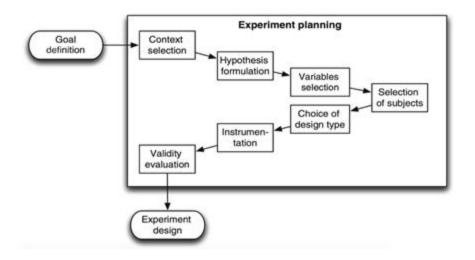
RQ2 How does performance vary between native and Web versions of the same app?

$$egin{aligned} H_0: \mu_{e_{ ext{native}}} &= \mu_{e_{ ext{Web}}} \ H_a: \mu_{e_{ ext{native}}}
eq \mu_{e_{ ext{Web}}} \end{aligned}$$

$$\begin{split} H_0 : \mu_{d_{\text{native}}} &= \mu_{d_{\text{Web}}} \ \, \forall d \in \{n,c,m,f\} \\ H_a : \mu_{d_{\text{native}}} \neq \mu_{d_{\text{Web}}} \ \, \exists d \in \{n,c,m,f\} \end{split}$$

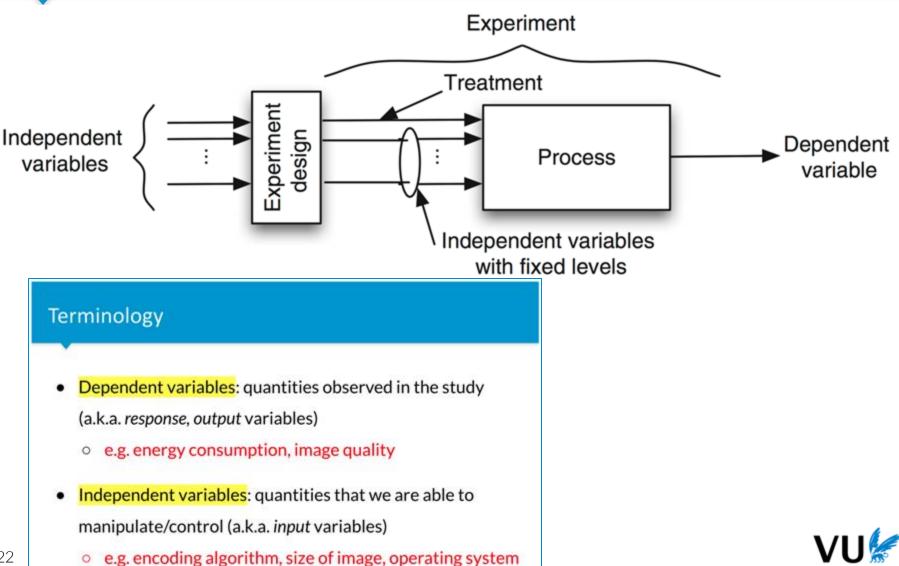


Variables selection





Recap



Variables selection

- The choice of independent and dependent variables is usually done **in parallel**
- Some variables cannot be measured directly (e.g. productivity, code quality, effort...)
 - We use **proxies** to estimate them
 - proxies may introduce a construct validity threat: is what we are measuring a good representation of our variable?



Variables selection

 Independent variables should have some effect on the dependent ones

 \rightarrow do not choose variables randomly, think about your RQs

 After choosing the variables you have to define their types, scales, ranges → this is part of measurement theory



Hypotheses formulation

- There is always only 1 dependent variable
 - e.g., power consumption
 - ... and 1 independent variable (the main factor)
 - often one treatment for the **control group**

e.g. use of traditional technique

- one or more treatments for experimental groups

e.g. use of new technique(s) tool(s)

Other independent variables are the co-factors



Co-factors

Our main factor is not the only variable influencing the dependent variable(s)

• e.g., network instability of your experimental environment, usage patterns of the analysed websites, skills of subjects, experience of developers, ...

We will never account for all possible co-factors

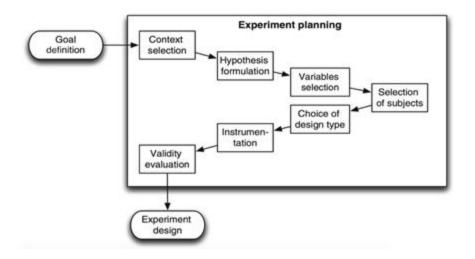
Your best friend here is randomization

In a good experiment:

- You **limit** the effect of co-factors through a good experimental design
- You are able to **separate** the effect of co-factors from main factors
- You analyze the **interaction** of co-factors with main factor



Subjects selection





Subjects selection

- **Population**: the complete set of items of interest for our experiment
 - e.g. open-source software applications
 - e.g., all existing progressive web apps
- **Sample:** representative selection of individuals for that population
 - e.g. Apache, MySQL
 - e.g. progressive web apps mined from the Tranco list



Sampling techniques

- **Probability Sampling:** the probability of selecting each subject in the population is <u>known</u>
 - Simple Random Sampling: random selection from the population, probability is 1/total
 - Stratified random sampling: the population is divided into groups with a known distribution between the groups. Random sampling is then applied within each group
- Non-probability sampling: the probability of selecting each subject out of the population is <u>unknown</u>
 - Convenience: the most convenient (cost/distance/ complexity) subjects are selected [usually it is the only way to go]
 - *Quota*: you select (usually by convenience) samples from groups of subjects (e.g. male vs females, open-source vs closed source)



How big should be a sample?

• Sample size: the larger, the better (more general results)

• If the population has a high *variation*, a **larger** sample size is needed

- Data analysis may influence sample size
 - some statistical tests have meaning only on large samples



Example of subjects selection

Selected Subjects and Usage Scenarios

Category	Subject	Usage scenario (looped)
News	• ESPN	Open news article, scroll down, continue with next article
	The Weather Channel	Check hourly forecast, check 10-day forecast, check radar
Social media	• LinkedIn	Scroll personal feed, scroll jobs
	Pinterest	Scroll posts, open post, go back
E-Commerce	Coupang	Open category, scroll products, open product page, check comments
	Shopee	Open category, scroll products, open product page, check comments
Audio streaming	 SoundCloud 	Listen to promoted song
	Spotify	Search for a playlist, listen to playlist
Video streaming	• Twitch	Search for channel, watch channel
	YouTube	Search for video, watch video

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Read the details in Section 3a in the MobileSoft 2023 paper on Canvas

What this module means to you?

You know how to:

- define the context of your experiment
- link research questions to statistical hypotheses
- define independent and dependent variables
- strategies for selecting subjects

Next step

Measurement theory \rightarrow how to define the "type" of variables



Readings

Claes Wohlin - Per Runeson Martin Höst - Magnus C. Ohlsson Björn Regnell - Anders Wesslén

Experimentation in Software Engineering

Chapter 8

Springer

+ MobileSoft 2023 paper

+ All papers in the "Articles on performed experiments" folder in Canvas (only the part related to subjects selection and variables definition)

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