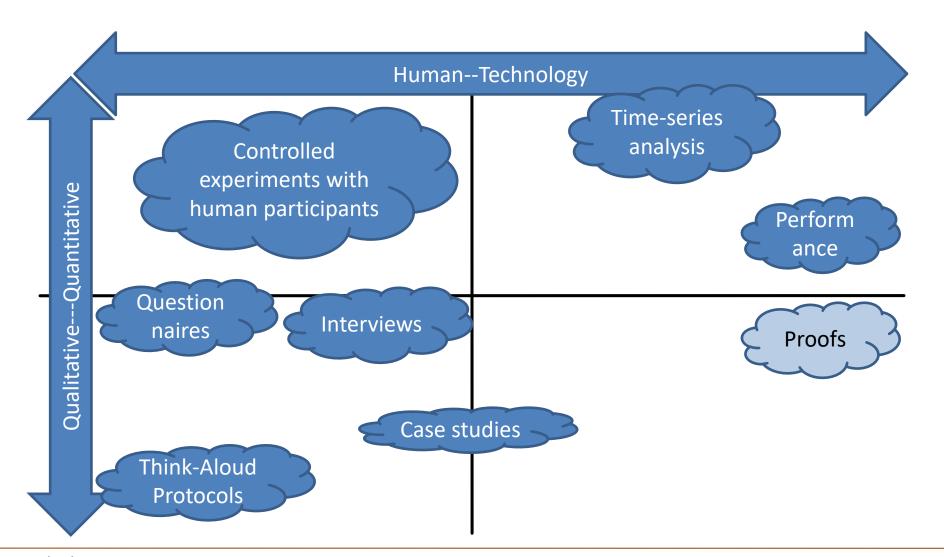


Controlled Experiments

Overview





Learning Goals

- Design good research hypotheses
- Design an experiment with high internal validity and high external validity

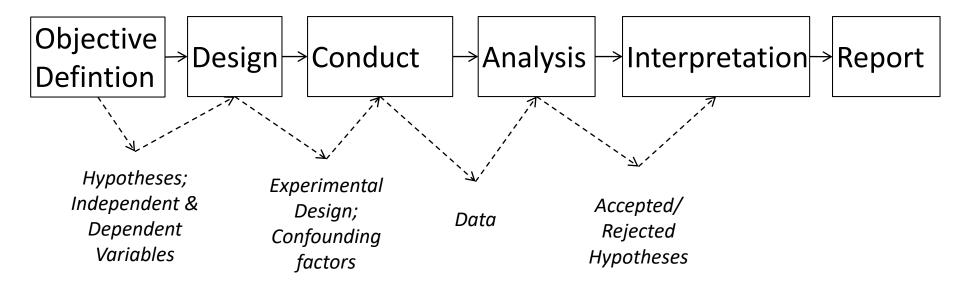


Controlled Experiment: Definition

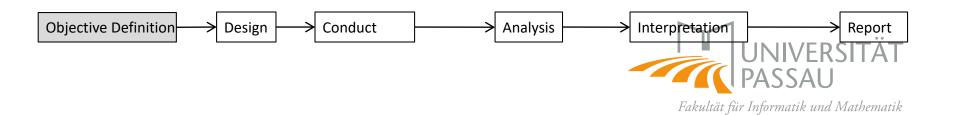
- Systematic study
- One or more factors are varied
- Everything else is held constant
- Result of systematic variation is observed



Experimental Phases







Variables

Independent Variables

- Varied by experimenter on purpose and systematic
- Also called factor or predictor (-variable)
- Has alternatives, levels, or treatments

- Examples:
 - Programming language or paradigm
 - User interfaces
 - Interaction methods



Dependent Variable

- Result of an experiment
- Depends on variation of independent variable
- Is observed

- Examples:
 - Productivity of programmer
 - Bugs in a program
 - Operator error



Latent Variables

- Construct
- Not observable directly

- Examples
 - Program comprehension
 - Intelligence
 - Mental model of user



Operationalization

- Defining operations that allows you to measure variables
- Must not contradict common sense

- Example:
 - Program comprehension
 - Number of bugs in a program
 - Development time



Task

- Find operational definition for the following variables:
 - Usability of new UI
 - Maintainability of a program



Hypotheses

- Expectations of results
- Expecations need to be justified, e.g., in theory or practice
- Hypotheses need to be simple and clear
- Hypotheses need to be falsifiable
- Falsifiability (Homework assignment: Make yourself familiar with the term and explain its role for experimental design)



Hypotheses- Bad Examples

- Bad source-code comments are bad for program comprehension
- Good source-code comments are good for program comprehension



How can we do better?

- Comments that describe every statement of source code do not affect the time developers need to understand a source-code snippet
- Comments that contain wrong information about source code increase the time developers need to understand a source-code snippet
- Comments describing the purpose of source-code statements decrease the time developers need to understand a source-code snippet



Why do we need Hypotheses?

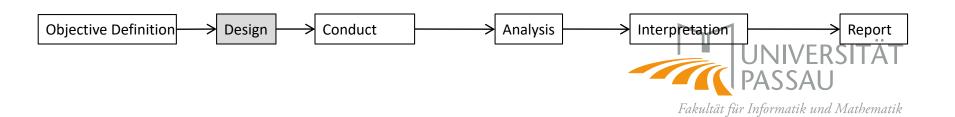
- They guide us when designing an experiment
- Prohibit Fishing for Results
- Connects theory and empirical research
 - Derived from theory
 - Evaluated with empirical research



Task

- State a hypothesis on the following research questions:
 - Does Python increase productivty of developers?
 - Is Pyhton better than C++?
 - Is the new UI more productive?
- Keep in mind that the hypothesis needs to be evaluated, justified, and the variables operationalized
- = Example for an exam question





Design

Validity

• Do we measure what we want to measure?



Internal Validity

 Amount to which the value of the dependent variable can be explained my the systematic variation of the independent variable



External Validity

- Amount to which results are transferable to other circumstances (participants, material,...)
- Generalizability



Homework Assignment

Research other kinds of validity



Threats to Validity

- Confounding factors:
 - Influence dependent variable in addition to indepedent variables
 - Learning effect
 - Hawthorne effect
 - Mono-method bias
 - Selection bias



Task

- Evaluating the effect of new UI: What confounding factors exist?
- How could the influence of these factors be controlled?



Confounding Factors

- There are numerous confounding factors
- Carefully identify and control their influence:
 - Randomization
 - Matching/parallelization/balancing
 - Define confounding factors as independent variable
 - Keep confounding factor constant
 - Analyze afterwards



Randomization

- Random numer generator
- Toss a coin
- Throw a dice

• ...

Issues:

- Groups need to be sufficiently large
- 5 per group seems to be too low, 10 seems to be sufficient



Matching/parallelization/balancing

Participant	Value
P5	65
P9	56
P3	42
P4	34
P10	24
P6	23
P7	21
P8	16
P2	12
P1	5

Group A	Group B
65	56
34	42
24	23
16	21
12	6

odd-even-even-odd/ ABBA



Matching/parallelization/balancing

- Drawback compared to randomization
 - Confounding factor needs to be measured
 - How to measure programming experience? How to measure intelligence?
- Advantage compared to randomization:
 - More detailed knowledge about parameter



Define confounding factors as independent variable

- Is varied systematically by experimenter
- Confounding factor is operationalized
- Experience with tool:
 - New UI/low experience
 - Old UI/much experience
 - New UI/low experience
 - Old UI/much experience



Doing the Math...

- 23 confounding factors, each with 2 levels
 - = 8 388 608 possible combinations
- How many participants do I need to cover each combination
 - at least 10 participants per group
 - 83 886 080 (i.e., Germany)



Keep confounding factor constant

- Only one leve of confounding factor
- Programming experience:
 - Only undergraduate students
 - Only programming experts
- Intelligence
 - Only students with a certain grade



Analyze afterwards

- Measure factor during experiment
- Analyze influence of a variable after the experiment
- Issues
 - Could show that results are useless



Recommendation

- In your experiment, maximize internal validity
- In your experiment, maximize external validity



Relation between Internal and External Validity

- Both kinds of validity request different things:
 - Internal: control everything
 - Extern: general setting
- And now?
 - First maximize internal validity
 - Then increase external validity step by step



Quality Criteria of Empirical Studies

- Validity
- Reliability
- Objectivity



Quality Criteria of Empirical Studies

Reliability:

Accuracy of measurement instrumenst

Objectivity:

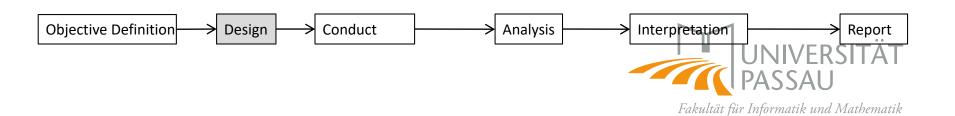
- Execution of experiment must need depend on person of experimenter
- The same experiment, conducted by a different experimenter, should produce the same result



Example

- Scale for measuring the weight:
 - Valid
 - Reliable depending on quality
 - Digital scale is more objective, as everyone sees the same number (analogous leaves more wiggle room)
- The same scale for measuring the height
 - Less valid
 - Reliable depending on quality





Experimental Designs

Designs

- Between vs. Within Subject
- = With vs. without repeated measures
- One-factorial vs. multi-factorial
- = One vs. several independent variables
- Univariate vs. Multivariate
- = One vs. several dependent variables



Why Experimental Designs?

- Instruction to act
- Makes communication easier
- Decision for statistical analysis



How to Select a Design

- Depends on:
 - Effect size
 - Sample size
 - > The bigger both are, the less the influence of confounding factors manifest
 - --> When both are small, a suitable design is very important
- Unfortunately, you will very often have small samples and unknown effect sizes





One-Factorial

Between-Subjects

- Participants are devided into two groups
- As many group as there are levels of the independent variable
- Results are compared between groups

```
Gruppe Stufen

A Textuelle
Annotationen

B Hintergrundfarben
```

```
//Add the core application commands always
    public static final Command viewCommand = ne
     public static final Command addCommand = new
    public static final Command deleteCommand = 1
27
     public static final Command backCommand = new
29
     public static final Command editLabelCommand
30
31
33
     public static final Command favoriteCommand
    public static final Command viewFavoritesCom
35
36
37
        * Constructor
38
39
     public PhotoListScreen() {
40
       super ("Choose Items", Choice. IMPLICIT);
```



Issues

- Variance between participants (i.e., inter individual differences) can be large
- -> 10x (What does 10x Mean? Measuring Variations in Programmer Productivity. Steve McConnell.)
- Sufficient number of participants
- Balancing between groups

Within-Subjects

- Inter individual differences need to be controlled for
- Each participant is exposed to all levels of an independent variable

One	Session 1	Session 2
Group	Background	Textual
	colors	Annotations



Issues

- Learning effects
 - Especially with creative tasks
 - You need different, but similar tasks at the same time
- Ordering effects
- Intra individual differences
 - Fatigue
 - Motivation
- Mortality



Crossover

- Each participant is exposed to all levels
- Comparison between and within groups is possible

Group	Session 1	Session 2
A	Background colors	Textual annotations
В	Textual annotations	Background colors



Issues

- Intra individual differences
- Inter individuelle differences
- Mortality



Benefits

- Check for learning effects:
 - Difference between both sessions for both levels
- Check for ordering effects:
 - Difference between both sessions for one level

Gruppe	Session 1	Session 2
Α	Background colors	Textual annotations
В	Textual annotations	ва kground colors



Comparison

Property	Between-Subjects	Within-Subjects	Cross-Over	
Sample size	2	1	2	
Group balancing	2	1	2	
Learning effects	2	3	1	
Ordering effects	2	3	1	
Mortality	1	2	2	
Motivation, fatigue	1	2	2	
Experiment duration	1	2	2	
Internal validity	2	2	1	
External validity	2	2	1	





Multi Factorial Designs

Latin Square

Group	Task 1	Task 2
A	Background colors	Textual annotations
В	Textual annotations	Background colors

- Special case of cross over
- But different task in sessions -> Task is second factor

Two-factorial, Between-subjects

Programming experience, Intelligence

Variables	Groups	
Background color/ novice	Group A	
Background color/ expert	Group B	
Textual/ novice	Group C	
Textual/ Expert	Group D	

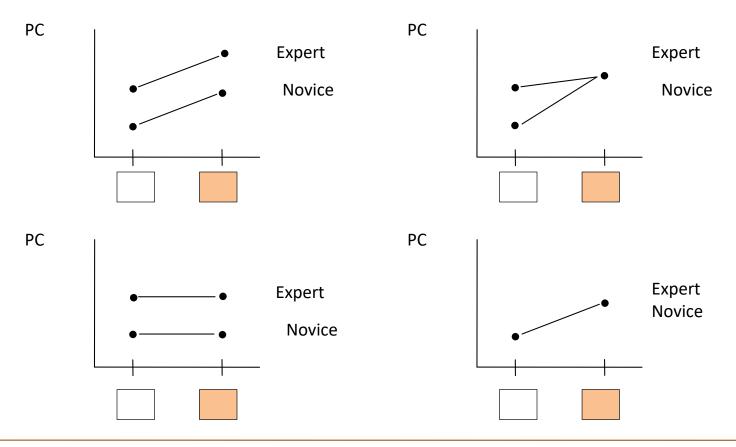


Two-factorial, Within-Subjects

Group	Session 1	Session 1 Session 2		Session 4	
Background color/ novice	Group A	Group D	Group C	Group B	
Background color/ expert	Group B	Group A	Group D	Group C	
Textual/ novice	Group C	Group B	Group A	Group D	
Textual/ Expert	Group D	Group C	Group B	Group A	



Main- and Interaction Effects





Multi-factorial Designs

- In case the shown designs are not sufficient
- 4-factorial design (2x2x3x2)
- Higher-order interaction

		C ₁		C ₂		C ₃	
		B ₁	B ₂	B ₁	B ₂	B ₁	B ₂
A_1	D ₁						
	D ₂						
A ₂	D ₁						
	D ₂						



Selecting a Design

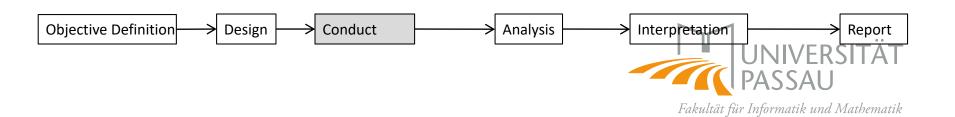
- Prefer a simple as possible design
- Carefully consider benefits and disadvantages
- Consider resource constraints



Homework assignments

- Why do we need hypotheses?
- What does falsifiability mean?
- Research other kinds of validity
- Design an experiment for a self-selected research question
 - Define hypotheses
 - Operationalize variables
 - Identify the most important confounding parameters (at least 5) and select suitable control techniques
 - (You do not need to think about statistical tests yet)





Conduct

What Can Go Wrong?

- Everything!
- Pilot studies:
 - Test material and tools
 - Check that data is actually stored
 - Check instructions for participants
 - **—** ...
- Exactly tell participants what they need to do
- Observe that participants do exactly what you told them they should do
- Do a warm-up task, so that participants can familiarze with everything



Ethics

- Effort for participants need to be in line with insights that you gain
 - Evaluating teaching methods
 - Evaluating medicine
- Ensure anonymity of participants
- Be nice to your participants, they voluntarily invest their time



Learning Goals

- Design good research hypotheses
- Design an experiment with high internal validity and high external validity



Task

- Following statements:
 - Program in Java can be easier debugged
 - Learning to program is the most easy with Haskell
 - Novice programmers should start with object orientation
 - **–** ...

• Define:

- Hyptoheses
- Dependent and independent variables and operationilzation
- Confounding factors and how to control them
- Experimental design



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