

# Better Empirical Science for Software Engineering

How not to get your empirical study rejected:  
we should have followed this advice

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Invited Presentation

International Conference on Software Engineering May 2006

# Motivation for this presentation

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- There is not enough good empirical work appearing in top SE conference venues
- Our goal is to help authors and reviewers of top SE venues improve this situation

# Presentation structure

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- Discuss the state of the art in empirical studies in software engineering
  
- Debate problems and expectations for papers with empirical components in top SE conference venues

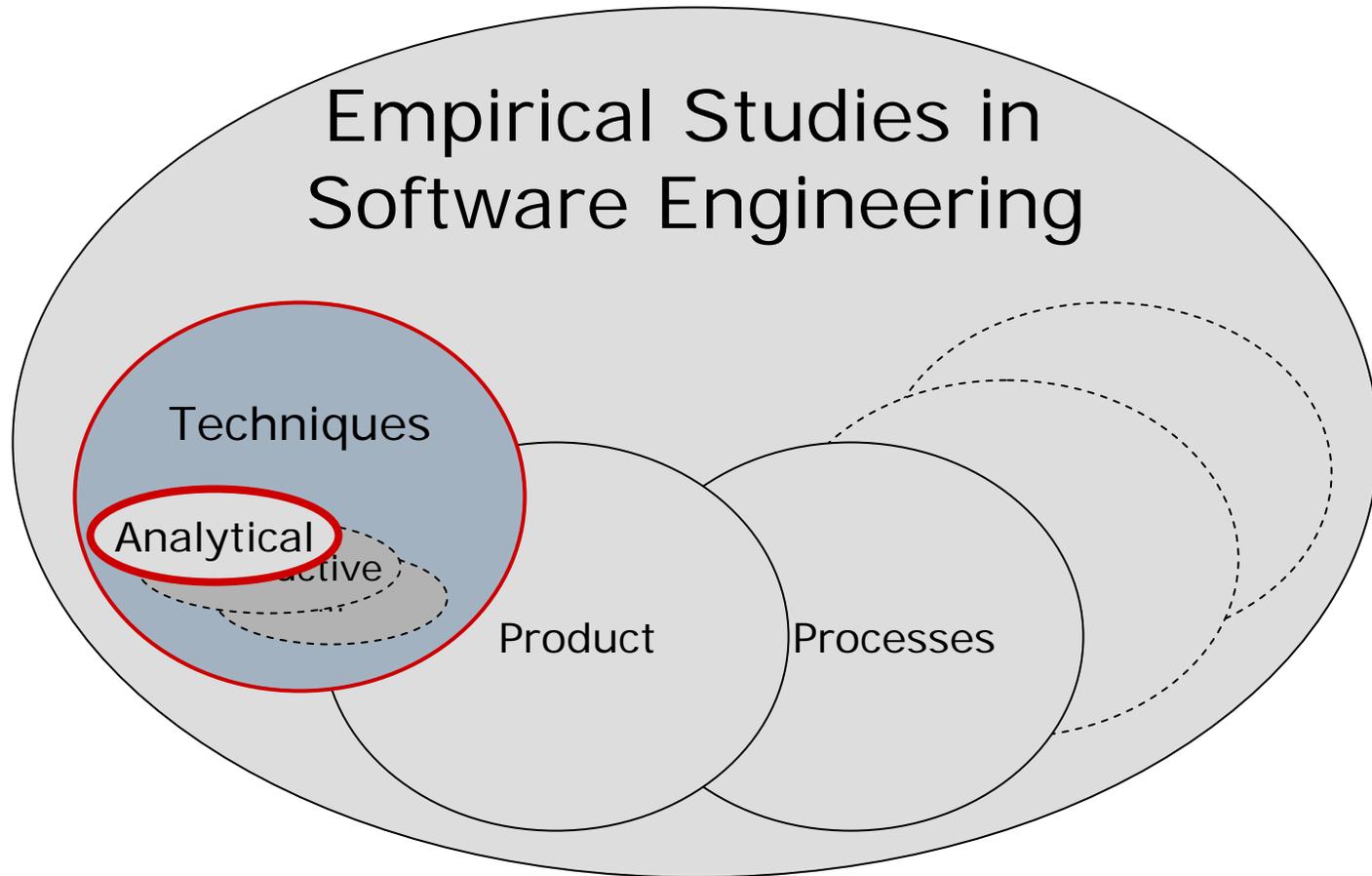
# What is an empirical study?

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*Empirical study* in software engineering is the scientific use of quantitative and qualitative data to understand and improve the software product and software development process.

# What are we studying?

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# Why study techniques empirically?

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- Aid the technique developer in
  - Demonstrating the feasibility of the technique
  - Identifying bounds and limits
  - Evolving and improving the technique
  - Providing direction for future work
- Aid the user of the technique in
  - Gaining confidence of its maturity for context
  - Knowing when, why and how to use it
- To learn and build knowledge

# How to study a technique?

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1. Identify interesting problem
2. Characterize and scope problem (stakeholders, context, impact, ...)
3. Select, develop, or tailor techniques to solve a part of problem
4. Perform **studies** to assess technique on a given artifact (feasibility, effectiveness, limits,...)
5. Evolve the studies (vary context, artifacts, ... and aggregate)

*Repeat steps as necessary **and disseminate results!***

# Why is repetition necessary?

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## Need accumulative evidence

- Each study is limited by goals, context, controls, ...
- Families of studies are required
  - Varying goals, context, approaches, types of studies, ...
  - Increase confidence, grow knowledge over time

## Need to disseminate studies

- Each paper is limited by length, scope, audience, ...
- Families of papers are required
  - Gain confidence through replications across community
  - Move faster or more meaningfully by leveraging existing work to drive future research

# Studies of Techniques

## Large variation across community

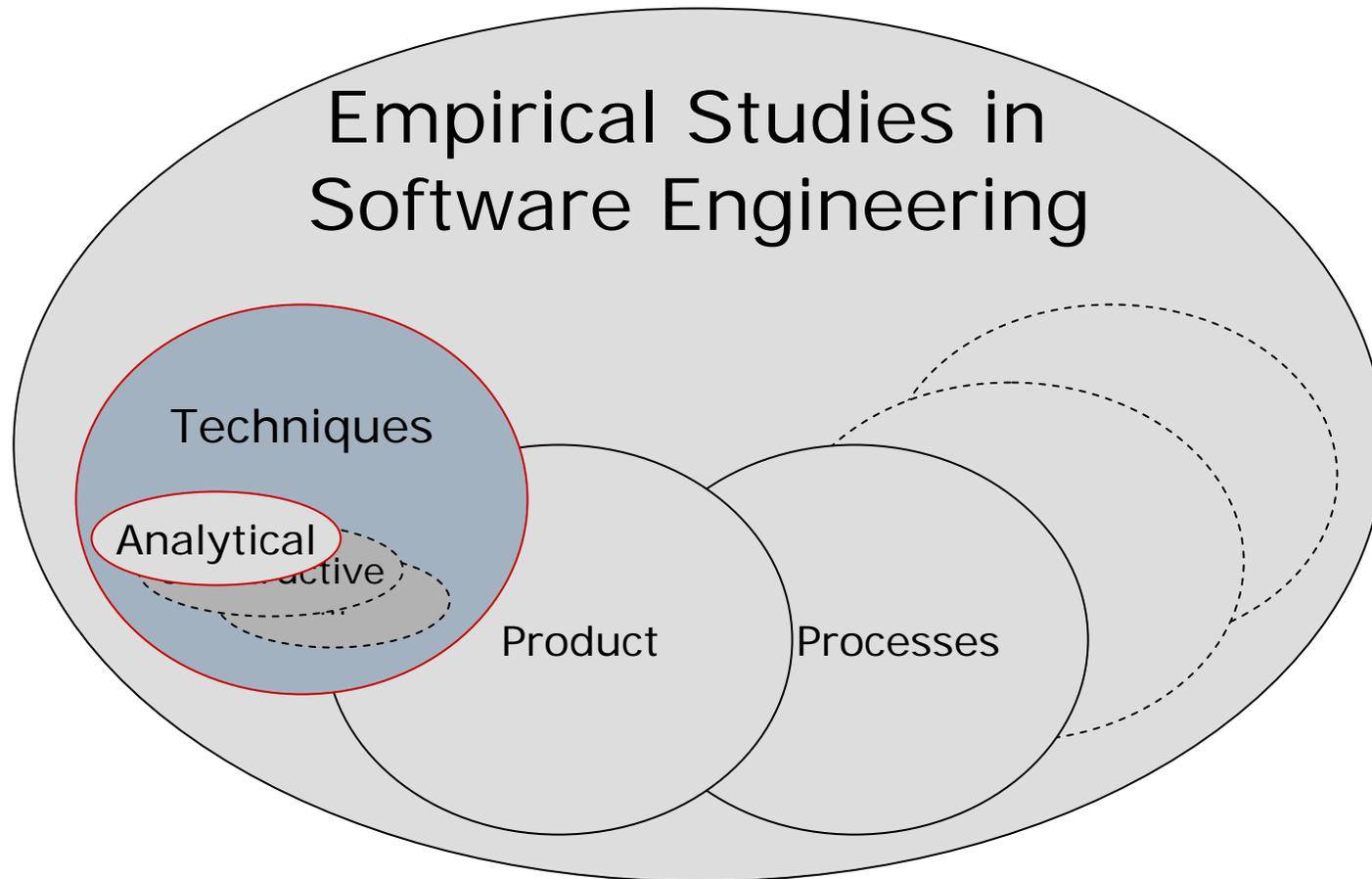
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- Is the human part of the study?
- What are the bounds on sample size?
- What is the cost per sample?
- What are the interests, levels of abstraction, model building techniques?
- What types of studies are used, e.g., qualitative, quantitative, quasi-experiments, controlled experiments?
- How mature is the area?

# Studies of Techniques

## Two Examples

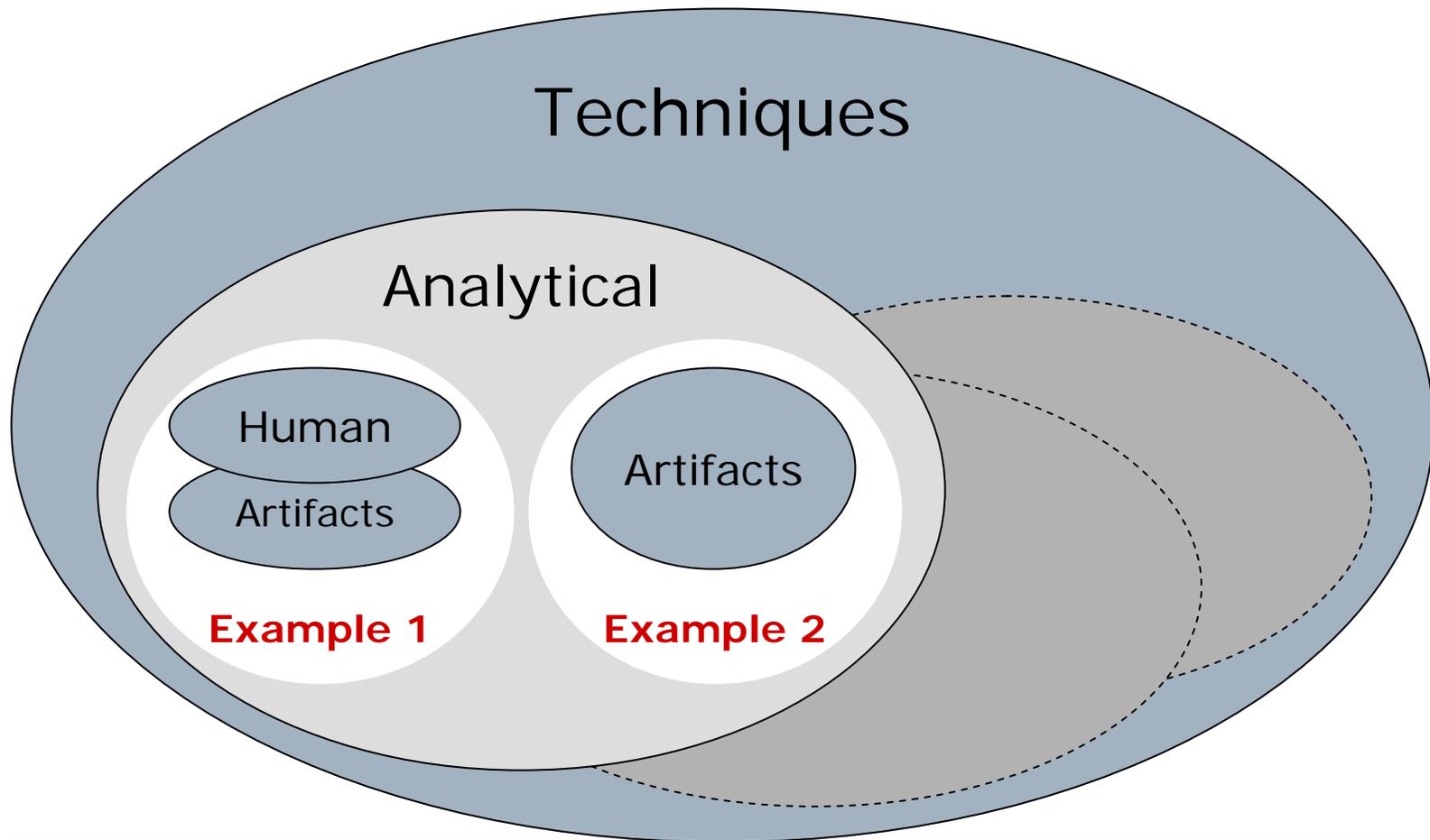
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# Studies of Techniques

## Two Examples

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# Example 1: Human Based Study on an analytic technique

## **Evaluating a code reading technique**

Initial version: rejected for ICSE 1984

Invited Talk: American Statistical  
Association Conference, July 1984

Published TSE 1987 (after much discussion)

# A study with human subjects

## Question and Motivation

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- Is a particular *code reading* technique effective?
  - State clearly what questions the investigation is intended to address and how you will address them, even if the study is exploratory.
  - How does it compare to various testing techniques?
  - Try to design your study so you maximize the number of questions asked in that particular study, if you can.
  - What can you discover?
  - What is the effect of experience, product type, ...?

# A study with human subjects

## Context and Population

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### Environment:

NASA/CSC

Text form

Seeded w

145 - 36

Specify as much context as possible... this is often hard to do so in a short conference paper.

database

### Experiment

Fractional

Three applications

74 subjects: 32 NASA/CSC, 42 UM

Student studies offer a lot of insights. This led to new questions for professional developers.

# A study with human subjects

## Variables and Metrics

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Independent (the technique)

Code Reading

Technique definition and process conformance need to be carefully specified in human studies.

Given: Spec

Functional Test

Given: Spec and Executables

Structural Testing: % statement coverage

Given: Source, Executables, Coverage tool, then spec

Dependent (effectiveness)

fault detection effectiveness, fault detection cost, classes of faults detected

# A study with human subjects

## Controlling Variation

		<u>Code Reading</u>			<u>Functional Testing</u>			<u>Structural Testing</u>											
		P1	P2	P3	P1	P2	P3	P1	P2	P3									
Advanced Subjects	S1	<div style="background-color: yellow; padding: 10px; border: 1px solid black;">                     The more people you can get to review your design, the better.                      It is easy to miss important points.                 </div>																	
	S2																		X
	S8																		
Inter- mediate Subjects	S9			X			X			X									
	S10		X		X														
	S19	<div style="background-color: yellow; padding: 10px; border: 1px solid black;">                     It is easy to contaminate subjects.                      It is hard to compare a new technique against the current technique.                 </div>									X								
S20																			
S21																		X	
Junior Subjects	S32	X				X			X										

**Blocking according to experience level and program tested**  
**Each subject uses each technique and tests each program**

# A study with human subjects

## Quantitative Results (NASA/CSC)

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### Fault Detection Effectiveness

- Code reading > (functional > structural)

### Fault

- Code reading > (functional ~ structural)

Student Study had weaker results but showed similar trends.

### Classes of Faults Detected

- Interface:

- code reading > (functional ~ structural)

- Control:

- functional > (code reading ~ structural)

# A study with human subjects

## Qualitative Results (NASA/CSC)

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- Code reviews performed by participants rated their performance better than when they were not performing best
- Participants believed that their code was working better than it actually was
- When inspections were applied on a live project, it was found that, if any errors were found, they were more likely to be found by the inspectors than by the programmers themselves
- Threat to validity.
  - External Validity: Generalization interaction of error
- Study
  - 32 professional programmers for 3 days

Empirical studies are important even when you believe the results should be self-evident.

It may be difficult to generalize from in vitro to in vivo.

Human subject studies are expensive. You cannot easily repeat studies.

# A study with human subjects

## New Ideas (NASA/CSC)

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- Reading using a defined technique is more effective than a test.  
It is important to make clear the practical importance of results independent of the statistical significance.
- Different techniques may be more effective than tests.  
Don't expect perfection or decisive answers. For example, insights about context variables alone are valuable.
- The reading technique may be different from the reading method.

# Studies with human subjects

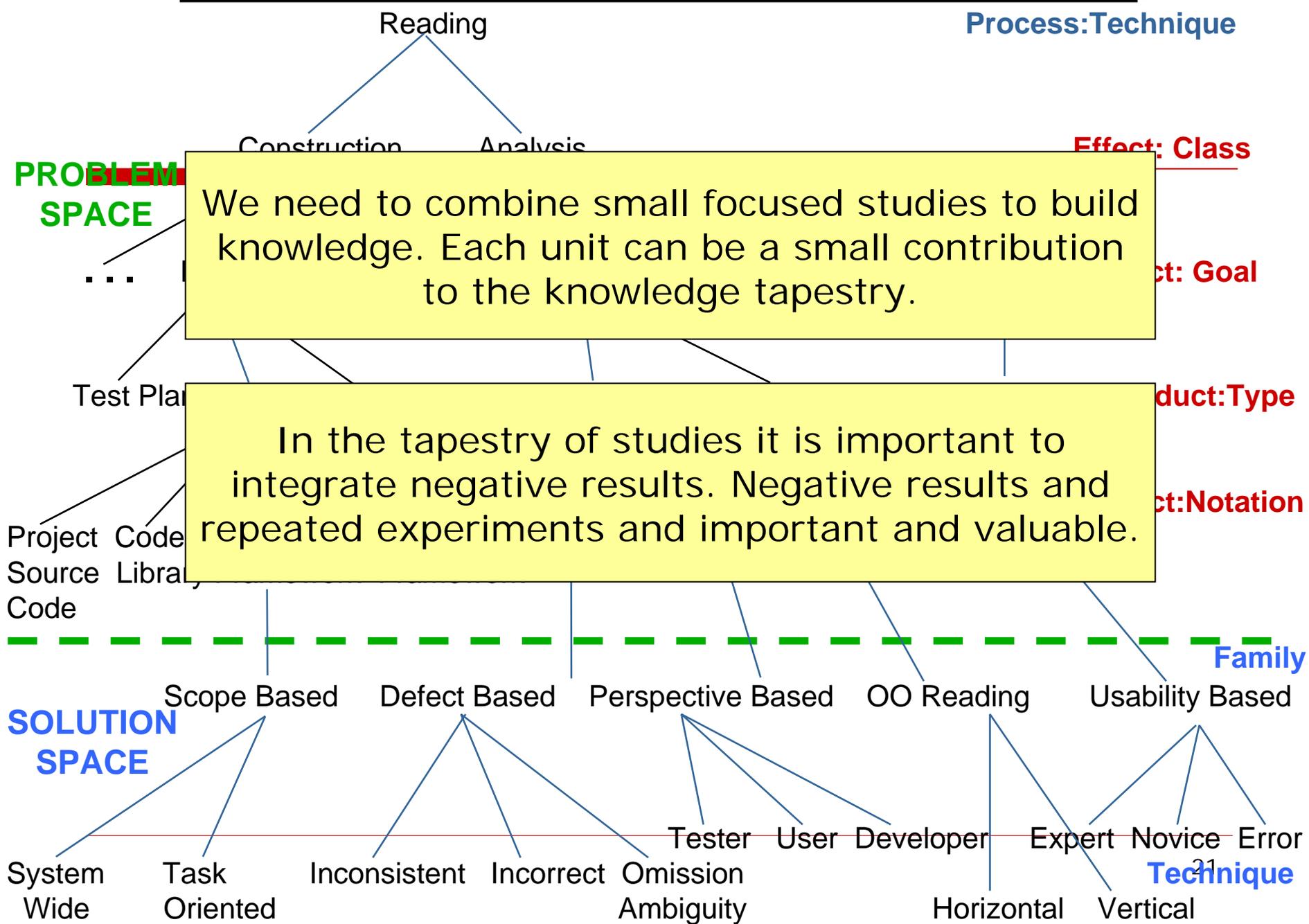
## Evolution of Studies

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Each study opens new questions.  
Scaling up is difficult and the empirical methods change.

		One	More than one
# of Teams	One	<b>3. Cleanroom (SEL Project 1)</b>	<b>4. Cleanroom (SEL Projects, 2,3,4,...)</b>
	per Project	<b>2. Cleanroom at Maryland</b>	<b>1. Reading vs. Testing 5. Scenario reading vs. ...</b>

# Evolution of Studies: Families of Reading Techniques



# Example 2: Artifact Based, Analytic

The Impact of Test Suite Granularity on the  
Cost Effectiveness of Regression Testing  
(ICSE 2002)

Evaluating the effects of test suite composition  
(TOSEM 2004)

# A study with artifacts

## Question and Motivation

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### □ How do we compose test suites?

- Separate beliefs from knowledge.

### □ Why

Experience can help to shape interesting and meaningful conjectures.

**Bor** *More tests than to do the job with fewer, grander tests.*

**Cem Kaner:** *Large tests save time if they aren't too complicated; otherwise, simpler tests are more efficient.*

**James Bach:** *Small tests cause fewer cascading errors, but large tests are better at exposing system level failures involving interactions.*

# A study with artifacts

## Context and Population

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### Context

- Identify context that is likely to have greatest impact!
- What tests should we re-run?
- In what order should we re-run them?

sion)

### Population

- We do not have a good idea of our populations...  
but this should not stop us from specifying scope of findings.
- Seeded faults
  - Non-seeded versions were the oracles
- Test suite
  - Original + enhanced

# A study with artifacts

## Type of Study

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### □ Family of controlled experiments

- Conjectures should lead to more formal and (likely more constrained) hypotheses. *ng*

- Measure effects on

- Carefully identify and explain dependent, independent, and fixed variables.

- *Does granularity and grouping matter?*

### □ High levels of controls

- Process, execution, replicability

# A study with artifacts

## Controlling sources of variation

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### □ Controlled manipulation



Controlling is not just about the chosen experimental design, is also about controlling noise so that we really measure the desired variables.

*of tests*

1. Start with a given test suite
2. Partition in test grains
3. To generate test suite of granularity  $k$   
Select  $k$  grains from pool

# A study with artifacts

## Controlling sources of variation

### □ Experimental designs

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Once automated, application of treatment to units is inexpensive. We can get many observations quickly and inexpensively.

ity,  
els, ...

Test Case Selection				Test Case Prioritization				
All				Feedback				
Granula				Granularity				
G1	G2	G4		G1	G2	G4	G8	G16

Provide detailed definition of data collection process, including costs and constrains that justify choices.

Empire  
(10 versions)

Bash  
(10 versions)

# A study with artifacts

## Analysis and Results

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### □ Analysis

- Richness of results may be in interactions between factors. Question is not really about "does it matter?" but "when does it matter?"

### □ Results

- Combine exploratory and formal data analysis.
- Test suite fault detection effectiveness improved at
  - Coarse granularity but only for easy-to-detect faults
  - Fine granularity when faults were detected by single grains

# A study with artifacts

## Qualified Implications

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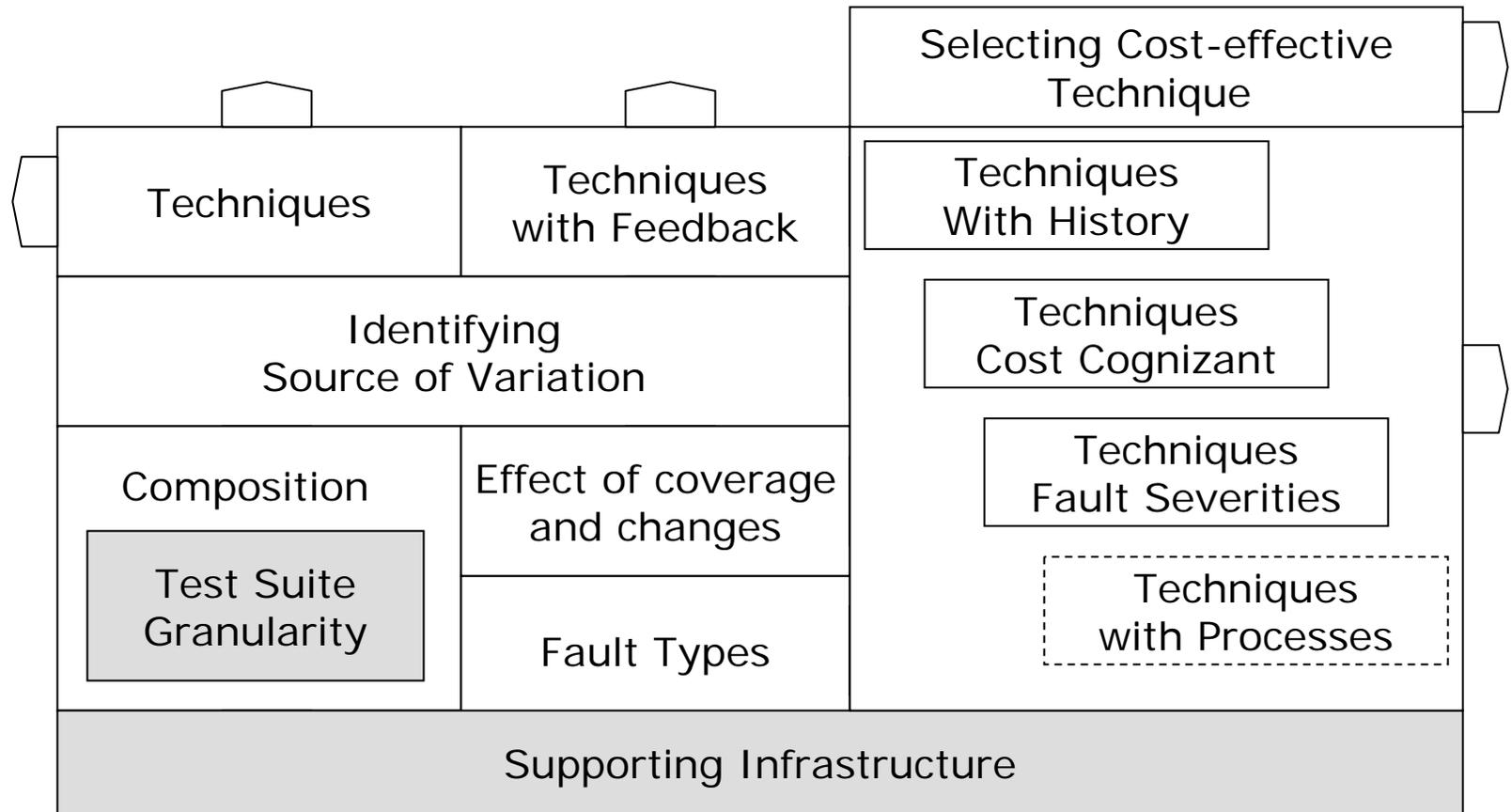
- Test suite ~~comp. mattered, especially for extremes~~
- But 

Keep “chain of significance” throughout the paper. Close with “distilled implications”.

  - Hard-to-detect faults
  - Aggressive test case selection or reduction techniques
- Threats
- Generalizations
  - Early testing, significant program changes: coarser suites
  - Mature stage, stable product: finer granularity

# A study with artifacts

## Building a Family for Regression Test Case Prioritization



A 6 year lifespan, over 15 researchers from many institutions,  
building knowledge incrementally.

# Looking at Some Recurring Issues

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- ❑ What is the target and scope?
- ❑ What is representative?
- ❑ What is an appropriate sample?
- ❑ What are the sources of variation?
- ❑ What infrastructure is needed?

# Recurring Issues

What is the **target and scope**?

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## □ With humans

- Effect of people applying technique
- Costly. Little margin for error in a single study
- Hard to replicate, context variables critical

## □ With artifacts

- Effect of technique on various artifacts
- Summative evaluations, confirmatory studies
- Replicable through infrastructure/automation

# Recurring Issues

## What is representative?

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### □ With humans

- **Participants'** ability, experience, motivation, ...
- **Technique** type, level of specificity,...
- **Context** for technique application

### □ With artifacts and humans

- **Product:** domain, complexity, changes, docs, ..
- **Fault:** actual or seeded, target, protocols, ...
- **Test Suite:** unit or system, original or generated,
- **Specifications:** notation, type of properties, ...
- ...

# Recurring Issues

What is an appropriate **sample**?

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- **With humans:** mostly opportunistic
  - Small data samples
  - Learning effect issues
  - Unknown underlying distributions
  - Potentially huge variations in behavior
- **With artifacts:** previously used artifacts/testbeds
  - Reusing “toy” examples to enable comparisons
  - Available test beds for some dynamic analysis
  - Not natural occurring phenomenon

# Recurring Issues

What are the **sources of variation**?

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## □ With humans

- Learning and maturation
- Motivation and training
- Process conformance and domain understanding
- Hawthorn Effect

## □ With artifacts

- Setup/clean residual effects
- Perturbations caused by program profiling
- Non-deterministic behavior

# Recurring Issues

How objective can we be?

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- Comparing a new technique with
  - Current practices is hard without contaminating subjects
  - Other techniques on same test bed can be suspect to “tweaking”
- Ideal is not to have a vested interested in techniques we are studying
  - But we are in the best position to identify problems and suggest solutions

# Recurring Issues

## How do we support empirical studies?

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- Need for infrastructure
  - Test beds are set of artifacts and support for running experiments
  - Testbeds are applicable to **limited** classes of techniques → need many testbeds
  - **Costly but necessary**
  - How do we **share** and **evolve** infrastructures?

# Success Story

## Aiding the Empirical Researcher

Software-artifact Infrastructure Repository

Search for Objects

Java C

Minimum Version Count

Maximum Version

Display

- Goal is to support controlled experimentation on
- Static and dynamic program analysis techniques
- Programs with faults, versions, tests, specs, ...
- + 30 institutions are utilizing and helping to evolve SIR!

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1. ant	Download: <a href="#">all platforms</a>	Updated: 2005-11-22 Downloads: 2 SIR Version: 1.0
<b>Language</b>	Java	
<b>Test Types</b>	unit	
<b>Fault Types</b>	seeded	
<b>Fault Matrices</b>	yes; unit class-level, method-level	
<b>Sequential Versions</b>	11	
<b>Size</b>	80500 LOC, 627 classes	
<b>Acknowledgements</b>		
2. galileo	Download: <a href="#">all platforms</a>	Updated: 2005-11-22 Downloads: 0 SIR Version: 1.0
<b>Language</b>	Java	
<b>Test Types</b>	tsl	
<b>Fault Types</b>	seeded	
<b>Fault Matrices</b>	no	
<b>Sequential Versions</b>	16	
<b>Size</b>	15200 LOC, 79 classes	
<b>Acknowledgements</b>		
3. jmeter	Download: <a href="#">all platforms</a>	Updated: 2005-11-22
<b>Language</b>	Java	

# Success Story:

## Aiding the Technique Developer

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- **Testbed** : TSAFE -a safety critical air traffic control software component

Trying out a technique on a testbed

- helps identify its bounds and limits
- focuses the improvement opportunities

- - provides a context for its interaction with other techniques
- helps build the body of knowledge about the class of technique

- **Results:** The experimental study resulted in a

- Better fault classification
- Identified strengths and weaknesses of the technology
- Helped improve the design for verification approach
- Recognized one type of fault that could not be caught

ign  
in

# Success Story:

## Aiding the Technique User

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- **Testbed** : a variety of class projects for high performance computing artifacts at UM, MIT, USC, UCSB, UCSD, MSU
  - **Evalu** models, e.g., t
  - **Result** It is important to build a body of evidence about a domain, based upon experience, recognizing what works and doesn't work under what conditions
  - On ce
  - Op
  - UPC/CAF requires around 5-35% less effort than OpenMP
  - XMT-C requires around 50% less effort than MPI.
  - For certain kinds of embarrassingly parallel problems, message-passing requires less effort than threaded.
  - The type of communication pattern does not have an impact on the difference in effort across programming models.
-

# Motivation for this presentation

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- Discuss the state of the art in empirical studies in software engineering
- Debate problems and expectations for papers with empirical components in top SE conference venues

# For the Author:

## How do we deal with reviews?

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- Like with any other review
  - The reviewer is right
  - The reviewer has misunderstood something
    - We led them astray
    - They went astray by themselves
  - The reviewer is wrong

# Review example

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"It is **well-known** that shared memory is easier to program than distributed memory (message passing). So well known is this, that numerous attempts exist to overcome the drawbacks of distributed memory."

- Issue: How do you argue that empirical evidence about known ideas is of value?

# Review example

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“... it is hard to grasp, from the way the results are presented, **what is the practical significance of the results**. This is mostly due to the fact that the analysis focuses on statistical significance and leaves practical significance aside. Though this, with substantial effort, can partially be retrieved from tables and figures, this burden should not be put on the reader.”

- Issue: analysis/results disconnected from practical goals

# Review example

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“There are two groups in the study with effective sizes of 13 and 14 observations. As the authors point out, the phenomena under study would need samples of more like 40 to 60 subjects given the variance observed. Thus the preferred approach would have been to either **treat this study as a pilot, or to obtain data from other like studies** to establish the needed sample size for the power needed.”

- Issue: How do you present and justify your empirical strategy?

# Review example

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“... (The technique) was tried on a single form page on five web applications. This is actually **quite a limited experiment**. Web sites such as those they mention have thousands of pages, and hundreds of those with forms. Perhaps a more extensive study would have produced more interesting results. ”

- Issue: how much evidence is enough?
  - Depends on ideas maturity and sub-community empirical expertise

# Review example

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“the population of inexperienced programmers make it likely that results may be quite different for expert population or more varied tasks”

- Issue: Are empirical studies of students of value?

# Review example

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“... **It is well-known** that the composition of the original test suite has a huge impact on the regression test suite. The authors say that they created test cases using the category partition method. Why was only one suite generated for each program? Perhaps **it would be better to generate several** test suites, and consider the variances.”

- Issue: what factors can and should be controlled?
  - We cannot control them all.
  - Tradeoffs: cost, control, representativeness

# Review example

---

- “The basic approach suggested in this paper is very labour intensive. *There would appear to be other less labour intensive approaches* that were not considered ... You have not presented a strong argument to confirm that your approach is really necessary.
  
- Issue: Have the steps been justified against alternatives?

# Review example

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“ ... This paper represents a solid contribution, even though the technique is lightweight ... 6 of the 10 submitted pages are about results, analysis of the results, discussion ... with only a single page required for the authors to describe their approach. Thus, the technique is straightforward and might be construed as lightweight! .”

- Issue: is there such a thing as too much “study” of a straightforward technique?

# From our experience

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- Ask questions that matter
  - Why do they matter? To Who? When?
- State tradeoffs and threats
  - Control versus exposure
  - Cost versus representativeness
  - Constructs versus variables
- Solicit/share expertise/resources with
  - Authors (as a reviewer)
  - Readers (as an author)
  - Researchers (as a researcher)
- Maintain chain of significance
  - Conjecture, Impact, Results, Impact, Conjecture

# For authors and reviewers

## Checklists

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One example: “Preliminary Guidelines for Empirical Research in Software Engineering” by B. Kitchenham et al. TSE 02

Relevant to previous reviews

- *Differentiate between statistical significance and practical importance.*
- *Be sure to specify as much of the context as possible.*
- *If the research is exploratory, state clearly and, prior to data analysis, what questions the investigation is intended to address, and how it will address them.*
- *If you cannot avoid evaluating your own work, then make explicit any vested interests (including your sources of support), and report what you have done to minimize bias.*
- *Justify the choice of outcome measures in terms of their relevance to the objectives of the empirical study.*

# For the Reviewer

Hints for Reviewing SE Empirical Work - Tichy, EMSE 2000

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- Don't expect perfection
- Don't expect a chapter of a statistics book
- Don't expect decisive answers
- Don't reject "obvious" results
- Don't be casual about asking authors to redo their experiment
- Don't dismiss a paper merely for using students as subjects (or small programs)
- Don't reject negative results
- Don't reject repetition of experiments

# Advice from our studies: About overall design

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- ❑ State clearly what questions the investigation is intended to address and how you will address them, especially if the study is exploratory
- ❑ Justify your methodology and the particular steps
- ❑ Justify your selection of dependent variables
- ❑ Try to design your study so you maximize the number of questions asked in that particular study
- ❑ Make clear the practical importance of the results independent of the statistical significance
- ❑ Specify as much context as possible; it is often hard to do so in a short conference paper
- ❑ The more people you can get to review your design, the better, it is easy to miss important points.

# Advice from our studies:

## About scope, sample, representation

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- ❑ Student studies can show trends that are of real value
- ❑ Student studies offer a lot of insights leading to improved questions for professional developers
- ❑ It is easy to contaminate subjects in human studies
- ❑ It is hard to compare a new technique against the current technique
- ❑ Technique definition and process conformance need to be carefully specified in human studies
- ❑ Human subject studies are expensive. You cannot easily repeat studies.
- ❑ Don't expect perfection of decisive answers, for example, insights about context variables alone are valuable

# Advice from our studies: About building a body of knowledge

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- ❑ Empirical studies are important even when you believe the results should be self-evident
- ❑ It may be difficult to generalize from in vitro to in vivo  
It is important to make clear the practical importance of the results independent of the statistical significance
- ❑ Each study open new questions scaling up is difficult and the empirical methods change
- ❑ We need to combine small focused studies to build knowledge, each unit can be a small contribution to the knowledge tapestry
- ❑ In the tapestry of studies it is important to integrate negative results; negative results and repeated experiments are important and valuable

# Improving the odds of getting a paper accepted at a conference

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- Define a complete story (motivation, design, analysis, results, practical relevance)
- Achieve a balance among the
  - Control on the context
  - Generalization of the findings
  - Level of detail in a 10 page paper
- Get as many reviews beforehand as possible

# Better Empirical Science for Software Engineering

How not to get your empirical study rejected:  
we should have followed this advice

Victor Basili



Sebastian Elbaum



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