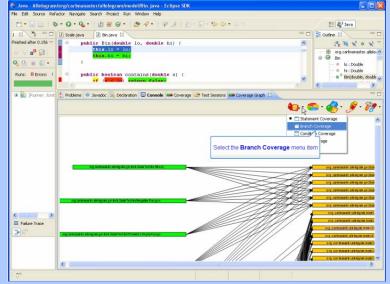


# An Empirical Evaluation to Study Benefits of Visual versus Textual Test Coverage Information

<u>Vahid Garousi</u> Negar Koochakzadeh

Software Quality Engineering Research Group University of Calgary, Canada





#### Acknowledging funding and support from:











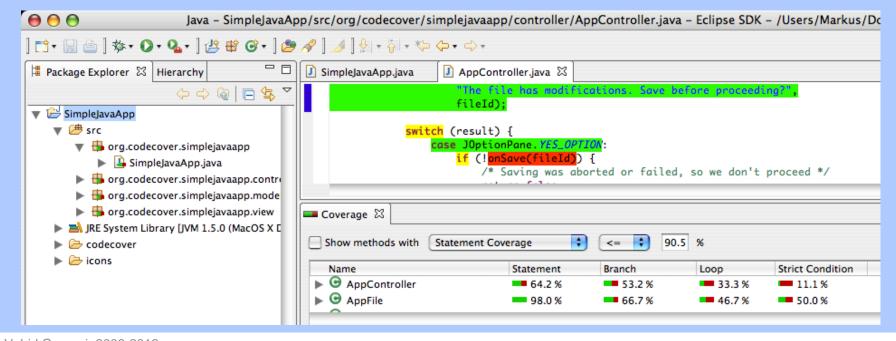
- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A

# **Background and Motivations: Existing Code Coverage Tools**



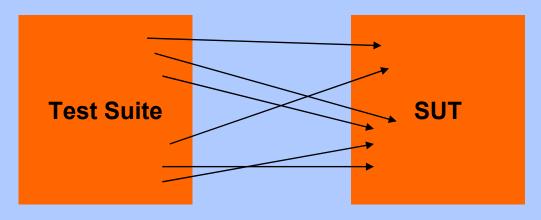
- To support automated code coverage measurement and analysis...
- test coverage values are conventionally shown in percentages and are visualized by progress-bar-like green/red boxes in the existing coverage tools
- e.g., the CodeCover plug-in for the Eclipse IDE



# **Background and Motivations: However... (The need for Test Visualization)**

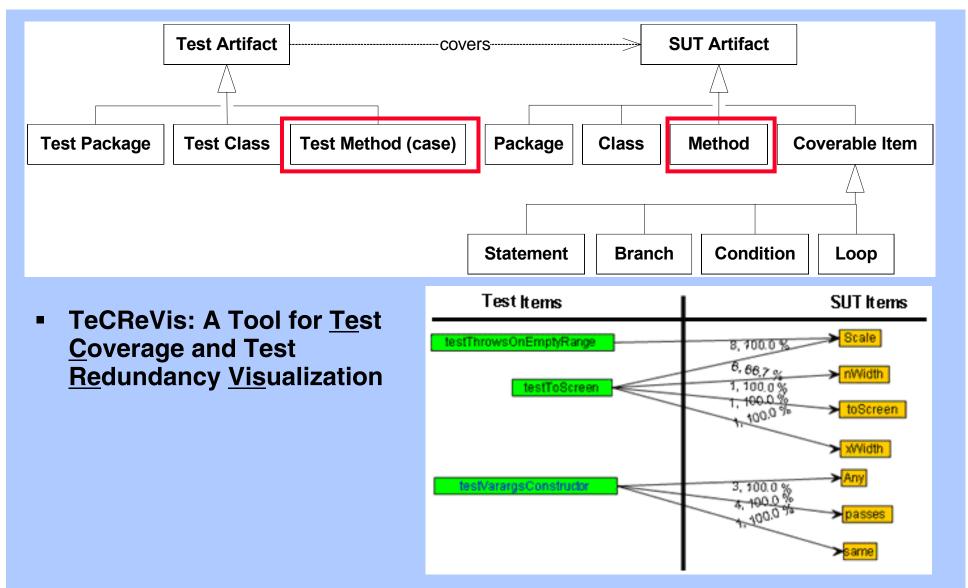


- However with increasing size and complexity of code bases of both systems under test and also their automated test suites (e.g., based on JUnit)
- there is a need for visualization techniques to enable testers to analyze code coverage in "higher" levels of abstraction and in holistic manners
- e.g., which packages of the SUT are covered by a specific set of test cases? Two domains...



# Background and Motivations: We have developed a tool to do that (an Eclipse plug-in)







- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A

# **Empirical Study - Goal**

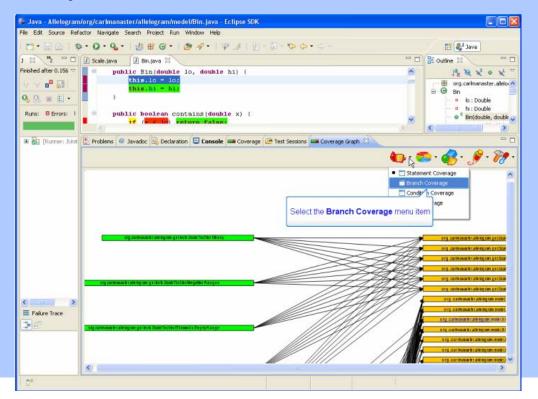


- We wanted to conduct an Empirical Evaluation to study benefits of visual versus textual test coverage information
- and to assess the usability, effectiveness and usefulness of our tool in unit testing and test maintenance tasks
- The goal (using the GQM template):
- To analyze the benefits of test coverage visualization, for the purpose of evaluating its effectiveness on fault localization from the point of view of project managers and software testers in the context of software maintenance.

### **Research Question**



Does the TeCReVis tool help human testers on average to localize faults more efficiently compared to the use of conventional code-coverage tools (which show only textual and progress-bar like coverage information)?





- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A

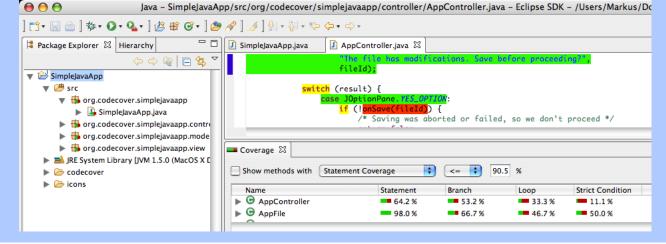
# **Empirical Study - Setup**



- Subjects: Eight graduate students (studying at the University of Calgary) in the field of software engineering
- The eight participants were divided into two groups
- TeCReVis was available only for the experimental group

while the control group used the CodeCover coverage

tool



# **Empirical Study - Setup**



- In grouping the participants, we utilized rigorous methods as defined by empirical software engineering experts
- e.g., random assignment and careful blocking
- We did our best to make sure that the accumulative testing knowledge and experience of both groups were almost equal
- Hypothesis (H1): TeCReVis helps human testers on average to localize faults more efficiently.
- Null Hypothesis (H0): TeCReVis does not assist human testers with fault localization.

# A Metric to measure Fault Localization Efficiency



$$FLE(d) = \sum_{i=1}^{n} \frac{1}{t_i}$$

- d is a human debugger and ti is the amount of time that he/she has spent to locate the i-th fault.
- More time spent would result in less efficiency.



- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A





- An open-source ATM machine simulation software
- 2,541 Java LOC



# **Object of the Study**

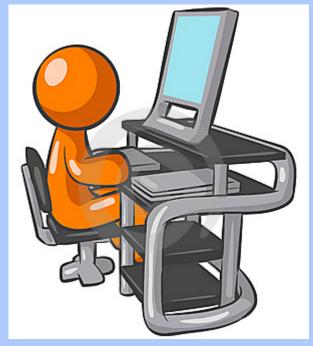


- To perform the fault localization process, we slightly revised this system by injecting into it three (realistic) faults.
- Since there was no unit test suite provided with the ATM implementation online, we created a test suite (containing 23 JUnit test methods) for version 1 of this system.
- This test suite was constructed to achieve full path coverage on the SUT's UML state-chart diagram.
- For replicability purposes, all of the developed JUnit test suite and the system's UML design models are available online. (see the URL in the paper)

# **Empirical Study - Execution**



- Participants were asked to find and locate three injected faults in the ATM system.
- Participants were asked to report the time of locating each fault, which were analyzed later by the authors to measure fault localization efficiency.





- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A

# **Results of the Experiment**

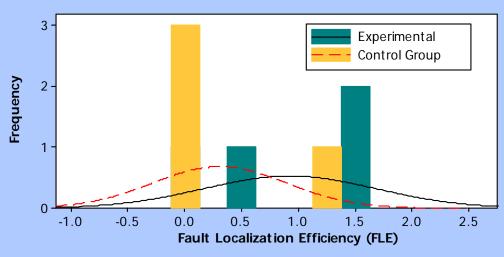


	All time	values are in r	minutes	(FLE)
P1	20	2	1	1.55
P2	24	*	*	0.04
P3	18	1	2	1.55
P4	23	2	*	0.54
P5 Control Group	*	*	*	0
	27	*	*	0.03
		7	1	1.18
		-	-	0
	P2 P3 P4 P5	P1 20 P2 24 P3 18 P4 23 P5 * P6 27 P7 22	P1       20       2         P2       24       *         P3       18       1         P4       23       2         P5       *       *         P6       27       *         P7       22       7	P1       20       2       1         P2       24       *       *         P3       18       1       2         P4       23       2       *         P5       *       *       *         P6       27       *       *         P7       22       7       1

### Results of the Experiment



- t-test was applied.
- Two types of experiment errors (α and β) were as follows:
- $\alpha$ =0.12 and  $\beta$ =0.47 (pass if only  $\alpha$ <0.05)
- Reminder: α = P(H0 is rejected | H0 is true) and β = P(H0 is accepted | H0 is false).
- → Null hypothesis (H0) cannot be rejected
- → It is possible to say with confidence that TeCReVis helps human testers on average to localize faults more efficiently.



#### **Lessons Learned and Future Works**



- We believe that, although we had tutorial part in our experiment first, learning curve in limited time of performing fault localization task in the experiment has affected our results.
- In other words, learning curve caused less effectiveness of using TeCReVis in localizing faults in limited time.
- All of the participants' answers were supportive of the usefulness of TeCReVis for fault localization.
- For instance, a participant of the experiment group said: "I feel that, in large systems, this graph-based visualization can be very useful".
- Repeating the experiment with more subjects and more control.



- Background and Motivations
- Empirical Study Goal
- Research Question
- Empirical Study Setup
- Object of the Study
- Empirical Study Execution
- Results
- Lessons Learned and Future Works

Q/A