# Comparing the Effectiveness of Automatically Generated Tests by $Randoop,\ JWalk$ and $\mu Java$ with jUnit Tests

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#### Abstract

This paper will explore and compare three automated testing techniques with manually contrived jUnit tests. These techniques and their respective tools are outlined, after which they are applied to a realistically complex system. Their advantages and disadvantages are discussed, together with possible trade-offs and the type of faults they are able to detect. In conclusion, the findings are summarized and suggestions for future research are made.

# 1 Introduction

Testing is essential in ensuring software quality and verifying its correct operation; it comes as no surprise then that the practice is widely applied by programmers and testers alike. However, writing these tests is often a tedious and uninspiring task, which potentially could be made easier by automation. Using test generating tools may save time, since they are able to test more cases and test these more more exhaustively, requiring less time or effort invested by the tester. They may even encounter faults that human testers would never think of, but is the quality of those generated tests on par with a carefully thought out test suite?

This paper will explore and compare three automated testing techniques with manually contrived jUnit tests. First, the selected testing techniques and representative tools are outlined. We then move on to the hypothesis and criteria that will be used when evaluating the tools the section thereafter, followed by the experiments and their results. The relative merits

and disadvantages of every technique are then compared, after which the findings are summarized and suggestions for future research are made.

# 2 Testing Techniques and Tools

Three different testing techniques (and their associated tools) were selected for comparison. They were chosen as to be unalike: every technique employs a different approach to the same problem.

## 2.1 Feedback-Directed Random Testing with Randoop

The  $Randoop^1$  tool requires no additional user input or previously existing test suites, but will automatically generate unit tests for a given set of classes during a limited time interval, which is preset by the tester. It does so by executing random sequences of methods from the class-under-test (CUT). Since the sequences are random, results may vary slightly when examining results from different runs.

Using prior results, more intelligent inputs and sequences will be chosen to avoid redundant or illegal sequences as much as possible, making up the feedback-directed aspect of the tool. For instance, a sequence containing code that will result in an exception is pruned from future runs.[1] The tests that are output can be classified into two categories: first regression tests that capture the current behavior of the code are generated, and secondly, also contract violating tests are produced, which could indicate potentially buggy code.

The author of this tool provides a more detailed and technical explanation of its workings in Sect. 2 of [2].

## 2.2 Lazy Systematic Unit Testing with JWalk

JWalk<sup>2</sup> is a lazy systematic unit testing tool. Lazy indicates that no formal specifications are required, since they will be inferred from the ever changing code by the tool itself. The lack of specifications is typical in agile, or XP development, for which this tool could prove a useful addition. Systematic testing is achieved by executing all transitions between states (caused by sequences of methods) in a replicable, deterministic manner, as opposed to

 $<sup>^1</sup>$ http://people.csail.mit.edu/cpacheco/randoop

<sup>2</sup>http://www.dcs.shef.ac.uk/~ajhs/jwalk/

random testing. The class-under-test is exhaustively exercised until a preset depth is reached.[3]

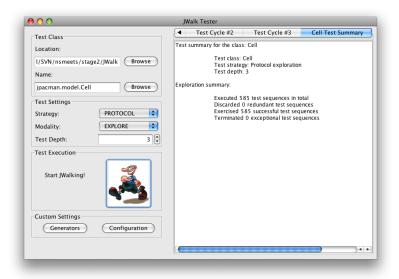


Figure 1: The JWalk tool.

The tool can be used for both exploring the class as for generating an oracle, based on automatic analysis, predictive rules and interaction with the tester. It distinguishes between mutator methods and observer methods, which respectively do and do not cause a change in state in the system, for faster and more reliable results. When the software, and therefore its specification, is modified, JWalk will only prompt for verification of altered properties, in this way providing a regression test for the CUT.

A notable detail is that this tool does not generate unit tests, only reports. [4]

## 2.3 Mutation Testing with $\mu Java$

 $\mu Java^3$  will introduce small modifications, or mutations, to the code, both on class and method level. Since the testing occurs on the implementation level, this is a white box technique. Its intent is to trace code that is rarely called or showing that the existing unit tests are inadequate (when the injected fault is not discovered by the test suite). In a way, this resembles coverage measurements – it also attempts to indicate which code needs improved testing. The difference is that mutation testing can find code that is executed (and thus is recorded as covered), but is not actually tested. Since

<sup>3</sup>http://cs.gmu.edu/~offutt/mujava/

this technique's concept is more the testing of the tests,  $\mu Java$  indeed does require preexisting tests. It will only generate a modified implementation of the system.[5] A screenshot is presented in Figure 2.

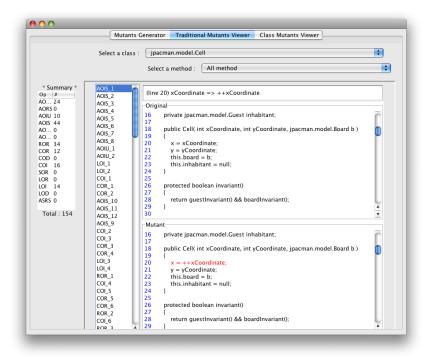


Figure 2: The  $\mu Java$  Mutant Generator screen.

# 3 Hypothesis and Criteria

As previously stated, it may prove interesting to see how well test generating tools perform when compared to an actual human tester. Though it is highly unlikely that any tool could ever fully replace an experienced software tester, it just might make the latter one's job easier.

Obviously, a sample software system is needed when comparing the tools. We opted for *JPacman*, an implementation of the famed game in *Java*, that is used as a teaching example at the Delft University of Technology. It was chosen for its well-documented unit and integration tests, known to be of sound quality from previous experience in testing class. Moreover, the system is one of relative complexity and thus serves as a more realistic sample system as opposed to e.g. a mere stack implementation.

Coverage is an unbiased, verifiable and easily measured criterion when using

tools such as  $Emma^4$ . It is however unsuitable in this case, since neither JWalk nor  $\mu Java$  actually generate unit tests. Besides,  $\mu Java$  is in a way similar to measuring coverage, since it indicates which sections of the code are insufficiently executed by tests, if executed at all.

Time (or effort) spent seems like a good alternative, but it is much more problematic to quantify precisely. For instance, measurements of the time invested in composing the non-generated *jUnit* tests are not available and impartially estimating the actual effort spent learning the tools and reading papers are particularly hard problems. Since running the tests usually happens very quickly, it is also not straightforward to reliably measure how long this takes. It is also biased, seeing that *Randoop* generates tests in a preset time limit, as opposed to the other tools.

Therefore, comparisons will primarily be made based on the type of faults the tools can detect. Further points of distinction may be the amount of extra work a tester has to invest when using one of these tools and maintainability of any software that needs to be manually written by the tester, or that was automatically generated by the tool. A potential user can then weigh the pros and cons of usage of the tool.

# 4 Experiments

## 4.1 Randoop

Both easy in installation and use, the *Randoop* tool randomly generates interleavings of methods, bounded by time, not depth. The unit tests it like so generates, essentially capture the current behavior of the class-under-test. The tool was ran with a time limit of 1 second, 10 seconds and 30 seconds respectively. Coverage measurements using Emma were conducted to get a quick overview of the quality of the generated regression tests. Figure 3 shows coverage achieved when running the test suite generated by *Randoop* in 30 seconds. Overall coverage in the jpacman.model package when using *Randoop*'s default 10 seconds setting is only slightly less (49.2%) than when using 30 seconds (50.5%). As can be expected, 1 second yielded far worse results (39.9% coverage).

Randoop's main attraction is the automated generation of a full regression test suite. This can be helpful when working on a project that has no tests whatsoever and for which it is crucial to ensure current behavior is preserved

<sup>4</sup>http://emma.sourceforge.net/

Element	Coverage	Covered Instructions	Total Instructions
▼ ⊕ jpacman.model	50.5 %	1379	2730
▶ 🗓 Board.java	40.5 %	133	328
▶ 🗓 Cell.java	39.7 %	122	307
Ingine.java	51.7%	231	447
▶ 🗓 Food.java	55.4 %	46	83
▶ ☐ Game.java	<b>59.9</b> %	404	674
▶ ☐ Guest.java	47.8 %	64	134
▶ ☑ Monster.java	51.2 %	22	43
▶ ☑ Move.java	52.8 %	181	343
▶       MovingGuest.java	<b>100.0 %</b>	3	3
▶ 🗓 Player.java	43.2 %	70	162
▶ ☑ PlayerMove.java	50.3 %	83	165
▶ ☑ Wall,java	48.8 %	20	41

Figure 3: Coverage achieved by running the test suite generated by Randoop in 30 seconds.

when updating the code. Even though one gets free and effortless regression tests, it comes at a excessively high cost in maintenance. When running Randoop for 30 seconds, it yields some 96,000 lines of testing code, accounting for more than 400 test cases. Some of these cases use just under a hundred variables that are named by the unhelpful convention var\_number, making any deep understanding of the tests practically impossible. Another limitation of Randoop is the lack of support for category partition testing. There is no way for the tester to supply carefully chosen values for arguments.

It is claimed in [1] that the efficiency and quality of the resulting test suite can be improved by using method level annotations and checkers. First, three method level annotations are announced to be available in Randoop: @Omit, @Observer and @ObjectInvariant, each respectively ignoring the method it is added to, indicating the method it is attached to has no side-effects and finally declaring that this parameterless method that returns a boolean value, is actually a class invariant. The latter seemed especially relevant to JPacman since it employs class invariants, but an implementation of this annotation was not found (the first and second ones are available though). Likewise, checkers should provide a way of translating the assert-statements that already occur in the JPacman code into contracts that Randoop can interpret and use. To make matters worse, the class that implements this was also nowhere to be found in this distribution of the tool, nor was there any documentation to be found concerning custom checkers.

The tests cannot be generated with the existing assertions in mind, but they can be switched on when executing the test suite. Some tests will return an AssertionException, as is to be expected, but this only occured in a small fraction of the tests.

#### 4.2 JWalk

JWalk recently hit version 1.0 and was easy to install and comprehend, through clear and detailed instructions on its website<sup>5</sup>. There are multiple options available for analyzing the selected class: first there are three modes, then there are three strategies.

The first mode is the *inspect* mode, which just gives a summarizing overview of the methods in the class. The results depend on which of the three strategies is chosen. The first and simplest strategy, the *protocol* strategy, takes into consideration all available methods in that class and the overview will only distinguish between constructors and public methods. The *algebra* strategy attempts to separate methods that have side-effects (mutator<sup>6</sup> or transformer operations) and observers, methods that do not change the current state of the system, e.g. getters, invariants, etc. Finally, the *states* mode attempts to guess the various high level states the class can be in.

The second mode is the *explore* mode. Here, paths from the constructor and all possible interleavings till the selected depth are generated. Using the protocol strategy, all distinct interleavings of the methods in the class may lead to a potentially different state, so, again, none are pruned. This is the most exhaustive, but also the most time consuming strategy. The algebra strategy applies only distinct sequences of constructors and mutators, allowing for faster results since observer methods need not to be considered. The states strategy attempts to find all abstract design states by executing every unique combination of state predicates over the objects variables.

All three strategies report the amount of discarded and non-redundant sequences that were found, and how many of the latter executed successfully or ended in an exception; an example is shown in Listing 1. All non-redundant sequences and their results can be browsed in the tool.

```
Test summary for the class: Cell
    Test class: Cell
    Test strategy: Algebraic exploration
    Test depth: 3

Exploration summary:
    Executed 9 test sequences in total
    Discarded 576 redundant test sequences
    Exercised 9 successful test sequences
    Terminated 0 exceptional test sequences
```

Listing 1: Example summary for jpacman.model.Cell generated by JWalk, using the algebraic setting and maximum test depth 3.

<sup>&</sup>lt;sup>5</sup>http://www.dcs.shef.ac.uk/~ajhs/jwalk/userguide.html

<sup>&</sup>lt;sup>6</sup>Not to be confused with mutators as employed in  $\mu Java$ .

The third and last mode is the validate mode. Here, the programmer is asked to validate the results of a sequence of methods that is shown in a pop up window. This way, one can concentrate on checking expected results, as opposed to writing unit tests. However, reading, understanding and validating the proposed result (for a sequence of methods taking parameters) is a tedious task at first, especially when a considerable depth has been selected. However, JWalk attempts to predict results, requiring less interaction. Also, when an oracle has been generated and the code is updated, only novel sequences need to be verified. So, effort is only substantial the first run, making this method scalable. If it encounters an unexpected exception, it prompts the tester – this could possibly indicate, for example, a new fault or a broken precondition. An example oracle that was generated by JWalk is printed in Listing 2. (The three strategies are fully analogous to the previous modes.)

```
new(1,2,Board#1)=Cell#1
new(1,2,Board#1).getY()=2
new(1,2,Board#1).getX()=1
new(1,2,Board#1).guestInvariant()=true
new(1,2,Board#1).getInhabitant()=null
new(1,2,Board#1).isOccupied()=false
new(1,2,Board#1).cellAtOffset(5,6)=null
new(1,2,Board#1).getBoard()=Board#1
new(1,2,Board#1).adjacent(Cell#1)=false
```

Listing 2: Oracle for jpacman.model.Cell generated by JWalk, using the algebraic setting and maximum test depth 2.

An downside to this tool is that the automatically generated values for parameters are practically useless for exhaustive testing. For instance, the tool will always generate a game board of size 1 by 2, then attempt to retrieve the cell located at position (3,4). This will always yield the same exception again and again; we are obviously also interested in the correct retrieval of cells that effectively are on the board, making category partition testing virtually impossible with JWalk (support for category partition is however currently under development [4]). This problem can be solved by implementing a custom generator (an example is shown in Listing 3).

```
package generators;
import org.jwalk.GeneratorException;
import org.jwalk.gen.CustomGenerator;
import org.jwalk.core.MasterGenerator;

public class BoardGenerator implements CustomGenerator {
    private int seed = -1;
    private int ctr = 0;
```

```
int[] num = {5, 10, 3, 5, 0, 0, 4, 9, -1, -1, 5, 10};

public boolean canCreate(Class<?> type) {
        return type = int.class;
}

public Object nextValue(Class<?> type) throws GeneratorException {
        if (type != int.class) {
            throw new GeneratorException(type);
        }

        Integer res = new Integer(num[ctr]);
        if (ctr == num.length -1) {
            ctr = 0;
        } else {
            ctr++;
        }
        return res;
}

public void setOwner(MasterGenerator generator) {
        // Null op
}
```

Listing 3: Custom generator for jpacman.model.Board

Again, these were easily written using the available instructions, but yet, they provide no elegant solution – a depth far greater than probably is required to achieve the same result in jUnit tests is needed here. Therefore, this testing technique and tool is not recommended for classes that rely heavily on category partition testing. It is, however, very useful for the exhaustive testing of all possible interleavings of methods, in just a fraction of the time it takes to do this manually. This makes effort for using this tool very low.

Also note that JWalk does not allow the option of enabling assertions.

## 4.3 $\mu Java$

The latest version of the tool (v.3) was downloaded from the  $\mu Java$  website<sup>7</sup>. Installation instructions can be retrieved when scrolling down the page, but these were difficult to follow at times and occasionally omitted details. A somewhat clearer guide is found in the appendices of [5].

Notwithstanding the fact that the  $\mu Java$  homepage claims there is support

<sup>&</sup>lt;sup>7</sup>http://cs.gmu.edu/~offutt/mujava/#Links

for Java~1.5 and 1.6, this proved to be untrue when applying the tool to the JPacman code. Very little, if any, debug output is provided to the tester (generally a mere "cannot parse" is output when encountering an error), making any attempts at locating the cause of the problems far more trying. After a good deal of study, it became apparent that there is no support for Java~1.5 or 1.6 language constructs such as annotations (for instance  ${\tt QOverride}$ ,  ${\tt QTest}$ , etc). It also does not allow generics (it should be noted that this issue is indicated on the website though), and worse,  ${\tt assert}$ -statements, which are plentiful in JPacman. Therefore, all files had to be rewritten and compiled to Java version 1.4. Fortunately, once this problem was fixed,  $\mu Java$  was fairly easy to work with.

First, mutants were generated. All available mutants were selected for all classes in the jpacman.model package. These mutators are listed in Tables 3 and 4 and the Java equivalents for the latter are found in Table 5 (compiled from the information found in [5] [6] [7]). The mujava.gui.GenMutantsMain GUI allows for easy comparison between the original and mutated class, as could be previously seen in Figure 2.

Class	Method-level	Class-level
Board	167	9
Cell	154	15
Engine	81	5
Food	28	1
Game	147	9
Guest	(abstract)	(abstract)
Monster	0	0
Move	(abstract)	(abstract)
MovingGuest	(abstract)	(abstract)
Player	49	5
PlayerMove	36	9
Wall	0	0

Table 1: Classes in the jpacman.model package and their associated number of mutants created by  $\mu Java$ .

Since  $\mu Java$  predates jUnit, the tests originally supplied with JPacman need to be rewritten as well. No modifications were made other than revising each test to now output a string, which will be used for comparison when running tests on both the original and the mutated code. This process is called  $killing\ mutants$  and is conducted using the mujava.gui.RunTestMain GUI by selecting the CUT and its associated unit test. However, JPacman does not supply a separate unit test for all of the classes, resulting in a very low number of mutants killed for some.

	Method-Level		Class-Level			Coverage	
Class	Killed	Total	Score	Killed	Total	Score	Instr.
Board	108	167	64%	5	9	55%	39%
Cell	109	154	70%	7	15	46%	46%
Engine	1	81	1%	0	5	0%	47%
Game	31	147	21%	0	9	0%	49%
PlayerMove	0	36	0%	0	9	0%	50%

Table 2: Percentages of mutants killed for classes that have an associated unit test in JPacman, contrasted with block coverage scores.

The higher the score, the better – it signifies that more mutations, or possible subtle mistakes by the programmer, can be detected by the test suite. It is fairly high for the Board and Cell classes (especially for mutations on the method level). This is not surprising, since these are conceived as actual unit tests. However, the Game and Engine tests do no provide an exhaustive testing of all possible paths or transitions between states, so it comes as no surprise their mutation scores are very low. Also, when comparing to traditional coverage results, one can deduce the individual tests are far from complete as well.

Mutation testing with  $\mu Java$  has clearly highlighted where there is room for improvement in the tests for the JPacman system. However, the time and effort invested by the tester is exceedingly high, since all code and tests have to be converted to Java 1.4. This operation may not even prove possible or feasible for other systems. The assert-statements provided with the system were lost as well. All this, and the fact that maintaining the same code base in multiple Java versions for the sole sake of using  $\mu Java$  borders on insanity, makes the tool practically unusable for realistic systems unless they are written in compliance with version 1.4 to start with.

# 5 Conclusions

In conclusion, one can state that the *Randoop* tool provides an easy regression test suite for testers that lack time to write one themselves, but it does come with the high cost of unmaintainable tests.

JWalk provides exhaustive testing by executing all methods, but the lack of support for sound category partition testing limits its practical uses. A solution is however in the make. Furthermore, only CUTs with 5 to 15 public operations, executed till a depth of 3 to 5 methods[4], are realistically

Category	Operator	Description	
Encapsulation	AMC	Access Modifier Change	
	IHD	Hiding Variable Deletion	
	IHI	Hiding Variable Insertion	
	IOD	Overriding Method Deletion	
T 1 1	IOP	Overriding Method Calling Position	
Inheritance		Change	
	IOR	Overriding Method Rename	
	ISI	super Insertion	
	ISD	super Deletion	
	IPC	Explicit Call To Parent's Construc-	
		tor Deletion	
	PNC	New Method Call With Child Class	
		Type	
	PMD	Member Variable Declaration With	
		Parent Class Type	
D-1	PPD	Member Variable Declaration With	
Polymorphism		Child Class Type	
	PCI	Type Cast Operator Insertion	
	PCC	Cast Type Change	
	PCD	Type Cast Operator Deletion	
	PRV	Reference Assignment With Other	
		Compatible Variable	
	OMR	Overloading Method Contents Re-	
		placement	
	OMD	Overloading Method Deletion	
	OAN	Arguments Of Overloading Method	
		Call Change	
	JTI	this Keyword Insertion	
	JTD	this Keyword Deletion	
	JSI	static Keyword Insertion	
	JSD	static Keyword Deletion	
Java-Specific Features	JID	Member Variable Initialization	
Java-specific reatures		Deletion	
	JDC	Java-Supported Default Construc-	
		tor Creation	
	EOA	Reference Assignment and Content	
		Assignment Replacement	
	EOC	Reference Comparison and Content	
		Assignment Replacement	
	EAM	Accessor Method Change	
	EMM	Modifier Method Change	

 ${\it Table 3: The 29 \ class-level \ mutation \ operators \ available \ in \ MuJava.}$ 

Category	Operator	Description	
$AOR_B$		Arithmetic Operator Replacement (binary)	
	$AOR_U$	Arithmetic Operator Replacement (unary)	
	$AOR_S$	Arithmetic Operator Replacement (shortcut)	
Arithmetic	$AOI_U$	Arithmetic Operator Insertion (unary)	
	$AOI_S$	Arithmetic Operator Insertion (shortcut)	
$AOD_U \ AOD_S$		Arithmetic Operator Deletion (unary)	
		Arithmetic Operator Deletion (shortcut)	
Relational	ROR	Relational Operator Replacement	
	COR	Conditional Operator Replacement	
Conditional   COI		Conditional Operator Insertion	
	COD	Conditional Operator Deletion	
Shift	SOR	Shift Operator Replacement	
	LOR	Logical Operator Replacement	
Logical	LOI	Logical Operator Insertion	
LOD		Logical Operator Deletion	
Assignment	$ASR_S$	Assignment Operator Replacement (shortcut)	

 $\label{thm:conditional} \begin{tabular}{ll} Table 4: The twelve traditional, or method-level, mutation operators available in $MuJava$. \end{tabular}$ 

Category	Operator	Description
	Binary: +, -,	addition, subtraction, multiplication,
Arithmetic	*, /, %	division and modulus of two numeri-
		cal variables
	Unary: +, -	indicating a positive or negative value
		for a number
	Shortcut: ++,	both pre- and post-increments
Relational	<, <=, >,	
	>=, ==, !=	
Conditional	Binary: &&,   ,	conditional and bitwise AND/ORs,
Conditional	&,  , ^	bitwise XOR
	Unary: !	negation
Shift	>>, <<, >>>	signed shift right/left, unsigned shift
		right
Logical	Binary: &,  , ^	AND, OR, XOR
Logical	Unary: ~	complement
Assignment	+=, -=, *=,	assignment shortcuts
	/=, %=,	

 ${\bf Table\ 5:}\ {\it MuJava's\ method-level\ mutation\ operators\ in\ Java.}$ 

feasible when deploying this tool.

Mutation testing provides a fresh view on assuring test quality and a helpful addition to coverage measurements, but the  $\mu Java$  tool is unfortunately in desperate need of an update to be of any realistic use.

Even though the tools outlined in this paper could be of use to an experienced manual tester, it is clear that there is still a lot of work that needs to be done before (if ever!) the latter can be fully replaced by them. The tools used only work on unit test level; there still will be a need for integration and system tests conducted by experts. Supporting fully automatic category partition testing will prove to be a complex challenge, especially when basic support for this does not yet exists. Also, these three techniques are but a small sample of the various approaches that exist to automated test generation, such as evolutionary testing, model-based testing, robustness testing, classification tree testing, assertion-based testing and many more.

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