

New Strategies for Automated Random Testing

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A thesis submitted for the degree of

Doctor of Philosophy

April 24, 2013

Abstract

This is where you write your abstract ...

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Acknowledgements

Several people have contributed to the completion of my PhD dissertation. However, the most prominent personality deserving due recognition is my worthy supervisor, Dr. Manuel Oriol. Thank you Manuel for your endless help, valuable guidance, constant encouragement, precious advice, sincere and affectionate attitude.

I thank my assessor Prof. John Clark for his constructive feedback on my various reports and presentations. I am also thankful and highly indebted to Prof. Richard Paige for his generous help, cooperation and guidance during my research at the University of York.

Special thanks to my father Prof. Mushtaq A. Mian who provided a conducive environment, valuable guidance and crucial support at all levels of my educational career and my very beloved mother whose love, affection and prayers have been my most precious assets. Also I am thankful to my elder brothers Dr. Ashfaq, Dr. Aftab, Dr. Ishaq, Dr. Afaq and my sister Dr. Haleema who have been the source of inspiration for me to pursue higher studies. My immediate younger brother Dr. Ilyas and my younger sister Ayesha studying in the UK, deserve recognition for their help, well wishes and moral support. Last but not the least I am very thankful to my dear wife Dr. Munazza for her company, help and cooperation throughout my stay at York.

I was funded by Departmental Overseas Research Scholarship (DORS), a financial support awarded to overseas students on the basis of outstanding academic ability and research potential. I am truly grateful to the Department of Computer Science for financial support that allowed me to concentrate on my research.

I feel it a great honour to dedicate my PhD thesis to my beloved parents
for their significant contribution in achieving the goal of academic
excellence.

Chapter 1

Introduction

In this chapter we give a brief introduction and motivation for the research work presented in this thesis. We commence by introducing the problems in random testing. We then describe the alternative approaches to overcome these problems, followed by our research goals and contributions. At the end of the chapter, we give an outline of the thesis.

1.1 The Problems

In software testing, one is often confronted with the problem of selecting a test data set, from a large or often infinite domain, as exhaustive testing is not always applicable. Test data set is a subset of domain carefully selected to test the given software. Finding an adequate test data set is a crucial process in any testing technique as it aims to represent the whole domain and evaluate the given system under test (SUT) for structural or functional properties [?], [?]. Manual test data set generation is a time-consuming and laborious exercise [?], therefore, automated test data set generation is always preferred. Test data generators are classified in to Pathwise, Goal-Oriented, Intelligent and Random [?]. Random test data generation generate test data set randomly from the whole domain. Unlike other approaches Random approach is simple, widely applicable, easy to implement in an automatic testing tool, fastest in computation, no overhead in choosing inputs and free from bias [9].

Despite the benefits random testing offers, its simplistic and non-systematic nature

expose it to high criticism [?]. Myers & Sandler [?] mentioned it as “Probably the poorest methodology of all is random-input testing...”. Where this statement is based on intuition and lacks any experimental evidence, it motivated the interest of research community to evaluate and improve random testing. Adaptive random testing [?], Restricted Random Testing [4], Feedback directed Random Test Generation [?], Mirror Adaptive Random Testing [6] and Quasi Random Testing [7] are few of the enhanced random testing techniques aiming to increase its fault finding ability.

Random testing is also considered weak in providing high code coverage [?], [?]. For example, in random testing when the conditional statement “if (x == 25) then ... ” is exposed to execution then there is only one chance, of the “then...” part of the statement, to be executed out of 2^{32} . If x is an integer variable of 32 bit value [?].

Random testing is no exception when it comes to the complexity of understanding and evaluating test results. Modern testing techniques simplifies results by truncating the lengthy log files and display only the fault revealing test cases in the form of unit tests. However efforts are required to show the test results in more compact and user friendly way.

1.2 Our Goals

The overall goal of this thesis is to develop new techniques for automated testing based on random strategy that addresses the above mentioned problems. Particularly,

1. We aim to develop an automated random testing technique which is able to generate more fault-revealing test data. To achieve this we exploit the presence of fault clusters found in the form of block and strip fault domains inside the input domain of a given SUT. Thus we are able to find equal number of faults in fewer number of test cases than other random strategies.
2. We aim to develop a novel framework for finding the faults, their domains and the presentation of obtained results on a graphical chart inside the specified lower and upper bound. It considers the correlations of the fault and fault domain. It also gives a simplified and user friendly report to easily identify the faulty regions across the whole domain.

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3. We aim to develop another automated testing technique which aims to increase code coverage and generation of more fault-revealing data. To achieve this we utilises Daikon— an automated invariant detector that reports likely program invariant. An invariant is a property that holds at certain point or points in a program. With these invariants in hand we can restrict the random strategy to generate values around these critical points. Thus we are able to increase the code coverage and quick identification of faults.

1.3 Contributions

To achieve the research goals described in Section xx, we make the following specific contributions:

1.3.1 Dirt Spot Sweeping Random Strategy

Development of a new, enhanced and improved form of automated random testing: the Dirt Spot Sweeping Random (DSSR) strategy. This strategy is based on the assumption that faults and unique failures reside in contiguous blocks and stripes. The DSSR strategy starts as a regular random+ testing strategy a random testing technique with preference for boundary values. When a failure is found, it increases the chances of using neighbouring values of the failure in subsequent tests, thus slowly sweeping values around the failures found in hope of finding failures of different kind in its vicinity. The DSSR strategy is implemented in the YETI random testing tool. It is evaluated against random (R) and random+ (R+) strategies by testing 60 classes (35,785 line of code) with one million (105) calls for each session, 30 times for each strategy. The results indicate that for 31 classes, all three strategies find the same unique failures. We analysed the 29 remaining classes using t-tests and found that for 7 classes DSSR is significantly better than both R+ and R, for 8 classes it performs similarly to R+ and is significantly better than R, and for 2 classes it performs similarly to random and is better than R+. In all other cases, DSSR, R+ and R do not perform significantly differently. Numerically, the DSSR strategy finds 43 more unique failures than R and 12 more unique failures than R+.

1.3.2 Automated Discovery of Failure Domain

There are several automated random strategies of software testing based on the presence of point, block and strip fault domains inside the whole input domain. As yet no particular, fully automated test strategy has been developed for the discovery and evaluation of the fault domains. We therefore have developed Automated Discovery of Failure Domain, a new random test strategy that finds the faults and the fault domains in a given system under test. It further provides visualisation of the identified pass and fail domain. In this paper we describe ADFD strategy, its implementation in YETI and illustrate its working with the help of an example. We report on experiments in which we tested error seeded one and two-dimensional numerical programs. Our experimental results show that for each SUT, ADFD strategy successfully performs identification of faults, fault domains and their representation on graphical chart.

1.3.3 Invariant Guided Random+ Strategy

Acknowledgement of random testing being simple in implementation, quick in test case generation and free from any bias, motivated research community to do more for increase in performance, particularly, in code coverage and fault-finding ability. One such effort is Random+ — Ordinary random testing technique with addition of interesting values (border values) of high preference. We took a step further and developed Invariant Guided Random+ Strategy (IGRS). IGRS is an extended form of Random+ strategy guided by software invariants. Invariants from the given software under test are collected by Daikon— an automated invariant detector that reports likely invariant, prior to testing and added to the list of interesting values with high preference. The strategy generate more values around these critical program values. Experimental result shows that IGRS not only increase the code coverage but also find some subtle errors that pure Random and Random+ were either unable or may take a long time to find.

1.4 Thesis Outline

The rest of the thesis is organised as follows: In Chapter 2, we give a thorough review of the relevant literature. We commence by discussing a brief introduction of software

testing and shed light on various techniques and types of software testing. Then, we extend our attention to automated random testing and the testing tools using random technique to test softwares. In Chapter 3, we present our first automated random strategy Dirt Spot Sweeping Random (DSSR) strategy based on sweeping faults from the clusters in the input domain. Chapter 4 describes our second automated random strategy which focus on dynamically finding the fault with their domains and its graphical representation. Chapter 5 presents the third strategy that focus on quick identification of faults and increase in coverage with the help of literals; Finally, in Chapter 7, we summarise the contributions of this thesis, discuss the weaknesses in the work, and suggest avenues for future work.

Chapter 2

Literature Review

Paul Ehrlich famous quote is, To err is human, but to really foul things up you need a computer. Since the programmers are ordinary human beings, it is most obvious that some errors remain in the software after its completion. Errors are not tolerated as they can cause great loss. According to the National Institute of Standard and Technology 2002, 10 report, software errors cost an estimated \$59.5 billion loss to US economy annually. The destruction of the Mariner 1 rocket (1962) that cost \$18.5 million was due to a simple formula coded incorrectly by a programmer. The Hartford Coliseum Collapse (1978) costing \$70 million, Wall Street crash (1987) costing \$500 billion, Failing of long division by Pentium (1993) costing \$475 million, Ariane 5 Rocket disaster costing \$500 million and many others are caused by minor errors in the software. To achieve high quality, the software has to satisfy rigorous stages of testing. The more complex and critical the software, the higher the requirements for software testing and the larger the damage caused if the bug remains in the software.

2.1 Software Testing

Software testing is the process of executing the software or part of the software with test data and then checking the test results to see whether the software or that particular part of software is performing according to its specification [24]. Tests pass if the output obtained from the Software Under Test (SUT) matches the proposed output for the test data. Test is more successful if it finds more errors in the software. Once

errors are found in the SUT, the software is given back to the developers for removing the found errors and after rectifying the mentioned errors the software is again handed over to the testers for retesting. The testing process starts from the very beginning of software development and remains continuous throughout the life of the software.

One thing to keep in mind is that program testing can be used to show the presence of bugs, but never to show the absence of bugs [11]. Which means SUT that passes all the tests without giving a single error is not guaranteed to contain no error. The testing process increases however the reliability and confidence of the users in the tested product.

2.1.1 Software Test Plan

As proper planning is the key to success for many projects this is often also true with software testing. A software test plan is a well defined document that defines the goal, scope, method, resources and time schedule of the testing.

2.1.2 Software Testing Approaches

The testing process starts from the very beginning of the System Development Life Cycle (SDLC) and is carried out in the following two ways.

1. Static Testing
2. Dynamic Testing

2.1.3 Static Testing

The term static means still or non executable. Static Testing is the process in which software documentation/source code is checked for errors without any execution. All high quality softwares will always be accompanied by documentation in addition to software code. These include requirements, design, technical, end user and marketing documentation. Reviews, walkthroughs or inspections are most commonly used for static tests. For instance it is necessary to do the static testing of user documentation for errors because software developed at the cost of millions of dollars have been neglected and abandoned for the only reason that end users were not able to find out the proper

way to operate it to do their routine business. Users tend not to think I will figure out how to operate this software; rather, they say, this software doesn't do what I need it to do, so I want the old system back [12].

2.1.4 Dynamic Testing

Dynamic means variable or changeable so Dynamic Testing is the process in which software code is executed and input is converted into output through processing. Results are analysed to find any error in the software. It is not necessary that dynamic testing start once the software is fully complete. Instead it can start from a single method/unit. Unit testing, integration testing, system testing, and acceptance testing are most commonly used as dynamic testing methods. Dynamic testing can be manual or automated. In manual testing the programmer develops the test cases which are executed by the developed software to find any error in processing or output. Similarly in automated testing the software or components of the software is given as input to testing software that automatically generates test cases and executes the SUT against them to find any errors. Manual testing typically consumes more time and resources than automated testing.

2.1.5 Methods of Dynamic Software Testing

There are three methods for testing the software dynamically.

1. Black-Box Testing
2. White-Box Testing
3. Grey-Box Testing

2.1.5.1 Black-Box Testing

Black-Box or Functional testing is the method in which the testers don't know about the structure of the software. Test cases are derived from the specifications of the SUT and tests pass only if the output checks, according to the specification no matter how it is internally processed by the software. The main emphasis of black-box testing is to check the functionality of the product.

2.1.5.2 White-Box Testing

White-Box or Structural testing is a method in which the testers must know about the complete structure of the software. Test cases are derived from the code structure of the SUT and the test is pass only if the results are correct according to the specification as well as the execution is carried out according to oracle. The main emphasis of the white-box testing is not only functionality but also code coverage [1].

2.1.5.3 Grey-Box Testing

Grey-Box testing is the combination of both black-box/functionality and white-box/structural testing. The tester knows about both the functionality and the internal structure of the SUT. Some of the test cases are based on the functionality and some of the test cases are based on the structure. Emphasis of grey-box testing is both on code coverage as well as functionality [23].

2.1.6 Software Testing Workflow

There are many software techniques like unit testing, integration testing, random testing, regression testing, system testing, acceptance testing, performance testing, load testing, stress testing, alpha testing, beta test etc. All testing techniques belong to black-box, white-box or grey-box approach. Each testing technique has its own strength and weaknesses but the technique in focus here is Random Testing.

We have explained software testing graphically with the help of plotting venn diagram on two dimensional axis. The positive x axis represent black-box while negative x axis represent white-box testing. Grey-box testing in the middle is represented by the overlapping of black-box and white-box testing. Similarly on positive y axis we have dynamic testing and on negative y axis we have static testing. Now if a test is black box and dynamic then the test will fall in 0 to 90 degree on the diagram and if the test is black-box and static then it will fall in 270 to 360 degree. On the other hand if the test is white-box and dynamic then it will fall in 90 to 180 degree and if the test is white-box and static then it will fall in 180 to 270 degrees.

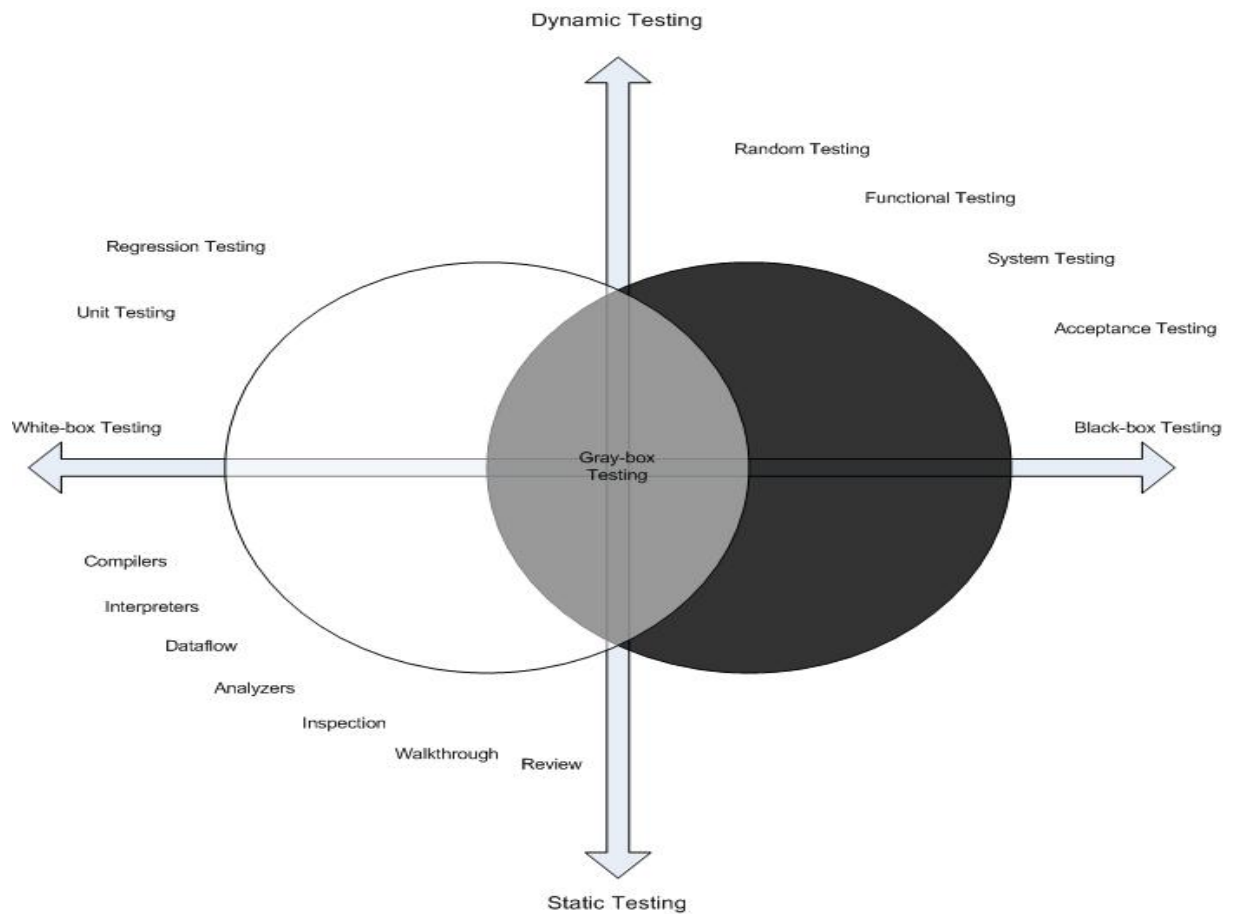


Figure 2.1: Software Testing Workflow

2.2 Software Testing

2.2.1 Manual Testing

2.2.2 Automated Testing

2.2.2.1 Random Testing

Random testing is a dynamic and black-box testing approach in which the software under test is tested by selecting random test cases from already specified input domain [4]. The input domain is the set of all possible inputs to the software and the num-

ber of test cases selected are dependent on the basis of desired reliability or available resources.



Figure 2.2: Random Testing

According to Richard Hamlet [13], to conduct random testing, we need to define an input domain, then we take test points randomly from the whole input domain through a random number/test case generator independently without any restrictions. The program under test is executed on these points and the results obtained are compared to the program specifications. The test fails if any input leads to incorrect results or otherwise it is successful. The internal working of random testing is described in figure 1.2.

2.2.2.2 Exhaustive Testing

2.3 Automated Random Testing

2.3.1 Test Data Generation

2.3.2 Test Execution

2.3.3 Test Oracle

2.3.4 Test Report

2.4 Variations in Random Testing

Different researchers tried various strategies to improve the performance of random testing. In order to better understand the topic we have studied each strategy in detail.

2.4.1 Adaptive Random Testing

Chen et al., [3] noticed that there are certain patterns across the input domain that have more chances of finding bugs in the system. They called it failurecausing inputs and classified them in three categories. The point, block and strip patterns as shown in the following figure.

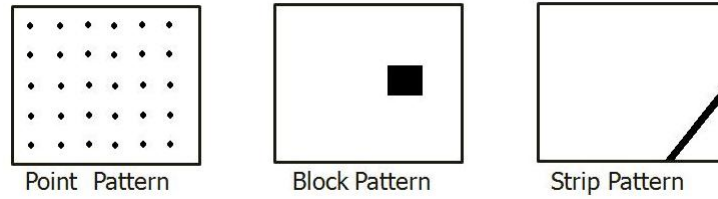


Figure 2.3: Patterns of failure causing inputs

In figure 2.3 the square indicates the whole input domain, in point pattern the failure causing inputs lies scattered in the whole input domain. In block pattern the fault revealing test cases lies in the form of block across the input domain and in the strip pattern the failure causing inputs form a strip across the input domain. Chen et al., [6] argued that we get more chances of hitting these failure patterns if test cases are selected at random but evenly across the input domain. To achieve an even distribution of test cases across the input domain they used two sets. The executed set having the test cases that have been executed by the system and the candidate set that contain the random selected test cases as candidates for execution. The first test case is selected at random from the candidate set and executed, the second test case is then selected from the candidate set based on the criteria that it is far away from the last executed test case. Thus the whole input domain can be tested and there are more chances of hitting the block and strip patterns. Results of the experiments performed on published program using ART showed up to 50% increase in the performance of than ordinary random testing.

2.4.2 Mirror Adaptive Random Testing

Mirror Adaptive Random Testing (MART) [6] was introduced to distribute even test cases across the input domain without complex computation. This decrease in compu-

tation is obtained by merging the technique of mirroring and ART.

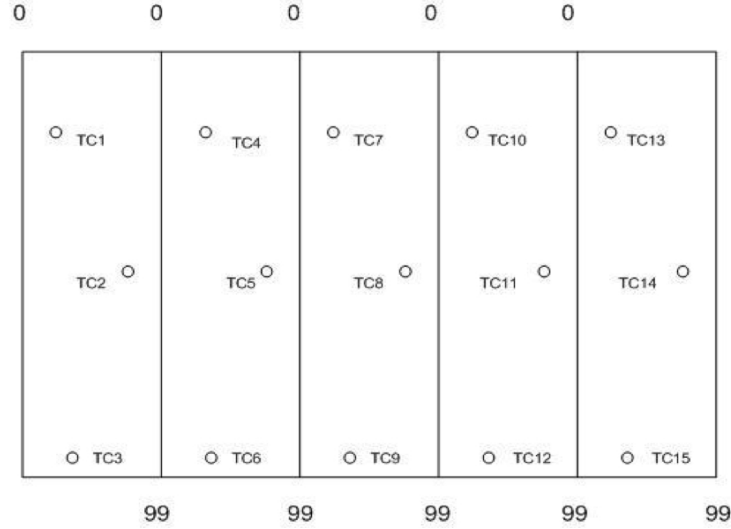


Figure 2.4: Mirror Adaptive Random Testing (0-500)

In this technique, the input domain is divided into n disjoint subdomains. One of the subdomain is called source subdomain while all the others are termed as mirror subdomains. Distance-based Adaptive Random Testing (D-ART) is then applied on source subdomain to select the test case which is executed by the system and the execution is stopped if a failure is detected. Since all other subdomains are the mirrors and similar to source subdomains therefore the same method is applied without too much computational header. This process is continuous sequentially until the first failure is obtained or the system reaches to its stopping point.

2.4.3 Directed Automated Random Testing

2.4.4 Quasi Random Testing

Quasi-random testing [7] is a technique developed to obtain an even better distribution of test cases across the input domain in less computation time with respect to ART. From various experiments Chen et al found out that the failure causing inputs not only form specific pattern but these patterns are continuous as well. Quasi-random testing don't restrict random selection of test cases like ART or RRT rather it uses

a class with a formula. This formula forms an s-dimensional cube in s-dimensional input domain and produces number with Quasi sequence (a sequence of numbers that have small discrepancy and low dispersion) for an s-dimensional input domain. These sequence of numbers are then used by the Quasi approach to select the test cases from s-dimensional input domain. For performance analysis the author compared Quasi approach with ART and random testing. Results showed that the approach is better than random testing but not than ART.

2.4.5 Monti Carlo Random Testing

2.4.6 Good Random Testing

2.4.7 Feedback-directed Random Test Generation

In a bid to improve random testing Pacheco et al., [22] developed a technique which produces unit tests randomly for object oriented programs which are later used for testing the units of the SUT. It is an incremental approach in which unit tests are created and executed against a set of contracts and filters. The feedback obtained from this execution serve as a basis for a sequence of new unit tests. The feedback of the unit test indicate that it is useful to create new input but if it is redundant or illegal like it throws `IllegalArgumentException` error then they are discarded and no unit test of similar nature is created based on its feedback. Thus it only selects unit tests which can be effective in finding bugs or can be used for regression testing. Results of the experiments adopting the technique of Feedback-directed random test generation shows that it can be more productive in code coverage and error detection than systematic and undirected random test generation.

2.4.8 Randoop: Feedback-directed Random Testing

Randoop stands for RANdom tester for Object Oriented Programs [21]. It tests software by using the principle of feedback-directed random test generation to produce unit tests for java and .NET. Randoops input is a set of classes that is to be tested within a certain time and optionally a set of contracts that extend the existing default contracts. After processing the input according to the method of feedback-directed

random testing it give two test suites as output. One is contract voilating tests and the other is regression tests.

2.4.9 Adaptive Random Testing for Object-Oriented

2.4.9.1 Object Distance and its application

To improve the performance of random testing the emphasis of ART was on the distance between the test cases. But this distance was defined only for primitive data types like integers and other elementary input. Ciupa et al defined the parameters that can be used to calculate distance between the composite programmer-defined types so that ART can be applicable to testing of todays object-oriented programs [8]. Two objects have more distance between them if they have more dissimilar properties. The parameters to specify the distance between the objects are dynamic types, values of its primitive and reference fields. Strings are treated as a directly usable values and Levenshtein distance [16] which is also known as edit distance is used as a distance criteria between the two strings. To implement object distance first all the distances of the objects are measured. Then two sets candidate- objects containing the all the objects ready to be run by the system and the used-objects set which is initially empty. First object is selected randomly from the candidate-object set and is moved to used-object set when executed by the system. Now the second object selected from the candidate set for execution is the one with the biggest distance from the last executed object present in the used-object set. This process is continue until the bug is found or the objects in the candidate-object set are finished.

2.4.9.2 ARTOO Tool

After the criteria to calculate the distance between the objects is defined [8], the same team implemented that model and performed several experiments to evaluate the proposed model. Adaptive Random Testing for Object Oriented (ARTOO) is a testing strategy, based on object distance, implemented in AutoTest tool [16]. ARTOO was implemented as a plug-in strategy in AutoTest. It only deals with creating and selecting inputs and all other functionality of the AutoTest was the same. Since ARTOO is based on object distance therefore the method for test input selection is to pick that

object from the candidate set (A pool of objects that is a potential candidate to be executed by the system) which has the highest average distance in comparison to the objects already executed. In the experiments classes from EiffelBase library [17] were used. To evaluate ARTOO the same tests were also applied to directed random strategy (RAND). The outcome of the experiments showed that ARTOO finds the first bug with fewer test cases than RAND. The computation to select test case in ARTOO is more than RAND and therefore ARTOO takes more time to generate a test input. The experiments also found few of the bug found by ARTOO were not pointed out by RAND furthermore ARTOO is less sensitive to the variation of seed value than RAND.

2.4.9.3 Experimental Assessment of Random Testing for Object-Oriented Software

In this research the effect of various parameters involved in random testing and its effect on efficiency is evaluated by performing various experiments on Industrial-grade code base. Large scale clusters of computers were used for 1500 hours of CPU time which resulted in 1875 test sessions for 8 classes under test. [9] The finding of the experiments are 1. Version of random testing algorithm that is efficient for smaller testing timeout is equally efficient for higher testing timeouts. 2. The value of seed for random testing algorithm plays a vital role in finding the number of bugs in specific time. 3. Most of the bugs are found in the first few minutes of the testing sessions.

2.4.10 Restricted Random Testing

Motivated from Adaptive random testing, aim of Restricted Random Testing (RRT) is the same that is selection of test cases from the input domain such that the whole input domain is represented [?]. The plan to achieve an even selection of test cases from the input domain is accomplished by forming an exclusion zone around the first random selected test case.

The next random test case then must be selected outside of this exclusion zone. It makes sure that there is enough distance between the two test cases. The exclusion zone is fixed around each test case and the area of each zone decreases with successive cases. Experimental results of seven error seeded program indicated that RRT is 55% more effective than ordinary/undirected random testing in terms of f-measure (Where

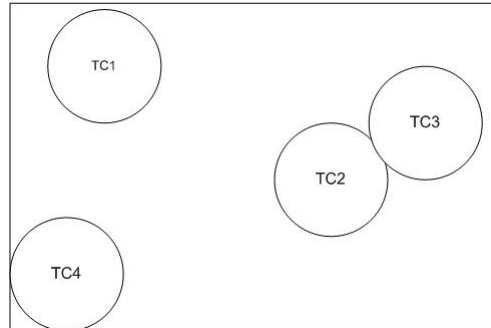


Figure 2.5: Input domain with exclusion zone around the selected test case

f-measure is the total number of test cases required to find the first failure).

2.5 Automated Random Testing Tools

2.5.1 JCrasher

JCrasher is an automatic testing tool that uses a random testing technique to test java classes/programs [21]. The main features of JCrasher are: 1. The randomly created test cases are according to the type and parameters of the methods under test. 2. It uses special heuristics rules, after the execution of the test cases, to see whether the given exceptions are real bugs or the generated input violated the pre-conditions of the program. 3. To clarify the testing from any old tests JCrasher make it sure that every test run on a clean state. 4. JCrasher also produces test cases for JUnit that can be integrated into IDEs like Eclipse. To use JCrasher we have to supply set of Java classes in byte code and testing time. JCrasher analyzes the classes and create test cases randomly with the same type and same parameter list. These test cases are only for public methods of the classes and they check for any system crash. List of exceptions is obtained as a result of execution of test cases which are differentiated as bugs and precondition violations by the input.

2.5.2 JArtage

Jartage (JAwa Random TEst GEnerator) is a tool that randomly generates unit tests for classes specified with JML (Java Modeling Language) [17]. The specification of

Java classes with JML serves two purposes. First, all the test cases generated by Jartege have to verify the conditions defined by JML and thus irrelevant test cases are eliminated. Secondly these JML specifications are also used as oracles. Apart from the JML specification which are made by hand it automates the whole testing process which include test case generation, execution, comparing it against oracle and using the generated test cases for future regression testing.

2.5.3 Eclat

Eclat [20] is a tool that automatically generates unit tests for Java. Eclat can be executed from both command line or from IDE where it can be installed as a plug-in. [28]. Eclat selects a sub-set of test inputs from a large domain, that is likely to reveal fault in the SUT. Eclat takes a correct execution of the SUT and on the basis of it creates an operational model. It then selects only these test inputs from the input domain which fail to comply with the model. A Reducer function removes the redundant test inputs and the remaining test inputs are likely to discover faults in the SUT. Based on the operational model it also produces an automated oracle. Various experiments results shows that Eclats is very effective in finding faults and the ratio of finding faults and test inputs is almost same.

2.5.4 JTest

Parasoft Jtest is a commercial tool that automatically generates and execute unit tests. It can be easily integrated to Java IDEs like Eclipse where it provide two main functionalities, i.e. Static Analysis, Unit testing and code coverage. [25] In static analysis Jtest takes a complete project or set of classes as input and compares it with a list of built-in rules. The statement violating any of these rules is an error. It also suggests probable fixes for the detected fault. For unit testing it takes a class as an input and processes a number of scenarios against it to generate and execute unit tests. Once unit tests are executed they become the part of regression test for future reference. Jtest also shows the code coverage of the program by colour coding the statements that are not executed by the unit tests.

2.5.5 QuickCheck

QuickCheck [10] is a light weight random testing tool that is developed specifically for testing of Haskell programs [14]. Haskell is a functional programming language where programs are evaluated using expressions rather than statements as in case of imperative programming. Therefore in this process the tester defines certain expressions for the functions that must hold for a large number of test cases to be correct. These test cases are generated automatically through generator function which can be set by the tester to generate random test cases or according to specific criteria. After processing all the generated test cases any test case that causes the expression to become false is considered faults.

2.5.6 AgitarOne

AgitarOne is a commercial tool that automatically generates unit tests. It has a Junit Generator engine that can create 25,000 lines or more of Junit per hour [29]. It can be easily integrated into famous IDE like Eclipse. It takes as input, classes under test, time and optionally any knowledge or test cases that has a positive influence on the performance of the testing process. The generated Junit tests can be run from the same IDE and can also be used for later regression testing. The GUI interface is called a dashboard which provides in depth knowledge of the tests conducted, failures detected, alerts and the archives of the tests conducted earlier. It also shows the coverage obtained after executing the Junits against the code under test.

2.5.7 Autotest

Based on Formal Automated testing AutoTest is a tool used for testing of Eiffel programs [9]. The Eiffel language use the concept of contracts (pre-conditions, postconditions and class invariants). Input can be a single class, method or a set of classes which is then processed by AutoTest to generate test cases. It generates both primitive and object type test cases. All the generated test cases are kept in a pool and then randomly a test case is selected from it for execution. A user can set the features of the AutoTest options include: Number of test cases to generate, whether to monitor pre or post condition, order of testing and the initial values of the primitives variables.

2.5.8 TestEra

TestEra [15] is a novel framework for testing Java applications. All the tests are produced and executed in an automated fashion. Tests are conducted on the basis of the method specifications [5]. TestEra takes methods specifications, integer value as a limit to the generated test cases and the method under test. It uses pre-conditions of a method from specifications to automatically generate test cases up to the specified limit. These test cases are then executed on the method and the result is compared against the post-conditions (oracle) of that method. Any test case that fails to satisfy postcondition is considered as a fault. The complete error log is displayed in the Graphical User Interface (GUI).

2.5.9 Korat

Korat [2] is an automated testing tool for Java programs that generates and execute test cases for a method based on its formal specification. To generate test cases for a method Korat makes use of its pre-condition. It then executes the generated test cases against the method specifications. Korat uses JML for specifications. In order to generate test cases for a method Korat constructs new methods that return a Boolean value (Java Predicate) from its pre-conditions. When given these Java predicates Korat generates all non isomorphic input for which the return value of predicate is true. To check correctness of the method, Korat executes the test cases on that method and analyzes the output with the post conditions of the method (oracle). A fault in a method under test throws an exception to indicate the violation of the post-condition.

2.5.10 YETI

The final tool we discuss is YETI (the York Extensible Testing Infrastructure) [33], which is entirely automated and freely available as open-source. It can be used for testing programs written in Java, JML, C, command-line and .Net [18]. It can also be run in cloud for faster execution [19]. It is implemented in Java and uses random technique for testing. It has GUI which makes it easy to diagnose problems at runtime. It is able to call up to one million calls per minute on Java code. YETI oracle is language dependant. If the specifications are available, YETI checks the code against

the specifications for any inconsistency. In case of programs having assertions YETI interprets violations as failures and in case there is no specifications or assertions, YETI performs robustness testing and considers undeclared run- time exceptions as faults. Errors-revealing test cases are reproduced at the end of each testing session. Experiments conducted with YETI showed significant number of bugs in Java.lang class (45 faults) and Itext (120 faults).

2.5.11 Tools for Automated Random Testing

From the literature we can find a number of open source and commercial testing tools that automatically generate unit tests. Each tool utilize different generation technique but the one we are interested in is random technique. We present the most well known tools.

2.6 Conclusion

Tool	Lang	Input	Strategy	Output	Benefit
QuickCheck	H	Specification & Functions	Specification hold to random TC?	Pass/Fail	Easy to Use, Program Doc
Jcrasher	J, JML	Program	Method Type to predict input, Randomly find values of crash	TC	Automated TC, Use of Heuristic Rules
Parasoft Jtest	J	Package	Static Analysis of Code & RT	Exceptions & TC	Eclipse plug-in, GUI & Quick
Jartage	J	Classes	Random strategy with controls like class weight	TC, RT	Quick, Automated
Randoop	J, N	Specification, Code & Time	Generate then Execute Methods & give Feedback for next generation	Faulty TC, RT	Quick, Easy to use
Eclet	J	Classes, Pass TC & candidate inputs	Create model from TC, classify each candidate as Pass/Fail	Faulty TC	Produce output as text, JML
AgitarOne	J	Package, Time & Manual TC	Analyze code with auto and provided data in given time	TC, RT	Eclipse plug-in, GUI & Easy to use
AutoTest	J	Classes, Time & Manuel tests	Heuristic Rules to evaluate Contracts	voilations, RT	GUI in HTML, Easy to use
Korat	J	Specification & Manual TC	Check contracts with specifications	Contracts violations	Give faulty TC
TestEra	J	Specifications, Integer & Manuel TC	Check contracts with specifications	Contracts voilations	GUI, give faulty example
YETI	J, N, JML	Code, Time	Random Plus, Pure Random	Traces of found faults	GUI, give faulty example, quick

Figure 2.6: Summary of automated testing tools

Appdx A

and here I put a bit of postamble ...

Appdx B

and here I put some more postamble ...

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