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A Smart Model for Web Programming Agile Testing

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Received: 14 December 2013 Accepted: 4 January 2014 Published: 15 January 2014

6 Abstract

7 A Model that Minimizing the number of test operations on Web Programming Exhaustive

⁸ Testing is introduced. The proposed model based for the first time to the best of our

⁹ knowledge, the Hebbian trained matrix algorithm. The model problem is formed as Agility

¹⁰ WEB Programming Methodologies optimization problem. Time and web developer efforts will

¹¹ be reduced on white box testing type. A full numerical solved example is introduced. Our

¹² numerical experience shows that our approach is promising especially for Web Programming

13 Exhaustive Testing.

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15 Index terms— exhaustive testing, Hebbian rules, white box, software testing, agile. i.

17 introduction and related works esting in agile software development should provide the information that stakeholders need to make decisions and steer the development into the right direction. We can increase the 18 19 value of testing most by improved intelligence earlier. The challenges of testing in agile development have to be solved! What information is the testing based on? What to be tested and what are the expected results? 20 How to make testing, development and business collaborate? How to involve customer and business people in 21 testing? How to test early so we can achieve the customer requirements in time? Can we use a method to do 22 that smoothly? How to match in between exhaustive testing which take time and to be delivered with customer 23 satisfaction?. The ultimate goal of function testing is to verify that the system performs its functions as specified 24 in the requirements and there are no undiscovered errors left. Since proving the code correctness is not feasible, 25 especially for large software systems, the practical testing is limited to a series of experiments showing the program 26 27 behavior in certain situations. Each choice of input testing data is called a test case. If the structure of the 28 tested program itself is used to build a test case, this is called a white box (or open-box) approach [1]. Several white-box methods for automated generation of test cases are described in literature. For example, the technique 29 of [4] uses mutation analysis to create test cases for unit and module testing. A test set is considered adequate 30 if it causes all mutated (incorrect) versions of the program to fail. The idea of testing programs by injecting 31 simulated faults into the code is further extended in [10]. Another paper [11] presents a family of strategies for 32 automated generation of test cases from Boolean specifications. However, as indicated by [10], modern software 33 systems are too large to be tested by the white-box approach as a single entity. White-box testing techniques can 34 work only at the subsystem level. In function tests that are aimed at checking that a complex software system 35 meets its specification, black-box (or closed box) test cases are much more common. The actual outputs of a 36 black-box test case are compared to expected outputs based on the tester's knowledge and understanding of the 37 38 system requirements. Since the testers have time for only a limited number of test cases, each test case should 39 have a reasonable probability of detecting a fault along with being non-redundant, effective, and of a proper 40 complexity [6]. It should also make program failures obvious to the tester who is supposed to know the expected 41 outputs of the system. Thus, selection of the tests and evaluation of their outputs are crucial for improving the quality of tested software with less cost. If the functional requirements are current, clear, and complete, 42 they can be used as a basis for designing black-box test cases. Assuming that requirements can be re-stated as 43 logical relationships between inputs and outputs, test cases can be generated automatically by such techniques as 44 cause-effect graphs (see [8]) and decision tables [2]. Another method for automatic generation of test vectors from 45 functional relationships is described in [3]. Several ways are proposed to determine, input-output relationships 46

in tested software. Thus, a tester can analyze system specifications, perform structural analysis of the system's 47 source code, and observe the results of system execution. While available system specifications may be incomplete 48 or outdated, especially in the case of a "legacy" application, and the code may be poorly structured, execution 49 data seems to be the most reliable source of information on the actual functionality of an evolving system. In this 50 paper, we extend the idea initially introduced [1] that input-output analysis of execution data can be automated 51 by Info-Fuzzy Network methodology of data mining [7] [9]. In [7] the proposed concept of IFN-based testing has 52 been demonstrated on individual discrete outputs of a small business program. The current study evaluates the 53 effectiveness of the Hebbian rules in Neural networks to let the system more intelligent (expert-system) and can 54 be learned by previous cases tested methodology on a complex application having multiple continuous outputs. 55 This is also deal with the question of determining the minimal number of training cases required to. The rest of 56 the paper is organized as follows. Section 2 provides the methodology on the process testing and derived required 57 paths. Section 3 presents the notation and definition of the proposed model. Section 4 describes a detailed of 58 the proposed methodology. Finally, Section 5 summarizes the paper with initial conclusions and directions for 59 future research and applications. 60 ii. 61

62 1 Methodology

Testing is the process of executing a program with the intent of finding errors." [13]. Assume that there are 520 63 possible different execution flows. If we execute one test per millisecond, it would take 3.170 years to test this 64 program. See figure (1). All tests should be traceable to customer requirements and the uncover errors will be 65 discovered then quickly. The system configuration can be represented mathematically by a graph, with nodes and 66 representing links. Seefigures (1), (2). Our system will minimize the number of 'things in processes, minimize 67 the size of 'things in processes establish a regular cadence deliver business value early, often and consistently, 68 empower the team to create software that meets the customer needs. The ultimate objective of this paper is to 69 give software developers procedures to enhance their ability to find acceptance testing -by customer-for which 70 71 testability is an important consideration. Ideally, one would like to generate an acceptance solution algorithms that take as input the characteristics customers requirements as well as needs criteria, and produce as output 72 an optimal path for solution, this is known as acceptance testing, and it is very difficult to achieve. However, 73 we consider set of paths that will execute all statements and all conditions in a program, at least once that is 74 already designed then we try to derive all the test cases, then the customer will select his steps of tests according 75 to his requirements then a test related path will be chosen and get the test result, if the solution face customer 76 satisfaction the test path will be selected and complete the software development. Customer partner ensure 77 customer gets the value they are paying for build a reference first instead of System requirements. Customer gets 78 what he wants and validates the expected results. Developers know the right answer Use the simplest technology 79 team capability ease of use case of Refactoring Flexibility to Change test automation empower developers to run 80 their own tests run tests regularly [13]. Our paper considers the customer to be member in the testing because 81 the expected results depend on his satisfaction and agreement. Our aim is to minimize the time consumed for 82 test according to our new method procedure. The customer choose path steps toward the objective need to be 83 achieved, then we can detect the suitable solution by multiply the steps with the trained matrix, the solution 84 wanted path will be the result iii. 85

⁸⁶ 2 Notation and Definitiond

Now we will illustrate in this section, all parameters which we used in our new model and we will define every
item: Si: Solution Number

⁸⁹ 3 Mathematical Model

⁹⁰ 4 General formulation of the problem:

91 Analysis function which can be derived for acceptance tests in Extreme Programming Exhaustive testing:?Wi =92 Si*Pi, n Wt =??Wi i=1

93 The accepted test will be:Ta = Wt * Pi

Which guarantee the solution to be assess customer visible functionality. We will present her an important assumptions to declare and describe the formulation of our new model. We derive test cases to exercise these paths. And convert each path to its matrix representation and derive the trained matrix. We have paths as follows:Path 1: [1,1,1,0,0,1,1,1] Path 2: [1,1,1,0,0,0,1,1] Path 3: [1,1,0,1,0,0,1,1] Path 4: [1,1,0,1,0,0,1,0] 1 0 (W1=[1,1,1,0,0,1,1,1]*0= 0 (W2=[1,1,1,0,0,0,1,1]**= (W3=[1,1,0,1,0,0,1,1])*= (W4=[1,1,0,1,0,0,1,0])*= Wt= (W1+(W2+(W3+(W3+(W4=

During the collaboration with customer to output the features and requirements needed a suite solution can be chosen which will achieve the problem and customer requirements. Assume that the customer choose the solution with feature binary: [0, 0,1,0] then to know what is the path to be tested to give solution, it can be derived as follows: Chosen pattern of solution multiply by trained matrix. By dividing the results on the maximum number and the integer values to be taken the new results will be the matrix value [0 0 1 0], this value represent the solution number 3 which will be under test and expected needs to the customer. The result represents the path of solution wanted the next step will be to be tested directly without any loose of time to discover with uncover error with needless paths. In this method we can minimize the Exhaustive testing and minimize the time taken and efforts which yield to produce the project in quick time and help in the highly iteration and incremental analysis.

111 5 Conclusion

112 Exhaustive tests should be planned long before testing begins. The Hebbian trained matrix algorithm applies

to Web programming excusive testing as well all testing should be traceable to customer requirements. Most effective testing should be performed by Hebbian trained matrix, time and web developer efforts will be reduced too.¹



Figure 1: Fig. 1 :

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Figure 2:

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