

# Extended Regression Test Model for Test Suite Reduction

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**Abstract**—Problem of test suite size increases in the process of software maintenance when numbers of revision are required can be solved by reduction algorithms such random and regression test reduction. Nevertheless, the whole performance need to be improved. Therefore, this research paper proposes the new model that is used for improving the regression test reduction, which guarantees the best results compared with traditional methods by adding four algorithms; testing, ordering, deleting test cases and fixing bugs. Accordingly, it gives smaller size and higher percent fixing bugs as about 40% and 200% respectively.

**Keywords**—test case; test suite; reduction; deletion

## I. INTRODUCTION

Test suite reduction techniques are provided for controlling the size of the modified software and to guarantee that after adaptation, it can work properly [1-2]. This means that the modification can be used to avoid problem of programming errors and support users regarding their requirement changes [3]. As we know that one of the main problems of software maintenance is the size of a new test suite that is used to guarantee the ability of using a modified program increases. Accordingly, it drops the performance of execution and processing times including producing a software bug after the adaptations. Therefore, size reduction methods are designed to fix this issue, which includes retest-all, random and regression test techniques [4-6]. Benefit of applying retest-all is given when the size is small regarding test suite reduction [7]. If the size gets larger, it increases the mentioned issues explained above [8-11]. To solve these problems, random technique is proposed but the weakness is that the accuracy of reducing the test suite size may not give an appropriate result, when dealing with numbers of bug [12]. Hence, the regression test model is applied for better performance compared with the traditional methods. This technique provides smaller size with less programming error by preparing iterations of selecting good representative by avoiding those bugs. However, we found that if we can reduce the size of test suite without the effect on the use of software, then it can be new alternative for program developer and tester to prepare effective modified software. However, the traditional methods [13] still need the improvement when the different size of test suites are given. Therefore, the proposed model is offered by improving the performance of regression test reduction to contribute two benefits, which are giving a lower size and higher percent of fixing bugs when compared to the original approach.

## II. TEST SUITE REDUCTION

Test suite reduction becomes the idea of removing irrelevant test cases under the field of software engineering, which the main objective is to increase ability of the modified program, while user requirements get involved. Accordingly, this paper focuses four techniques, which are retest-all, random, and regression test including the extended regression test model.

### A. Retest-all test cases ( $R_1$ )

It is the traditional method that verifies and validates a new version of the software modification by testing all test cases in a specified test suite of each program. Besides,  $R_1$  is a well-known technique that is useful since this study becomes the hot issue of software maintenance because the algorithm is simple to use. But, when records of test suite increase that means the selected test cases also grow as well. Thus, this difficulty cannot be fixed whereas the size is too big. Accordingly, the different selection approaches are proposed for this reason, e.g., random technique, regression test model, minimization and prioritization.  $R_1$  tests and chooses all test cases ( $t_i$ ) in a test suite ( $T'$ ) after modifying program ( $P'$ ) [1]. It works very well when amounts of  $t_i$  or the size of  $T'$  are small, while  $\forall t_i$  are tested and chosen. However, the weakness is that the ability of  $P'$  drops when dealing with a large  $T'$ .

### B. Random Deletion ( $R_2$ )

Using  $R_1$  cannot guarantee the good results when the size of  $T'$  is to big. Therefore, random Deletion ( $R_2$ ) is proposed to fix the problem of  $R_1$  by randomly removing  $\exists t_i$  in  $T'$  [2]. One of the objectives of applying  $R_1$  is reduce only the size of  $T'$  without avoiding bugs ( $b$ ) that could be occurred when  $\exists t_i$  fail after test. Moreover, this technique can help developer to save testing time as well regarding the reason explained at the beginning. However, the performance of  $R_2$  is better than  $R_1$  whereas  $T'$  gets bigger but another problem of using  $R_2$  is that bugs will affect a new program when they are not fixed. This is the reason why regression test selection is proposed.

According to this, numbers of the selected test cases are reduced with less errors.

### C. Regression Test Reduction ( $R_3$ )

It is a better technique used for choosing a set of test cases that avoid programming errors, while the size is control. It can solve the critical issue in software maintenance.  $R_3$  is concerned as a suitable technique that reduces the size of  $T'$  by picking  $\exists t_i$ , which has no  $b$ , while  $T'$  is being controlled. Accordingly,  $R_3$  gives better results when compared the traditional methods ( $R_1$  and  $R_2$ ). There are five steps to run  $R_3$  listed as follows [4-6];

- (i) execute  $P'$  to find  $b$  by using  $R_1$ ,
- (ii) build generate  $\forall t_i$  or  $T'$  for the executed  $P'$  based upon the functional and non-functional requirements,
- (iii) design the coverage matrix to identify relevant and relevant  $t_i$  in  $T'$ ,
- (iv) determine  $b$  in  $T'$  to confirm that the coverage matrix is a well design, and
- (v) remove  $\exists t_i$  that doesn't meet the conditions, however, revising  $\exists t_i$  may be required when  $t_i = \phi$ .

The objective of this paper is to improve ability of  $R_3$  by adding an effective methods at the last step, which will be explained in section III.

### III. THE PROPOSED MODEL

The proposed model ( $R_4$ ) is created to improve ability of the last process of  $R_3$  regarding (i)  $T'$  of  $P'$  are given, (ii)  $\forall t_i$  are tested to avoid  $b$ , (iii)  $\forall t_i$  are ordered, (iv) irrelevant test cases are deleted, and (vi) checking bugs. However, this paper presents three algorithms for extending the regression test, which are described as follows,

#### Algorithm 1: Testing $\forall t_i$

Input:  $T'$

Output:  $t_p, t_f$ , and  $t_v$

- (1) Let  $t_i \in T'$ ,  $t_p \in t$  and  $t_f \in t$ , whereas  $t_i$  comprises passed ( $t_p$ ) and failed ( $t_f$ ) test case respectively.
- (2) If  $t_f \in t$  then  $t_f$  will be removed, or else if  $t_p \in t$  then keep  $t_p$  for the next classification.
- (3) Moreover, if  $t_v \in t$  then we have to rewrite  $t_v$ , whereas  $t_v$  is a revised test case. These results that  $T'$  combines  $t_p, t_f$ , and  $t_v$ .

#### Algorithm 2: Ordering $t_u$ and $t_d$

Input:  $t_p, t_f$ , and  $t_v$

Output:  $t_u$  and  $t_d$

Where,  $\{t_o, t_e, t_c\} \in t_u$  and  $\{t_g, t_n\} \in t_d$

- (4) If  $t_u \in t_p$  then  $t_p$  is ordered to give a set of  $t_u$  and  $t_d$ , which are denoted as unused and used test cases.
- (5) Ordering  $t_u$  presents an obsolete ( $t_o$ ), redundant ( $t_e$ ), and complex ( $t_c$ ) test cases, while  $t_d$  concludes  $t_g$  and  $t_n$ , which are reused and new test cases with faultless, respectively.

#### Algorithm 3: Deleting $t_u$

Input:  $t_u$  and  $t_d$

Output:  $t_g$  and  $t_n$

- (6)  $t_o, t_e$ , and  $t_c$  will be removed from  $T'$
- (7) if " $t_o \in t_u$ " then delete  $t_o$  or
- (8) else if " $t_e \in t_u$ " then delete  $t_e$  or
- (9) else if " $t_c \in t_u$ " then delete  $t_c$ .

#### Algorithm 4: Finding bugs

(10) For  $\exists b \in t_d$  and  $b \neq \phi$ , Delete  $b$

(11) While  $\exists b \notin t_d$  or  $b = \phi$ , Receive  $t_d = t_d - 0$

Evaluation of the performance of test suite reduction can be considered by using percent Size Reduction (%S), Percent bugs (%B), and percent efficiency of removing bugs(%R) shown as follows;

$$\%S = 100\% \times \left( \frac{T' - t_d}{T'} \right) \quad (1)$$

$$\%B = 100\% \times \left( \frac{b}{t_d} \right) \quad (2)$$

$$\%R = 100\% \times \left( \frac{b - b'}{b} \right) \quad (3)$$

Where,  $b$  and  $b'$  is amounts of bugs and fixed bugs in the modified program.

As shown by Fig. 1,  $R_4$  focuses the result of  $R_3$  by adding three algorithms discussed previously to determine new  $T'$  that combines  $t_p, t_f$ , and  $t_v$ . For  $T'$  combines  $t_f$ , and  $t_v$ , they will be removed from the system in order to avoid redesigning test cases, which may take long time. The main objective is to find  $t_p$  that combines  $t_u$  and  $t_d$ . However,  $t_u$  gives  $t_o, t_e$ , and  $t_c$ , which will be removed from new  $T'$

Finally,  $t_d$  will be classified into  $t_g$  and  $t_n$  for to be chosen as a good representative by selection strategy.

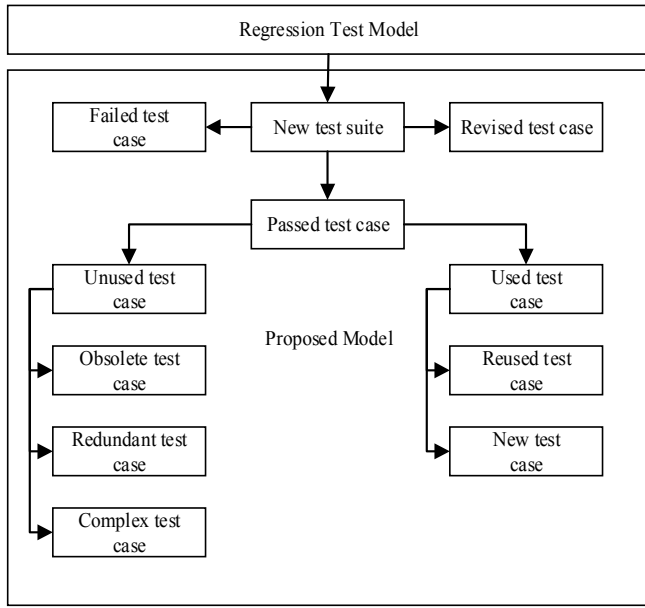


Figure 1: Structure of the proposed model

#### IV. RESULTS AND DISCUSSION

There are five factors discussed in this section, which can show ability of the proposed model to control the size of a new test suite of each selected program (Tcas, Totinfo, Schedule, Schedule2, Print-tokens, Print-tokens2, and Replace). The results of  $R_4$  will be compared with the three traditional algorithms explained as follows;

##### A. Size Reduction

Table I gives the amounts  $t_p$ ,  $t_f$ , and  $t_v$ . On the part of doing the experiments by checking the results of testing  $t$ , which getting “passed” will be typed as  $p$ . Accordingly, this gives the outputs as follows;  $\alpha = 1,267$ ,  $\alpha_2 = 717$ ,  $\beta = 2,308$ ,  $\beta_2 = 2,401$ ,  $\chi = 3,802$ ,  $\chi_2 = 3,722$ , and  $\delta = 5,251$ . For the cases that getting “failed” or  $f$  refer to investigating the programming errors, which cannot be integrated to the modified programs. It shows the value of  $\alpha = 241$ ,  $\alpha_2 = 276$ ,  $\beta = 271$ ,  $\beta_2 = 289$ ,  $\chi = 255$ ,  $\chi_2 = 297$ , and  $\delta = 270$ . Table II reports the results of running  $R_4$ , which are the amount of unused and used test cases. For studying  $t_v$ , it is not detailed in this paper because revising test cases interferes the design of appropriate criteria, e.g., setting scopes and limitations of modifying codes, which takes time. Moreover, data in Table II shows more information for determining  $t_u$  and  $t_d$ , which are member of  $t_p$  as;

$$t_p = t_u + t_d \quad (4)$$

Example of computing  $p$  by using (4) for dataset named  $\alpha$  gives that  $p = 1,138 + 129 = 1,267$ . The same procedure can be

applied for the rest. For the details of finding a set of  $t_u$  and  $t_d$  use (5) and (6) respectively.

$$t_u = t_o + t_e + t_c \quad (5)$$

$$t_d = t_g + t_n \quad (6)$$

Therefore, the outcomes of choosing  $d$  by different four studies offers that  $R_4$  can prepare the smaller size that others as shown in Table IV. Accordingly, the average %S of  $R_2$ ,  $R_3$ , and  $R_4$  are 28.43%, 74.14%, and 93.56% respectively. The lowest performance of decreasing the size of the selected test case refer to using  $R_2$ . This means that  $R_2$  technique does not appropriate for the test suite size, where the number of test cases are in the range 1,000-5,000. Besides, Table V compares performance of reducing the size of test suite of each program between  $R_4$  versus  $R_2$  and  $R_3$  respectively. According to this,  $R_4$  gives better results than others.

##### B. Bugs

For  $d$  or selected test cases from different technique explained in this research have no bugs when they are tested individually. However, after the integration, some bugs are found reported in Table VI. Besides, Table VI and VII presents amount and percentage of bugs by those studies. It shows that the satisfied reports can be done by the proposed model, which better than the comparative studies as 2-25 times. Therefore, to apply  $R_4$  for the test suite size between 1,000-5,000 cases can give the best answer, which the traditional methods cannot offer the small size because if the size becomes smaller, numbers of bugs may increase.

##### C. Efficiency

Fixing bugs as reported in Table VIII will be held to make the program works properly. It reports that some number of bugs can be solved. Unfortunately that this part has been done under the limitation of testers and supports. As shown by Table IX,  $R_4$  can give 100.00% bugs removal, while  $R_2$  and  $R_3$  mostly offer lower percentage. Probably, the problem of fixing bugs after the selection deals the complexity of preparing technique, e.g., code deletion and/or addition. Therefore, designing test cases at the beginning of the process need to be done carefully.

TABLE I. TYPES OF TEST CASE

Name	Program	$T'$	$t_p$	$t_f$	$t_v$
Tcas	$\alpha$	1,608	1,267	241	100
Totinfo	$\alpha_2$	1,052	717	276	59
Schedule	$\beta$	2,650	2,308	271	71
Schedule2	$\beta_2$	2,710	2,401	289	20
Print-tokens	$\chi$	4,130	3,802	255	73
Print-tokens2	$\chi_2$	4,115	3,722	297	96
Replace	$\delta$	5,542	5,251	270	21

TABLE II. UNUSED AND USED TEST CASE

Program	$t_u$	$t_o$	$t_e$	$t_c$	$t_d$	$t_g$	$t_n$
$\alpha$	1,138	310	365	463	129	108	21
$\alpha 2$	643	52	205	386	74	69	5
$\beta$	2,149	423	259	1,467	159	127	32
$\beta 2$	2,265	361	264	1,640	136	95	41
$\chi$	3,595	259	314	3,022	207	181	26
$\chi 2$	3,352	397	202	2,753	370	343	27
$\delta$	4,974	456	317	4,201	277	247	30

TABLE III. SIZE

Program	$R_1$	$R_2$	$R_3$	$R_4$
$\alpha$	1,608	1,126	354	129
$\alpha 2$	1,052	736	305	74
$\beta$	2,650	1,882	636	159
$\beta 2$	2,710	1,897	650	136
$\chi$	4,130	3,015	1,156	207
$\chi 2$	4,115	3,045	1,029	370
$\delta$	5,542	4,046	1,607	277

TABLE IV. PERCENT REDUCING SIZE

Program	$R_2$	$R_3$	$R_4$
$\alpha$	29.98	77.99	91.98
$\alpha 2$	30.04	71.01	92.97
$\beta$	28.98	76.00	94.00
$\beta 2$	30.00	76.01	94.98
$\chi$	27.00	72.01	94.99
$\chi 2$	26.00	74.99	91.01
$\delta$	26.99	71.00	95.00

TABLE V. PERCENT REDUCING SIZE OF  $R_4$  VERSUS  $R_2$  AND  $R_3$ 

Program	$R_2$	$R_3$
$\alpha$	62.00	13.99
$\alpha 2$	62.93	21.96
$\beta$	65.02	18.00
$\beta 2$	64.98	18.97
$\chi$	67.99	22.98
$\chi 2$	65.01	16.02
$\delta$	68.01	24.00

TABLE VI. NUMBERS OF BUG

Program	$R_2$	$R_3$	$R_4$
$\alpha$	60	10	3
$\alpha 2$	66	17	2
$\beta$	67	15	1
$\beta 2$	63	10	2
$\chi$	69	15	2
$\chi 2$	68	16	3
$\delta$	65	20	1

TABLE VII. PERCENT BUGS

Program	$R_2$	$R_3$	$R_4$
$\alpha$	5.33	2.82	2.33
$\alpha 2$	8.97	5.57	2.70
$\beta$	3.56	2.36	0.63
$\beta 2$	3.32	1.54	1.47
$\chi$	2.29	1.30	0.97
$\chi 2$	2.23	1.55	0.81
$\delta$	1.61	1.24	0.36

TABLE VIII. FIXED BUGS

Program	$R_2$	$R_3$	$R_4$
$\alpha$	50	10	3
$\alpha 2$	50	6	2
$\beta$	24	6	1
$\beta 2$	22	9	2
$\chi$	32	9	2
$\chi 2$	21	6	3
$\delta$	34	6	1

TABLE IX. EFFICIENCY OF FIXING BUGS

Program	$R_2$	$R_3$	$R_4$
$\alpha$	83.33	100.00	100.00
$\alpha 2$	75.76	35.29	100.00
$\beta$	35.82	40.00	100.00
$\beta 2$	34.92	90.00	100.00
$\chi$	46.38	60.00	100.00
$\chi 2$	30.88	37.50	100.00
$\delta$	52.31	30.00	100.00

## V. CONCLUSION

Problem of offering test suite reduction claims knowledge of evading bugs and reducing size of modified programs. Random technique cannot solve problem of bugs that occurred in a program and cannot give low percent size reduction, while regression test reduction give the better results. When compared all performances with the proposed model, we found that the traditional methods offers greater size and less percent of fixing bugs. These will affect executing, processing, and testing times throughout the whole processes of software maintenance.

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