

The Design of Neural Network Based Accuracy Prediction for Software Testing Black-Box A Method with the Equivalence Partitions Technique

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Abstract— The absence of a *neural network algorithm* model to predict the level of accuracy in terms of black-box software testing, *equivalence partitions* technique is a problem in this research. In this case, the algorithm used for predicting software errors by researchers is the *neural network algorithm* and testing the software uses the *black-box* method with the *equivalence partitions* technique. The neural network algorithm is an artificial neural system, or neural network are the physical cellular system that can acquire, store and use the knowledge gained from experience for activation using bipolar sigmoid output values which range between -1 to 1. Software testing black-box methods is a testing approach where the data comes from defined functional requirements regardless of the final program structure, and the technique used is equivalence partitions. The design prediction accuracy of this research is by determining the college application to be the software to be tested, then tested using the *black-box* method with the equivalence partitions technique (this test chosen because it will find software errors in several categories, including functions error or missing, interface errors, errors in data structures or external database access, performance errors, initialization errors and terminations), from the *black-box* test the dataset obtained. This dataset measures the accuracy of the *real output* and prediction output. The last step is calculating the *error*, *RSME* from the *real output* and *prediction output*. The results of this research show that the neural network algorithm was being to measure the accuracy level of software testing applied to determine the prediction of the accuracy level of *black-box* software testing with the *equivalence partitions* technique, and the average accuracy results are above 80%.

Keywords—Neural Network, Black-Box, Equivalence Partitions.

I. INTRODUCTION

Software testing is a program execution process that has a research objective is to find errors [1]. Software testing should find unintended errors, and the test is declared successful if it is successful in correcting those errors. *Black-box* software testing defined as a testing process that tries to find errors in several categories, including malfunction, interface error, data structure error, performance error, initialization error, and termination [2]. The absence of a design prediction model of the level of accuracy based on the *neural network* algorithm for testing software *black-box* method *equivalence partitions* technique. From the background above, the researcher will make “the design of the neural network algorithm based accuracy prediction for software testing black-box method with the equivalence partitions technique.”

II. METHOD

The following is a picture of the method in this research, and the higher education of application will be the software to be tested.

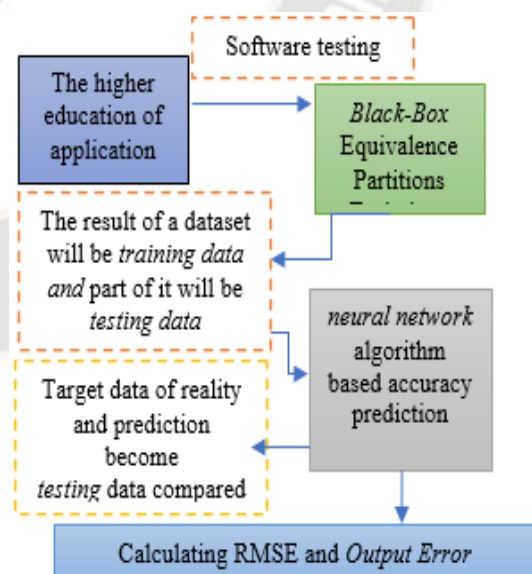


Fig. 1. The method of software testing *black-box* method of *equivalence partitions* technique

Based on the figure above illustrates the method used in this research, with the following stages: the higher education of application is software that was being as a software to be tested, then the software will be tested using the *black-box* method of *equivalence partitions* technique, the results of software testing, afterward this dataset become *training data*, and part of it will be *examining data*, then this dataset will be predicted for accuracy with a *neural network* algorithm by comparing reality target data with predictive data from *testing data*, then finally counting the *output error* value and RSME.

A. The Design of Software Testing Prediction Model

The design of the software testing prediction model can be seen in Figure 2, namely:

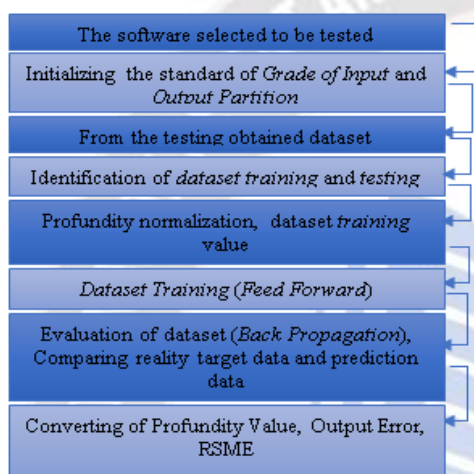


Fig. 2. The design of the software testing prediction model

Based on fig. 2 describes that the design stages of the prediction model for software testing, the first stage begins with software selected to was being testing, namely the higher education of application. The black box method is applied by initializing the standard grade of the input and output partitions. The black box method is applied by initializing the standard grade of the input and output partitions. From the test results obtained a dataset, and this dataset will be identified, and this dataset will become training data and test data. The data set will become a neural network algorithm. To change this, what is done is first to normalize the training data set and depth values, then enter the feed-forward stage to the training data. The dataset was evaluating (*back-propagation*) of this process will produce reality target data and prediction data. For the last step, changes to the depth value, from the depth value conversion, the calculated output error value, and the RSME value.

III. EXPERIMENT RESULT

The steps were taken before formatting your paper are the first to write and save the content as a separate text file. Keep you text and graphic files separate until after the text has been formatting and to. The following stages of applying the *black-box* method of *equivalence partitions* techniques and *neural network* algorithms[12], here are the results of this research, namely:

a. Initialization the Standard Grade Partition of Input and Output

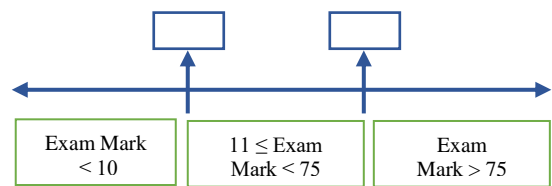


Fig. 3. Equivalence partitions for Exam mark input

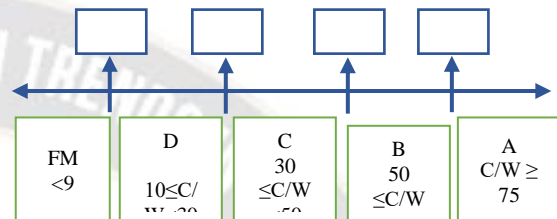


Fig. 4. Equivalence partitions for grade outputs

b. Dataset Testing with the black-box method of equivalence partitions technique

In software testing documentation of the black-box method will be carried out. At this stage. At the grade values that are found to be errors and in each form will be divided into five error models, including errors in function, data structure, interface, initialization, and performance. As for the score grade, the error value found in each form, namely:

Second Form of the Fifth Testing	
Test Case Value Input (Error)	Score Error
Input (Exam Mark Performance)	0
Total Error (as Calculated)	10
Partiton Tested (Of Exam Mark)	$10 \leq C/W < 30$
Expected Output	D

Based on the second form of the fifth testing, the input value calculation = 10, the output value is obtained (Total Error = 0, and the Partition Grade is D).

Second Form	
Test Case Value Input (Error)	Score Error
Function	0
Structure	0
Interface	0
Initialization	0
Performance	10
Total Error (as Calculated)	10
Partiton Tested (Of Exam Mark)	$10 \leq C/W < 30$
Expected Output	D

From the second form, the calculation of the input value (0 + 0 + 0 + 0 + 10), the output value is obtained (Total Error = 10, and the Grade Partition is D).

Table 1: a dataset of software testing results by using *black-box* method *equivalence partitions* technique

a	b	c	d	e	f	g	Testing Status
1	1	1	1	1	10	1	Correct
2	1	2	1	2	20	1	Correct
3	1	3	1	3	30	1	Correct
4	1	4	1	4	40	1	Correct
198	40	3	1	0	0	0	InCorrect
199	40	4	1	0	0	0	InCorrect
198	40	3	1	0	0	0	InCorrect
199	40	4	1	0	0	0	InCorrect

a=no, b=form, c= testing, d= tester, e= total error, f= equivalence partitions, g= defect.

Table 1 shows that the *dataset* produced from software testing with the *black-box* method of *equivalence partitions* technique, for the *correct test status* label it means that the tester test results have the same results as the *defect* test results, while the *incorrect* meaning of the *tester* test results is not the same as the *defect* results.

c. Profundity normalization, Dataset Training Values

In normalization of the profundity values on the *dataset training* is carried out, for the process of converting the testing data into range data to 0.1 and 0.9 because the activation function used is the sigmoid function, where the function value never reaches 0 or 1[11], namely:

$$X' = \frac{0.8 * (Xa)}{b * a} + 0,1$$

Table 2: Dataset of Normalization Results

a	b	c	d	e	f	Testing Status
2	1	0,12	0,1	0,1	0,1	InCorrect
2	2	0,12	0,1	0,1	0,1	InCorrect
2	3	0,12	0,1	0,1	0,1	InCorrect
2	4	0,12	0,1	0,1	0,1	InCorrect
2	5	0,12	0,12	0,3	0,12	Correct
25	1	0,1	0,1	0,1	0,1	Correct
25	2	0,1	0,1	0,1	0,1	Correct
25	3	0,1	0,1	0,1	0,1	Correct
25	4	0,1	0,1	0,1	0,1	Correct
25	5	0,1	0,1	0,1	0,1	Correct

a=no, b=form, c= testing, d= tester, e= total error, f= equivalence partitions, g= defect.

Table 3: Dataset of *neural network algorithm* based accuracy prediction of software testing by using *Black-Box* method and *Equivalence partitions* technique, RMSE value = 0.00142438 in 481 iterations

a	Reality		Prediction		Error	
	b	c	d	e		
2	0	InCorrect	0,006872	InCorrect	- 0,00687	0,993128
2	0	InCorrect	0,014721	InCorrect	- 0,01472	0,985279
2	0	InCorrect	0,013219	InCorrect	- 0,01322	0,986781
2	0	InCorrect	0,012621	InCorrect	- 0,01262	0,987379
2	1	Correct	0,994963	Correct	0,005037	0,994963
2	0	InCorrect	0,006872	InCorrect	- 0,00687	0,993128
25	1	Correct	0,99241	Correct	0,00759	0,99241
25	1	Correct	0,99647	Correct	0,00353	0,99647
25	1	Correct	0,996065	Correct	0,003935	0,996065
25	1	Correct	0,995877	Correct	0,004123	0,995877
25	1	Correct	0,995469	Correct	0,004531	0,99547

a=no, b=defect, c= target, d= defect, e= output

Based on the graph of *Dataset 4 Hidden Layer Prediction Results*, at *Epoch = 900*, *Learning rate = 0.1*, and the number of testing data = 20 data (20% of the 100 total training data) can be seen in the figure below:

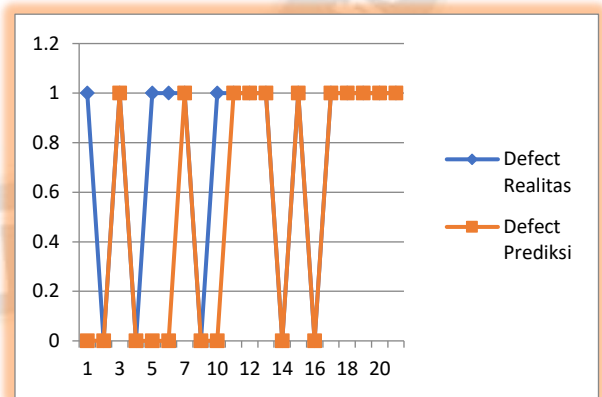


Fig. 5. Graph of *Dataset 4 Hidden Layer Prediction Results*, at *Epoch = 900*, *Learning rate = 0.1*, and the number of testing data = 20 data (20% of the 100 total training data)

d. The result of Accuracy Prediction Level based on the *Neural Network Algorithm* of Software Testing with the *Black-Box Method* and *Equivalence Partitions* Technique

Software testing by using the *black-box* method of the *equivalence partitions* technique based on the *neural network algorithm* has an accuracy of 95% out of 100%, and this is very accurate.

Total Hidden Layer	Epochs	Learning rate	Confusion Matrix (Accuracy)	AUC
4	900	0,1	82.00%	0.840 +/- 0.071
4	900	0,1	95.00%	0.954 +/- 0.071
5	900	0,1	81.00%	0.814 +/- 0.071

IV. CONCLUSION

Based on the results of this research, it will be concluded as follows:

Making a prediction model of the level of accuracy using the neural network algorithm, which is used for software testing using the black-box method is comparable to the partition technique method and can be applied to determine the prediction of the accuracy level of black-box software testing with the equivalence partition technique with very accurate accuracy because the prediction shows the average value. The average above 80%, namely: 82% (4 hidden layers, epoch = 900, learning rate = 0.1), 95% (4 hidden layers, epoch = 1000, learning rate = 0.1), and 81% (5 hidden layer, epoch = 1000, learning rate = 0.1), and the most accurate *neural network training* design model is with 4 *hidden layers*, epoch = 1000, learning rate = 0.1) with an accuracy rate of 95%.

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