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COMPARISON OF TEST CASES USING GENETIC ALGORITHM IN SOFTWARE TESTING

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ABSTRACT

The limit of particularly offering encoded data to different customers through open cloud storage may This paper presents a set of methods that uses a software testing technique in genetic algorithm for a automatic test-data generation. These test case generations are used in evaluate the fitness function in genetic algorithm for selecting the best test case methods. In use a genetic algorithm to generate test cases for evaluating the fitness function in the case generation methods. In use of test cases in genetic algorithm to calculate the fitness function to improve the performance. To aim of the research paper is to implement in genetic algorithm to reduce the cost, time and effort and to minimization of test cases to deliver the good quality software.

INTRODUCTION

Software testing is a term to Quality and Reliability. And the technique used to validating and verifying to developing the software Quality. In software testing technique to remains the integral part of (SDLC) Software Development Life Cycle. Then it has done important phase is to remove the errors and to deliver a good Quality of the software. The first publication on 'search-based software testing' appeared in 1976, and was the work of two American researchers Webb Miller and David spooner[1].

INTRODUCTION TO GENETIC ALGORITHM

Genetic Algorithm is an optimization technique. In mostly used in software testing is to reduce the effort, time and cost and to deliver a good quality of the software. Genetic Algorithm is to find the solution of modified optimization problem. In a genetic algorithm to generate an automatic test cases in the random testing. The genetic algorithm using and finding the evaluating a fitness functions, in the two test cases are generated

INTRODUCTION TO GENETIC ALGORITHM

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Initialization population;
Randomly generated n number of individuals;
Evaluate fitness;
while(fitness value!=termination of criteria)do
{
    Selection;
    Crossover;
    Mutation;
    Calculate fitness function;
}

```

GENETIC ALGORITHMS USE IN THREE OPERATIONS

- SELECTION
- CROSSOVER
- MUTATION

Selection

Selection is the process to analyzing the chromosome or the population and to generate a fitness value. In one of the various selection method in RANDOM SELECTION



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Crossover

Crossover is a technique to used to generate a new chromosome in the existing values and to generate a new population.

Mutation

In mutation process to evaluate a fitness values in one by one and to execute the average fitness value in the existing function

First case:

Population size = 50

Number of generations = 1000

Crossover rate=0.7

Mutation rate = 0.001

Table 1: Average fitness value with mutation rate = 0.001

Generation	Average Fitness	Max Fitness
1	31.88	33
2	32.88	34
3	32.67	34
4	32.75	35
5	32.96	34
10	34.17	38
25	37.29	42
50	41.21	44
100	39.67	47
150	44.33	49
200	46.33	47
250	47.63	49
500	50.23	50
750	51.79	51
1000	52	51
	607.79	638
	Average Fitness Value	95%

Second Case

Population size = 50

Number of generations = 500

Crossover rate=0.7

Mutation rate = 0.01

Table 2: Average fitness value with mutation rate = 0.01

Generation	Average Fitness	Max Fitness
1	32.25	32
2	32.13	32
3	33.92	33
4	32.42	33
5	33.79	33
10	34.71	34
25	38.42	36
50	39.08	37
100	37.42	36
150	35.54	40
200	38.79	40
250	41.83	39
500	42.18	43
	472.48	468
	Average Fitness Value	0.990518117

Population size = 50

Number of generations = 250

Crossover rate=0.7

Mutation rate = 0.001

Table 2: Best fitness value generation

Best Fitness	Average Fitness	Standard Deviation
16.77	11.59	5.18
20.39	15.09	5.3
20.39	15.71	4.68
22.99	15.84	7.15
22.99	16.03	6.96
23.24	17.65	5.59
25.12	20.87	4.25
25.89	21.42	4.47
26.27	20.88	5.39
26.77	23.28	3.49
26.77	20.38	6.39
26.78	22.41	4.37
26.98	22.68	4.3



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IMPLEMENTATION OF GENETIC ALGORITHM IN R PROGRAMMING

In implementing the genetic algorithm in R tool, to generate a chart for the fitness values

Case 1

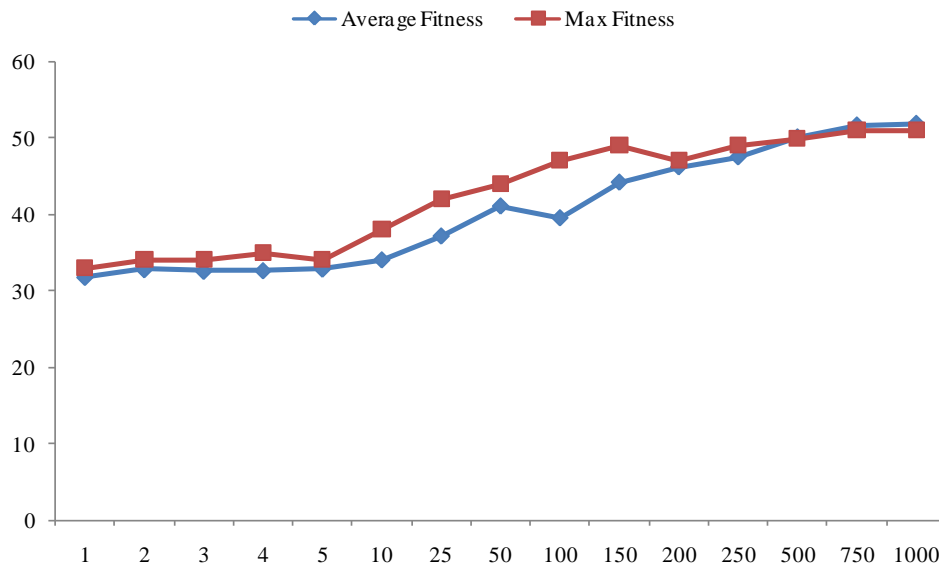


Fig 1 : Average fitness value with mutation rate = 0.001

Case 2

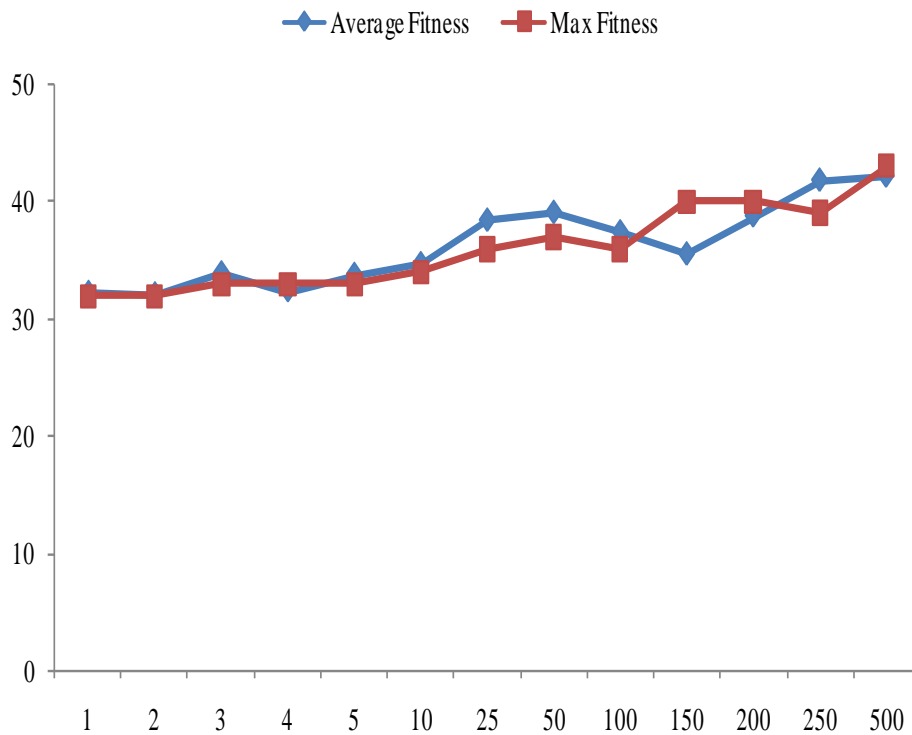


Fig 2 : Average fitness value with mutation rate = 0.01

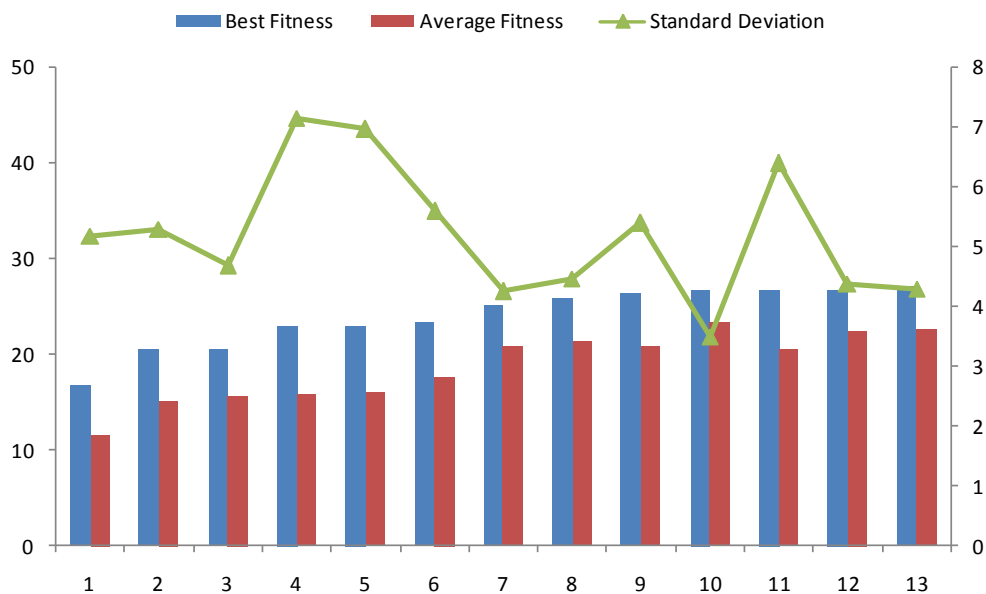
Best Fitness


Fig 3 : Best fitness value generation

CONCLUSION

In this paper to generate test cases, in implementing the genetic algorithm and to evaluate the fitness values in the Minimum values in the Minimum values. To find the optimum solution for the fitness function then to use the test cases in Genetic Algorithm. To most commonly use of a Genetic Algorithm is to initialize the population or chromosome, and to do the selection process and after is also done a crossover and mutation. Its a genetic operators then evaluating mechanism and the optimum solution is found. To implementing the test cases in the use of genetic algorithm as to give the optimal solution. When the help of R tool is used to generate a graph and to result as the optimal solution values. In future work is to implement the test case values are higher, and to use this test cases are in use of another algorithm and to use of PSO (Particle Swarm Optimization). Then it has to be concluded has the method of test case generation and to implementing the fitness values is optimized solution.

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