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## Lightweight Automated Testing for TSTL

----- Milestone 1 for CS569 Project

### 1 Introduction

For the Lightweight methods, it is involved some concepts, such as the random test, adaption-based programming, reinforce learning and so on. First, let me ask what is a lightweight automated testing method? 1. Easy enough to implement that it is essentially available for all languages and environments: if it does not exist, anyone can code it up in an afternoon. 2. Easy enough to use that any programmer interested in writing automated tests can quickly code up a harness for small, moderate complexity, modules. 3. Fast enough to produce results quickly, so automated testing can be pursued if useful and abandoned if not productive. Next, I need to mention that the archetypal lightweight automated testing method is random testing. Last, testing with adaptation-based programming is like random testing, but different. The idea of testing with adaptation-based programming is that replace calls to pseudorandom number generator with calls to library for reinforcement learning.

### 2 Reinforcement Learning Algorithm

First,  $s$  means the state,  $a$  means action,  $Q(s, a)$  show the overall report of a state  $s$  return to give an estimation of the action  $a$ ,  $r$  is the immediately report.

1. For every  $s$  and  $a$ , initialize the  $Q(s, a)$  to 0.
2. Observe the current state.
3. Repeat doing as following:  
Choose one action  $a$  and execute it, this action is action  $a$  that make  $Q(s, a)$  is the maximum .  
Receive the immediately report.  
Observe the new state  $s'$ .  
For the  $Q(s', a')$ , upload the data as following:  
$$Q(s, a) = r(s, a) + \gamma \max_{a'} Q(s', a').$$
  
 $s = s'$ .

The goal of Reinforcement Learning is to construct the control strategy to get the maximum performance.

In my project, I use reinforcement learning algorithm called SARSA( $\lambda$ ) that is a fairly easy algorithm to implement in almost language. The framework for ABP-based testing is almost identical to that used for random testing.

### 3 Operations

- 1- Timeout: It will show the time to cover a test with second.
- 2- Seed: Generate the random number from code for the Python random objects.
- 3- Depth: Show the maximum length of a test in the algorithm.
- 4- Width: Show the maximum memory/BFS queue with searching width.
- 5- Faults: Checking the faults in SUT either is 1 or 0.

- 6- Coverage: It should be reported coverage of testing as a final coverage report by using TSTL's internalReport() function.
- 7- Running: After running either 1 or 0 by using TSTL randomtester.py, generate the branch of coverage.

#### **4 Explanation for the work**

In this milestone 1, I just do the random testing work, because the ABP-based testing is like to that used for random testing, but they are different. In this project, I will implement it at last. But, so far, I need to finish the random test first, and then improved it to the ABP-based testing in the following work. Above, I did not show the random testing algorithm, because it is just a part of my work, not all. And I think most important and hard work is reinforce learning and its' algorithm.

#### **Reference**

- [1] Groce, A. Fern, J. Pinto & T. Bauer etl. "Lightweight Automated Testing with Adaptation-Based Programming" in International Conference on Software Engineering, 2007.
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- [4] <http://baike.baidu.com/view/1882942.htm>
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