Al for Software Testing and Reverse Engineering Lab 3

INTRODUCTION

Evolutionary algorithms (EAs) are population-based algorithms that are widely used to generate software inputs or tests in an automated fashion.

EAs methods use a guided random process for constructing new inputs/tests that try to trigger new code branches. They recombine previously tried inputs/tests guided by a heuristic that measures the proximity to conditions that trigger new branches (approach level and branch distance).

EAs, in particular genetic programming, are also used to patch existing software. The key idea is that given a set of pre-existing tests, some of which fail, we modify the code until all tests are satisfied. You use the operations from EAs such as crossover and mutation to perform these modifications, and the number of tests satisfied as fitness function.

In this lab, you will construct and experiment EAs, understand how they work, and learn how to use them to perform automated software patching.

LEARNING OUTCOMES

After completing this assignment, you will be able to:

- Automatically build test sets that maximize branch coverage
- Use test sets to locate faults in software
- Build EAs for software patching
- Use and understand modern EA-based patching tools such as Astor

INSTRUCTIONS

For this assignment, we will again use the RERS problems. Similar to the previous assignment, we have provided you with the instrumentation and you should only need to modify *PatchingLab.java* for this assignment. Before you begin with this assignment, make sure you have pulled the latest changes from the *JavaInstrumenation* repository:

- · First navigate to the *JavaInstrumentation* folder.
- Then pull the latest changes by running the following command: git pull

Before beginning with this assignment, make sure you have read the instructions on how to instrument the RERS problem for this lab assignment. The instructions can be found on the README of the GitHub repository:

• https://github.com/apanichella/JavaInstrumentation/blob/main/README.md

Task 1: building a test-set (Optional)

To see whether our EA is doing well in the patching, we need test cases for the RERS problems. As each RERS problem outputs a string based on the given input, we can use the input-output behaviour of the RERS problem to generate test-cases. We have already generated test cases for each RERS problem for you. The test cases can both be found on BrightSpace and on Github (in the folder named rers2020_test_cases). However, if you want you can use your own solution from Labs 1-2 to generate the cases yourself. Tip: use your best performing fuzzer to generate the test cases, as the best performing fuzzer has achieved the highest branch coverage.

You can feel free to play around with the tool that we have used to generate the test cases:

• https://github.com/TCatshoek/AISTRTestCaseGenerator

We believe that the test cases that we have generated are complete but we might have overdone it or miss some cases. So if you have a better generator, feel free to generate test cases using your own generator.

The test-cases have the following form:

```
iA, iB, iC->oXoYoZ
iA, iB, iD->oXoXoZ
```

i.e, the RERS input symbols, separated by "," characters, followed by "->", followed by a concatenation of all produced outputs. For this task, you are allowed to build your own test cases for the RERS problem but you have to make sure that you use the same format. *OperatorVisitor.java* contains code to add instrumentation to track the generated outputs:

This calls the *output* method in *OperatorTracker.java* whenever something is printed in the standard out. You may use this method but feel free to implement your own method to collect input-output pairs that together cover as many branches as you are able to.

Task 2: locating bugs

You are provided with buggy versions of each Reachability problem, where only operators are modified. Your task is to automatically locate the faulty operators using the test sets that were provided to you in Task 1 or use the test sets that you have generated yourself. Of course, one could simply use a code difference method to locate the changes, but in practice you will only have access to failing tests and not a bug-free implementation. Start your task by computing the initial fitness function for the sets you generated on the buggy Reachability problems, i.e., compute the number of failing tests for each problem. The test set is loaded using the *readTests* method.

In *OperatorTracker*, there are two methods that you can use to run an individual test (*runTest*) or all the tests (*runAllTests*) that are listed in the given test set. You can call these methods from *PatchingLab.java*.

First, we compute fault localization to determine which operators are more likely to be faulty by computing how frequently an operator is used in a failing test. *OperatorTracker.java* keeps track of all encountered binary operators in if-statements (>, <, >=, <=, ==, and !=). For every encounter it calls

```
boolean encounteredOperator(String operator, int left, int right, int
operator_nr)

or

boolean encounteredOperator(String operator, boolean left, boolean
right, int operator_nr)
```

Where *operator* is the encountered operator, *left* and *right* are the left and right hand sides of the operator and *operator_nr* is an identifier for the operator (from top to bottom, all operators in the instrumentation are assigned a unique number). In contrast to lab assignments 1 and 2, in this lab the instrumentation directly influences the functionality of the code. The returned boolean is used instead of the encountered operator in the instrumented if-statement.

Implement the fault localization strategy that uses the tarantula score and use this as part of your genetic algorithm to fix bugs. Describe in pseudocode (max 15 lines) how you utilise the Tarantula score to patch the buggy operators.

Task 3: fixing bugs

Use the number of failing tests as a fitness value for a genetic algorithm with fault localization to guide mutations. The instrumented code is set up in such a manner that you do not need to recompile the code in order to test mutations of operators (which only modified these operators) without the need to recompile. The current operator list is maintained in OperatorTracker:

OperatorTracker.operators[operator nr]

The result of each test is stored in *OperatorTracker.test_result*. Use this to compute a fitness function. Your algorithm needs to contain the following components:

- Mutation is a random change or one operator into another, guided by fault localization.
- Implement any selection strategy e.g. tournament selection.
- Use as the number of passing tests as the search heuristic.

Experiment with mutation rates and show the performance of your genetic algorithm by plotting a convergence graph (best fitness over time) and what percentage of faults you have correctly patched for Reachability problems that you have chosen to run your algorithm on.

Task 4: Genetic Programming using ASTOR

You are asked to use the tool ASTOR (https://github.com/SpoonLabs/astor) to find a patch for buggy versions of the RERS problems. ASTOR is already available in the docker container we provided for this course. Detailed instructions on how to run ASTOR on the three RERS problems above are also posted on Brightspace.

Step 1: First, you have to generate test cases for the original version (without bugs) of the three RERS problems above. You can use the test cases we provide. These are included in the packages on Brightspace, generated using EvoSuite.

Step 2: Your job is to run ASTOR to find a patch to the target bug. For the assignment, you need to run ASTOR using Genetic Programming as patch engine. You will set this engine using the parameter **-mode jgenprog**. More details information on how to run ASTOR are available on Brightspace and on GitHub.

Step3: For each buggy version of the RERS problem, run ASTOR once. Then, report on statistics regarding:

- The time it needed to find the patches (if any).
- How many patches were found by ASTOR.

Then, manually analyze the generated patches and answer the following questions:

- 1. Does ASTOR generate a meaningful patch? Use a diff-tool to analyze the code between the original RERS problems, their buggy versions, and the patched versions.
- 2. Did ASTOR generated meaningful patches? Is genetic programming effective at finding patches, why (not)?

RESOURCES

Slides from Lectures 5

Study:

- (1) Le Goues, Claire, et al. "Genprog: A generic method for automatic software repair." Ieee transactions on software engineering 38.1 (2012): 54-72
- (2) Martinez, Matias, and Martin Monperrus. "Astor: A program repair library for java." Proceedings of the 25th International Symposium on Software Testing and Analysis. ACM, 2016.

PRODUCTS

A small report of max two A4 pages (excluding visualizations) answering the questions from tasks 2, 3 and 4. The report can be extended with at most two A4 pages if visualizations are used.

An archive (tar/zip) containing the code for computing the results.

Make sure you have also provided some instructions on how to run your code.

ASSESSMENT CRITERIA

You can either pass or fail this assignment. We expect everyone to pass. You have to complete all lab assignments. Submissions of which the report text does not fit into two A4s (excluding visualizations) will not be evaluated; the submission will automatically result in a fail.

Code that does not compile/run will not be evaluated; the submission will automatically result in a fail.

Your work will be evaluated based on its completeness (having done all tasks), the correctness of the implementation, and demonstration that you understand the results in the analysis. When deemed insufficient, you will receive feedback and will be given a one-week grace period to fix any shortcomings.

SUPERVISION AND HELP

There will be lab sessions every Wednesday, where the teachers and TAs will be available to answer any questions you may have. The preferred way to ask questions is through Mattermost.

SUBMISSION AND FEEDBACK

The submission is through Brightspace. You will receive feedback within one week after the deadline.