

# Automated Debugging

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Prof. Dr. Andreas Zeller  
Paul Zhu  
Marius Smytzek

## Exercise 7 (10 Points)

**Due: 27. January 2023**

Submit your solutions as a Zip file on your status page in the [CMS](#).

We will provide you a structure to submit your solutions where each task has a dedicated file. You can add new files and scripts if you want, but you must not delete any provided ones. You can verify whether your submission is valid by calling `python3`.

```
python3 verify.py
```

The output provides an overview if a required file, variable, or function is missing and if a function pattern was altered. If you do not follow this structure or change it, we cannot evaluate your submission. A non evaluable exercise will result in 0 points, so make sure to verify your work before submitting it. Note that the script does not reveal if your solutions are correct.

### Exercise 7-1: No Silver Bullet (5 Points)

Delta debugging and abstract failure-inducing input are powerful in many scenarios. But you can always find scenarios where these two technologies become not so helpful. One example is heartbeat ( `exercise_1.py` ), which we have seen several times in the previous exercises:

```
In [1]: data = 'password:hjasdiebk456jhaccount:smytzek'

def store_data(payload: str):
    global data
    data = payload + data

def get_data(length: int) -> str:
    return data[:length]

def heartbeat(length: int, payload: str) -> str:
    assert length == len(payload)
    store_data(payload)
    assert data.startswith(payload)
    r = get_data(length)
    assert r == payload
    return r
```

To better understand why these two technologies become less helpful on the heartbeat example, let's conduct an experiment as follows:

1. Build a grammar for the `payload` argument that can be of arbitrary length.
2. Use the `GrammarFuzzer` introduced in the lecture to generate a random input for the `heartbeat` function (you may fix the `length` argument, say fix it to 5, to avoid fuzzing integers) so that it fails (i.e., to violate the assertions).
3. Apply `DeltaDebugger` on the random failure input you found: What minimized input do you get? How well does it explain the failure reason?
4. Apply `DDSetDebugger` on the random failure input you found: What generalized pattern do you get? Based on this pattern:

- 4.1. Do you think the pattern is correct, that is, all of its instantiations are failure inputs to `heartbeat` ?
- 4.2. (Now, to examine your thoughts on 4.1) Invoke the `fuzz_args` method of `DDSetDebugger` to produce 10,000 instantiations of the generalized pattern, calculate how many of them are failure inputs to `heartbeat` -- let's call this rate the *success rate*. Is the success rate high?
- 4.3. To enhance the informativity/expressiveness of `DDSetDebugger` -- so that it becomes more helpful for the `heartbeat` case -- what do you think can be a potential improvement or extension?

Please submit a **report** that describes how you conduct the experiment following the above steps. Specifically, your report should include:

- The Python code (including the grammar of step 1) you use for the experiment.
- The outputs of your code: the random input (step 2), the minimized input (step 3), the generalized pattern and the success rate (step 4).
- Your answers to the questions mentioned above.
- Optionally, any explanation to any piece of hard-to-understand code/function.

The report can be in **either** format:

- A markdown document `exercise_1.md` consisting of code and text (for your outputs, answers and additional explanations).
- Or, a Jupyter notebook (if you learned how to use it -- but not required for this course) consisting of executable code, running outputs, and markdown blocks for the answers and additional explanations. Remember to rename `exercise_1.md` to `exercise_1.ipynb`.

Keep your report clear, structured (with titles numbering the steps), easy-to-follow, and even better -- simpler and shorter. Misunderstanding may lead to low scores for this exercise.

## Exercise 7-2: Semantic-Preserving Program Slicing (5 Points)

As mentioned in the book, one particularly fun application of delta debugging is to reduce program code -- a fancier terminology for this operation is *program slicing*. You learned flow-analysis-based slicing in [this chapter](#). In this exercise, you are going to investigate another kind of program slicing via a combination of AST-based delta debugging and testing.

Recall in the delta debugging chapter, [this section](#) introduces how to reduce program code on AST nodes so that we can find a minimized program that triggers the failures (i.e., failed tests). Thinking oppositely, we could also find a minimized program that **passes** all the tests is found, and this minimized program can be regarded approximately semantically equivalent to the original program since they have the **same behaviors** according to the tests. The minimized program is thus regarded as the semantic-preserving slice.

To implement such a slicing functionality, we can reuse the `DeltaDebugger`, in a similar way to what we learned from the lecture:

```
with DeltaDebugger() as dd:
    compile_and_test_ast(fun_tree, fun_nodes, test_tree)
```

But here, the `compile_and_test_ast` function should execute a piece of code (manually constructed from AST) that only fails when: all the tests are **passed**, i.e., the function represented by `test_tree` does not raise any exception. In this way, the delta debugger will be able to find a minimized program that still fails `compile_and_test_ast` -- but, this is equivalent to the situation where all the tests defined in `test_tree` get passed. Finish your implementation of this function in `exercise_2.py`:

```
def compile_and_test_ast(tree: ast.Module, keep_list: List[ast.AST],
                        test_tree: ast.FunctionDef) -> None:
    pass # YOUR CODE HERE
```

As an example, you can test your `compile_and_test_ast` on slicing the Fibonacci function:

```
In [7]: def fib(n: int) -> int:
        if n == 0 or n == 1:
            return 1
        return fib(n - 1) + fib(n - 2)

def fib_test_simple():
```

```
assert fib(0) == 1
assert fib(1) == 1
```

```
fib_test_simple()
```

First, create the input ASTs:

```
In [8]: import inspect, ast
from debuggingbook.DeltaDebugger import NodeCollector, DeltaDebugger
from debuggingbook.bookutils import print_content

fun_tree: ast.Module = ast.parse(inspect.getsource(fib))
fun_nodes = NodeCollector().collect(fun_tree)
test_tree: ast.FunctionDef = ast.parse(inspect.getsource(fib_test_simple)).body[0]
```

Then, apply the `DeltaDebugger` :

```
with DeltaDebugger() as dd:
    compile_and_test_ast(fun_tree, fun_nodes, test_tree)

reduced_nodes = dd.min_args()['keep_list']
reduced_fun_tree = copy_and_reduce(fun_tree, reduced_nodes)
print_content(ast.unparse(reduced_fun_tree), '.py')
which finds the following slice:
```

```
def fib(n):
    return 1
```

This makes sense as the test `fib_test_simple` only covers the cases when `n == 0` or `n == 1`. If you add another testcase, say `assert fib(4) == 5`, then all branches of `fib` is covered so that nothing can be reduced for the `fib` function.