Automatically Finding Patches Using Genetic Programming

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Motivation

- Software Quality remains a key problem
 - Over one half of 1 percent of US GDP each year
 - Programs ship with known bugs
- Reduce debugging costs
 - Bug reports accompanied by patches are addressed more rapidly
- Automated Patch Generation
 - Transform a program with a bug
 - Into a program without the bug
 - By modifying relevant parts of the program

The Cunning Plan

- We can automatically and efficiently repair certain classes of bugs in off-the-shelf, unannotated legacy programs.
- Basic idea: Biased search through the space of all programs until you find a variant that repairs the problem. Key insights:
 - Use existing **test cases** to evaluate variants.
 - Search by perturbing parts of the program **likely** to contain the error.

The Process

• Input:

- The program source code
- Some regression test cases passed by the program
- A test case failed by the program (= the bug)
- Genetic Programming Work:
 - Create variants of the program
 - Run them on the test cases
 - Repeat, retaining and combining variants
- Output:
 - New program source code that passes all tests
 - or "no solution found in time"

This Talk

- Genetic Programming
- Weighted Paths
- Our Technique
- Example
- Repair Experiments
- Big Finish

What's In A Name?

- Genetic programming is the application of evolutionary or genetic algorithms to program source code.
 - Population of variants
 - Mutation, crossover
 - Fitness function
- Similar in ways to searchbased software engineering:
 - Regression tests to guide the search



Two Secret Sauces

- In a large program, not every line is equally likely to contribute to the bug.
- Insight: since we have the test cases, run them and collect coverage information.
- The bug is more likely to be found on lines visited when running the failed test case.
- The bug is less likely to be found on lines visited when running the passed test cases.
- Also: Do not try to invent new code!

The Weighted Path

- We define a weighted path to be a list of <statement, weight> pairs.
- We use this weighted path:
 - The statements are those visited during the failed test case.
 - The weight for a statement S is
 - High (1.0) if S is not visited on a passed test
 - Low (0.1, 0.0) if S is also visited on a passed test

Genetic Programming for Program Repair: Mutation

- Population of Variants:
 - Each variant is an <AST, weighted path> pair
- Mutation:
 - To mutate a variant $V = \langle AST_v, wp_v \rangle$, choose a statement S from wp_v biased by the weights
 - Delete S, replace S with S1, or insert S2 after S
 Choose S1 and S2 from the entire AST
 - Assumes program contains the seeds of its own repair (e.g., has another null check elsewhere).

Genetic Programming for Program Repair: Fitness

- Compile a variant
 - If it fails to compile, Fitness = 0
 - Otherwise, run it on the test cases
 - Fitness = number of test cases passed
 - Weighted: passing the bug test case is worth more
- Selection and Crossover
 - Higher fitness variants are retained and combined into the next generation
- Repeat until a solution is found



Example: Abstract Syntax Tree



Example: Weighted Path (1/3)



Example: Weighted Path (2/3)



Example: Weighted Path (3/3)



Example: Mutation (1/2)



Mutation Source: Anywhere in AST Mutation Destination: Weighted Path

Example: Mutation (2/2)



Mutation Source: Anywhere in AST Mutation Destination: Weighted Path

Example: Final Repair



Minimize The Repair

- Repair Patch is a diff between orig and variant
- Mutations may add unneeded statements
 - (e.g., dead code, redundant computation)
- In essence: try removing each line in the diff and check if the result still passes all tests
- Delta Debugging finds a 1-minimal subset of the diff in $O(n^2)$ time
 - Removing any single line causes a test to fail
- We use a tree-structured diff algorithm (diffX)
 - Avoids problems with balanced curly braces, etc. 19

Experimental Results

Program	LOC	Path	Time (s)	Success	Bug
gcd	22	1.3	149	54%	inf loop
uniq	1146	81.5	32	100%	segfault
ultrix look	1169	213.0	42	99%	segfault
svr4 look	1363	32.4	51	100%	inf loop
units	1504	2159.7	107	7%	segfault
deroff	2236	251.4	129	97%	segfault
nullhttpd	5575	768.5	502	36%	buffer overrun
indent	9906	1435.9	533	7%	inf loop
flex	18775	3836.6	233	5%	segfault
atris	21553	34.0	69	82%	buffer overrun
average		881.4	184.7	58.7%	

Average minimization time: 12 seconds. Total: **10** repaired programs, over **63,000** lines of code.

Repair Quality

- Repairs are typically *not* what a human would have done
 - Example: our technique adds bounds checks to one particular network read, rather than refactoring to use a safe abstract string class in multiple places
- Recall: any proposed repair must pass all regression test cases
 - When POST test is omitted from nullhttpd, the generated repair eliminates POST functionality
 - Tests ensure we do not sacrifice functionality
 - Minimization prevents gratuitous deletions
 - Adding more tests helps rather than hurting

Technique Limitations

- May not handle nondeterministic faults
 - Difficult to test for race conditions, etc.
 - Long term: put scheduler constraints into the variant representation.
- Assumes bug test case visits different lines than normal test cases
- Assumes existing statements can form repair
 - Current work: repair templates
 - Hand-crafted and mined from CVS repositories
- Slower on large test suites: test case selection₂₂

Want to hear more?

ICSE 2009

- Formal algorithm, crossover, mutation
- Representation, parsing, stmt details
- Test cases used
- Sensitivity
- Repair quality
- "Does it work?"

GECCO 2009

- Evolutionary questions
 - nonstandard crossover
 - really evolutionary?
 - operator frequency
- Effect of more test cases
- Scaling behavior
- "Why did it work?"

Conclusions

- We can automatically and efficiently repair certain classes of bugs in off-the-shelf legacy programs.
 - Ten programs totaling 63kloc in about 6 minutes each, on average
- We use regression tests to encode desired behavior.
 - Existing tests encode required behavior
- The genetic programming search focuses attention on parts of the program visited during the bug but not visited during passed test cases.

Questions

• I encourage difficult questions.

Bonus Slide: Test Cases

```
#!/bin/sh
1
  # Positive Test Case for nullhttpd (POST data)
2
3
  ulimit -t 5
  /usr/bin/wget --tries=1 --post-data 'name=my name&submit=submit'
4
5
     "http://localhost:\$PORT/cgi-bin/hello.pl"
   if diff hello.pl ../known-good-hello.pl-result ; then
6
     # if the current output matches the known-good output
7
     echo "passed hello.pl test case" >> ../list-of-tests-passed
8
9
   fi
```

Figure 2: Positive test case for nullhetpd. wget is a command-line HTTP client; ulimit cuts the test off after five seconds. The test assumes that the sandboxed webserver is accepting connections on **PORT** and has its own copy of htdocs, including cgi-bin/hello.pl. Note the oracle comparison using diff against known-good-hello.pl-result on line 6.

```
#!/bin/sh
1
  # Negative Test Case for nullhttpd
2
 ulimit -t 5
3
4 ../nullhttpd-exploit -h localhost -p $PORT -t2
5 /usr/bin/wget --tries=1 "http://localhost:$PORT/index.html"
  if diff index.html ../known-good-index.html-result ; then
6
    # if the current output matches the known-good output
7
    echo "passed exploit test case" >> ../list-of-tests-passed
8
  fi
9
```

Figure 3: Negative test case for nullhttpd. If the exploit (line 4) disables the webserver then the request for index.html (line 5) will fail.