Static analyzer debugging and quality assurance approaches

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About the author & the project

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- Software engineer.
- (ex. security analyst; participated in commercial debugger project).

Equid<sup>1</sup> — a static analyzer for C/C++/RuC based on Model Checking and Abstract Interpretation. It verifies contracts and finds common defects.

<sup>&</sup>lt;sup>1</sup>Engine for performing queries on unified intermediate representations of program and domain models

What's special in analyzers?

Not much.

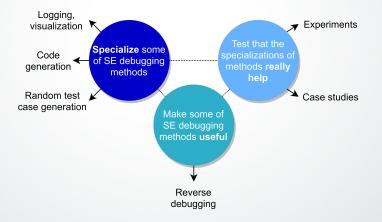
- Many equivalent transformations: input format ≠ intermediate format ≠ output format.
- Intermediate representations are mostly internal.
- The code is usually consistent and has high integrity, but there are logical mistakes, unprocessed parts → the biggest defects are logical.

# The problem

- There are many debugging & quality assurance methods.
- None of them are specialized enough for static analysis.
- Every project brings its own set of hardly formalized methods.

What if we find a right specialization of the methods to the static analysis field?

# The paper's goal



# Defect sources (observations)

- Missing support for the specific syntax/intermediate representation (IR) construction in submodules.
- Small differences in implementations for repeating parts (classes).
- Transformation and ordering issues.

## Defect reasons (observations)

- Low visuality of the transformation passes and the development process.
- Unattainable cross-dependencies between modules.
- Low quality of tests.

# **Proposed solutions**

Proposing the solutions of these three groups:

Code generation:

Generated code usage verification.

• Testing:

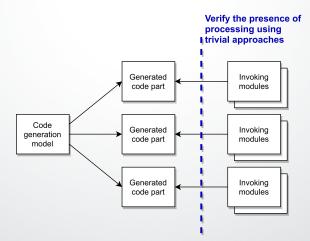
Goal-driven random test case generation.

• Logging:

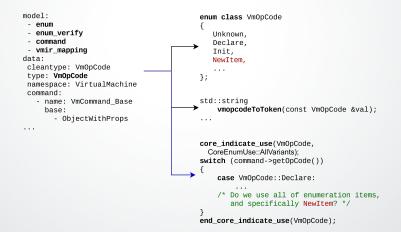
Log fusion and visual representation.

# Code generation

- One model, several interpretations, many output source files.
- Perform a simple integrity check.

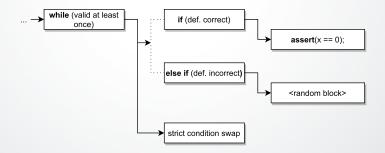


# Code generation: enumeration example



Goal-driven random test case generation

The idea is: generate input programs with an integrated verification goal (assertion).



Goal-driven random test case generation

- 1. The tool generates a random goal and asserts it  $\rightarrow$  an expression.
- The expression is repeatedly rolled into if/switch/for/while/... random blocks → a block.
- 3. The meaningful blocks are shuffled using equivalent transformations.

Goal-driven random test case generation

In result we get:

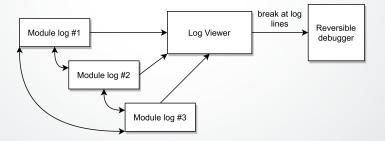
- 1. A completely random program.
- 2. A set of shuffled random programs.

By that, it is possible to verify:

- 1. Logical issues in transformations.
- 2. Ordering issues.
- 3. Runtime failures.

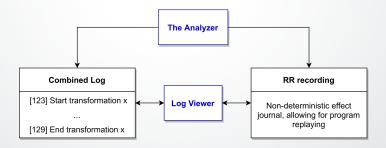
# Log fusion

Fuse separate logs, set up cross-references, so the final log is a technical documentation of the run. Allows for easy navigation.



# Log fusion: reverse recording assistance

The Log Fusion also helps break right after the specific log line using reversible debugger like RR<sup>2</sup>, UndoDB, etc. That is achieved using logging engine traps and GDB scripts.



<sup>2</sup>https://rr-project.org

# Visual representation: steps

Visualize steps — present all transformations in one window, allow to debug specific transformations.

Log	Performance Transformation	Logs	
Source:	y = 3;	Resources Containers	
VM IR:	assign (y:int res) = 3	Name assign (Lnopint(fres) = 0 branch ((Lnopint(fres) = 1 assign (Lnopint(fres) = 1 assign (Lnopint(fres) = 1 assign (Lnopint(fres) = 2 end branch branch ((Lnopint(fres) = 1) branch ((Lnopint(fres) = 1))	
CVC4:	(var_1_6 == 1    var_1_6 == 0 && var_1_2 == 5) => (var_1_4 = 3 && addrof(var_1_4) = var_345 && deref(var 345) = var 1 4)		
AI:	var_1_4 = [3, 3]	Fragments Stages	

# Visual representation: log health

Log health — visualize the time allocation for different modules. In result, it is possible to determine whether the specific part is unintentially skipped.

aisolver	9%	=
cvc4solver	86%	8_
cvc4solverType	9%	
debug	86%	
diagmodel	9%	
executionmodel	0%	
flattening	0%	
fragmentvisitor	12%	
hypervisor	0%	
multisolver	9%	
output	0%	
parsing	13%	₩_₩=
resourcevisitor	12%	
stpsolver	0%	
tagvisitor	11%	
virtualprocessor	0%	

# Random test case generation: discovered issues & their severity

The method detected many performance, ordering, logical issues, and even runtime failures.

Defect type	Number of issues	Severity
Performance	3	Medium
Ordering	5	High
Runtime failure	1	High
Logical issues	1	Medium

# Log fusion: (rough) time to resolve the issues

#### Average improvement rate: 2.8.

Defect type	Time to resolve before (h)	Time to resolve after (h)
Performance	25	13
Ordering	5	1
Runtime failure	1	0.3
Logical issues	1	1

Results: code generation and visual representation

The improvement is hard to examine.

In our experiment, developing the same feature twice took 7 times less time than on previous iteration - thanks to code generation.

Visual representation allowed to discover at least 2 performance issues, and overall provided an enormous help during defect resolution.

### Conclusion

- The specialization of the proposed methods helps find real issues in the static analyzer.
- The combination of approaches dramatically decreases the defect resolution time.