

A Review on “AUTOVERUS: Automated Proof Generation for Rust Code”

Gijung Im

FCAI Lab
Yonsei University

Jan 26, 2026



Paper Reading

- **Title:** AUTOVERUS: Automated Proof Generation for Rust Code
- **Source:** OOPSLA 2025
- **Authors:** Microsoft Research (Weidong Cui, et al.), UIUC, and Columbia Univ.
- **Presenter:** Gijung Im



Table of Contents

1 Introduction

- Motivation
- Background: Verus
- Unique Challenges

2 Overview

- Architecture
- Procedure

3 Experiment & Evaluation

- Evaluation Setup
- Experimental Results

4 Conclusion



Motivation

1. Generative AI in Software Engineering:

- LLMs excel at code generation (e.g., GitHub Copilot)
- but lag in generating formal proofs.

2. The Need for Correctness:

- Developers often distrust AI-generated code.

3. Goal:

~~ To automatically generate correctness proofs for Rust code,
allowing developers to enjoy both **productivity** (AI) and **reliability** (Verification).



Verus: Verified Rust for Low-level Systems

- A static verification tool for Rust that uses an SMT solver (Z3).
- Allows developers to write specifications and proofs directly in Rust-like syntax.
- Key Philosophy:
 1. *Implementation and Proof in the same language.*
No need to learn separate languages like Rocq or Dafny.
 2. *Zero-overhead.*
All “ghost code” (specifications/proofs) are erased at compile time, leaving only optimized Rust executables.



The Mode System (spec vs. proof vs. exec)

```

1 spec fn is_digit(c: u8) -> bool {
2     c >= 48 && c <= 57
3 }
4
5 spec fn cnt_dig(seq: Seq<u8>) -> int
6     decreases seq.len(),
7 {
8     if seq.len() == 0 {
9         0
10    } else {
11        cnt_dig(seq.drop_last()) +
12        if is_digit(seq.last()) {
13            1 as int
14        } else {
15            0 as int
16        }
17    }
18}

```

(a) The spec functions.

```

19 fn count_digits(text: &Vec<u8>) -> (ret: usize)
20     ensures ret == cnt_dig(text@),
21 {
22     let mut count = 0;
23     let mut i = 0;
24     while i < text.len()
25         invariant
26         i <= text.len(),
27         count <= i,
28         count == cnt_dig(text@.subrange(0, i as int)),
29     {
30         if text[i] >= 48 && text[i] <= 57
31             {count += 1;}
32         i += 1;
33         assert(text@.subrange(0, i - 1 as int)
34             == text@.subrange(0, i as int).drop_last());
35     }
36     assert(text@ == text@.subrange(0, i as int));
37     count
38}

```

(b) The implementation.

Fig. 1. A Rust function, in gray background, with Verus annotations. The dark-yellow background highlights the proof annotation needed by Verus to prove the specifications highlighted in the light-yellow.



Challenges in Verus Proof Generation

1. Data Scarcity.

- Verus is young (≤ 3 years old), #GitHub repos ≤ 10 .
- Unlike Rocq or Isabelle, there is no massive dataset for training.

2. Syntax Subtleties.

- Verus introduces specific syntax extensions—e.g., the `@` operator, `int` vs. `u64`.
- LLMs often confuse ***ghost code*** rules with standard Rust rules.

3. Feedback Ambiguity.

- Unlike Interactive Theorem Provers (ITP) that show proof state, SMT solvers essentially return either “Pass” or “Fail”.
- It is difficult to measure progress when verification fails.



Workflow

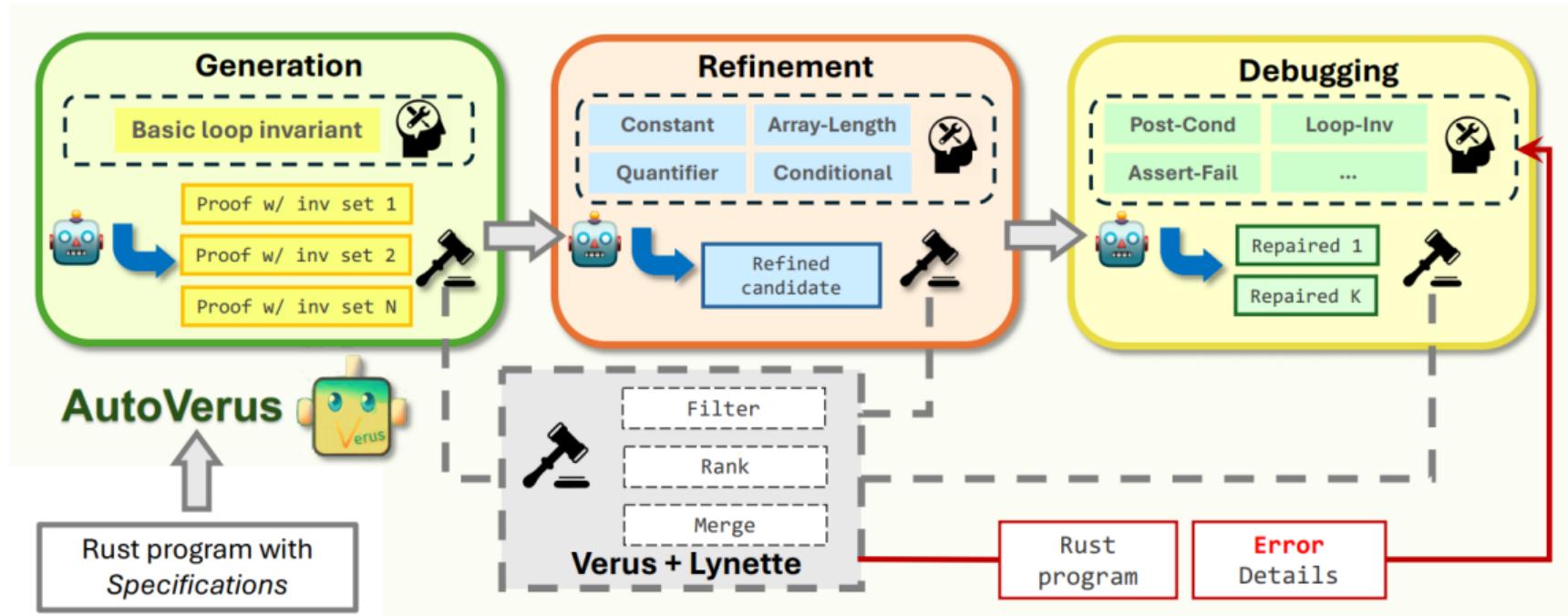


Fig. 2. The workflow of AUTOVERUS



Phase I—Generation (Preliminary Proof)

- **Objective:** Generate initial loop invariants.
- **Post-Processing.**

```
1 def is_correct(p):
2     return verus_correct(p) or verus_correct(houdini(p))
3     # The Houdini algorithm deletes the unprovable proof annotations from p.
4
5 def GenPrelimProof(program):
6     pCanidates = LoopInvAgent(program) # proof candidates
7     safeCandidates = { p ∈ pCanidates | is_safe(program, p) }
8     Sort(safeCandidates, comparison=(V,E)-rank) # V: verified, E: erroneous
9     mergeP = program # merged proof
10    for p in safeCandidates:
11        mergeP = merge(mergeP, p)
12        if is_correct(mergeP):
13            return mergeP
14    return safeCandidates[0] # first element
```

Listing 1: Preliminary proof generation with post-processing



Example—Diffy Benchmark

```

1 invariant
2 N > 0,
3 i <= N as usize,
4 sum.len() == 1,
5 sum[0] <= N + i,
6 forall|k: int|0<=k<N ==> b[k]==1,
7 b.len() == N,
8// (V, E) = (4, 2)

```

(a) Loop invariants of best-so-far P_B .

```

1 invariant
2 N > 0,
3 i <= N as usize,
4 sum.len() == 1,
5 sum[0] <= N + i,
6 forall|k: int|0<=k<N ==> b[k]==1,
7 N < 1000,
8// (V, E) = (3, 2)

```

(b) Loop invariants of a lower-ranked P_i .

```

1 i = 0;
2
3 while (i < N as usize)
4   invariant
5   N > 0,
6   i <= N as usize,
7   sum.len() == 1,
8   sum[0] <= N + i,
9   forall|k: int|0<=k<N ==> b[k]==1,
10  b.len() == N,
11  N < 1000,
12{
13  sum.set(0, sum[0] + b[i]);
14  i = i + 1;
15}
16// (V, E) = (5, 0)
17// Verified

```

(c) The merged program P' which is *verified*.



Fig. 5. An example of a merged program for a problem from Diffy benchmark.



Phase II–Refinement (Generic Fixes)

- **Objective:** Correct common mistakes before consulting the verifier.
- **4 Specialized Agents.**
 - *Constant-Propagation:* Propagates preconditions into every loop.
 - *Array-Length:* Adds invariants about every array/container sizes (e.g., `len()`).
 - *Quantifier:* Checks whether quantifier-related invariants are used correctly.
 - *Conditional-Loop-Invariant:* Adjusts invariants for specific iterations of a loop.

```
1 def RefineProof(program):
2     best = program
3     # from the simplest (i.e., Constant-Propagation) to the more complex ones
4     for refineAgent in refineAgentList:
5         p = refineAgent(best)
6         if is_correct(p):
7             return p
8         if accept_refine(p):
9             best = p
10    return best
```

Listing 2: Proof Refinement



Phase III—Debugging (Error-Driven)

- **Objective:** Iteratively fix verification errors reported by Verus.

```

1 def DebugProof(program):
2     for _ in range(MaxIter):
3         if is_correct(program):
4             return program
5         errors = verus_compile(program)
6         e = select_error(errors)
7         repair_agent = select_repair_agent(e)
8         repairedCandidates = repair_agent(program, e)
9         for p in repairedCandidates:
10            if is_correct(p):
11                return p
12            if accept_repair(p):
13                program = p
14                break
15
    return program

```

Listing 3: Debug Proof Generation

repair_precond_error
repair_precond_veclen
repair_postcond_error
repair_invfail_front
repair_invfail_end
repair_assertion_error
repair_arithmetic_flow
repair_mismatch_type
repair_assertion_error_with_proof_func
repair_default

Fig. 6. AUTOVERUS repair agent list



Fibonacci Sequence (Iteration 1 & 2)

```

1 fn fibonacci(n: usize) -> (ret: Vec<i32>)
2   requires
3     fibo_fits_i32(n as int),
4     n >= 2,
5   ensures
6     forall|i: int| 2 <= i < n ==>
7       #[trigger] ret@[i] == fibo(i),
8     ret.len() == n,
9 {
10   let mut fib = Vec::new();
11   fib.push(0);
12   fib.push(1);
13   let mut i = 2;
14   +1 // Assert the invariant right before the loop
15   +1 assert(fib.len() == 2);
16   while i < n
17     invariant
18       0 <= 2 <= i <= n,
19       fib.len() == i,
20       fibo_fits_i32(n as int),
21       forall|j: int| 0 <= j < i ==>
22         # [trigger] fib[j] == fibo(j),

```

```

23 -1   fib.len() == n;
24   {
25 +2 // Assert the addition won't overflow an i32
26 +2 assert(fib[i-1] as int + fib[i-2] as int
27 +2           < 0x8000_0000) by {
28 +2   assert(fib[i-1] == fibo(i-1));
29 +2   assert(fib[i-2] == fibo(i-2));
30 +2   assert(fibo(i-1)+fibo(i-2)==fibo(i));
31 +2   lemma_fibo_monotonic(i, n);
32 +2 };
33   let next_fib = fib[i-1] + fib[i-2];
34   fib.push(next_fib);
35   i += 1;
36 }
37 fib
38}
39 +2 proof fn lemma_fibo_monotonic(..) {}

```

- (a) Repair the “the loop invariant not satisfied before loop” and “arithmetic overflow” errors.



Fibonacci Sequence (Iteration 3)

```

1spec fn fibo(n: int) -> nat
2  decreases n,
3{
4    if n <= 0 { 0 }
5    else if n == 1 { 1 }
6    else { fibo(n - 2) + fibo(n - 1) }
7}
8spec fn fibo_fits_i32(n: int) -> bool {
9    fibo(n) < 0x8000_0000
10}

```

(b) The spec functions.

```

1proof fn lemma_fibo_monotonic(n:int, m:int)
2  requires n <= m,
3  ensures fibo(n) <= fibo(m),
4  decreases m - n
5{
6  if n < m {
7    lemma_fibo_monotonic(n, m - 1);
8+ assert(fibo(n) <= fibo(m - 1));
9+ assert(fibo(m - 1) <= fibo(m));
10 }
11}

```

(c) Repair the post-condition not satisfied error.



Fibonacci Sequence (Iteration 4 & 5)

```
1proof fn lemma_fibo_monotonic(n:int, m:int)
2  ...
3{
4  if n < m {
5    lemma_fibo_monotonic(n, m - 1);
6    assert(fibo(n) <= fibo(m - 1));
7- assert(fibo(m - 1) <= fibo(m));
8+ if m > 1 { // Avoiding the case m == 1 to
9+   prevent negative indexing in fibo(m-2)
10+  assert(fibo(m-2)+fibo(m-1)==fibo(m));
11+  assert(fibo(m - 1) <= fibo(m));
12}
13}
```

(d) Repair the assertion failure error.



Verus-Bench

- A curated suit of 150($= 78 + 38 + 11 + 23$) non-trivial proof tasks.
 - *MBPP*: 78 tasks (Python \rightarrow Rust)
 - *Diffy*: 38 tasks (C benchmarks)
 - *CloverBench*: 11 tasks
 - *Misc*: 23 tasks (Algorithms)
- Excludes trivial tasks that Verus can verify without annotations.
- *Comparison Baseline*: GPT-4o with sophisticated prompting (repeated invocations).

Table 1. Summary of Verus-Bench

Benchmark Sources	CloverBench	Diffy	MBPP	Misc	Total
# of Proof Tasks	11	38	78	23	150
Executable LOC	175	951	1,333	390	2,849
Specification LOC	80	265	700	207	1,252



Results

- **Success Rate:** AutoVerus = 91.3% vs. Baseline = 44.7%.
- **Efficiency:**
 - Solved > 50% of tasks within 30[sec] or 3 LLM calls.
 - Baseline required significantly more time and calls.
- **Cost:** Total cost for the entire benchmark was only $\leq \$37$.



Summar & Significance

Summary.

- *Synergy of Techniques:* Combines LLM creativity with formal rigor.
- *Workflow Mimicry:* Imitated the “Draft-Refine-Debug” loop used by human experts.
- *Overcoming Data Scarcity:* Demonstrated that In-Context Learning + Static Analysis can solve problems in few resources languages without fine-tuning.

Significance.

- *Democratizing Verification:* Lowers the barrier to entry for verifying Rust systems.
- *Practicality:* Works on standard hardware with low cost and high speed.



Future Work

Future Work.

1. Scaling to large-scale system verification with complex dependencies.
2. Automating specification generation.



Thank you for listening!

