

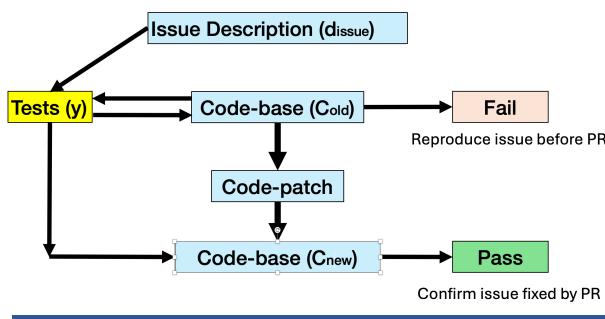
Otter: Generating Tests from Issues to Validate SWE Patches

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Problem Statement

Let d_{issue} = issue description, c_{old} = old code before PR, and c_{new} = new code after PR. The problem is to generate tests y given as input only $x = (d_{issue}, c_{old})$, without access to c_{new} .



Contributions

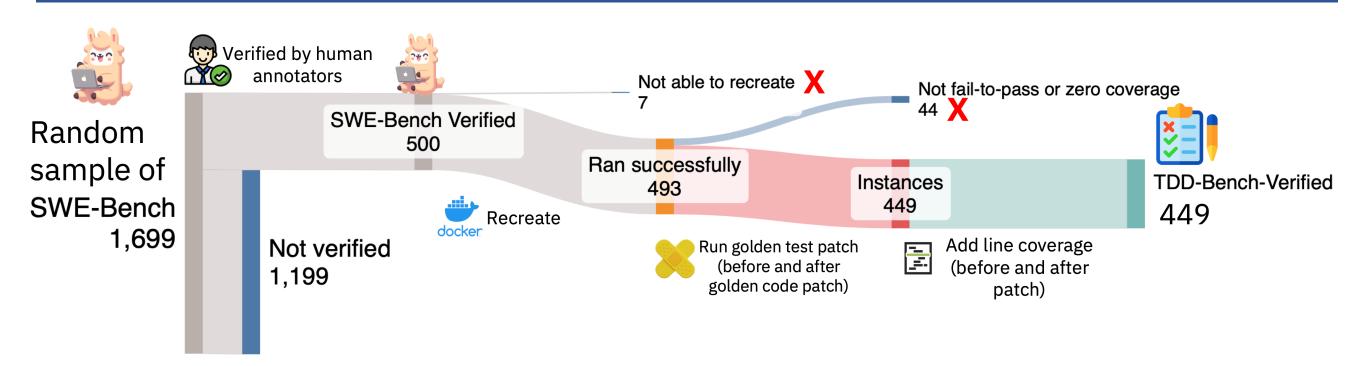
- Two bug reproduction test generation approaches: i) Otter ii) Otter++
- Benchmark to evaluate reproduction tests: TDD-Bench-Verified.

Motivation

- To systemically evaluate test generation tools (using Benchmark)
- Improve precision of SWE-agents by validating SWE-patches
- Support Test Driven Development (TDD)
 - make requirement more precise
 - easy to maintain codebase

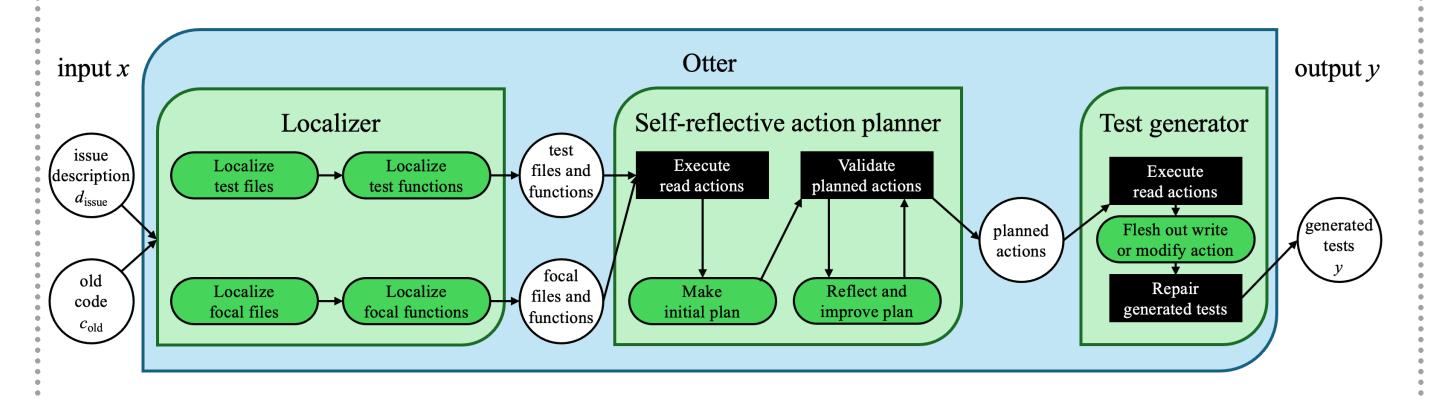
TDD-Bench-Verified: https://github.com/IBM/TDD-Bench-Verified

Benchmark Construction



- Started with 500 samples from SWE-Bench-Verified
- Ended up with 449 after all filtering process (based on coverage and f2p property)
- We propose a new metric tddScore (consider coverage also)

Otter: Overview



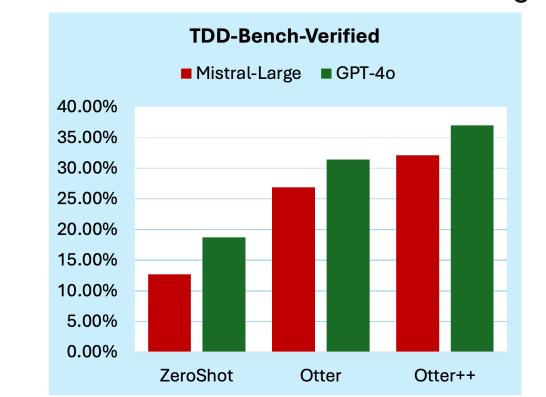
Legend: data LLM call Rule-based

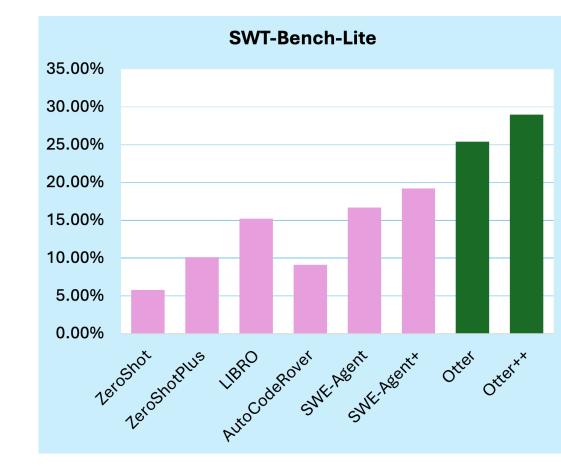
We have three components:

- i. Localizerii. Self-reflective action planner
- iii. Test generator

Performance Evaluation

Otter++ is Otter with inference scaling

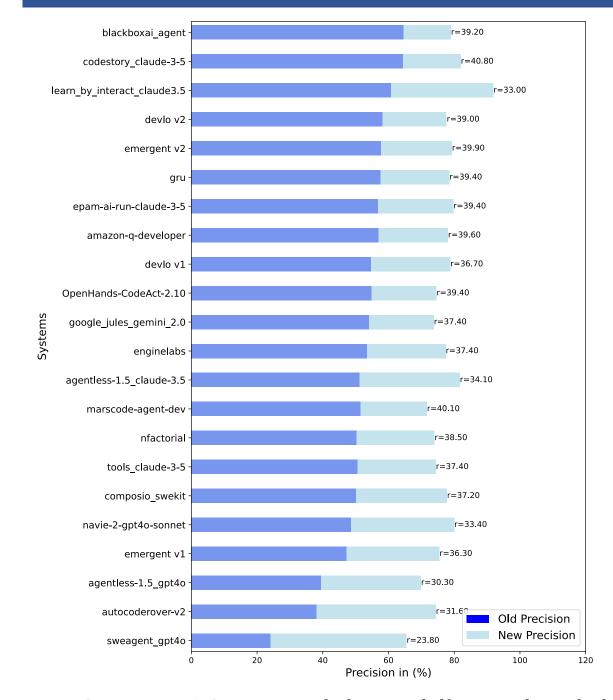




Key Insights

- Without self reflective action planning, we lose more than 14%-20% of f2p tests for GPT-4o and 21%-36% for Mistrallarge model (see details in the paper).
- For each instance, Otter costs \$0.06 and Otter++ costs \$0.09

Validating SWE-Patches



• 65% to 92% precision while maintaining a decent recall of 30%-41%.

Conclusion

- Proposed Otter, a system that generates tests from issues, using LLMs with a novel self-reflective action planner
- Open-sourced TDD-Bench-Verified, a benchmark for test driven development
- An empirical study on using tests generated from issues to filter SWE-Patches

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