Code Agents are State of The Art Software Testers

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We propose SWT-Bench, a new test generation benchmark based on issue descriptions and real world code bases. We show that LLM-based Code Agents outperform zero-shot baselines and prior specialized methods at this task.

SRILAB

logic*

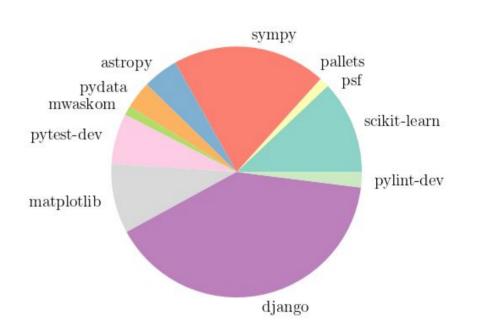
Benchmark Overview

Task inputs Pre PR **P**ost PR Generated Tests isValid currently allows Fail isValid("name\n") == False Pass trailing newlines but only (!)alphanumeric characters isValid("name") == True Pass Pass should be accepted. isValid("name") == False Fail Fail Codebase (Pre PR)

We derive a Software Testing dataset from SWE-Bench [1]. Target: predict a test patch that reproduces a reported user issue.

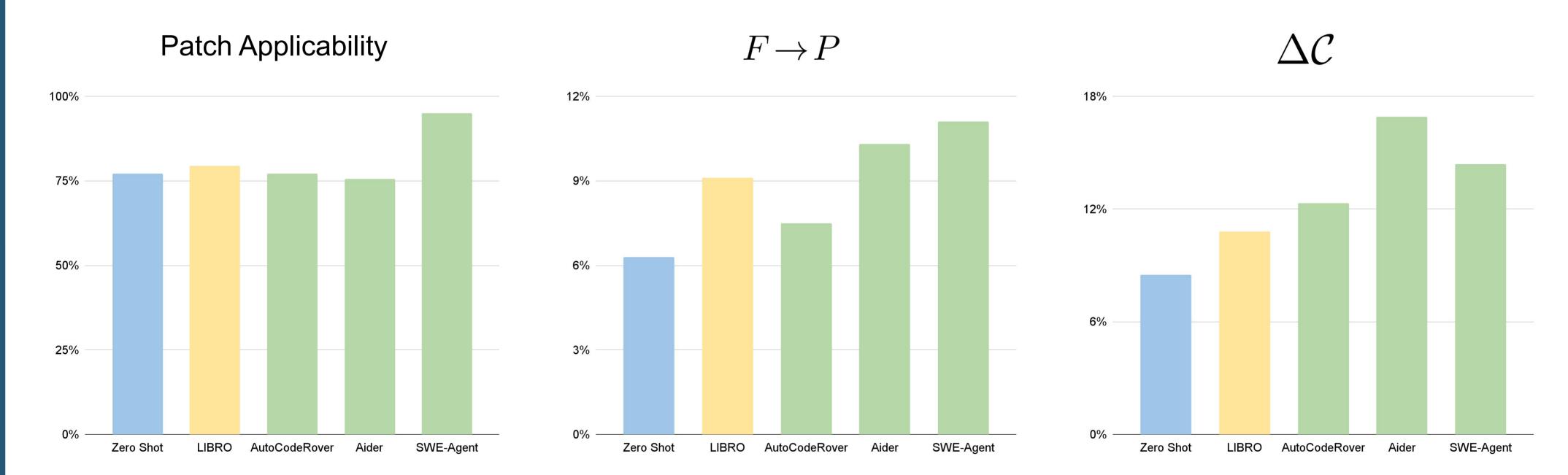
- Patch Applicability: Whether the prediction is a valid patch.
- Fail-to-Pass rate (F→P): Cases where i) at least one test fails before golden code patch and ii) all tests pass after.
- Coverage (ΔC): Line coverage of the golden code patch.

		Mean	Max
Issue Text	# Words	315.1	8756
Codebase	# Files	210.1	384
	# Lines	52330.8	122605
Existing Tests	$\# F \! \rightarrow \! P$	0.05	55
	$\# F \!\rightarrow\! F$	1.5	98
	$\# P \!\rightarrow\! P$	91.4	4837
	$\# P \!\rightarrow\! F$	0.3	40
	# total	105.1	4842
	Coverage	32.3%	67.7%
Golden Tests	$\# F \!\rightarrow\! P$	1.5	952
	$\# F \! \to \! F$	0.0	5
	$\# P \!\rightarrow\! P$	1.6	766
	$\# P \!\rightarrow\! F$	0.0	0
	# added	2.8	750
	# removed	0.3	104



E *H* zürich

Experimental Results



Agentic approaches perform as well or outperform zero-shot baseline and specialized previous methods



% correct patches of proposed patches

All experiments based on GPT-4. Zero Shot Baseline: GPT-4 + BM25 retrieved files, Previous Method: LIBRO [2], heuristic filtering of proposed tests. Code Agents: Aider [3], AutoCodeRover [4] and SWE-Agent [5] with modified prompts (i.e. "generate a test case", "run the test suite before submission")

[1]: Jimenez et al: Can language models resolve real-world GitHub issues?, 2024 [2] Kang et. al: Large language models are few-shot testers, 2023
[3] https://aider.chat [4] Zhang et al: AutoCodeRover, 2024 [5] Yang et al: SWE-Agent, 2024