Large Language Models of Code Fail at Completing Code with Potential Bugs

Tuan Dinh, Jinman Zhao, Samson Tan, Renato Negrinho, Leonard Lausen, Sheng Zha, and George Karypis

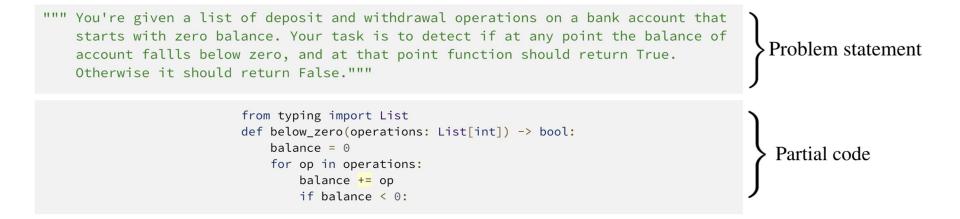
University of Wisconsin-Madison

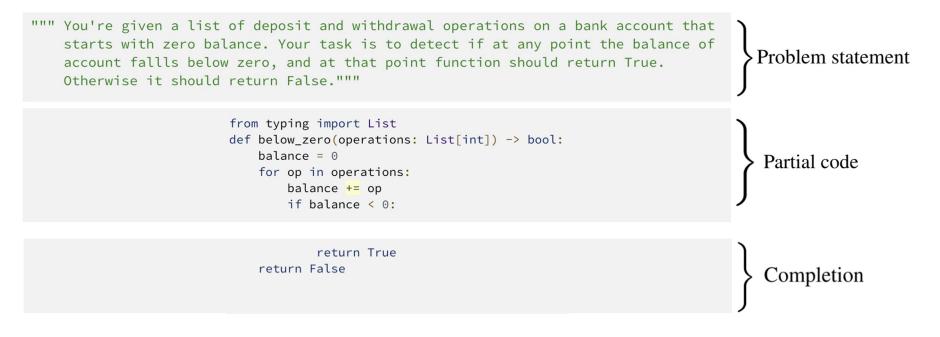
AWS AI Research and Education



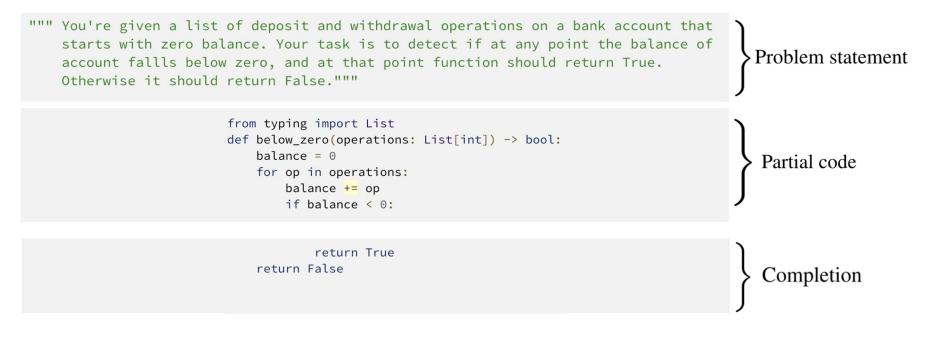








Code-LLMs achieve > 50% pass rate on various benchmarks



Code-LLMs achieve > 50% pass rate on various benchmarks

Existing completion models assume error-free inputs ...

return Tru False

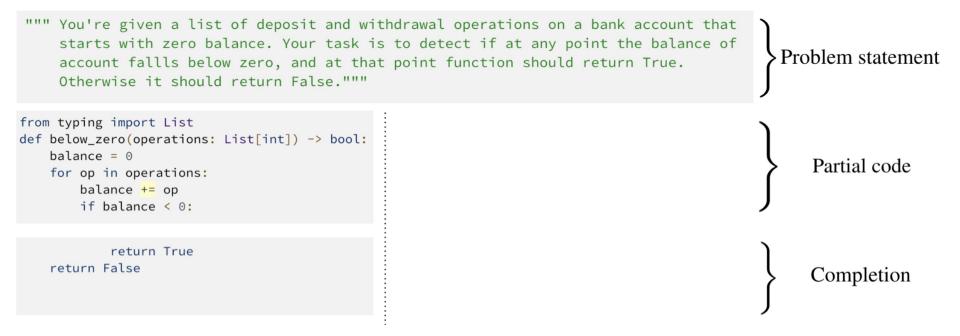
Completion

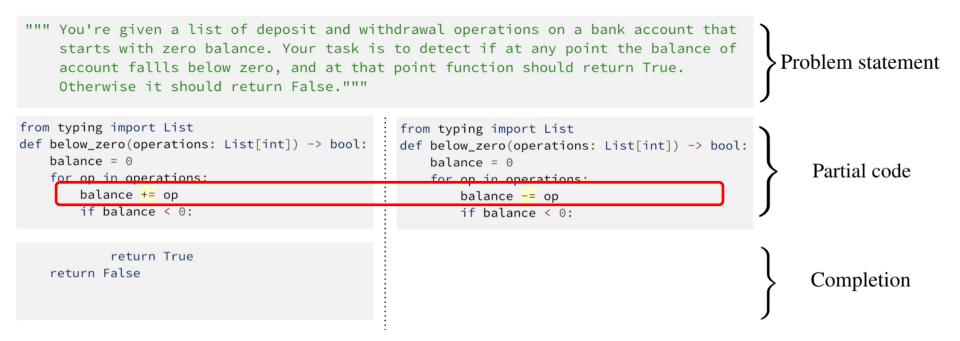
Code-LLMs achieve > 50% pass rate on various benchmarks

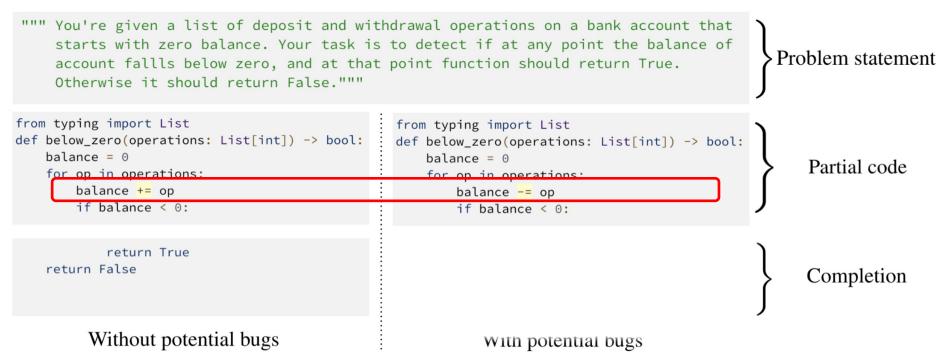
Existing completion models assume error-free inputs ...

Bugs in code are inevitable!

(esp for the in-progress partial code)

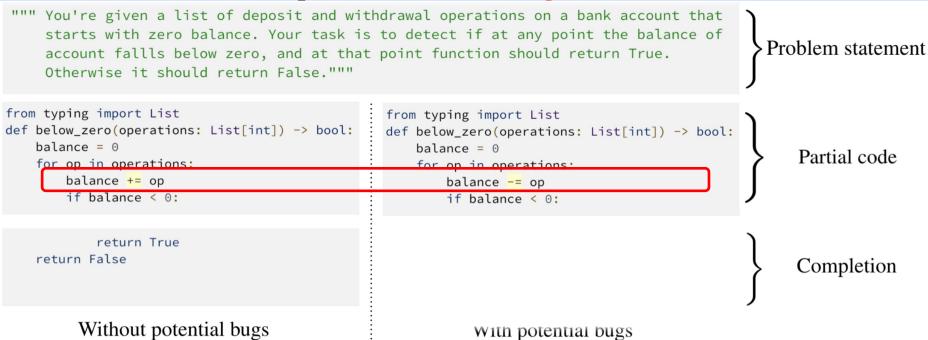






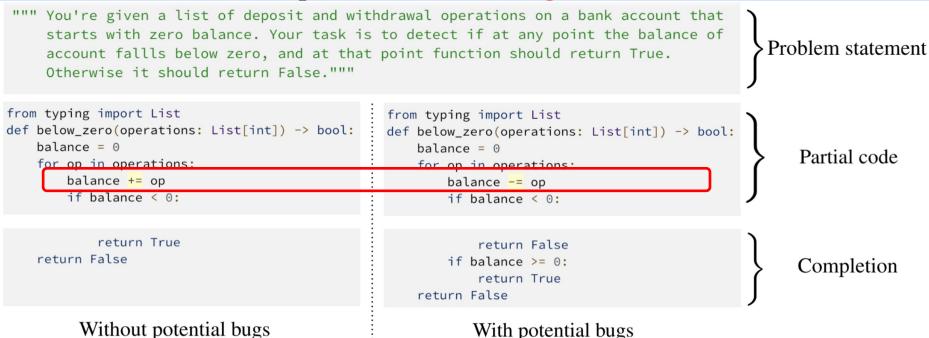
A change from $+= \rightarrow -=$ results in a **potential bug** \rightarrow partial code + original completion fails test: below_zero(1, 2) == False

\rightarrow the code completion should change



A change from $+= \rightarrow -=$ results in a **potential bug** \rightarrow partial code + original completion fails test: below_zero(1, 2) == False

\rightarrow the code completion should change



New Benchmarks for Buggy Code Completion

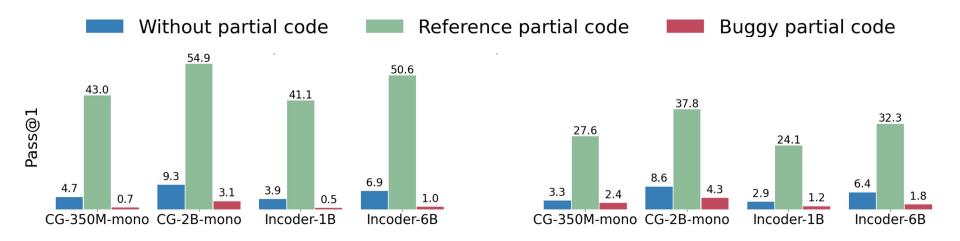
Buggy-HumanEval

- Artificial potential bugs
- Constructed from HumanEval

Buggy-FixEval

- Potential bugs procured from user submissions to coding problems
- Constructed from FixEval and CodeNet

Analysis 1: Failures on Buggy-Code Completion



Performance drops hugely when potential bugs are present!

[CTJ+'21] Chen et al., 2021. Evaluating large language models trained on code. [HALB'22] Haque et al., 2022. Fixeval: Execution-based evaluation of program fixes for competitive programming problems.

5

Mitigation Methods for Completion

Strategy 1: **Removal** \rightarrow **Completion**

Idea: remove the partial code to guarantee no potential bug

exists! Strategy 2: Completion \rightarrow Rewriting

Idea: treat the completion as buggy and attempt to fix.

Strategy 3: **Rewriting** → **Completion**

Idea: locate and rewrite potential bugs before being completed

[FAL+'22] Fried et al., 2022. Incoder: A generative model for code infilling and synthesis.

Mitigation Methods for Completion

Strategy 1: **Removal** \rightarrow **Completion**

Idea: remove the partial code to guarantee no potential bug

$\begin{array}{l} \substack{\text{exists!}\\ \text{Strategy 2: Completion} \rightarrow \text{Rewriting} \end{array}$

Idea: treat the completion as buggy and attempt to fix.

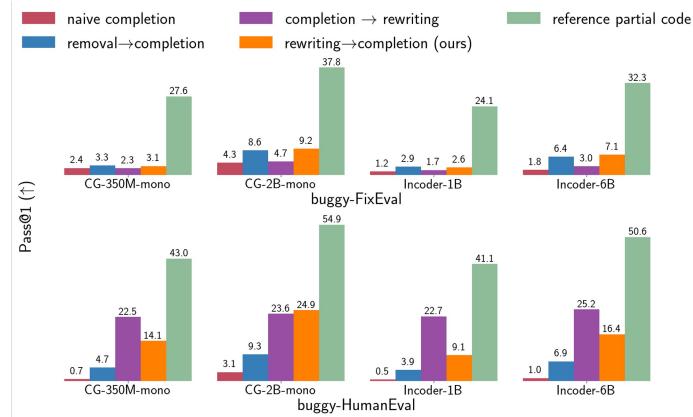
Strategy 3: **Rewriting** → **Completion**

Idea: locate and rewrite potential bugs before being completed Consider potential bug as distributional outliers using an infilling model [FAL+'22]

[FAL+'22] Fried et al., 2022. Incoder: A generative model for code infilling and synthesis.

Analysis 2: Mitigation Methods for Completion

Methods improve the completion, but remain huge gap to the inference



7

Ablation and Case Studies

When do Code-LLMs surpass?

60%: fails to react

Check if in given list of numbers, are any two numbers closer to each other than given threshold.

<pre>from typing import List def has_close_elements(numbers: List[float],</pre>	÷	<pre>from typing import List def has_close_elements(numbers: L</pre>
distance = abs(elem - elem2)	į	continue

if distance < threshold: return True

return False

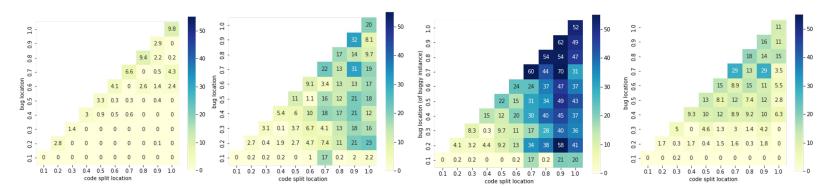
Without potential bugs

List[float], loat) -> bool: pers): rate(numbers):

if abs(elem - elem2) < threshold: return True return False

With potential bugs

Bug and split location can affect the performance



Thank you!

Contact information: Tuan Dinh: <u>tuan.dinh@ucsf.edu</u> Jinman Zhao: <u>jinmaz@amazon.com</u>

> Poster: #539 (Wed 13 Dec 5 -- 7 p.m. CST) Paper: <u>https://neurips.cc/virtual/2023/poster/70988</u> Github: <u>https://github.com/amazon-science/buggy-code-completion</u>