

Reducing Large Language Model Hallucinations for Test Generation

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1. INTRODUCTION

UTGen combines a search-based software testing[1] tool, EvoSuite[2], with LLMs to improve the understandability of the generated tests.

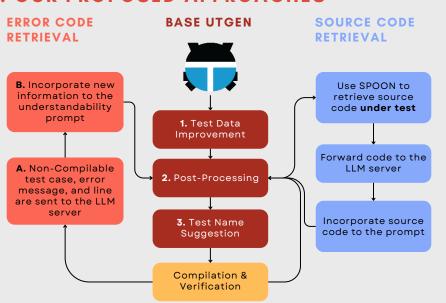
The LLM may "hallucinate" and generate test code that is not compilable or code that is too dissimilar to the initial EvoSuite test. UTGen handles this by re-prompting, making the process inefficient and expensive.

The purpose of our research is to decrease the number of reprompts, by engineering an enhanced prompt for the LLM using information retrieval.

2. RESEARCH QUESTIONS

- 1. **RQ1:** Is it possible to reduce the hallucination of LLMs used in UTGen during the Post-Processing phase, to minimise the need for re-prompting, using prompt engineering with *source code retrieval*?
- 2. **RQ2:** Is it possible to reduce the hallucination of LLMs used in UTGen during the Post-Processing phase, to minimise the need for re-prompting, using prompt engineering with *error code retrieval*?

3. OUR PROPOSED APPROACHES

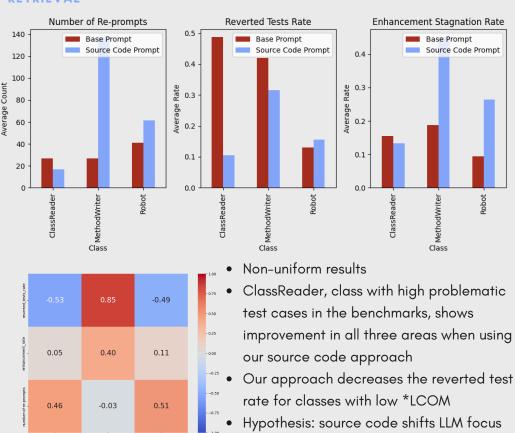


4. METHODOLOGY

- Comparison Study: hallucination performance on same EvoSuite test suite between our respective methods and base UTGen
- <u>Metrics:</u> Number of re-prompts, Enhancement Stagnation rate (LLM did not generate test similar to EvoSuite), Reverted tests rate (LLM did not generate compilable test), Source code complexity.
- <u>Dataset:</u> 4 available benchmarks of UTGen on 204 classes of SF110 of DynaMOSA[4] dataset. From those we select 3 classes according to benchmark enhancement stagnation and reverted test performance.
- <u>LLM:</u> code-llama:7b-instruct model, as provided by Hugging Face

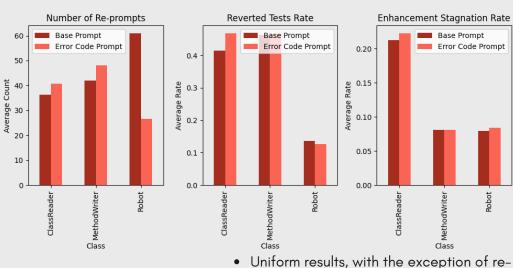
5. RESULTS & CONCLUSIONS

SOURCE CODE RETRIEVAL



to compilability

ERROR CODE RETRIEVAL



- prompt count drop for Robot (high *LCOM)

 Mirrored correlation between reverted test
 - rate and number of re-prompts
 Expected no correlation with enhancement stagnation rate, since the prompt is not
 - Hypothesis: High prompt complexity does not allow for improvement

used between similarity evaluations

6. LIMITATIONS

- Lack of Documentation on UTGen Implementation
- Limited technical & financial resources for utilizing the LLM
- Measuring the Number of Re-Prompts effectively

7. FUTURE WORK

-0.09

-0.14

-0.11

- Investigate conciser and more focused prompts to ground the LLM further
- Upgrade to LLM with more training parameters for higher accuracy
- Investigate similarity between EvoSuite and error code approach test suites

8. REFERENCES

[1] Shaukat Ali, Lionel C. Briand, Hadi Hemmati, and Ra- jwinder Kaur Panesar-Walawege. A Systematic Re- view of the Application and Empirical Investigation of Search-Based Test Case Generation. IEEE Transac- tions on Software Engineering, 36(6):742-762, Novem- ber 2010.
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