Tech Debt Transformer Pipeline

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Recap: Transformers!

A good way to capture a lot of information!

Classification

Generation

Recap: Technical Debt

Definition: "Implied cost of future reworking required when choosing an easy but limited solution instead of a better approach that could take more time" Wikipedia.

```
₩ README.md × | # highlight.lua i4 ×
     local Config = require("todo-comments.config")
              nmmm, this looks a bit funky
       for kw in pairs(Config.keywords) do
         local start, finish = str:find("(" .. kw .. "):")
          return start, finish, kw
        main 04 highlight.lua
                                                                                                               lua Top @ 15:51

✓ of lua/todo-comments/config.lua 1

       TODO: add support for markdown todos
        TODO: do something
       FIX: this should be fixed
       HACK: weird code warning

✓ d lua/todo-comments/highlight.lua 9

    PERF: fully optimised
       HACK: hmmm, this looks a bit funky
seratchpad nvim ~/p/todo-comments.nvim
```

Existing Solutions, Their Limitations, and Novelty

- Existing solutions usually utilize rules to find technical debt (more have started to use machine learning recently)
- Only one transformer based approach: finding technical debt in issue trackers. (Skryseth, Daniel, et al.)

 This study will be focused on directly looking at code from git commits to get insights into technical debt!

Problem Statement

In open source projects, managing technical debt can be a challenging process. This study explores how fine-tuned transformer models can help improve the identification and classification of these issues.

Research Questions

- 1. How well can a fine-tuned transformer model classify technical debt, given code from a commit?
- 2. How well can a fine-tuned transformer model generate labels for technical debt?

Proposed Architecture



The Technical Debt Dataset

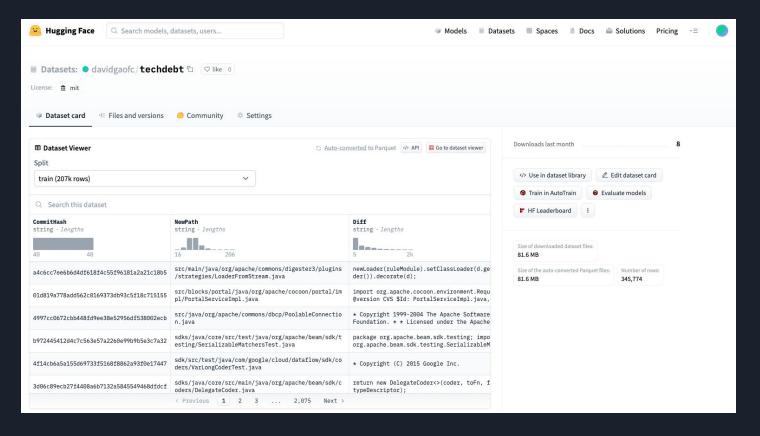
- 78K commits from 33 Java Projects
- SZZ algorithm annotations
 - find the fix, trace backwards
- SonarQube
 - static code analysis, rule based

Classification Task

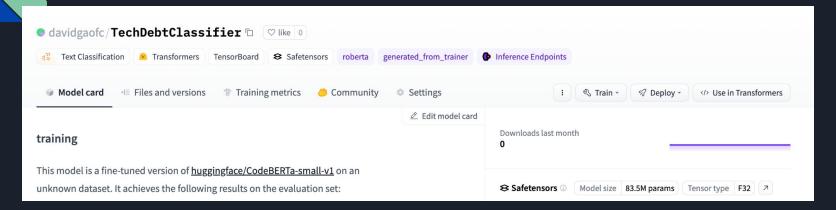
How well can a fine-tuned transformer model classify technical debt, given code from a commit?

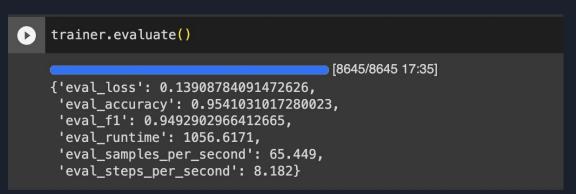
Metrics: Accuracy, F1 Score

Preliminary Work



Preliminary Work (continued)





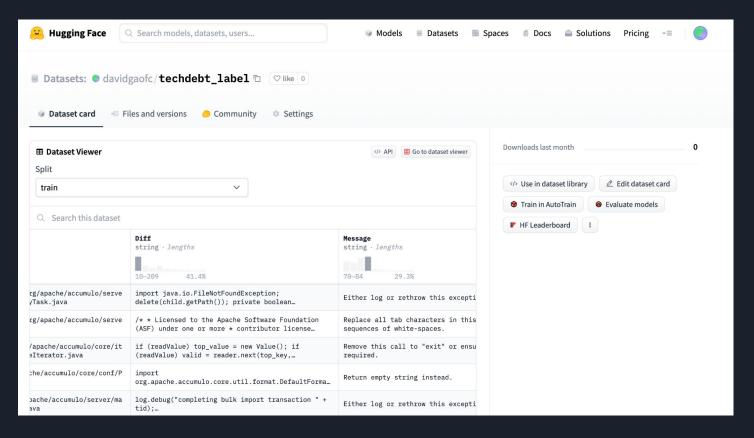
F1 Score =
$$\frac{2}{\frac{1}{\frac{1}{\text{Precision}} + \frac{1}{\text{Recall}}}}$$
$$= \frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$$

Generation Task

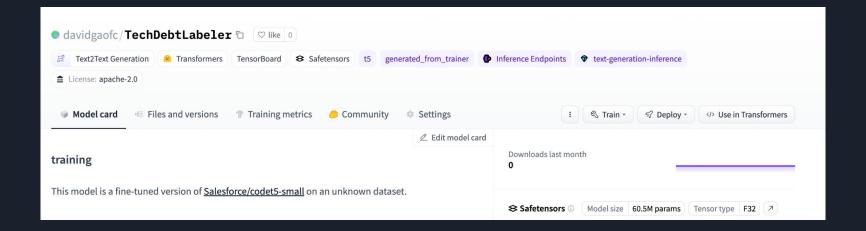
How well can a fine-tuned transformer model generate labels for technical debt?

Metrics: BLEU, ROUGE

Preliminary Work



Preliminary Work (continued)



Pipeline



```
output

[[{'label': 'LABEL_1', 'score': 0.9938503503799438}, {'label': 'LABEL_0', 'score': 0.00614961888641119}]]
```



```
output

[{'generated_text': "Rename this local variable to match the regular expression '^[a-z][a-zA"}]
```

Questions/Comments?

References

Valentina Lenarduzzi, Nyyti Saarimäki, and Davide Taibi. 2019. The Technical Debt Dataset. In Proceedings of the Fifteenth International Conference on Predictive Models and Data Analytics in Software Engineering (PROMISE'19). Association for Computing Machinery, New York, NY, USA, 2–11. https://doi.org/10.1145/3345629.3345630