Mining Mutations from Public Python Projects

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Problems

- 1. Mutation testing relies heavily on the quality of mutation operators
- 2. Common mutation operators may not be representative of actual bugs
- 3. It's hard to come up with enough mutation operators manually

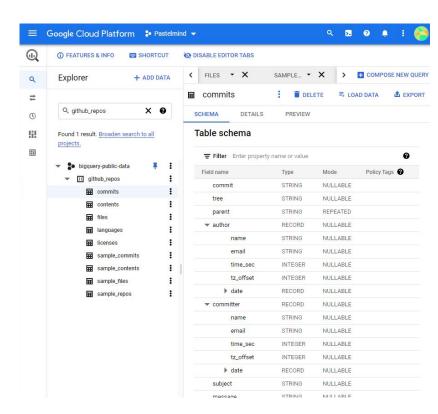
Solution: Automated mutation mining from real-world bugs

Collect commits with Google BigQuery

bigquery-public-data.github_repos.commits

- A public dataset mined from open-source GitHub repositories
- 870GB table containing 250 million commits

We needed **bug fixes** for **Python** code



Collect commits with Google BigQuery

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Observation:

- By convention, developers use certain words in commit messages that fix bugs
- e.g. "This fixes #23"

We selected commits that...

- Change > 1 Python file
- The message contains a keyword often used when fixing a commit (bug, fix, issue, error)

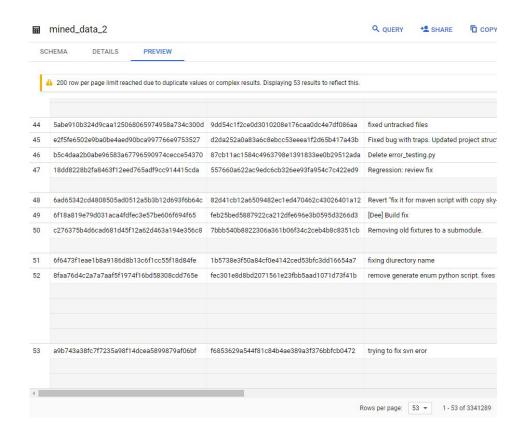


```
SELECT commit, subject, repo_name, ARRAY(
    SELECT AS STRUCT *
    FROM UNNEST(difference)
    WHERE (new_path LIKE "%.py")
) AS difference
FROM `bigquery-public-data.github_repos.sample_commits`
WHERE EXISTS (
    SELECT new_path FROM UNNEST(difference)
    WHERE (new_path LIKE "%.py")
)
AND regexp_contains(subject, 'bug|fix|issue|error')
```

Collect commits with Google BigQuery

End result: 3.3 million commits with

- Repository name
- SHA-1 hash of the commit
- SHA-1 hash of the parent commit
- Paths of all Python files changed



```
1 class GithubFilesSpider(scrapy.Spider):
       name = f"github_files_chunk{CHUNK NUMBER}"
        allowed_domains = ["raw.githubusercontent.com"]
       CHUNK_FILE_NAME = CHUNK_FILE_NAME
       OUTPUT FILE PATH = OUTPUT FILE PATH
       def start requests(self):
           # a list. This is because the chunk files eat up a large amount of
           # memory (~300 MB observed).
           ROW_COUNT = sum(1 for _ in load_gzipped_lines(file_name=self.CHUNK_FILE_NAME))
           # For resuming interrupted crawling sessions
           downloaded changes = load downloaded changes(file path=self.OUTPUT FILE PATH)
            logging.info(
               f"Found {len(downloaded_changes)} change entries that were downloaded in a previo
            skipped_file_count = 0
            for row index, row in enumerate(
                load_gzipped_lines(file_name=self.CHUNK_FILE_NAME)
                file change entries = list(
                    generate file change data(row=row, row_index=row_index)
               for fc_num, fc_data in enumerate(file_change_entries, start=1):
                   if is_already_downloaded(downloaded_changes, fc_data):
                       logging.debug(
                            f"Skipping already downloaded commit {row index + 1} / {ROW COUNT}, f
   ile {fc_num} / {len(file_change_entries)}"
                       skipped file count += 1
                       if skipped file count % 1000 == 0:
                           logging.info(
                                f"Skipped {skipped file count} files that were already downloade
```

Crawling GitHub with Scrapy

Use Scrapy to download commits directly from GitHub

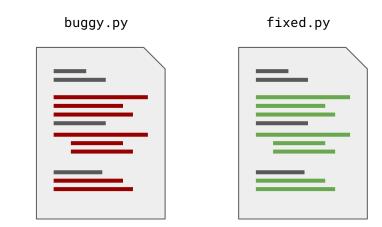
Downloaded 9.3 million Python files (200GB) over 3 weeks

(GitHub limits server hits to 5000/hour 😡)

https://scrapv.org/

Preprocessing: Pairing

A file may contain changed & unchanged code regions

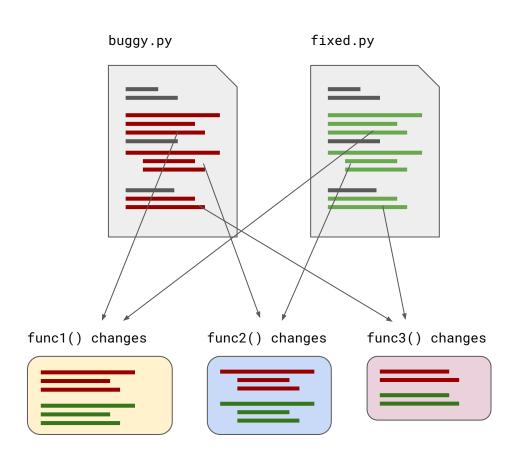


Preprocessing: Pairing

A file may contain changed & unchanged code regions

⇒ Extract function-level changes

- Pick functions and methods
- Find matching pairs of functions by name
 - Python functions are rarely overloaded, so function names are usually unique
- Ignore function pairs that are unchanged



Preprocessing: Extraction

How to describe changes in code?

⇒ GumTreeDiff, an AST comparison tool that generates a sequence of edit actions

```
-self.discovered_filesystems = []
+self.discovered_filesystems = set()

"insert-node",
"insert-node",
"insert-node",
"delete-node"
]
```

https://github.com/GumTreeDiff/gumtree

Preprocessing: Normalization

Python code contains many distinct values and identifiers

- Large "vocabulary" of words, hard to generalize
- ⇒ Replace variables and constants with placeholders
- ⇒ Preserve frequently used "idioms": x, i, print, ...

```
def add_details(self, info):
   for (k, v) in info.items():
        try:
        setattr(self, k, v)
        except AttributeError:
        pass
```

```
def IDENTIFIER_0(self, IDENTIFIER_1):
    for (k, v) in IDENTIFIER_1.IDENTIFIER_2():
        try:
        IDENTIFIER_3(self, k, v)
        except IDENTIFIER_4:
        pass
```

Preprocessing: Filtering

Many code changes are too large for analysis

- ⇒ Select functions that have ≤ 50 lexer tokens per function
 - We wanted to train an RNN model with seq2seq, which accepts 50 "words" per sentence
 - Unfortunately, we didn't have enough time to actually use RNN...

⇒ Processed and filtered 475,961 function pairs

Data format

1. data.jsonl

Fields of each dictionary:

- name
- before_code: source code before bugfix
- after_code: source code after bugfix
- before_code_normalized,
 after_code_normalized:
 Abstracted & normalized code
- edit_actions: List of AST edit actions
- replacement_map

```
"name": "intel-hpdd/intel-manager-for-lustre:ff337f5ac809a9dbf2cefb988e595fad696ac
    f8a:chroma-manager/chroma_core/lib/detection.py:DetectScan.__init__",
      "before code": "def init (self, step):\n self.created filesystems = []...",
      "after_code": "def __init__(self, step):\n self.created_filesystems = []...",
      "before code normalized": "def init ( self , IDENTIFIER 0 ) : ...",
      "after code normalized": "def __init__ ( self , IDENTIFIER_0 ) : ...",
      "edit actions": [
        "insert-node",
        "insert-node",
        "insert-node",
        "delete-node"
      "replacement_map": {
        "identifiers": {
          "IDENTIFIER_0": "step",
          "IDENTIFIER 1": "created filesystems",
          "IDENTIFIER 2": "discovered filesystems",
          "IDENTIFIER 3": "created mgts"
        "floats": {},
        "ints": {},
        "strings": {},
        "f strings": {}
25 }
```

Data format

2. buggy.txt

Each line is abstracted buggy code

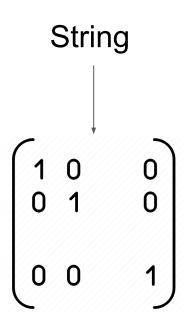
3. fixed.txt

Each line is abstracted fixed code

(whitespace replaced with special tokens)

Vectorization / Embedding

- Edit actions between buggy and fixed code are tokens like [delete-node], [move-tree], etc.
- Train Gensim's Doc2Vec model on sequence of edit actions.
- 3. Doc2Vec produces vectors of fixed size such that similar sequences of edit actions can be calculated to be close (e.g. using cosine distance)...hopefully



Clustering algorithms

Several options:

K-Means

Gaussian Mixture

Why?

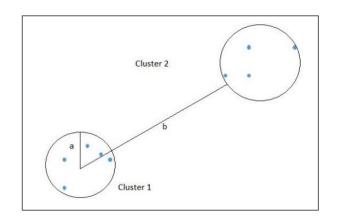
A simple and powerful classic go-to

Why?

As opposed to K-Means, utilizes the variance of the data and not just the mean.

Cluster evaluation (Silhouette score)

- A metric used to calculate the goodness of a clustering technique. Its value ranges from -1 to 1
- 1 means clusters are well apart from each other and clearly distinguished.
- 0 means clusters are indifferent, or in other word, the distance between clusters is **not significant**
- -1 means clusters are assigned in the wrong way



Silhouette Score =
(b-a)/max(a,b)

a= average intra-cluster distance
b= average inter-cluster distance

Cluster evaluation (Silhouette score)

For vector size = 5:

Number of clusters	Silhouette score
2	0.36445
3	0.42311
4	0.38316
5	0.34782
6	0.35342

For vector size = 20:

Number of clusters	Silhouette score
2	0.35168
3	0.41482
4	0.39859
5	0.37131
6	0.34029

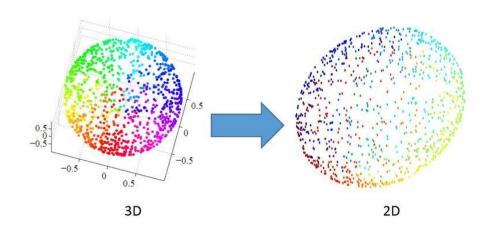
For vector size = 40:

Number of clusters	Silhouette score
2	0.35446
3	0.41928
4	0.40131
5	0.37247
6	0.35874

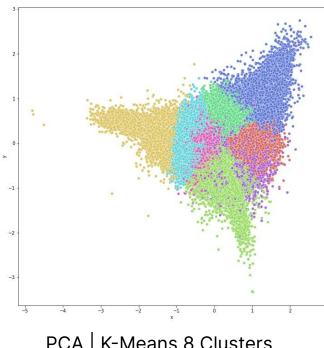
Clustering visualized

Data is not in 2D! How do we visualize?

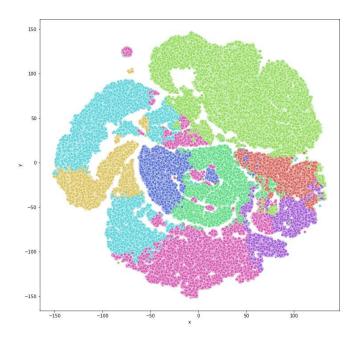
Use t-SNE and PCA algorithms for **dimensionality reduction**.



t-SNE + PCA **PCA**

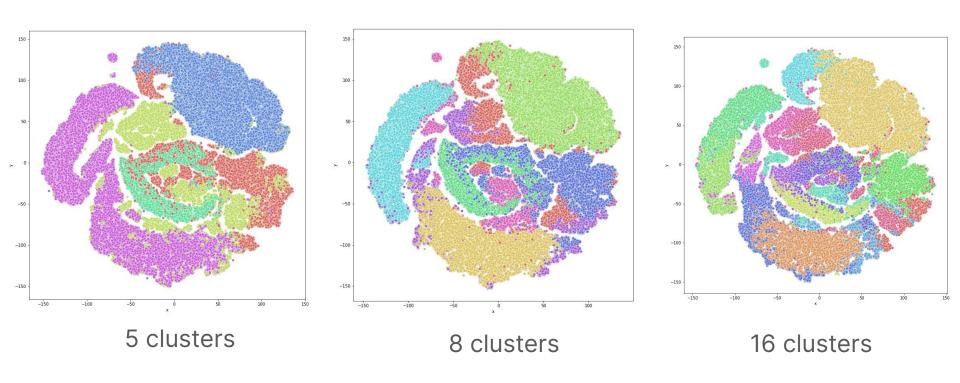


PCA | K-Means 8 Clusters

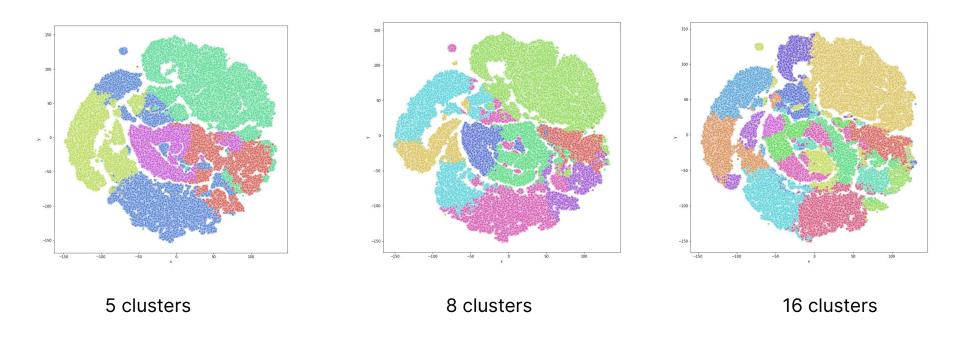


PCA + t-SNE | K-Means 8 Clusters

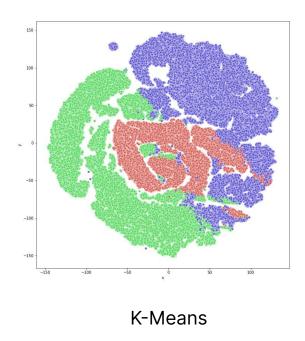
Clusters | Gaussian Mixture Model



Clusters | K-Means



K = 3, Best Silhouette score



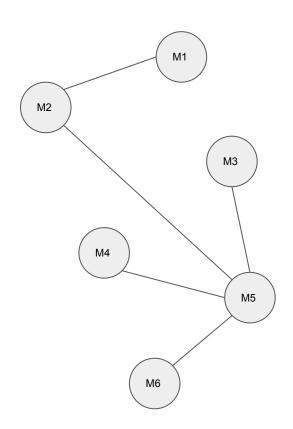
Gaussian Mixture

Sampling algorithm

For each cluster, build a graph. Each "mutation" is a node.

How do we add edges?

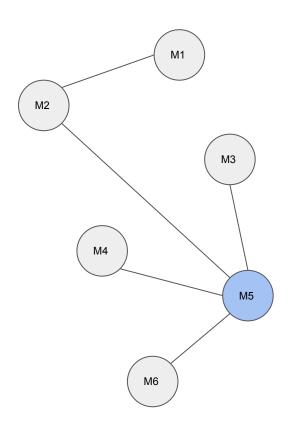
- We calculate the cosine distance between all pairs of points in the set of embeddings
- 2. For pairs with cosine distance larger than a certain threshold, the edge is added



Sampling algorithm (cont.)

Two ways to rank nodes:

- 1. Choose one with the largest degree (no. of edges)
- 2. Use Page Rank algorithm



Samples | ["update-node"]

```
def IDENTIFIER_0 ( self ) :
    self. IDENTIFIER_1( )
    IDENTIFIER_3 . IDENTIFIER_2 ( STR_0 )
    print ( IDENTIFIER_4 ( IDENTIFIER_6 . IDENTIFIER_5 () ) )

def IDENTIFIER_0 ( self ) :
    self . IDENTIFIER_7 ( )
    IDENTIFIER_3 . IDENTIFIER_2 ( STR_0 )
    print ( IDENTIFIER_4 ( IDENTIFIER_6 . IDENTIFIER_5 () ) )
AFTER
```

Samples | ["insert-node", "delete-node"]

Samples ["insert-node", "insert-node", "move-tree"]

```
def IDENTIFIER_0 ( self , IDENTIFIER_1 ) :
   x = IDENTIFIER_1 . IDENTIFIER_2 ( int , str )
                                                      BEFORE
   if IDENTIFIER_3 ( x ) is int :
      x = IDENTIFIER_4 [x]
   IDENTIFIER_1 . IDENTIFIER_5 ( x )
def IDENTIFIER_0 ( self , IDENTIFIER_1 ) :
   x = IDENTIFIER_1 . IDENTIFIER_2 ( int , str )
                                                      AFTER
   if IDENTIFIER_3 ( x ) is int :
      x = IDENTIFIER_1 . IDENTIFIER_4 | x |
   IDENTIFIER_1 . IDENTIFIER_5 ( x )
```

Samples | ["delete-node"]

```
def IDENTIFIER_0 ( self , name , value ) :
                                                                   BEFORE
    if value == None :
       value = ''
    value = str ( value )
    self . IDENTIFIER_1 . append ( ( name , value ) )
    return
def IDENTIFIER_0 ( self , name , value ) :
    if value is None :
       value = ''
                                                                   AFTER
    value = str ( value )
    self . IDENTIFIER_1 . append ( ( name , value ) )
    return
```

Current Difficulties or Limitations

- Hard to automate evaluation of whether these clustering and sampling techniques produce useful mutations
- 2. Hard to evaluate usefulness of mutations (could be subjective)
- 3. Hyperparameters for dimensionality reduction, clustering, sampling and embedding model are not optimized

Possible Improvements

- Utilize GumTree better
- 2. Utilize normalized before and after code. (Utilize cosine distance between the embedding of before and the embedding of the after code)
- 3. Tune hyperparameters
- 4. Train an ML model that given fixed code as input, the model can learn to generate buggy code as output.

Conclusion

- 1. The technique is valid and has potential
- 2. With the aforementioned improvements, we're confident that a powerful neural mutation testing tool is possible
- 3. It will allow for world domination:)

Summary

- Problem: Mutation testing tools offer a very limited set of mutation operators
- 2. **Approach**: Mine bug-fixes from public GitHub commits, group them into semantically close groups and choose the best samples from each group
- 3. **Results**: Valid mutations could be found, although many of them are not too different from regular mutation operators. The ones that are different may or may not be hard to implement.

QUESTIONS?