

PREDICTING TEST CASE VERDICT USING TEXTUAL ANALYSIS OF COMMITTED CODE CHURNS

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About me

- Personal profile
 - □ Khaled Al-Sabbagh
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- Academic background

MSc degrees Management
 MSc degree in Software Engineering
 BSc Information Technology Engineering

Current work

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Context: testing in continuous integration

- Continuous integration often includes (regression) testing after every build
 - Frequent (every 10 minutes) integrations results in large number of test executions
 - Regression suites need to be small in order to reduce the cost of testing
- Continuous testing is often organized in several suites
 - Minimal suite after every build
 - Larger daily suite
 - Even larger weekend suite
- Developers need feedback about their code (from testing) as soon as possible
 - We should strive to execute the test that have the highest probability of failure as quickly as possible after code commit



Goals for this research

- To reduce the time for testing?
 - Reduce the time for test execution to shorten the feedback loop
 - Reduce the risk of re-introducing new defects when fixing the existing ones
- To increase the rate of fail/executed test cases?
 - Reduce the number of test cases that are executed and do not trigger any failures











Problem formulation:

- How to predict which test case would fail for a given line of code?
- How can we predict whether a given test case will fail/pass for a given line of code?
- How do we optimize the "limited test scope" for <u>each</u> build, so that no unknown errors are found when we run "full test scope"?











Code Churn Builds Main branch #bb3ed #53ada.... #37baa....

The amount of changes made to software between two points in time is referred to as code churn.

//pointer declaration. Int *p;	//pointer declaration. Int *p;
 int age[100]	 int age[100]
 char vowels[][5] = { {'A', 'E', 'l', 'O', 'U'}, {'a', 'e', 'i', 'o', 'u'} };	 char vowels[][5] = { {'A', 'E', 'I', 'O', 'U'}, {'a', 'e', 'i', 'o', 'u'} };
	//array declarations int person[100]
	person[0]= p
	······





Method using Bag of Words for Test Selection (MeBoTS)







Step 1: (Data Extraction)





Step 2: (Features Extraction)



csv file





Step 3: (Classification)



GOT



Evaluation – Case and Dataset

- Company: Software Telecommunication in Sweden
- Dataset: 12 test cases, 82 executions
- Original Dataset:
 - Mix of small and large churns
 - 1.4m lines of code, 500 features
- Curated Dataset:
 - <120k lines of code per churn</p>
 - 290k lines of code, 500 features







Evaluation - Metrics

• Precision: how many test cases identified as passing will pass?

 $precision = \frac{|TruePositive|}{|TruePositive| + |FalsePositive|}$

 Recall: How many test cases passing, will be identified as such?

$$recall = \frac{|TruePositive|}{|TruePositive| + |FalseNegative|}$$

- Goal:
 - High recall to identify many test cases that need no execution
 - High precision to be sure about them





Evaluation - Results

• Before data curation

• After data curation

Result:

- Medium recall: we already identify many test cases that need no execution
- High precision: we are sure about them in >7 of 10 cases







Threats to validity and mitigation

- Small sample size of test executions (7 test cases).
- Test failures may be caused by an environment upgrade or defect in the test scripts.
- Non-deterministic behavior of test cases (flaky tests)
- Different architecture and configuration of the networks' hyperparameters may result in higher prediction performance.





Conclusion and future work

- More data to evaluate the effectiveness of MeBoTS in practice.
- The prediction performance showed a precision rate of 73% and a medium recall.
- Using the method with small code churns showed an overall improvement in precision and recall.
- Evaluate other textual analysis techniques for better prediction.
- Evaluate the method on different software systems and contexts.
- Evaluate the trained model on code changes from outside the extracted sample.
- Measure the required time to retrain the model for better accuracy.

