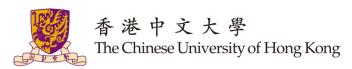


A survey of root cause analysis

霍茵桐 2021/06/25 Workshop







Problem Formulation

Target to identify the underlying causes leading to a failure that has affected end users, which is often closely related to the failure diagnosis.

In the broad domain of root cause analysis, log-based failure diagnosis is now a standard practice.

CHALLENGES

- The complexity of modern software systems grows rapidly, making too complex to efficiently disclose relations of fault, failure and human-observed symptoms.
- As software systems become more mature, failures are becoming more and more hard to detect and diagnose.

In the very beginning...



Traditional rule-based methods are adopted.

 which heavily rely on a set of predefined rules (if-then logic) from the expert knowledge to diagnose failures.



Limitations:

Rule-based methods cannot be well generalized to unseen failures that are not included in the rules.

Outline

#	1: Execution Replay Methods
	#2: Model-based Methods
#	3: Statistics-based Methods
#	4: Retrieval-based Methods

Outline

• Infer execution flow ****** #1: Execution Replay Methods • Trace back to software failure

Execution Replay Methods

Approach

Automatically infer the execution flow from logs and trace back to the software system failure.

Input:

- Log recorded during the failed execution
- Source code of the target program

Output:

Derive the potential code paths and execution path

Execution Replay Methods (1)

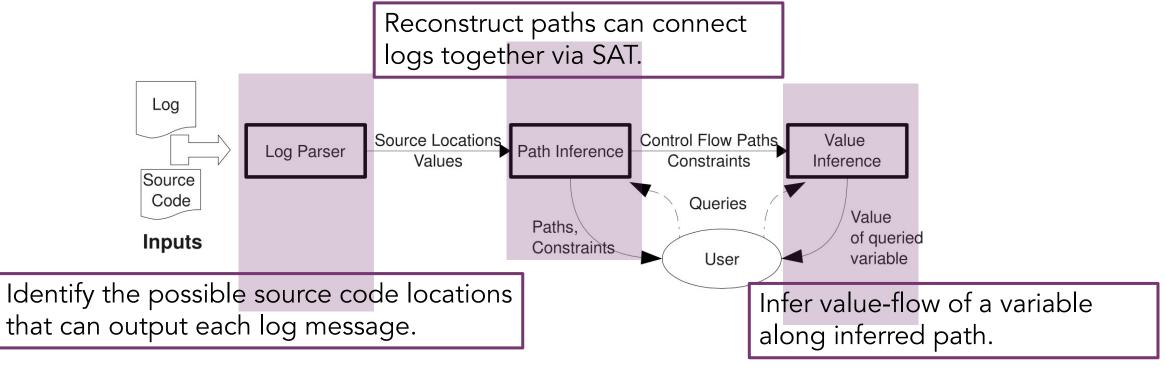
LogMap Approach

- 1. LogMap retrieves log messages from bug reports.
- 2. Apply static analysis technique to *identify corresponding logging lines* in source code.
- 3. Traverse through logging lines to *derive the potential code paths*, which help reconstruct the execution path and assist debugging process.

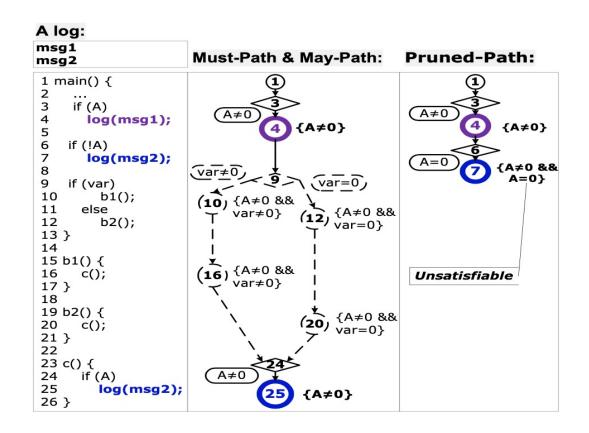
Execution Replay Methods (2)

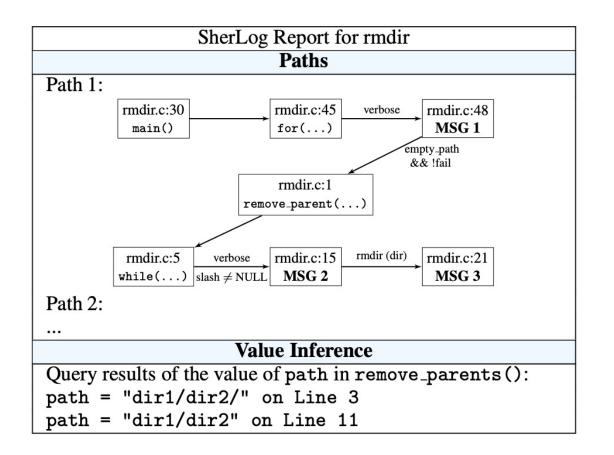
Goal: Find information from printed log

- Must-have: Partial execution paths that were definitely executed.
- May-Have: Partial execution paths that may have been executed
- Must-Not-Have: Execution paths that were definitely NOT executed.

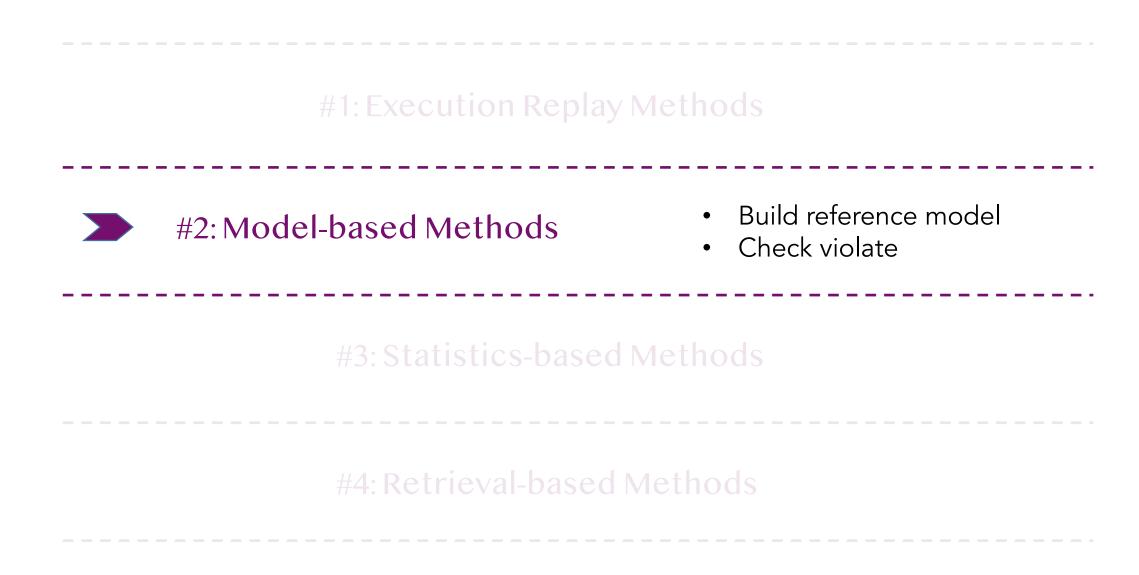


Execution Replay Methods (2)





Outline

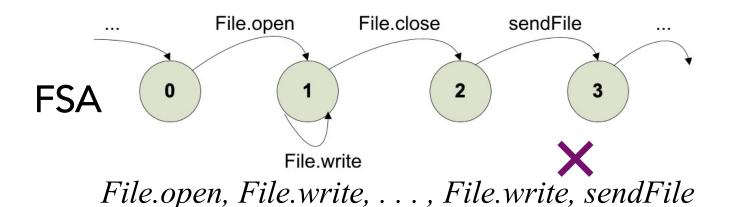


Model-based Methods

Approach

Utilize logs to build the reference model for a software system, then check which log events violate the reference model.

- Generate a model of the legal behavior of a target system by tracing successful executions at testing-time.
- Compare the set of events detected during failing executions with the generated models.



Model-based Methods (1)

How to generate reference model from log?

How to compare?

```
Raw Logs:
(1)2016-10-10T01:00:18.866Z parse_params: {"inline-relations-depth"=>"1"}
(2)2016-10-10T01:00:18.867Z parse_params: {"inline-relations-depth"=>"1", "q"=>"name:test-app-1476011110087"
(3)2016-10-10T01:00:18.870Z cc.dispatch host: 9e65a679-f602-4366-bfe8-24b3efa73230
(4)2016-10-10T01:00:18.957Z Request failed: 400: {"code"=>170002, "description"=>"App has not finished staging'} parameters
                                                                                                                                             flow 1
(5)2016-10-10T01:00:18.964Z cc.dispatch host: 9e65a679-f602-4366-bfe8-24b3efa73230
                                                                                                                                            flow 2
(6)2016-10-10T01:00:18.987Z Completed 200 vcap-request-id: d89bb92b-bcc5-467b-4120-54e82eac7873::2faebb11-7343-4d95-bb75-a7e7e59eedea flow 2
(7)2016-10-10T01:00:18.987Z Completed 400 vcap-request-id: f010f420-04f5-425a-4aab-a38bbcc18a9f::bf38597a-acc8-4180-8bb6-943c356b5d40
                                                                                                                                            flow 1
 Log Templates:
                                                                             Ground Truth time-weighted CFG:
 1)parse_params: {*}
                                                     T1
                                                                                         flow 1
                                                     T1
 (2)parse params: {*}
                                                     T2
  (3)cc.dispatch host: *
 (4)Request failed: 400:{"code"=>*, "description"=>"*"
                                                     T3
 (5)cc.dispatch host: *
 (6)Completed 200 vcap-request-id: *
                                                                                    flow 2
 (7)Completed 400 vcap-request-id: *
```

Mine time-weighted control flow graphs (TCFG) from interleaved logs of a system.

Node: Template

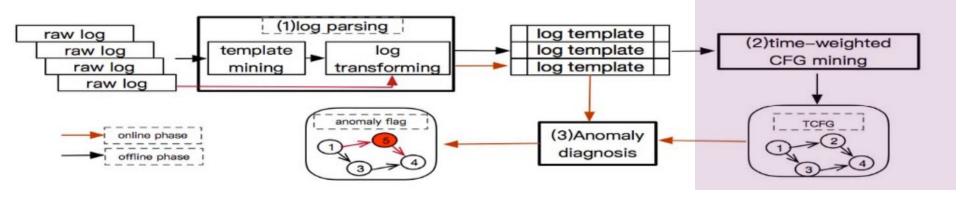
Edge: Transition from one template to another

Weighted: Execution time between templates

Model-based Methods (1)

How to generate reference model from log?

How to compare?



Challenges: Without a certain transaction ID.

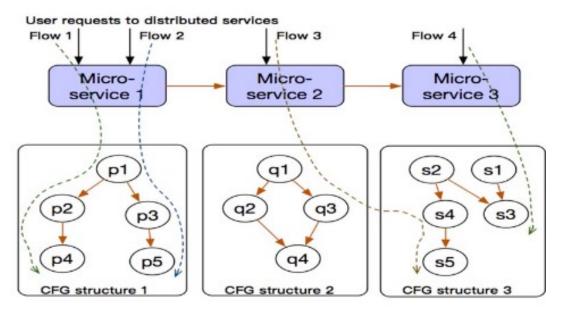
Motivation: Logs of correct immediately succeeding template will appear in a short period.

- **Two Stages:** (1) Compute frequent successor groups called FS groups of each template.
 - (2) Leverage this FS group to mine only the immediate successors of each template.

Model-based Methods (2)

How to generate reference model from log?

How to compare?



Model the causal relationships between services as a service topology.

Node: Service

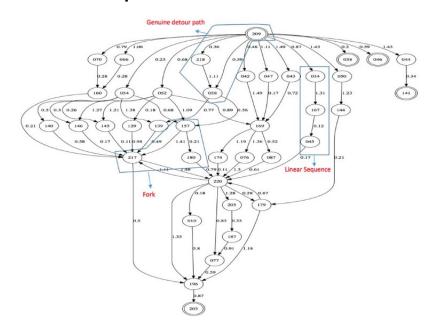
Edge: Causal relationship between two services

Mining approach: PC-algorithm

Model-based Methods

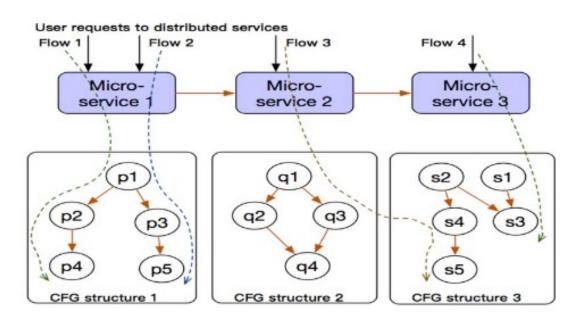
How to generate reference model from log?

How to compare?



TCFG

Diagnose for sequence anomaly, redundancy anomaly and latency anomaly.



Service topology.

Examine the TCFGs of its predecessor services in the service topology.

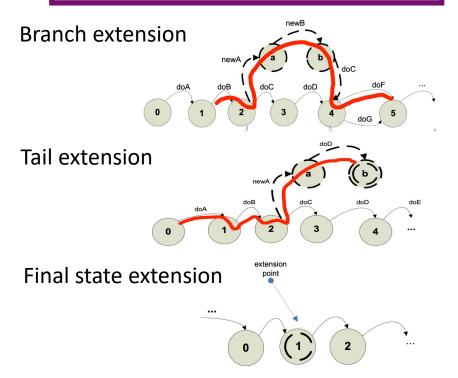
Model-based Methods

I

How to generate reference model from log?

How to compare?

Select the expected behaviors to be considered for local analysis.



(delete, insert, replace and final state)

Compare expected behavior with the observed failed sequence.

Example insert interpretation:

expected sequence: doA doB - - - doC observed sequence: doA doE doF doG doH doC

Example delete interpretation:

expected sequence: dol doL doF doG - doM observed sequence: dol - - doN doM

Sets specific to anticipation

I.newEvents = doE doF doG doH
D.removedEvents = doL doF doG

Aligned sequences:

doE doF doG doH doL doF doG -

General sets

I.m = doA doC D.m = doI doM

score =
$$\frac{2*2 \cdot 2*1 \cdot 1 + 2 + 2 \cdot 0 \cdot 1}{6 + 6} = 0.33$$

confidence value =(0.33+1)/2=0.665 (normalized score)

Outline

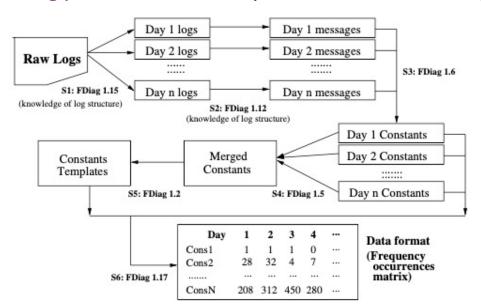
#3: Statistics-based Methods • Statistical analysis

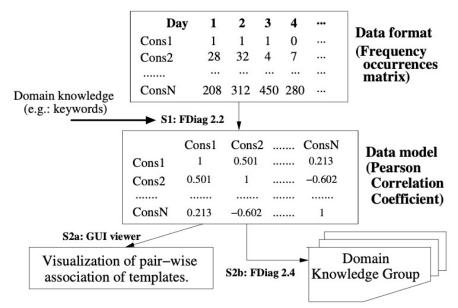
Statistics-based Methods

Approach

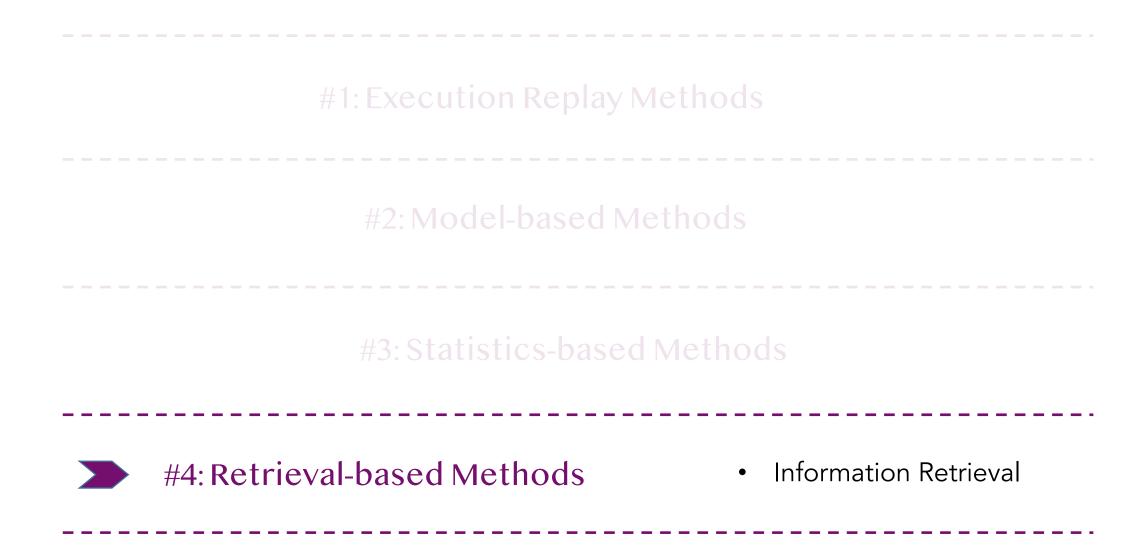
Employ some statistical techniques to capture the relationships between logs and the consequent failures.

Given knowledge of any system event, find all the system events that are associated with the given event, and how strongly associated these system events are to the given event.





Outline



Approach

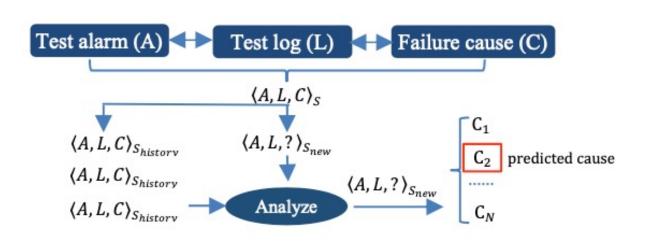
Retrieve similar failures in a knowledge base composed of failures in history.

Motivation: In practice, failures that previously occurred are valuale since they can aid developers in better diagnosing newly-occurred failures

Basic Idea: Detect the test logs of historical test alarms that may share the same causes with the new test logs.

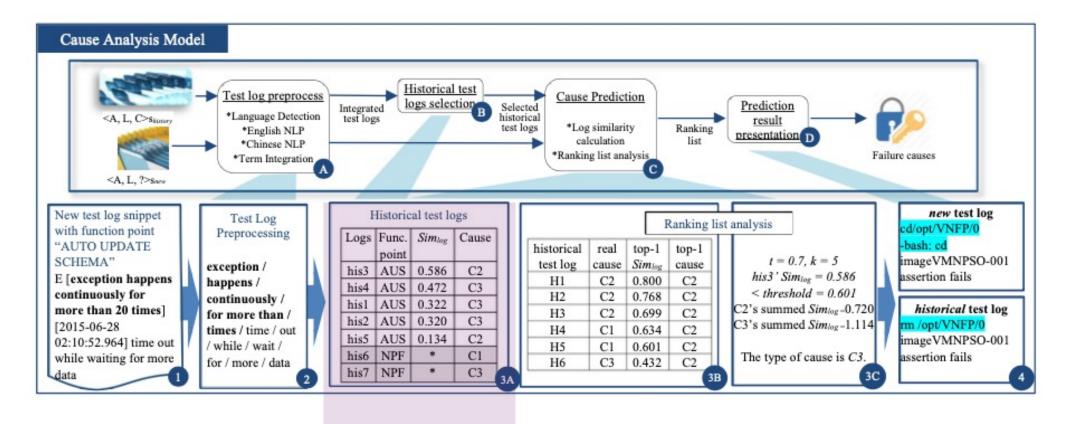


Bug Report

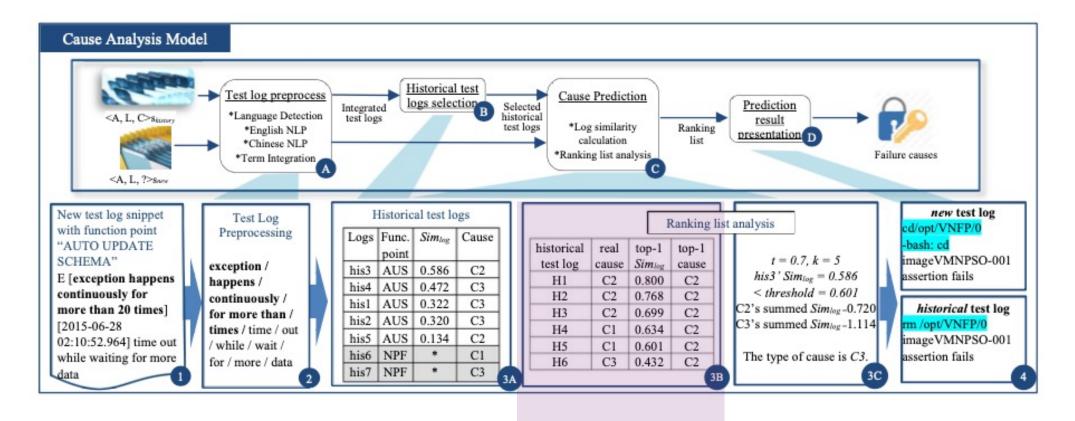


Task Formulation: Give <A, L, ?>, predict correct C

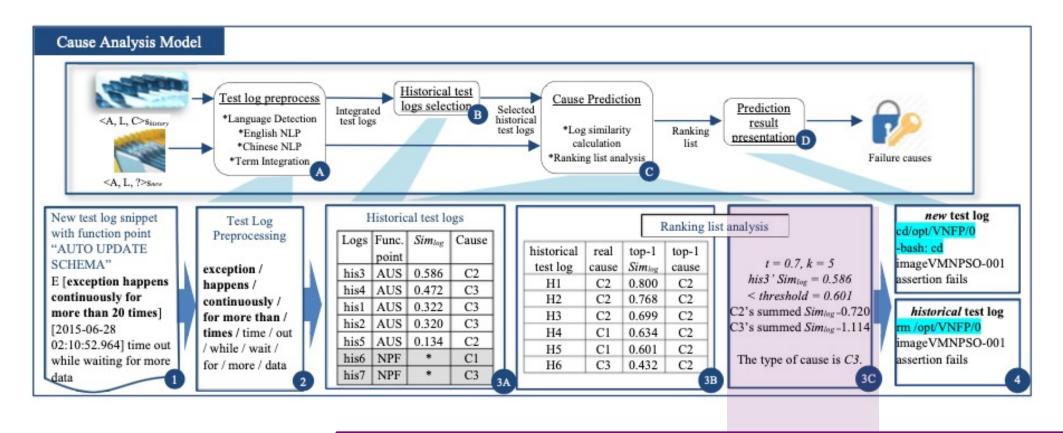
Evaluation: Test in two industrial testing projects at Huawei-Tech Inc.



Selects historical test logs by examining the *function points of test scripts*, since the authors find that test scripts with the same function point usually target the same functionalities to check.



Predict the cause of a new test alarm by first ranking the selected historical test logs according to their similarities with the new test log.

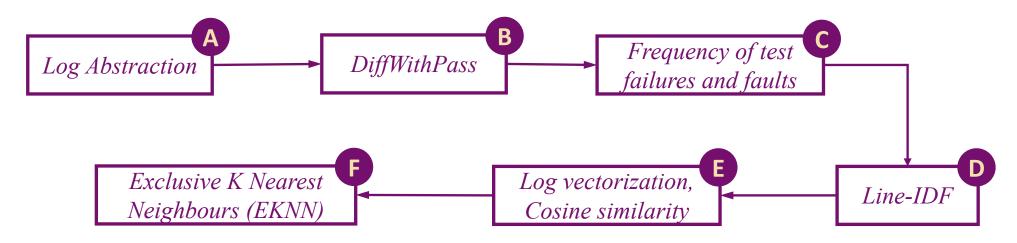


Normally predict the test alarm cause with the top-1 cause in the ranking list. If the similarity is low, use KNN strategy.

Previous limitation and new challenge:

- The complex test environment introduces many non-product test failures. #FaultsFound
- The logs contain an overwhelming amount of detailed information.
 #LogLinesFlagged

Approach:

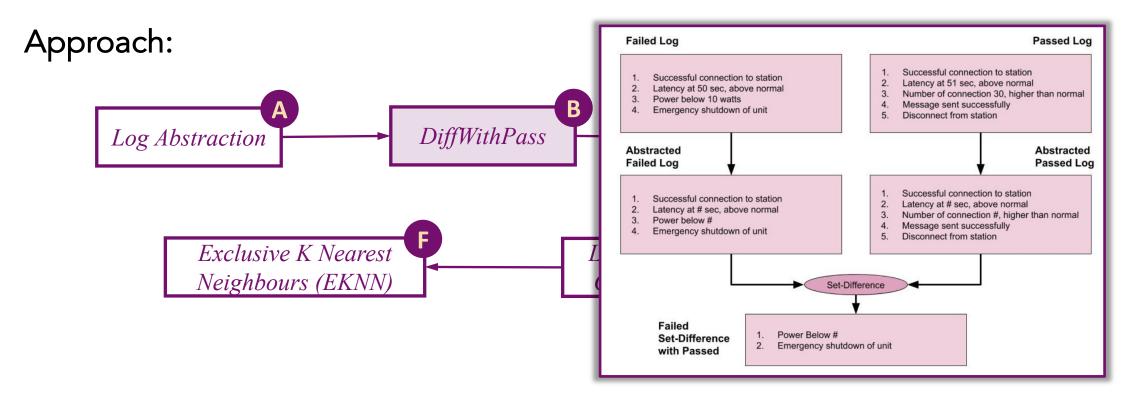


Previous limitation and new challenge:

- The complex test environment introduces many non-product test failures.
- The logs contain an overwhelming amount of detailed information.

#LogLinesFlagged

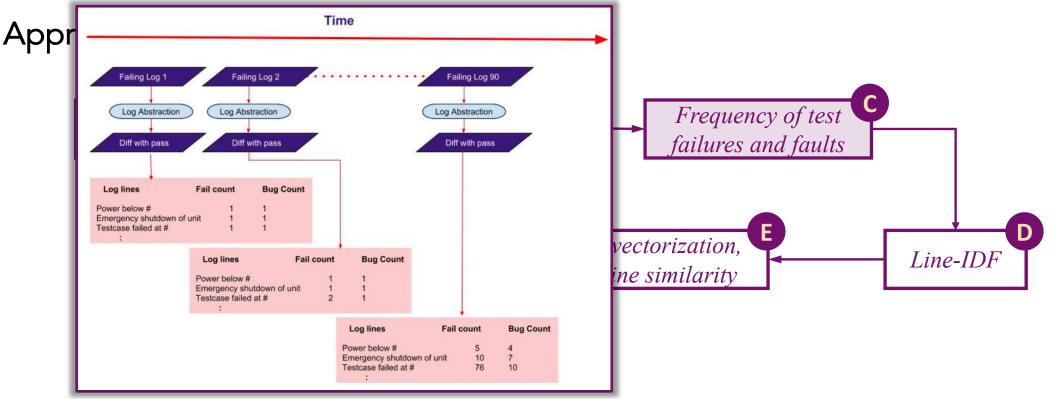
#FaultsFound



Previous limitation and new challenge:

- The complex test environment introduces many non-product test failures.
- The logs contain an overwhelming amount of detailed information.

#FaultsFound #LogLinesFlagged

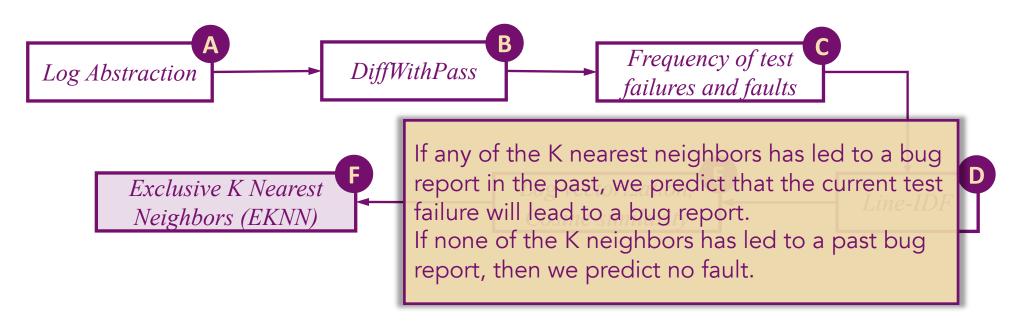


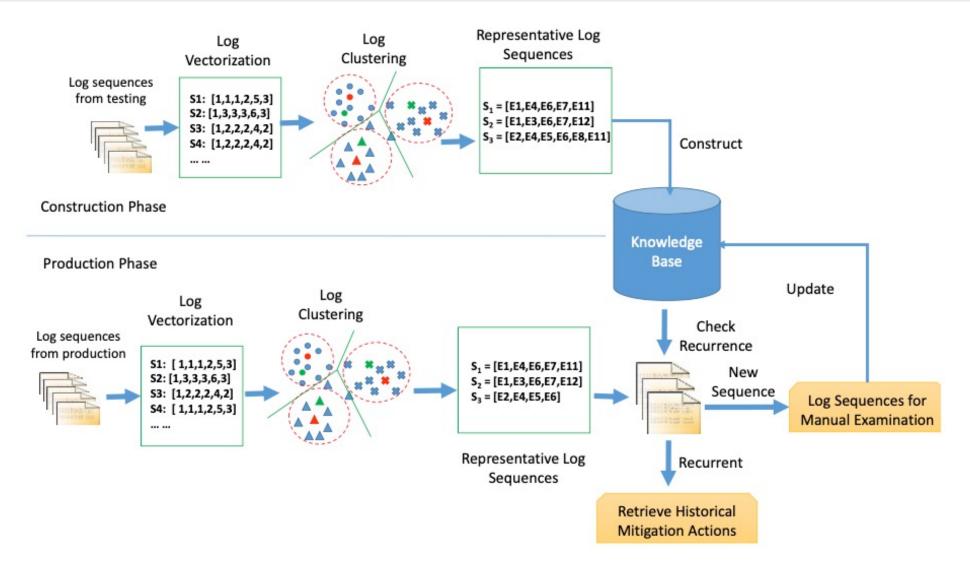
Amar A, Rigby P C. Mining historical test logs to predict bugs and localize faults in the test logs[C]//2019 IEEE/ACM 41st International Conference on Software Engineering (ICSE). IEEE, 2019: 140-151.

Previous limitation and new challenge:

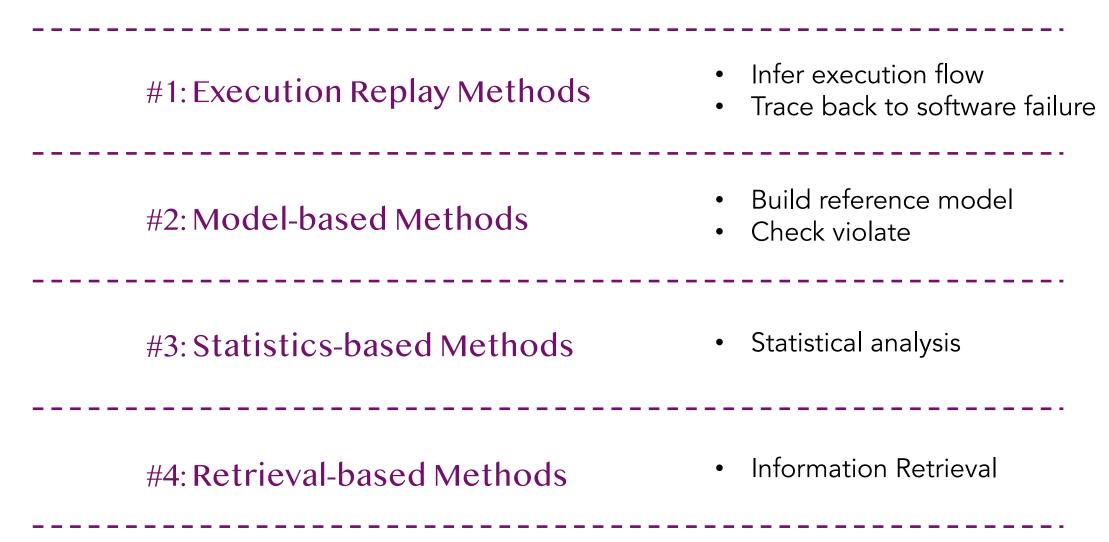
- The complex test environment introduces many non-product test failures. #FaultsFound
- The logs contain an overwhelming amount of detailed information. #LogLinesFlagged

Approach:





Conclusion



SherLog: Error Diagnosis by Connecting Clues from Run-time

Thank you!



