Accurate Troubleshooting Recommendations Using Online Forum and Language Models

온라인 포럼과 언어 모델을 활용한 시스템 문제 해결 방법 추천

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Problems in Cloud Systems

- Various cause of software problems
 - Misconfiguration
 - Mistaken by human, Environment changes, ...
 - Hardware failure
 - Cpu/memory/storage/switch failure, temporary network partitioning
 - Resource shortage
 - Insufficient memory space, storage space, cpu capacity, network saturation
 - Network failure
 - Bug in source code

Researches on System Problems

Anomaly detection

- Building anomaly propagation graph [1]
- Applying bayesian linear attribution [2]
- Time-series metrics of sensor, network fingerprint, system calls [3,4,5]
- Analyzing flow and pattern of log [6,7]
- Workflow monitoring via interleaved logs [8]

Fault localization

- Network monitoring to localize fault component for OpenStack [9]
- Combining multiple dimensions value to analyze abnormal KPI [10]
- Building dependency graph using system calls [11]

Definition of troubleshooting

Most of the researches focus on detecting anomalies & fault localization. Still, not enough work on finding root cause or direct solution of problem

- [1] Automated anomaly detection and root cause analysis in virtualized cloud infrastructures (NOMS'16)
- [2] BALANCE: Bayesian Linear Attribution for Root Cause Localization. (arxiv'23)
- [3] Root cause detection in a service-oriented architecture. (SIGMETRICS'13)
- [4] Sieve: Actionable insights from monitored metrics in distributed systems (Middleware'17)
- [5] Root cause localization for unreproducible builds via causality analysis over system call tracing (ASE'19)
- [6] Deeplog: Anomaly detection and diagnosis from system logs through deep learning (CCS'17)

[7] Execution anomaly detection in distributed systems through unstructured log analysis (ICDM'09)

- [8] CloudSeer: Workflow monitoring of cloud infrastructures via interleaved logs (ASPLOS'16)
- [9] HANSEL: Diagnosing faults in OpenStack [CoNEXT'15]
- [10] CMMD: Cross-Metric Multi-Dimensional root cause analysis [KDD'22]
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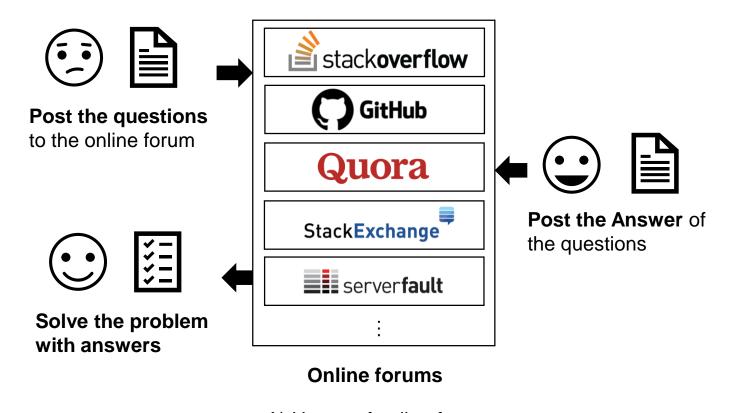
Research Problem

Goal: Finding the <u>root cause or direct solution</u> to a given system problem

- Motivation:
 - Finding out the root cause or solution is too difficult and require high domainknowledge
 - We often use sensor, log or system calls, but <u>online forum has already vast</u> <u>amount of troubleshooting data.</u>
 - However, no suitable approaches exists to use forum data effectively.
 - Nowadays, new NLP and AI techniques help us to use online forum data effectively to system troubleshooting.
- Approach: <u>Using NLP and AI techniques</u>, <u>retrieving the most problem-relevant post</u> <u>which has root cause or solution in online forum</u>

What is Online Forum?

- An online discussion site where people can hold conversations in the form of posted messages.
 - Popular Online Forum: Stack Overflow, GitHub Issue, Quora, Serverfault, Stack Exchange, ...





Total:

23,772,352 Questions # 35+ million Answers # 21+ million Users

Per day:

7.6K new Questions # 5.8M visitors

What Kinds of Questions are in Online Forum?

Types of Information in Online Forum

Problems of system

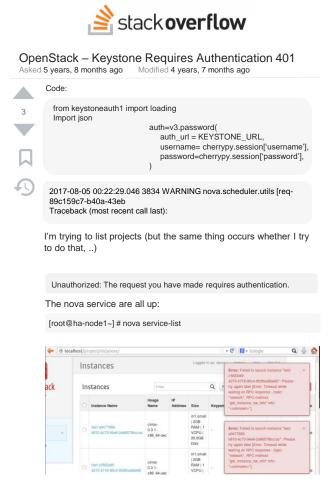
- Sql, mongodb, docker, apache-spark,...
- Q1. Openstack error when launching an instance
- Q2. Spark.java.lang.OutOfMemoryError: java heap space

Coding problem

- Javascript, python, java, c++, ...
- Q1. How do I merge two dictionaries with one expression?
- Q2. How can I remove a specific item from an array?

Conceptual question

- Q1. What and where are the stack and heap?
- Q2. Is java 'pass-by-value' or 'pass-by-reference'?



A) Example of stackoverflow questions about problem of system

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Types of Information in Online Forum

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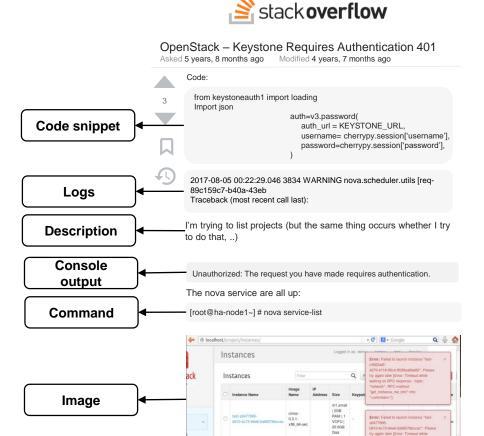
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A) Example of stackoverflow questions about problem of system

Researches on Online Forum

- Various kinds of research on online forum
 - Understanding characteristics of online forum [1, 2]
 - Analyzing coding aspects [3, 4]
 - Reformulating query [5]
 - What developers ask [6]

· Not many researches of online forum for troubleshooting

 ^[1] Tagging and linking web forum posts (ACL'10)

^{• [2]} Crowdsourced knowledge on stack overflow: A systematic mapping study. (EASE'17)

^[3] Sotorrent: Reconstructing and analyzing the evolution of stack overflow posts (MSR'18)

^{• [4]} From query to usable code: an analysis of stack overflow code snippets (MSR'16)

^{• [5]} Automated query reformulation for efficient search based on query logs from stack overflow (ICSE'21)

^[6] Going big: A large-scale study on what big data developers ask (FSE'19)

• Goal: Retrieving the most relevant online forum post related to system problem

Goal: Retrieving the most relevant online forum post related to system problem

Failure-case data (user-side data)

Command

Code

yyk@DGX-Station~\$: openstack image create -file cirros "cirros"

```
yyk@DGX-Station~$ python3 spark.py

spark = SparkSession.builder.getOrCreate()

data = [("a1", 2), ("b1", 4)]

df = spark.createDataFrame(data, ["fruit_1", "quantity_1"])

condition = df.select("fruit_1"), when(col("quantity_1") > 2, col("fruit_1") == "a1")

df = df.withColumn("new_column_1", when(condition == "many", "Lots of ").otherwise("Not many") + col("fruit_1"))
```

Goal: Retrieving the most relevant online forum post related to system problem

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2017-08-05 00:22:29.246 WARNING nova.scheduler.utils

yyk@DGX-Station~\$ python3 spark.py

Log

[req-89c159c7-b40a-43eb-8]

Failed to compute_task_build_instances: No valid host was found. There are not enough hosts available. Setting instance to ERROR state.

Console output

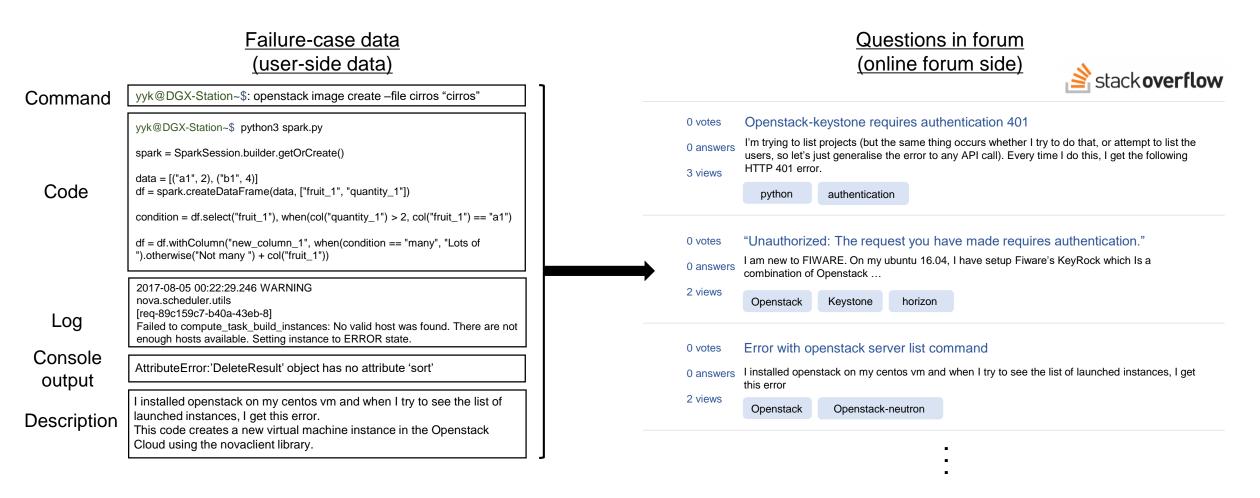
AttributeError:'DeleteResult' object has no attribute 'sort'

Description

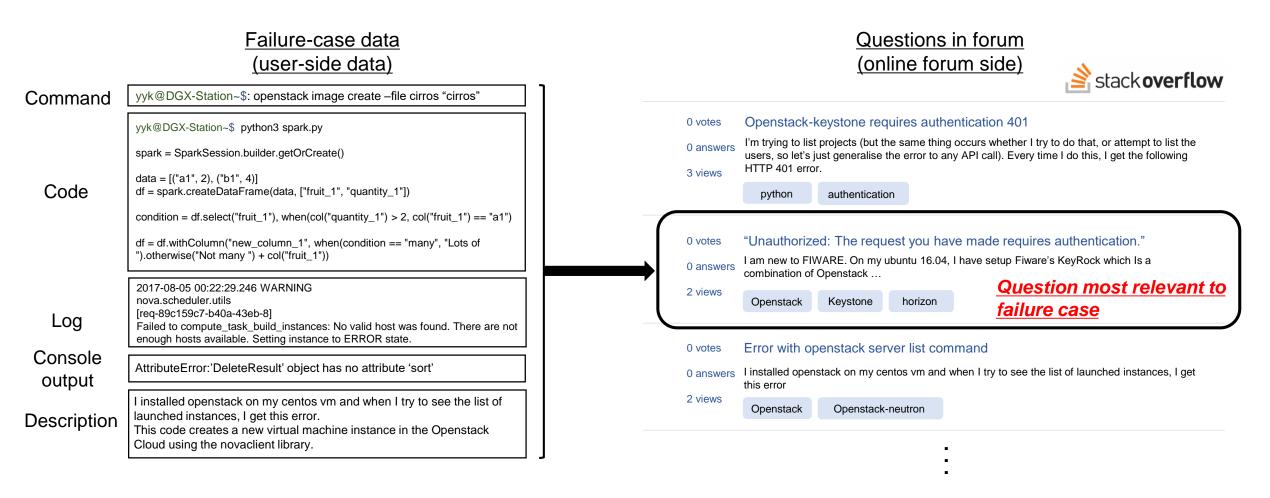
I installed openstack on my centos vm and when I try to see the list of launched instances, I get this error.

This code creates a new virtual machine instance in the Openstack Cloud using the novaclient library.

Goal: Retrieving the most relevant online forum post related to system problem



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Naïve Search in Online Forum

- Naïve search 1. Searching using <u>log or console output from failure case</u>
 - Problem: Irrelevant document can have same logs, because one log appears in multiple operation

One log appears when multiple commands and codes are executed in OpenStack

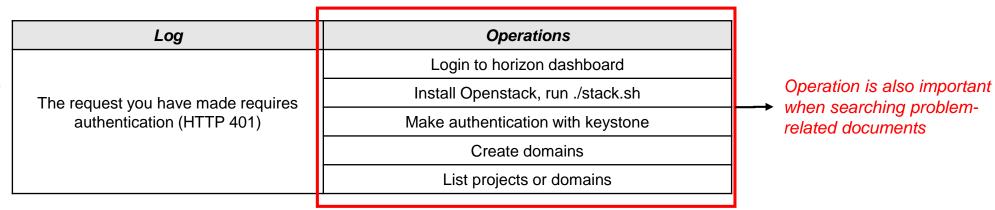
Log	Operations					
	Login to horizon dashboard					
The request you have made requires authentication (HTTP 401)	Install Openstack, run ./stack.sh					
	Make authentication with keystone					
	Create domains					
	List projects or domains					

- Naïve search 2. Searching <u>using code and command string</u>
 - Problem: Code and command can be changed because of version update.
 - Problem: Code does not share the same word between the same tasks
- Naïve search 3. Searching with <u>user-written description</u>
 - Problem: Search results vary depending on how detailed the user explaining the situation.

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Does 3 Types of Naïve Search Works?

Naïve search does not find relevant document.

Naïve search 1. Logs:

File "/usr/lib/python3/site-packages/keystoneauth1/session.py, line 484 in request raise exceptions.from_response(resp, method, url)

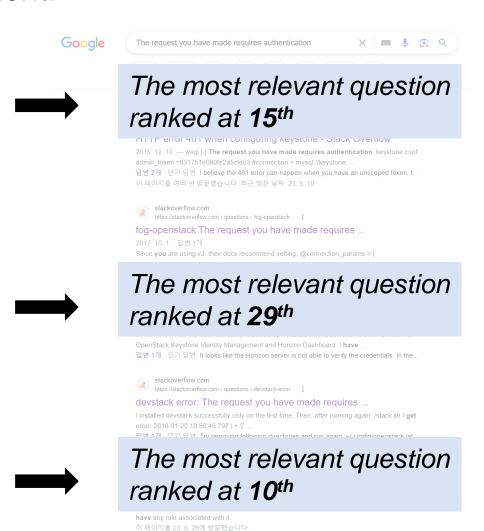
. .

Unauthorized: The request you have made requires authentication (HTTP 401)

Naïve search 2. Code:

Naïve search 3. User-written description

I tried to authenticate and list project in openstack.



- Because of the assumption that 'the same data will exist in relevant forum posts'.
 - When searching with command, user thinks that same command will exist in relevant post.
- But <u>relevant posts do not always contain the same data</u>, such as...
 - Users search with <u>command</u>, but same information is <u>in the description</u>.
 - Users search with <u>description</u>, but same information <u>exists by code</u>.

Searching with **Command**:

Openstack create server -image -flavor <server-name>

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Same data in post as **Description**:

I installed openstack on my centos vm and when I try to launch the instance I get this error.

a) Command and description share same information

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Same data in post as Code:

def list_servers_nova(conn):
print("list servers with nova:")
for server in conn.compute.servers():
print(server)
list_servers_nova(conn)

b) Description and code share same information

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Searching with Command:

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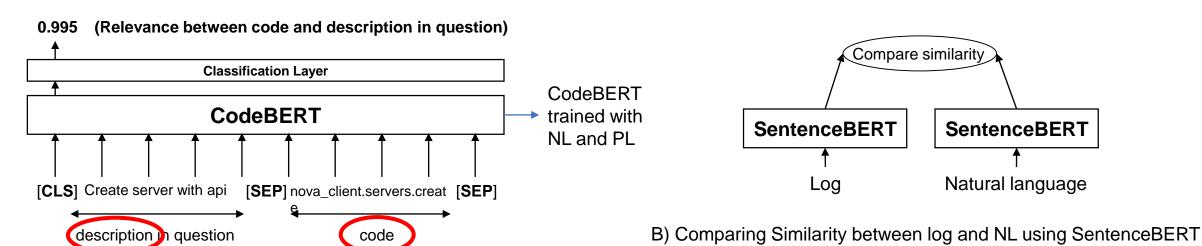
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How Can We Solve This Problem?

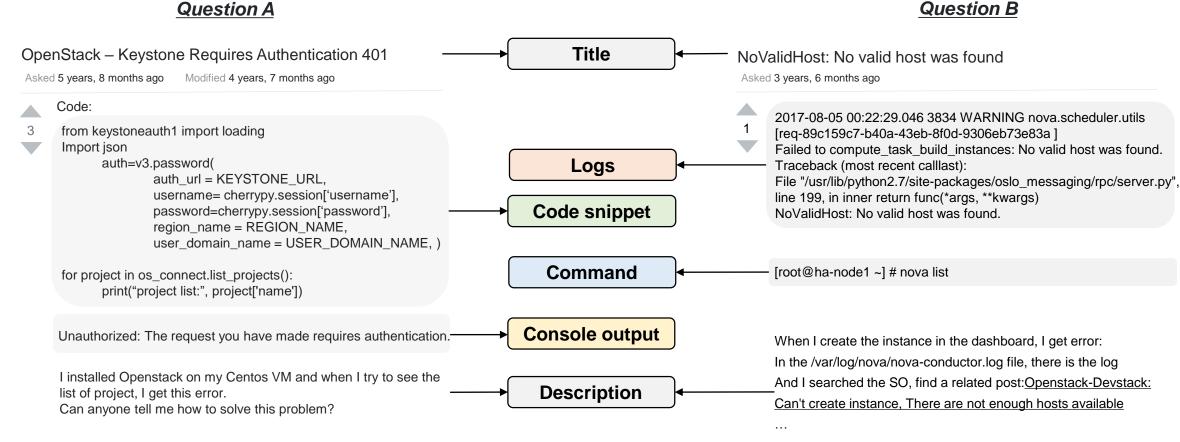
- To resolve previous limitation, we need to compare between disparate data type.
 - comparison between command and description, or comparison between code and description
- Nowadays, bi-modal model such as **CodeBERT and SentenceBERT** enables the semantic search (enables to compute relevance) between disparate data types.
- We treat the online forum data <u>as combination of multiple disparate data</u>.
- We use this technique to <u>compare disparate types of data</u> for retrieving relevant online forum post



A) CodeBERT compute relevance between description and code

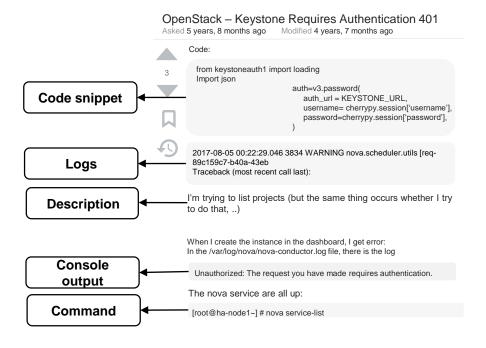
Multi-Modality of Online Forum Data

- Special aspect of online forum is inclusion of multiple types of data
 - Same type of data in failure case exist in forum posts.
 - machine languages (code), short phrases (command), natural language (log, console output, description)



Approach: Separating forum posts and applying 3 types of model

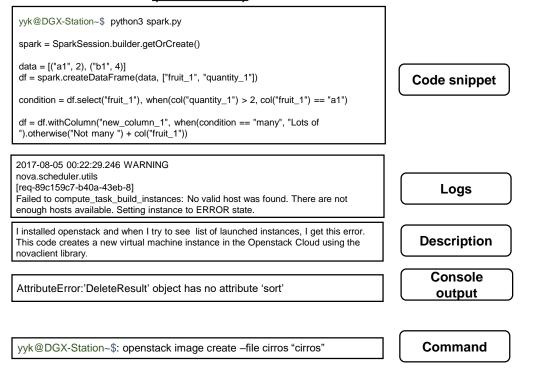




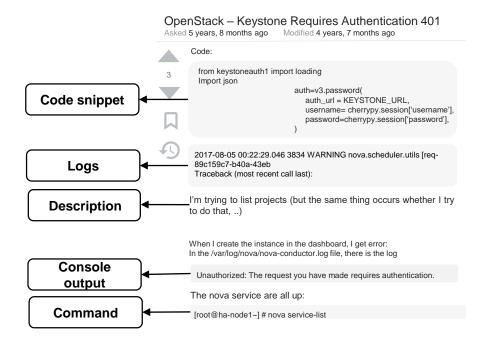
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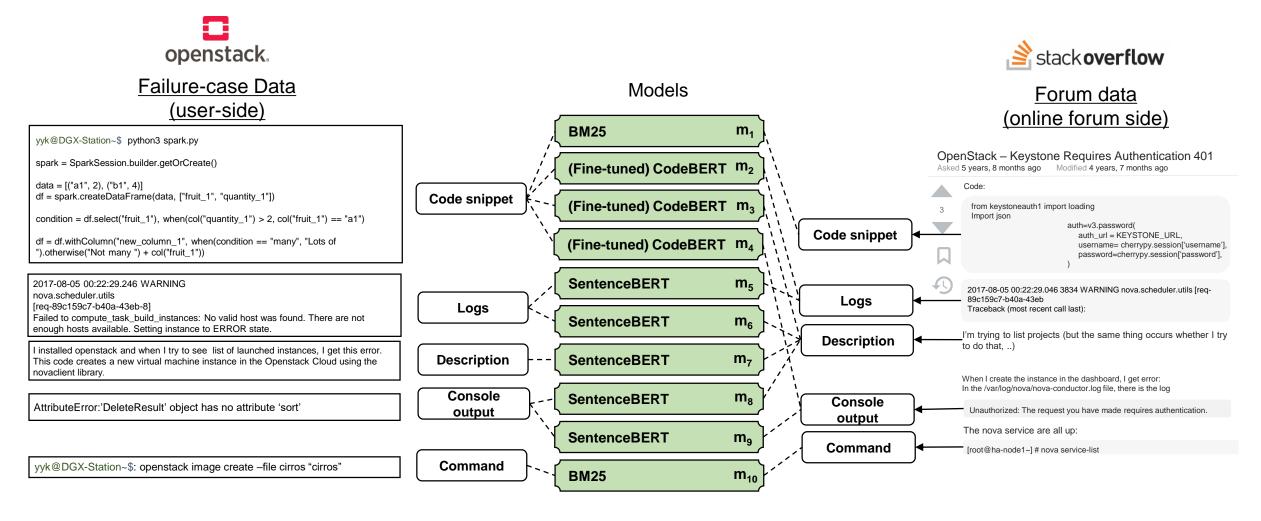
Failure-case Data (user-side)



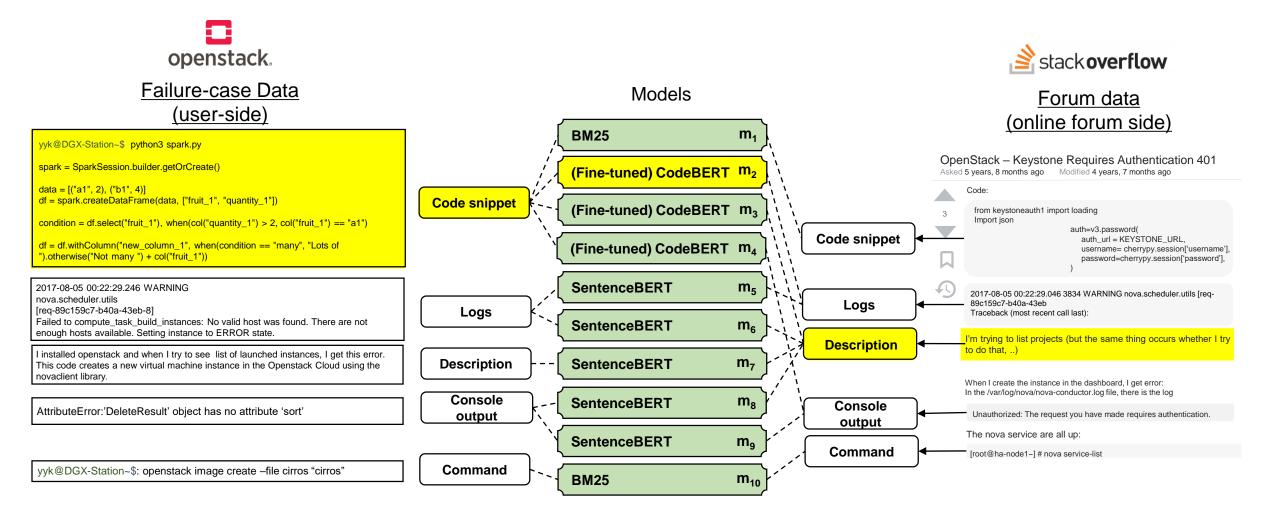




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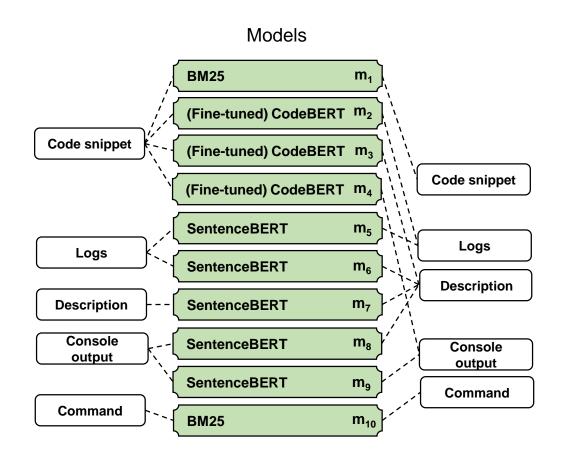


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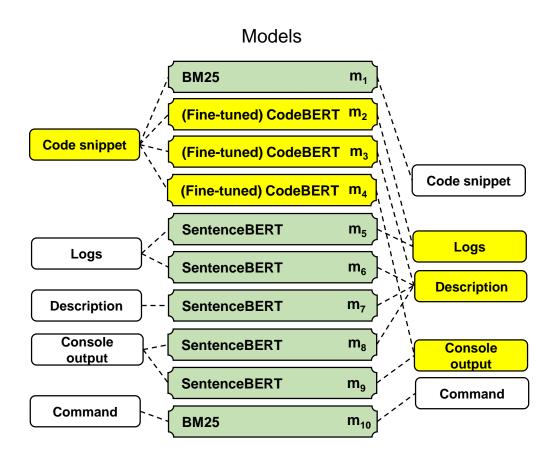


- Let's learn what characteristics of each data make it use CodeBERT, S-BERT, BM25
- Relevance score computing mechanism

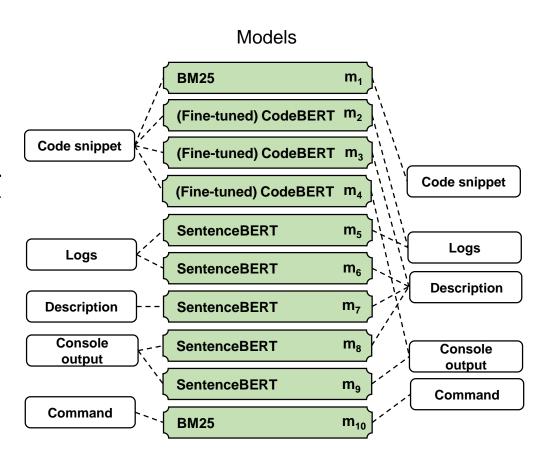
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 - Cod:Tnd, Cod:Log, Cod:Cns



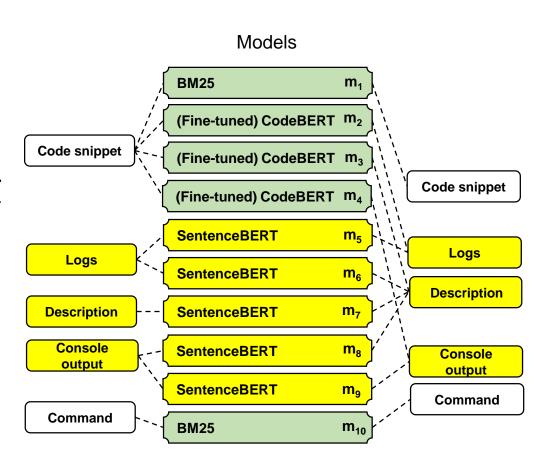
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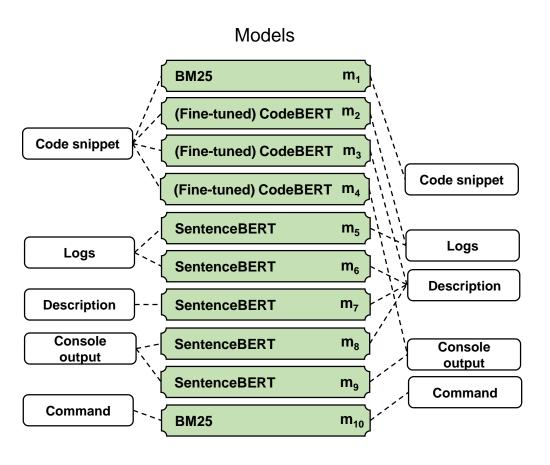
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 - Searching similar sentence in post: SentenceBERT
 - Log, Cns: The same text is posted in an online forum
 - Comparing data such as
 - Log:Log, Log:Tnd
 - Cns:Cns, Cns:Tnd
 - Des:Tnd



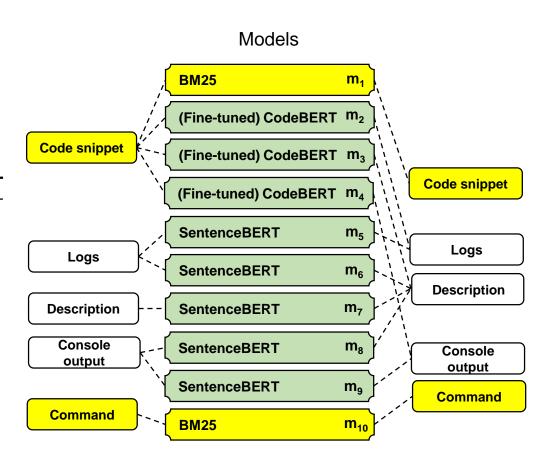
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 - Searching same token in post: <u>BM25</u>
 - Command: Fixed format so same data exist in post
 - e.g. "Openstack create image"
 - Code: Sharing same word when using API
 - Comparing data such as Cod:Cod, Cmd:Cmd

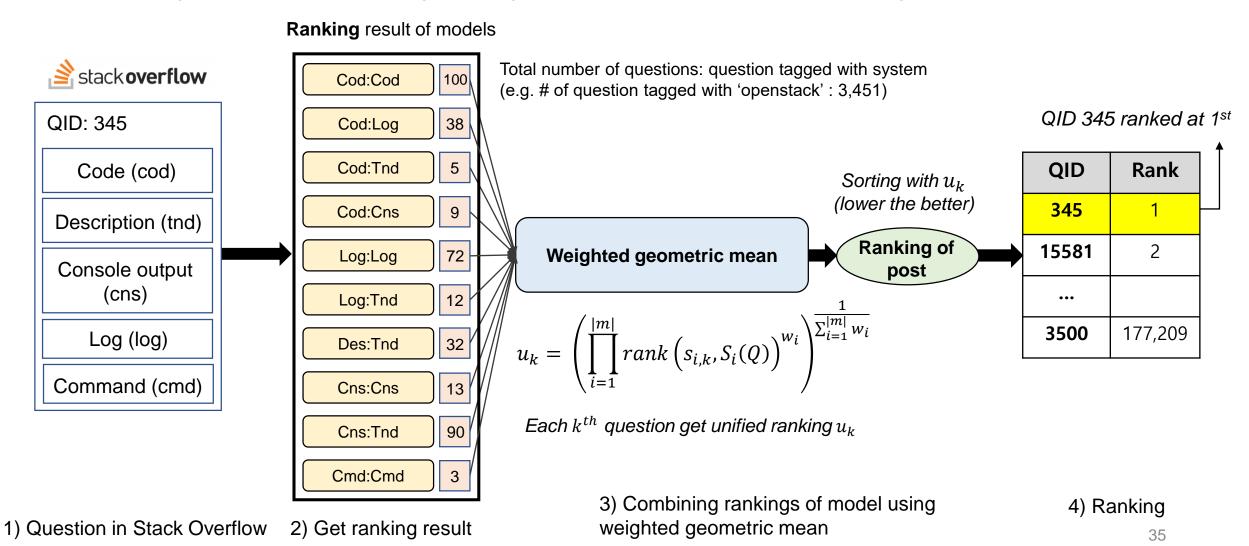


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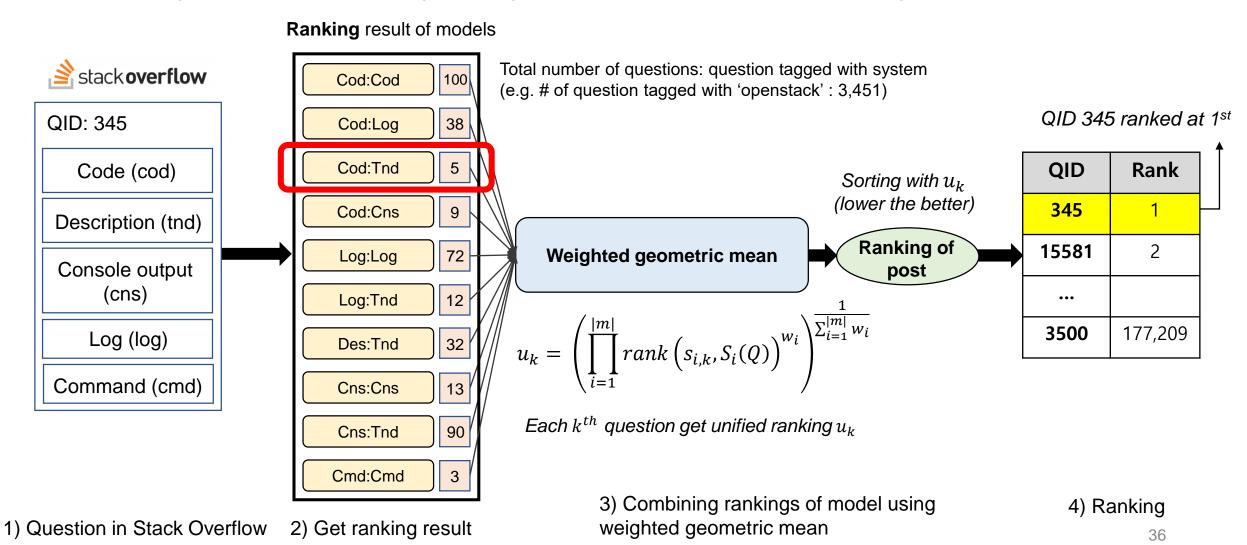
How to Combine Results of Multiple Models?

Combining results with weighted geometric mean and ranking documents



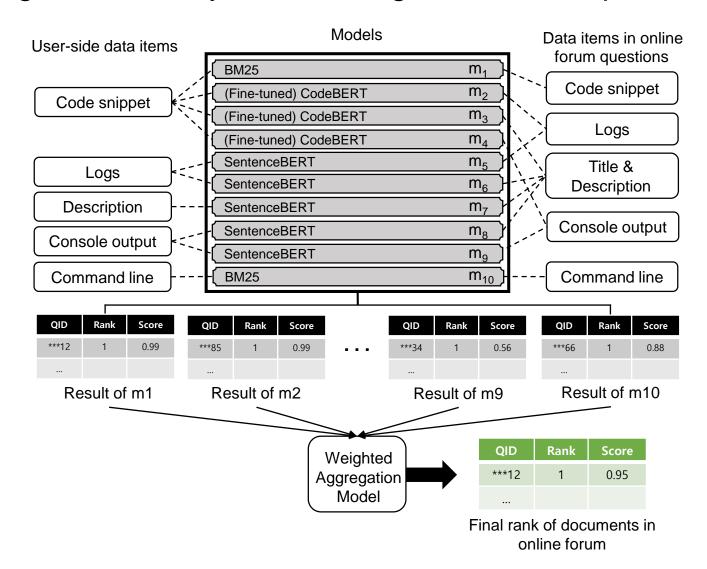
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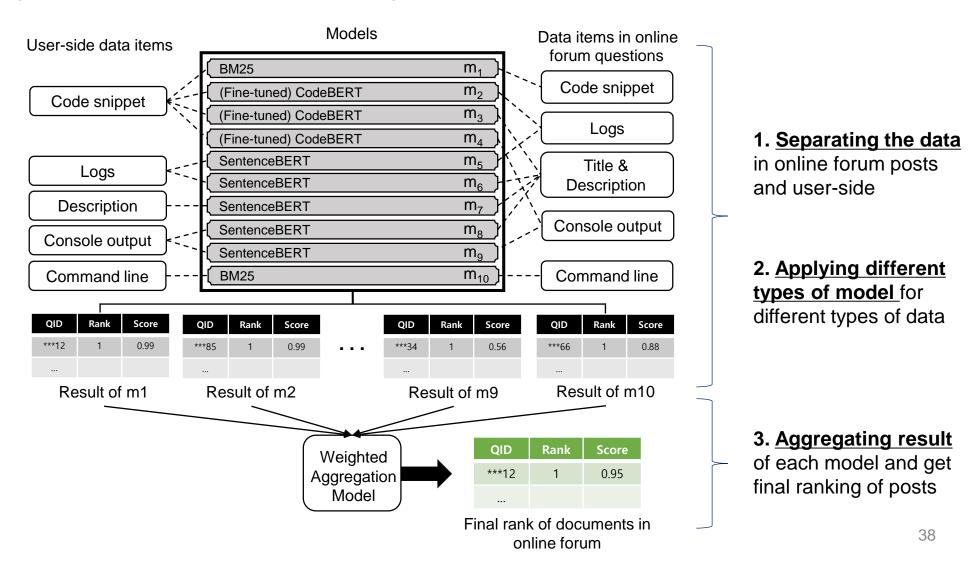
Overall Architecture of FDD

Goal: Improving the accuracy of searching online forum post related to problem



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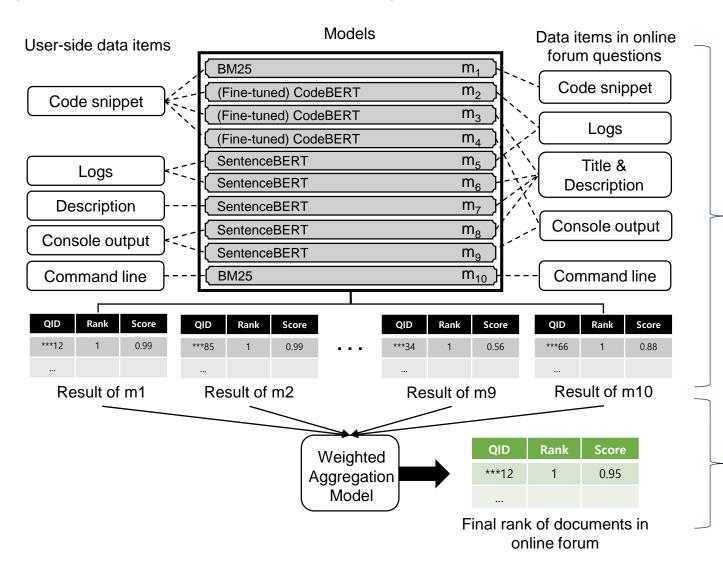
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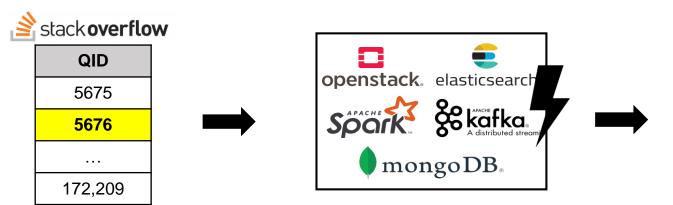


- 1. <u>Separating the data</u> in online forum posts and user-side
- 2. Applying different types of model for different types of data

3. <u>Aggregating result</u> of each model and get final ranking of posts

How to Evaluate FDD?

- Benchmark cases
 - 5 Target application: Openstack, MongoDB, Spark, Elasticsearch, Kafka
 - 77 Failure cases are manually reproduced



Case No. 15							
Command	Openstack create image –file cirros						
Console output	The request you have made requires authentication						
Description	I want to create image in openstack						

1) Selecting question

2) Reproducing failure cases

3) Collecting operation context & operation output

• Evaluation metric: MRR / SuccessRate@K

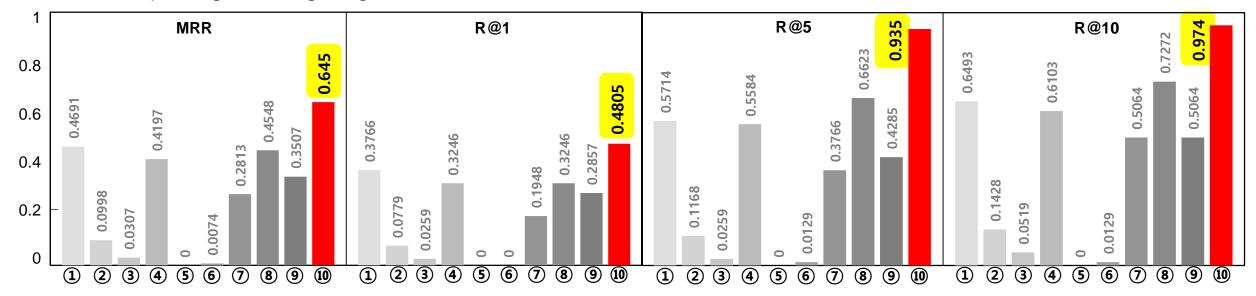
MRR	SuccessRate@K						
$ ext{MRR} = rac{1}{ Q } \sum_{i=1}^{ Q } rac{1}{ ext{rank}_i}$	$SuccessRate@K = \frac{1}{ C } \sum_{c=1}^{c} \delta(Rank_c \le k), k \in \{1,5,10\}$						
 How high the question is ranked If all questions ranked 1st, MRR = 1 Higher the better 	 How many questions are ranked within K If all questions ranked 1st, SuccessRate@K = 1 Higher the better 						

Result of Spark Using FDD (Total postings: 100,729)

CaseNo.	Type of Case	BM25	Doc2vec	BERT	SBERT	BERT(F)	CBERT(F)	G (a)	G (b)	ST	ChatGPT	FDD
S1	Kmeans with python	10	486	18078	35	71576	656	NF	7	1	X	2
S2	Connect Mongodb	1	33857	28359	18	86309	3099	20	1	NF	X	1
S3	Calling java/scala function	17	3119	11332	1	67103	8053	69	69	4	X	3
S4	Create dataframe	3	7025	6975	867	31954	7731	NF	NF	NF	X	1
S5	Use multiple conditions	6	271	354	1	19115	23058	26	26	6	X	1
S6	Pyspark mongodb connect	432	14512	2360	4	80537	3795	13	13	NF	0	4
S 7	Check dataframe	21340	288	74353	4462	29628	12809	23	23	14	X	3
S8	Create spark session	1	13857	8987	1	66978	17214	3	1	1	X	2
S9	Create udf	1	4526	47966	39	75390	421	10	4	NF	0	1
S10	Read and write table	6	3101	36404	2	76320	16773	1	1	1	0	1
S11	Set spark configuration	2	86	21152	158	86366	32885	1	1	1	X	1
S12	Delete and recreate context	5	27741	6652	87	63470	27724	21	2	17	X	4
S13	Create dataframe	1	4536	3538	94	78517	38046	NF	NF	NF	X	4
S14	Run logistic regression	140	1786	21206	26	73910	4474	1	1	1	0	1
S15	Run mllib in pyspark	55988	19610	37652	189	76428	8223	NF	NF	NF	X	5
S16	Initializing spark context	1	235	4189	2	50009	7538	4	3	5	0	1
S17	Run spark on remote cluster	2	911	4689	5	82692	12404	6	2	9	X	1
S18	Run spark in cluster mode	47350	82365	29435	16193	98211	26704	40	12	NF	X	3

Improved Searching Accuracy With FDD

- Ranking quality comparison
 - Comparing ranking of ground-truth with 9 baselines
- 1 BM25 2 Doc2vec 3 BERT(CLS) 4 SentenceBERT 5 BERT(Fine-tuned)
 6 CodeBERT(Fine-tuned) 7 Google Search (a) 8 Google Search (b)
- (9) Stack Overflow Search (10) FDD

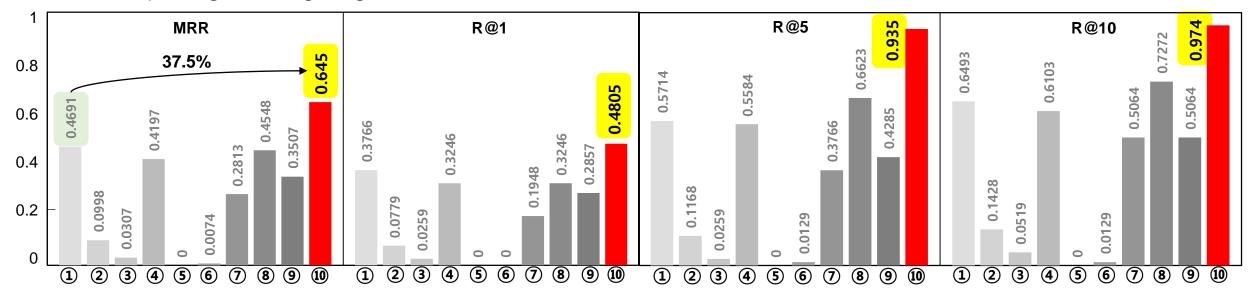


Result analysis

- FDD produced better results than all baselines. FDD <u>improves accuracy by 37.5%</u> compared with BM25.
- FDD ranked most of the cases(75/77) within 10th.
- BM25 showed the highest MRR among the baseline. High ranking only in command-based cases.
- By Google Search (b), 30% of cases were ranked at 1st. But the variance is too high (too many 'Not Found')

Improved Searching Accuracy With FDD

- Ranking quality comparison
 - Comparing ranking of ground-truth with 9 baselines
- 1 BM25 2 Doc2vec 3 BERT(CLS) 4 SentenceBERT 5 BERT(Fine-tuned)
 6 CodeBERT(Fine-tuned)
 7 Google Search (a)
 8 Google Search (b)
- (9) Stack Overflow Search (10) FDD



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Conclusion

- We try to troubleshoot by finding the most relevant online forum post with system failure.
- The approach of <u>recognizing multiple disparate data items within the</u> <u>online posting data</u> and <u>performing cross-data search</u> shows high effectiveness.
 - On 5 diverse applications, for 75 reproduced failure cases, our technique gave single-digit ranking of true online posts.
 - The accuracy improved 37.5% compared against best performing competitor.
- Limitation of FDD
 - Cross-data type search may not always work
 - logs or console output evolved too much across versions
 - Our accuracy is bound by the performance of the modeling technique
 - CodeBERT in FDD only accept code with less than 512 tokens.