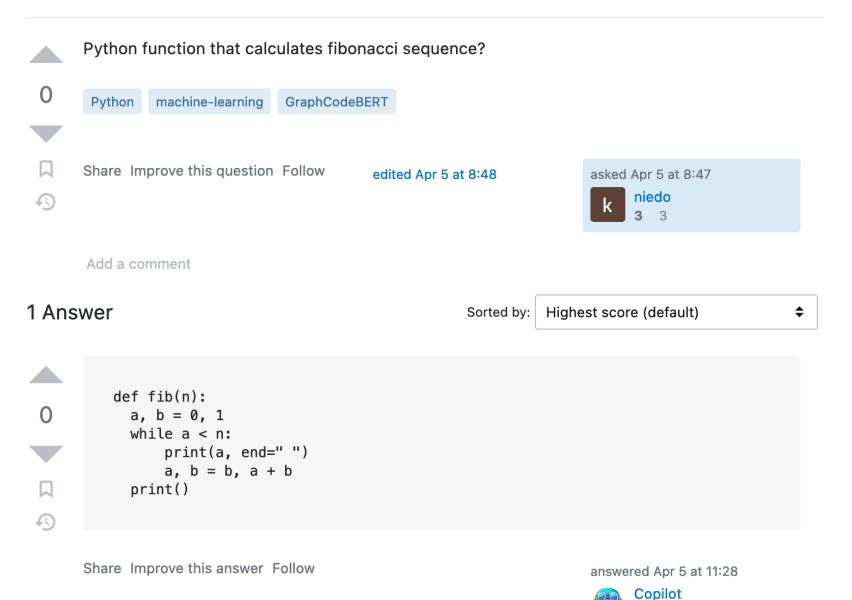
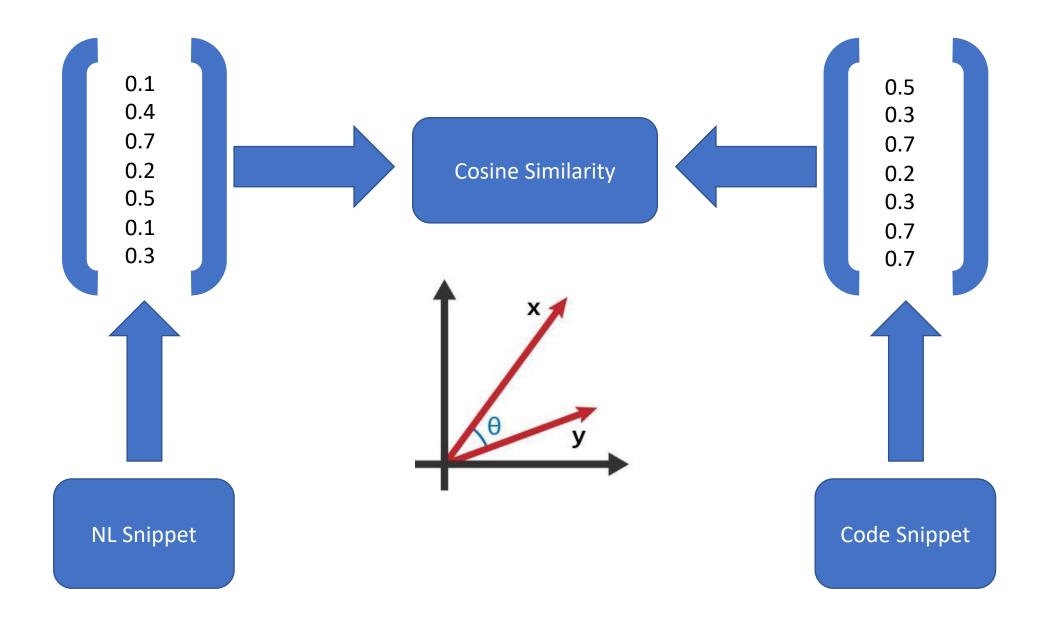
Fibonacci sequence calculation in python?

Asked 10 days ago Modified 9 days ago Viewed 29 times



1,186 1 14 22

Natural Language Code Search



https://www.learndatasci.com/glossary/cosine-similarity/

Natural Language vs. Code

- Very different structure
- Code has underlying logic, which strongly differs from NL

Criteria	Natural Language	Programming Language
Syntax	Unstructured and often ambiguous	Precise and structured
Vocabulary	Vast and diverse	Limited and well-defined
Ambiguity	Common	Rare
Error Tolerance	Tolerant of errors	Strict

Syntax/Ambiguity



- 1 def myFunction(myArguments):
 - 2 # do something

4

- 3 myResult = process(myArguments)
 - **return** myResult

Vocabulary

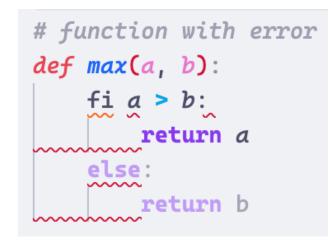
- Over 150k english words
- Python has 33 reserved keywords

Python Keywords			
False	def	if	raise
None	del	import	return
True	elif	in	try
and	else	is	while
as	except	lambda	with
assert	finally	nonlocal	yield
break	for	not	
class	from	or	
continue	global	pass	

RealPython: https://realpython.com/lessons/reserved-keywords/

Error Tolerance

Yuo cna porbalby raed tihs esaliy desptie teh msispeillgns



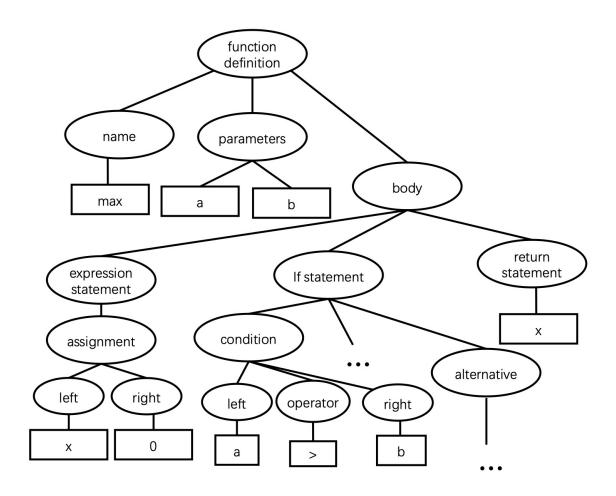
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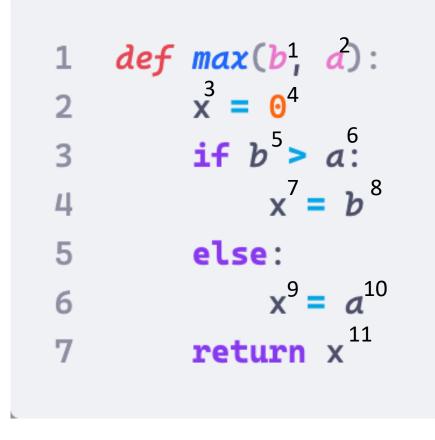
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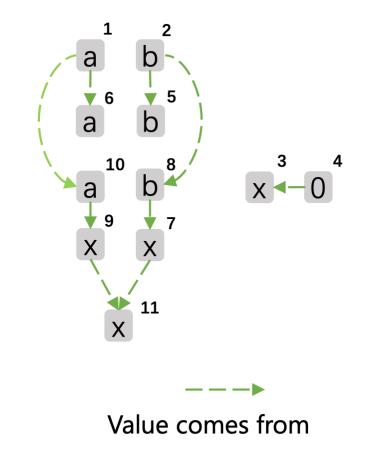
Abstract Syntax Tree

1	<pre>def max(b, a):</pre>
2	x = 0
3	if b > a:
4	x = b
5	else:
6	x = a
7	return x

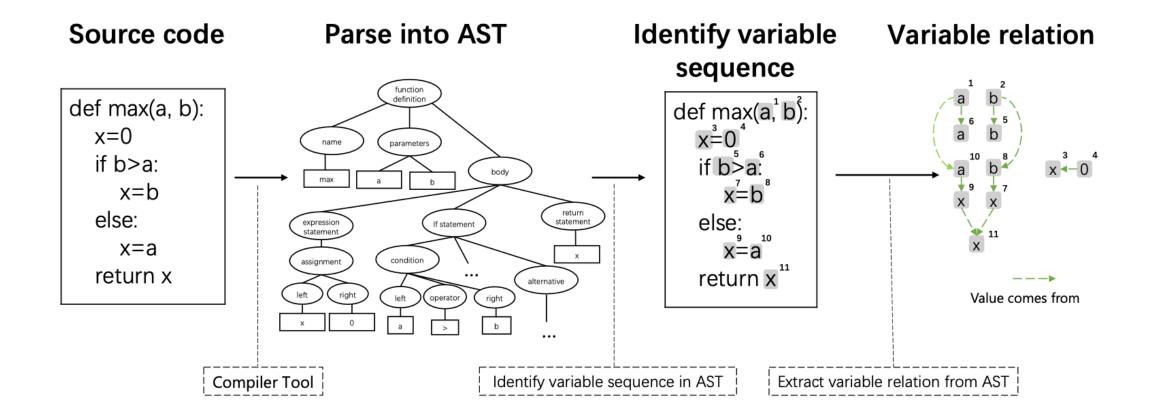


Data Flow

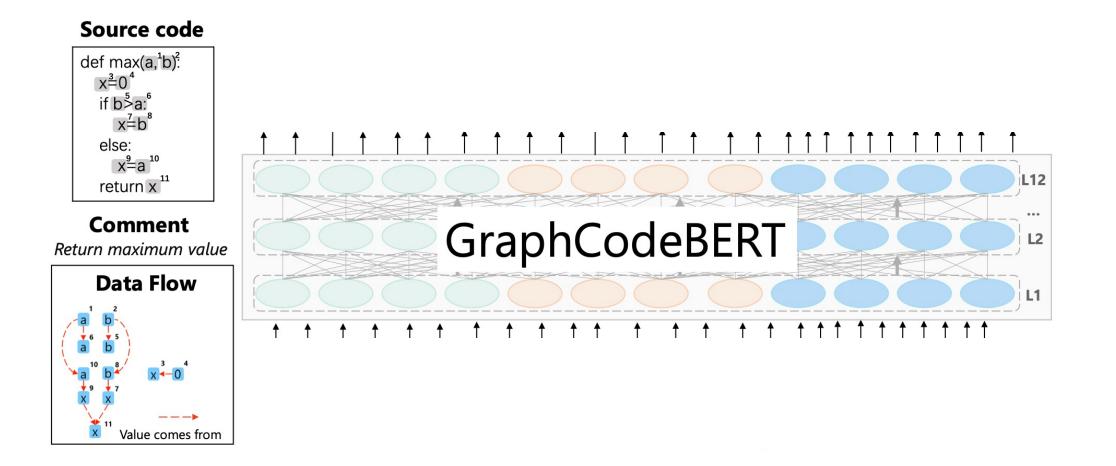




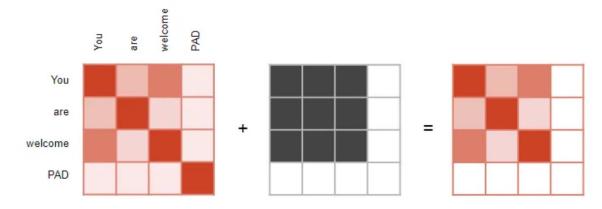
Data Flow Derivation



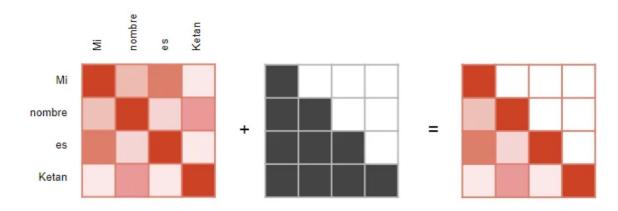
GraphCodeBERT architecture



Attention Mask



Encoder Self-Attention Scores



Ketan Doshi: https://towardsdatascience.com/transformers-explained-visually-part-2-how-it-works-step-by-step-b49fa4a64f34

PreTraining Tasks

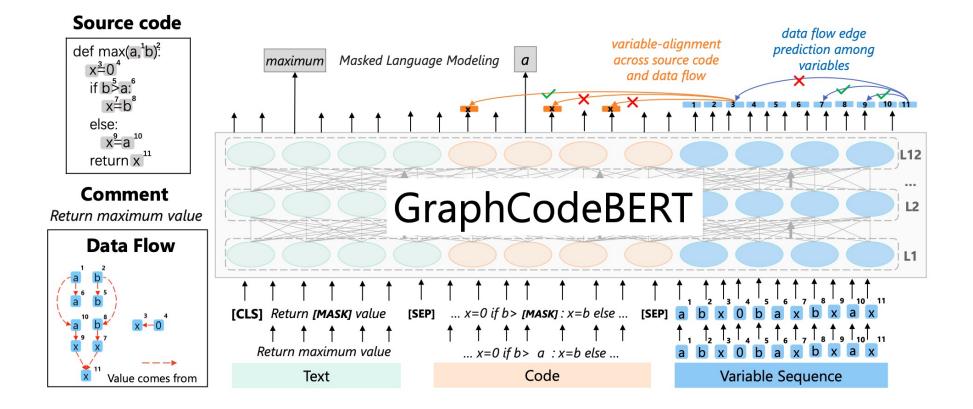
- Masked Language Modeling
- Edge Prediction
- Node Alignment

Masked Language Modelling

- Common way of pretraining
- Follows BERT approach

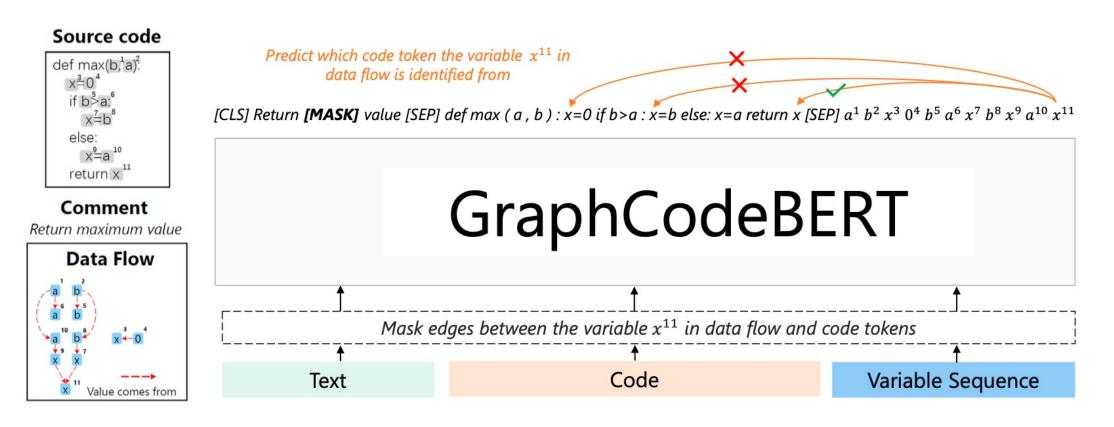
1	
1	def max(a, b):
2	"""Return [MASK] value"""
3	x = 0
4	if $b > a$:
5	x = b
б	else:
7	x = a
8	return x

Edge Prediction Pretraining



$$loss_{EdgePred} = -\sum_{e_{ij} \in E_c} [\delta(e_{ij} \in E_{mask}) logp_{e_{ij}} + (1 - \delta(e_{ij} \in E_{mask})) log(1 - p_{e_{ij}})]$$

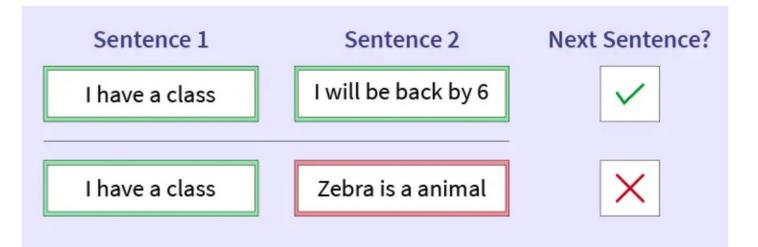
Node Alignment Pretraining



$$loss_{NodeAlign} = -\sum_{e_{ij} \in E_{c}^{'}} [\delta(e_{ij} \in E_{mask}^{'}) logp_{e_{ij}} + (1 - \delta(e_{ij} \in E_{mask}^{'})) log(1 - p_{e_{ij}})]$$

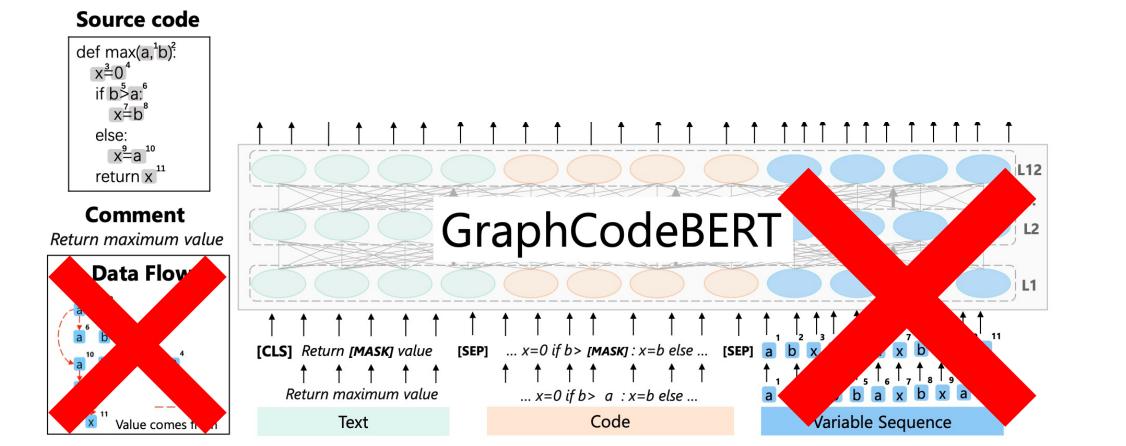
RoBERTa

- Train longer, bigger batches and more data
- Remove NSP
- Train on longer sequences
- Dynamically change masking pattern

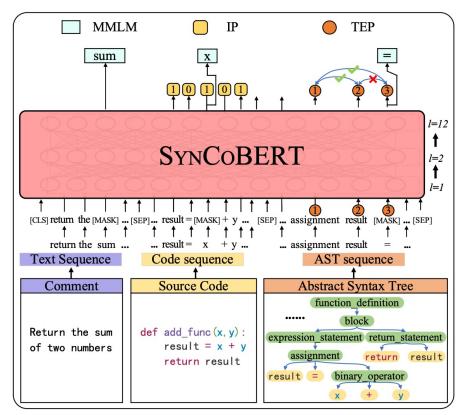


Scaler Topics: https://www.scaler.com/topics/nlp/bert-next-sentence-prediction/

CodeBERT



SynCoBERT



(a) SYNCOBERT pre-training over MMLM, IP and TEP objectives

SynCoBERT paper: https://arxiv.org/abs/2108.04556

MRR Score

$$\mathrm{MRR} = rac{1}{|Q|} \sum_{i=1}^{|Q|} rac{1}{\mathrm{rank}_i}.$$

Query	Proposed Results	Correct response	Rank	Reciprocal rank
cat	catten, cati, cats	cats	3	1/3

NL Code Search Results

model	Ruby	Javascript	Go	Python	Java	Php	Overall
NBow	0.162	0.157	0.330	0.161	0.171	0.152	0.189
CNN	0.276	0.224	0.680	0.242	0.263	0.260	0.324
BiRNN	0.213	0.193	0.688	0.290	0.304	0.338	0.338
selfAtt	0.275	0.287	0.723	0.398	0.404	0.426	0.419
RoBERTa	0.587	0.517	0.850	0.587	0.599	0.560	0.617
RoBERTa (code)	0.628	0.562	0.859	0.610	0.620	0.579	0.643
CodeBERT	0.679	0.620	0.882	0.672	0.676	0.628	0.693
GraphCodeBERT	0.703	0.644	0.897	0.692	0.691	0.649	0.713

SynCoBERT NL Code Search Results

Model	AdvTest			С	odeSearch			
WIUUCI	Python	Ruby	Javascript	Go	Python	Java	PHP	Average
NBow	-	16.2	15.7	33.0	16.1	17.1	15.2	18.9
CNN	-	27.6	22.4	68.0	24.2	26.3	26.0	32.4
BiRNN	-	21.3	19.3	68.8	29.0	30.4	33.8	33.8
Transformer	-	27.5	28.7	72.3	39.8	40.4	42.6	41.9
RoBERTa	18.3	58.7	51.7	85.0	58.7	59.9	56.0	61.7
RoBERTa (code)	-	62.8	56.2	85.9	61.0	62.0	57.9	64.3
CodeBERT	27.2	67.9	62.0	88.2	67.2	67.6	62.8	69.3
GraphCodeBERT	35.2	70.3	64.4	89.7	69.2	69.1	64.9	71.3
SYNCOBERT	38.1	72.2	67.7	91.3	72.4	72.3	67.8	74.0

Ablation Study

Methods	Ruby	Javascript	Go	Python	Java	Php	Overall
GraphCodeBERT	0.703	0.644	0.897	0.692	0.691	0.649	0.713
-w/o EdgePred	0.701	0.632	0.894	0.687	0.688	0.640	0.707
-w/o NodeAlign	0.685	0.635	0.887	0.682	0.690	0.640	0.703
-w/o Data Flow	0.679	0.620	0.882	0.672	0.676	0.628	0.693

Code Clone Detection

- Measure similarity between two code segments
- Code segments have similar output for the same input
- Easier software maintenance and to prevent bugs
- BigCloneBench dataset

```
protected String downloadURLtoString(URL url) throws IOException {
 1
        BufferedReader in = new BufferedReader(new InputStreamReader(url.openStream()));
 2
        StringBuffer sb = new StringBuffer(100 * 1024);
 3
        String str;
 4
        while ((str = in.readLine()) != null) {
 5
            sb.append(str);
 6
 7
        in.close();
 8
        return sb.toString();
 9
10 }
```

```
BLEU Score: 0.983
```

```
public static String fetchUrl(String urlString) {
 1
        try {
 2
            URL url = new URL(urlString);
 3
            BufferedReader reader = new BufferedReader(new InputStreamReader(url.openStream()));
 4
            String line = null;
 5
            StringBuilder builder = new StringBuilder();
 6
            while ((line = reader.readLine()) != null) {
 7
                builder.append(line);
 8
            }
 9
            reader.close();
10
            return builder.toString();
11
        } catch (MalformedURLException e) {
12
        } catch (IOException e) {
13
14
        return "";
15
16 }
```

Clone Detection Results

Model	Precision	Recall	F1
Deckard	0.93	0.02	0.03
RtvNN	0.95	0.01	0.01
CDLH	0.92	0.74	0.82
ASTNN	0.92	0.94	0.93
FA-AST-GMN	0.96	0.94	0.95
RoBERTa (code)	0.949	0.922	0.935
CodeBERT	0.947	0.934	0.941
GraphCodeBERT	0.948	0.952	0.950

Code Translation

- Migrate code to different language
- Mostly used for legacy software
- BLEU score
- Dataset crawled from open source projects

BLEU Score

- Bilingual evaluation understudy
- Measure for similarity of machine translated text
- Based on frequency of shared words and phrases
- Frequencies are compared with reference corpus

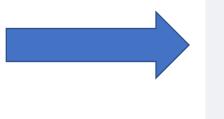
The closer a machine translation is to a professional human translation, the better it is

BLEU Paper: https://aclanthology.org/P02-1040.pdf

Java

C#

1	<pre>public void print(boolean b){</pre>
2	<pre>print(String.value0f(b));</pre>
3	}



1

public void print(bool b){ print(b.toString()); 2 3 }

Code Translation Results

Method	Java→C#		C#→Java	
	BLEU	Acc	BLEU	Acc
Naive	18.54	0.0	18.69	0.0
PBSMT	43.53	12.5	40.06	16.1
Transformer	55.84	33.0	50.47	37.9
RoBERTa (code)	77.46	56.1	71.99	57.9
CodeBERT	79.92	59.0	72.14	58.8
GraphCodeBERT	80.58	59.4	72.64	58.8

Code Refinement

- Automatically fix bugs
- Reduces cost of bug fixes
- Dataset by Tufano et al.

Code Refinment Results

Method	small		medium	
	BLEU	Acc	BLEU	Acc
Naive	78.06	0.0	90.91	0.0
LSTM	76.76	10.0	72.08	2.5
Transformer	77.21	14.7	89.25	3.7
RoBERTa (code)	77.30	15.9	90.07	4.1
CodeBERT	77.42	16.4	91.07	5.2
GraphCodeBERT	80.02	17.3	91.31	9.1

Conclusion/Remarks

- Small improvement compared to CodeBERT
- SynCoBERT outperformed the model
- Not clear benefits of Data Flow when compared with AST
- BLEU score is not really justified