CodeBPE: Investigating Subtokenization Options for Large Language Models TLDR: carefully choosing Pretrained on Source Code tokenization in LLMs

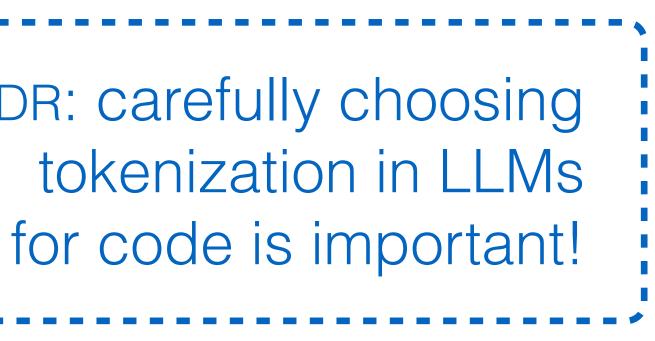


Nadezhda Chirkova Naver Labs Europe*



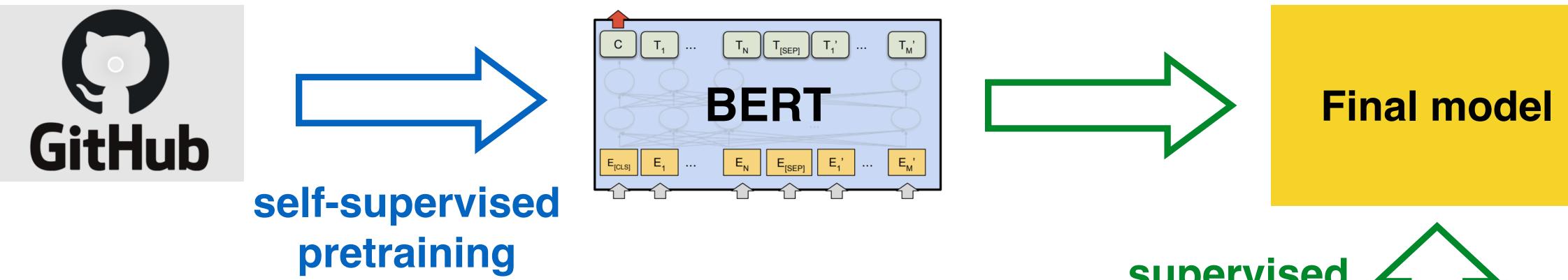
Sergey Troshin University of Amsterdam*

* Work done at HSE University





Pretrained language models (PLMs)



- Learn domain specifics from large code corpora during pretraining
- Often outperform models developed specifically for applied tasks



Task-specific data

PLM pipeline

FreqLists = [[0, 0] for i in range(vocSz)] **J** subtoken segmentation Freq List s = [[0, 0] for i in range (vo c S z)]

Feed Forward

Inputs

Positional Encodina

Masked Multi-Head Attention

Output Embedding

Outputs

(shifted right



Masked Language Modeling objective (or other)

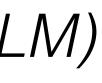


Task-specific objective

Research on PLMs for Code

- **Existing work:** investigation of models and pretraining objectives: CodeBERT (Feng20), CuBERT (Kanade20)
- CodeGPT (Lu21) \bullet
- PLBART (Ahmad21)
- GraphCodeBERT (Guo21) with data flow prediction objective CodeT5 (Wang21b) with variable naming and identifier tagging objectives
- \bullet DOBF (Roziere21) with variable naming objective

Our work considers another dimension: *subtokenization options* (e. g. BPE vocabulary size or BPE vs UnigramLM)



Overview

Main goals:

- \bullet
- \bullet

Considered options:

- Subtokenization granularity ullet
- UnigramLM vs BPE \bullet
- Vocabulary size
- Transferability between lacksquareprogramming languages

Methodology: start from UnigramLM (50k vocab) and add one modification at a time All experiments with PLBART

choose the most effective subtokenization (maximize downstream performance) choose the most length-efficient subtokenization without downstream performance drop

Downstream tasks:

- Code translation
- Code summarization
- Code generation
- Clone detection

Various levels of including spaces and punctuation in tokens:

Level	Example
0	['for','i','in','range', 'print','(','i',')','NEW
1	['for', 'i', 'in', 'range 'print', '(', 'i', ') NEW _
2	['for', 'i', 'in', 'range' '(', 'i', ')
3	['for i in range' , '(df' , '. sh 'print', '(df', '. column',
4	['for i in range', '(df', 'NEW_LINE print', '(df',

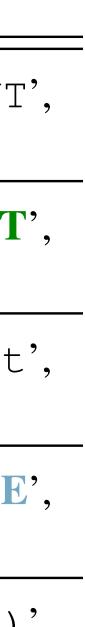
', '(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', W_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', _LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', '])']

', '(', 'df', '**.shape**', '[', '1', ']):', 'NEW_LINE INDENT', 'print', print', '(', 'df', '.columns', '[', 'i', '])']

hape [1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', ['], **'S [i**['], **']**)[']]

". shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', , **'. columns'**, **'[i]**)']



Various levels of including spaces and punctuation in tokens:

	Level	Example
conventionally used	0	['for','i','in','range', 'print','(','i',')','NEW_
	1	['for', 'i', 'in', 'range' 'print', '(', 'i', ')
	2	['for', 'i', 'in', 'range', '(', 'i', ') NEW_LINE ', 'p:
	3	['for i in range' , '(df', '. sha ' print', '(df', '. column' ,
	4	['for i in range', '(df', ', 'NEW_LINE print', '(df', '

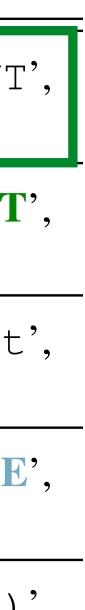
'(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', N_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', **LINE**', 'print', '(', 'df', '.', 'columns', '[', 'i', '])']

', '(', 'df', '.shape', '[', '1', ']):', 'NEW_LINE INDENT', 'print', print', '(', 'df', '.columns', '[', 'i', '])']

ape [1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', **'s**[**i**', **'**])']

". shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', **'. columns'**, **'[i]**)']



Various levels of including spaces and punctuation in tokens:

	Level	Example
allow	0	['for', 'i', 'in', 'range', 'print', '(', 'i', ')', 'NEW
merging of punctuation	1	['for', 'i', 'in', 'range' 'print', '(', 'i', ') NEW _
chars	2	['for', 'i', 'in', 'range', '(', 'i', ')
	3	['for i in range', '(df', '. sha 'print', '(df', '. column',
	4	['for i in range', '(df', ' 'NEW_LINE print', '(df',

, '(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', W_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', _LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', '])'] ←

', '(', 'df', '.shape', '[', '1', ']):', 'NEW_LINE INDENT', 'print', print', '(', 'df', '.columns', '[', 'i', '])']

hape [1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', , **'S [i**', **'])'**]

". shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', **'. columns'**, **'[i]**)']



Various levels of including spaces and punctuation in tokens:

	Level	Example
+ allow merging of dots with text	0	['for', 'i', 'in', 'range', 'print', '(', 'i', ')', 'NEW
	1	['for', 'i', 'in', 'range' 'print', '(', 'i', ') NEW _
	2	['for', 'i', 'in', 'range', '(', 'i', ')
	3	['for i in range ', '(df', '. sha 'print', '(df', '. column',
	4	['for i in range', '(df', ' 'NEW LINE print' '(df'

, '(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', W_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', _LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', '])']

', '(', 'df', '.shape', '[', '1', ']):', 'NEW_LINE INDENT', 'print', print', '(', 'df', '.columns', '[', 'i', '])']

nape [1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', **'S**[**i**', **'**])']

'. shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', 'NEW_LINE print', '(df', '. columns', '[i])']



Various levels of including spaces and punctuation in tokens:

	Level	Example
	0	['for', 'i', 'in', 'range', 'print', '(', 'i', ')', 'NEW
	1	['for', 'i', 'in', 'range' 'print', '(', 'i', ') NEW _
	2	['for', 'i', 'in', 'range', '(', 'i', ') NEW_LINE ', 'p
allow spaces inside tokens	3	['for i in range ', '(df', '. sha 'print', '(df', '. column' ,
	4	['for i in range', '(df', ' 'NEW LINE print' '(df'

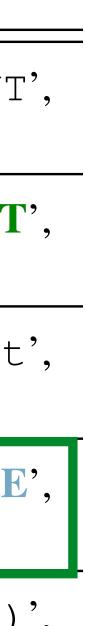
'(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', W_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', _LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', '])']

', '(', 'df', '**.shape**', '[', '1', ']):', 'NEW_LINE INDENT', 'print', >rint', '(', 'df', '.columns', '[', 'i', '])']

ape[1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', **'s**[**i**', **'**])']

". shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', **'NEW_LINE print'**, **'(df'**, **'. columns'**, **'[i]**)']



Various levels of including spaces and punctuation in tokens:

Level	Example
0	['for','i','in','range', 'print','(','i',')','NEW
1	['for', 'i', 'in', 'range' 'print', '(', 'i', ') NEW _
2	['for', 'i', 'in', 'range', '(', 'i', ')
3	['for i in range ', '(df', '. sha 'print', '(df', '. column' ,
4	['for i in range', '(df', ' 'NEW_LINE print', '(df',

allow new lines and; inside tokens

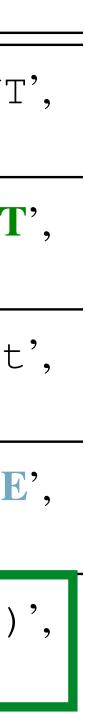
'(', 'df', '.', 'shape', '[', '1', ']', ')', ':', 'NEW_LINE', 'INDENT', N_LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', ']', ')']

e', '(', 'df', '.', 'shape', '[', '1', ']):', 'NEW_LINE INDENT', [LINE', 'print', '(', 'df', '.', 'columns', '[', 'i', '])']

', '(', 'df', '**.shape**', '[', '1', ']):', 'NEW_LINE INDENT', 'print', print', '(', 'df', '.columns', '[', 'i', '])']

ape [1', ']):', 'NEW_LINE INDENT', 'print', '(i', ') NEW_LINE', **'s**[**i**', **'**])']

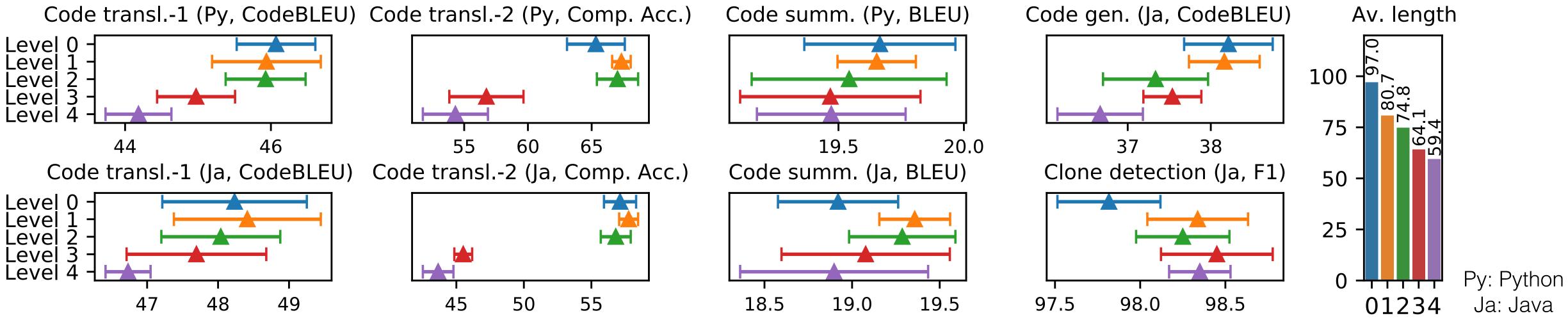
f. shape', '[1]', ')', ': NEW_LINE', 'INDENT print', '(i)', '. columns', **'[i]**)']



Subtokenization granularity Various levels of including spaces and punctuation in tokens:

Level	Example
0	['for', 'i', 'in', 'range', '(', 'df', '.', 'shape', '[', '1', ' '(', 'df', '.', 'columns', '[', 'i', ']', ')']
1	['for', 'i', 'in', 'range', '(', 'df', '.', 'shape', '[', '1', '] '.', 'columns', '[', 'i', '])']
2	<pre>['for', 'i', 'in', 'range', '(', 'df', '.shape', '[', '1', ']); '.columns', '[', 'i', '])']</pre>
3	['for i in range', '(df', '. shape [1', ']) :', 'NEW_LINE IND
4	['for i in range', '(df', '. shape', '[1]', ') ', ': NEW_LINE', '

Code transl.-1 (Py, CodeBLEU) Code transl.-2 (Py, Comp. Acc.)



Main conclusion: Level 1 compresses lengths by 17% without performance drop, comp. to Level 0

']', ')', ':', 'NEW_LINE', 'INDENT', 'print', '(', 'i', ')', 'NEW_LINE', 'print',

']):', 'NEW_LINE INDENT', 'print', '(', 'i', ') NEW_LINE', 'print', '(', 'df',

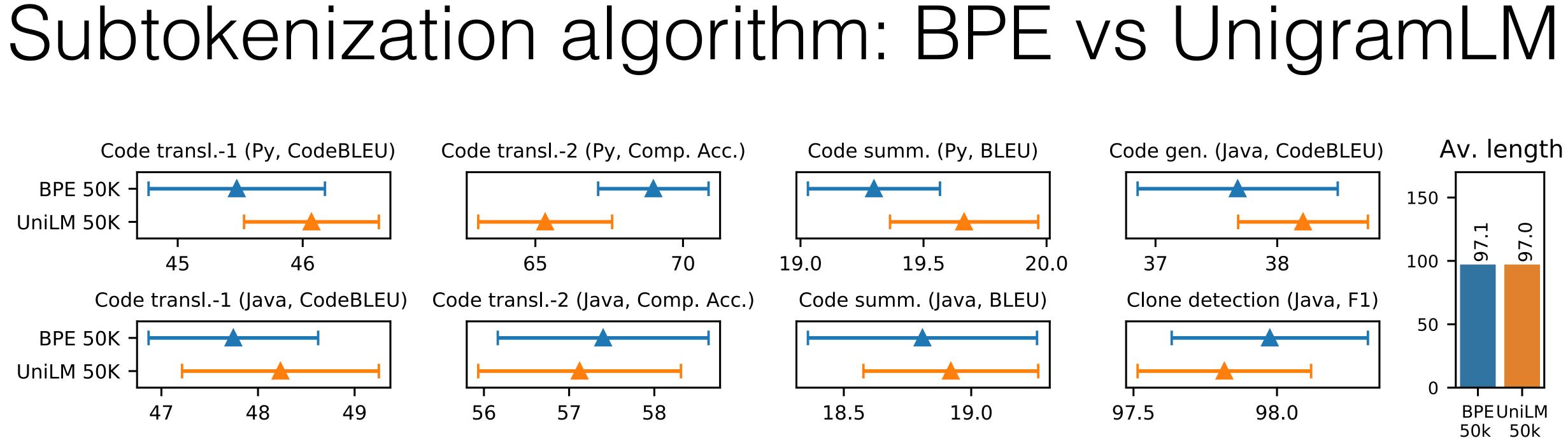
:', 'NEW_LINE INDENT', 'print', '(', 'i', ') NEW_LINE', 'print', '(', 'df',

DENT', 'print', '(i', ') **NEW_LINE**', 'print', '(df', '. column', 's[i', '])']

'INDENT print', '(i)', 'NEW_LINE print', '(df', '. columns', '[i])']







Main conclusion: UnigramLM slightly outperforms or performs on par with BPE in 7 tasks

Subtokenization algorithm: BPE vs UnigramLM

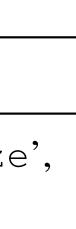
UnigramLM is better aligned with splitting identifiers by CamelCase and snake_case:

Original token	UnigramLM subtok-	BPE subtokenization	Native subtokenization
	enization		(Camel- or snake_case)
fromDottedString	['from', 'Dotted', 'String']	['from', 'Dot', 'ted', 'String']	['from','Dotted','String']
isInstantiated	['is','Instantiate','d']	['isIn', 'stanti', 'ated']	['is','Instantiated']
GridBagConverter	['Grid', 'Bag', 'Converter']	[' GridBag ', 'Converter']	['Grid', 'Bag', 'Converter']
isSameSize Horizontally	['isSame', 'Size', 'Horizontally']	['isSame', 'Size', ' H ', 'orizontally']	['is', 'Same', 'Size 'Horizontally']
PA_Hierarchy_ID	['PA', '_', 'Hierarchy', '_ID']	['PA', '_H', 'ierarchy', '_ID']	['PA', '_', 'Hierarchy', '_', 'ID']



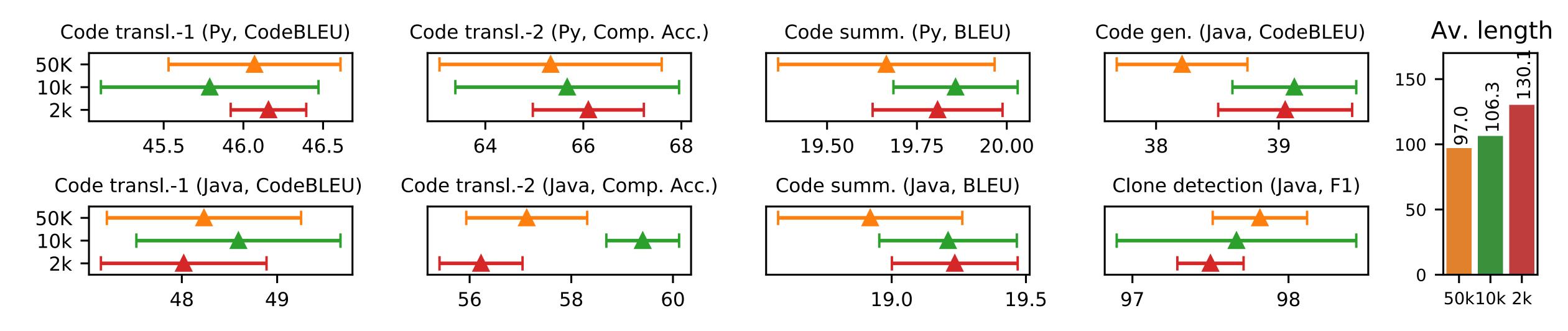






Vocabulary size

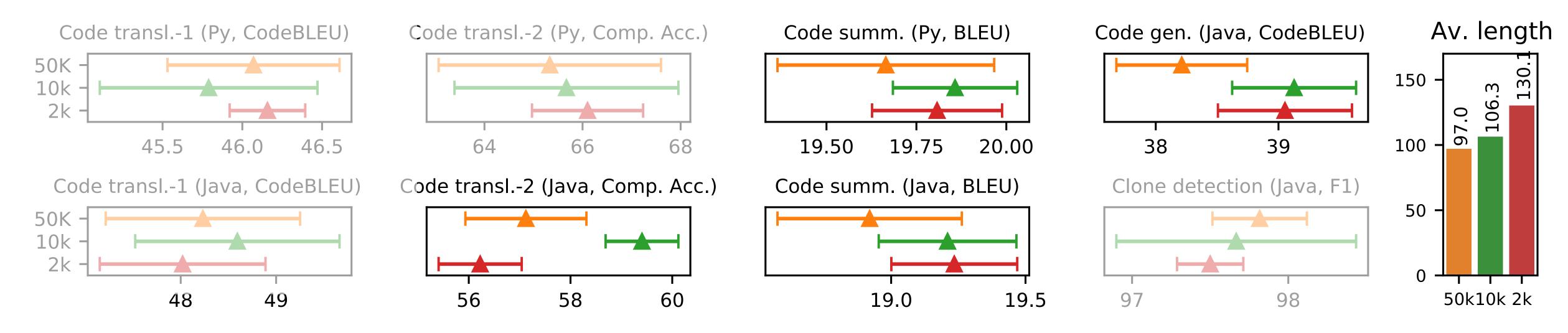
Unigram LM: **50k**, **10k**, **2k**



Main conclusion: **10k** > **50k** in 4 tasks, in other tasks similar performance

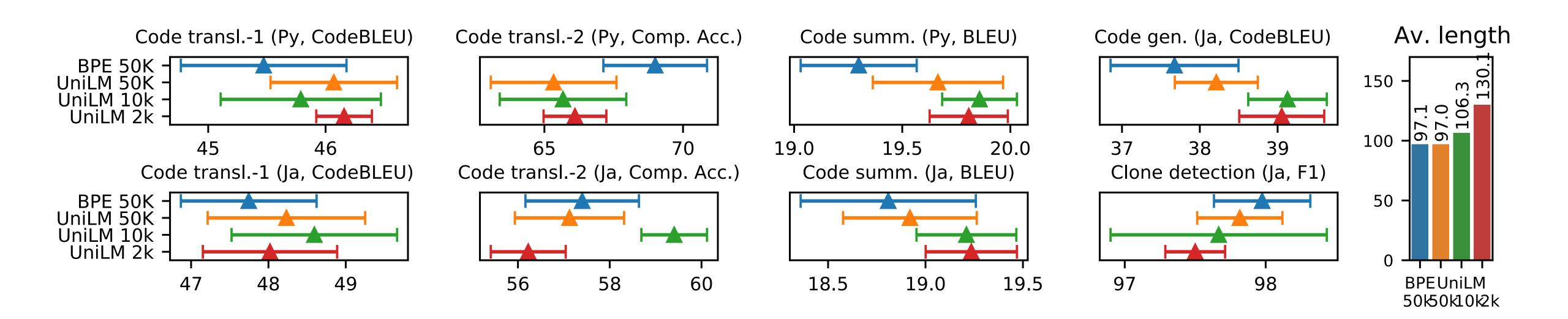
Vocabulary size

Unigram LM: **50k**, **10k**, **2k**



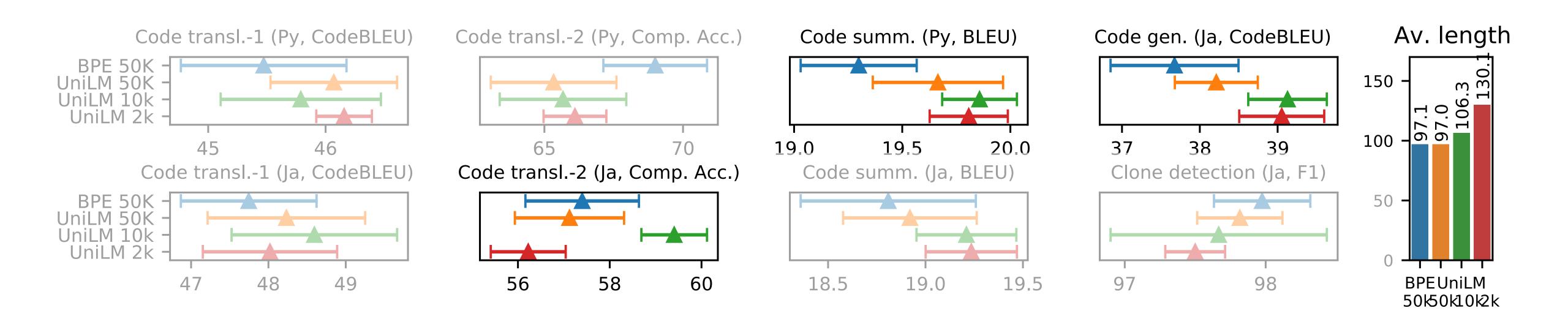
Main conclusion: **10k** > **50k** in 4 tasks, in other tasks similar performance

Subtokenization algorithm + vocabulary size



UnigramLM 10k > commonly used **BPE 50k** in 3 tasks substantially and 2 tasks by one std

Subtokenization algorithm + vocabulary size



UnigramLM 10k > commonly used **BPE 50k** in 3 tasks significantly and 2 tasks by one std

Summary

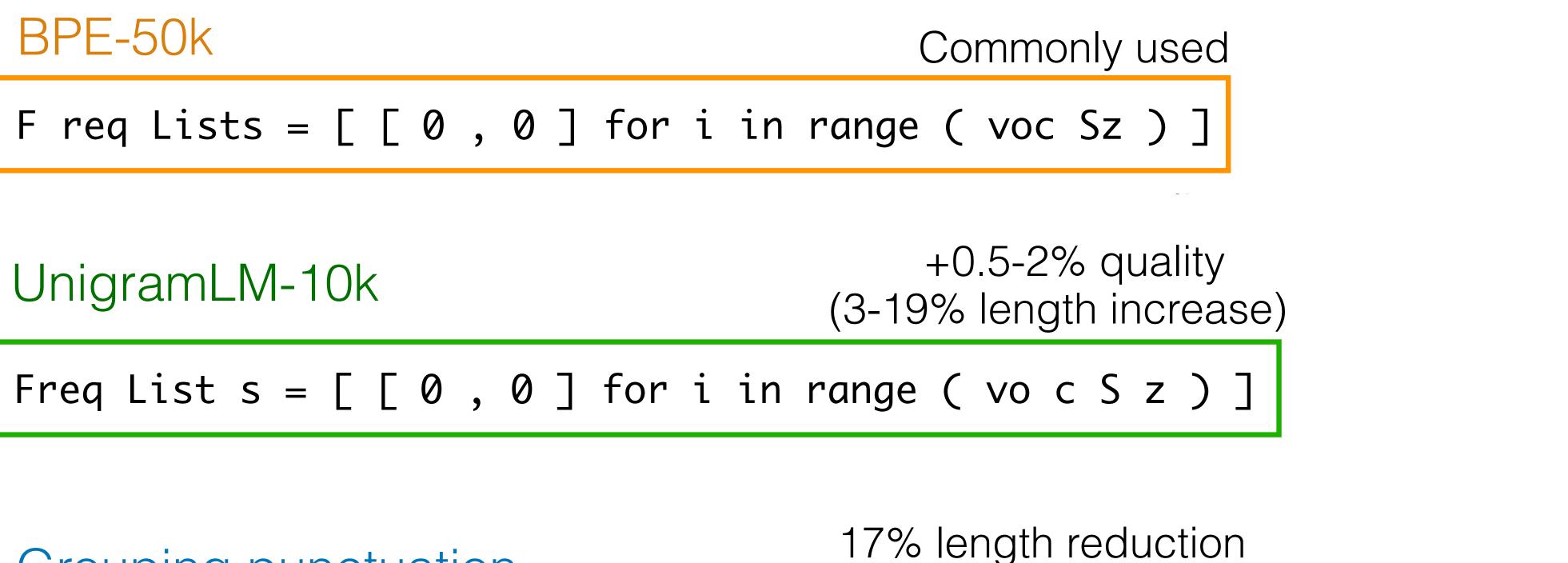
BPE-50k

F req Lists = [[0 , 0] for i in range (voc Sz)]

UnigramLM-10k

Grouping punctuation

Freq Lists = [[0, 0] for i in range (voc S z)]



without quality drop

CodeBPE /

CodeUnigramLM