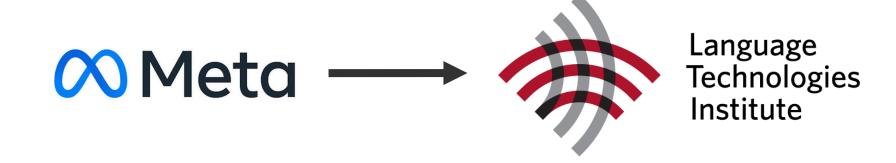
Contextual Communication in Programming

Daniel Fried



Today's Question

Are bigger models the solution for Al-assisted programming?

Posing This Question in 2012...

On the Naturalness of Software

Abram Hindle, Earl Barr, Mark Gabel, Zhendong Su, Prem Devanbu

[ICSE 2012; Most Influential Paper 2022]

Natural languages like English are rich, complex, and powerful. We begin with the conjecture that <u>most software is also natural</u>, in the sense that it is created by <u>humans at work</u>, with all the attendant constraints and limitations—and thus, like natural language, <u>it is also likely to be repetitive and predictable</u>. We then proceed to ask whether a) code can be usefully modeled by statistical language models and b) such models can be leveraged to support software engineers.

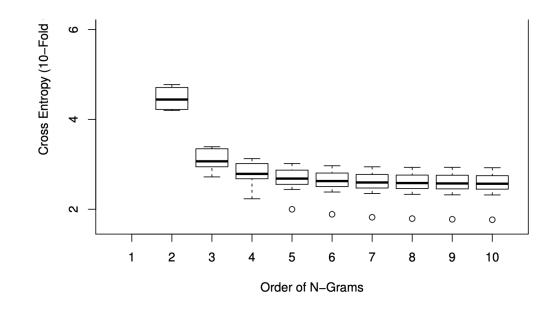
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On the Naturalness of Software

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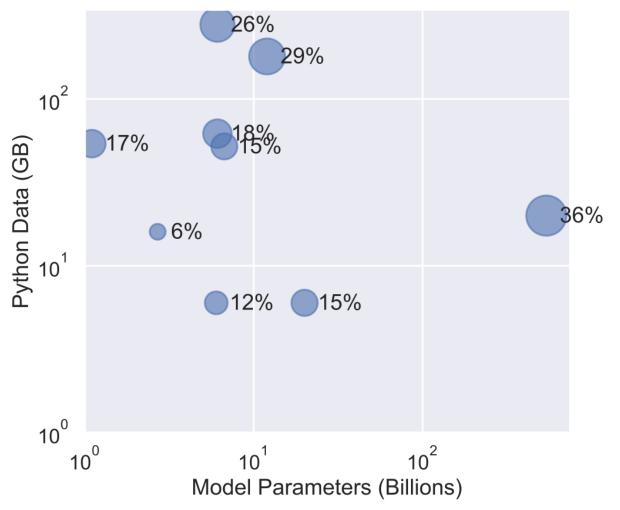
[ICSE 2012; Most Influential Paper 2022]

- ▶ n-gram models trained on ~25 million lines of code
- Substantial improvements to Eclipse's autocomplete
- ▶ But, 3-4 orders of magnitude less data than modern neural models



... and now

Function pass rate on a Python synthesis dataset [Chen et al. 2021] by data & model scale:



[Compiled from Chen et al. 2021, Xu et al. 2021, Li et al. 2021, Fried et al. 2022, Nijkamp et al. 2022, Chowdhery et al. 2022]

Today's Question

Are bigger models the solution for Al-assisted programming?

YES

Programming as Communication

We begin with the conjecture that most software... is created by humans at work, with all the attendant constraints and limitations

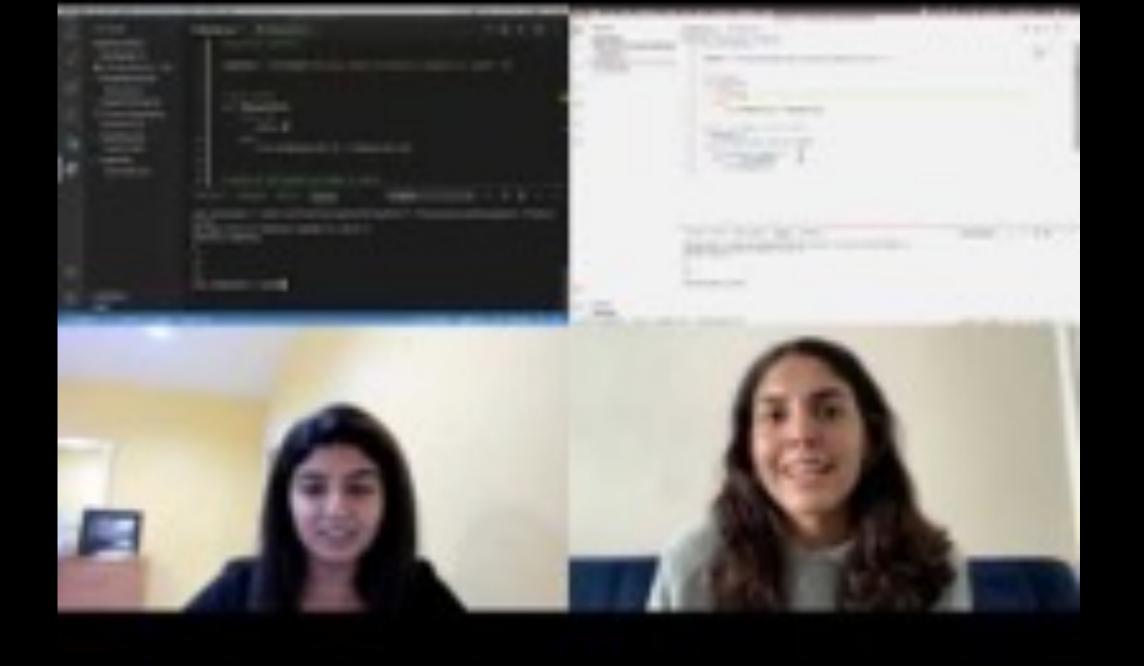
communicating with the compiler, other developers, and themselves,

and thus, like natural language,

it is also likely to be repetitive and predictable.

writing software is a form of contextual and interactive communication.

We then proceed to ask whether a) code can be usefully modeled by statistical language models and b) such models can be leveraged to support software engineers.



Communicating with Multiple Modalities

Natural Language Partial Code

Tests & Execution

Edits to Code

Deictic (Pointing / Highlighting)

As Inputs

[Fried & Aghajanyan et al., 2022]

[Fried & Aghajanyan et al., 2022]

[Shi et al. 2022]

As Outputs

[Fried & Aghajanyan et al., 2022]

[Wallace et al., in progress]

Modality Choice [Lin et al., 2022]

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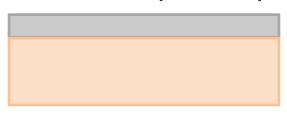
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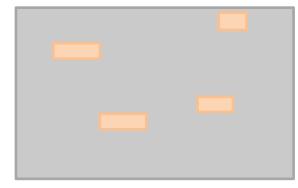
Modality Choice [Lin et al., 2022]

Neural Code Model Objectives

"Causal" (L-to-R)

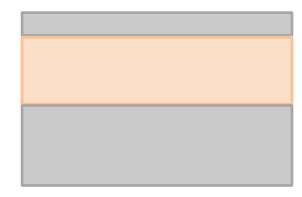


Masked Infilling



[e.g. BERT, CodeBERT]

Causal Masking



[Donahue+ 2020, Aghajanyan+ 2022, ours, Bavarian+ 2022]

InCoder: Code Generation and Infilling

Training

Original Document

Masked Document

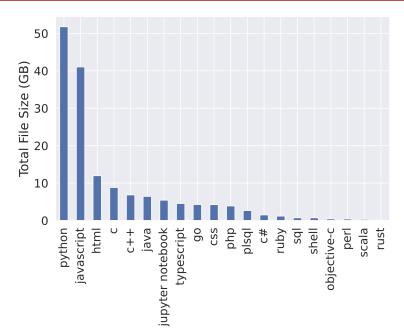
InCoder: Code Generation and Infilling

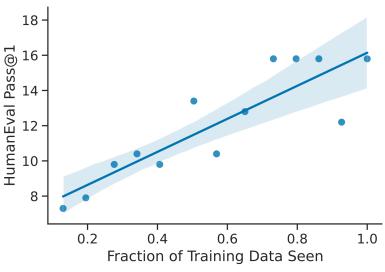
Data

- ▶ 600K permissively-licensed repositories from GitHub & GitLab
- StackOverflow: questions, answers, comments

Models

- Standard transformer LM
- ▶ 1B model: ~1 week on 128 GPUs
- ▶ 6B model: ~3 weeks on 240 GPUs





InCoder: Code Generation and Infilling

Zero-shot Inference

Docstring Generation

Multi-Region Infilling

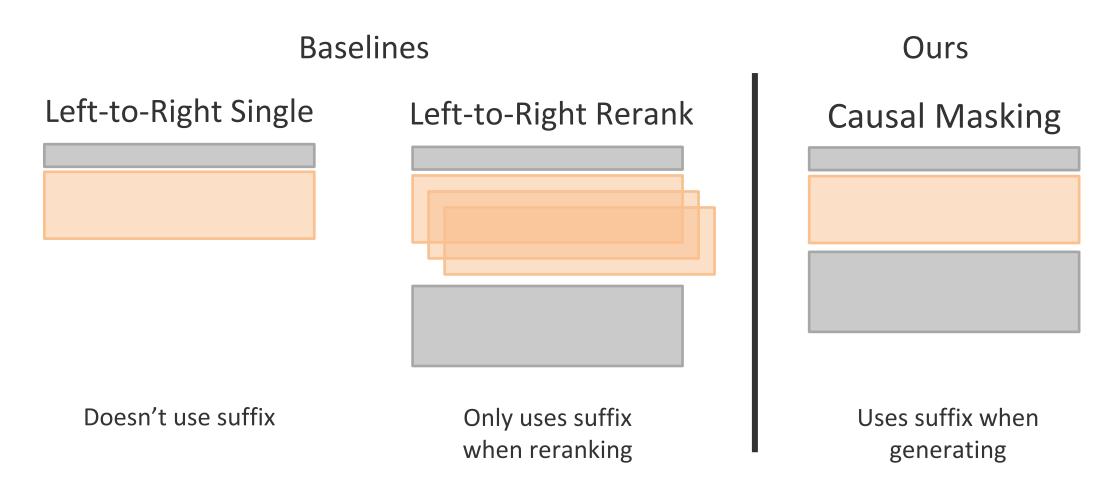
```
from collections import Counter

def word_count(file_name):
    """Count the number of occurrences of each word in the file."""
    words = []
    with open(file_name) as file:
        for line in file:
            words.append(line.strip())
    return Counter(words)
```

Also usable as a left-to-right generation model with no apparent loss in performance [see also Bavarian et al. 2022]

Evaluation

- Zero-shot evaluation on realistic code infilling tasks
- Compare the model in three different modes to evaluate benefits of suffix context



Evaluation

Docstring Generation

```
def count_words(filename: str) -> Dict[str, int]:
    0.00
    Counts the number of occurrences of each word in the given file.
    :param filename: The name of the file to count.
    :return: A dictionary mapping words to the number of occurrences.
    0.00
    with open(filename, 'r') as f:
           word counts = {}
           for line in f:
               for word in line.split():
                   if word in word_counts:
                       word_counts[word] += 1
                   else:
                       word counts[word] = 1
       return word_counts
```

Method	BLEU
Ours: L-R single Ours: L-R reranking Ours: Causal-masked infilling	16.05 17.14 18.27
RoBERTa (Finetuned) CodeBERT (Finetuned) PLBART (Finetuned) CodeT5 (Finetuned)	18.14 19.06 19.30 20.36

Evaluation

Type Inference

Method	F1
Ours: Left-to-right single Ours: Left-to-right reranking Ours: Causal-masked infilling	30.8 33.3 59.2
TypeWriter (Supervised)	48.3

Demo

```
Num Tokens:
Temperature:
 Extend
           Add <infill> mask
Syntax: Python
   1 <| file ext=.py |>
      from collections import Counter
      def <infill>
          """Count the number of occurrences of each word in the file."""
          <infill>
```

Communicating with Multiple Modalities

NL Instructions

Partial Code

Tests & Execution

Edits to Code

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• •

As Inputs

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[Shi et al. 2022]

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Modality Choice [Lin et al., 2022]

Using Test Inputs

Description:

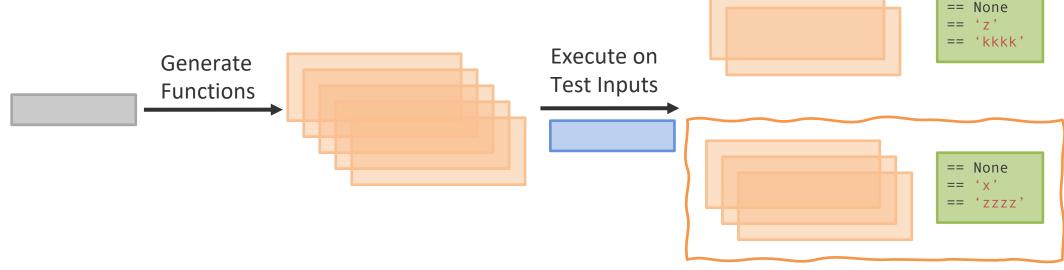
```
def longest(strings: List[str]) -> Optional[str]:
    """ Out of list of strings, return the longest one.
    Return the first one in case of multiple strings of
    the same length. Return None if the list is empty."""
```

Test Inputs:

```
longest([]) ==
longest(['x', 'y', 'z']) ==
longest(['x', 'yyy', 'zzzz', 'www', 'kkkk', 'abc']) == ____
```

Minimum Bayes Risk with Execution:

Cluster by Outputs



[Shi et al. 2022. See also AlphaCode, Li et al. 2022]

Other Features of Communication

Communicative cost

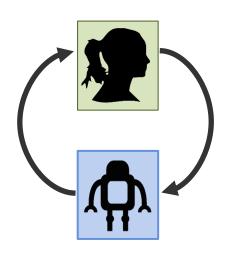
- Copilot outputs can be hard to understand [Vaithilingam et al. 2022]
- Would a user rather type a comment or edit code?

Resolving uncertainty

- Disambiguate by prompting with test inputs [Zhong et al. 2022]
- How to convey uncertainty to the user & build trust?

Adaptation

- Acceleration vs exploration modes for using Copilot [Barke et al. 2022]
- API preferences, functional vs imperative, design patterns, documentation style ...



Collaborators



Armen Aghajanyan



Anca Dragan



Marjan Ghazvininejad



Dan Klein



Mike Lewis



Jessy Lin



Freda Shi



Eric Wallace



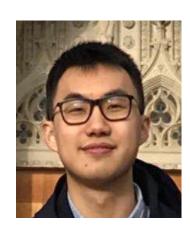
Sida Wang



Scott Yih



Luke Zettlemoyer



Ruiqi Zhong

Thanks!

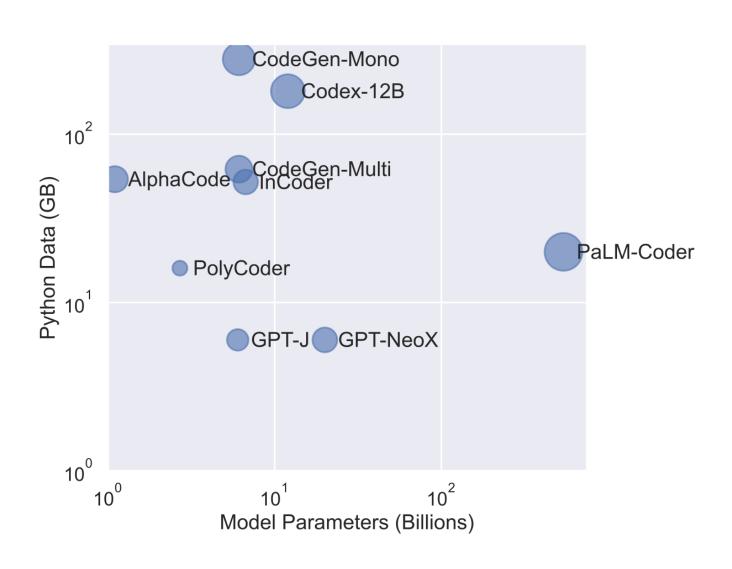
dfried@cs.cmu.edu
dpfried.github.io

Backup Slides

Scale and Performance

Model	Size (B)	Python Code (GB)	Other Code (GB)	Other (GB)	Code License	Infill?	HE @1	HE @10	HE @100	MBPP @1
Released	-						-			
CodeParrot [61]	1.5	50	None	None	_		4.0	8.7	17.9	
PolyCoder [68]	2.7	16	238	None	_		5.6	9.8	17.7	
GPT-J [63, 18]	6	6	90	730	_		11.6	15.7	27.7	_
InCoder-6.7B	6.7	52	107	57	Permissive	✓	15.2	27.8	47.0	19.4
GPT-NeoX [14]	20	6	90	730			15.4	25.6	41.2	_
CodeGen-Multi [46]	6.1	62	375	1200	-		18.2	28.7	44.9	
CodeGen-Mono [46]	6.1	279	375	1200	_		26.1	42.3	65.8	_
CodeGen-Mono [46]	16.1	279	375	1200	_		29.3	49.9	75.0	_
Unreleased										
LaMDA [10, 60, 21]	137	None	None	???	_		14.0		47.3	14.8
AlphaCode [44]	1.1	54	660	None			17.1	28.2	45.3	
Codex-12B [18]	12	180	None	>570			28.8	46.8	72.3	-
PaLM-Coder [21]	540	~20	~200	~4000	Permissive		36.0	_	88.4	47.0

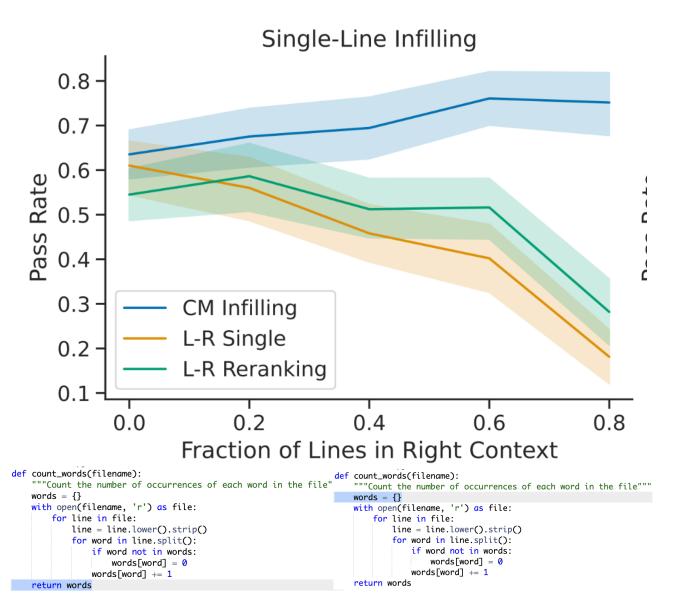
... and now



Details

- ▶ Remove duplicate files using exact match on alphanumeric token sequences
- ▶ Tokenization: retrain byte-level BPE, modified to allow merging across spaces
- e.g. import numpy as np is a single token
 - ▶ 40% reduction in token count compared to GPT-2's tokenizer
 - Longer effective contexts & more efficient training
 - ▶ But, less compute spent in training may affect performance
- Meta-data conditioning and prediction
- <| file source=github stars=high filename=setup.py ext=.py |>
- ▶ Attributes can appear at beginning of the file (conditioning) or end (prediction)

Function completion



Function completion

