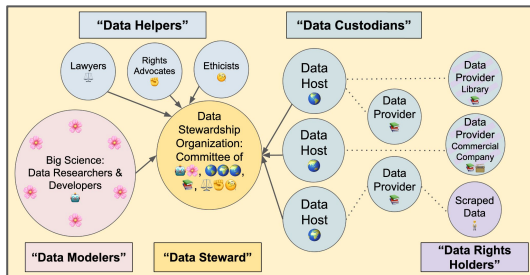


# Building Better Language Models: *Insights from BigScience* 🌸

Colin Raffel

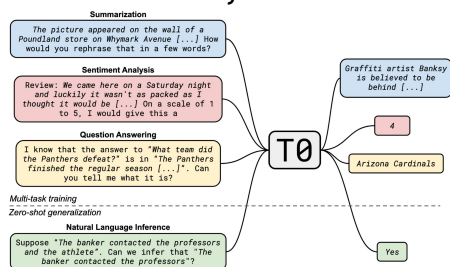
## Data governance



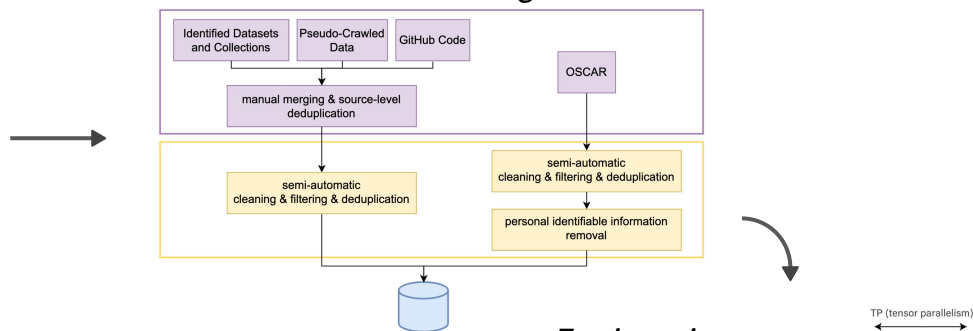
## Evaluation



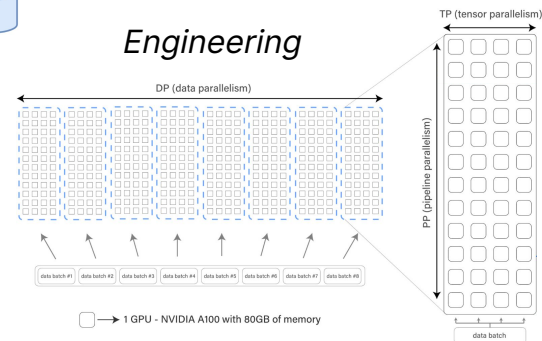
## Objective



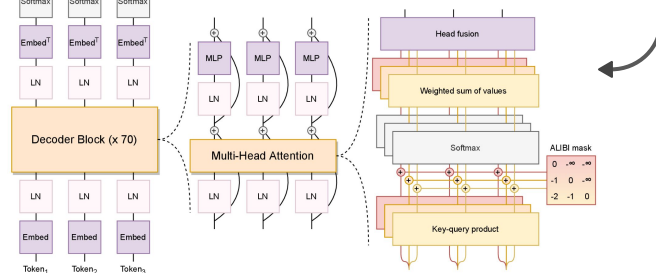
## Data sourcing

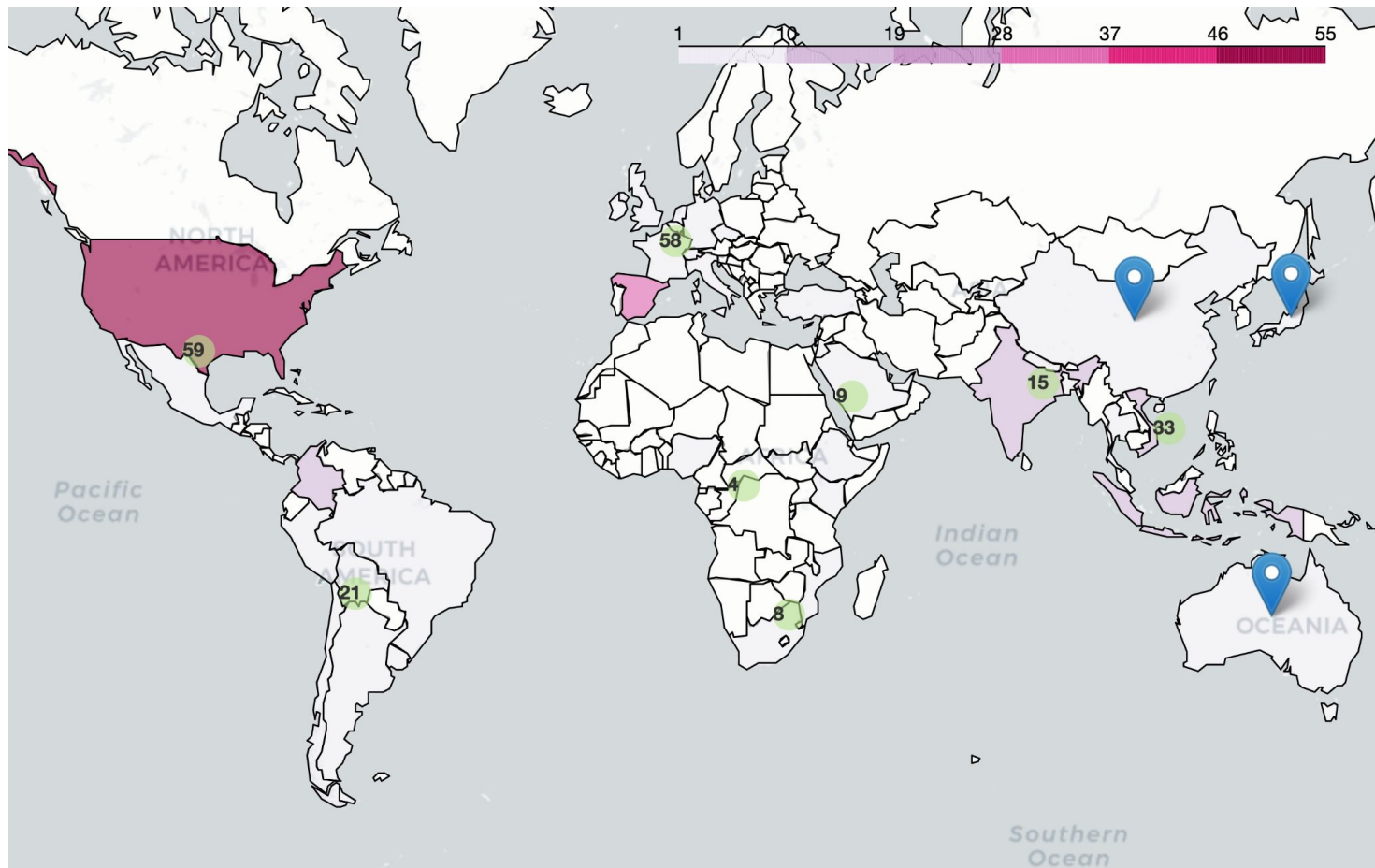


## Engineering



## Modeling

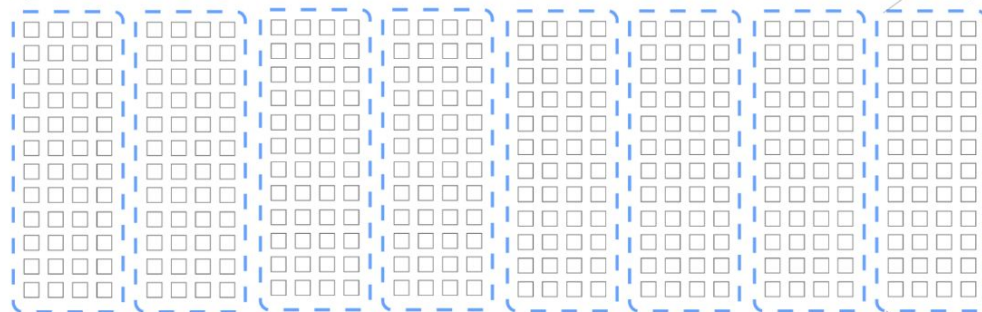




From <https://huggingface.co/spaces/bigscience/SourcingCatalog>

8 copies of the model are trained in parallel  
on a total of 384 GPUs (data parallelism = 8)

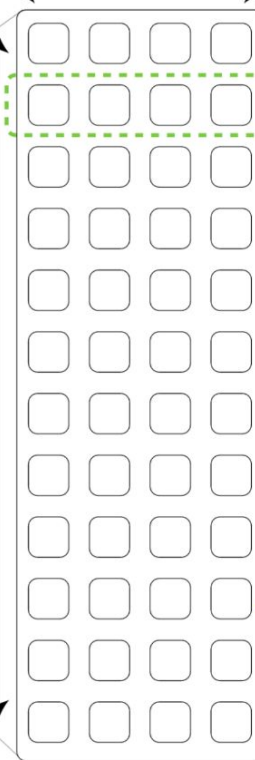
DP (data parallelism)



 → 1 GPU - NVIDIA A100 with 80GB of memory

TP (tensor parallelism)

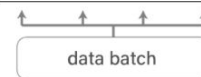
PP (pipeline parallelism)

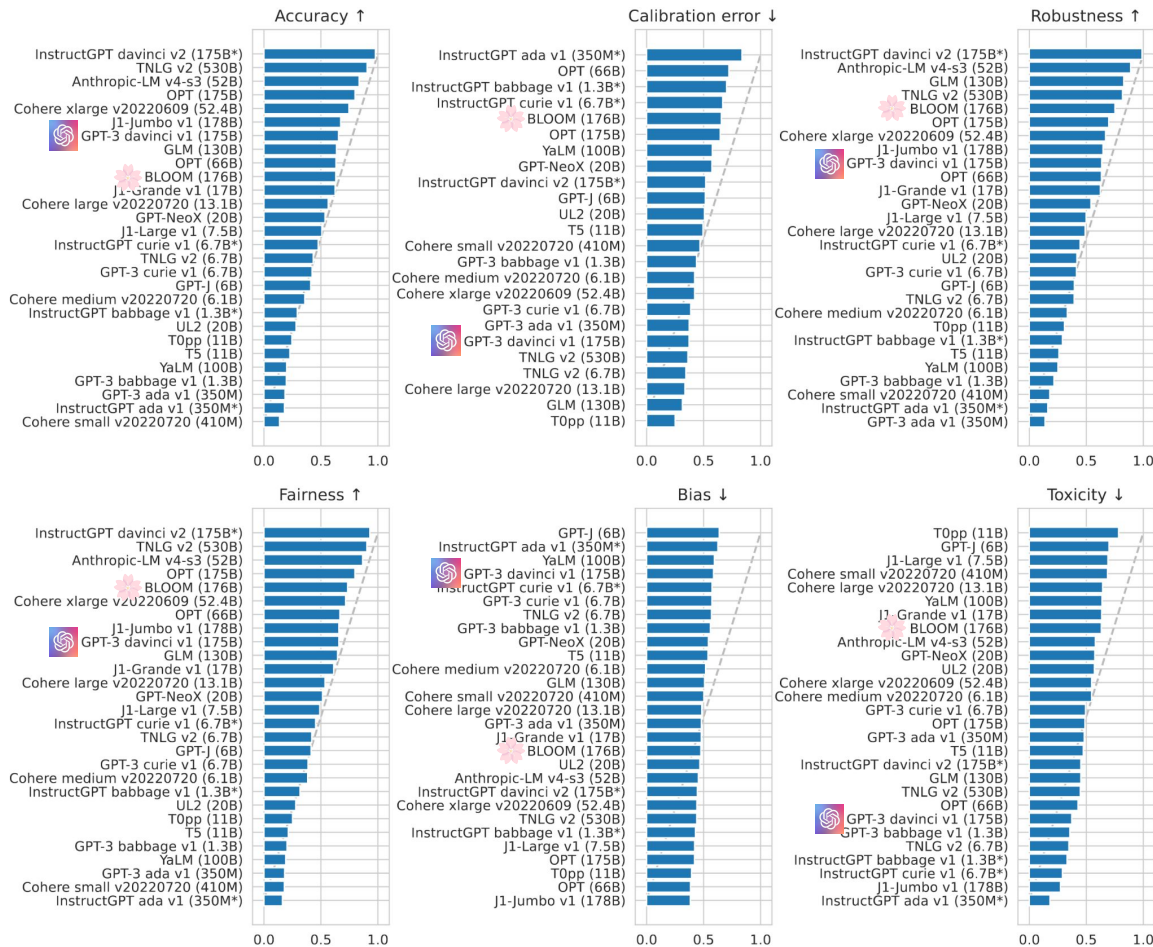


Model parameters  
are divided across 4 GPUs  
(tensor parallelism = 4)

The layers of the model  
are spread across  
12 groups of GPUs  
(pipeline parallelism = 12)

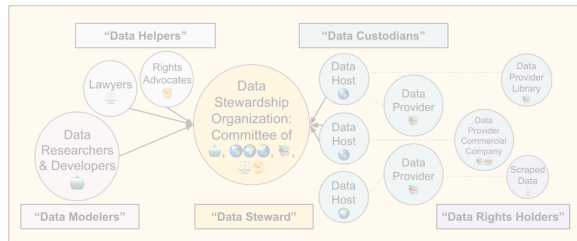
One full copy («replica»)  
of the model takes  
48 GPUs



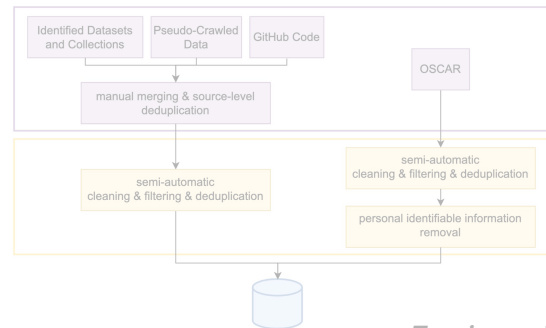


From "Holistic Evaluation of Language Models" by Liang et al.

## Data governance



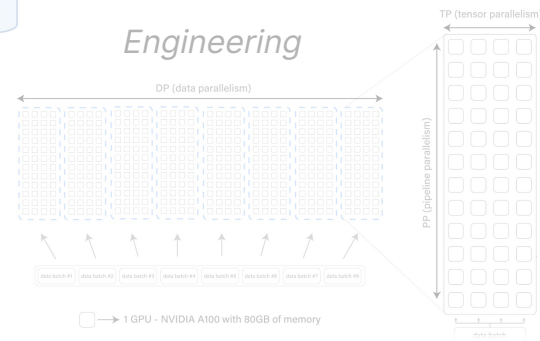
## Data sourcing



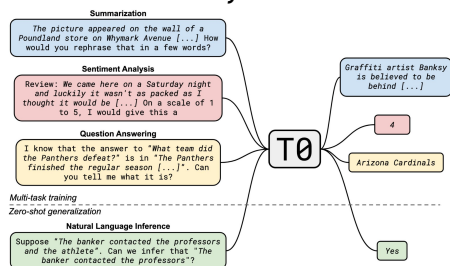
## Evaluation



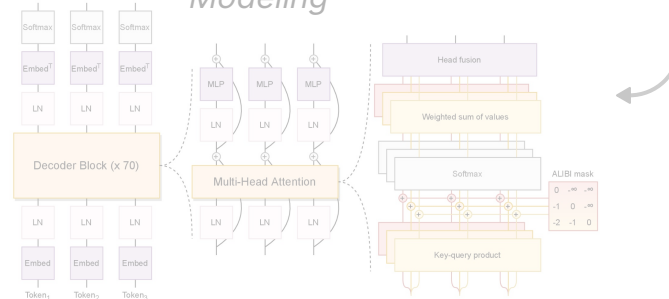
## Engineering



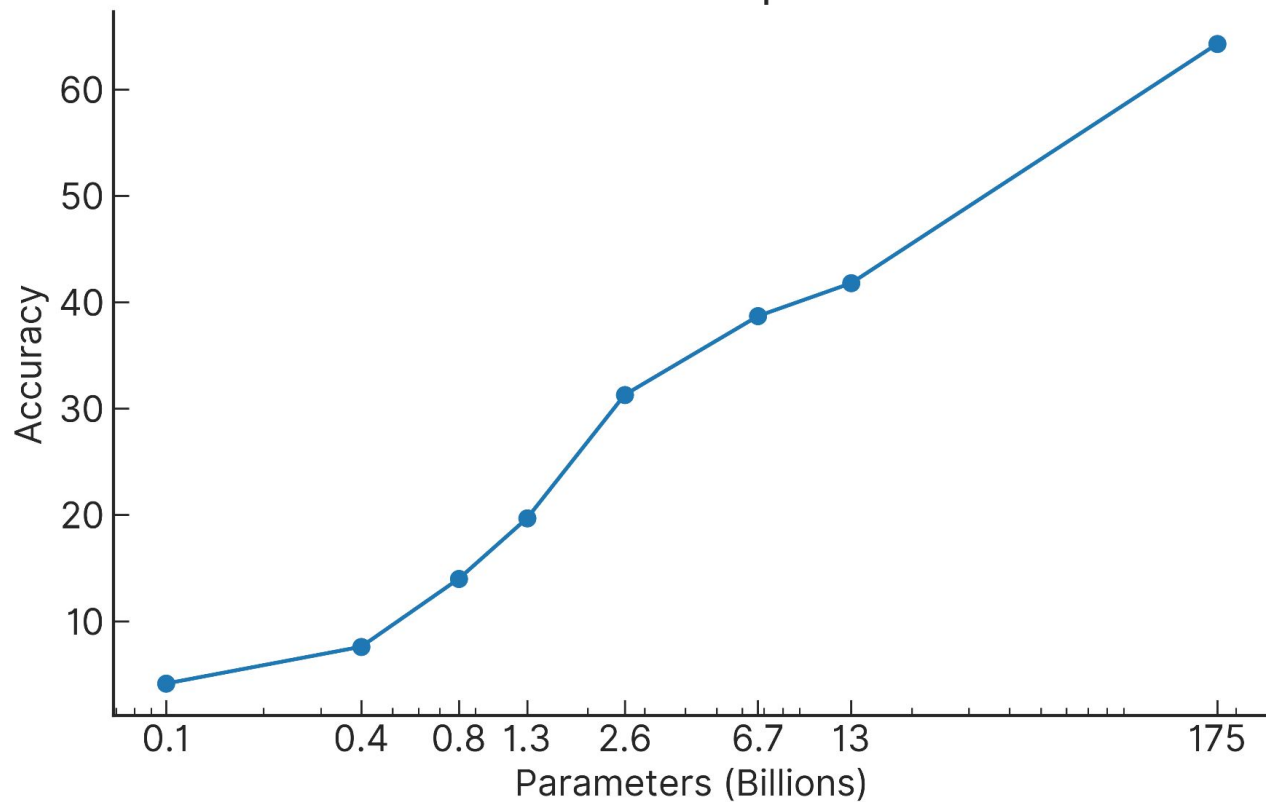
## Objective



## Modeling



## TriviaQA zero-shot performance



*From "Language Models are Few-Shot Learners" by Brown et al.*

### **Closed-book question answering**

<http://www.autosweblog.com/cat/trivia-questions-from-the-50s>

who was frank sinatra? a: an american singer, actor, and producer.

### **Paraphrase identification**

<https://www.usingenglish.com/forum/threads/60200-Do-these-sentences-mean-the-same>

Do these sentences mean the same? No other boy in this class is as smart as the boy. No other boy is as smart as the boy in this class.

### **Natural Language Inference**

<https://ell.stackexchange.com/questions/121446/what-does-this-sentence-imply>

If I say: He has worked there for 3 years. does this imply that he is still working at the moment of speaking?

### **Summarization**

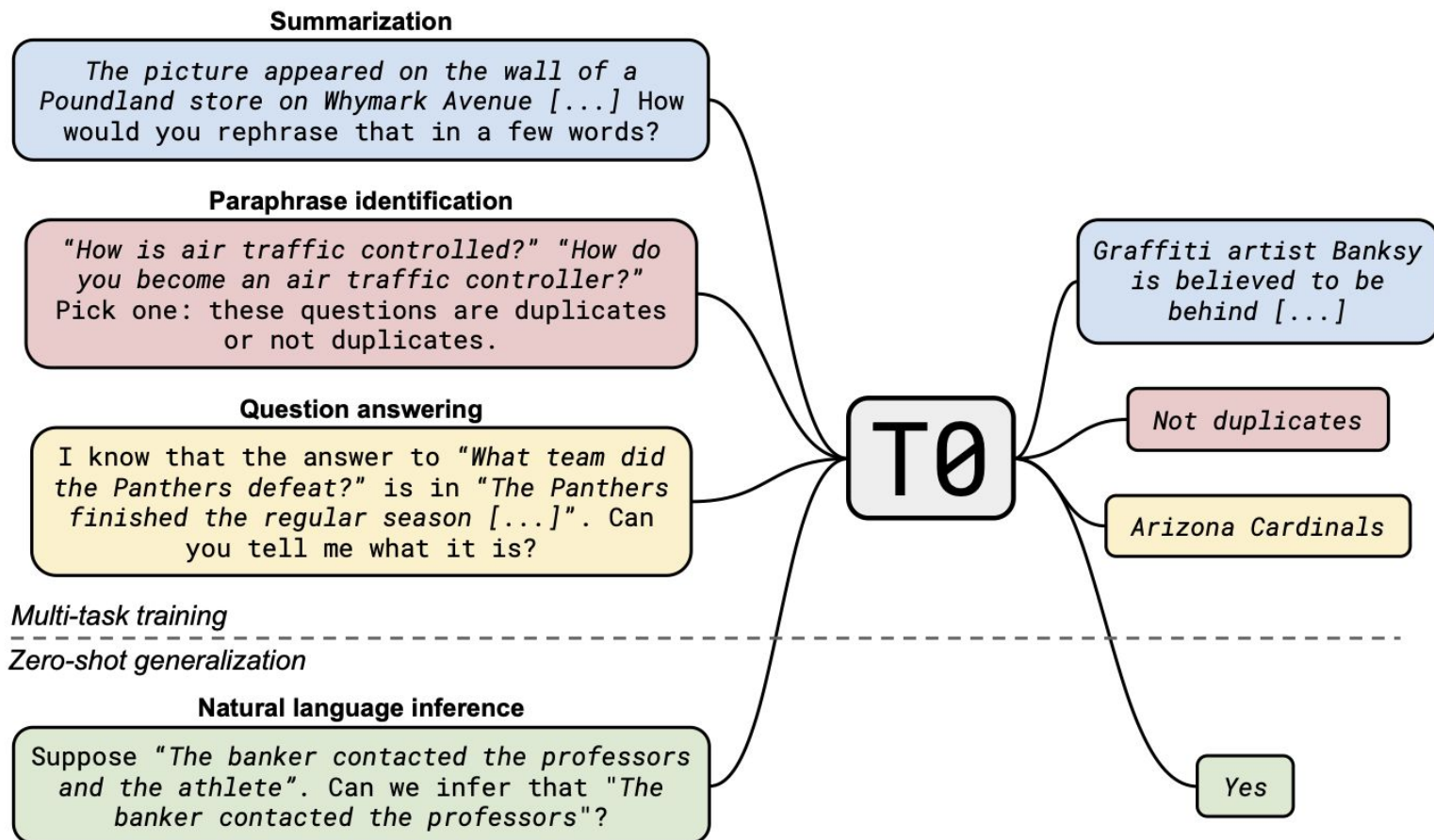
<https://blog.nytsoi.net/tag/reddit>

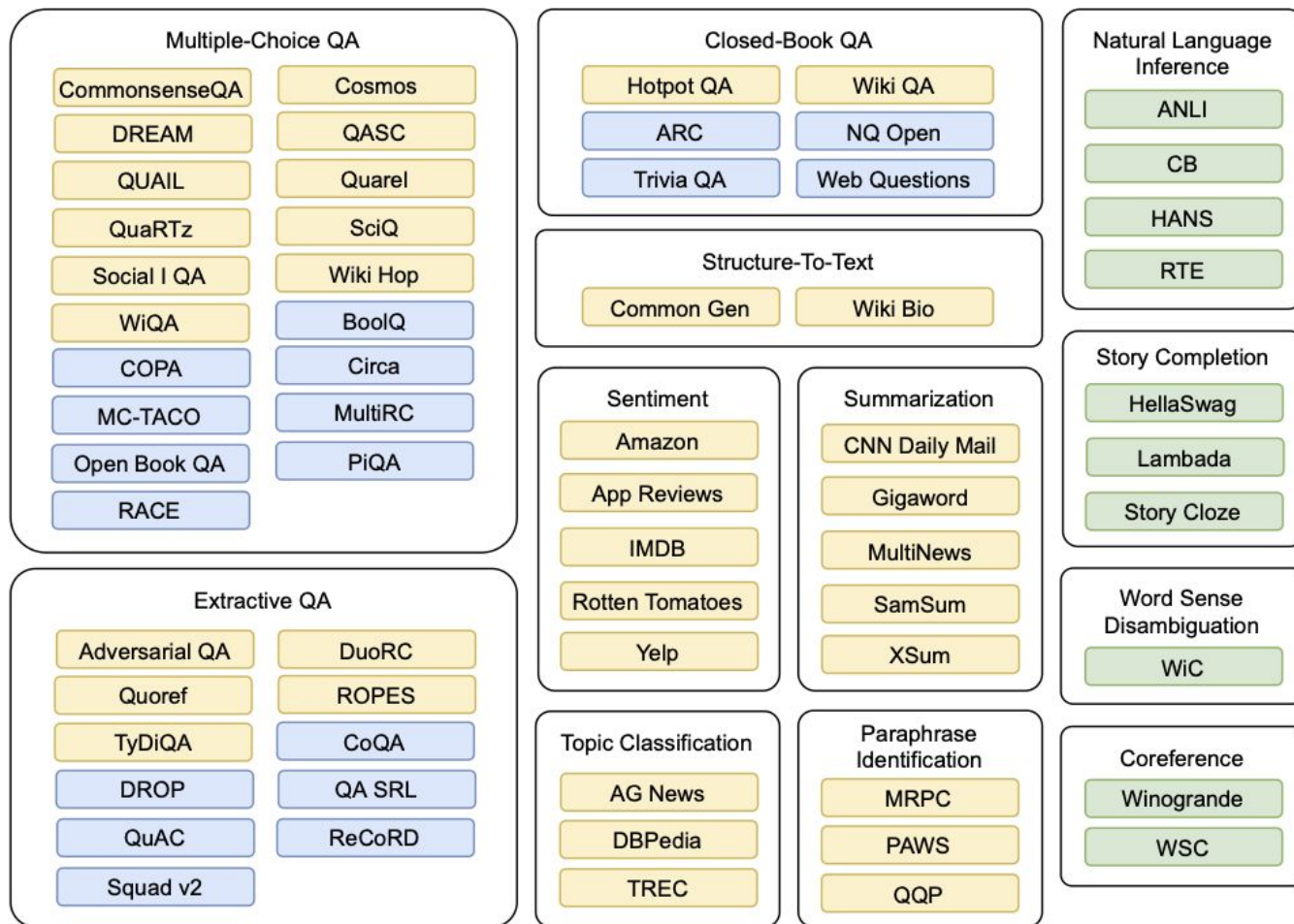
... Lately I've been seeing a pattern regarding videos stolen from other YouTube channels, reuploaded and monetized with ads. These videos are then mass posted on Reddit by bots masquerading as real users. tl;dr: Spambots are posting links to stolen videos on Reddit, copying comments from others to masquerade as legitimate users.

### **Pronoun resolution**

<https://nursecheung.com/ati-teas-guide-to-english-language-usage-understanding-pronouns/>

Jennifer is a vegetarian, so she will order a nonmeat entrée. In this example, the pronoun she is used to refer to Jennifer.





From "Multitask Prompted Training Enables Zero-Shot Task Generalization" by Sanh et al.

### QQP (Paraphrase)

Question1	How is air traffic controlled?
Question2	How do you become an air traffic controller?
Label	0

{Question1} {Question2}  
Pick one: These questions  
are duplicates or not  
duplicates.

{Choices[label]}

I received the questions  
"{Question1}" and  
"{Question2}". Are they  
duplicates?

{Choices[label]}

### XSum (Summary)

Document	The picture appeared on the wall of a Poundland store on Whymark Avenue...
Summary	Graffiti artist Banksy is believed to be behind...

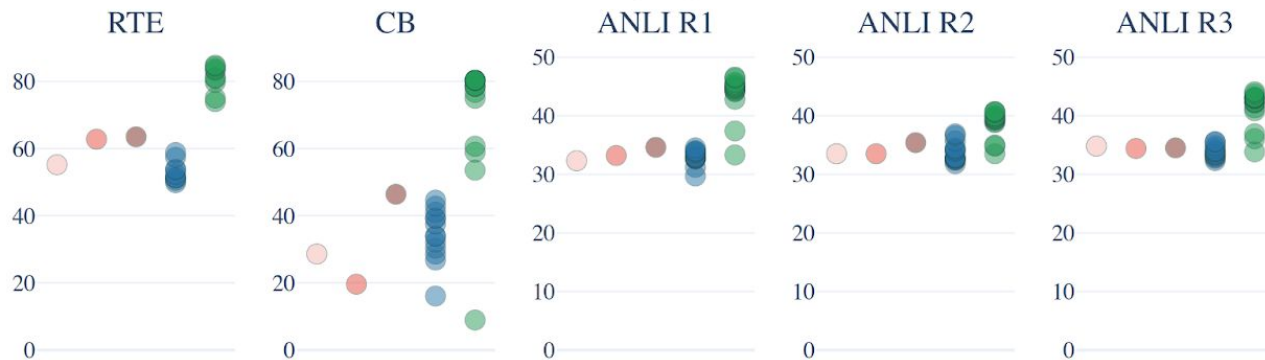
{Document}  
How would you  
rephrase that in  
a few words?

{Summary}

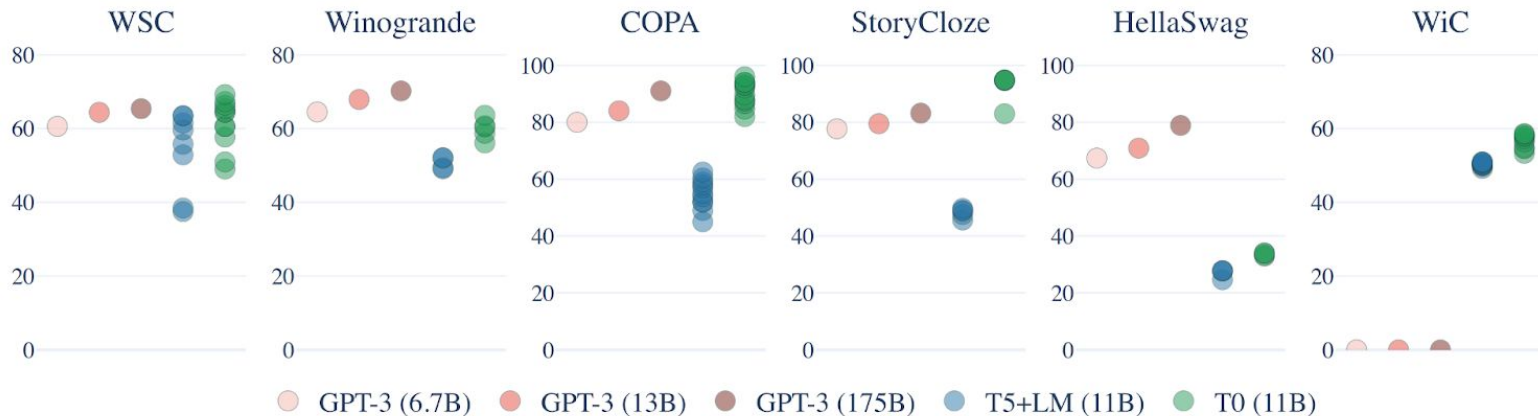
First, please read the article:  
{Document}  
Now, can you write me an  
extremely short abstract for it?

{Summary}

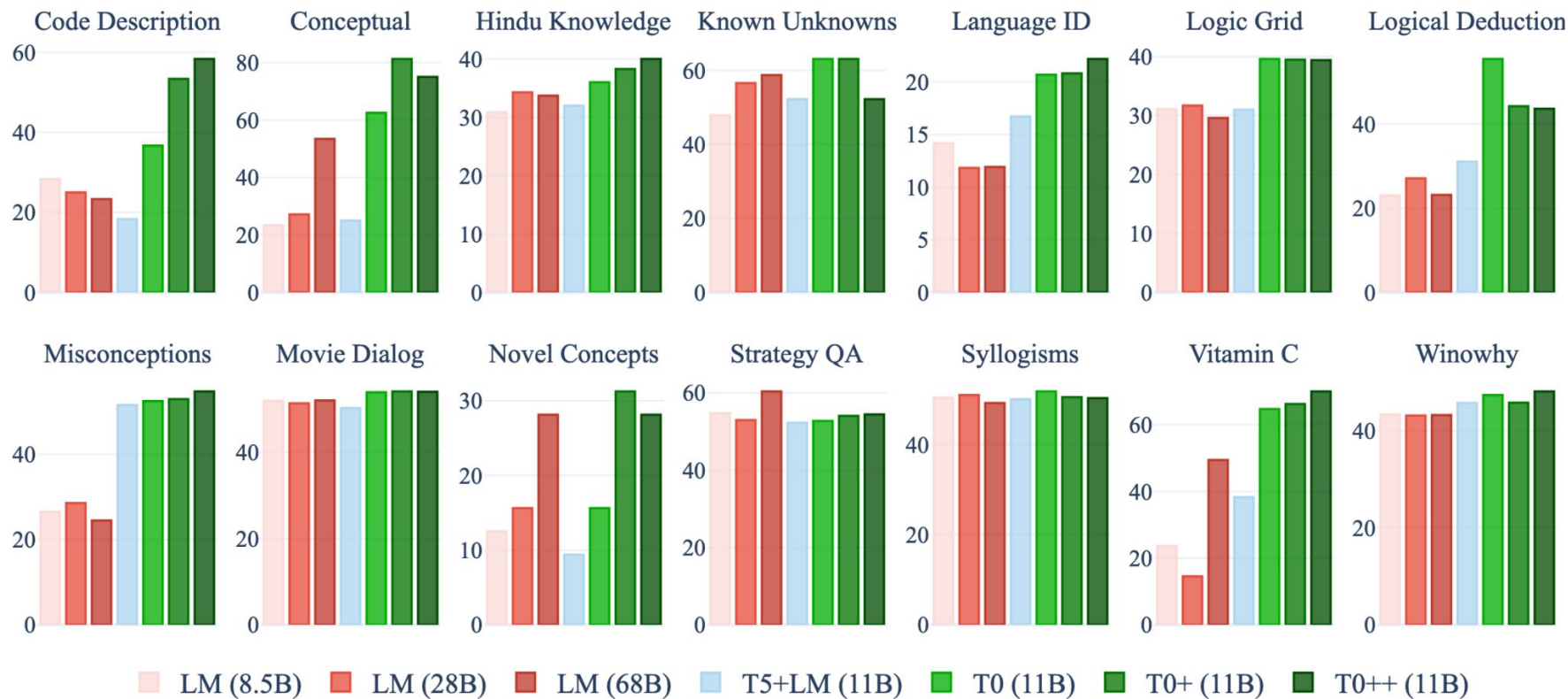
## Natural Language Inference



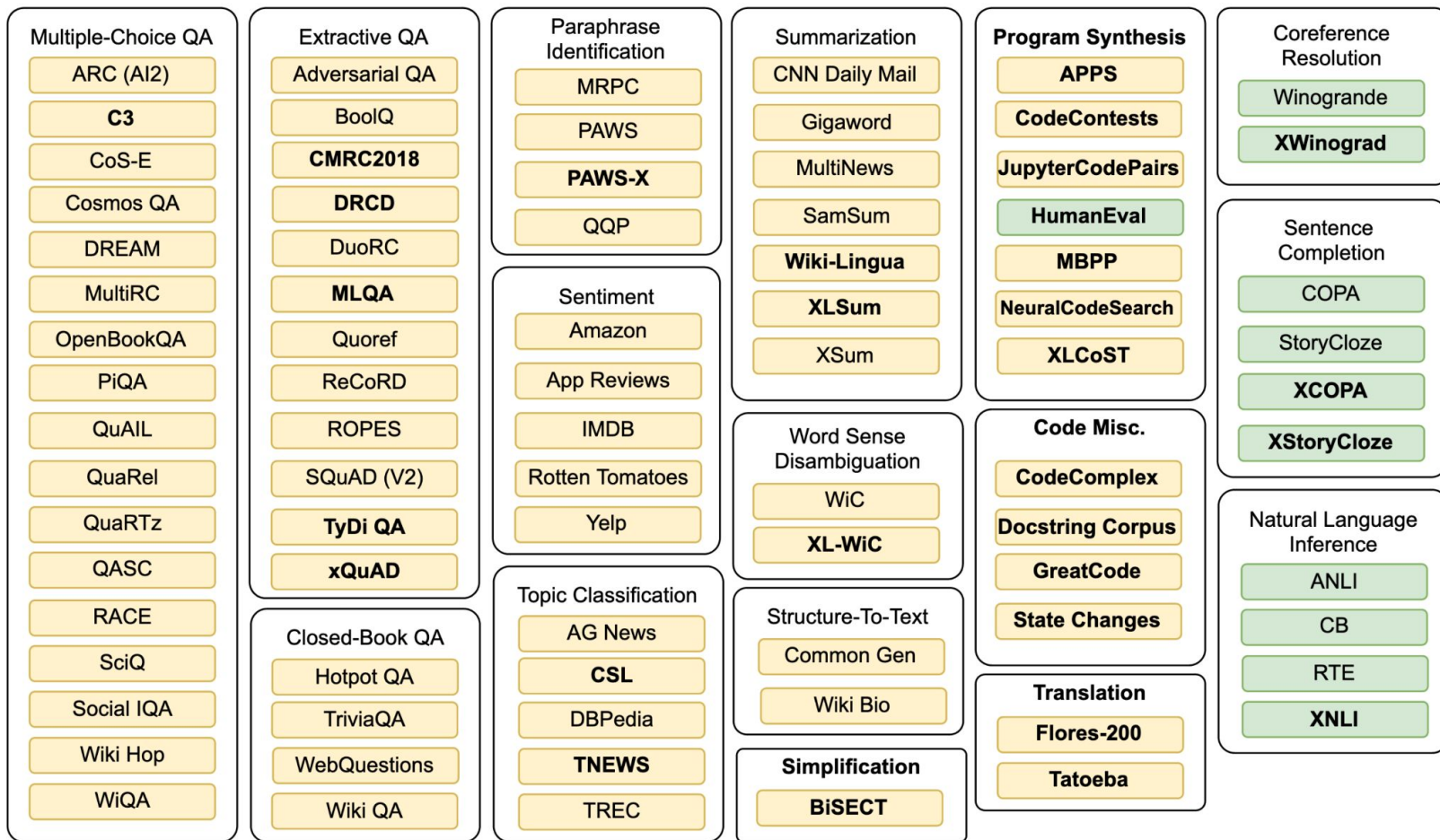
## Coreference Resolution



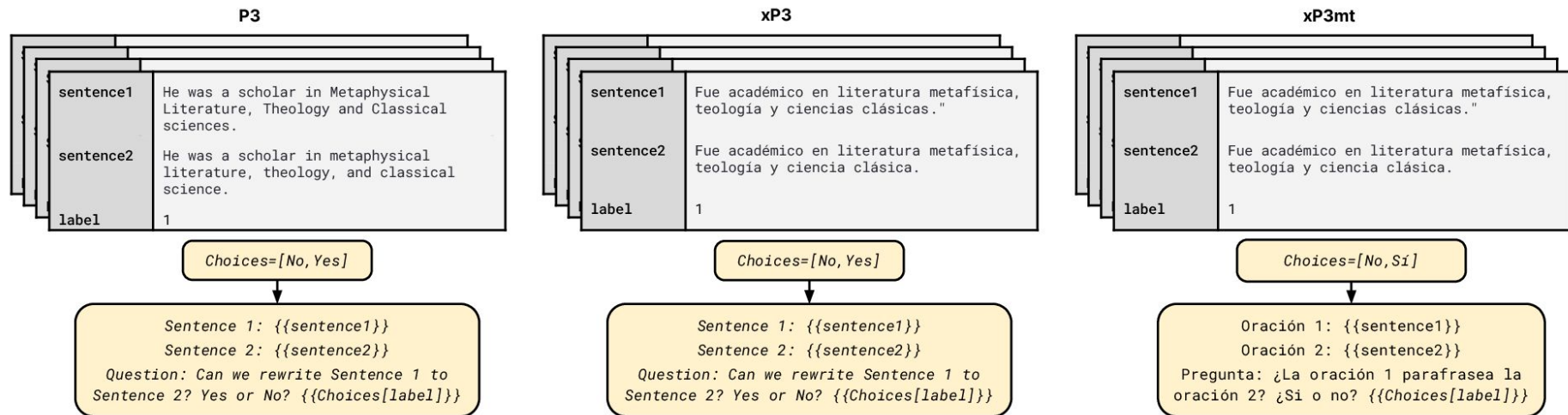
From "Multitask Prompted Training Enables Zero-Shot Task Generalization" by Sanh et al.



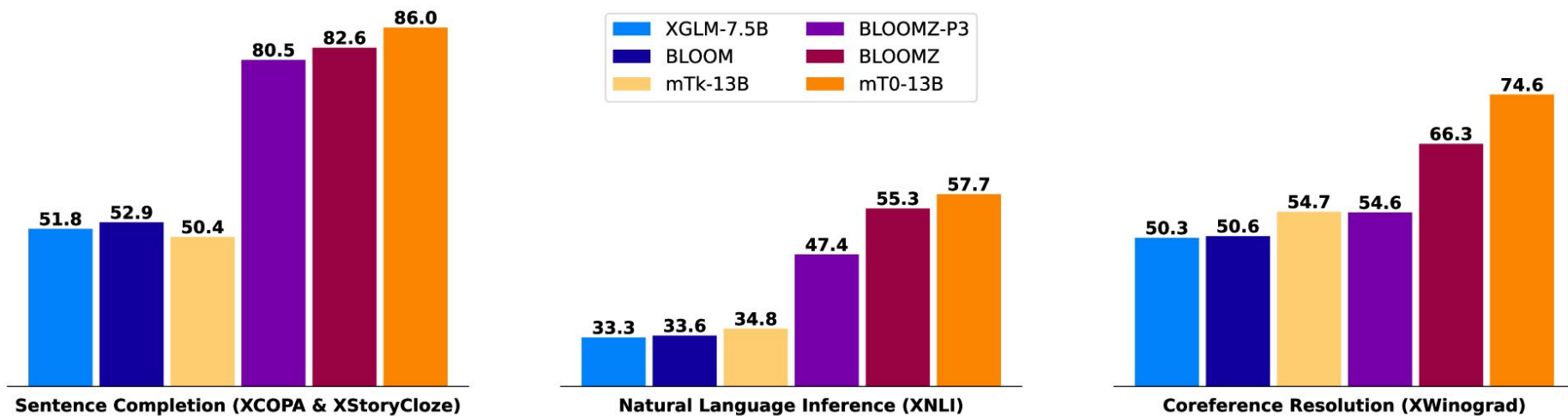
From "Multitask Prompted Training Enables Zero-Shot Task Generalization" by Sanh et al.



From “Crosslingual Generalization through Multitask Finetuning” by Muennighoff et al.

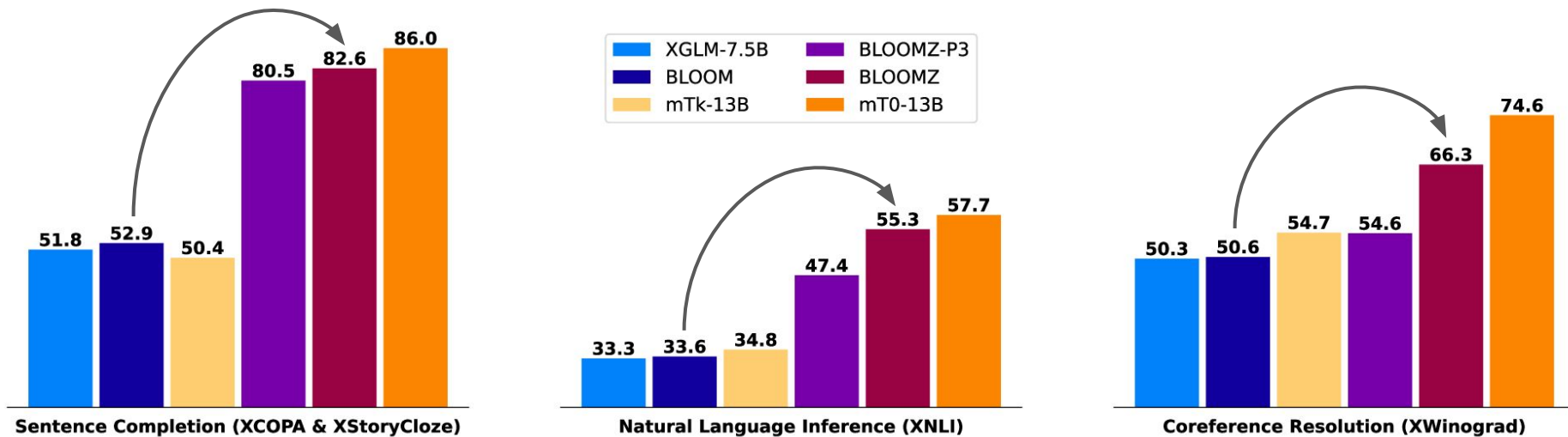


## Multilingual Multitask Generalization

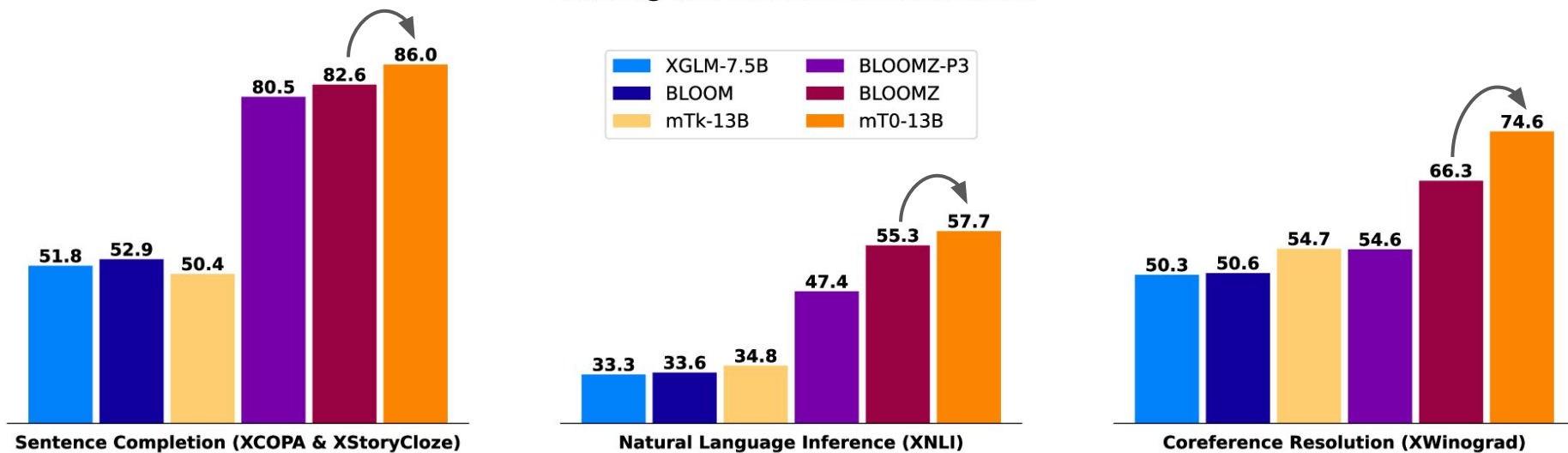


*From "Crosslingual Generalization through Multitask Finetuning" by Muennighoff et al.*

## Multilingual Multitask Generalization

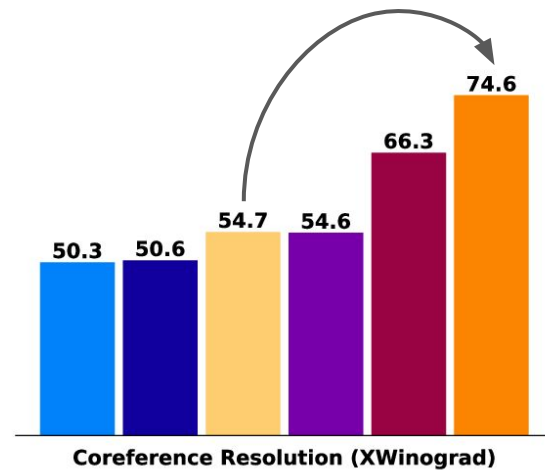
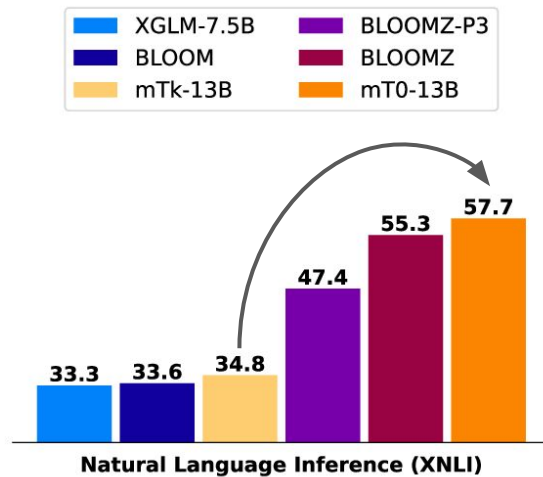
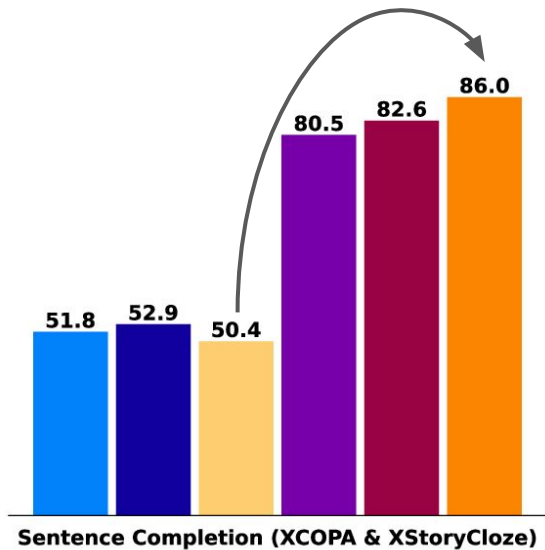


## Multilingual Multitask Generalization

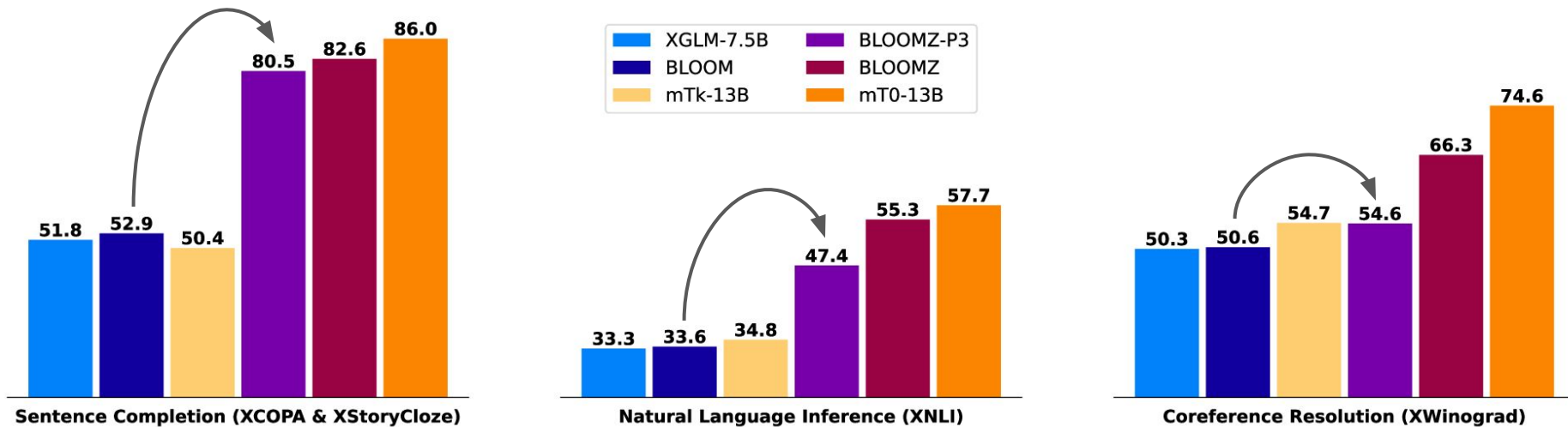


*From "Crosslingual Generalization through Multitask Finetuning" by Muennighoff et al.*

## Multilingual Multitask Generalization

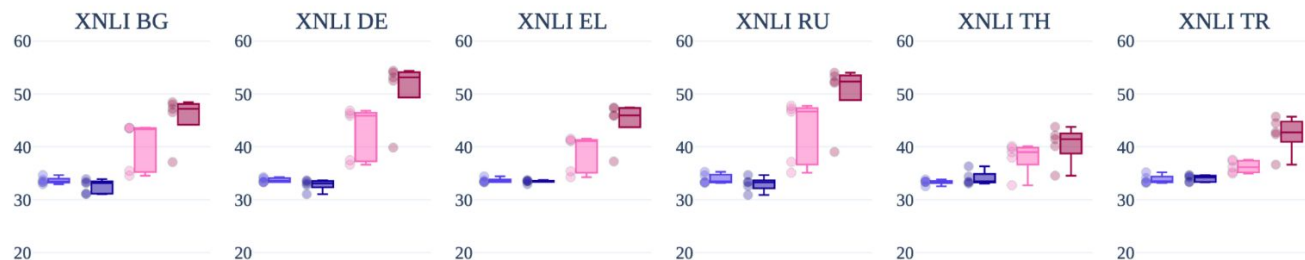


## Multilingual Multitask Generalization

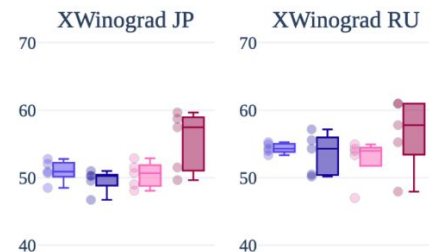


# Performance on languages that were never intentionally trained on

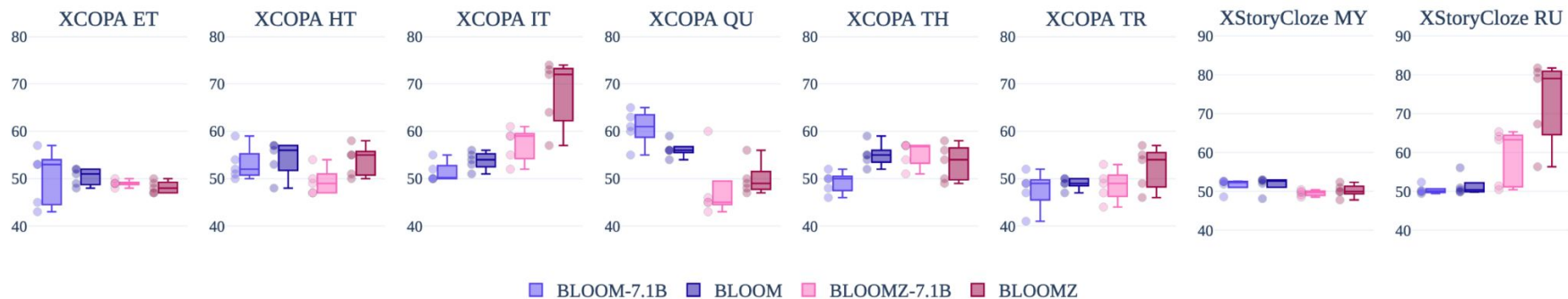
## Natural Language Inference

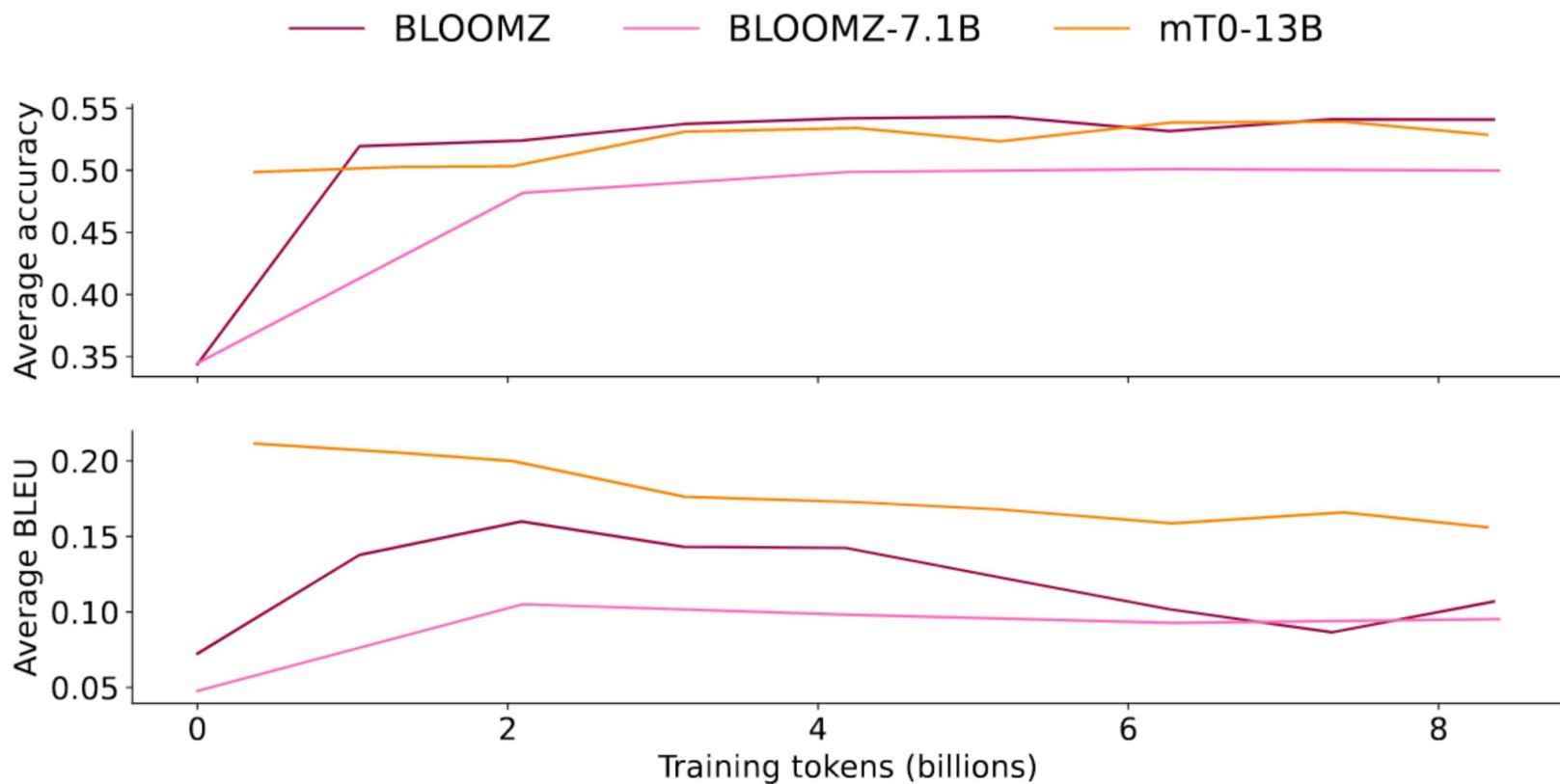


## Coreference Resolution



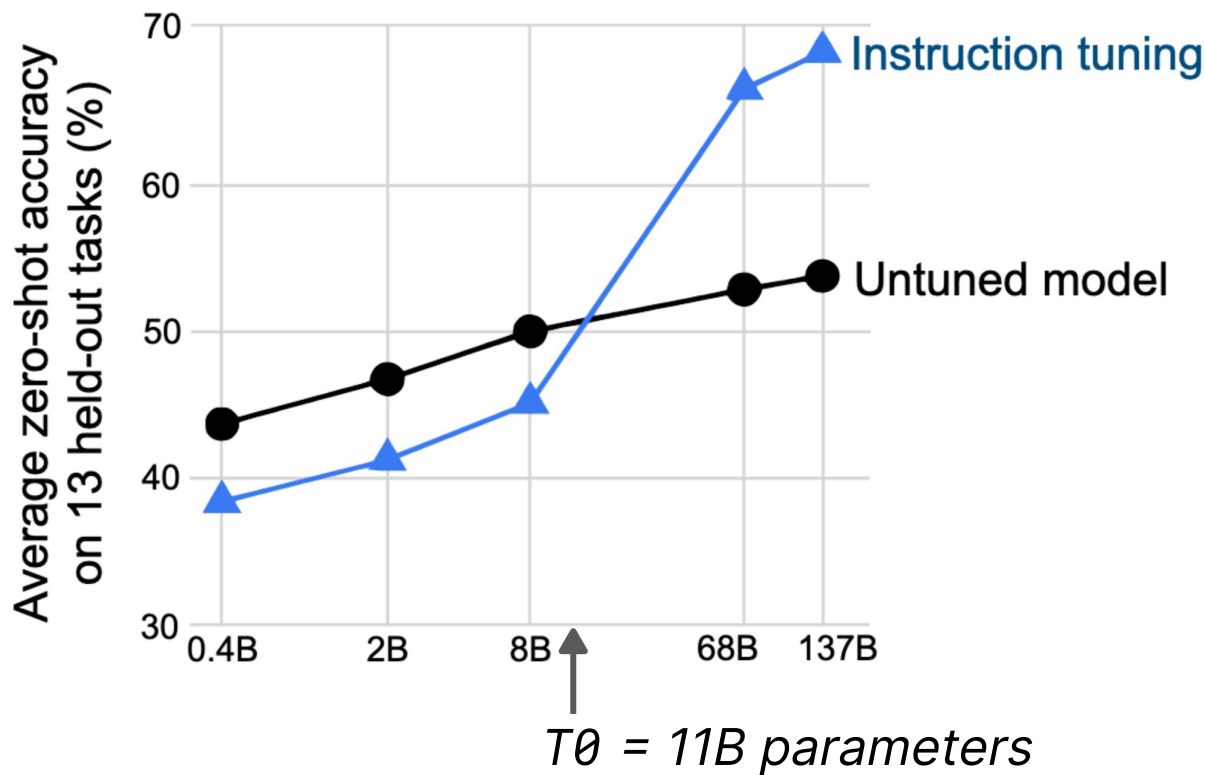
## Sentence Completion



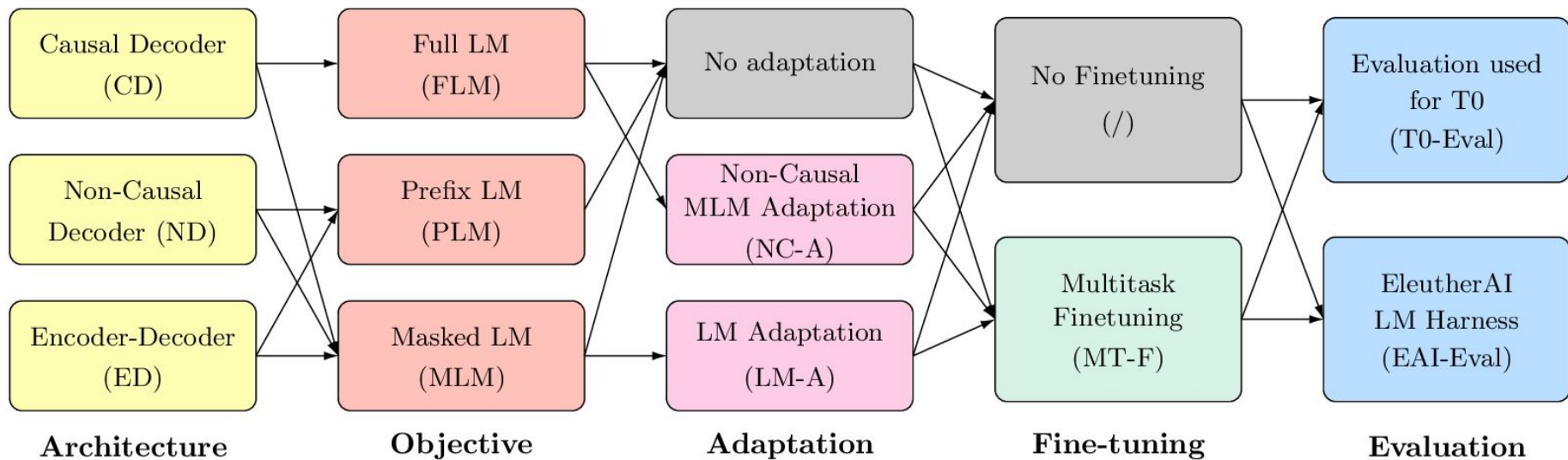


*From "Crosslingual Generalization through Multitask Finetuning" by Muennighoff et al.*

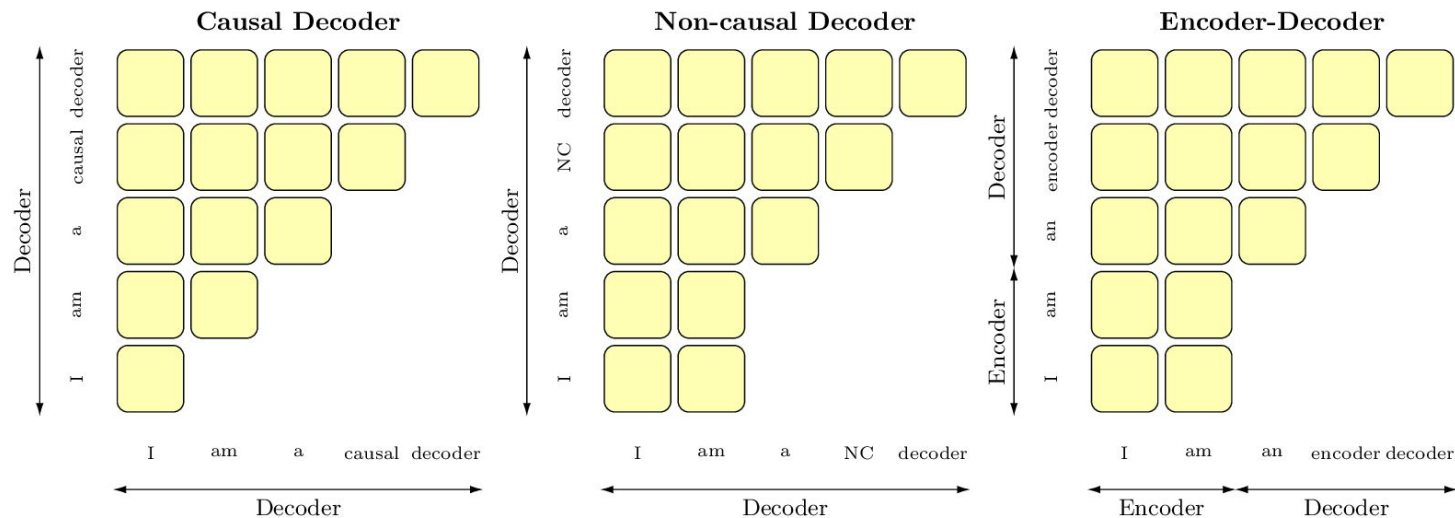
## Performance on held-out tasks



From "Fine-Tuned Language Models are Zero-Shot Learners" by Wei et al.



From "What Language Model Architecture and Pretraining Objective Work Best for Zero-Shot Generalization?" by Wang et al.



**Full Language Modeling**

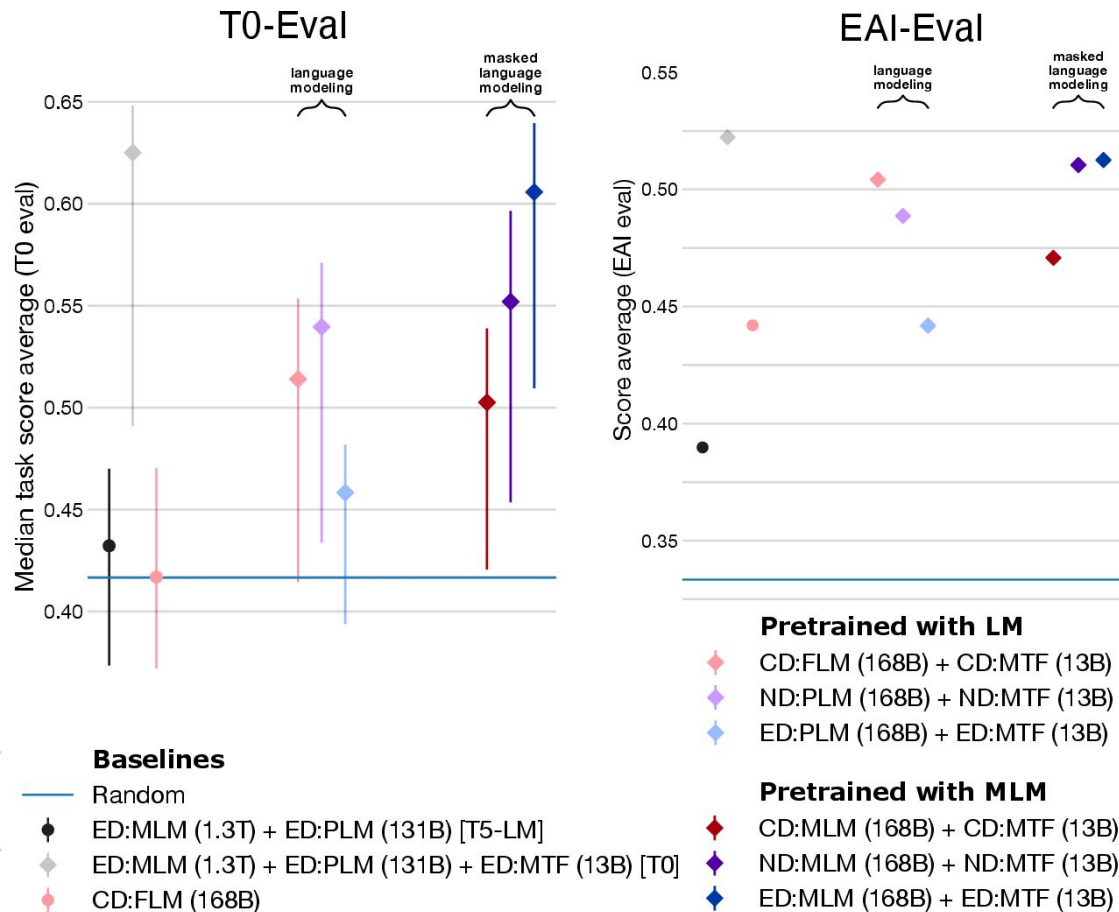
May <sup>targets</sup> the force be with you

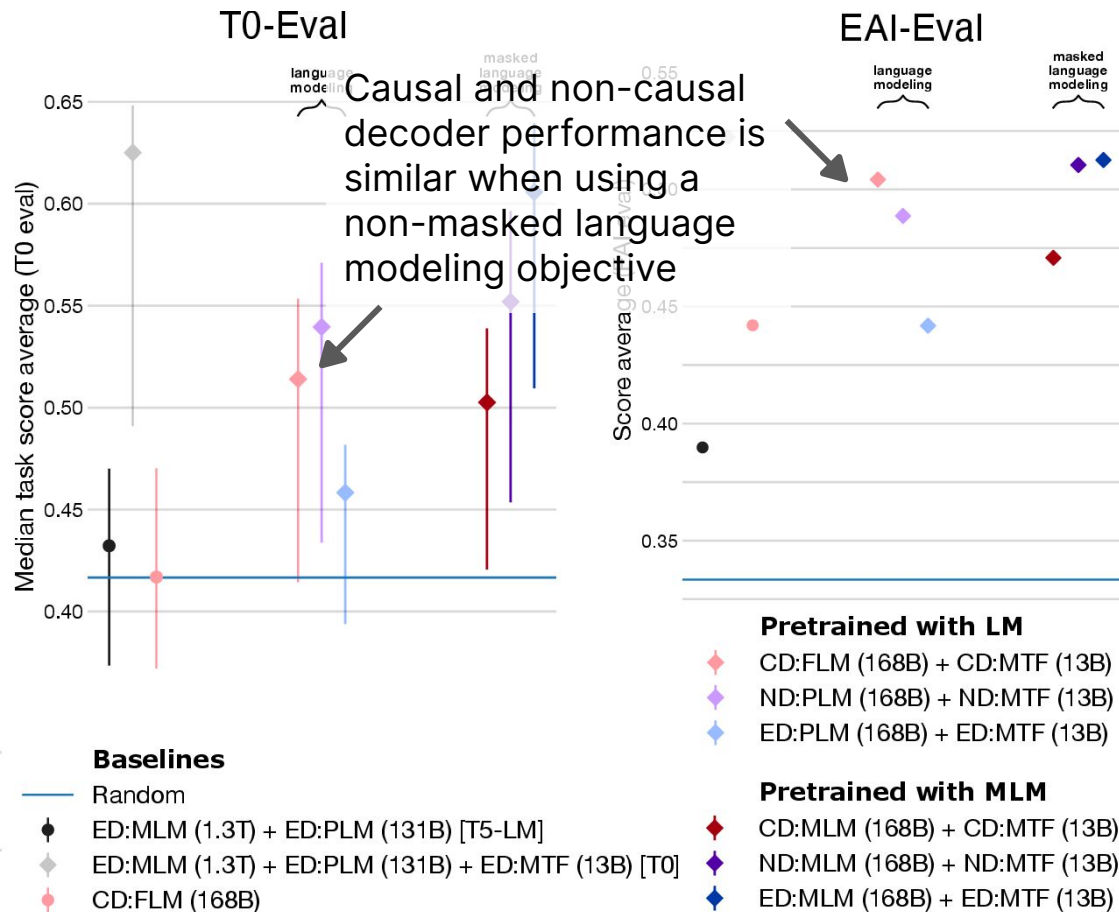
**Prefix Language Modeling**

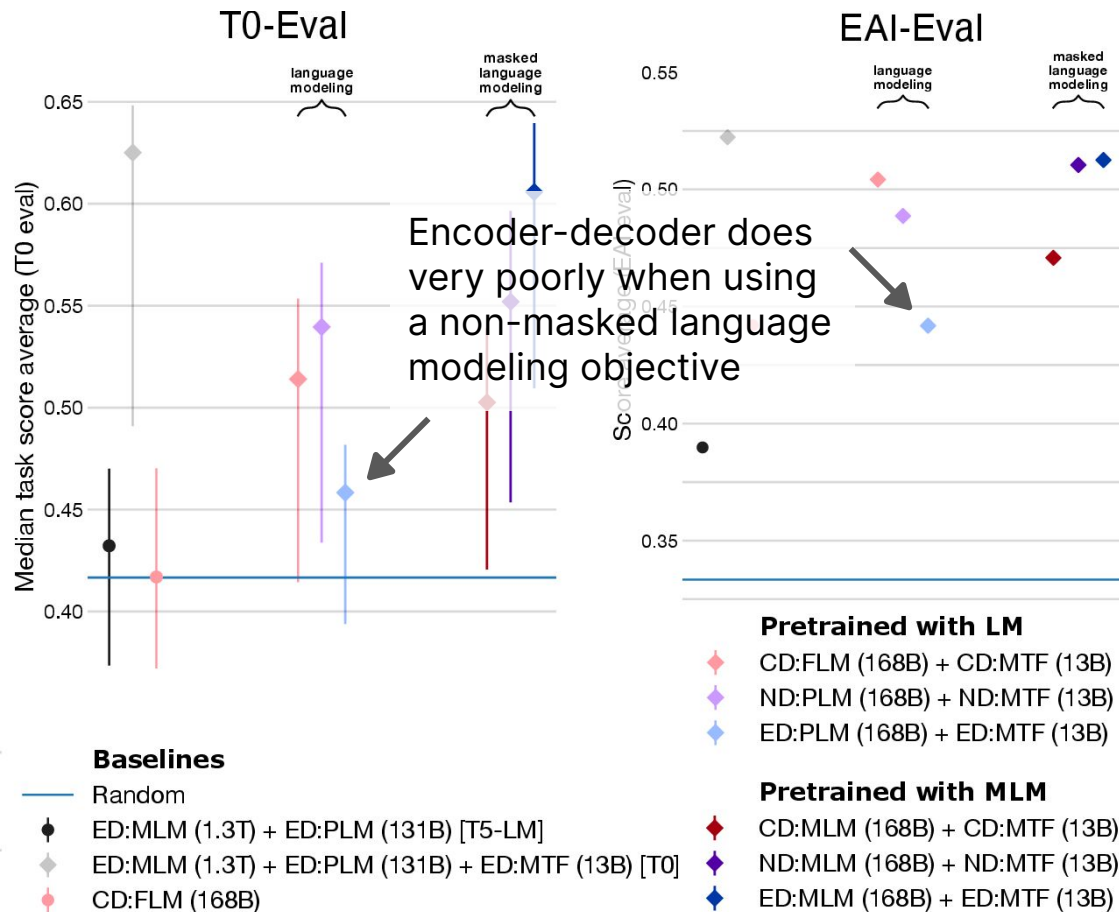
May the force <sup>targets</sup> be with you

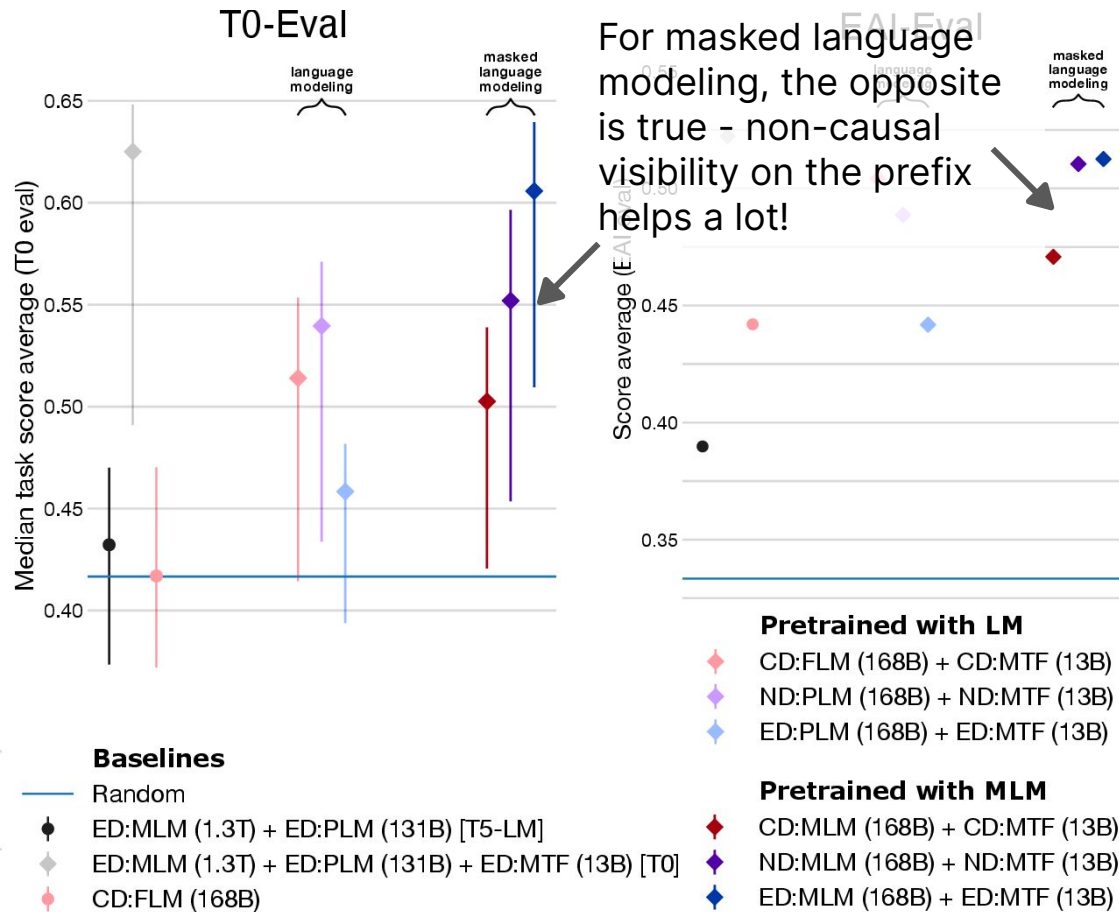
**Masked Language Modeling**

May <sup>targets</sup> the force be with you









## Zero-shot

The model predicts the answer given only a natural language description of the task. No gradient updates are performed.

```
1 Translate English to French: ← task description
2 cheese => ..... ← prompt
```

---

## Few-shot

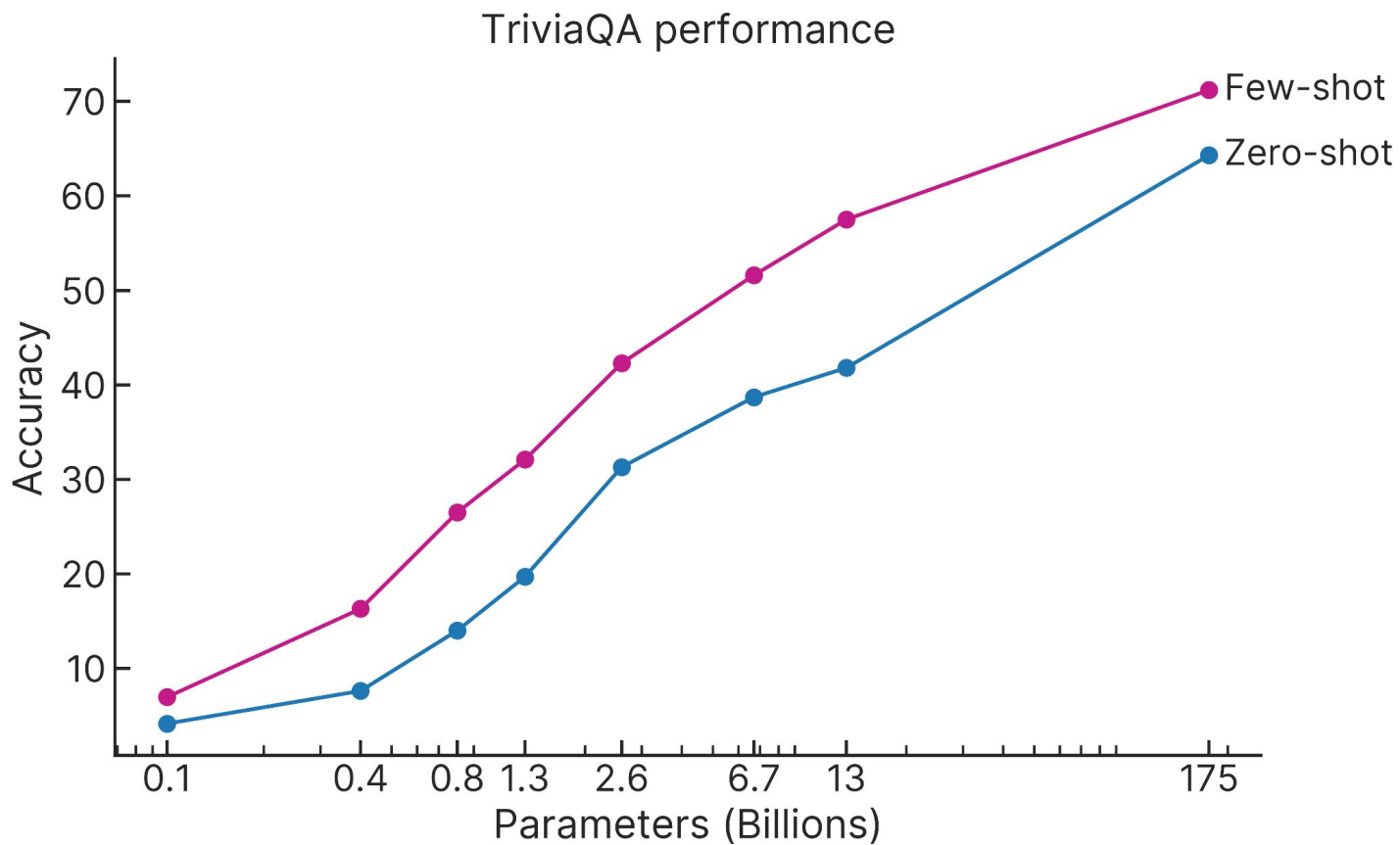
In addition to the task description, the model sees a few examples of the task. No gradient updates are performed.

```
1 Translate English to French: ← task description
2 sea otter => loutre de mer ← examples
3 peppermint => menthe poivrée ←
4 plush girafe => girafe peluche ←
5 cheese => ..... ← prompt
```

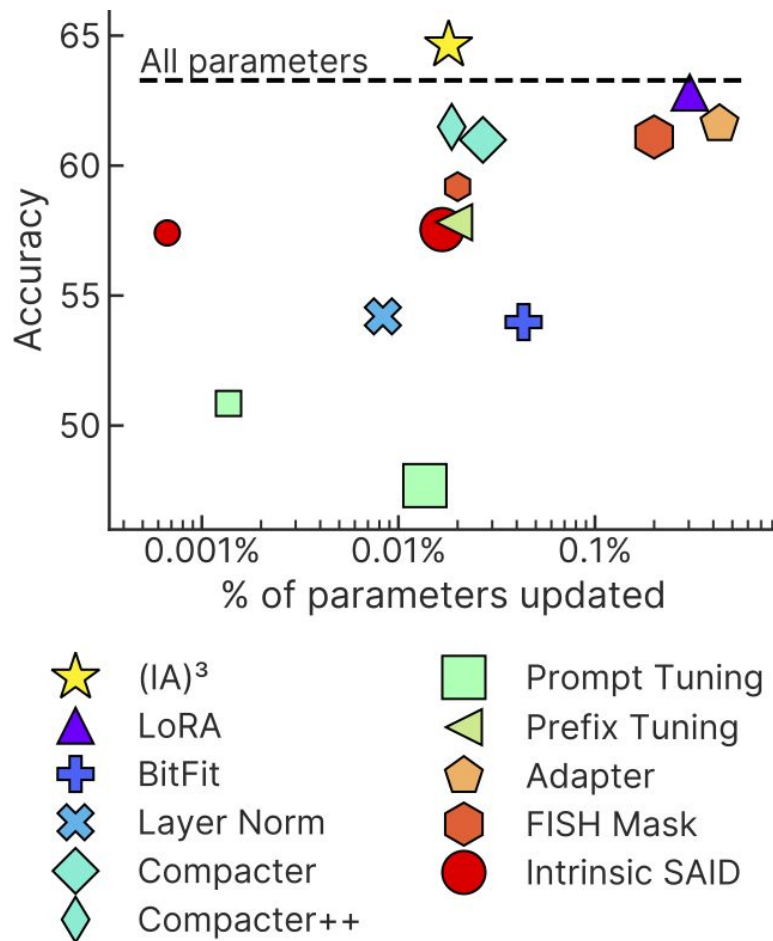
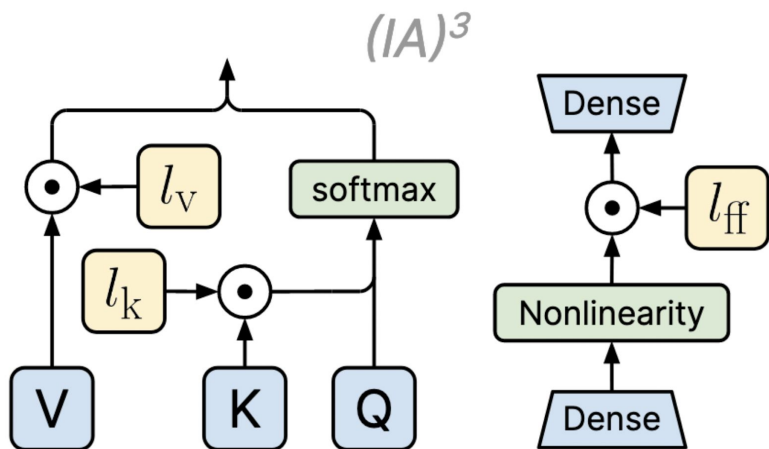
## Fine-tuning

The model is trained via repeated gradient updates using a large corpus of example tasks.

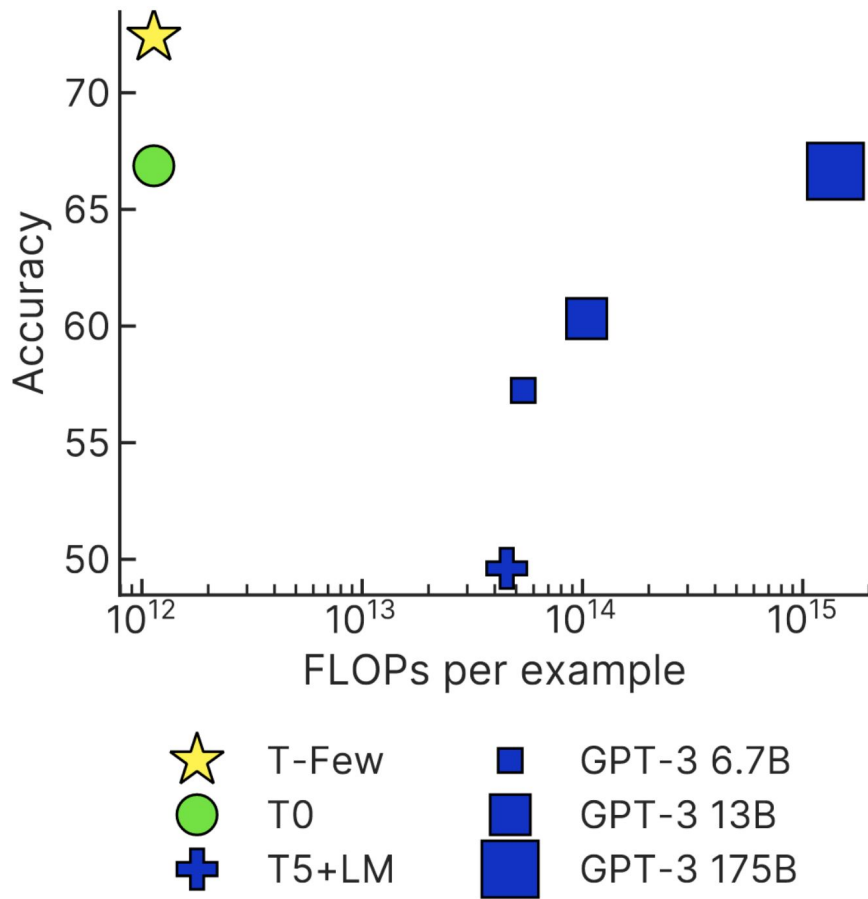




*From "Language Models are Few-Shot Learners" by Brown et al.*



Method	Inference FLOPs	Training FLOPs	Disk space
T-Few	1.1e12	2.7e16	4.2 MB
T0 [1]	1.1e12	0	0 B
T5+LM [14]	4.5e13	0	16 kB
GPT-3 6.7B [4]	5.4e13	0	16 kB
GPT-3 13B [4]	1.0e14	0	16 kB
GPT-3 175B [4]	1.4e15	0	16 kB



Method	Acc.
T-Few	75.8%
Human baseline [2]	73.5%
PET [50]	69.6%
SetFit [51]	66.9%
GPT-3 [4]	62.7%

Table 2: Top-5 best methods on RAFT as of writing. T-Few is the first method to outperform the human baseline and achieves over 6% higher accuracy than the next-best method.

## References

[Multitask Prompted Training Enables Zero-Shot Task Generalization](#)

[Crosslingual Generalization through Multitask Finetuning](#)

[What Language Model Architecture and Pretraining Objective Work Best for Zero-Shot Generalization?](#)

[Few-Shot Parameter-Efficient Fine-Tuning is Better and Cheaper than In-Context Learning](#)

[BLOOM: A 176B-Parameter Open-Access Multilingual Language Model](#)

Please give me feedback:

<http://bit.ly/colin-talk-feedback>

Thanks!