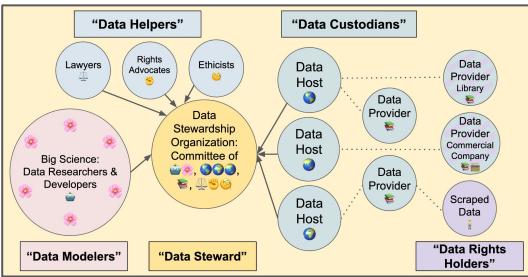


# Building Better Language Models: *Insights from BigScience*

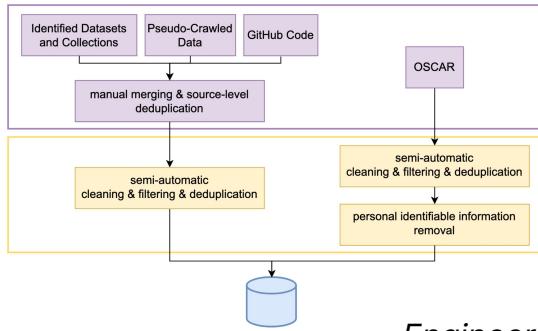


Colin Raffel

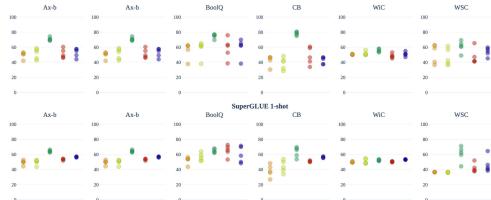
## Data governance



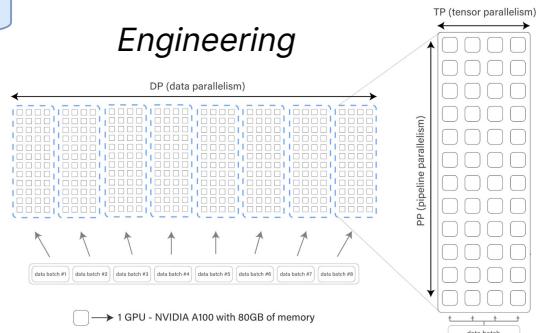
## Data sourcing



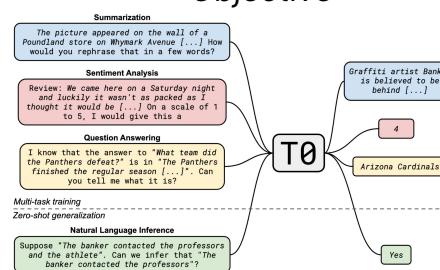
## Evaluation



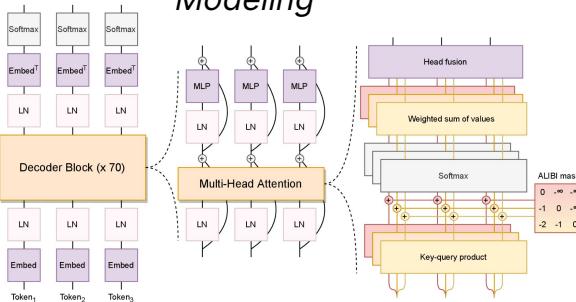
## Engineering

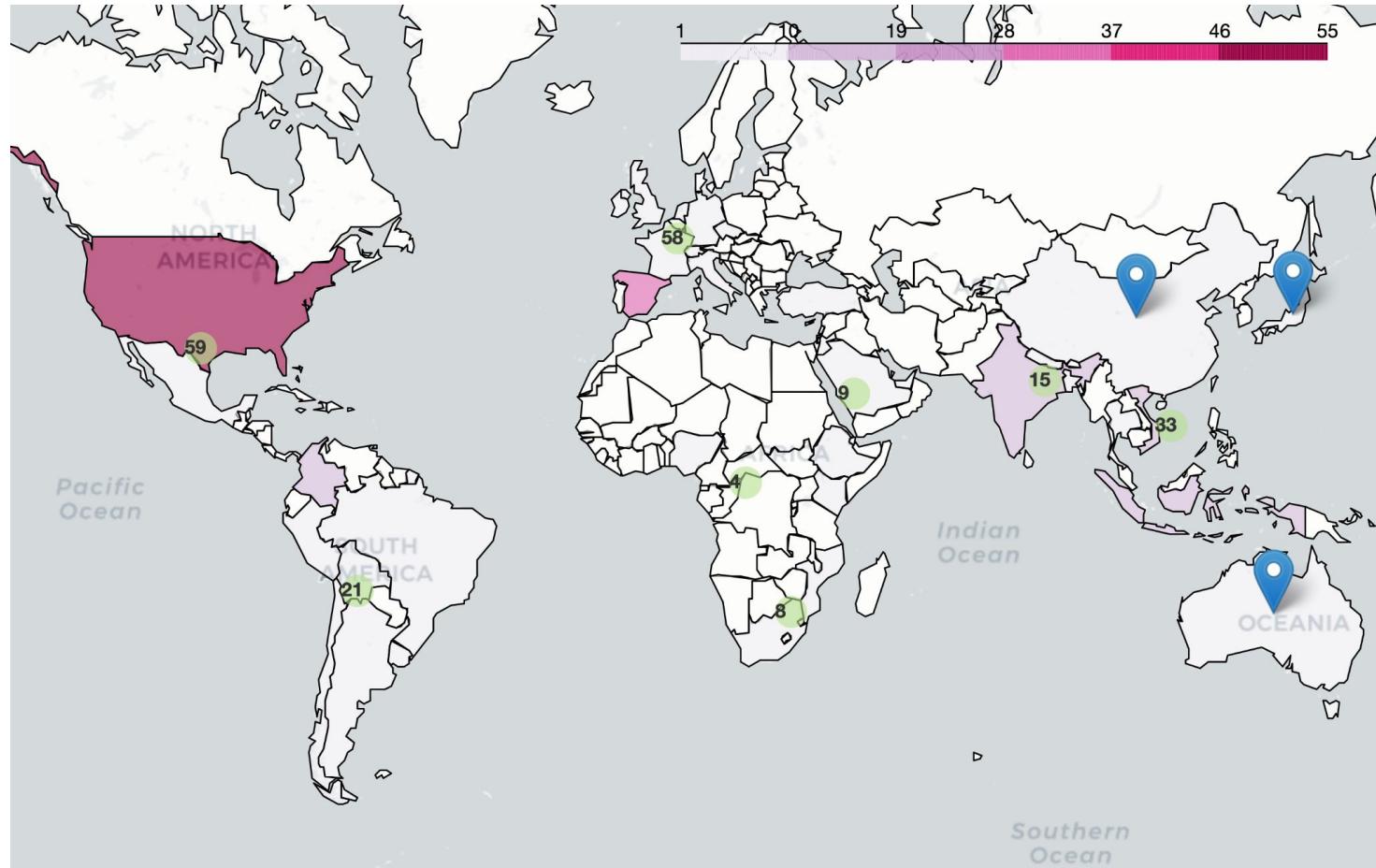


## Objective



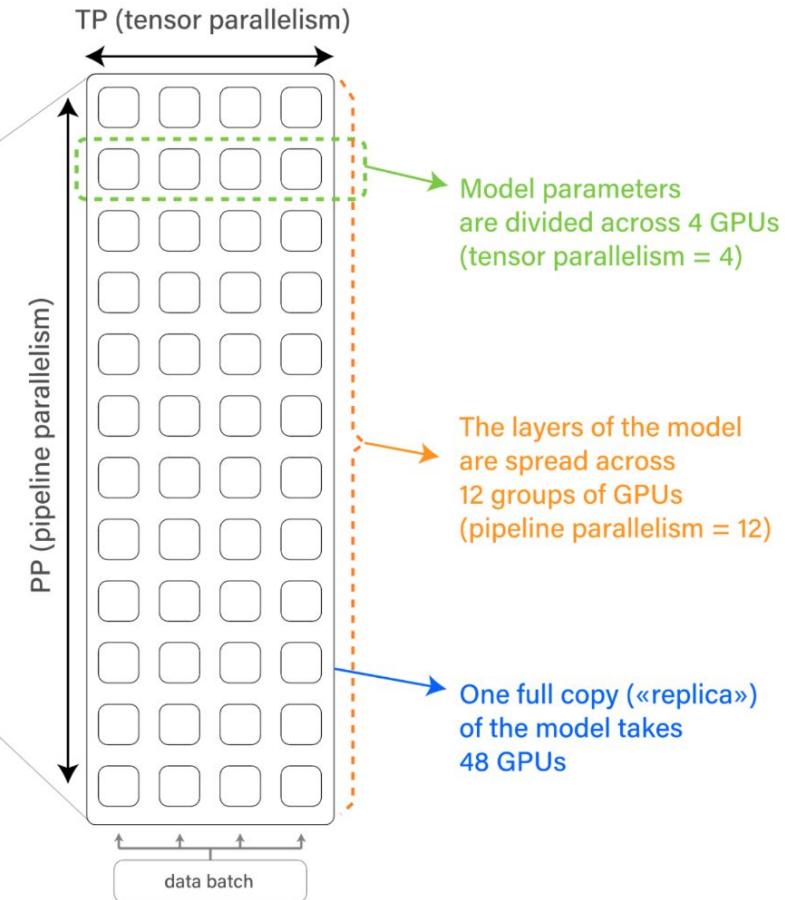
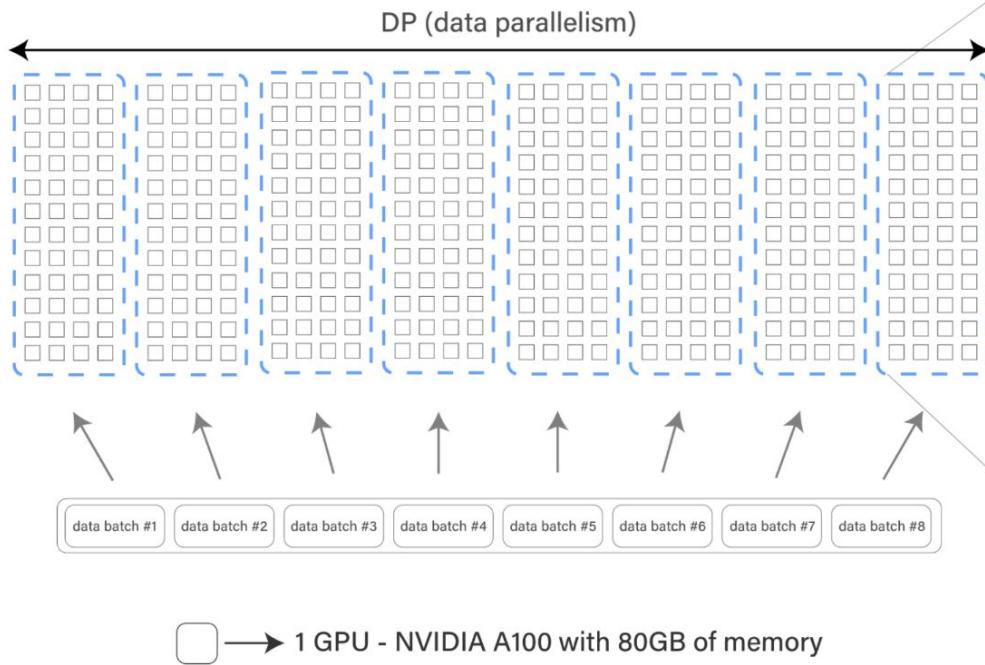
## Modeling

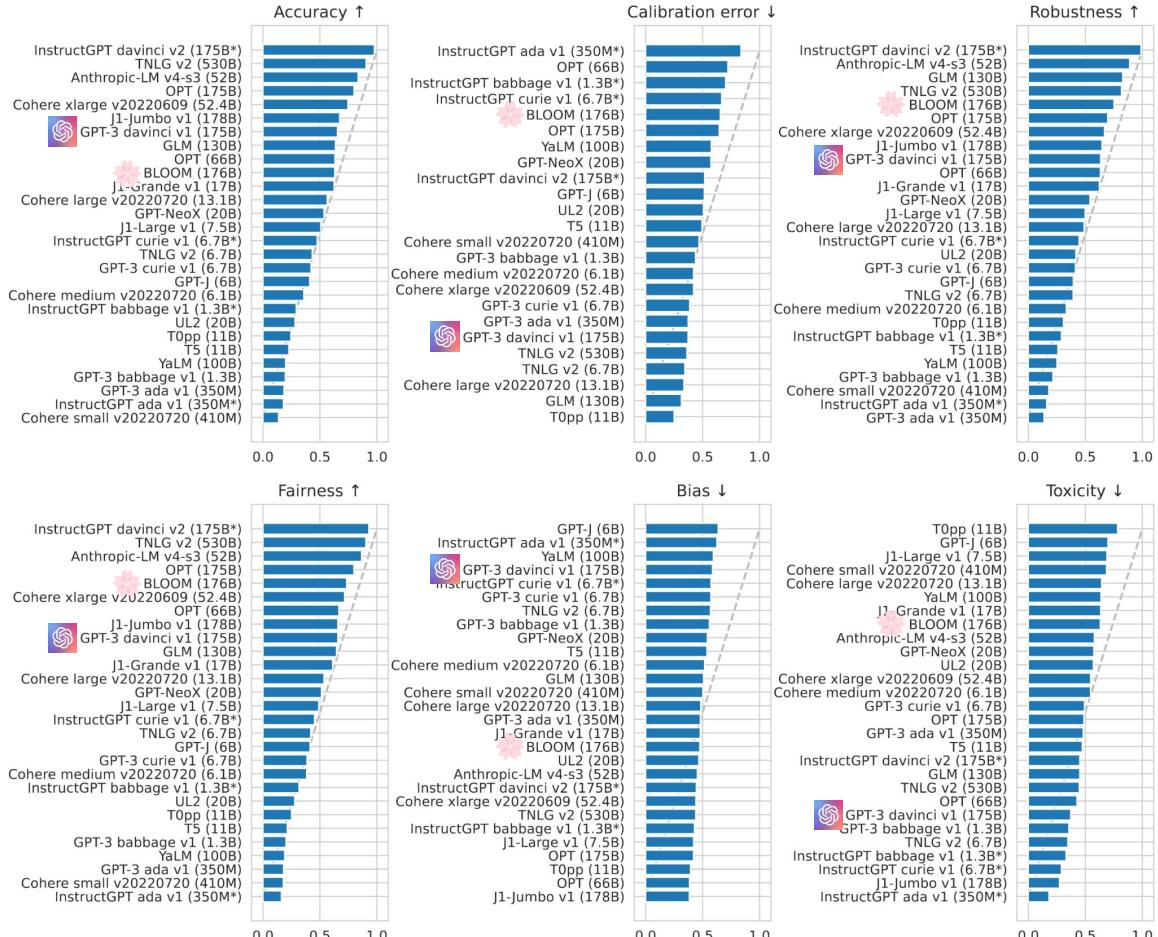




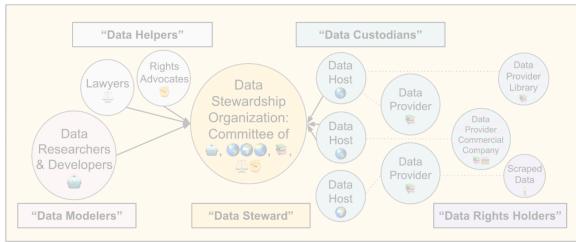
From <https://huggingface.co/spaces/bigscience/SourcingCatalog>

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

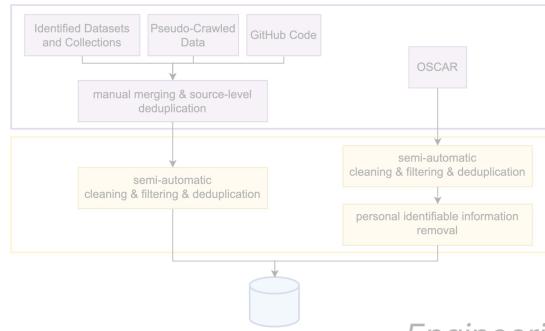




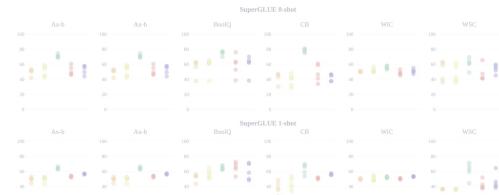
## Data governance



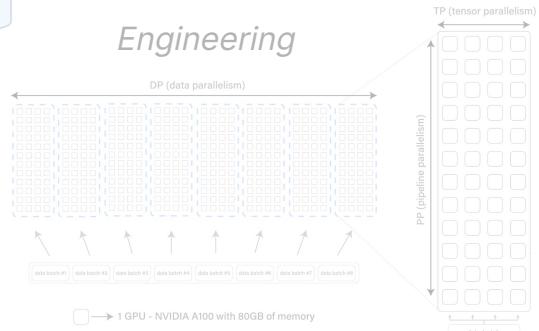
## Data sourcing



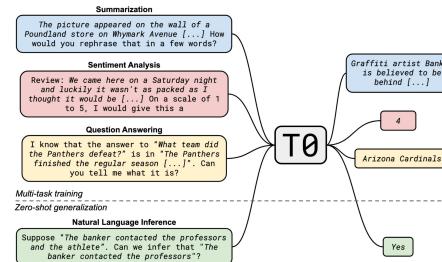
## Evaluation



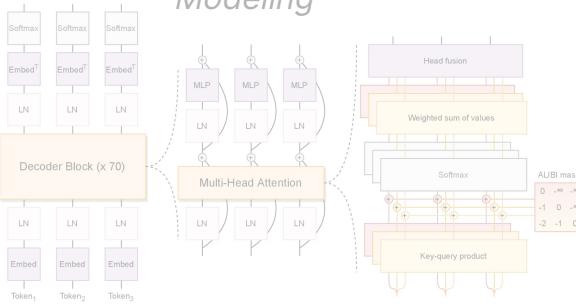
## Engineering



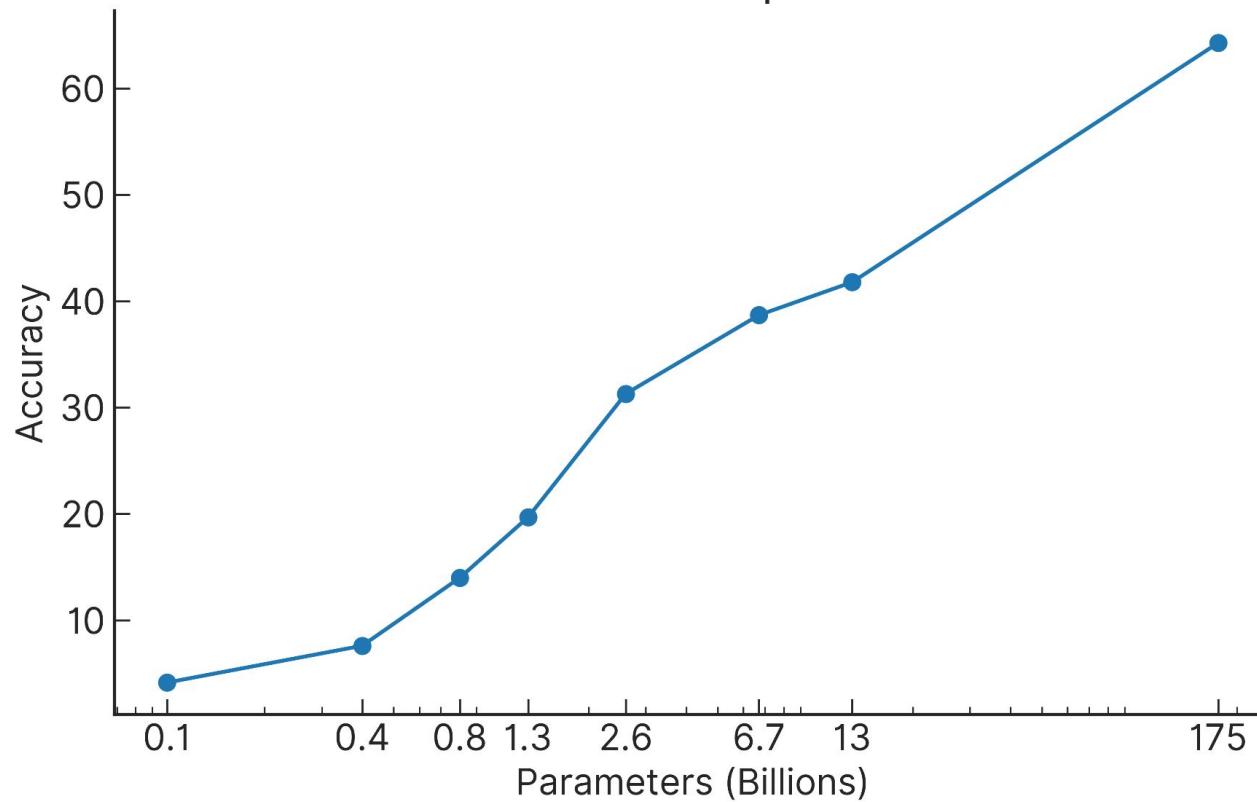
## Objective



## Modeling



## TriviaQA zero-shot performance



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

## Closed-book question answering

<http://www.autos weblog.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.

## Summarization

*The picture appeared on the wall of a Poundland store on Whymark Avenue [...] How would you rephrase that in a few words?*

## Paraphrase identification

"How is air traffic controlled?" "How do you become an air traffic controller?"  
Pick one: these questions are duplicates or not duplicates.

## Question answering

I know that the answer to "What team did the Panthers defeat?" is in "The Panthers finished the regular season [...]" . Can you tell me what it is?

## Multi-task training

## Zero-shot generalization

## Natural language inference

Suppose "The banker contacted the professors and the athlete". Can we infer that "The banker contacted the professors"?

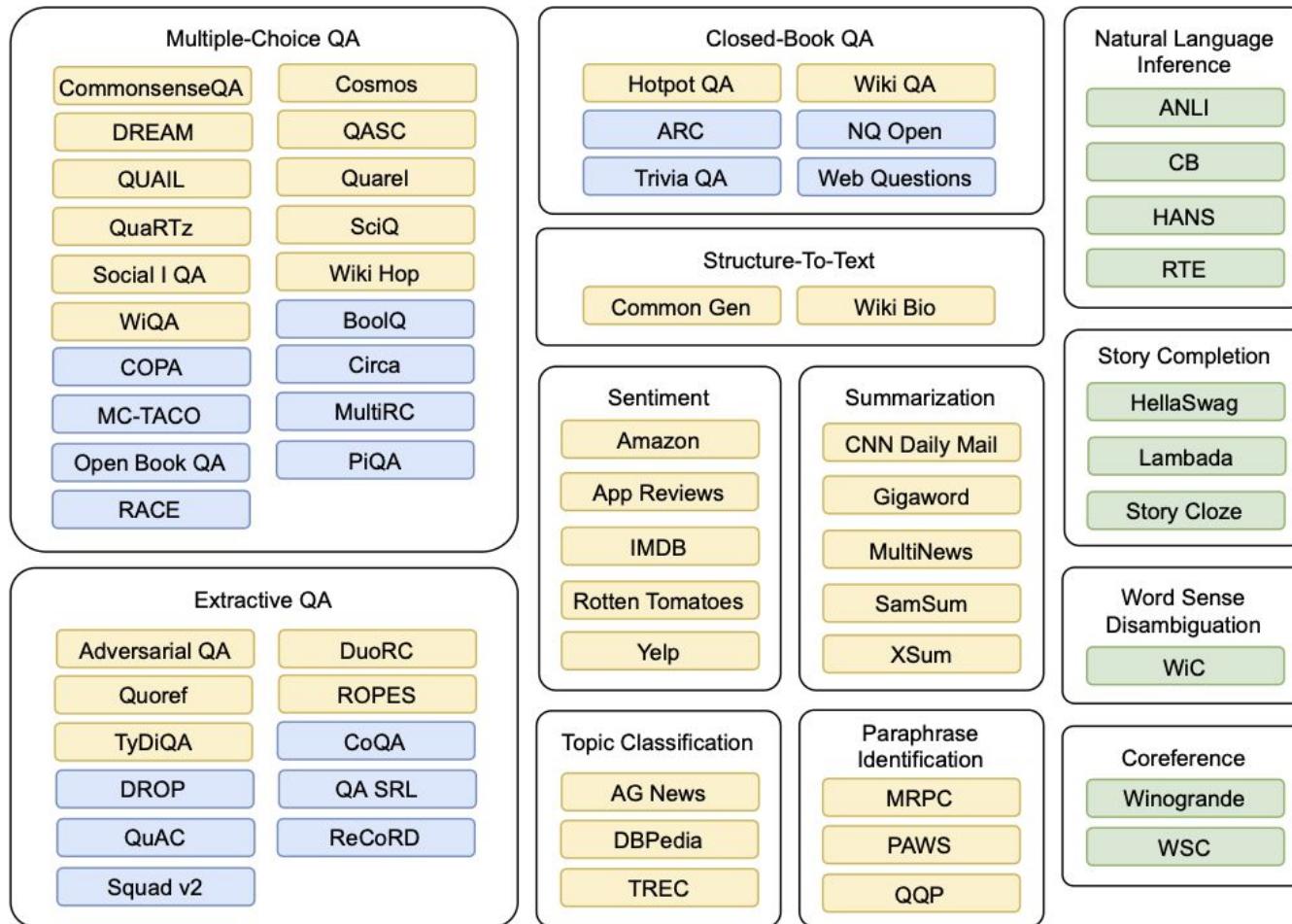
T0

Graffiti artist Banksy is believed to be behind [...]

Not duplicates

Arizona Cardinals

Yes



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...

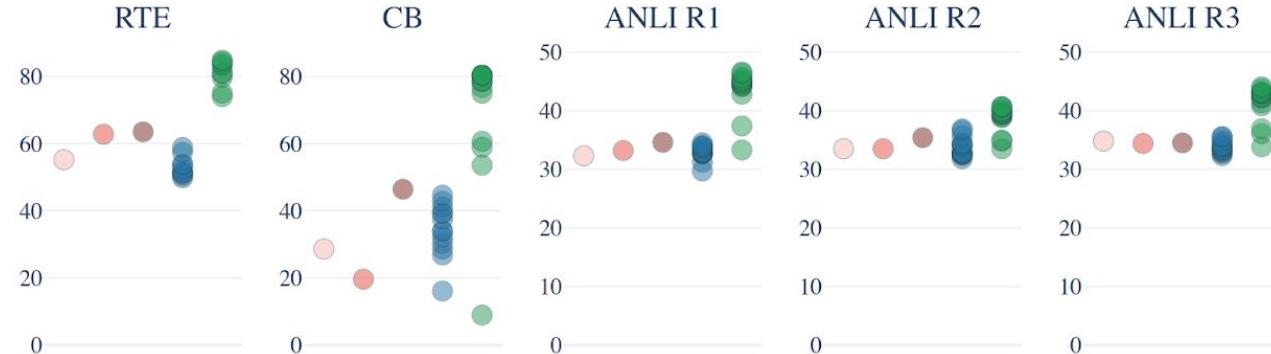
{Document}  
How would you  
rephrase that in  
a few words?

{Summary}

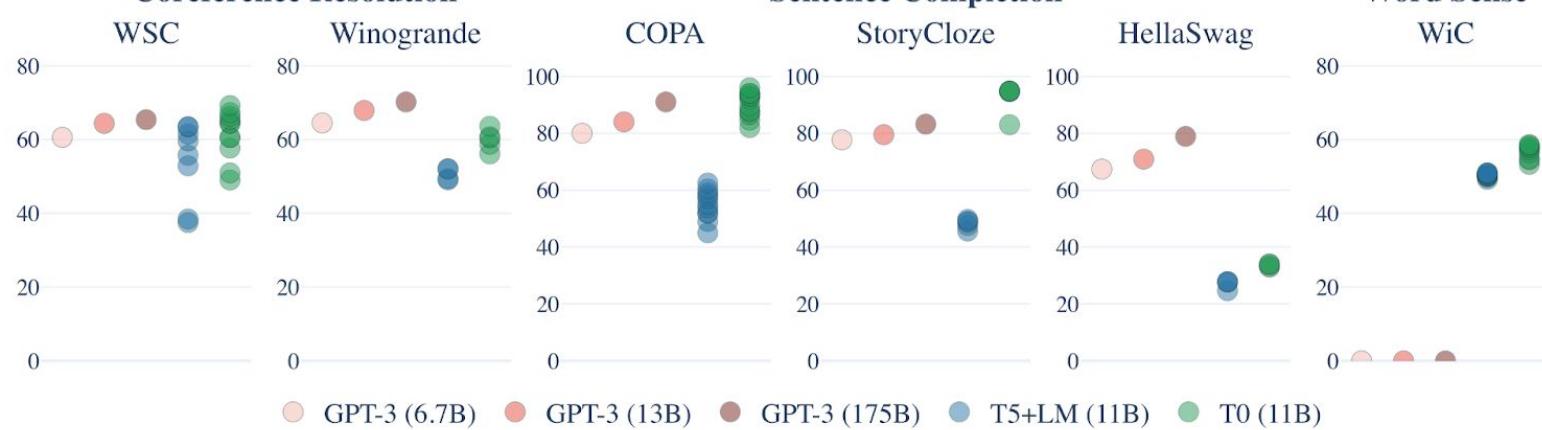
First, please read the article:  
{Document}  
Now, can you write me an  
extremely short abstract for it?

{Summary}

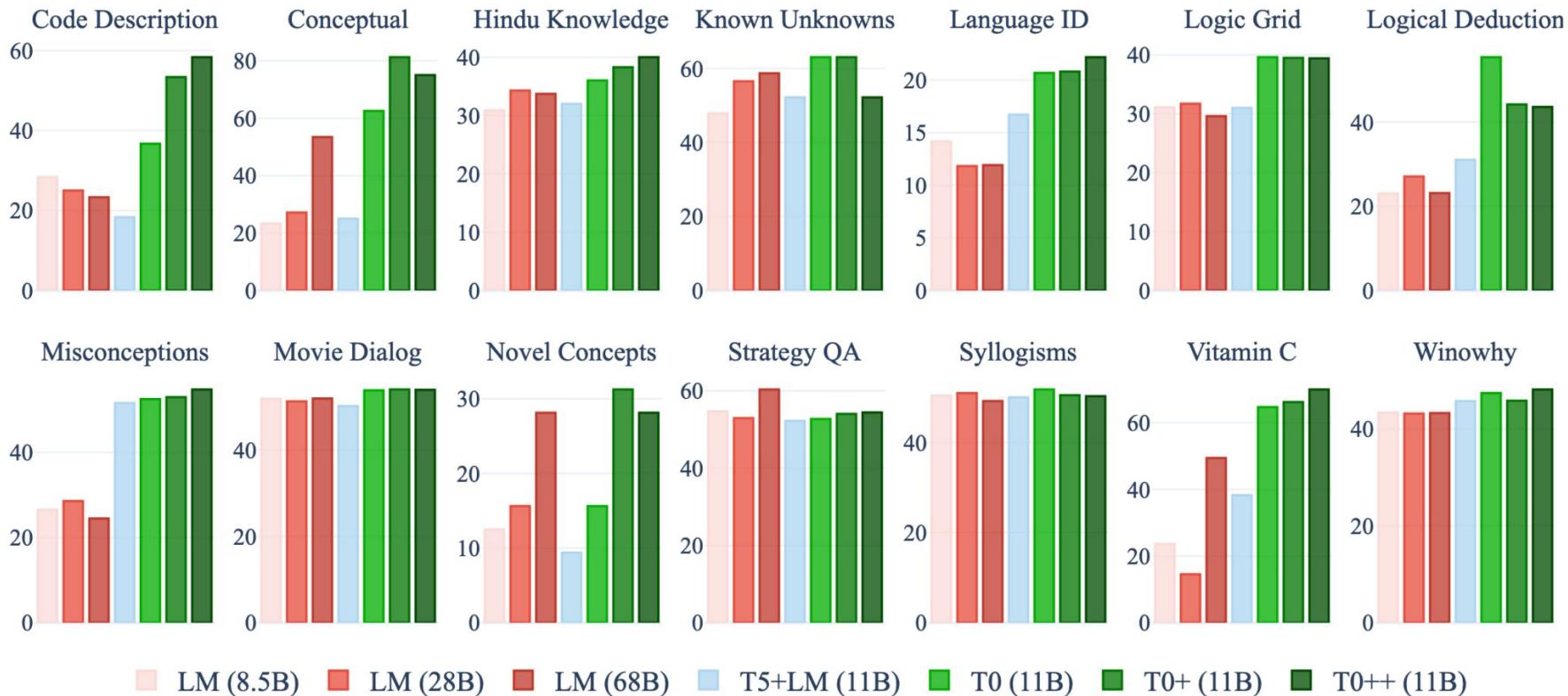
## Natural Language Inference



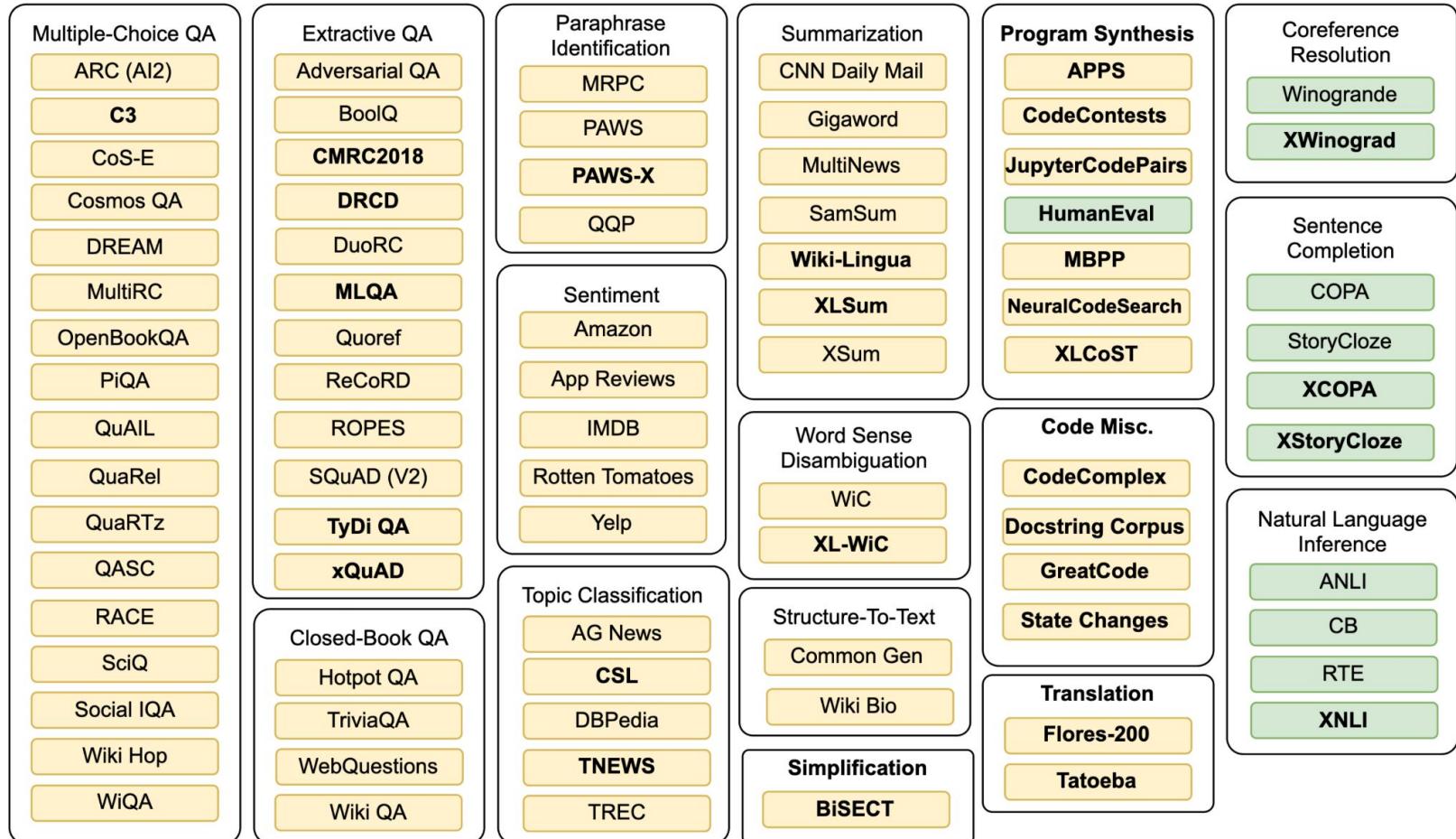
## Coreference Resolution



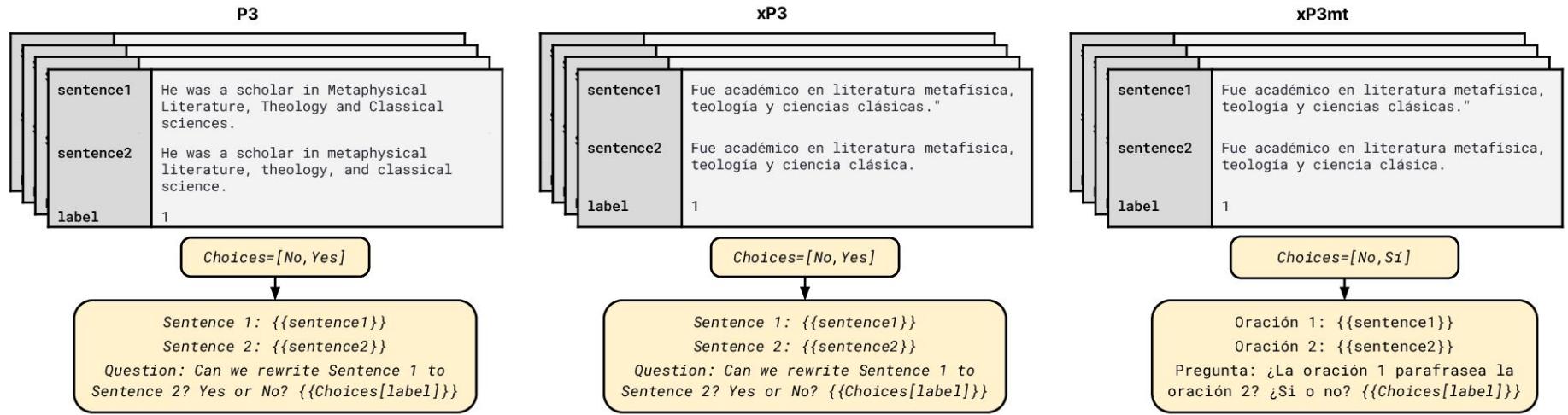
From "Multitask Prompted Training Enables Zero-Shot Task Generalization" by Sanh et al.



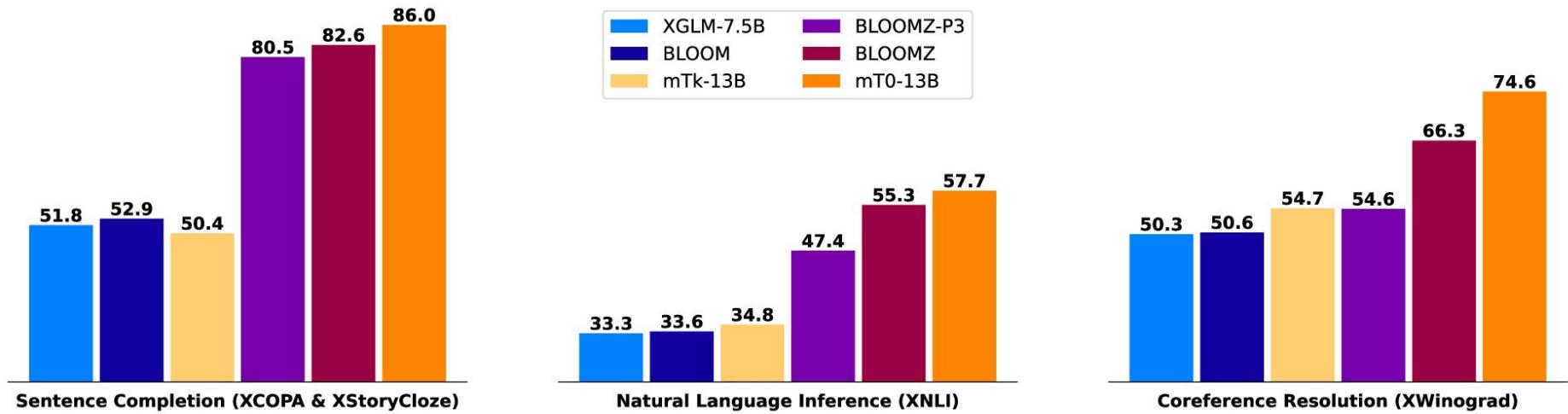
From “Multitask Prompted Training Enables Zero-Shot Task Generalization” by Sanh et al.



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

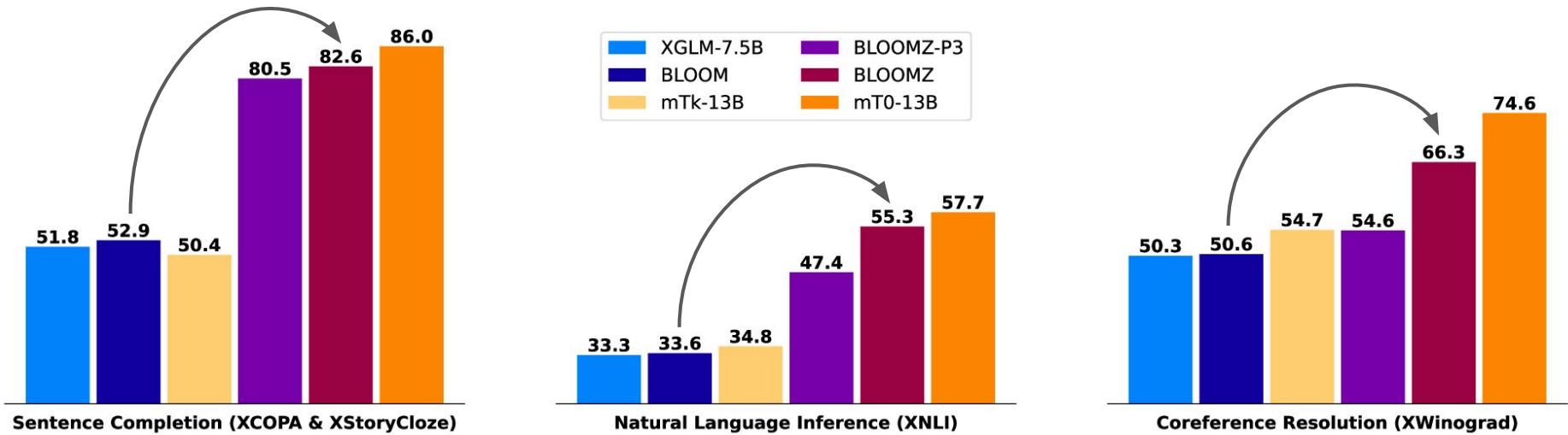


## Multilingual Multitask Generalization

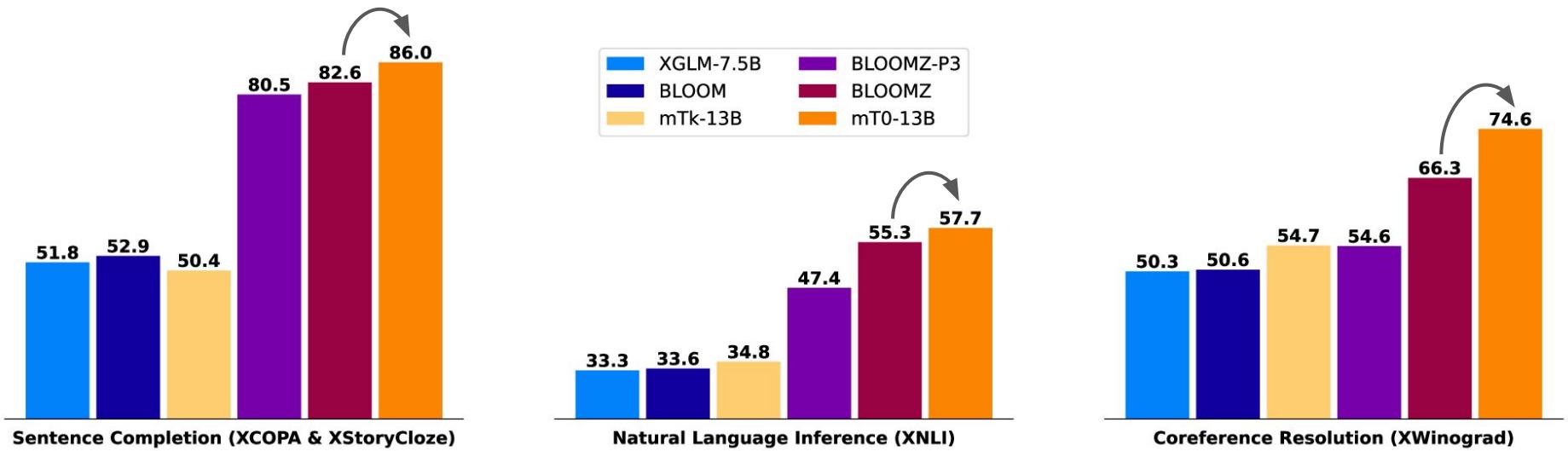


*From “Crosslingual Generalization through Multitask Finetuning” by Muennighoff et al.*

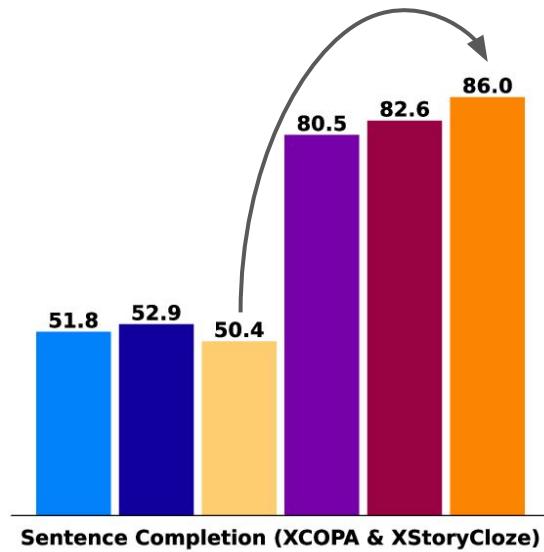
## Multilingual Multitask Generalization



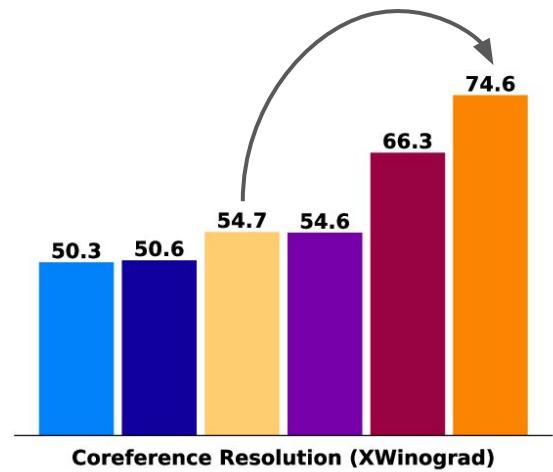
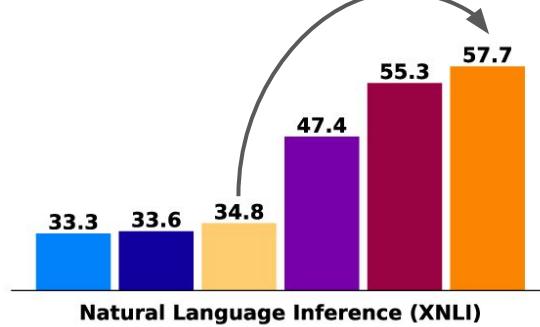
## Multilingual Multitask Generalization



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

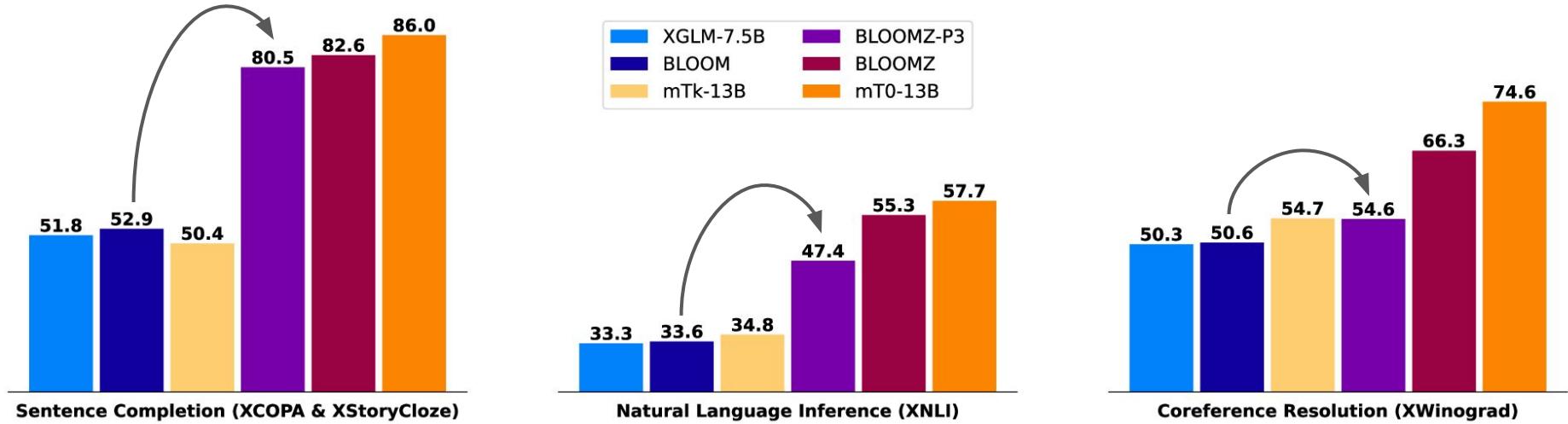


### Multilingual Multitask Generalization

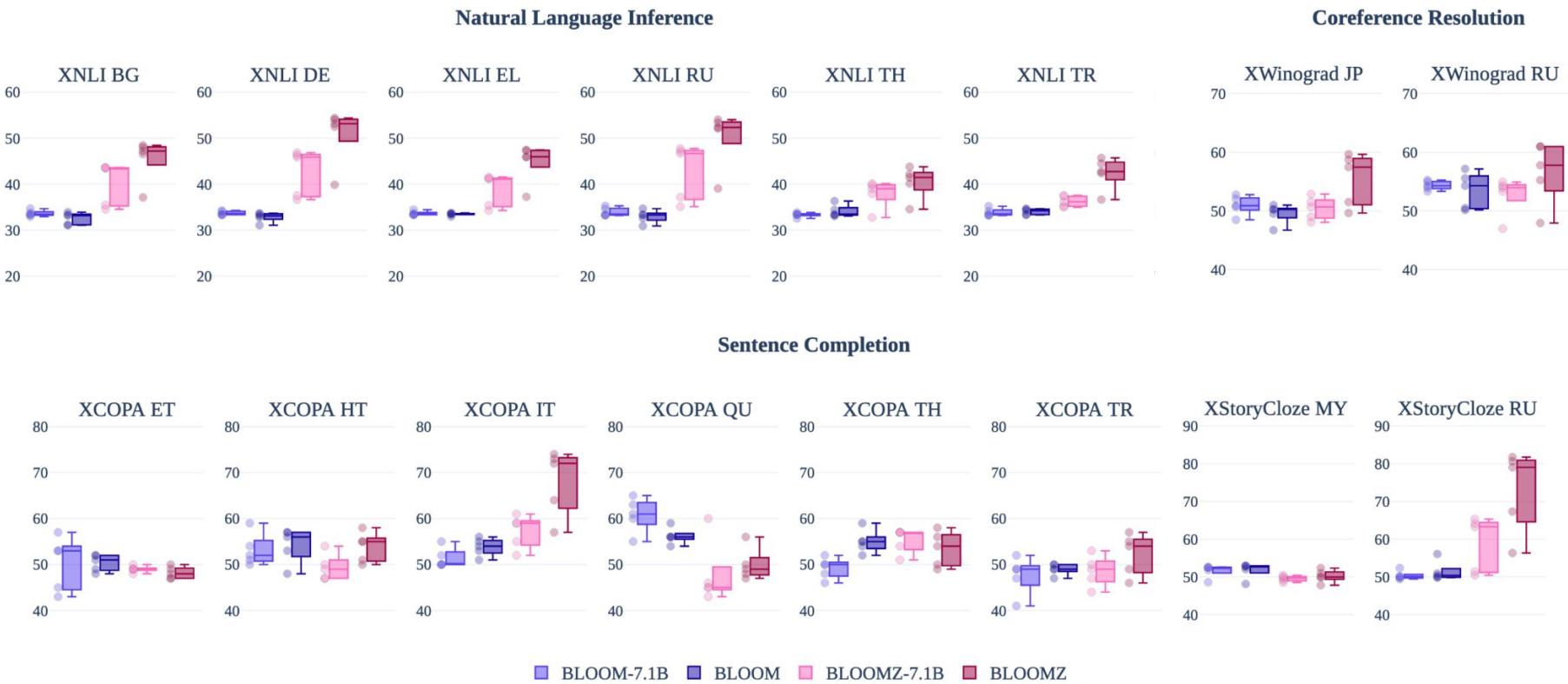


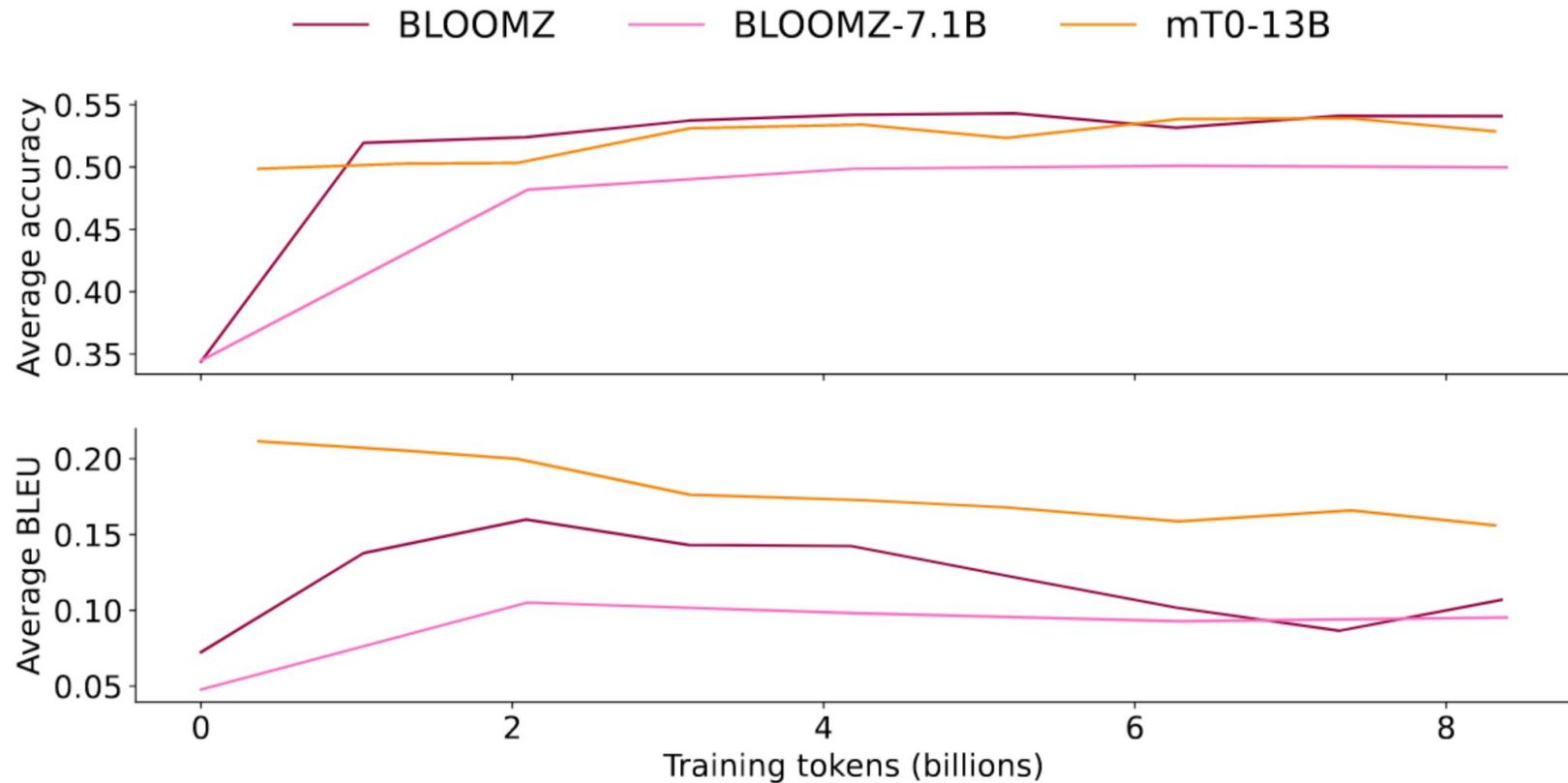
*From “Crosslingual Generalization through Multitask Finetuning” by Muennighoff et al.*

## Multilingual Multitask Generalization



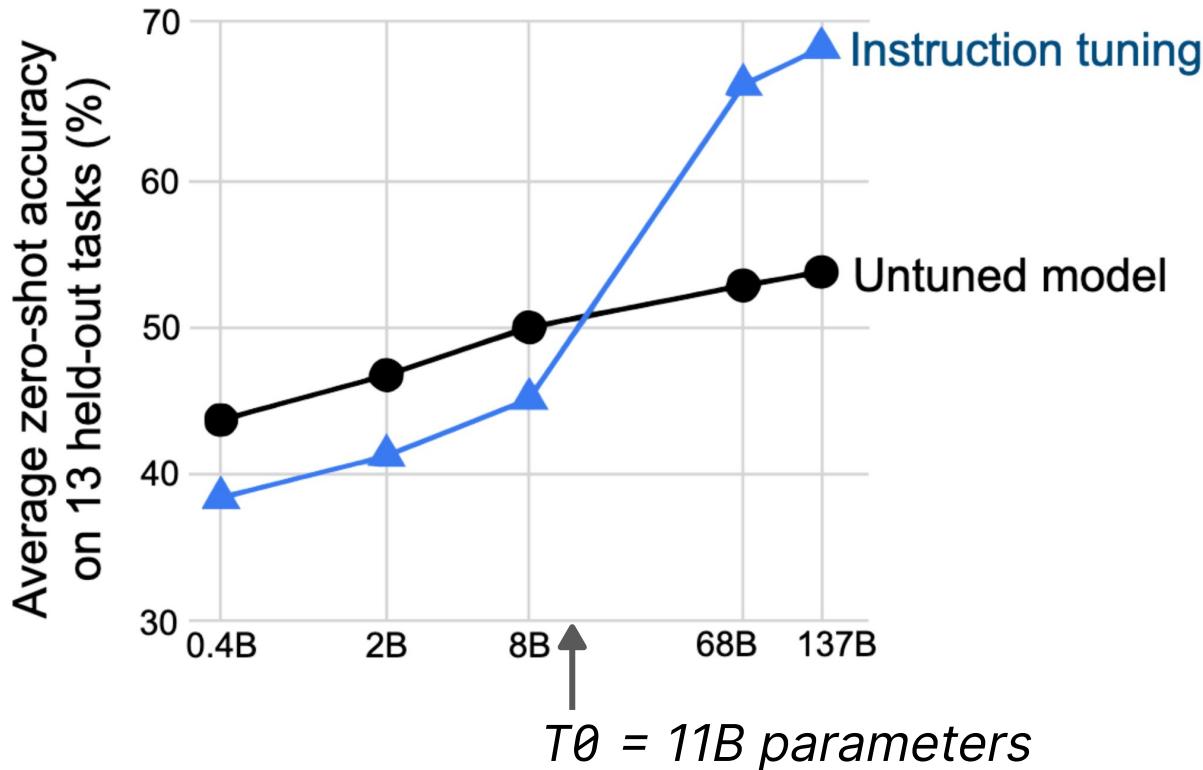
# *Performance on languages that were never intentionally trained on*

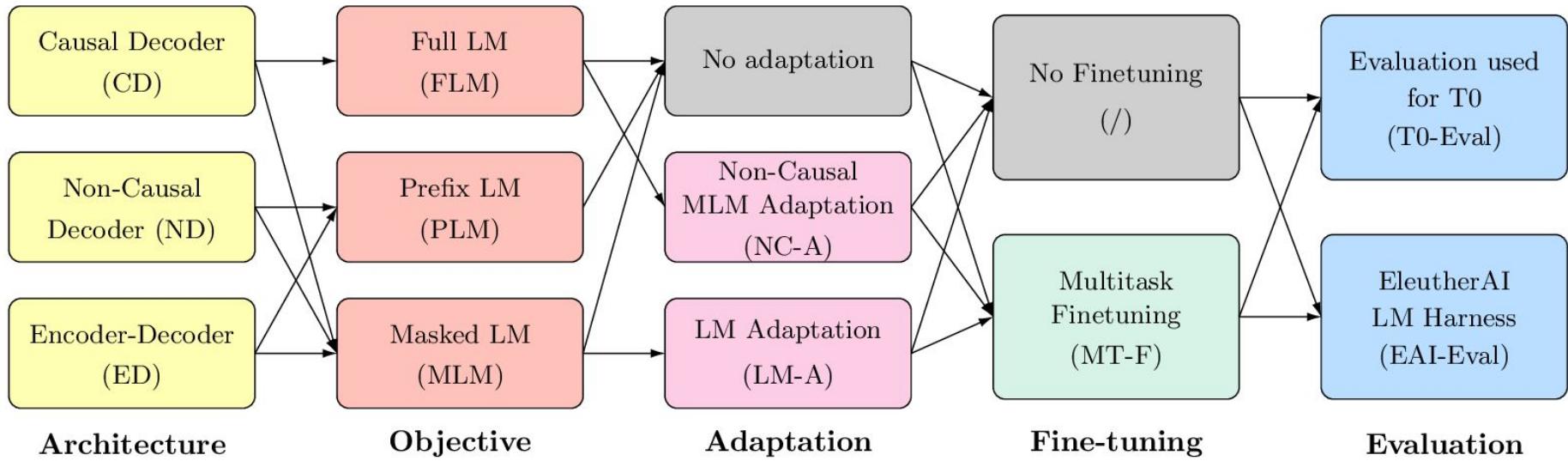




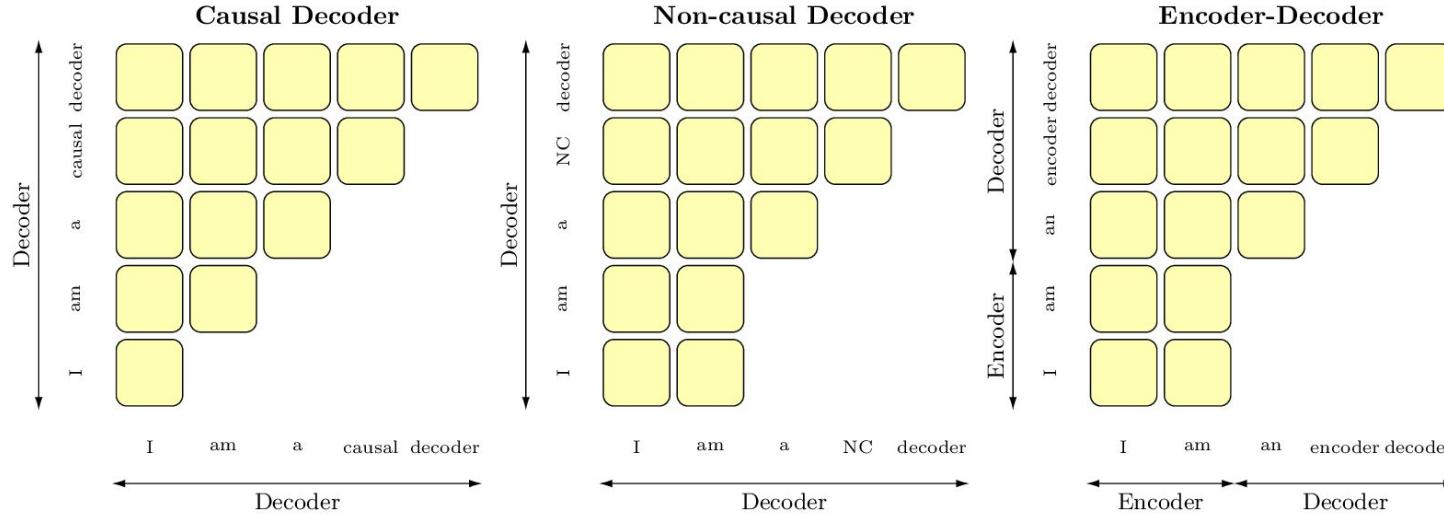
*From “Crosslingual Generalization through Multitask Finetuning” by Muennighoff et al.*

## Performance on held-out tasks





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



**Full Language Modeling**

May the force be with you

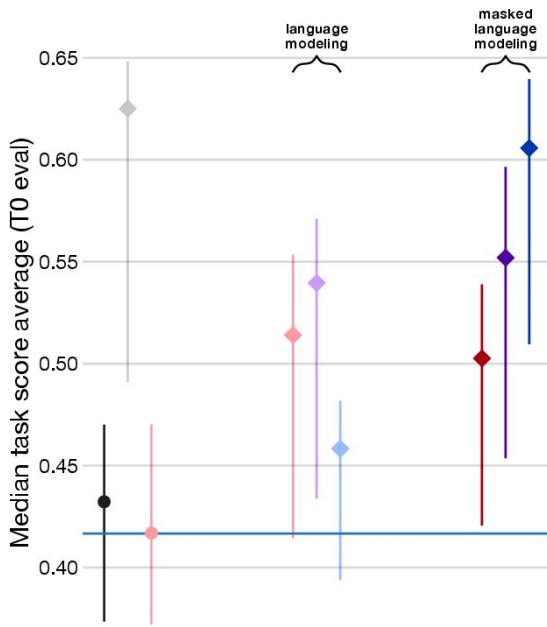
**Prefix Language Modeling**

May the force be with you

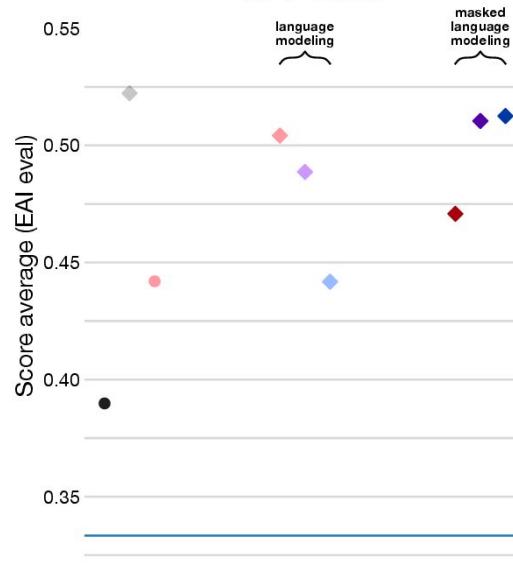
**Masked Language Modeling**

May the force be with you

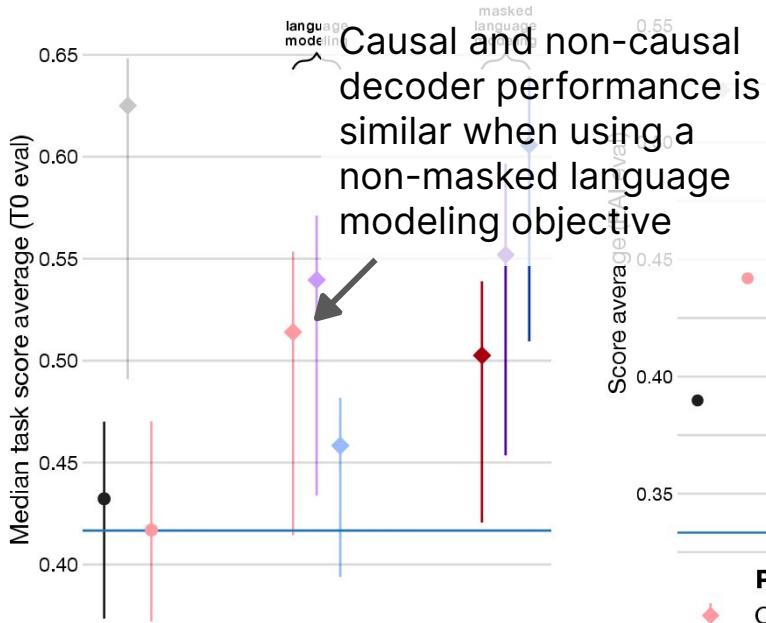
T0-Eval



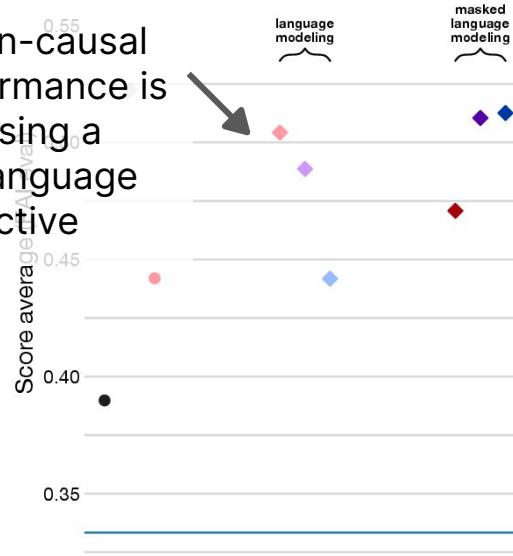
EAI-Eval



T0-Eval



EAI-Eval



Causal and non-causal decoder performance is similar when using a non-masked language modeling objective

### Baselines

- Random
- ◆ ED:MLM (1.3T) + ED:PLM (131B) [T5-LM]
- ◆ ED:MLM (1.3T) + ED:PLM (131B) + ED:MTF (13B) [T0]
- ◆ CD:FLM (168B)

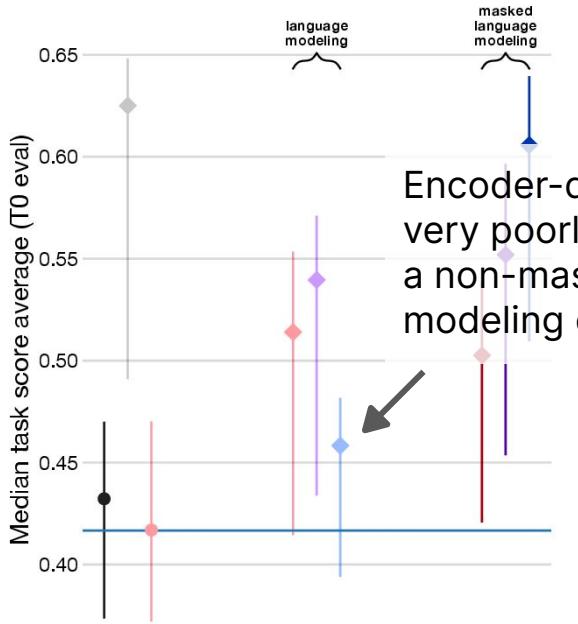
### Pretrained with LM

- ◆ CD:FLM (168B) + CD:MTF (13B)
- ◆ ND:PLM (168B) + ND:MTF (13B)
- ◆ ED:PLM (168B) + ED:MTF (13B)

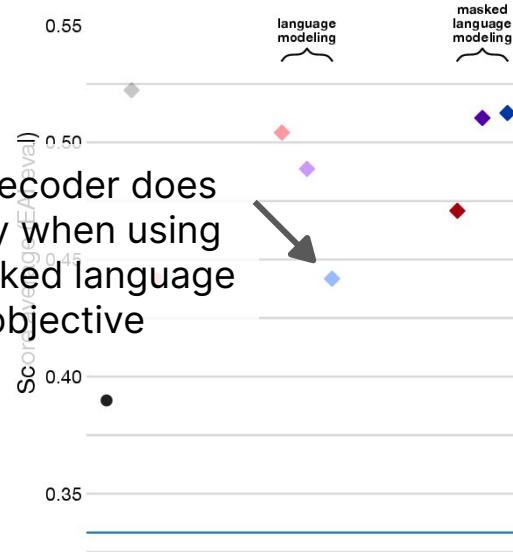
### Pretrained with MLM

- ◆ CD:MLM (168B) + CD:MTF (13B)
- ◆ ND:MLM (168B) + ND:MTF (13B)
- ◆ ED:MLM (168B) + ED:MTF (13B)

T0-Eval



EAI-Eval



Encoder-decoder does very poorly when using a non-masked language modeling objective

### Baselines

- Random
- ◆ ED:MLM (1.3T) + ED:PLM (131B) [T5-LM]
- ◆ ED:MLM (1.3T) + ED:PLM (131B) + ED:MTF (13B) [T0]
- ◆ CD:FLM (168B)

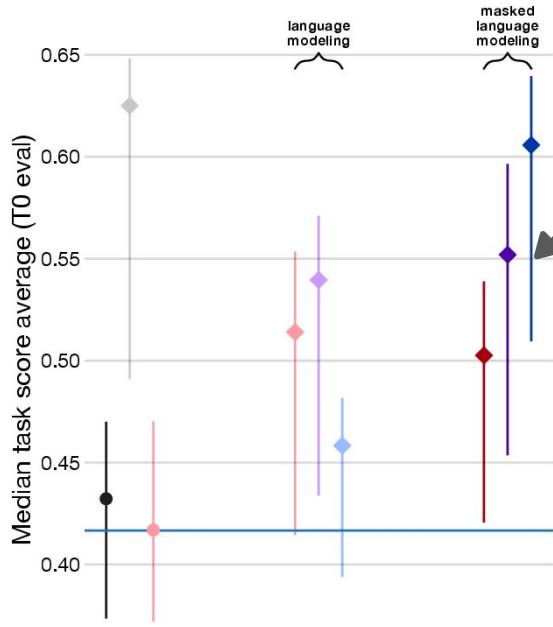
### Pretrained with LM

- ◆ CD:FLM (168B) + CD:MTF (13B)
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- ◆ ED:PLM (168B) + ED:MTF (13B)

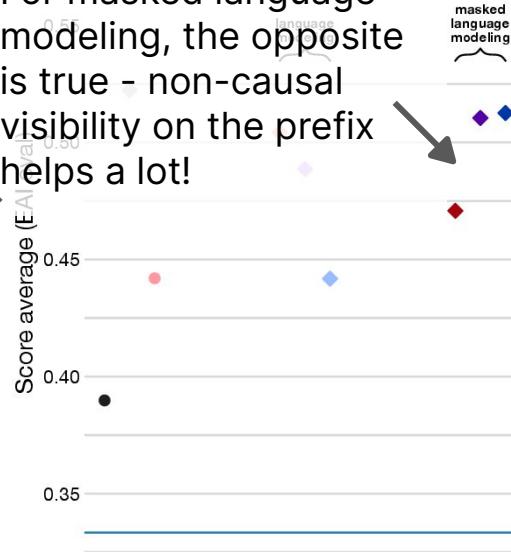
### Pretrained with MLM

- ◆ CD:MLM (168B) + CD:MTF (13B)
- ◆ ND:MLM (168B) + ND:MTF (13B)
- ◆ ED:MLM (168B) + ED:MTF (13B)

## T0-Eval



For masked language modeling, the opposite is true - non-causal visibility on the prefix helps a lot!



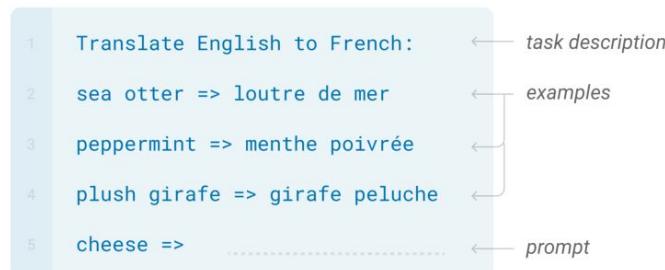
## Zero-shot

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



## Few-shot

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

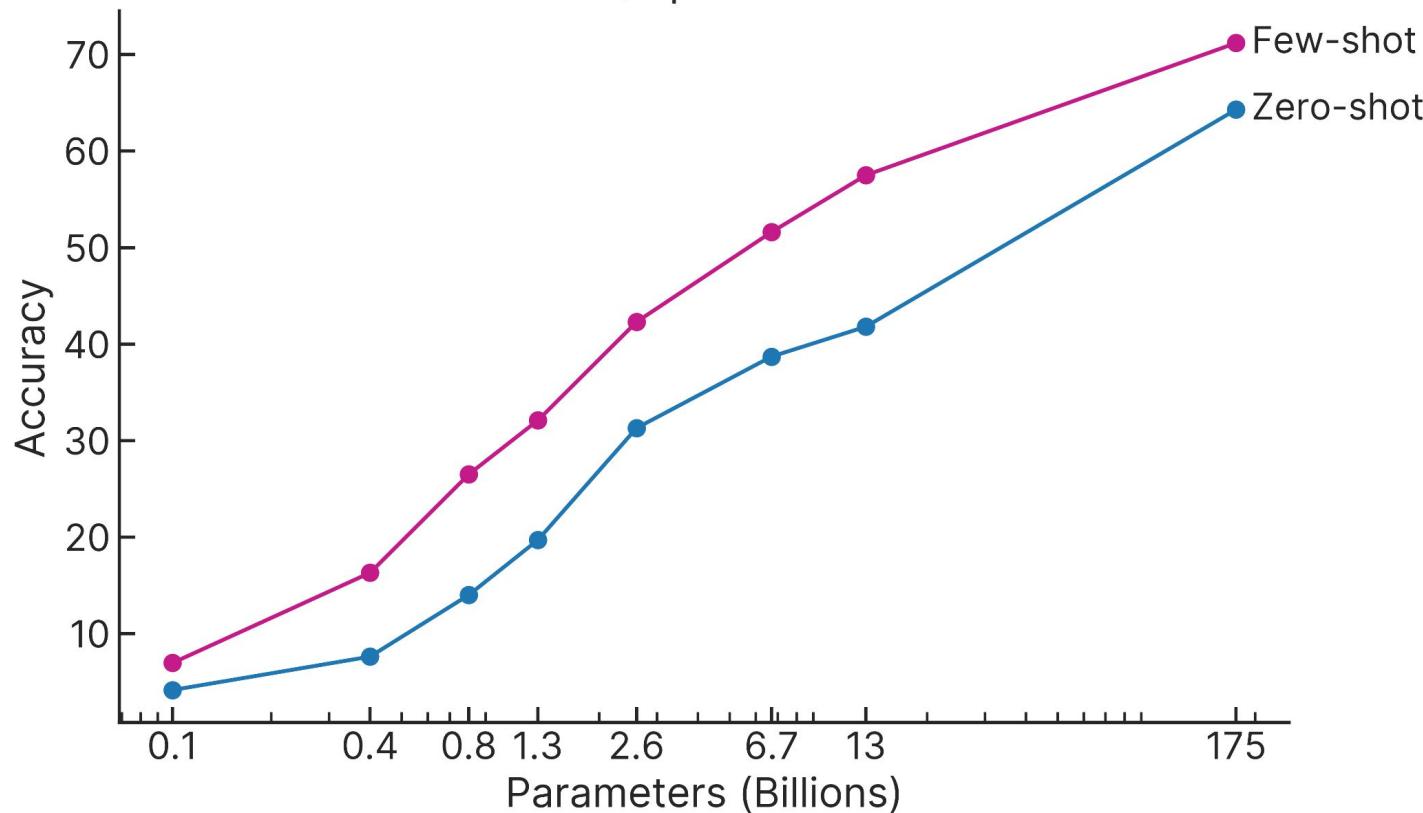


## Fine-tuning

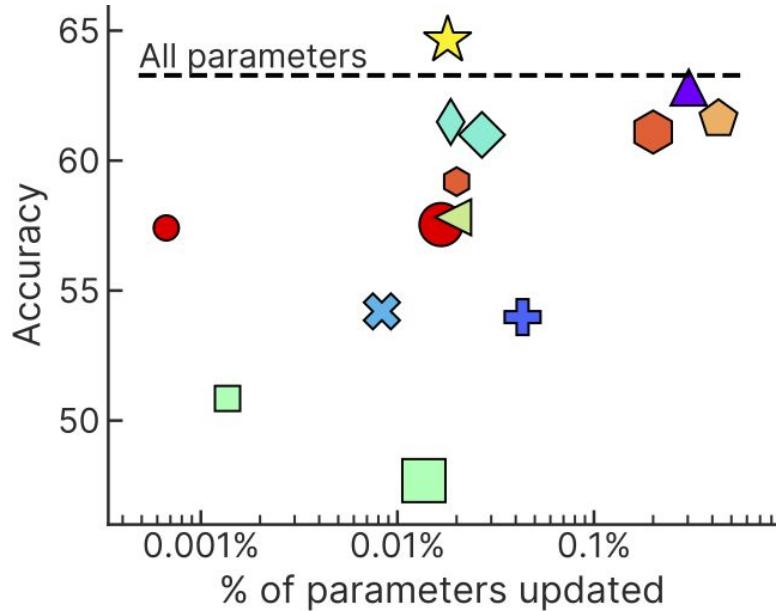
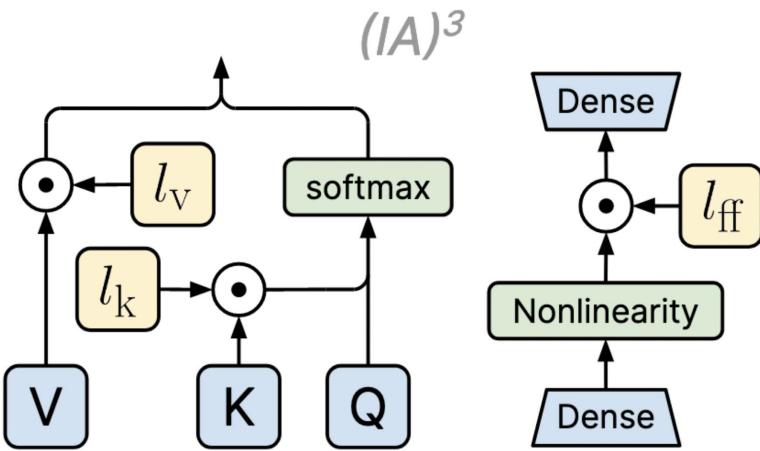
The model is trained via repeated gradient updates using a large corpus of example tasks.



## TriviaQA performance

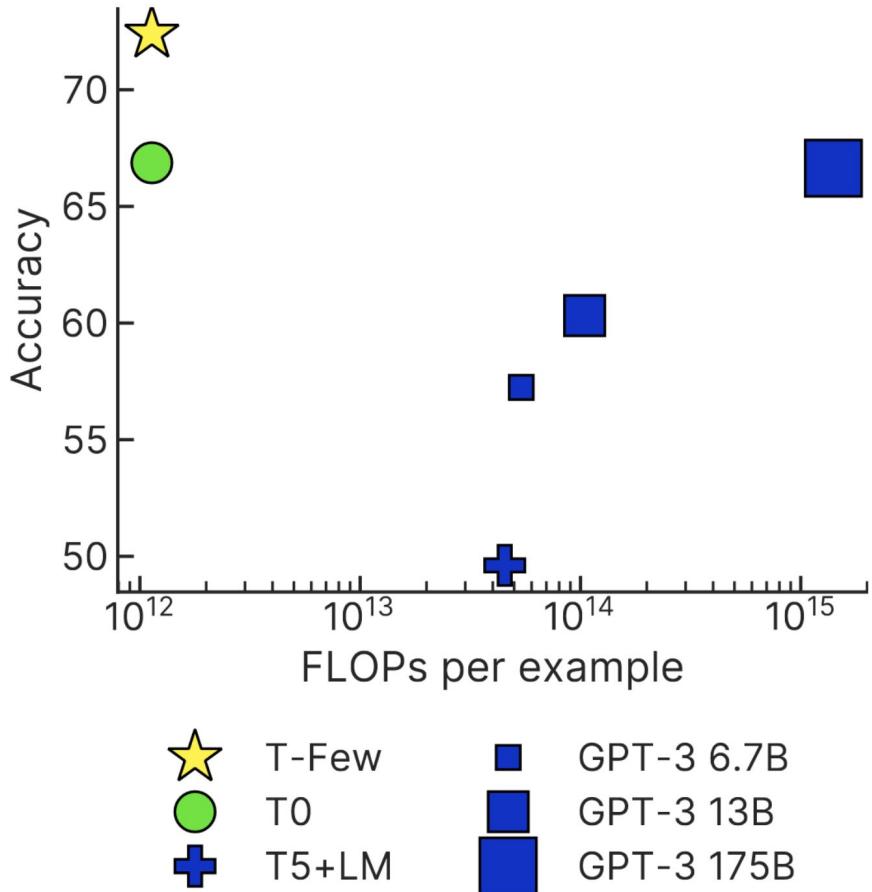


From "Language Models are Few-Shot Learners" by Brown et al.



- $\star$   $(IA)^3$
- $\blacktriangle$  LoRA
- $\textcolor{blue}{+}$  BitFit
- $\textcolor{lightblue}{\times}$  Layer Norm
- $\diamond$  Compacter
- $\textcolor{lightcyan}{\diamond}$  Compacter++
- Prompt Tuning
- △ Prefix Tuning
- Adapter
- FISH Mask
- Intrinsic SAID

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!