

Language Models are Unsupervised Multi-task Learners

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Qutline

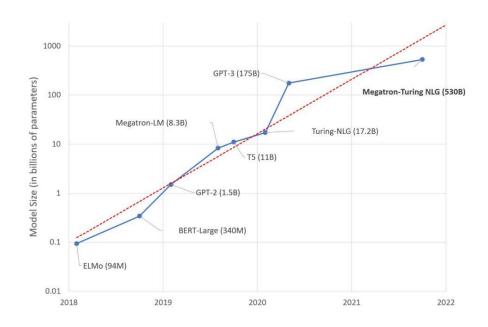
- Overview & Background
- Architecture
- Training & Data
- Experiments
- Limitations
- Discussion

Est. time: 15 min talk + 15 min discussion



Overview & Background

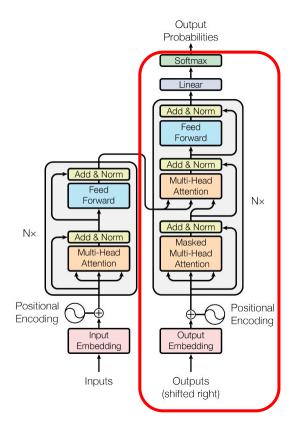
- Lack of Generalization of SOTA.
- Self-supervised learning.
- Respond to BERT paper.
- Zero-shot transfer.





Architecture

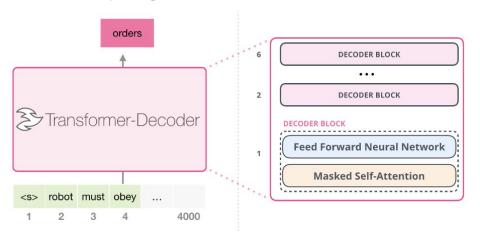


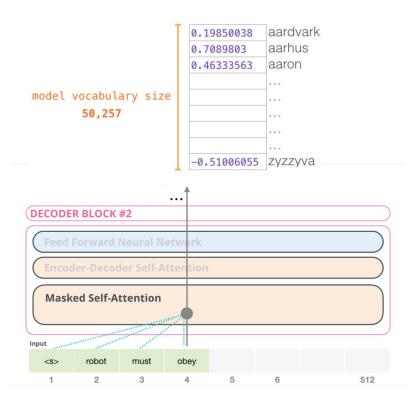




Architecture

- Transformer Decoder
 - 12 ~ 48 layers
- Masked Self-Attention
 - Compute efficient
- Sampling

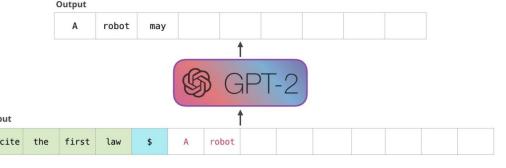




Objective

- Language modeling objective
 - i.e. NLL, MLE objective
 - "Predict the next word"
- vs. MLM (BERT)

$$p(x) = \prod_{i=1}^n p(s_n|s_1,...,s_{n-1})$$
 recite the first law



$$L_1(\mathcal{U}) = \sum_{i} \log P(u_i|u_{i-k}, \dots, u_{i-1}; \Theta)$$



Architecture

- 4 model variants.
- Compare to GPT-1
 - 10x parameters (XL model)
 - Vocabulary: $40k \rightarrow 50k$
 - Context size: 512 → 1024
 - Batch size: $64 \rightarrow 512$
 - Layer norm position
 - Weight initialization

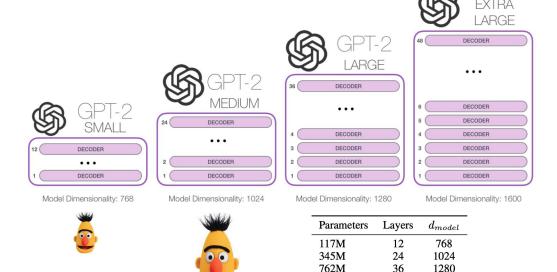


Table 2. Architecture hyperparameters for the 4 model sizes.

48

1600

1542M



- Data plays a big role.
- WebText
 - ~ 8 million web pages scraped and filtered from reddit.
 - (> 3 "karma" upvote score)
 - wikipedia explicitly filtered out.
- 40GB of data.
- Naturally occurring task-related data



"I'm not the cleverest man in the world, but like they say in French: Je ne suis pas un imbecile [I'm not a fool].

In a now-deleted post from Aug. 16, Soheil Eid, Tory candidate in the riding of Joliette, wrote in French: "Mentez mentez, il en restera toujours quelque chose," which translates as, "Lie lie and something will always remain."

"I hate the word 'perfume," Burr says. 'It's somewhat better in French: 'parfum.'

If listened carefully at 29:55, a conversation can be heard between two guys in French: "-Comment on fait pour aller de l'autre coté? -Quel autre coté?", which means "- How do you get to the other side? - What side?".

If this sounds like a bit of a stretch, consider this question in French: As-tu aller au cinéma?, or Did you go to the movies?, which literally translates as Have-you to go to movies/theater?

"Brevet Sans Garantie Du Gouvernement", translated to English: "Patented without government warranty".

Table 1. Examples of naturally occurring demonstrations of English to French and French to English translation found throughout the WebText training set.

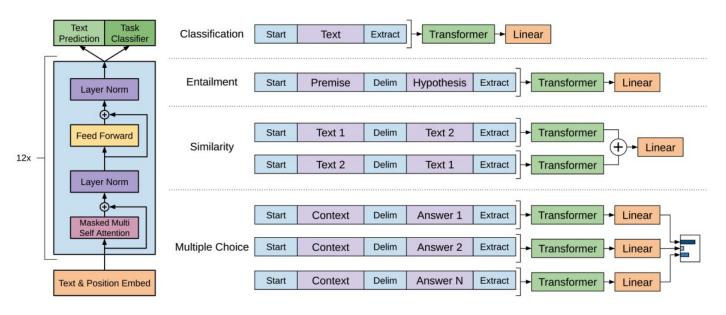
Better Language Models and Their Implications

We've trained a large-scale unsupervised language model which generates coherent paragraphs of text, achieves state-of-the-art performance on many language modeling benchmarks, and performs rudimentary reading comprehension, machine translation, question answering, and summarization—all without task-specific training.





Input Formulation (GPT-1)



Input Formulation (GPT-2)

NMT: "Translate to french," < English text>, < French text>.

QA: "Answer the question," < Document>, < Question>, < Answer>.

SUMM: < Document > "TL; DR:" < Summarization >

... ANTHROP\C

Prompt Engineer and Librarian

APPLY FOR THIS JOB



Experiments

- Zero-shot domain transfer (Language modeling tasks).
 - LAMBADA, CBT...
- Zero-shot NLU tasks.
 - MT, QA, CLS, WSC ...
- Misc.
 - Underfit WebText.
 - Model Memorization.



Experiments

Language Models are Unsupervised Multitask Learners

	LAMBADA	LAMBADA	CBT-CN	CBT-NE	WikiText2	PTB	enwik8	text8	WikiText103	1BW
	(PPL)	(ACC)	(ACC)	(ACC)	(PPL)	(PPL)	(BPB)	(BPC)	(PPL)	(PPL)
SOTA	99.8	59.23	85.7	82.3	39.14	46.54	0.99	1.08	18.3	21.8
117M	35.13	45.99	87.65	83.4	29.41	65.85	1.16	1.17	37.50	75.20
345M	15.60	55.48	92.35	87.1	22.76	47.33	1.01	1.06	26.37	55.72
762M	10.87	60.12	93.45	88.0	19.93	40.31	0.97	1.02	22.05	44.575
1542M	8.63	63.24	93.30	89.05	18.34	35.76	0.93	0.98	17.48	42.16

Table 3. Zero-shot results on many datasets. No training or fine-tuning was performed for any of these results. PTB and WikiText-2 results are from (Gong et al., 2018). CBT results are from (Bajgar et al., 2016). LAMBADA accuracy result is from (Hoang et al., 2018) and LAMBADA perplexity result is from (Grave et al., 2016). Other results are from (Dai et al., 2019).



Experiments

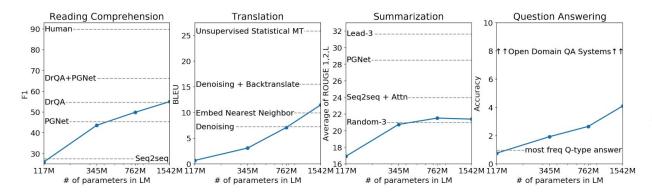


Figure 1. Zero-shot task performance of WebText LMs as a function of model size on many NLP tasks. Reading Comprehension results are on CoQA (Reddy et al., 2018), translation on WMT-14 Fr-En (Artetxe et al., 2017), summarization on CNN and Daily Mail (See et al., 2017), and Question Answering on Natural Questions (Kwiatkowski et al., 2019). Section 3 contains detailed descriptions of each result.

Figure 3. Performance on the Winograd Schema Challenge as a function of model capacity.

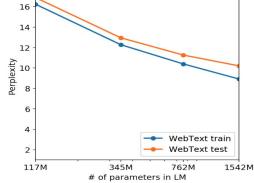


Results

- GPT-2 improved the then existing state-of-the-art for **7 out of 8** language modelling datasets in zero shot setting.

"Larger the better"

Underfitting on WebText.



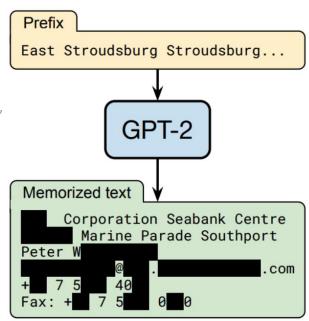
- The performance increases in log-linear fashion as model scales.
- Building even larger language models would reduce the perplexity and make language models better at natural language understanding.







- Limited performance
 - E.g. Summarization
 - Inefficiencies of uni-directional representations.
 - ! given the model size, data, and compute
 - "The zero-shot performance of GPT-2 is still far from usable."
- Memorization / Safety issues
 - Verbatim memorization of private or IP information
 - (Carlini et al. 2020): "We find that at least 0.1% percent of its text generations contain long verbatim strings in its train set."





Discussion

- Anything I missed? Corrections?
- Discussion points
 - How do we understand and distill what's learned by the model?
 - ...