

Workshop

NLP using Generative Al

Module 3

Speaker

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Building a startup (stealth mode) Ex - Head of AI and R&D, funtoot

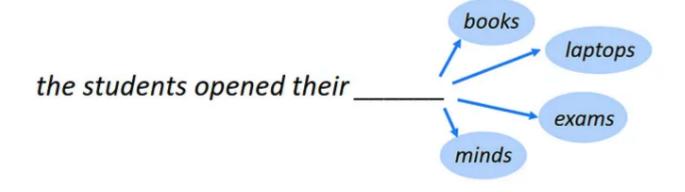


Module 3

Diving into GPT



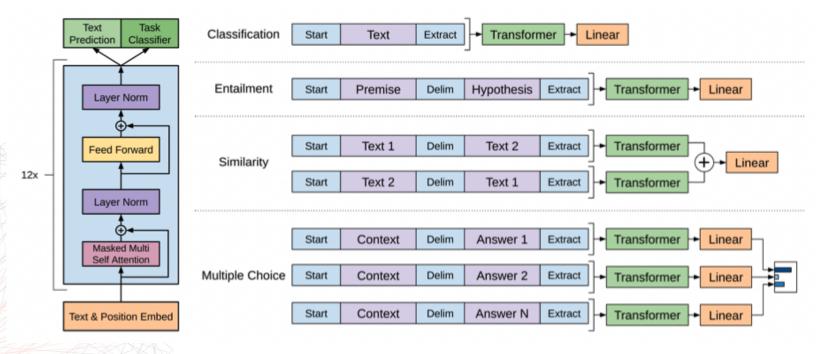
Sophisticated Next Word Predictors





Improving Language Understanding by Generative Pre-Training

followed by (task-agnostic) Discriminative Fine-Tuning



https://s3-us-west-2.amazonaws.com/openai-assets/research-covers/language-unsupervised/language_understanding_paper.pdf

GPT-2: Unsupervised Multitask Learner



- Pre-training transformer language models
- Task specific fine-tuning
 - Eliminating need of task specific architectures
- Limitation: while architecture is task-agnostic, still task specific datasets and fine-tuning needed
 - → For each task, large high quality dataset needed. May not be possible always
 - → Out of distribution generalization decays with
 - Expressiveness of the model
 - Narrowness of the training distribution
- Humans to do not require large training datasets to learn most natural language tasks
 - ♦ Brief directive in natural language or at-most a tiny number of demonstrations is enough
 - This allows humans to seemingly mix together or switch between various tasks and skills

GPT-3: Few-shot Learner



The three settings we explore for in-context learning

Zero-shot

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

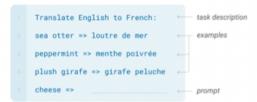
One-shot

In addition to the task description, the model sees a single example 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.



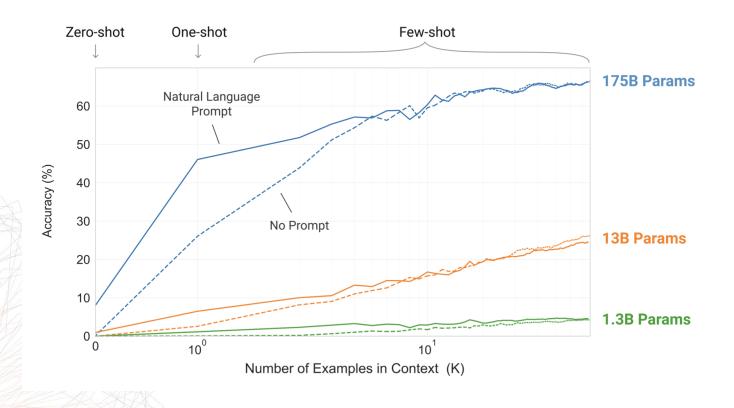
Traditional fine-tuning (not used for GPT-3)

Fine-tuning

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

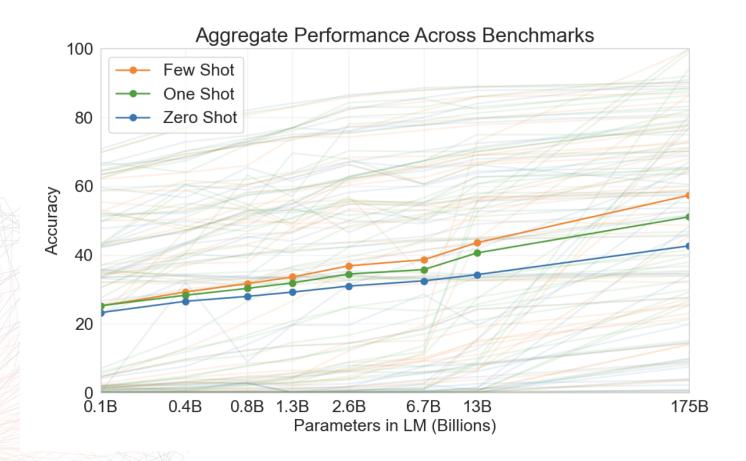


https://arxiv.org/abs/2005.14165



GPT-3: Few-shot Learner





Performance: Power Law



