

The scientist
named the
population, after their
distinctive horn,
Ovid's Unicorn.

GPT2: Zero Shot的崛起

Language Model

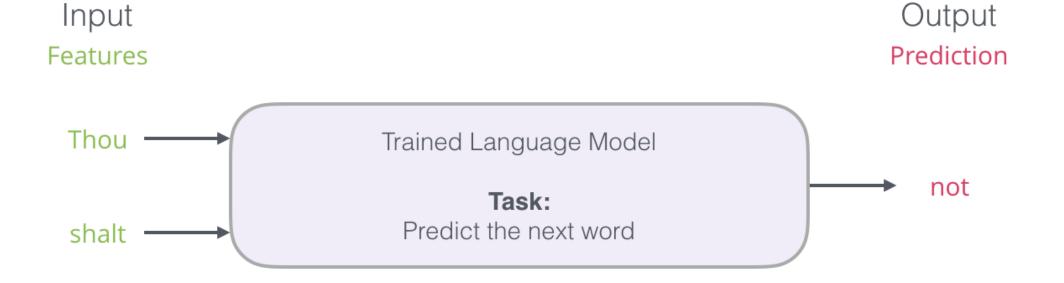
input/feature #1

input/feature #2

output/label

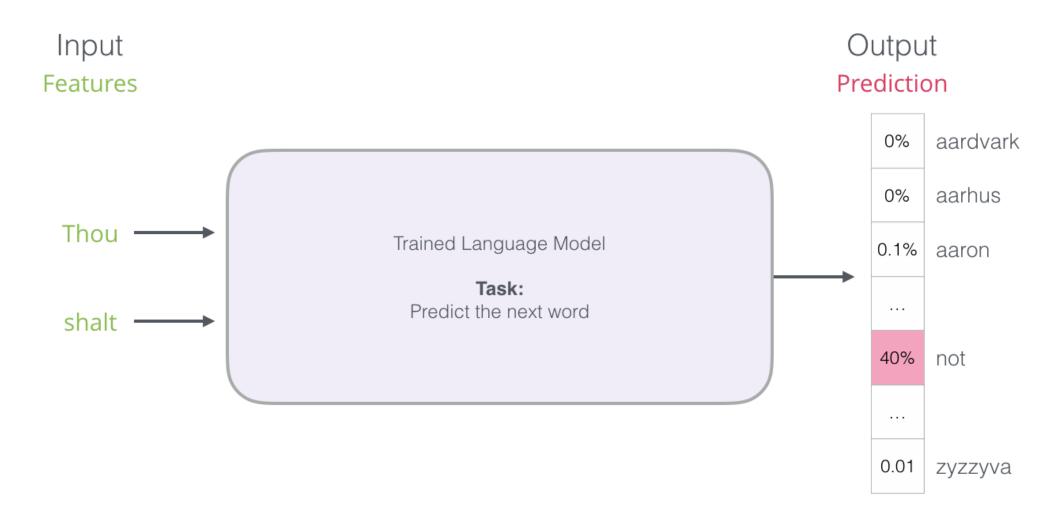
Thou shalt



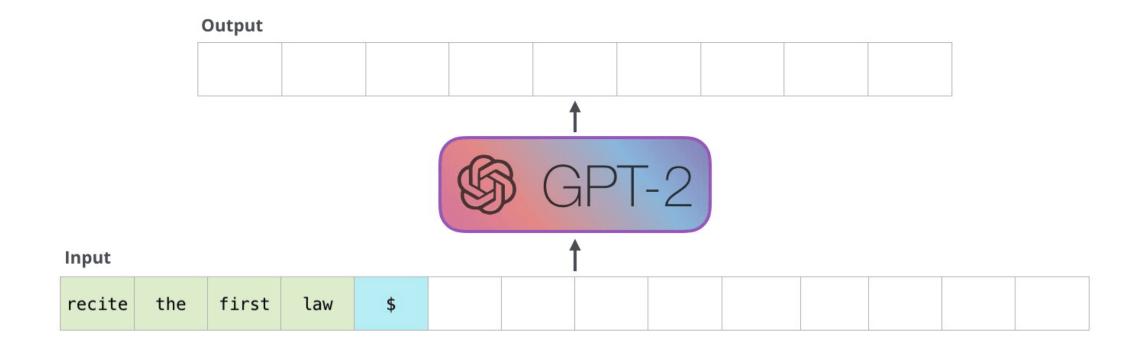


https://jalammar.github.io/illustrated-word2vec/

Language Model



Auto Regressive



Learning Objectives and Concepts

- Task Conditioning
- Zero Shot Learning and Zero Short Task Transfer

Task Conditioning

Training objective of language model is formulated as:

P(output|input)

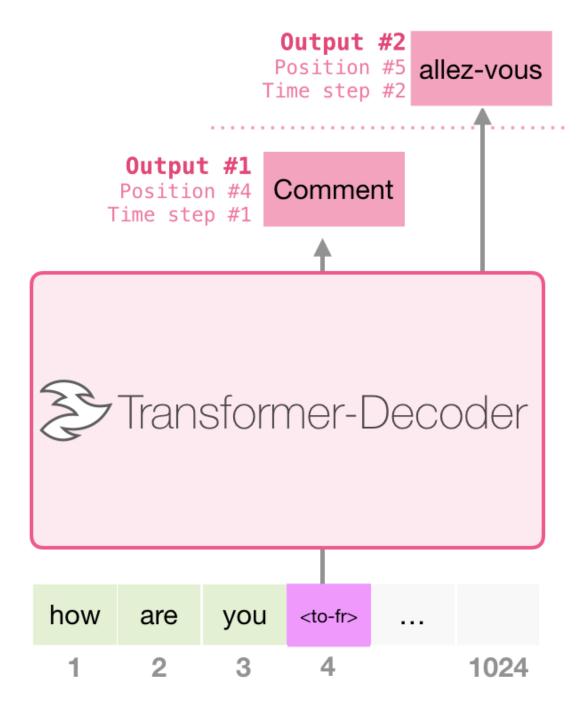
However, GPT-2 aimed at learning multiple tasks using the same unsupervised model. To achieve that, the learning objective should be modified to:

P(output|input,task)

This modification is known as task conditioning, where the model is expected to produce different output for same input for different tasks. Some models implement task conditioning at an architectural level where the model is fed both, the input and the task. For language models, the output, input and task, all are sequences of natural language.

Training Dataset

I	am	а	student	<to-fr></to-fr>	je	suis	étudiant
let	them	eat	cake	<to-fr></to-fr>	Qu'ils	mangent	de
good	morning	<to-fr></to-fr>	Bonjour				



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Positronic brain

From Wikipedia, the free encyclopedia

This article is about a fictional technological device. For the manufacturing company based in Springfield, Missouri, see Postronic (company).



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A positronic brain is a fictional technological device, originally conceived by science fiction writer Isaac Asimov, [137] It functions as a central processing unit (CPU) for robots, and, in some unspecified way, provides them with a form of consciousness recognizable to humans. When Asimov wrote his first robot stories in 1939 and 1940. the positron was a newly discovered particle, and so the buzz word positronic added a contemporary gloss of popular science to the concept. The short story "Runaround", by Asimov, elaborates on the concept, in the context of his fictional Three Laws of Robotics.

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Conceptual overview [658]

Asimov remained vague about the technical details of positronic brains except to assert that their substructure was formed from an alloy of platinum and indium. They were said to be vulnerable to radiation and apparently involve a type of volatile memory (since robots in storage required a power source keeping their brains "alive"). The focus of Asimon's stories was directed more towards the software of robots - such as the Three Laws of Robotics - than the hardware in which it was implemented. although it is stated in his stories that to create a positronic brain without the Three Laws, it would have been necessary to spend years redesigning the fundamental approach towards the brain itself

Within his stories of robotics on Earth and their development by U.S. Robots, Asimov's positronic brain is less of a plot device and more of a technological item worthy of

A positronic brain cannot ordinarily be built without incorporating the Three Laws; any modification thereof would drastically modify robot behavior. Behavioral dilemmas resulting from conflicting potentials set by inexperienced and/or malicious users of the robot for the Three Laws make up the bulk of Asimon's stories concerning robots. They are resolved by applying the science of logic and psychology together with mathematics, the supreme solution finder being Dr. Susan Calvin, Chief Robopsychologist of U.S. Robots.

The Three Laws are also a bottleneck in brain sophistication. Very complex brains designed to handle world economy interpret the First Law in expanded sense to include humanity as opposed to a single human; in Asimov's later works like Robots and Empire this is referred to as the "Zeroth Law". At least one brain constructed as a calculating machine, as opposed to being a robot control circuit, was designed to have a flexible, childlike personality so that it was able to pursue difficult problems without the Three Laws inhibiting it completely. Specialized brains created for overseeing world economics were stated to have no personality at all.

Under specific conditions, the Three Laws can be obvisted, with the modification of the actual robotic design.

- . Robots that are of low enough value can have the Third Law deleted; they do not have to protect themselves from harm, and the brain size can be reduced by half. . Rebots that do not require orders from a human being may have the Second Law deleted, and therefore require smaller brains again, providing they do not require
- . Robots that are disposable, cannot receive orders from a human being and are not able to harm a human, will not require even the First Law. The sophistication of

positronic circuitry renders a brain so small that it could comfortably fit within the skull of an insect.

Robots of the latter type directly parallel contemporary industrial robotics practice, though real-life robots do contain safety sensors and systems, in a concern for human safety (a weak form of the First Law: the robot is a safe tool to use, but has no "sudgment", which is implicit in Asimov's own stories).

In Allen's trilogy [set]

Several robot stories have been written by other authors following Asimov's death. For example, in Roger MacBride Alleri's Caliban trilogy, a Spacer roboticist called Gubber Anshaw invents the gravitonic brain. It offers speed and capacity improvements over traditional positronic designs, but the strong influence of tradition make rebotics labs reject Anshaw's work. Only one reboticist, Fredda Leving, chooses to adopt gravitorics, because it offers her a blank slate on which she could explore alternatives to the Three Laws. Because they are not dependent upon centuries of earlier research, gravitonic brains can be programmed with the standard Laws, variations of the Laws, or even empty pathways which specify no Laws at alt.



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Conceptual overview :

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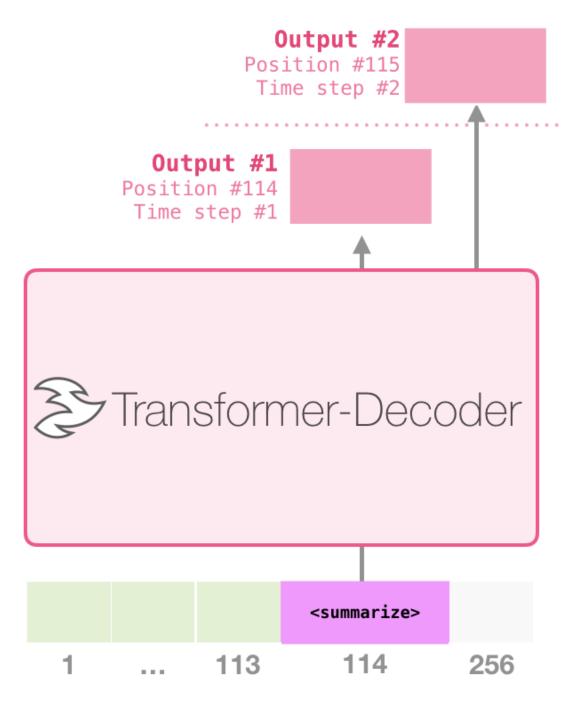
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Training Dataset

Article #1 tokens			<summarize></summarize>		Article #1 Summary	
Article #2 tokens	<summarize></summarize>	Article #2 Summary		padding		
Article #3	<summari< th=""><th>ze></th><th>Article #3 Summary</th></summari<>	ze>	Article #3 Summary			



Zero Shot Learning and Task Transfer

- An interesting capability of GPT 2 is zero shot task transfer. Zero shot learning is a special case of zero shot task transfer where no examples are provided at all and the model understands the task based on the given instruction. Instead of rearranging the sequences, as was done for GPT-1 for fine-tuning, input to GPT-2 was given in a format which expected the model to understand the nature of task and provide answers. This was done to emulate zero-shot task transfer behavior.
- E.g. for English to French translation task, the model was given an English sentence followed by the word French and a prompt (:). The model was supposed to understand that it is a translation task and give French counterpart of English sentence.

Model architecture

- GPT-2 had 48 layers and used 1600 dimensional vectors for word embedding.
- Larger vocabulary of 50,257 tokens was used.
- Larger batch size of 512 and larger context window of 1024 tokens were used.

```
class GPT2Config(PreTrainedConfig):
   Configuration for gpt2-base
   pretrained config archive map = CONFIG ARCHIVE MAP
   def init (
        self,
       vocab_size=50257,
       n positions=1024,
       n embd=768,
       n layer=12,
       n head=12,
       n inner=None,
       activation function="gelu new",
       resid pdrop=0.1,
       embd_pdrop=0.1,
       attn pdrop=0.1,
        layer norm epsilon=1e-5,
        initializer range=0.02,
        summary_type="cls_index",
        summary use proj=True,
        summary activation=None,
        summary proj to labels=True,
       summary first dropout=0.1,
       scale attn weights=True,
       use cache=True,
       bos token id=50256,
       eos token id=50256,
        scale_attn_by_inverse_layer_idx=False,
        reorder and upcast attn=False,
        **kwargs,
```

Model architecture

```
def construct(
        self,
        hidden states: Tuple[Tensor],
        layer past: Optional[Tuple[Tensor]] = None,
        attention mask: Optional[Tensor] = None,
        head_mask: Optional[Tensor] = None,
        encoder hidden states: Optional[Tensor] = None,
        encoder attention mask: Optional[Tensor] = None,
        use cache: Optional[bool] = False,
):
    residual = hidden states
    hidden states = self.ln 1(hidden states)
    attn outputs = self.attn(
        hidden states,
        layer past=layer past,
        attention mask=attention mask,
        head_mask=head_mask,
        use cache=use cache,
    attn output = attn outputs[0] # output attn: a, present, (attentions)
    outputs = attn outputs[1:]
     residual connection
    hidden_states = attn_output + residual
```

 Layer normalization was moved to input of each sub-block

Model architecture

```
for i, block in enumerate(self.h):
   if self.output hidden states:
       all hidden states = all hidden states + (hidden states,)
   outputs = block(
       hidden_states,
       layer_past=past_key_values[i],

    an additional layer

       attention mask=attention mask,
       head_mask=head_mask[i],
                                                                                   normalization was added after
       encoder hidden states=encoder hidden states,
       encoder attention mask=encoder attention mask,
       use cache=self.use cache,
                                                                                  final self-attention block.
   hidden_states = outputs[0]
   if self.use_cache:
       presents = presents + (outputs[1],)
   if self.output attentions:
       all_self_attentions = all_self_attentions + (outputs[2 if self.use_cache else 1],)
       if self.add cross attention:
           all_cross_attentions = all_cross_attentions + (outputs[3 if self.use_cache else 2],)
hidden_states = self.ln_f(hidden_states)
```

Special Scaled Initialization

```
# Reinitialize selected weights subject to the OpenAI GPT-2 Paper Scheme:

# > A modified initialization which accounts for the accumulation on the residual path with model depth. Scale

# > the weights of residual layers at initialization by a factor of 1/VN where N is the # of residual layers.

# > -- GPT-2 :: https://openai.com/blog/better-language-models/

#

# Reference (Megatron-LM): https://github.com/NVIDIA/Megatron-LM/blob/main/megatron/model/gpt_model.py

for name, p in module.named_parameters():
    if name == "c_proj.weight":
        # Special Scaled Initialization --> There are 2 Layer Norms per Transformer Block
        p.data.normal_(mean=0.0, std=(self.config.initializer_range / math.sqrt(2 * self.config.n_layer)))
```

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报名截止: 2023-09-30 23:30:00

竞赛状态: 进行中

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竞赛状态: 进行中

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作品提交时间 2023.4.25 - 8.30

第一批任务发布 2023.5.12

第二批任务发布 2023.6月中旬

全国赛时间 2023年10月

全国赛颁奖时间 2023年10月

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数据打榜赛题 (每题)

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算法创新赛题 (每题)

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