**Natural Language Processing** 

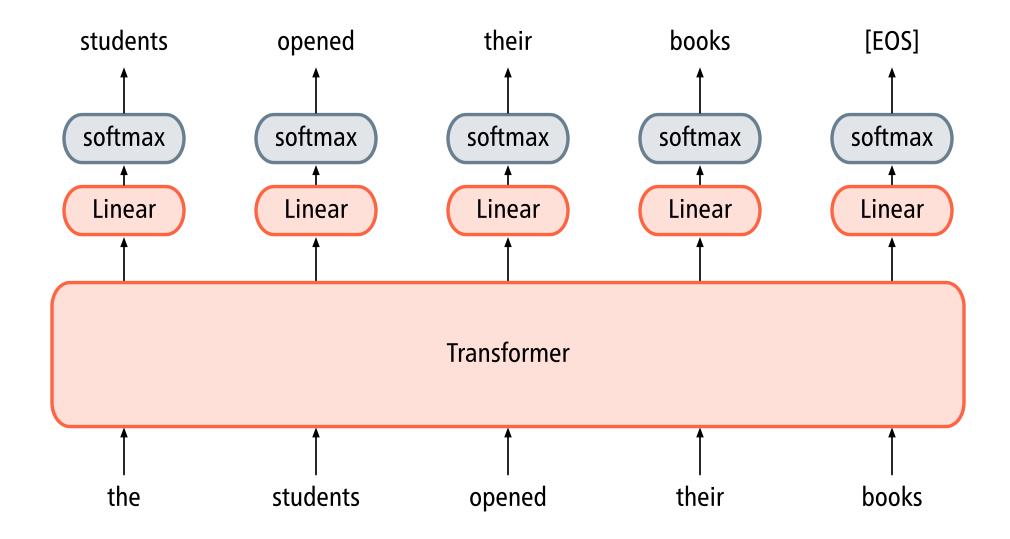
# Emergent abilities of LLMs

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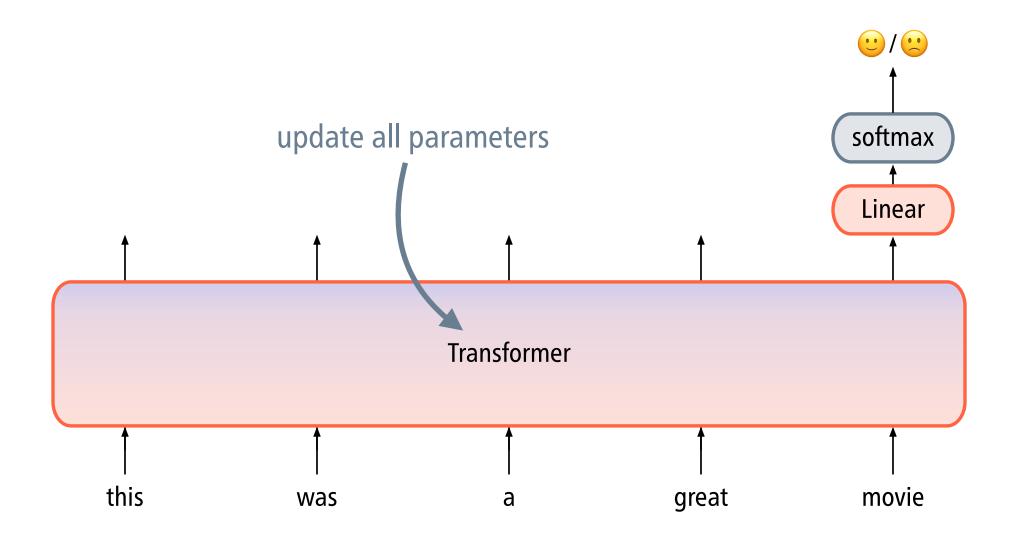
# Pretraining and finetuning



**Step 1: Pretrain on language modelling** 

Large quantities of text, general facts about language

# Pretraining and finetuning



**Step 2: Finetune on specific tasks** 

Small quantities of labelled data, task-specific knowledge

# Model growth

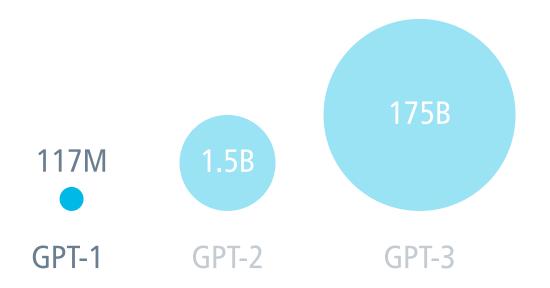
	GPT-1	GPT-2	GPT-3	GPT-4
Number of dimensions	768	1,600	12,288	?
Number of layers	12	48	96	120
Trainable parameters	0.117 B	1.542 B	175 B	1,800 B
Training data size	4 GB	40 GB	570 GB	?

Radford et al. (2018), Radford et al. (2019), Brown et al. (2020)

# **GPT-1: Effective pretraining**

Language modelling is an effective pretraining method for a broad range of tasks in natural language understanding.

Radford et al. (2018)



# Effective pretraining

#### **Natural Language Inference (NLI)**

**Premise:** A man inspects the uniform of a

figure in some East Asian country.

**Hypothesis**: The man is sleeping.

Label: contradiction

#### **Question answering**

**Question:** The first postage stamp was

made ...

Candidate answers: A. in England,

B. in America, C. by Alice, D. in 1910

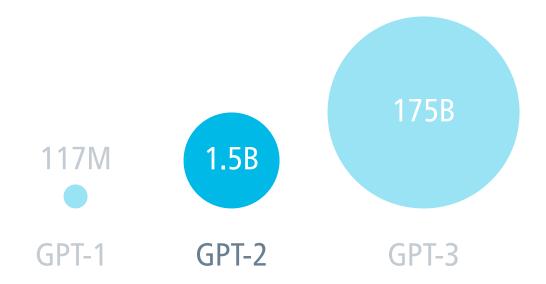
Method	MNLI +	MNLI –	QNLI	RTE	SNLI
Previous state-of-the-art	80.6	80.1	82.3	61.7	89.3
GPT-1 ( <u>Radford et al., 2018</u> )	82.1	81.4	88.1	56.0	89.9

# GPT-2: Emergent zero-shot learning

### zero-shot learning

the ability of a machine learning model to solve tasks out-of-the-box, with no examples and no gradient updates

Radford et al. (2019)



# Zero-shot learning

### **Sequence prediction**

Question: Who took the first steps on

the moon in 1969?

model output

**Answer: Neil Armstrong** 

Question: Which Stanford University

alumna co-founded Coursera?

**Answer: Daphne Koller** 

### **Sequence modelling**

The trophy doesn't fit into the brown suitcase because it is too large.

it = ... p(trophy) > p(suitcase)

The trophy doesn't fit into the brown suitcase because it is too small.

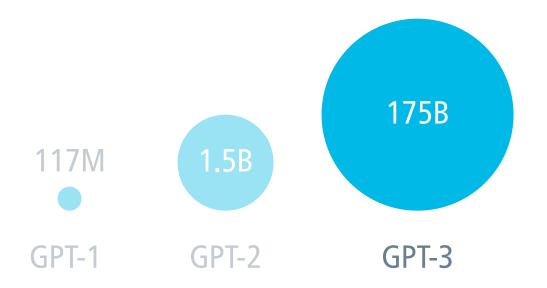
it = ... p(trophy) < p(suitcase)

## GPT-3: Emergent in-context learning

### in-context learning

the ability of a machine learning model to learn tasks from a few examples, with no gradient updates

Brown et al. (2020)



### **Word unscrambling**

gaot => goat

sakne => snake

brid => bird

fsih => fish

dcuk => duck

cmihp => chimp

in-context learning

#### **Machine translation**

thanks => merci

hello => bonjour

mint => menthe

wall => mur

otter => loutre

bread => pain

in-context learning

# Chain-of-thought prompting

#### **Standard prompting**

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 balls. How many tennis balls does he have now?

A: The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

A: The answer is 11.

#### **Chain-of-thought prompting**

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 balls each is 6 balls. 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they had 3 + 6 = 9. The answer is 9.

# Zero-shot chain-of-thought prompting

#### **Chain-of-thought prompting**

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 balls. How many tennis balls does he have now?

A: Roger started with 5 balls. 2 cans of 3 balls each is 6 balls. 5 + 6 = 11. The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

A: The cafeteria had 23 apples originally. They used 20 to make lunch. So they had 23 - 20 = 3. They bought 6 more apples, so they had 3 + 6 = 9. The answer is 9.

#### **Zero-shot chain-of-thought prompting**

Q: Roger has 5 tennis balls. He buys 2 more cans of tennis balls. Each can has 3 balls. How many tennis balls does he have now?

A: The answer is 11.

Q: The cafeteria had 23 apples. If they used 20 to make lunch and bought 6 more, how many apples do they have?

A: Let's think step by step. The cafeteria had 23 apples originally. They used 20 to make lunch, so they had 23 - 20 = 3. They bought 6 more apples, so they had 3 + 6 = 9. The answer is 9.

# Prompt engineering

### designed by a LM

Prompt	Accuracy
Let's work this out in a step by step way to be sure we have the right answer.	82.0
Let's think step by step.	78.7
First,	77.3
Let's think about this logically.	74.5
Let's solve this problem by splitting it into steps	72.2
Let's be realistic and think step by step.	70.8
Let's think like a detective step by step.	70.3
(Zero-shot)	17.7