Al and Large Language Models

October 2024

Accelerate Programme for Scientific Discovery







Welcome!

About the course

- Introduction to LLMs and how they work
- What is available?
- How to use them

Code and slides available on:

Accelerate Science GitHub repo





Today's Schedule

- Introduction to the Accelerate Science Programme
- Introduction to LLMs
- BREAK
- Augmenting LLMs
 - Finetuning
 - Prompting
- LUNCH
- Small language models
- Quantization
- Data Ethics
- BREAK
- What's out there?
 - Popular models
 - APIs
 - No Code





Who are we?

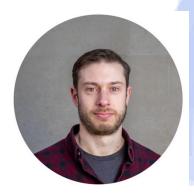




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Ryan Daniels





Accelerate Science Programme

"Accelerate Science pursues research at the interface of AI and the sciences, generating new scientific insights and developing AI methods that can be deployed to advance scientific knowledge."





Accelerate Science Programme

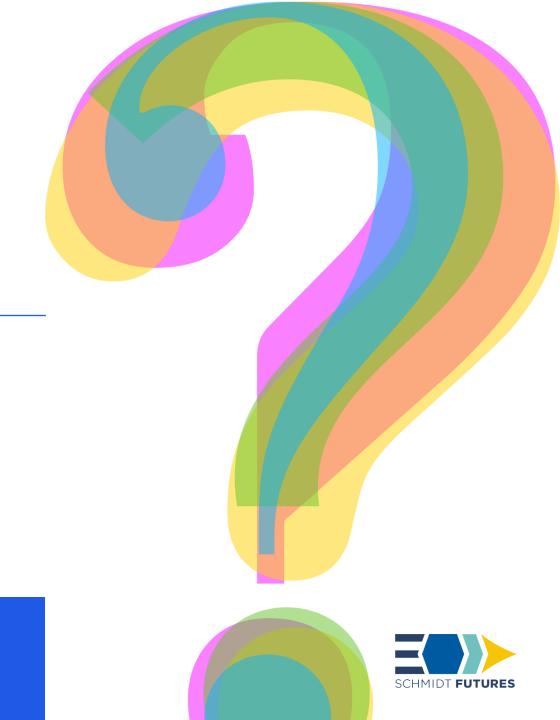
LLM Workshop Goals

We want to:

- Support researchers across the university to use AI that's LLMs in this group.
- Better understand the challenges that researchers face.
- Identify what training courses or software resources we might want the Accelerate Science Programme to create.
 - Including shared code
- Start to build a community of like-minded researchers across the university.









Overview

- What is a large language model?
- How do they work?
- o How are they trained?

But first...

What do you know about LLMs...?





What do you know about LLMs...?

- o How do they work?
- Can you name any?
- What can you use them for?
- What are potential problems?
- What Can YOU potentially use them for?







First, some definitions...

- Model The thing you train and use to generate text.
- Architecture The high-level structure of the model.
- Parameters (or weights) The internal knobs and dials that get altered during the course of training.
- Context The input to the LLM.
- Embedding A high-dimensional representation of text.
- Token The smallest unit of text fed into a model.
- Pretraining The first stage of training the LLM, and usually the most expensive.
- Finetuning Training the LLM to do a particular task.
- Inference The act of calling the model for text generation.
- Quantization Reducing the size of a model.
- Source The availability of the model (can be open or closed).
- Modality Defined by the type of data the model is trained on (e.g. text or images).
- API (Application Programming Interface) The means by which you can interact with a model.





What are LLMs?

- LLMs are conditioned on a massive amount of text.
- Given some input tokens, and given what we know from the training data, what is the most likely next token?
- An analogy...







What is a large language model?

A language model is a high-parameter model that is trained on massive amounts of text with the end goal of some kind of text prediction or generation.

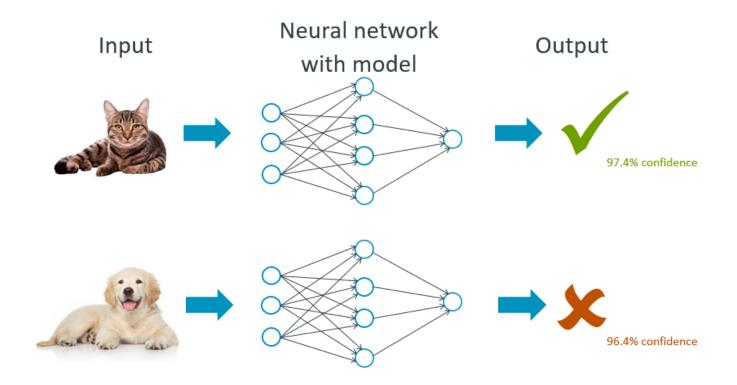
- Not all language models have the same architecture
- Not all language models are built for next-token prediction
 - GPT is a next token predictor
 - BERT is a masked-language model
- Not all language models are chatbots.





What is a large language model?

A normal neural network



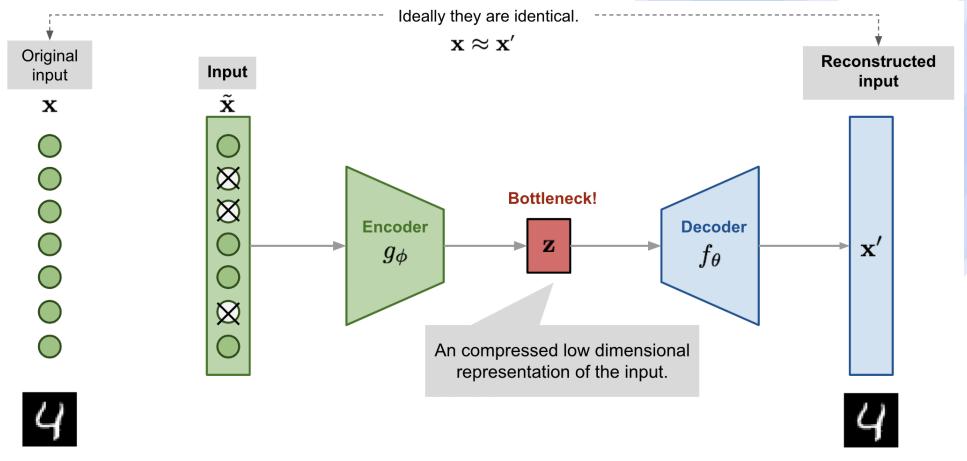
- 1. Picture goes in
- 2. Pixel values propagate through the weights of the network
- Network outputs either "cat" or "dog"
- 4. Update the weights of the network based on how wrong the prediction is
- 5. Do many times





What is a large language model?

Encoder vs Decoder









Tokenization

The machine starts from scratch

```
text = "The cat sat on the mat."
tokens = text.split()
print(tokens)
# Output: ['The', 'cat', 'sat', 'on', 'the', 'mat.']
```





Tokenization

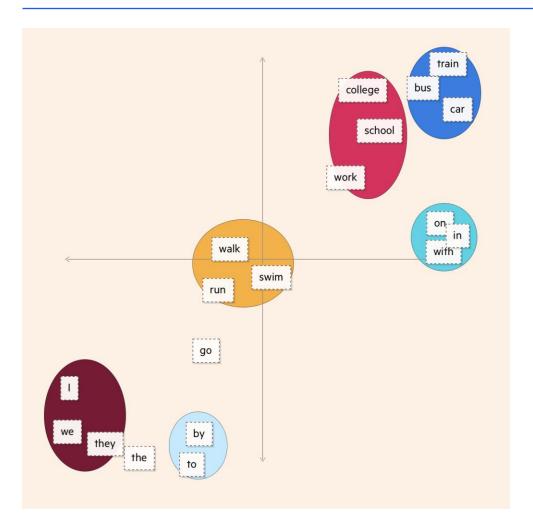
The machine starts from scratch

```
# Build vocabulary
vocab = sorted(set(tokens))
# Create token to index mapping
token to index = {token: index for index, token in enumerate(vocab)}
# Create index to token mapping
index_to_token = {index: token for token, index in token_to_index.items()}
print(token_to_index)
# Output: {'cat': 0, 'mat': 1, 'on': 2, 'sat': 3, 'the': 4}
# Convert tokens to indices
indices = [token_to_index[token] for token in tokens]
print(indices)
# Output: [4, 0, 3, 2, 4, 1]
```





Embeddings



- Embeddings are vectors (lists of numbers)
- https://ig.ft.com/generative-ai/
- There are a number of different types of embeddings:
 - Learned
 - Positional
 - Combinations





Embeddings

Four score and seven years ago our

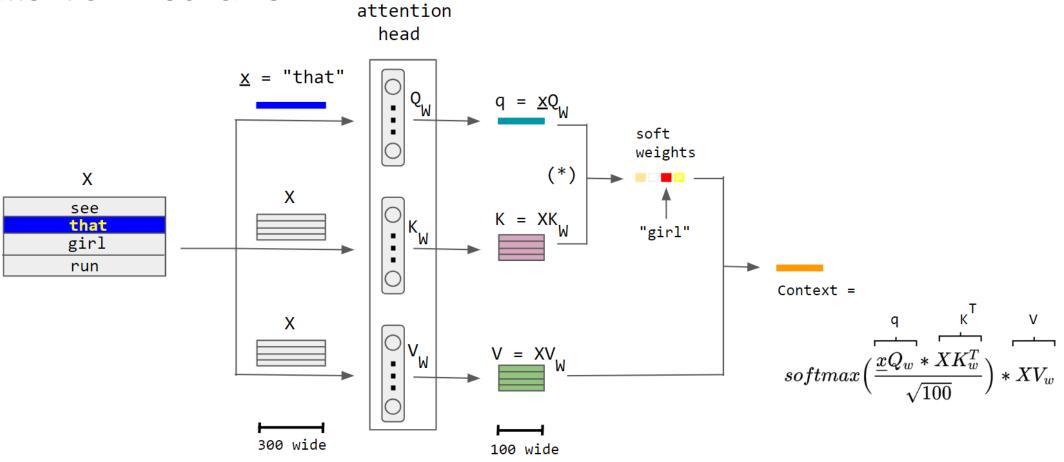
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How do they work?

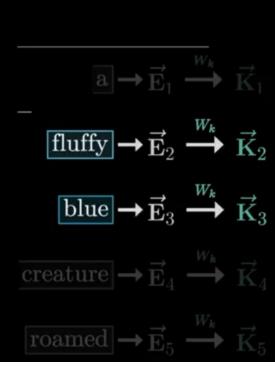
The attention mechanism

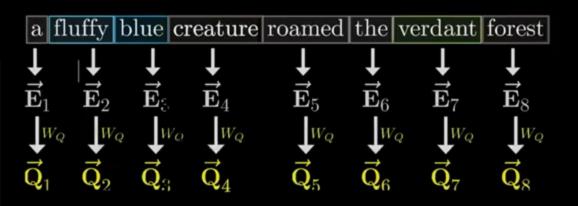






How do they work?



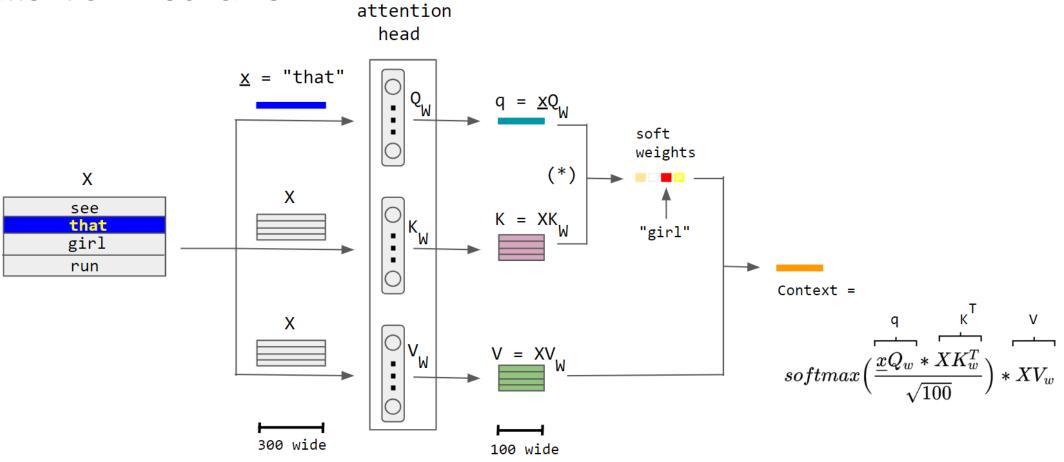






How do they work?

The attention mechanism



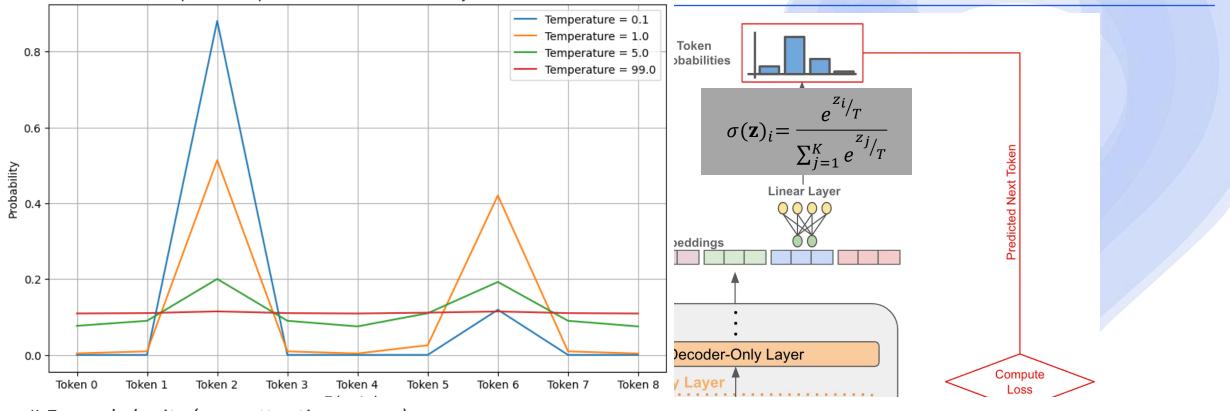






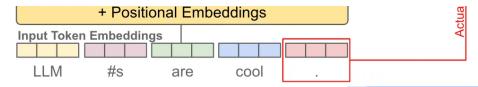
How are they trained?

Impact of Temperature on Softmax Probability Distribution



Example logits (e.g., attention scores)
logits = np.array([0.2, 1.0, 5.0, 1.0, 0.1, 2.0, 4.8, 1.0, 0.1])





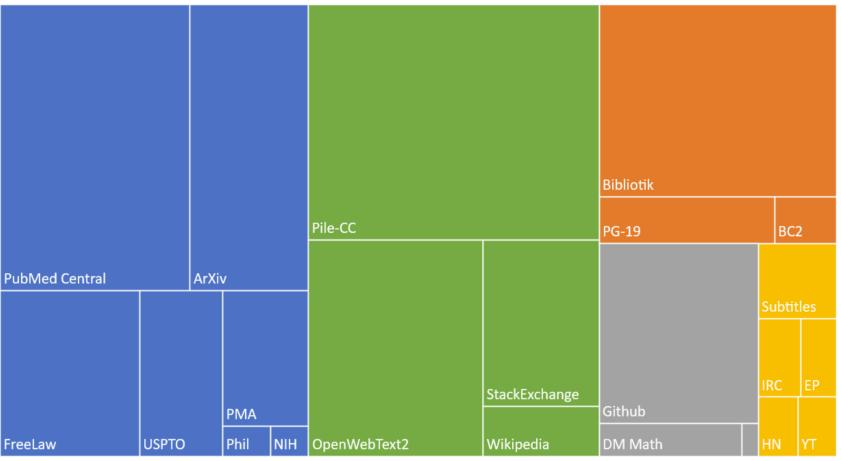




How are they trained?

Composition of the Pile by Category

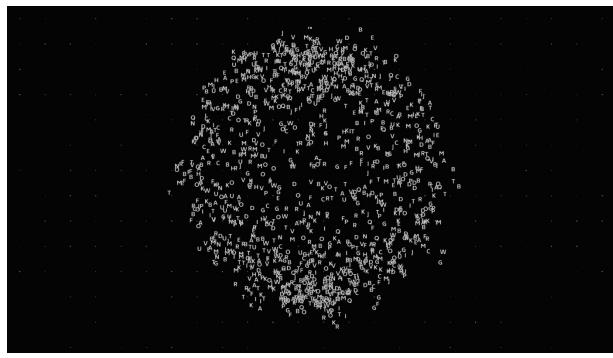
Academic Internet Prose Dialogue Misc







LLMs as a compressed version of the internet



https://www.newvorker.com/tech/annals-of-technology/chatgpt-is-a-blurry-ipeg-of-the-web

Some like to think of an LLM as a compressed version of the internet

One way of compressing images is to take every other pixel

When you decompress, you interpolate between the pixels

The inclination is to think that some kind of interpolation is happening in LLMs...

...but it's more complicated than this.





Scaling Laws

Chinchilla scaling laws



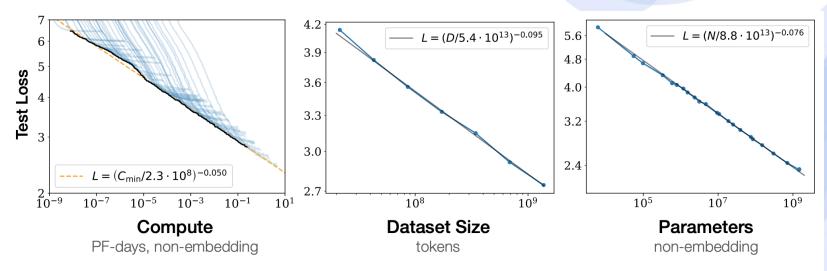


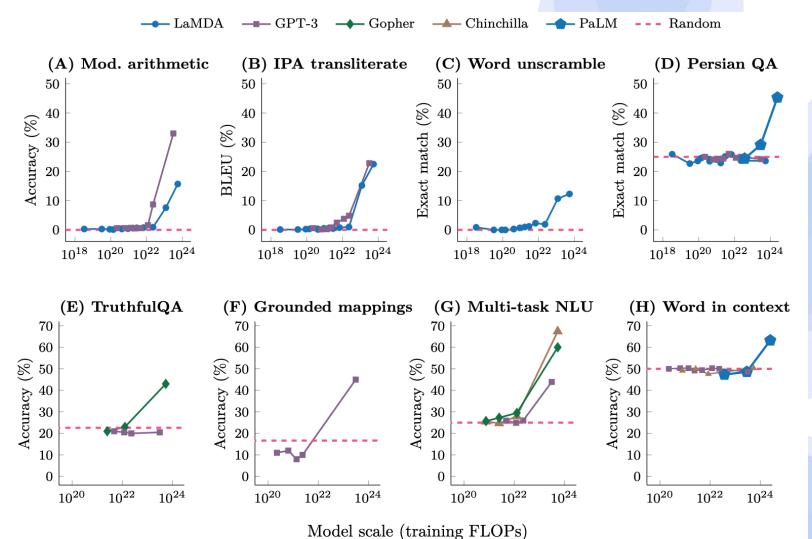
Figure 1 Language modeling performance improves smoothly as we increase the model size, datasetset size, and amount of compute² used for training. For optimal performance all three factors must be scaled up in tandem. Empirical performance has a power-law relationship with each individual factor when not bottlenecked by the other two.





Emergence

Maybe...







Resources

Jargon-free explanation of LLMs

Generative AI exists because of the Transformer

The dangers of Stochastic Parrots

The Pile

Survey of LLMs

Attention is all you need

Intro to Large Language Models





10 min BREAK

