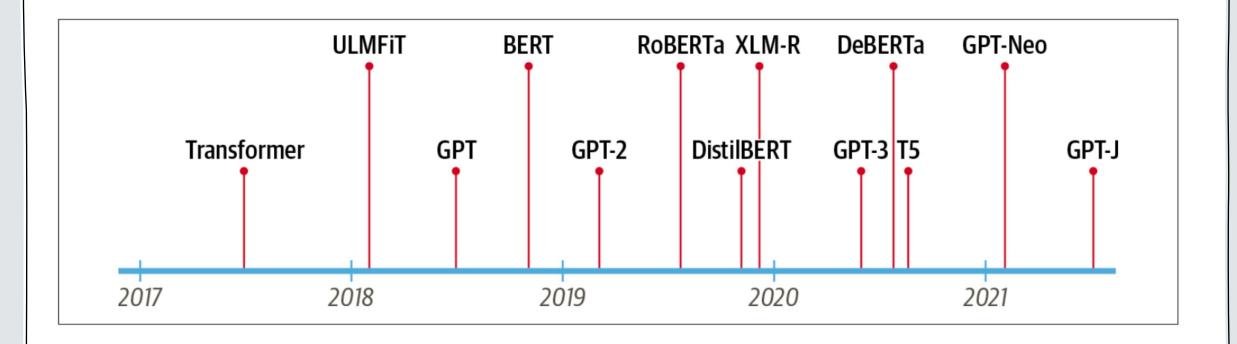
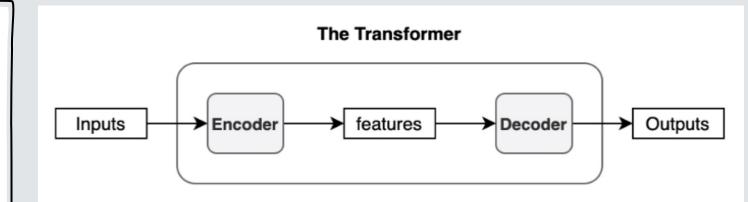
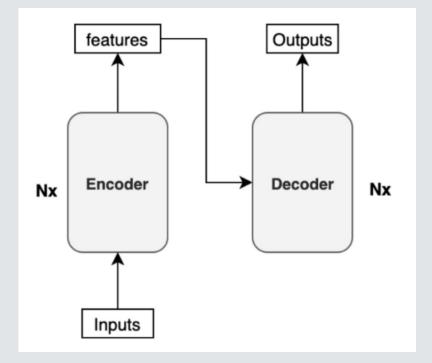


THE TRANSFORMER TIMELINE

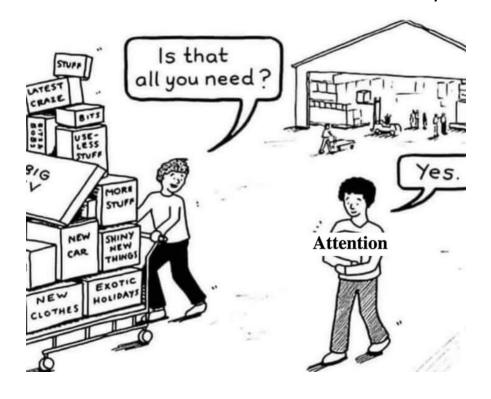


TRANSFORMER ARCHITECTURE



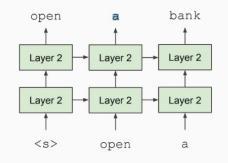


TRANSFORMER >> RNN, LSTM FOR NLP

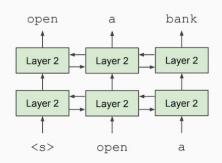


Attention is all you need. Vaswani et al., NeurIPS 2017

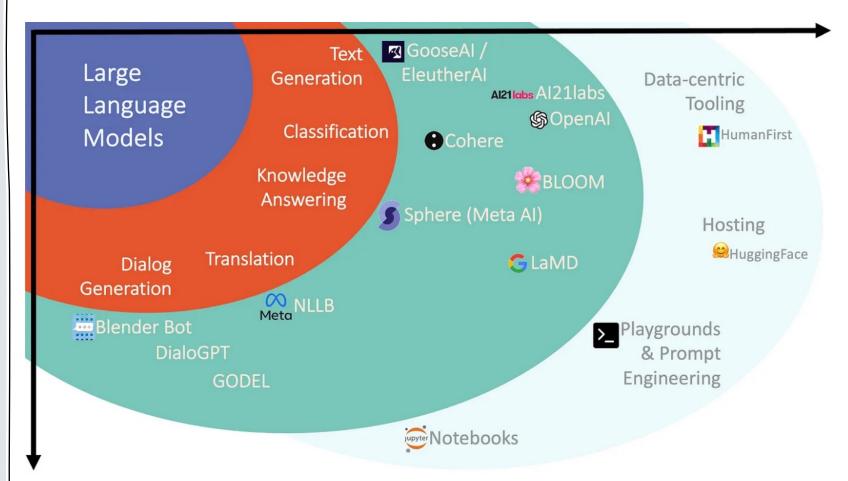
Unidirectional contextBuild representation incrementally

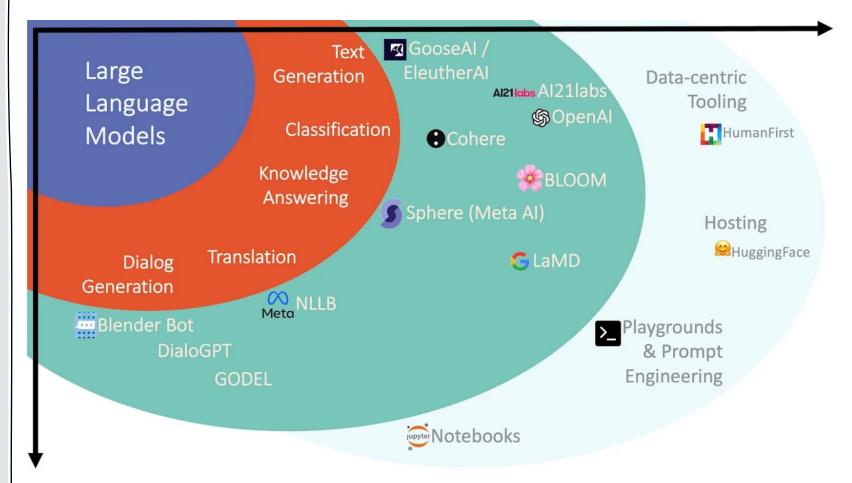


Bidirectional context Words can "see themselves"

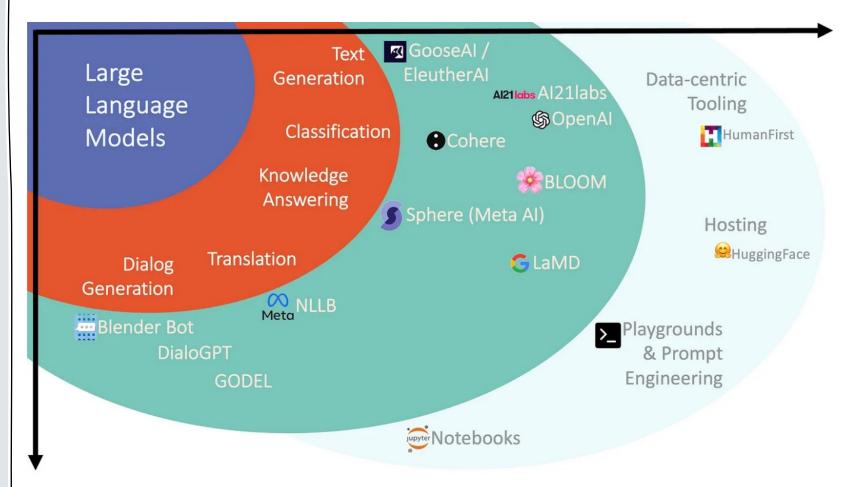


- ✓ Parallel Processing
- ✓ Bidirectionality
- ✓ Less labelled data required



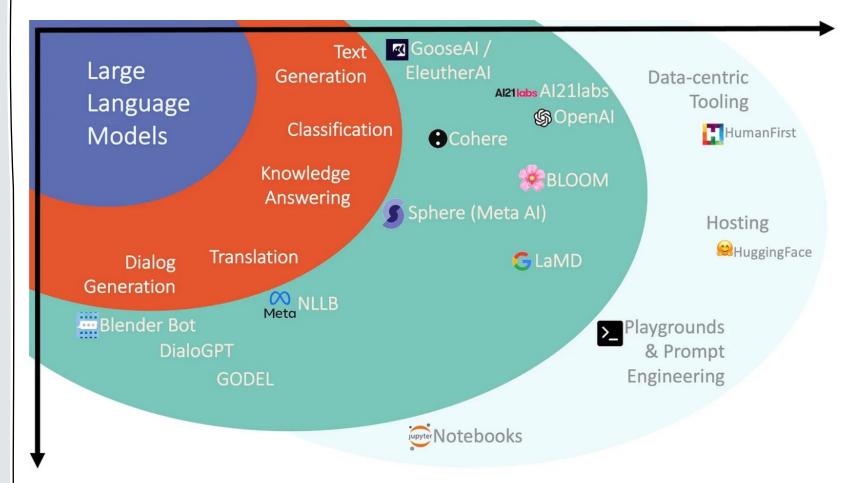








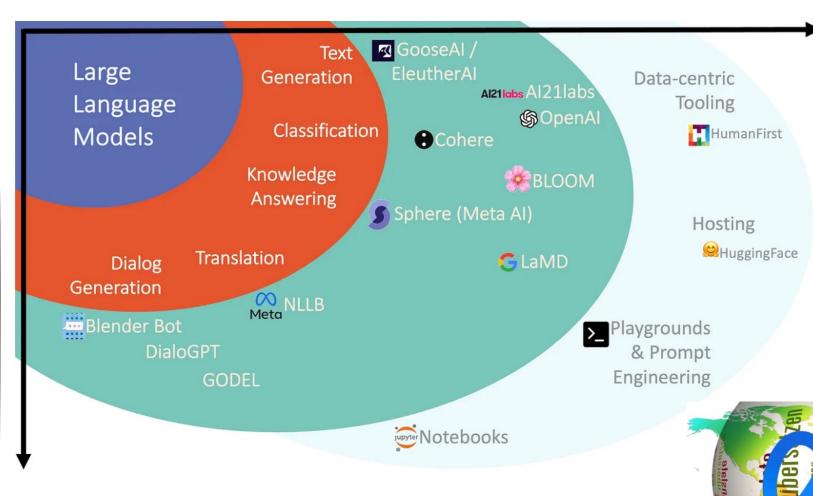










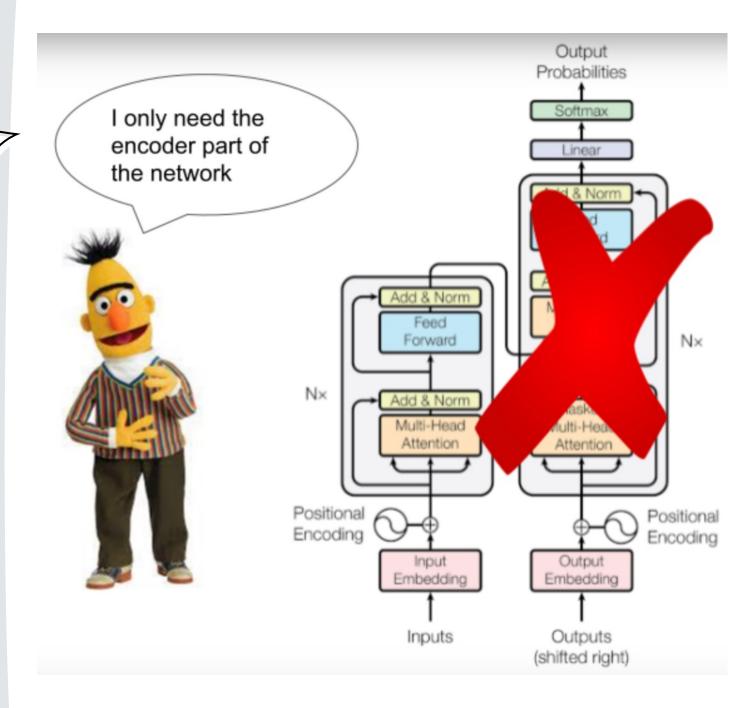




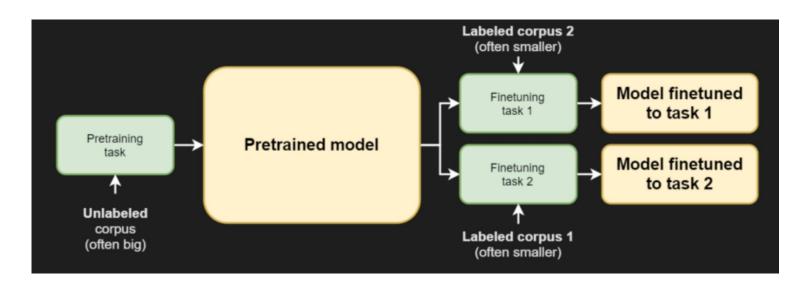




BIDIRECTIONAL
ENCODER
REPRESENTATION
for
TRANSFORMER



Bidirectional Encoder Representation for Transformer



BERT = fine-tuning & transfer learning i.e. pre-train a model on the large unlabelled corpus <u>and</u> finetune to a specific language task.

BERT pre-training has two objectives:

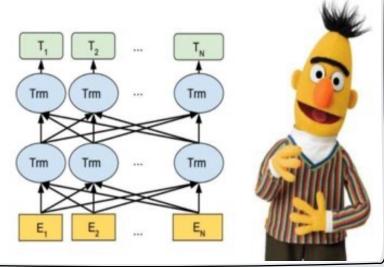
1) Predict masked tokens in texts (Masked Language Modelling)

LET'S PRETEND WE'RE BERT...

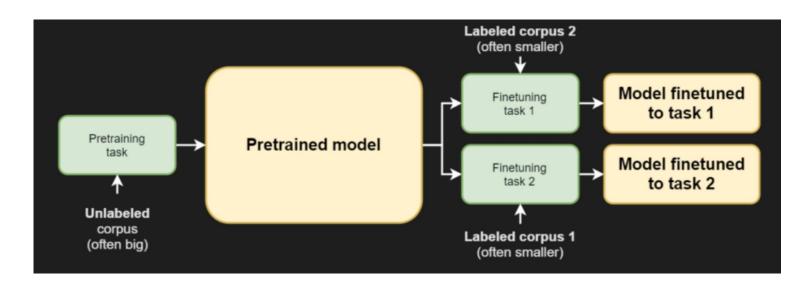
...and play a fill-in-the-blank game:

"Is _____ learning going to solve natural ____ processing and allow communication between ____ and machines?"

→ Which words do you think go in the blanks?



Bidirectional
Encoder
Representation
for
Transformer



BERT pre-training has two objectives:

- 1) Predict masked tokens in texts (Masked Language Modelling)
- 2) Determine if one text passage is likely to follow another (Next Sentence Prediction)

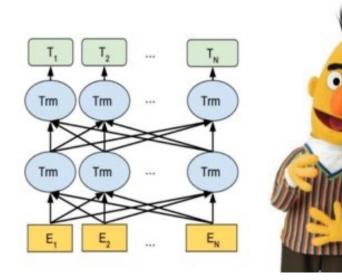
LET'S PRETEND WE'RE BERT...

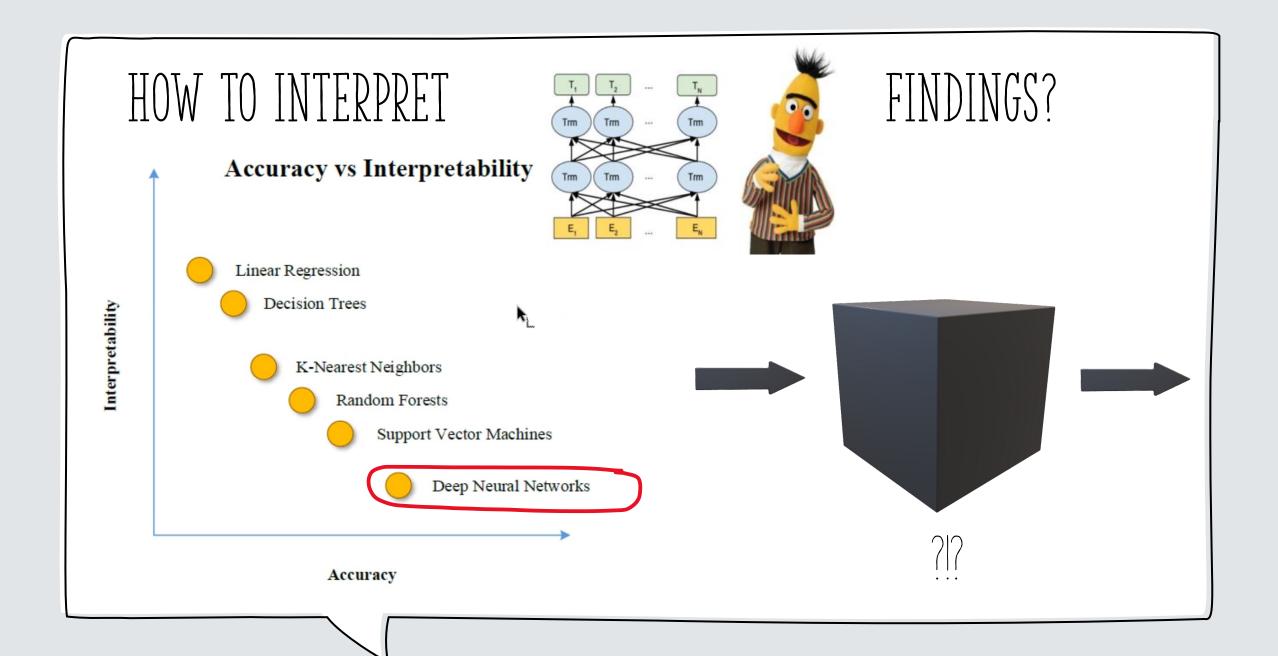
...and check whether a pair of sentences are absolute nonsense or not.

Sentence 1: "When I was younger, I dreamt of flying to Jupyter."

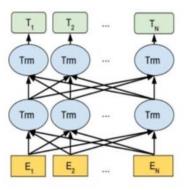
Sentence 2: "but Peking ducks taste better."

Is <u>Sentence 2</u> related to <u>Sentence 1</u>?





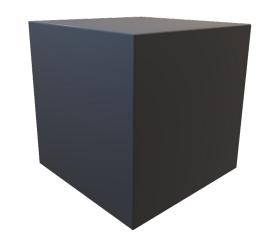
HOW TO INTERPRET

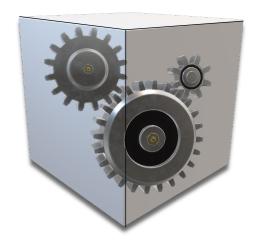




FINDINGS?

XAI





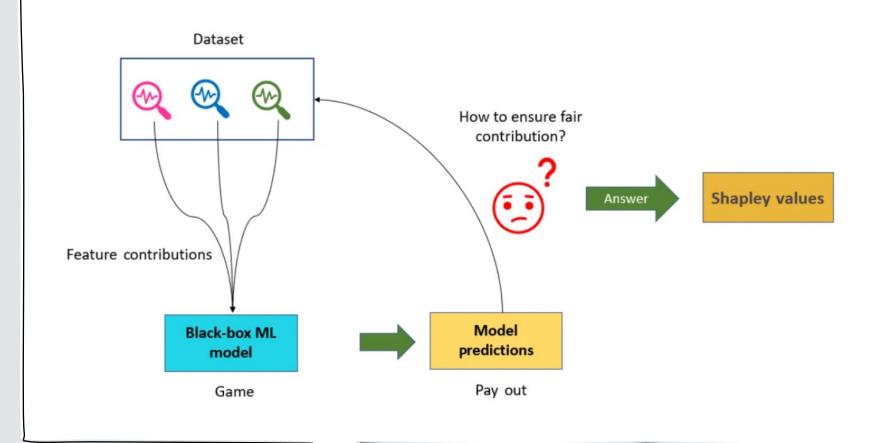


Which tokens in the input are important?

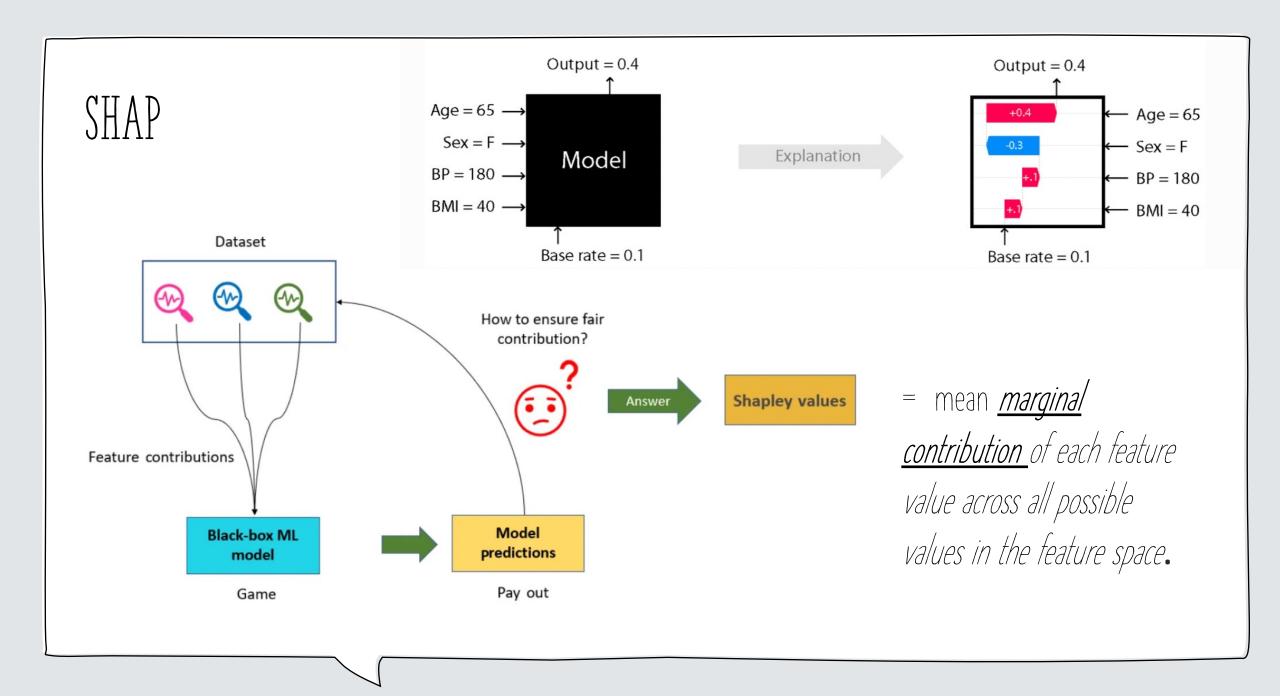


Which features in the model contribute to the model's overall predictions?

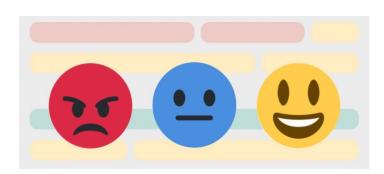
SHAP = SHapley Additive exPlanations

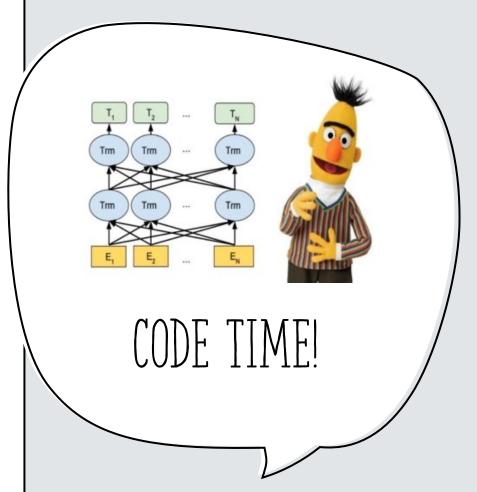


= mean <u>marginal</u>
<u>contribution</u> of each feature
value across all possible
values in the feature space.









REFERENCES, FURTHER READINGS & TUTORIALS

Vaswani et al. 2017 NeurlPS Attention is all you need.

Devlin et al. 2018 arxiv BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding.

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