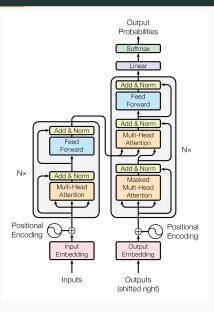
# Syntax and grounding in adjective learning General supplementary material

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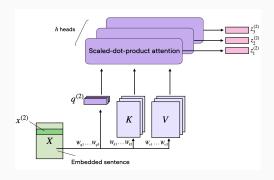
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Queen Mary University of London

#### **Full Transformer architecture**



### Multi-head self attention

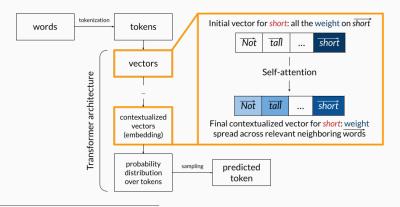


# **Model parameters**

Model	Structure	Parameters	Layers	Hidden dimension
GPT2-Large	Decoder	774M	36	1280
BERT-Large	Encoder	340M	24	1024
RoBERTa-Large	Encoder	355M	24	1024
XLNet-Large	Decoder	340M	24	1024
Mistral 7B v0.1	Decoder	7B	32	4096

# Self-attention creates context-sensitive word representations

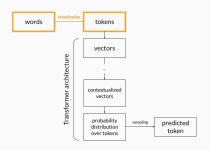
 Self-attention is at the core of Transformers, and should allow them to grasp the contextualized meaning of antonymic adjectives, and the functional behavior of negation.<sup>1</sup>



<sup>&</sup>lt;sup>1</sup>Though see Peng et al. (2024).

## Tokenization as a proxy for morphological decomposition

- Tokenization maps sentences into tokens, which represent words or pieces of words.
- Tokens are determined based on character coocurrences.
- Tokens may therefore reflect morphology,<sup>2</sup> and provide LLMs with useful distributional cues to derive ITA contrasts.



<sup>&</sup>lt;sup>2</sup>Nair and Resnik, 2023