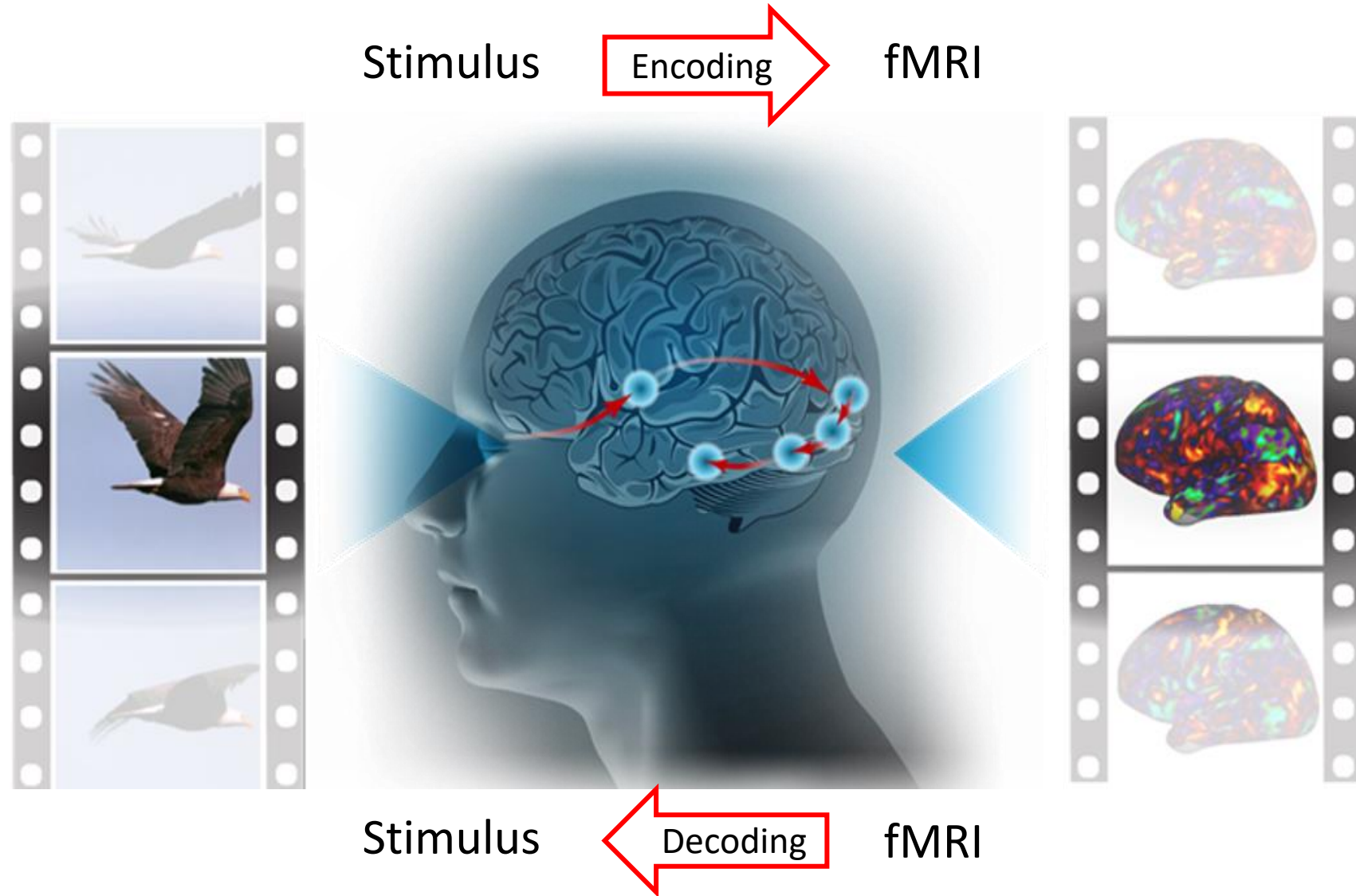


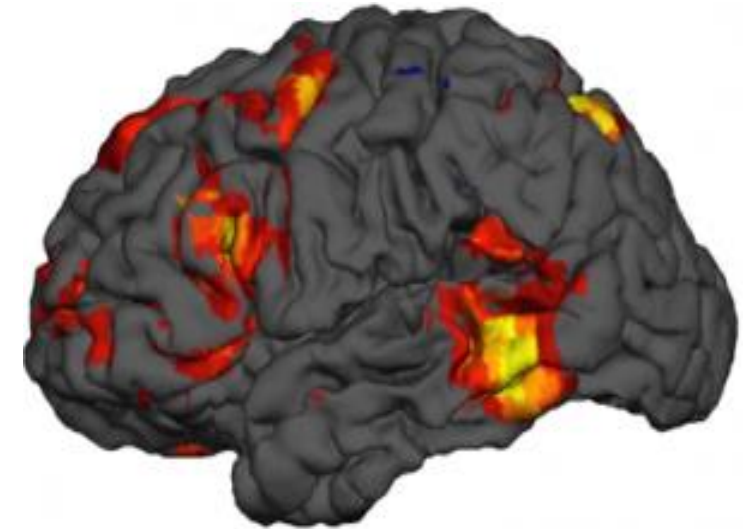
Predicting Brain Activity while Listening to Stories

Subba reddy Oota, Frederic Alexandre, Xavier Hinaut

Brain Encoding vs Decoding

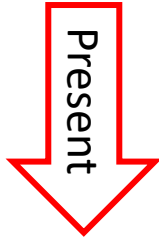


What is fMRI?

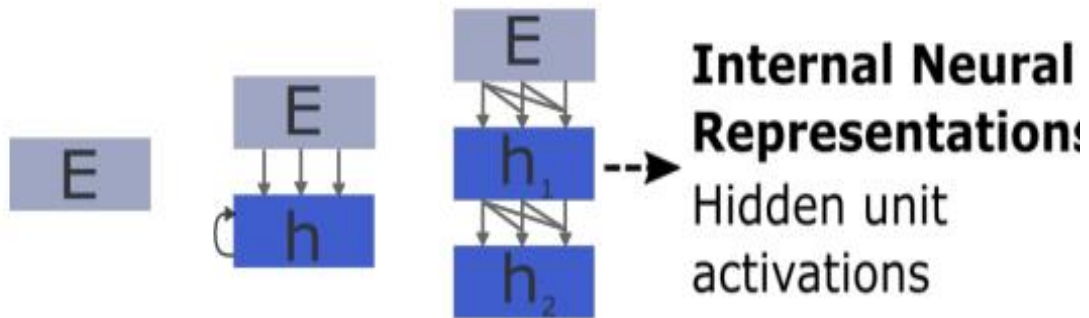


Brain Encoding

Stimulus

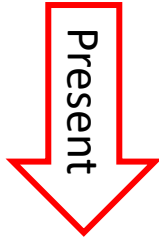


Models

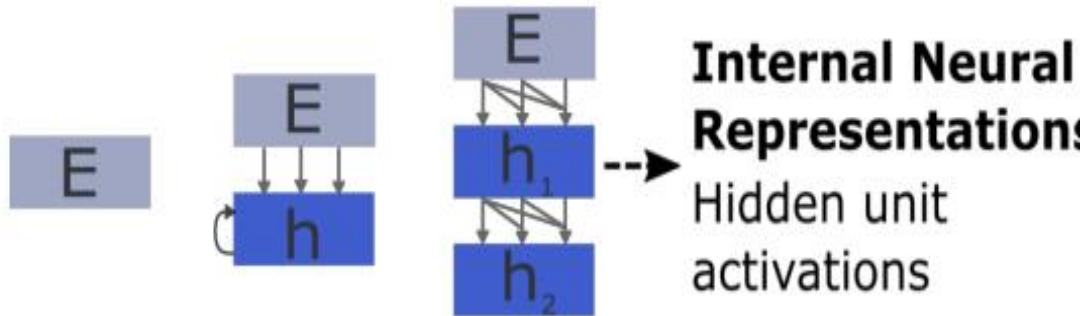


Brain Encoding

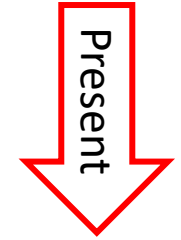
Stimulus



Models

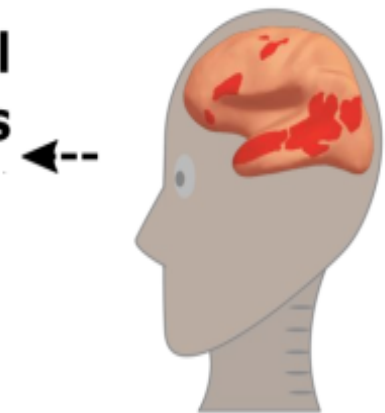


Stimulus

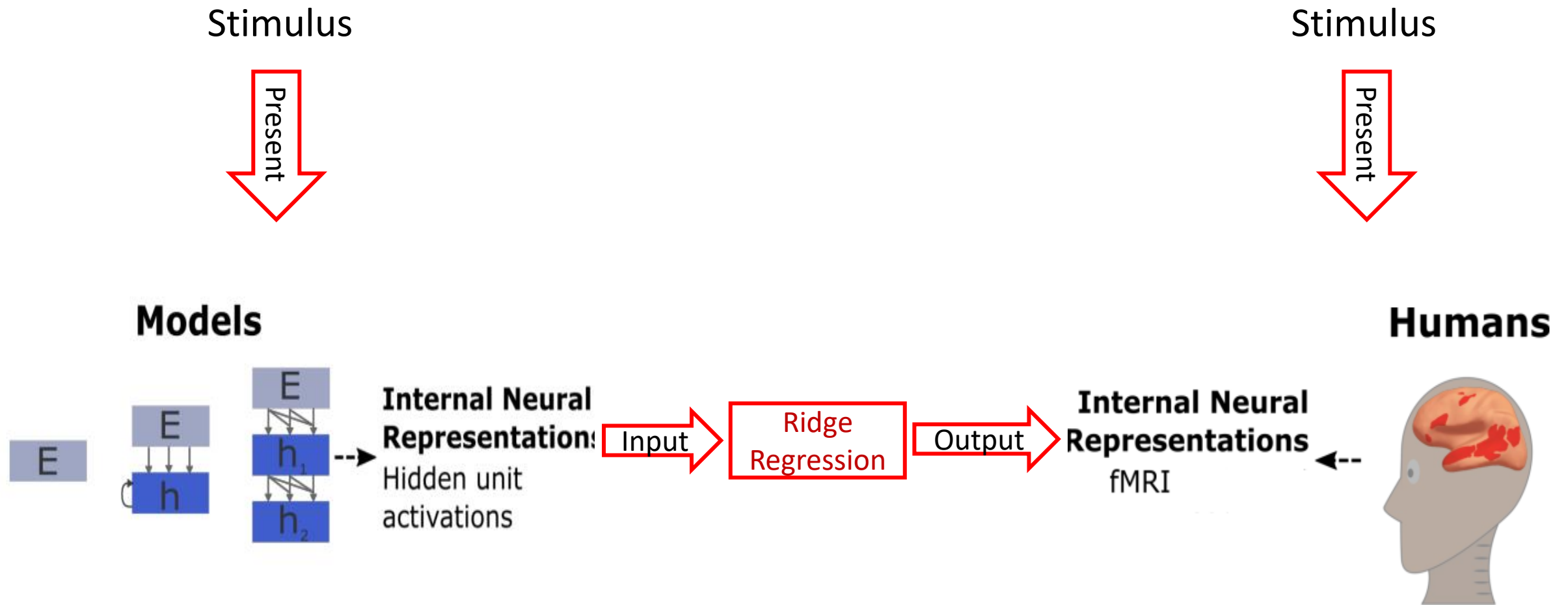


Humans

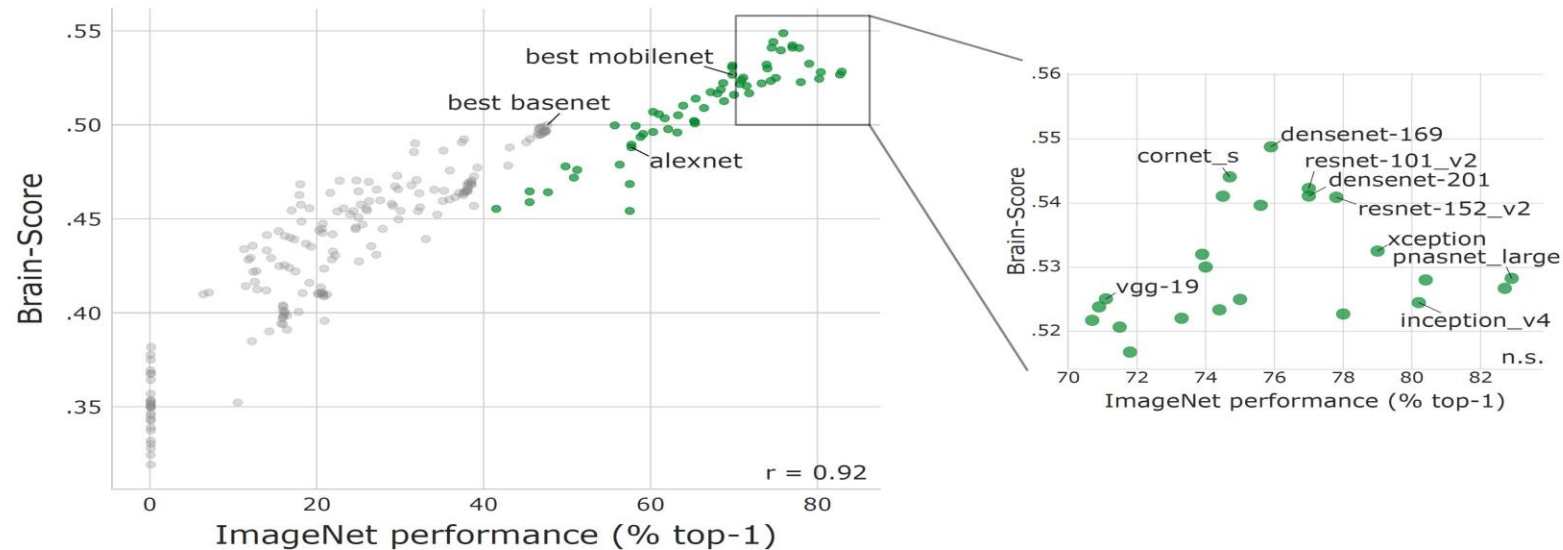
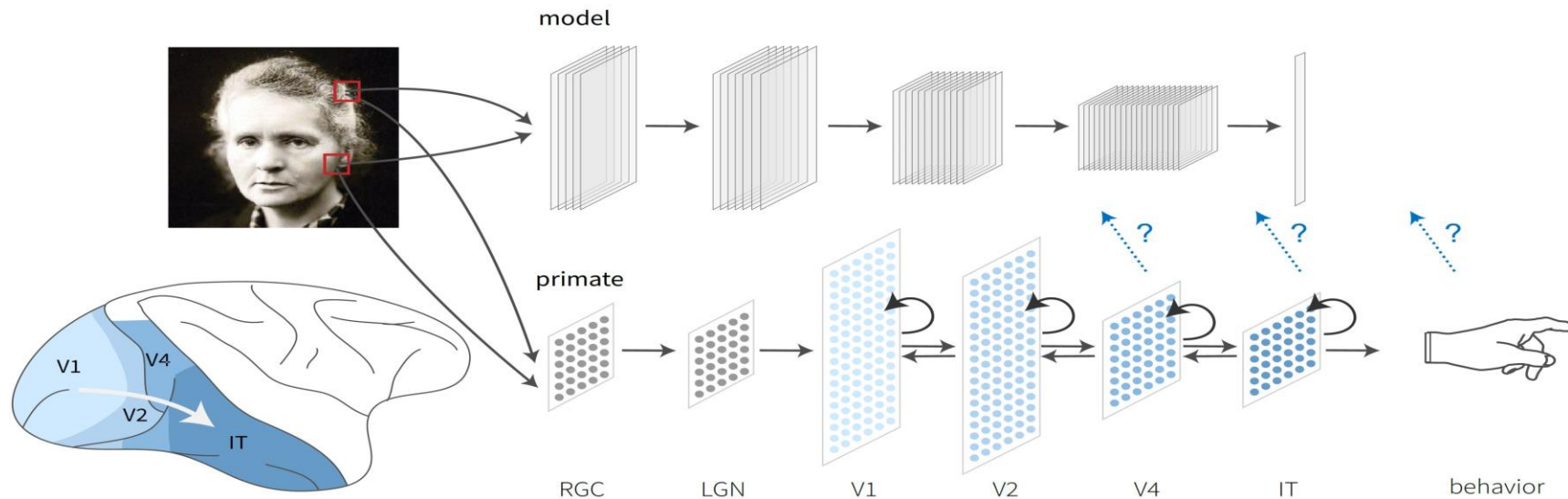
**Internal Neural
Representations**
fMRI



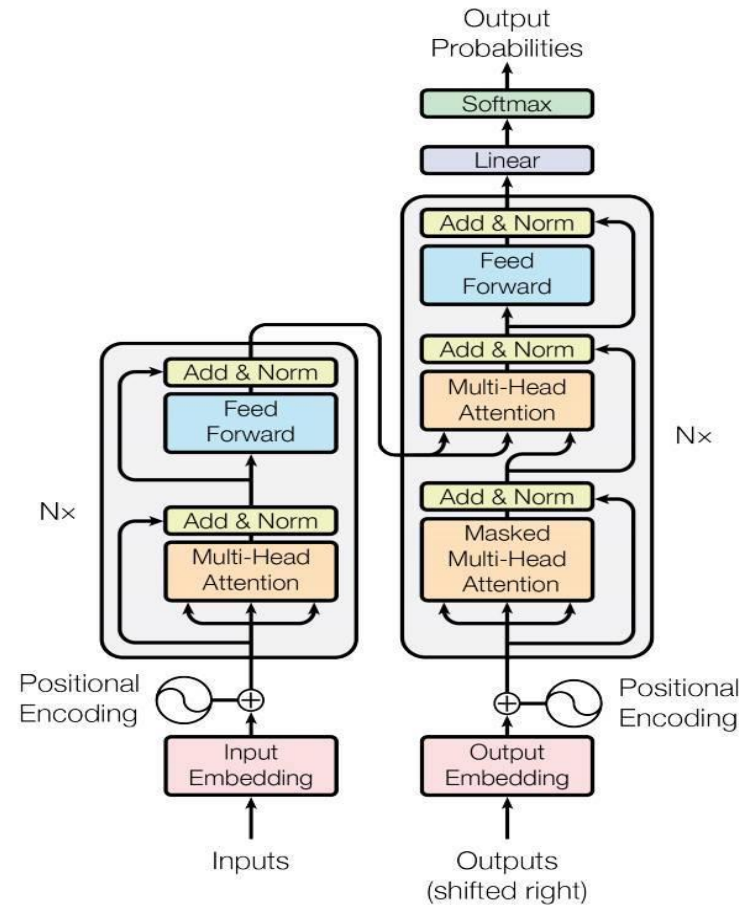
Brain Encoding



Vision models accurately predict human vision recordings

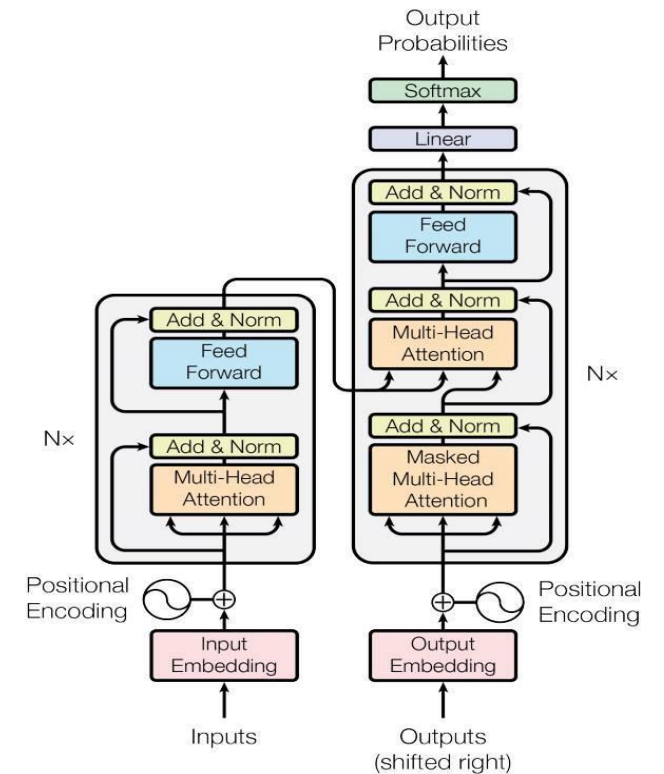
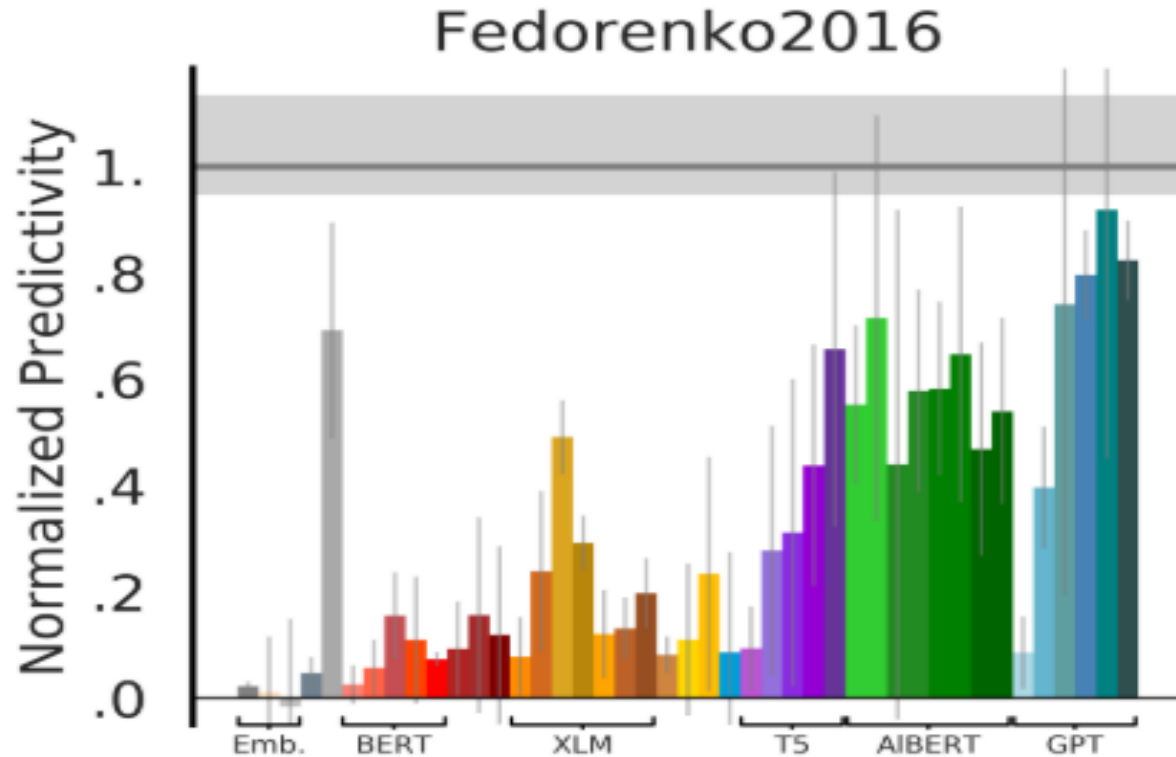
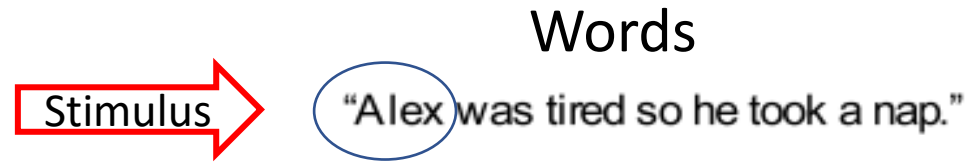


Most popular language models are Transformers



Transformer language models
(BERT, XML, GPT,...)

Language models based on words



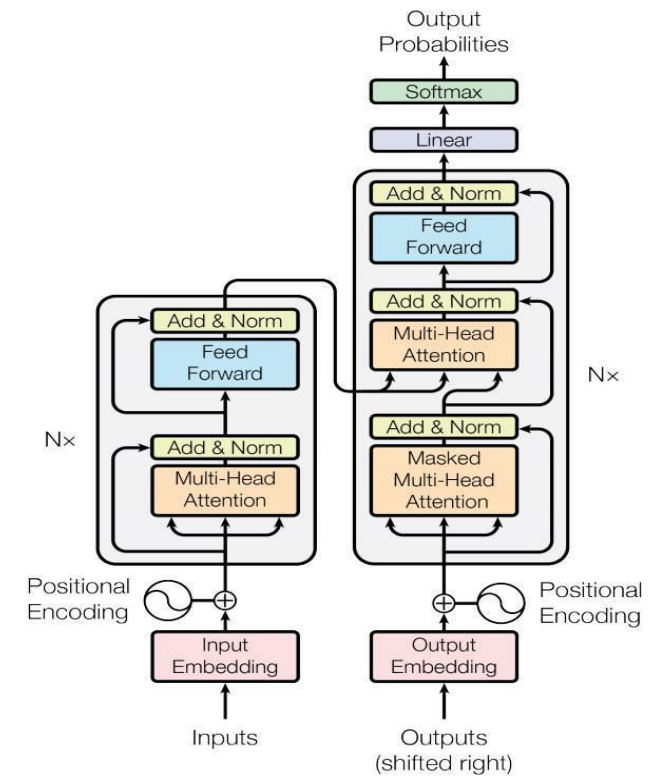
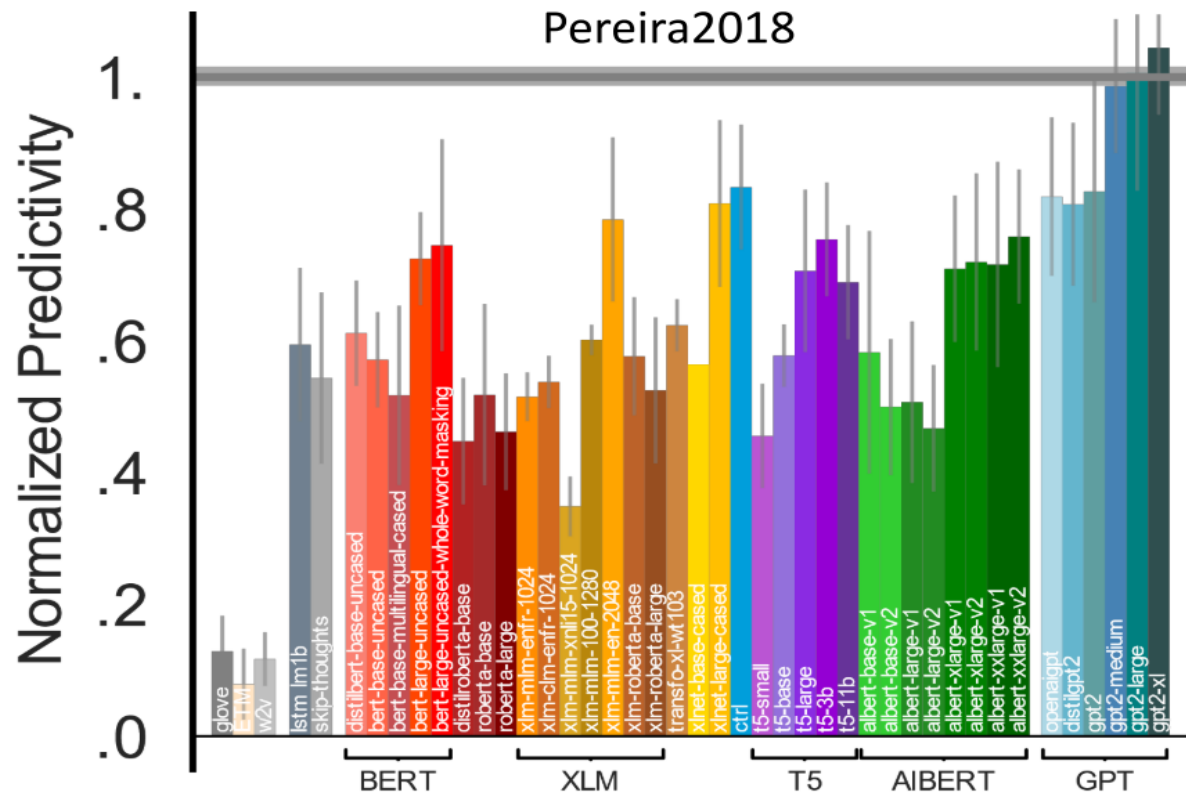
Transformer language models
(BERT, XML, GPT,...)

Language models based on sentences

Sentences

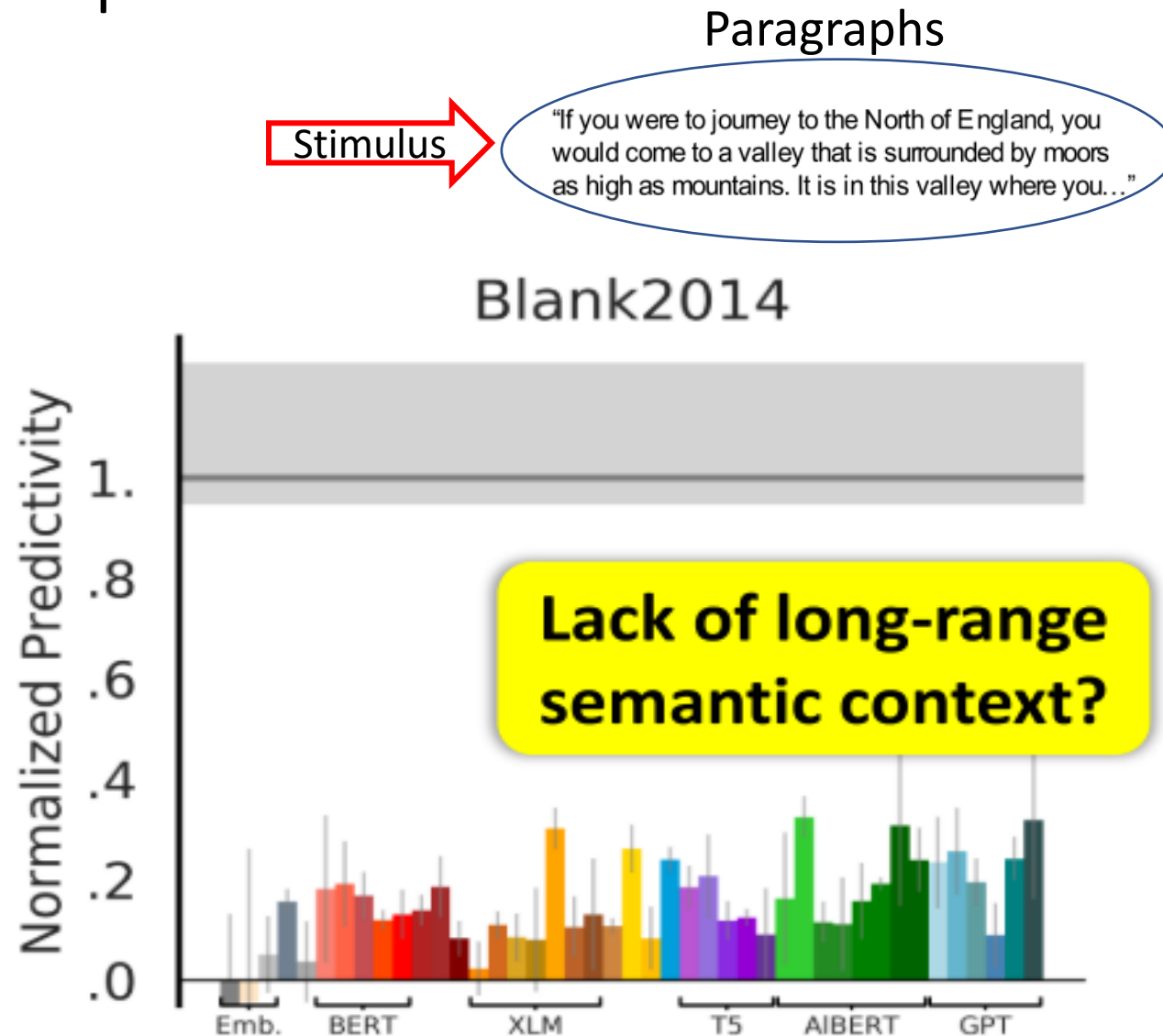
Stimulus

"Beekeeping encourages the conservation of local habitats. It is in every beekeeper's interest..."



Transformer language models
(BERT, XLM, GPT,...)

Can current language models deal with long-term dependencies?



- Transformer language models are unable to handle the long-term dependencies
 - **sequence** length is fixed to 512 words
- LSTMs still lacks investigation of the long-term memory cognitive plausibility and its link to fMRI data

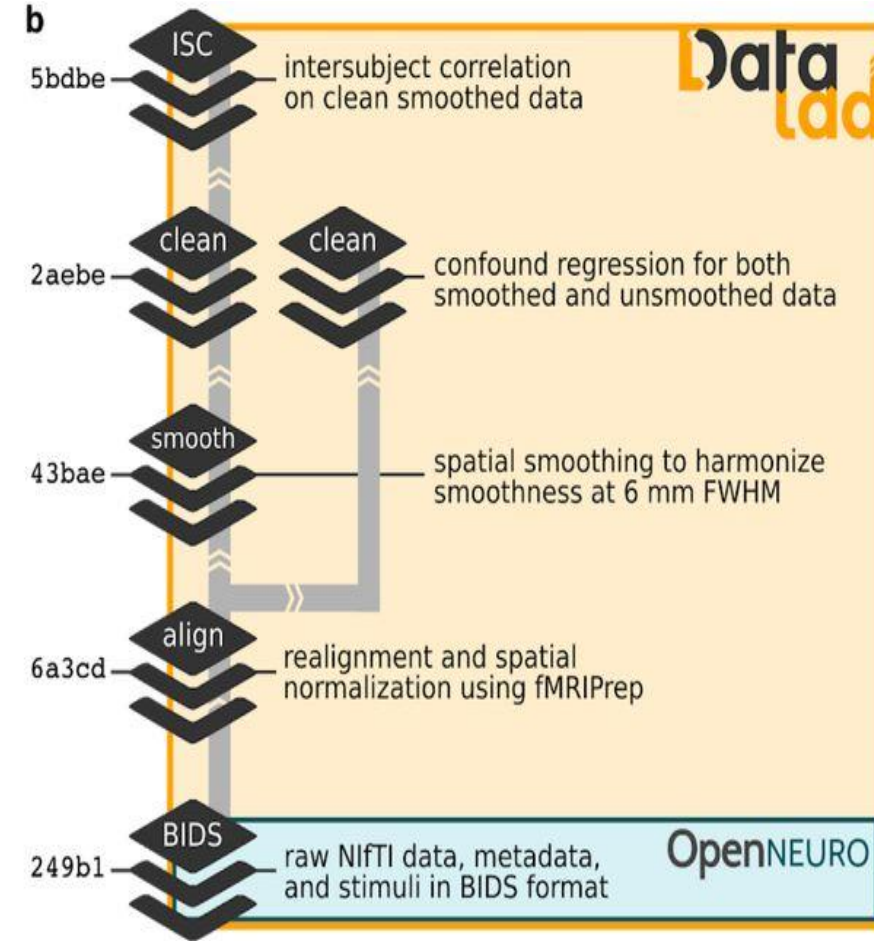
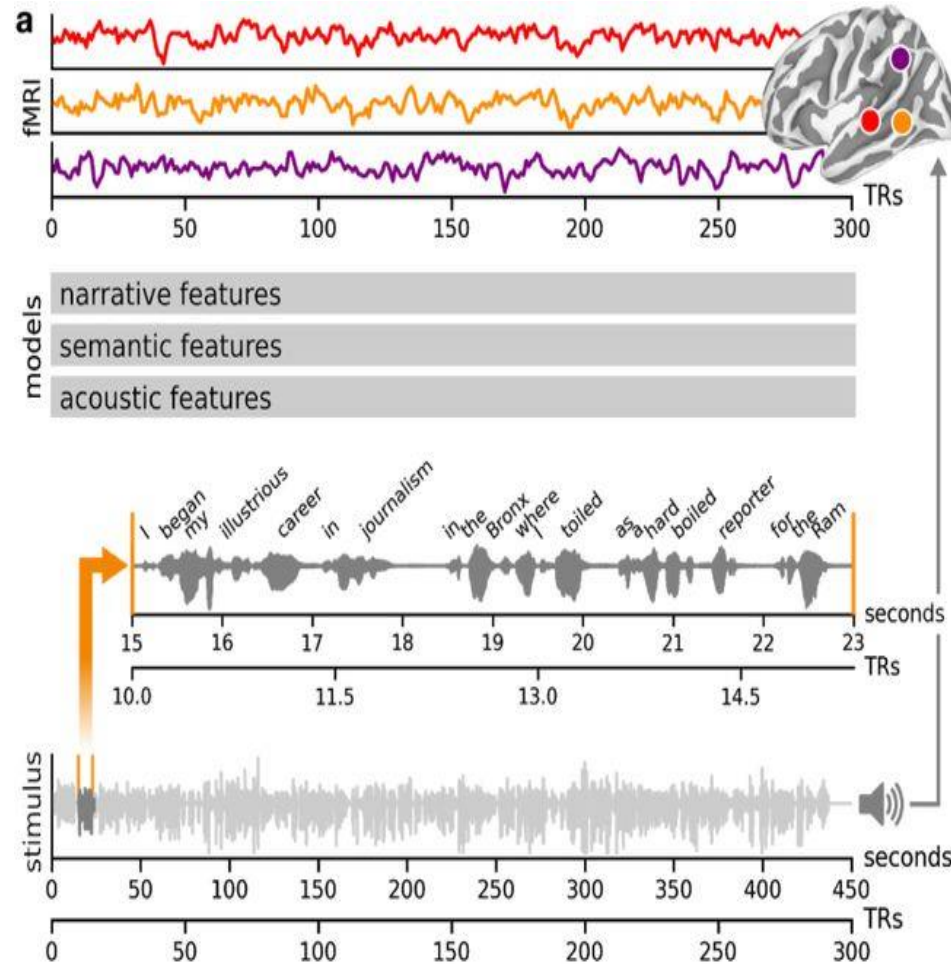
What kind of language models can represent long-term dependencies?

could they also predict higher cognition while subjects are engaged in longer stories ?

The data target: human brain recordings

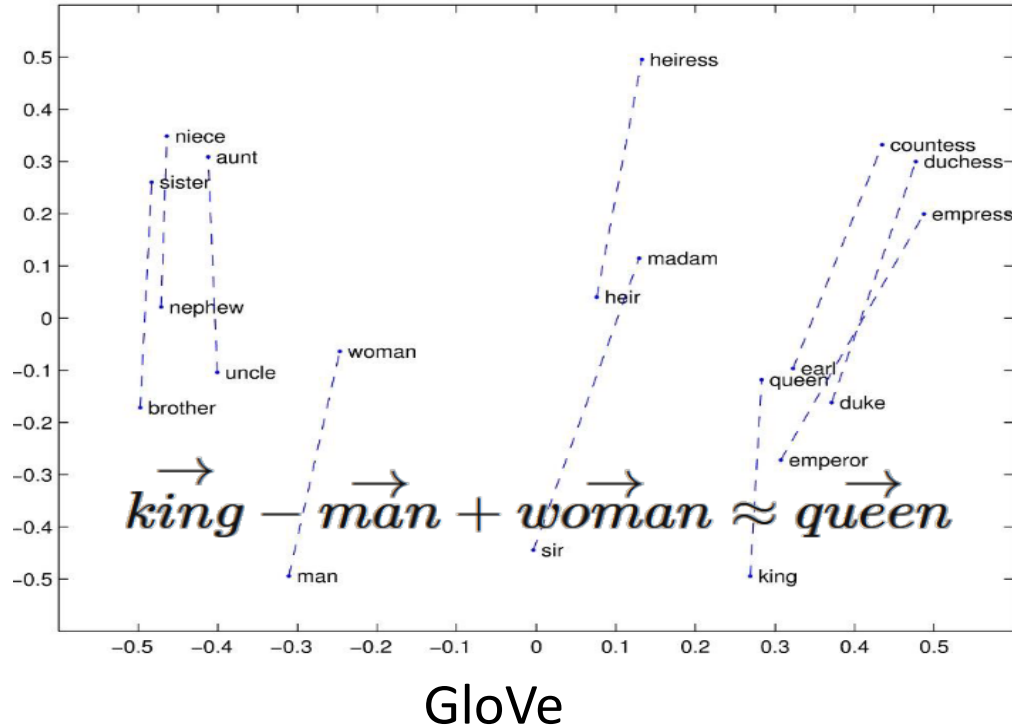
- We use Pieman story listening:
 - 82 subjects,
 - 259 TRs (repetition time)
 - here it is 1.5 sec.

Example: "I began my illustrious carrier in journalism in the Bronx where toiled as a hard boiled reporter for the team"



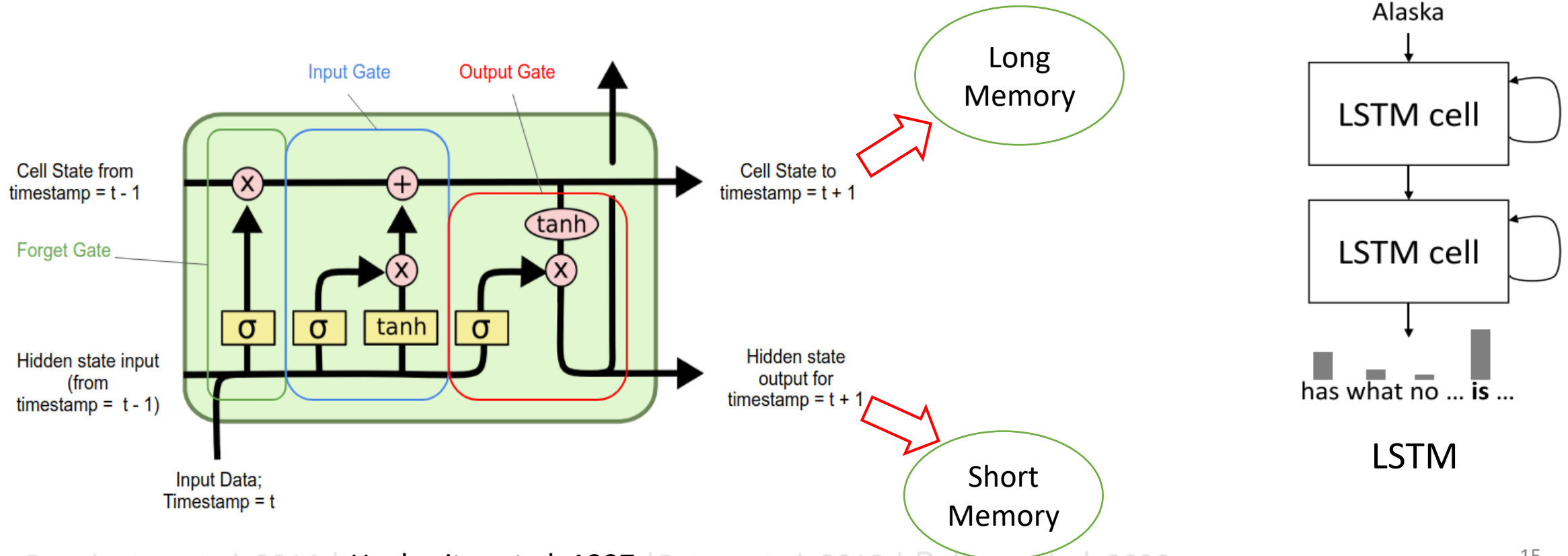
Models considered (n = 5)

- Embedding type models: GloVe



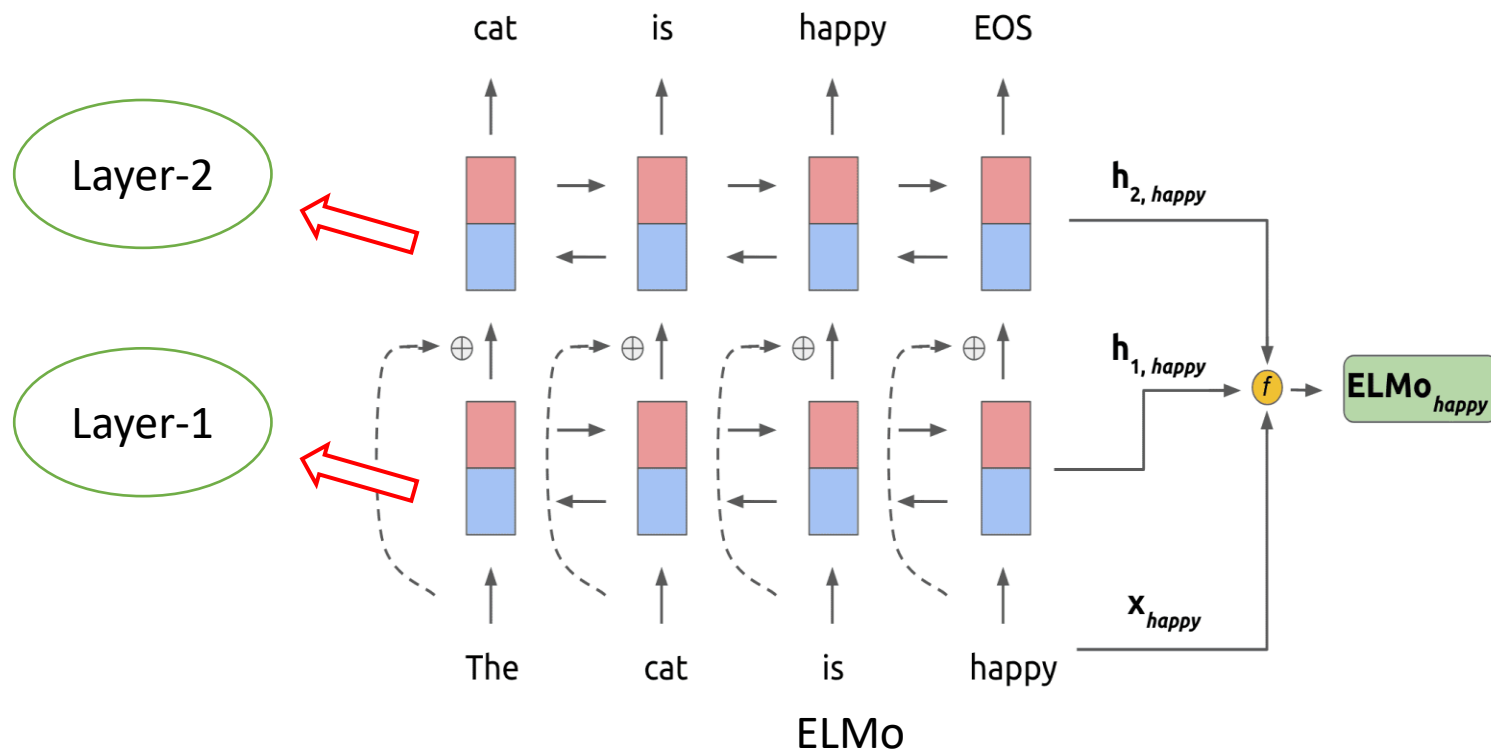
Models considered (n = 5) Contd.

- Embedding type models: GloVe
- Recurrent type models: LSTM, RandLSTM



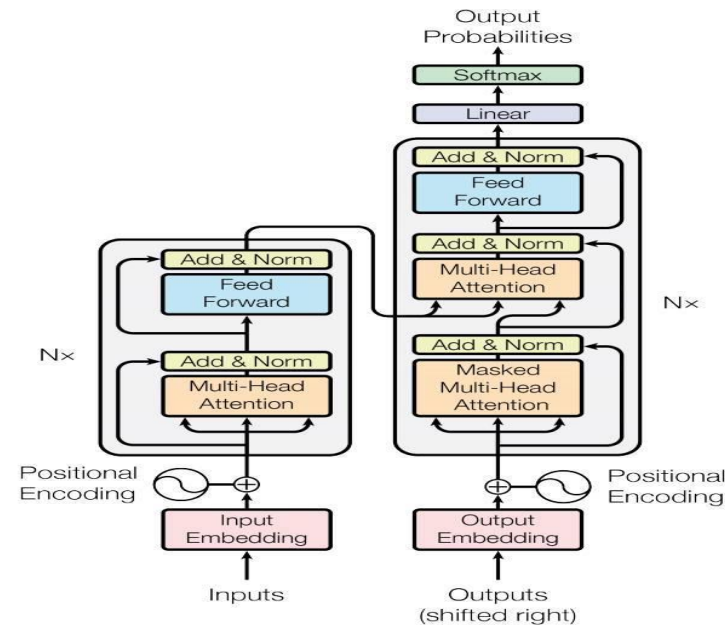
Models considered (n = 5) Contd.

- Embedding type models: GloVe
- Recurrent type models: LSTM, RandLSTM
- Contextualized pretrained models: ELMo, Longformer



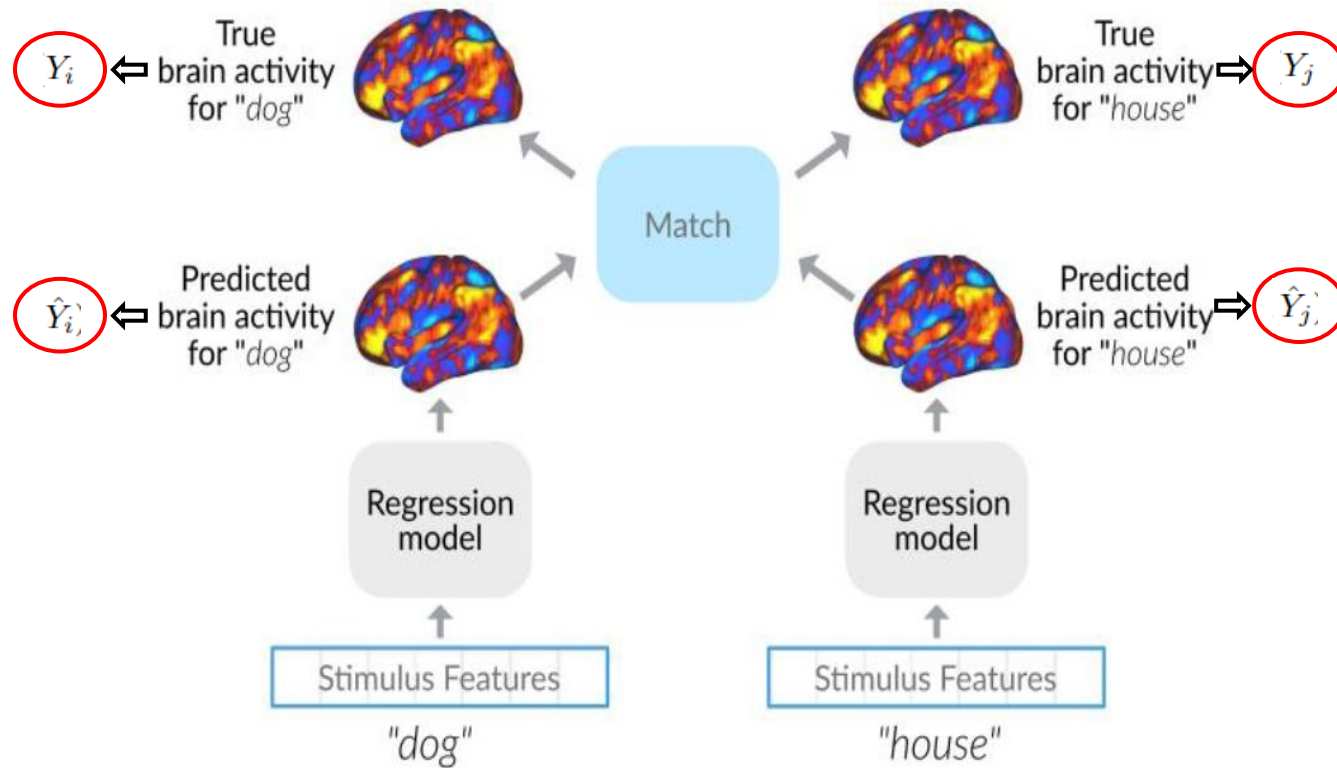
Models considered (n = 5) Contd.

- Embedding type models: GloVe
- Recurrent type models: LSTM, RandLSTM
- Contextualized pretrained models: ELMo, Longformer



Longformer: Transformer with
Global + local sliding window

Evaluation Metrics: 2V2 and Pearson



2V2 Accuracy

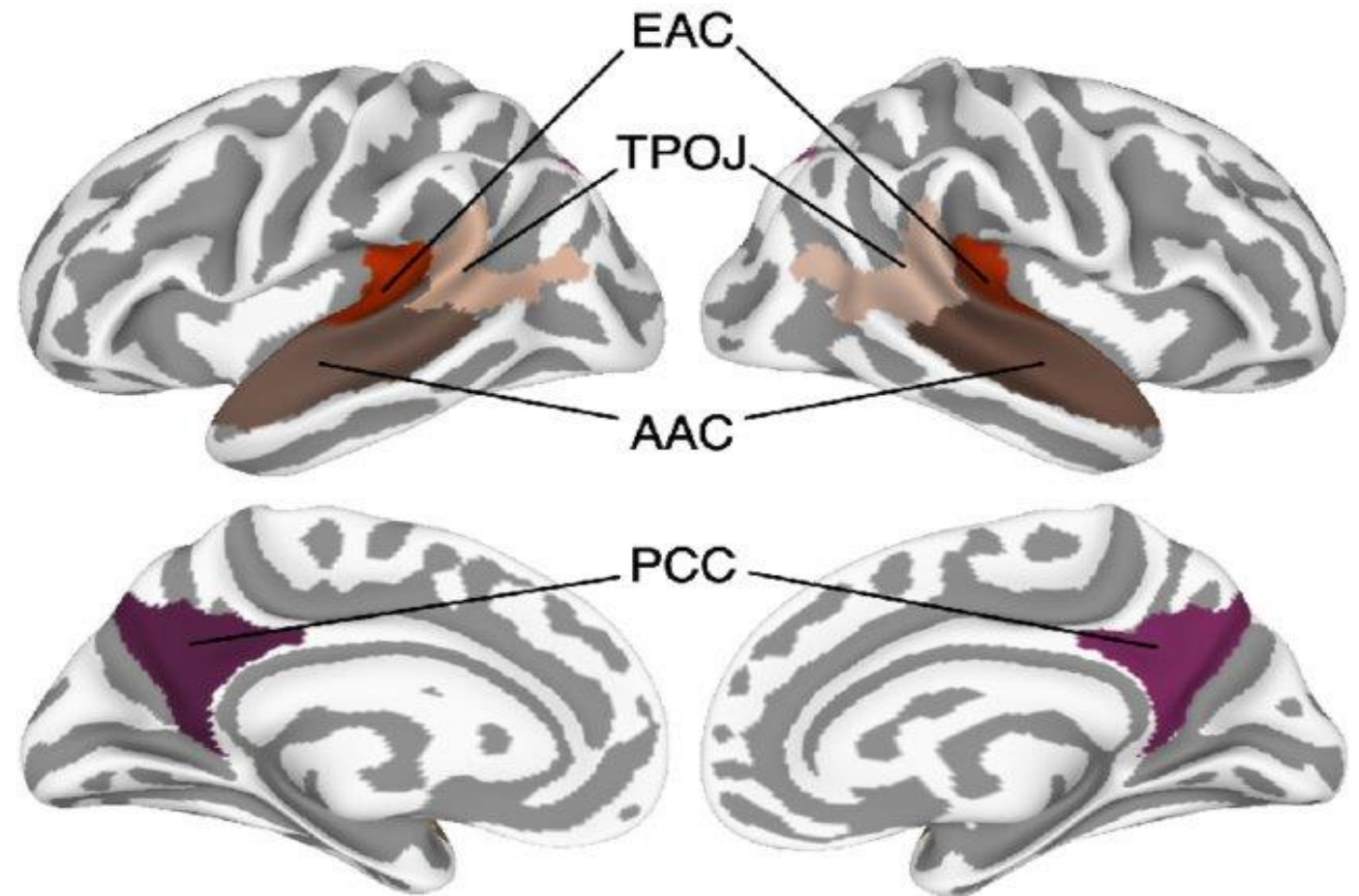
2V2 Accuracy =

$$\frac{1}{N_{C_2}} \sum_{i=1}^{N-1} \sum_{j=i+1}^N I[\{ \cos D(Y_i, \hat{Y}_i) + \cos D(Y_j, \hat{Y}_j) \} < \{ \cos D(Y_i, \hat{Y}_j) + \cos D(Y_j, \hat{Y}_i) \}]$$

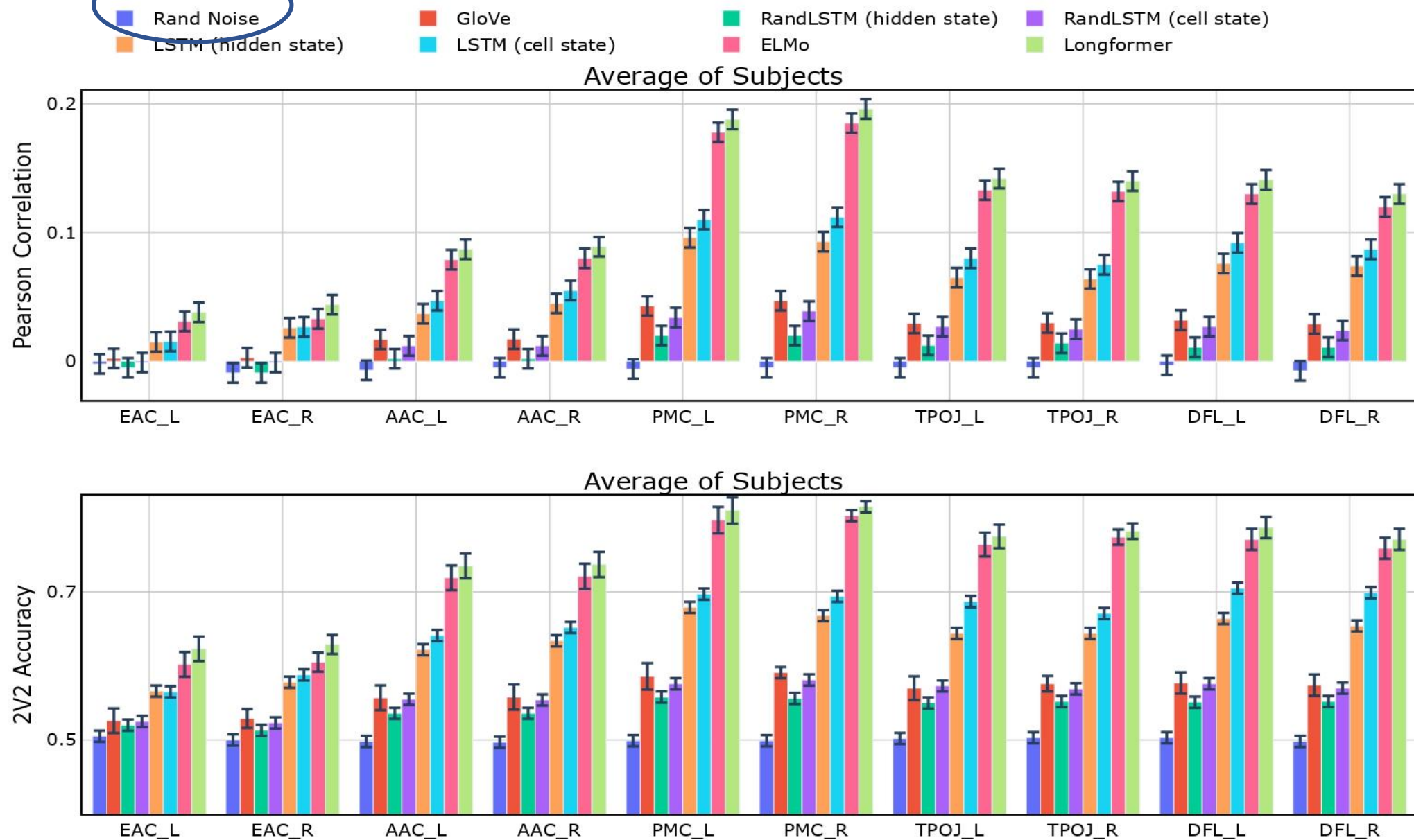
↙
Cosine distance

Language Hierarchy in Brain

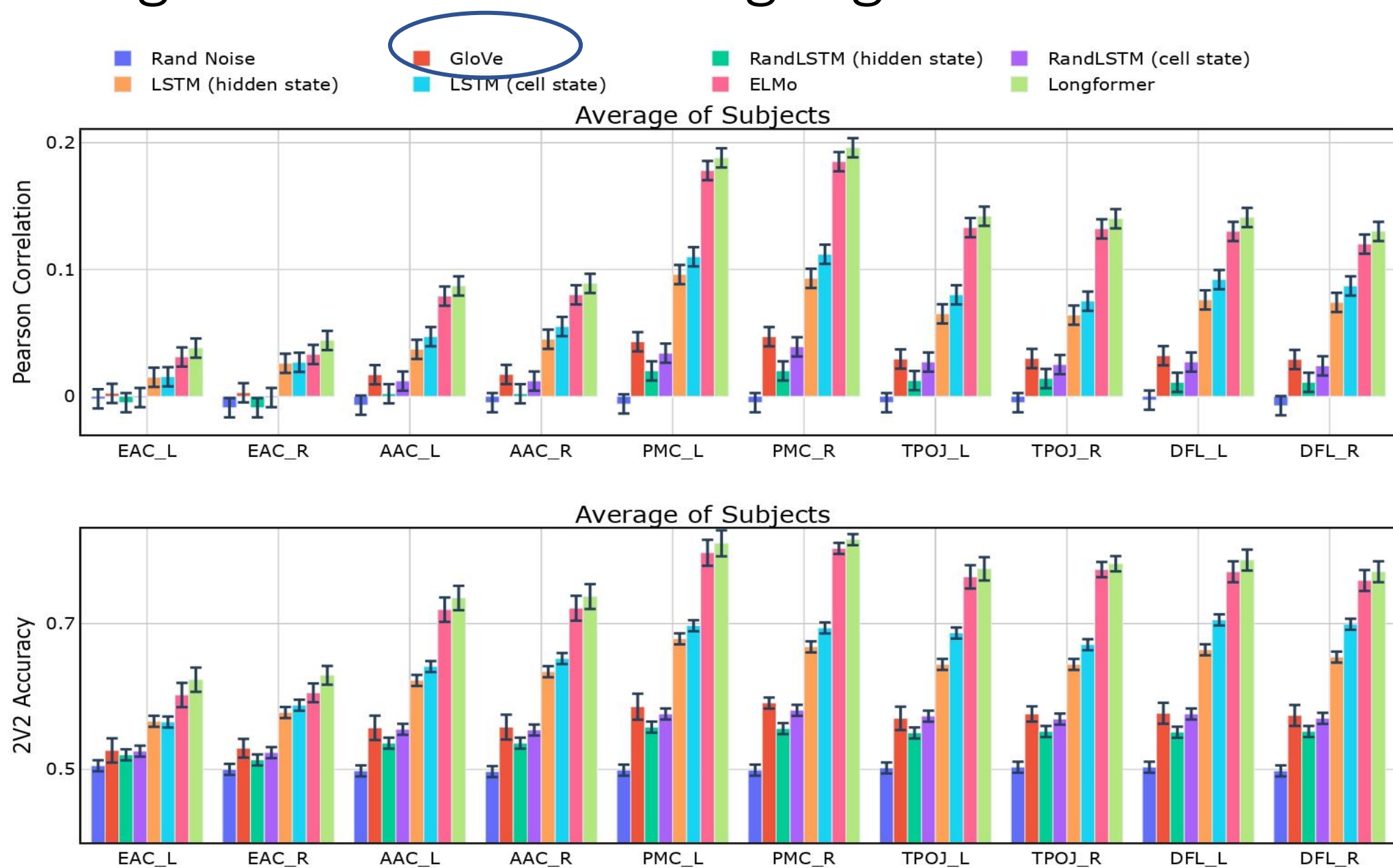
- EAC – Early auditory cortex
- AAC – Auditory association cortex
- TPOJ – Temporo-parieto-occipital junction
- DFL – Dorsal frontal lobe
- PMC – Posterior medial cortex



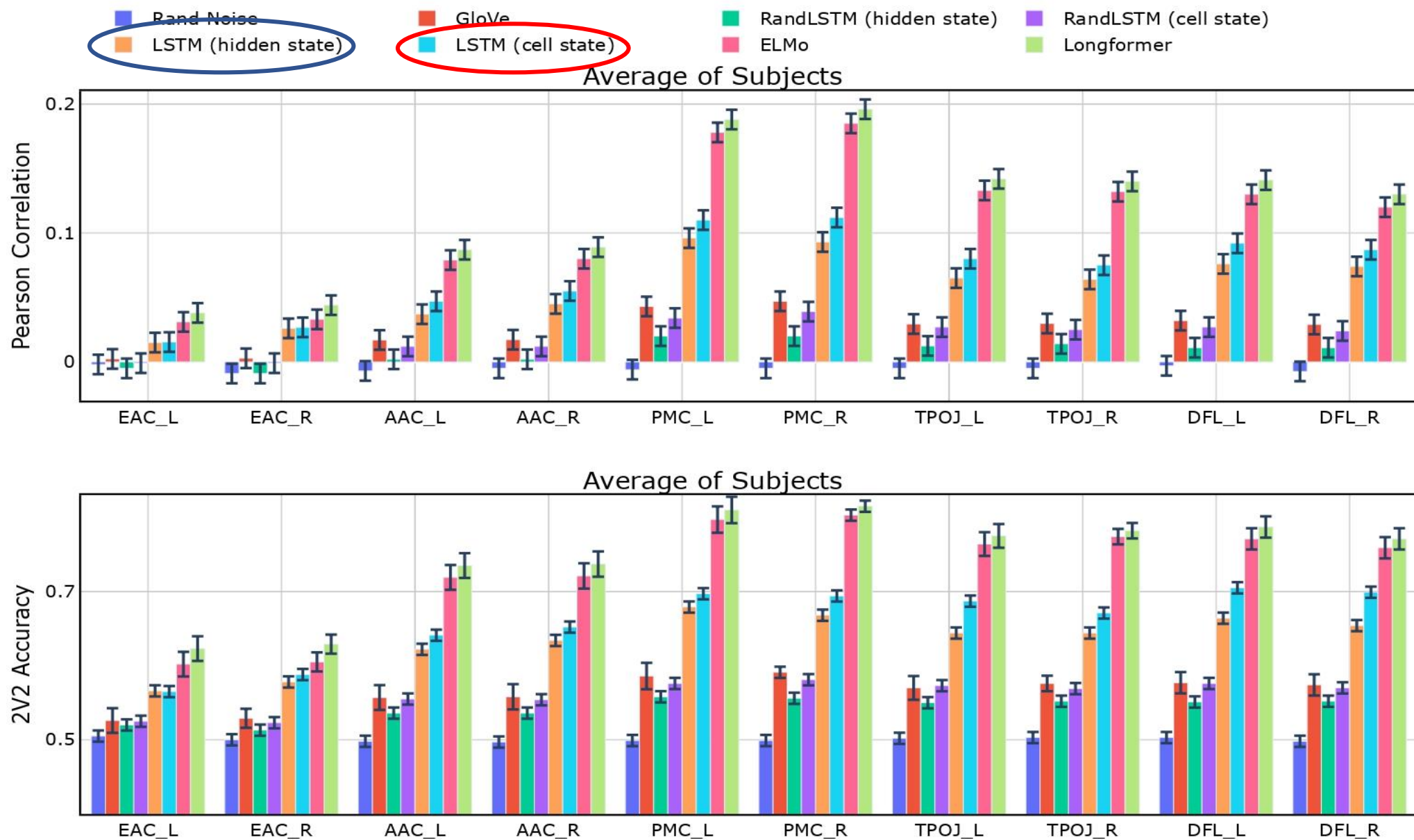
Encoding Performance of language models



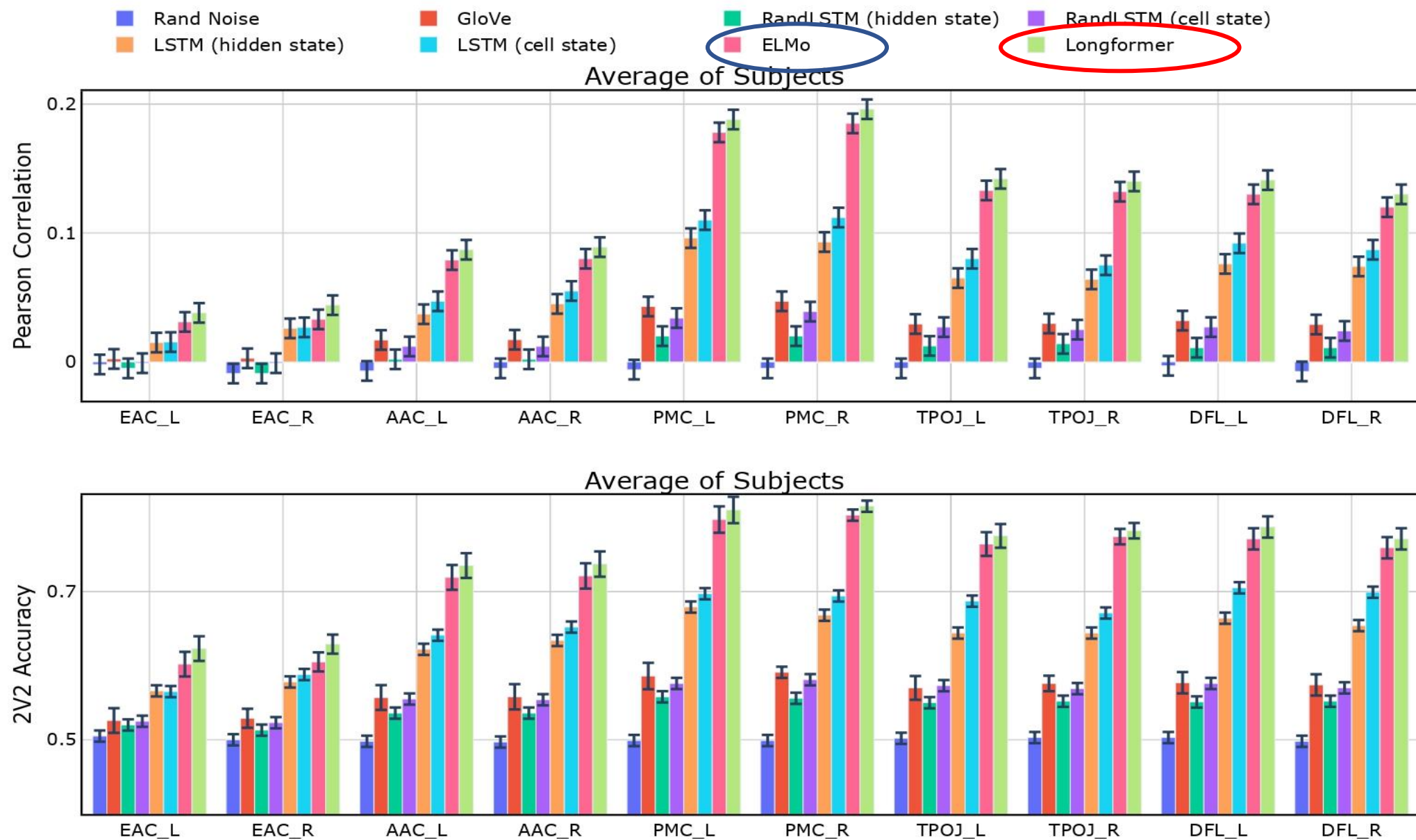
Encoding Performance of language models



Encoding Performance of language models



Encoding Performance of language models



Summary

- Language models with longer context better predict brain activity while subjects listening to stories.
- Future Directions:
 - Use more plausible language models than Transformers
 - Make hierarchical language models
 - Sub-part of the model predict different language levels: phonemes, words, POS, SRL

THANK YOU