

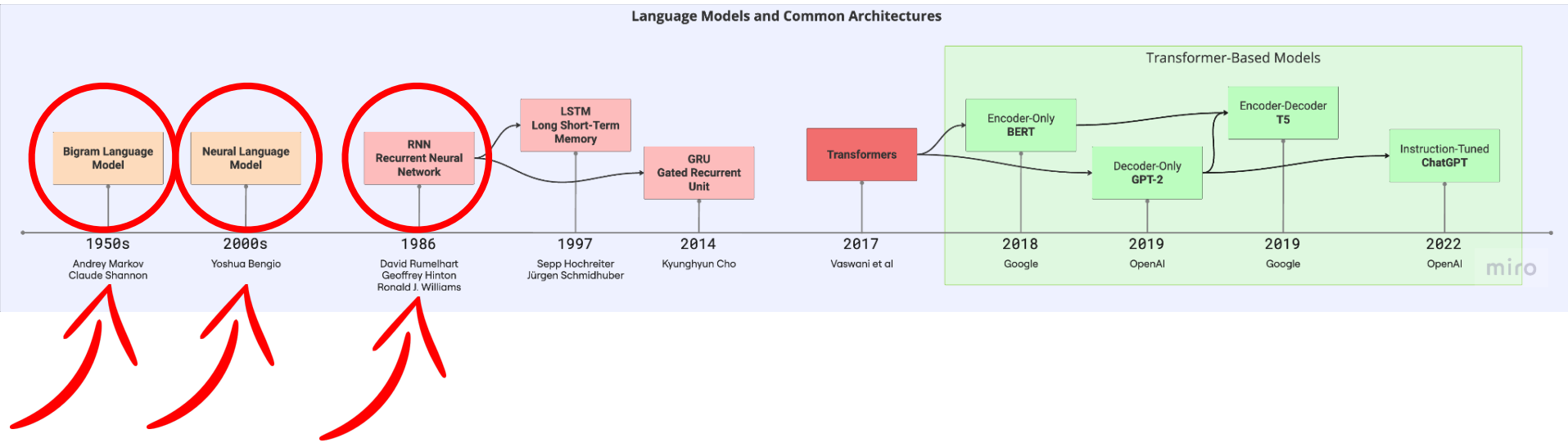
# NLP4Web

# Practice Session 9

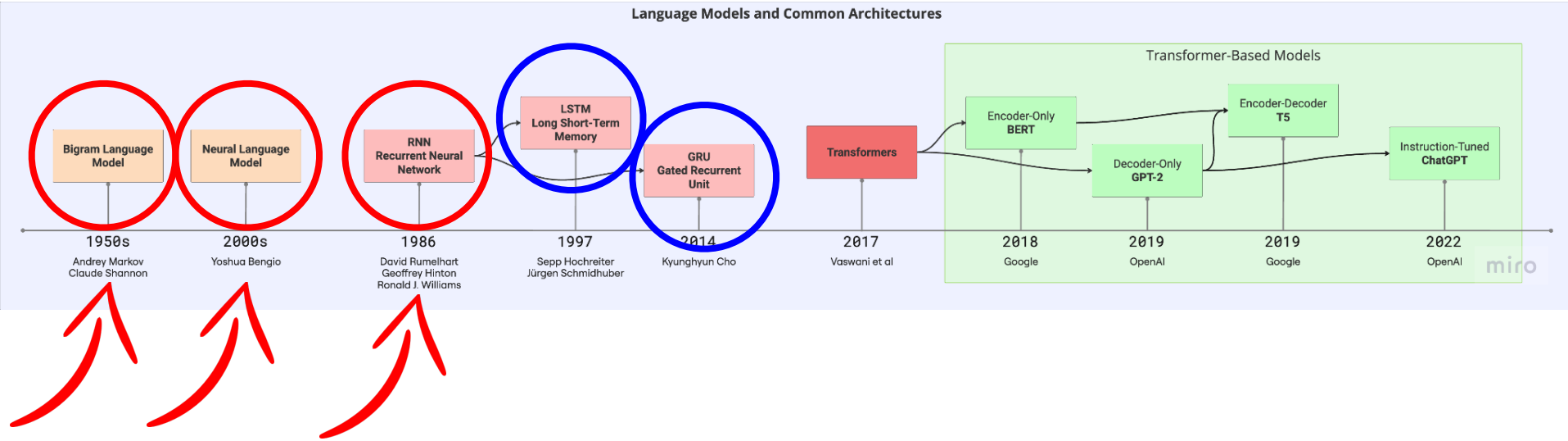
Transformers  
Decoder-only (GPT)

To not get lost in space over time, let's  
Use a **mind map**

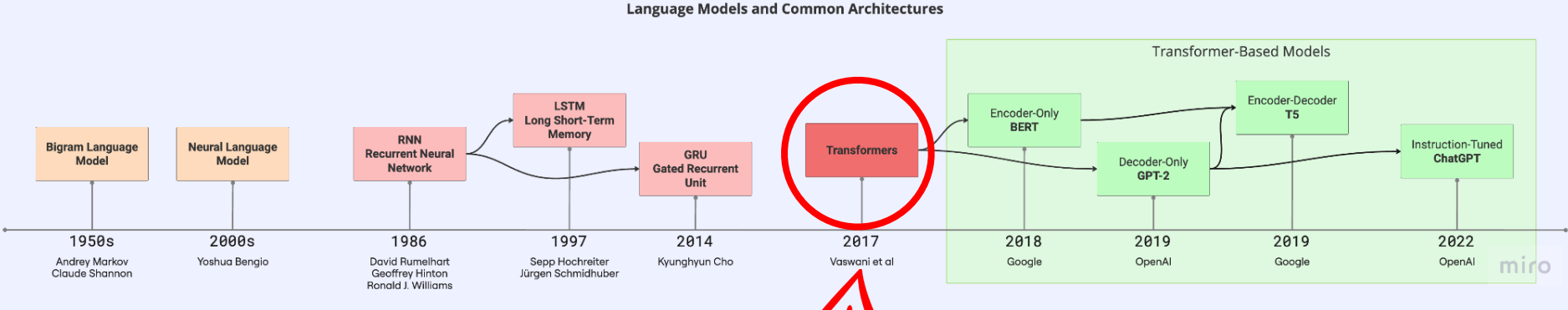
# Last time we covered: Bigram LM, NLM, RNN



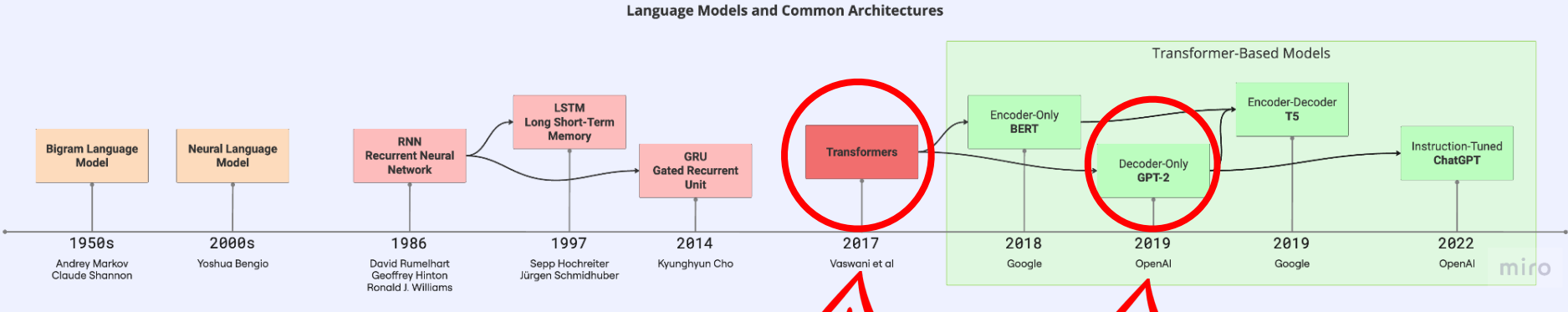
# LSTM and GRU are for HW4



# Today's subject: Transformers (Decoder-only)

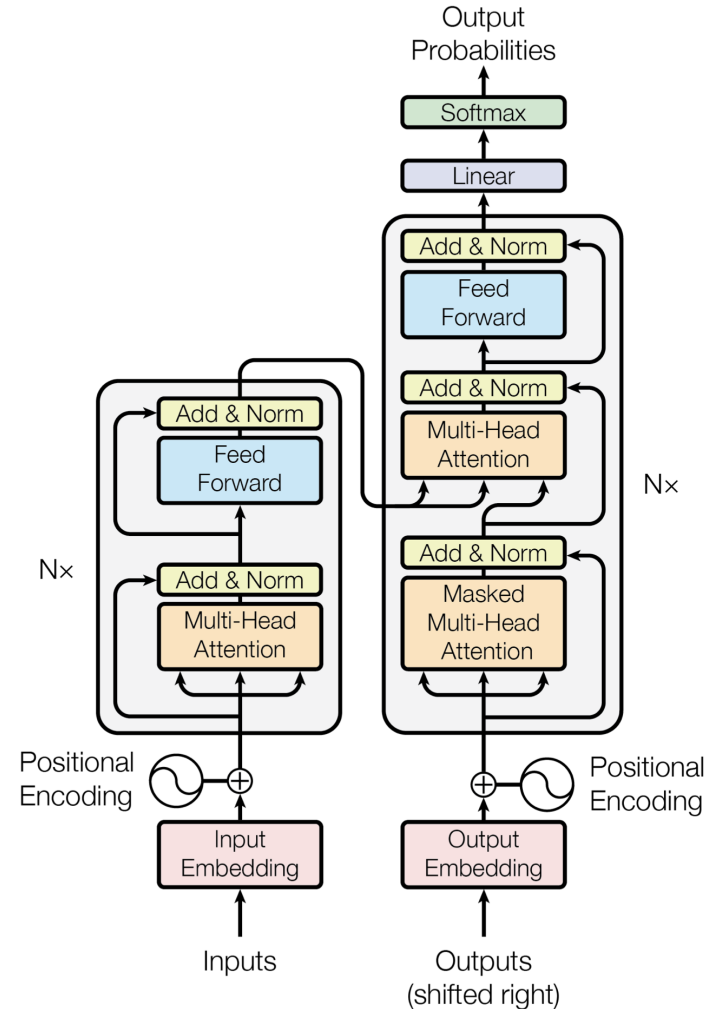


# Today's subject: Transformers (Decoder-only)



# Recap of Transformer architecture

- The main components
  - Embedding
  - **Positional Encoding**
  - **Self-Attention**
  - Feed Forward
  - **Layer Normalization**
  - Residual Connections

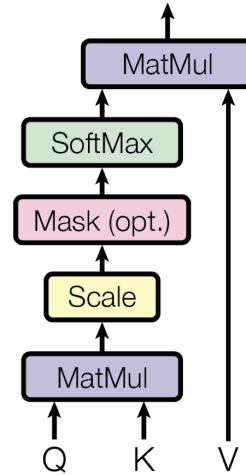


# Recap of Attention mechanism

- Scaled Dot-Product attention
- where  $\sqrt{d_k}$  is the dimension of the key vector  $k$  and query vector  $q$

$$\text{Attention}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right)V$$

Scaled Dot-Product Attention



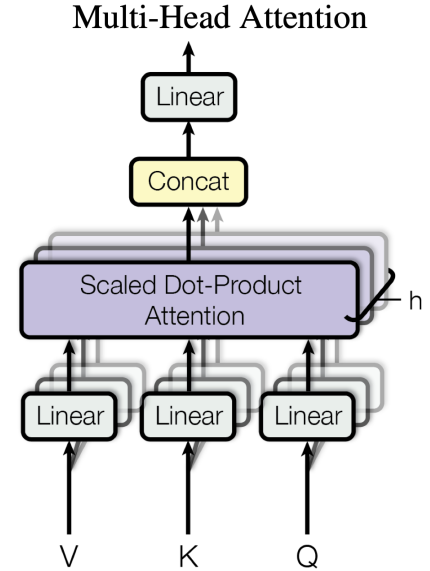


# Recap of Attention mechanism

- Multi-head attention

$$\text{MultiHead}(Q, K, V) = \text{Concat}(\text{head}_1, \dots, \text{head}_h)W^O$$

$$\text{head}_i = \text{Attention}(QW_i^Q, KW_i^K, VW_i^V)$$



# GPT: Decoder-Only (Autoregressive Architecture)

- **Decoder-only:** generates text by predicting the next word in a sequence based on prior context
- **Unidirectional:** processes text left-to-right, predicting one token at a time
- **Causal Language Modeling (CLM):** trains by predicting the next word in a sequence, avoiding future context
- **Self-attention:** focuses on relevant context in the input to determine the next output word
- **Pre-trained:** fine-tuned for specific tasks with minimal additional training
- **Text Generation:** excels at producing coherent and contextually relevant text based on prompts

