

Lecture 6 – Transformers

Provide short answers (max. 25 words each) to the following questions.

- What is the key mechanism in Transformers?
- Why is attention important in language tasks?
- What does a self-attention block compute?
- What are the three main components in self-attention?
- Why is attention computation nonlinear?
- Why do we scale the dot product in attention?
- Why do we need positional encodings in Transformers?
- What is absolute positional encoding?
- What is relative positional encoding?
- What is multi-head attention?
- Why use multiple heads in self-attention?
- How are outputs from multiple heads combined?
- What are the main components of a Transformer layer?
- What is tokenization?
- Why is tokenization challenging?
- What is a word embedding?
- How are embeddings learned in Transformers?
- What is the role of an encoder in Transformers?
- What is the role of a decoder in Transformers?
- What is an encoder-decoder architecture used for?
- What type of Transformer model is BERT?
- What is the main training strategy for BERT?
- What type of Transformer model is GPT?
- Why does GPT use masked self-attention?
- What is KV-cache in GPT?
- What is cross-attention in encoder-decoder models?
- Why is self-attention expensive for long sequences?
- What is Nyströmformer?

Multiple-choice questions

What key mechanism enables Transformers to handle long-range dependencies?

- A) Convolution
- B) Pooling
- C) Self-attention
- D) Dropout

What is the main advantage of attention in NLP tasks?

- A) Reduces vocabulary size
- B) Focuses on relevant parts of the input
- C) Removes positional encoding
- D) Speeds up tokenization

What does a self-attention block output?

- A) A single scalar
- B) Weighted sums of values for each input
- C) Only queries
- D) Only keys

Which components define attention?

- A) Queries, Keys, Values
- B) Tokens, Layers, Heads
- C) Embeddings, Dropout, Normalization
- D) Encoder, Decoder, Feedforward

How are attention weights computed?

- A) Softmax of query-key dot products
- B) Sigmoid of values
- C) ReLU of embeddings
- D) Max pooling

Why is scaling applied in dot-product attention?

- A) To increase variance
- B) To prevent large values destabilizing softmax
- C) To normalize embeddings
- D) To reduce sequence length

What is the complexity of self-attention with respect to sequence length?

- A) Linear
- B) Quadratic
- C) Cubic
- D) Logarithmic

What ensures attention weights sum to one?

- A) Normalization layer
- B) Dropout
- C) Softmax function
- D) Batch normalization

Why do Transformers need positional encoding?

- A) To reduce parameter count
- B) To normalize embeddings
- C) To compute attention faster
- D) Because self-attention ignores input order

Which positional encoding uses sinusoidal patterns?

- A) Absolute encoding
- B) Relative encoding
- C) Learned encoding
- D) Attention-based encoding

What does relative positional encoding capture?

- A) Absolute token positions
- B) Vocabulary size
- C) Offsets between query and key positions
- D) Embedding normalization

What is multi-head attention?

- A) Multiple feedforward layers
- B) Parallel attention computations with different parameters
- C) Multiple positional encodings
- D) Multiple tokenizers

Why use multiple heads?

- A) To reduce memory usage
- B) To speed up tokenization
- C) To eliminate positional encoding
- D) To capture diverse relationships between tokens

How are outputs from multiple heads combined?

- A) Averaged
- B) Concatenated and linearly transformed
- C) Summed directly
- D) Normalized only

Which components form a Transformer layer?

- A) Multi-head attention, feedforward network, normalization
- B) Convolution, pooling, dropout
- C) RNN, GRU, LSTM

D) Tokenizer, embedding, decoder only

What is the role of the feedforward network?

- A) Adds positional encoding
- B) Applies nonlinearity and increases capacity
- C) Computes attention weights
- D) Normalizes embeddings

Why is layer normalization used?

- A) To reduce vocabulary size
- B) To stabilize training
- C) To compute attention faster
- D) To eliminate Dropout

What is tokenization?

- A) Splitting text into smaller units
- B) Normalizing embeddings
- C) Computing attention weights
- D) Adding positional encoding

Why is tokenization challenging?

- A) Due to handling punctuation and word variations
- B) Requires convolution
- C) Needs positional encoding
- D) Requires dropout

What is a word embedding?

- A) One-hot representation
- B) Dense vector representation of a token
- C) Attention weight
- D) Positional encoding

How are embeddings learned?

- A) Fixed during training
- B) Computed by softmax
- C) Through trainable parameters
- D) Generated by dropout

What does an encoder do?

- A) Generates next token
- B) Applies Dropout only

- C) Converts input tokens into contextual representations
- D) Normalizes embeddings

What does a decoder do?

- A) Computes attention weights only
- B) Removes embeddings
- C) Adds positional encoding
- D) Generates output tokens based on previous outputs and encoder outputs

What is encoder-decoder architecture used for?

- A) Image classification
- B) Tokenization
- C) Sequence-to-sequence tasks like translation
- D) Dropout regularization

What type of Transformer is BERT?

- A) Decoder-only
- B) Encoder-only
- C) Encoder-decoder
- D) Convolutional

What is BERT's training strategy?

- A) Supervised only
- B) Pre-training with self-supervision and fine-tuning
- C) Reinforcement learning
- D) Dropout-based training

Name one pre-training task for BERT.

- A) Next token prediction
- B) Image patch prediction
- C) Masked language modeling
- D) Speech synthesis

How is BERT adapted for downstream tasks?

- A) By adding task-specific layers and fine-tuning
- B) By freezing all layers
- C) By removing embeddings
- D) By using convolution

What type of Transformer is GPT?

- A) Encoder-only

- B) Decoder-only
- C) Encoder-decoder
- D) Convolutional

What is GPT's main objective?

- A) Masked token prediction
- B) Image patch prediction
- C) Autoregressive next-token prediction
- D) Sentence classification

Why does GPT use masked self-attention?

- A) To prevent access to future tokens
- B) To allow future tokens
- C) To compute embeddings
- D) To normalize attention

What is KV-cache used for?

- A) Storing embeddings
- B) Storing previous key-value pairs for faster generation
- C) Computing positional encoding
- D) Normalizing attention

What is cross-attention?

- A) Decoder attends to encoder outputs
- B) Encoder attends to decoder outputs
- C) Attention between tokens in same layer
- D) Attention between positional encodings

Why is self-attention expensive for long sequences?

- A) Linear complexity
- B) Quadratic complexity
- C) Cubic complexity
- D) Logarithmic complexity

What is Nyströmformer?

- A) A convolutional Transformer
- B) A vision Transformer
- C) A low-rank approximation of self-attention
- D) A recurrent Transformer