

# CS371 N Lecture 11

## Transformers, Transformer Language Modeling

### Announcements

- A3 out

Recap Attention: places a probability distribution over a sequence of  $n$  tokens with embeddings  $e_1, \dots, e_n$

Simplified version:

① Form keys  $K_i = W^K e_i$

$$K = E(W^K)^T$$

query  $q$

$S [0 \ 0 \ 1 \ 0]$



② Compute scores  $s_i = k_i^T q \propto \left[\frac{1}{6} \ \frac{1}{6} \ \frac{1}{2} \ \frac{1}{6}\right]$

③ Compute attn weights  $\alpha = \text{softmax}(s)$

④ Result (output) =  $\sum \alpha_i e_i$

## Self-attention

$E$  now gives rise to

$E$ : seq len  $\times d$

$q_i$  and  $k_i$  for each word

$W^K$ :  $d \times d$  matrix  $K = E(W^K)^T$

$W^Q$ :  $d \times d$  matrix  $Q = E(W^Q)^T$

$K, Q$ : seq len  $\times d$

$S = QK^T$        $S_{ij} = q_i \cdot k_j$

Suppose  $E = \begin{bmatrix} 1 & 0 \\ 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$        $W^Q = W^K = I$

$S = \begin{bmatrix} 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \end{bmatrix}$  matrix of similarities

$A = \text{softmax}(S) = \begin{bmatrix} 3/10 & 3/10 & 1/10 & 3/10 \\ - & - & - & - \\ 1/6 & 1/6 & 1/2 & 1/6 \\ - & - & - & - \end{bmatrix}$

Last step:

$w^v$  produces  
"values"

$$\text{Output} = A \underbrace{E(w^v)^T}_{(seq\ len \times seq\ len) \cdot (seq\ len \times d)}$$

$$\text{Out} = \text{Seq len} \times d$$

A takes a weighted sum of values according to attention weights at each position

$$\text{First row of output} = \frac{3}{10} \cdot v_1 + \frac{3}{10} \cdot v_2 + \frac{1}{10} \cdot v_3 + \frac{3}{10} \cdot v_4$$

$$\text{Third row} = \frac{1}{6} \cdot v_1 + \frac{1}{6} v_2 + \frac{1}{2} v_3 + \frac{1}{6} v_4$$

