

Natural Language processing

- ↳ Machine translation
- ↳ Parts of speech tagging
- ↳ Generating missing words
- ↳ Autocomplete

Understanding language

vectors
or a sequence of number

embedding vector

x

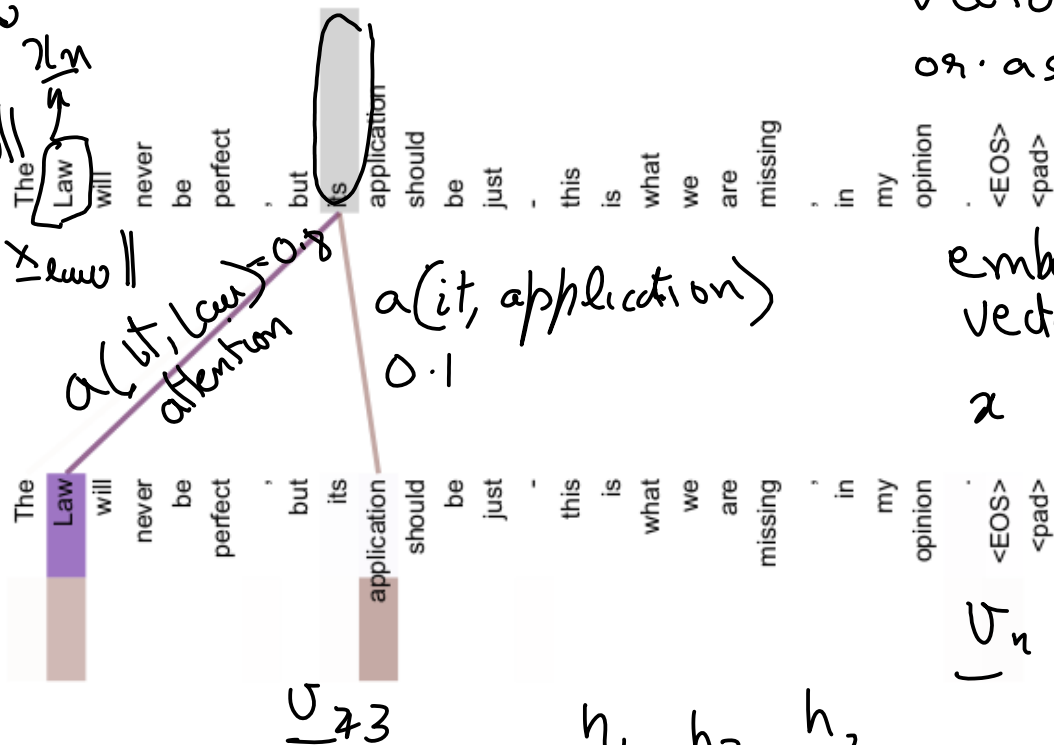
Rule $\rightarrow x_n$

$|x_{rule} - x_{law}|$

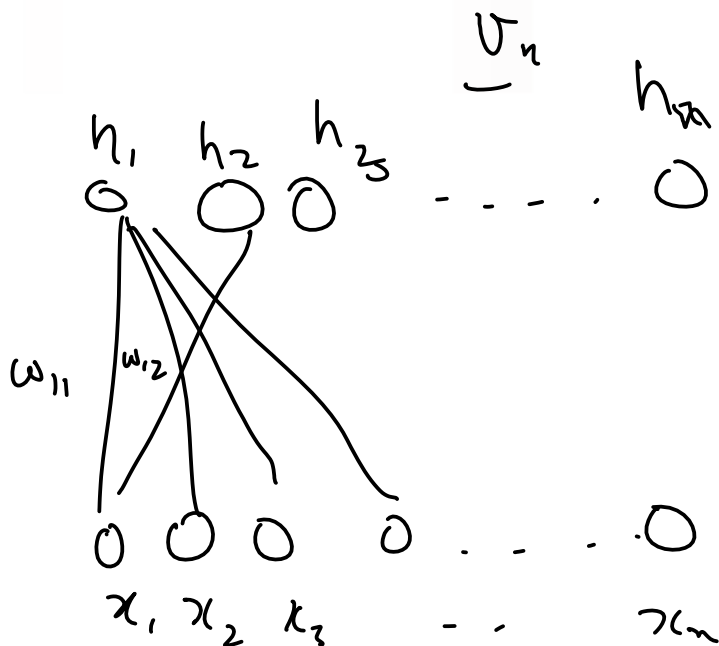
$<|x_{chous} - x_{law}|$

$a(i, l_{law}) = 0.8$
attention

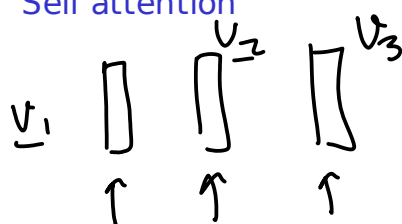
$a(i, application) = 0.1$



$$h = a(Wx + b)$$



Self attention



$$\underline{v}_n = \int_v \underline{x}_n + \beta_v$$

\uparrow \uparrow \uparrow \uparrow
 $D \times 1$ $D \times D$ $D \times 1$ $D \times 1$

$$sa(\underline{x}_m) = \sum_{n=1}^N a(\underline{x}_n, \underline{x}_m) \underline{v}_n$$

$D=4$

$N=3$

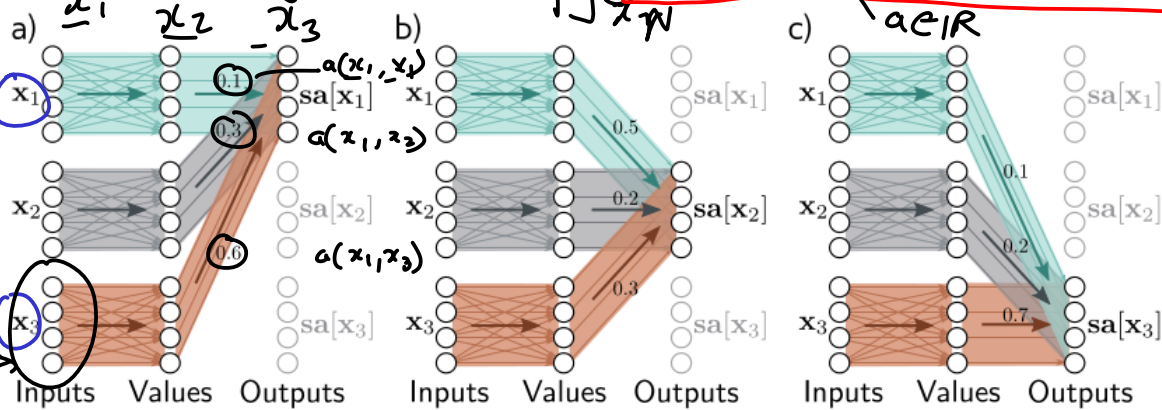
word
embedding

$$a(\underline{x}_n, \underline{x}_m) \in [0, 1]$$

$$\forall n \in [1, N] \text{ and } m \in [1, N]$$

$$\sum_{m=1}^N a(\underline{x}_n, \underline{x}_m) = 1$$

\uparrow
input



Dot product self attention

$$a(\underline{x}_n, \underline{x}_m)$$

value
vector

$$\underline{v}_n = \underline{\omega}_v \underline{x}_n + \underline{\beta}_v$$

for each word in the
sentence

$$\forall n \in [1, N]$$

key
vector

$$\underline{k}_n = \underline{\omega}_k \underline{x}_n + \underline{\beta}_k$$

query
vector

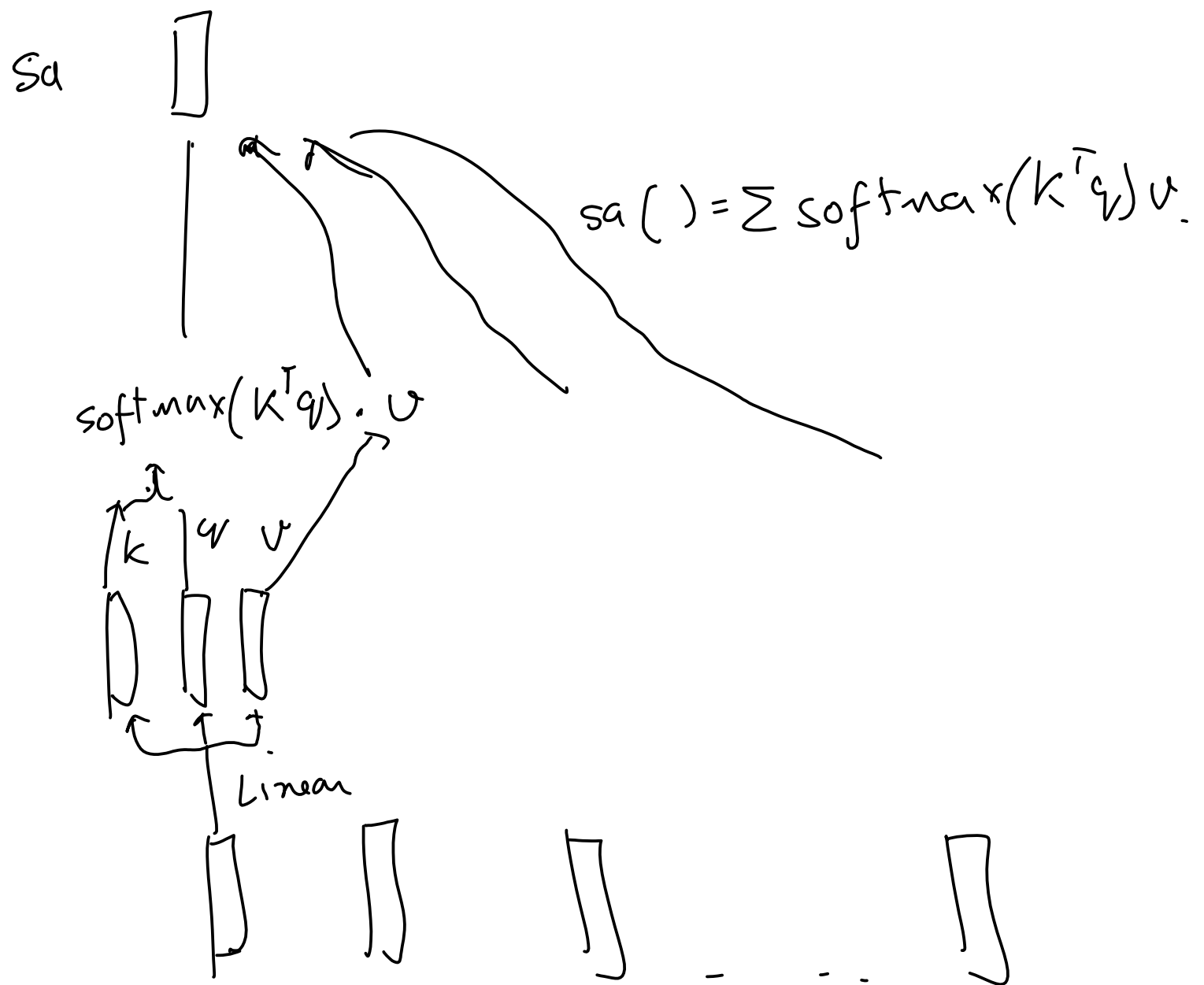
$$\underline{q}_n = \underline{\omega}_q \underline{x}_n + \underline{\beta}_q$$

$$a(\underbrace{\underline{x}_n}_{\text{output}}, \underbrace{\underline{x}_m}_{\text{input}}) = \text{softmax}_m \left(\underline{k}_m^T \underline{q}_n \right)$$

$$= \frac{\exp(\underline{k}_m^T \underline{q}_n)}{\sum_{m'=1}^N \exp(\underline{k}_{m'}^T \underline{q}_n)}$$

$$\underline{sa}(\underline{x}_n) = \sum_{m=1}^N \underbrace{\text{softmax}_m(\underline{k}_m^T \underline{q}_n)}_{\in \mathbb{R}} \underline{v}_m$$

$$\underline{v}_m = \underline{\omega}_v \underline{x}_m + \underline{\beta}_v$$



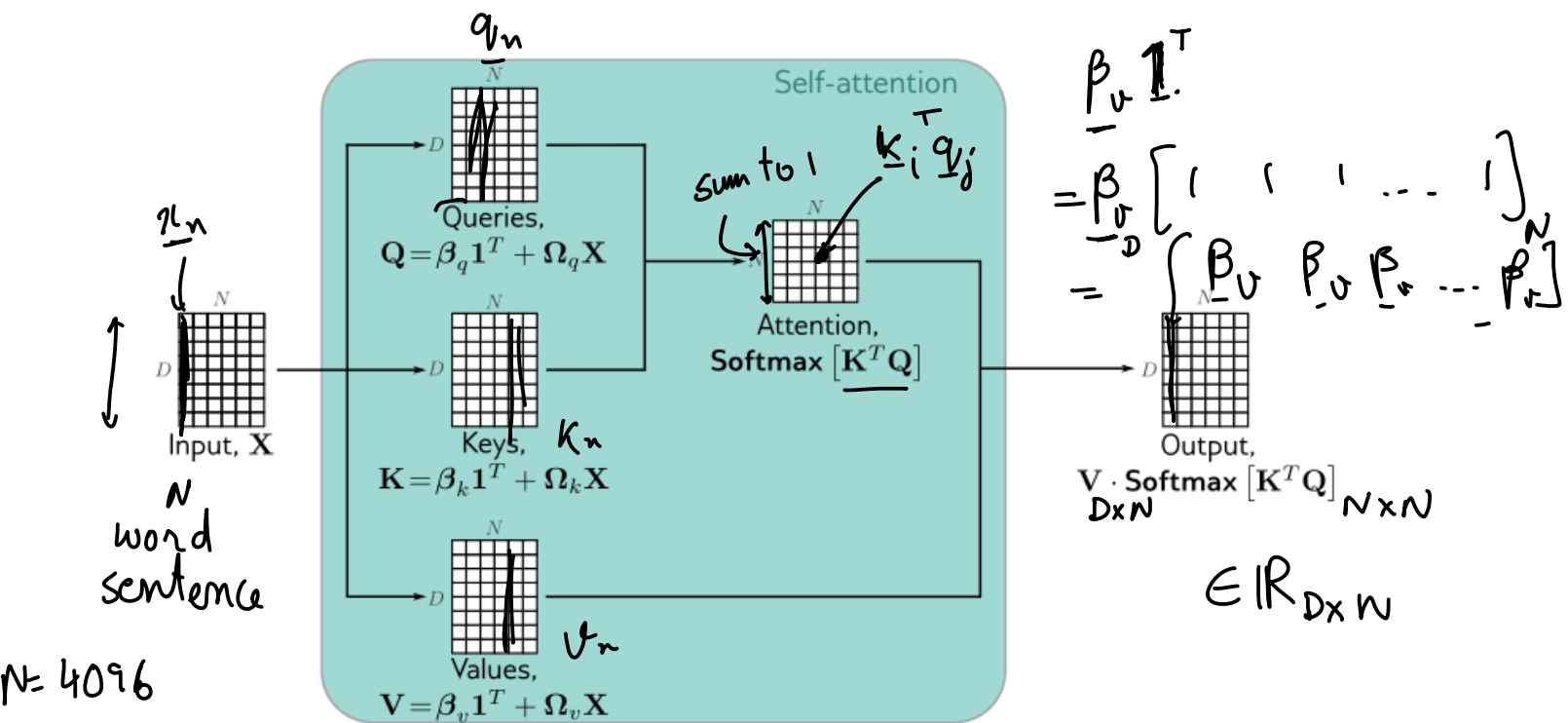
Database

key - value pair
 question - answer pair

key
 query
 value

softmax
 $(K^T q)$

query is a new key that you want to search in the database



$$sa(\underline{a}_n) = \sum_{m=1}^N \underbrace{\text{softmax}_m(\underline{k}_m^T \underline{q}_n)}_{\in \mathbb{R}} \underline{v}_m$$

Scaled Dot product self attention

$$sa[X] = V[X] \text{Softmax}(K(X)^T Q(X))$$

$$\left. \begin{array}{l} \text{Var}(K(X))_{D \times N} = 1 \\ \text{Var}(Q(X))_{D \times N} = 1 \end{array} \right\} \text{Var}(K(X)^T Q(X))_{N \times D, D \times N} = \mathbb{I}$$

$$S_a[x] = V[\hat{X}] \operatorname{softmax} \left(\frac{K(x)^T Q(x)}{\sqrt{D}} \right)$$

Positional encoding

The sentence The woman ate the raccoon has a quite different meaning to
The raccoon ate the woman.

\underline{x}_1 for The

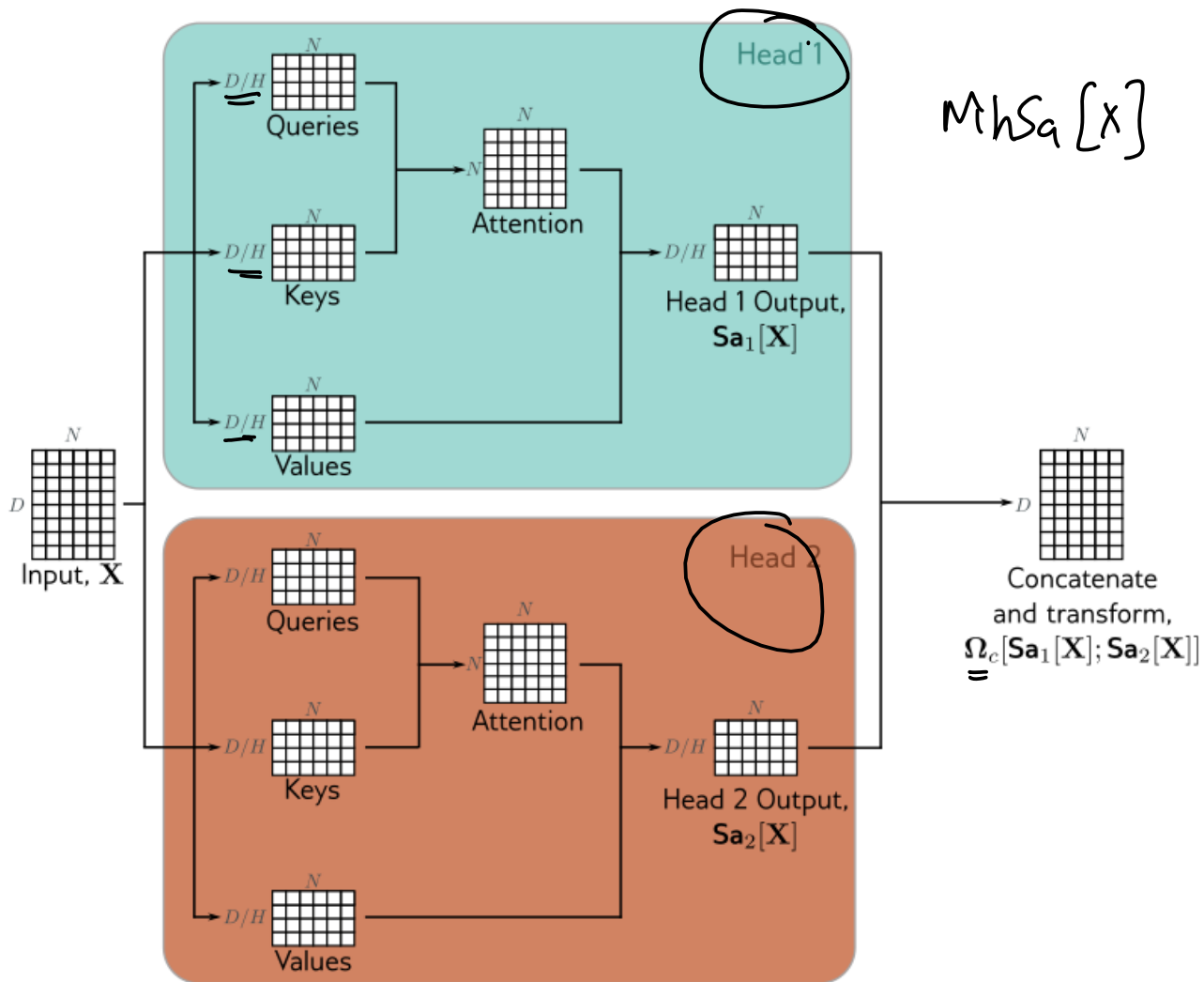
\underline{x}_2 for Raccoon

$$\underline{x}_i = \begin{bmatrix} \text{WE} \\ \vdots \\ \text{PE} \end{bmatrix}$$

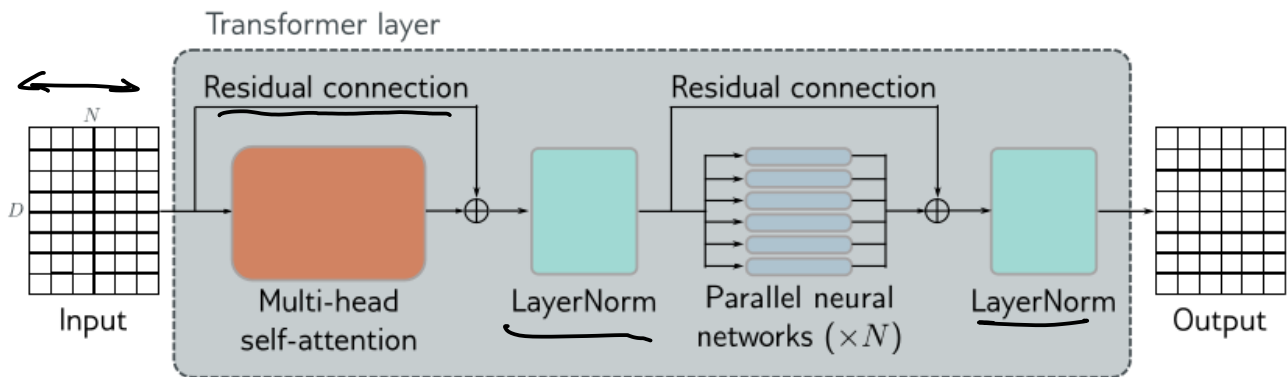
$$\text{PE}(\text{pos}, 2i) = \sin(\text{pos} / 10000^{2i/D'})$$

$$\text{PE}(\text{pos}, 2i+1) = \cos(\text{pos} / 10000^{2i+1/D'})$$

Scaled dot product self-attention



Transformer Layer



Layer Norm

$$\text{Batch Norm}(x_n) = \frac{x_n - \mu}{\sigma}$$

$$\mu = \frac{1}{B} \sum_{b=1}^B x_b$$

$$\sigma^2 = \frac{1}{B} \sum_{b=1}^B (x_b - \mu)^2$$

$$x_1, x_2, x_3 \dots x_B$$

Layer Norm

the mean and variance are computed over the "channel" dimension

a) Linear layer

$$x_1 = \begin{matrix} \text{height} \\ \text{width} \end{matrix}$$

$$h_l = \begin{matrix} \text{height} \\ \text{width} \end{matrix} \quad \begin{matrix} \uparrow \\ \text{\# of} \\ \text{hidden} \\ \text{units} \end{matrix}$$

b) Conv layer

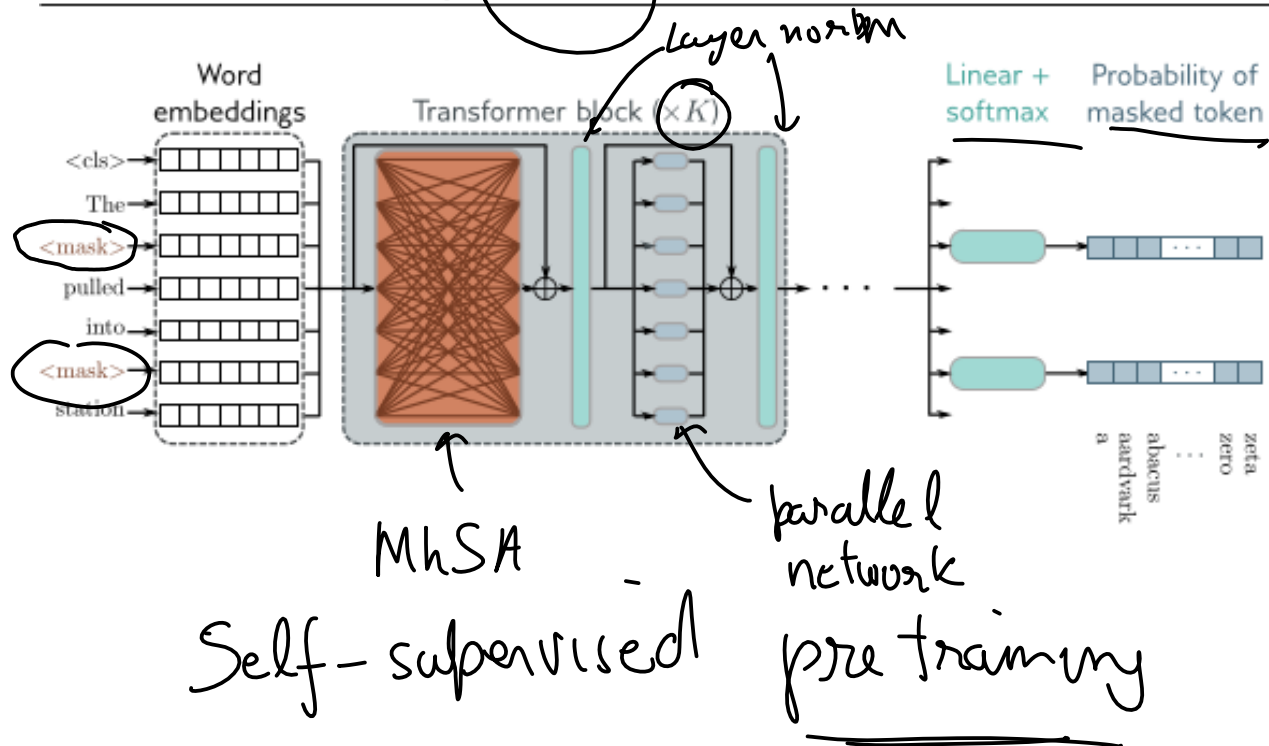
$$\begin{matrix} w \\ H \end{matrix} \quad \text{RGB} \equiv \text{channels} = 3$$

$$\text{Conv2D}(3, 16)$$

$$16 = \text{channels}$$

c) NLP vs Sq

Channels = N = number of word in
the sentence



GPT = Generative Pre trained Transformers