

Our tensor notation is discussed in Section ?? of Bayesuvius.

ℓ = number of words in a sentence segment. $\alpha \in [\ell]$, $\ell \sim 100$

L = number of words in vocabulary, $\beta \in [L]$, $L \gg \ell$

$d = d_{\underline{q}} = d_{\underline{k}} = d_{\underline{v}} = 64$, hidden dimension per head, $\delta \in [d]$.

$n_{\underline{h}} = 8$, number of heads, $\nu \in [n_{\underline{h}}]$

$D = n_{\underline{h}}d = 8(64) = 512$, hidden dimension for all heads. $\Delta \in [D]$

$\Lambda = 6$, number of layers in plate (a.k.a., stack)

reshaping

$$T^{\nu,\delta} \rightarrow T^{\Delta(\nu,\delta)} \quad (T^{[n_{\underline{h}}],[d]} \rightarrow T^{[D]}) \quad (1)$$

$$T^{\Delta(\nu,\delta)} \rightarrow T^{\nu,\delta} \quad (T^{[D]} \rightarrow T^{[n_{\underline{h}}],[d]}) \quad (2)$$

concatenation

$$T^{[n]} = (T^0, T^1, \dots, T^{n-1}) = (T^\nu)_{\nu \in [n]} \quad (3)$$

Hadamard product (element-wise, entry-wise multiplication)

$$T^{[n]} * S^{[n]} = (T^\nu S^\nu)_{\nu \in [n]} \quad (4)$$

Matrix multiplication ($T^{[n]} = T^{[n],[1]}$ is a column vector)

$$(T^{[n]})^T S^{[n]} = \text{scalar} \quad (5)$$

$$T^{[a],[b]} S^{[b],[c]} = \left[\sum_{\beta \in [b]} T^{\alpha,\beta} S^{\beta,\gamma} \right]_{\alpha \in [a], \gamma \in [c]} \quad (6)$$

$$e^{\delta,\alpha} = \sum_{\beta} E^{\delta,\beta} x^{\beta,\alpha} \quad (e^{[d],[\ell]} = E^{[d],[L]} x^{[L],[\ell]}) \quad (7)$$

$$Q^{\nu,\delta,\alpha} = \sum_{\delta'} W_{\underline{q}}^{\nu,\delta,\delta'} e^{\delta',\alpha} \quad (Q^{[D],[\ell]} = W_{\underline{q}}^{[D],[d]} E^{[d],[\ell]}) \quad (8)$$

$$K^{\nu,\delta,\alpha} = \sum_{\delta'} W_{\underline{k}}^{\nu,\delta,\delta'} e^{\delta',\alpha} \quad (K^{[D],[\ell]} = W_{\underline{k}}^{[D],[d]} E^{[d],[\ell]}) \quad (9)$$

$$V^{\nu,\delta,\alpha} = \sum_{\delta'} W_{\underline{v}}^{\nu,\delta,\delta'} e^{\delta',\alpha} \quad (V^{[D],[\ell]} = W_{\underline{v}}^{[D],[d]} E^{[d],[\ell]}) \quad (10)$$

$$B^{\nu,\alpha,\alpha'} = \frac{1}{\sqrt{d}} \sum_{\delta} Q^{\nu,\delta,\alpha} K^{\nu,\delta,\alpha'} \quad \left(B^{[n_{\underline{h}}],[\ell],[\ell]} = \left[\frac{1}{\sqrt{d}} (Q^{\nu,[d],[\ell]})^T K^{\nu,[d],[\ell]} \right]_{\nu \in [n_{\underline{h}}]} \right) \quad (11)$$

$$A^{[n_h],[d],[\ell]} = \left[\sum_{\alpha} V^{\nu,[d],\alpha} \underbrace{\text{softmax}(B^{\nu,\alpha,[\ell]})}_{(B^*)^{\nu,\alpha,[\ell]}} \right]_{\nu \in [n_h]} \quad (12)$$

$$= [V^{\nu,[d],[\ell]} (B^*)^{\nu,[\ell],[\ell]}]_{\nu \in [n_h]} \quad (13)$$

$$A^{[n_h],[d],[\ell]} \rightarrow A^{[D],[\ell]} \quad (14)$$

- **Positional Encoding Matrix**

$E_{pos}^{[d],[\ell]}$

$$E_{pos}^{\delta,\beta} = \begin{cases} \sin\left(\frac{\beta}{10^{4\delta/d}}\right) = \sin\left(2\pi \frac{\beta}{\lambda(\delta)}\right) & \text{if } \delta \text{ is even} \\ \cos\left(\frac{\beta}{10^{4(\delta-1)/d}}\right) = \cos\left(2\pi \frac{\beta}{\lambda(\delta)}\right) & \text{if } \delta \text{ is odd} \end{cases} \quad (15)$$

$E_{pos}^{\delta,\beta}$ changes in phase by $\pi/2$ every time δ changes by 1. Its wavelength λ is independent of β , but increases rapidly with δ , from $\lambda(\delta = 0) = 2\pi * 1$ to $\lambda(\delta = d) = 2\pi * 10^4$.

- **ReLU**

For a tensor T of arbitrary shape

$$ReLU(T) = (T)_+ = \max(0, T) \quad (16)$$

max element-wise

- **Feed Forward neural net**

$$F(x^{[1],[\ell]}) = ReLU(x^{[1],[\ell]} W_1^{[\ell],[d]} + b_1^{[1],[d]}) W_2^{[d],[\ell]} + b_1^{[1],[\ell]} \quad (17)$$

$$F(x^{[\ell]}) = W_2^{[\ell],[d]} ReLU(W_1^{[d],[\ell]} x^{[\ell]} + b_1^{[d]}) + b_1^{[\ell]} \quad (18)$$

- **softmax**

$\text{softmax}()$ takes a vector and returns a vector of probabilities of the same length

$$x^{[n]} \rightarrow P^{[n]} \quad (19)$$

where

$$P^\alpha = \frac{\exp(x^\alpha)}{\sum_{\alpha \in [n]} \exp(x^\alpha)} \quad \left(P^{[n]} = \frac{\exp(x^{[n]})}{\| \exp(x^{[n]}) \|_0} \right) \quad (20)$$

For example,

$$(1, 0, 0) \rightarrow (e, 1, 1)/norm \tag{21}$$

$$(10, 0, 0) \rightarrow (e^{10}, 1, 1)/norm \approx (1, 0, 0) \tag{22}$$

For any $a \in \mathbb{R}$,

$$(a, a, a) \rightarrow (1, 1, 1)/3 \tag{23}$$