

Transformer Attention Derivative

Question 1. Attention Derivative W^O

Before starting our calculation, we should know what's attention is used in The Transformer (Vaswani et al.):

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

$$\begin{aligned} \text{MultiHead}(Q, K, V) &= \text{Concat}(\text{head}_1, \dots, \text{head}_H)W^O \\ \text{where head}_h &= \text{Attention}(QW_h^Q, KW_h^K, VW_h^V) \end{aligned}$$

$$\begin{aligned} df(X) &= \text{tr}\left(\frac{\partial f(X)}{\partial X}^T d(\text{Concat}(\text{head}_1, \dots, \text{head}_H)W^O)\right) \\ &= \text{tr}\left(\frac{\partial f(X)}{\partial X}^T \text{Concat}(\text{head}_1, \dots, \text{head}_H)dW^O\right) \end{aligned}$$

$$\frac{\partial f(X)}{\partial W^O} = \text{Concat}(\text{head}_1, \dots, \text{head}_H)^T \frac{\partial f(X)}{\partial X}$$

Question 2. Attention Derivative W^v

$$\begin{aligned} df(X) &= \text{tr}\left(\frac{\partial f(X)}{\partial X}^T dX\right) \\ &= \text{tr}\left(\frac{\partial f(X)}{\partial X}^T d\text{Concat}(\text{head}_1, \dots, \text{head}_H)W^O\right) \\ &= \text{tr}\left(\frac{\partial f(X)}{\partial X}^T d\left(\text{softmax}\left(\frac{QW_1^Q(W_1^K)^T K^T}{\sqrt{d_k}}\right)VW_1^V, \dots, \text{softmax}\left(\frac{QW_H^Q(W_H^K)^T K^T}{\sqrt{d_k}}\right)VW_H^V\right)W^O\right) \\ &= \text{tr}\left(W^O \frac{\partial f(X)}{\partial X}^T d\left(\text{softmax}\left(\frac{QW_1^Q(W_1^K)^T K^T}{\sqrt{d_k}}\right)V, \dots, \text{softmax}\left(\frac{QW_H^Q(W_H^K)^T K^T}{\sqrt{d_k}}\right)V\right)\text{diag}(W_1^V, \dots, W_H^V)\right) \end{aligned}$$

let

$$A = \left(\text{softmax}\left(\frac{QW_1^Q(W_1^K)^T K^T}{\sqrt{d_k}}\right)V, \dots, \text{softmax}\left(\frac{QW_H^Q(W_H^K)^T K^T}{\sqrt{d_k}}\right)V\right)$$

$$W^v = \begin{bmatrix} W_1^V & & \\ & \ddots & \\ & & W_H^V \end{bmatrix}$$

We can get this formula:

$$\begin{aligned}
df(X) &= \text{tr} \left(W^O \frac{\partial f(X)}{\partial X}^T d(AW^v) \right) \\
&= \text{tr} \left(W^O \frac{\partial f(X)}{\partial X}^T AdW^v \right) \\
&= \text{tr} \left((A^T \frac{\partial f(X)}{\partial X} (W^O)^T)^T dW^v \right)
\end{aligned}$$

So

$$\frac{\partial f(X)}{\partial W^v} = A^T \frac{\partial f(X)}{\partial X} (W^O)^T$$

Question 3. Attention Derivative W^K, W^Q

$$\begin{aligned}
df(X) &= \text{tr} \left(\frac{\partial f(X)}{\partial X}^T dX \right) \\
&= \text{tr} \left(\frac{\partial f(X)}{\partial X}^T d\text{Concat}(\text{head}_1, \dots, \text{head}_H) W^O \right) \\
&= \text{tr} \left(\frac{\partial f(X)}{\partial X}^T d(\text{softmax}(\frac{QW_1^Q (W_1^K)^T K^T}{\sqrt{d_k}}) VW_1^V, \dots, \text{softmax}(\frac{QW_H^Q (W_H^K)^T K^T}{\sqrt{d_k}}) VW_H^V) W^O \right) \\
&= \text{tr} \left(W^O \frac{\partial f(X)}{\partial X}^T d(\text{softmax}(\frac{QW_1^Q (W_1^K)^T K^T}{\sqrt{d_k}}), \dots, \text{softmax}(\frac{QW_H^Q (W_H^K)^T K^T}{\sqrt{d_k}})) \text{diag}(VW_1^V, \dots, VW_H^V) \right)
\end{aligned}$$

let

$$\begin{aligned}
S &= (\text{softmax}(A_1), \dots, \text{softmax}(A_H)) \\
&= \left(\text{softmax}(\frac{QW_1^Q (W_1^K)^T K^T}{\sqrt{d_k}}), \dots, \text{softmax}(\frac{QW_H^Q (W_H^K)^T K^T}{\sqrt{d_k}}) \right) \\
A &= (A_1, \dots, A_H)
\end{aligned}$$

$$V' = \begin{bmatrix} VW_1^V & & \\ & \ddots & \\ & & VW_H^V \end{bmatrix}_{Hn \times d_{model}}$$

$$\begin{aligned}
df(X) &= \text{tr} \left(W^O \frac{\partial f(X)}{\partial X}^T d(SV') \right) \\
&= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T dS \right) \\
&= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T d(\exp A \odot \Upsilon(\exp A \mathbb{I})) \right)
\end{aligned}$$

where

$$\mathbb{I} = \begin{bmatrix} I & & \\ & \ddots & \\ & & I \end{bmatrix}_{Hn \times Hn}$$

$$I = \begin{bmatrix} 1 & \cdots & 1 \\ \vdots & \ddots & \vdots \\ 1 & \cdots & 1 \end{bmatrix}_{n \times n}$$

$$\Upsilon(X_{m \times n}) = \begin{bmatrix} 1/x_{11} & \cdots & 1/x_{1n} \\ \vdots & \ddots & \vdots \\ 1/x_{m1} & \cdots & 1/x_{mn} \end{bmatrix}_{m \times n}$$

$$\begin{aligned} df(X) &= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T d(\exp A \odot \Upsilon(\exp A \mathbb{I})) \right) \\ &= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T (d(\exp A) \odot \Upsilon(\exp A \mathbb{I}) + \exp A \odot d\Upsilon(\exp A \mathbb{I})) \right) \\ &= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T (\exp A \odot \Upsilon(\exp A \mathbb{I}) \odot dA - \exp A \odot \Upsilon(\exp A \mathbb{I}) \odot \Upsilon(\exp A \mathbb{I}) \odot d(\exp A \mathbb{I})) \right) \\ &= \text{tr} \left(V' W^O \frac{\partial f(X)}{\partial X}^T (S \odot dA - S \odot \Upsilon(\exp A \mathbb{I}) \odot d(\exp A \mathbb{I})) \right) \\ &= \text{tr} \left(\left(\left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \right)^T dA - \left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \odot \Upsilon(\exp A \mathbb{I}) \right)^T (\exp A \odot dA) \mathbb{I} \right) \right) \\ &= \text{tr} \left(\left(\left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \right)^T dA - \mathbb{I} \left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \odot \Upsilon(\exp A \mathbb{I}) \right)^T (\exp A \odot dA) \right) \right) \\ &= \text{tr} \left(\left(\left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \right)^T dA \right) \right) \\ &\quad - \text{tr} \left(\left(\left(\left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \odot \Upsilon(\exp A \mathbb{I}) \right) \mathbb{I}^T \right) \odot \exp A \right)^T dA \right) \end{aligned}$$

So we can get

$$\frac{\partial f(X)}{\partial A} = \left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S - \left(\left(\left(\frac{\partial f(X)}{\partial X} (W^O)^T (V')^T \right) \odot S \odot \Upsilon(\exp A \mathbb{I}) \right) \mathbb{I}^T \right) \odot \exp A$$

$$\begin{aligned} df(X) &= \text{tr} \left(\frac{\partial f(X)}{\partial A}^T dA \right) \\ &= \text{tr} \left(\frac{\partial f(X)}{\partial A}^T dA \right) \\ &= \text{tr} \left(\gamma \frac{\partial f(X)}{\partial A}^T d(QW_1^Q (W_1^K)^T K^T, \dots, QW_H^Q (W_H^K)^T K^T) \right) \end{aligned}$$

where

$$\gamma = \frac{1}{\sqrt{d_k}}$$

let

$$\mathbb{P}_h = (\dots, E_{n \times n}^h, \dots)_{n \times nH}$$

E is the $n \times n$ identity matrix which is located in the h -th column.

$$\begin{aligned} df(X) &= \text{tr} \left(\gamma \frac{\partial f(X)}{\partial A}^T d(QW_h^Q(W_h^K)^T K^T \mathbb{P}_h) \right) \\ &= \text{tr} \left(\gamma K^T \mathbb{P}_h \frac{\partial f(X)}{\partial A}^T Qd(W_h^Q(W_h^K)^T) \right) \\ &= \text{tr} \left(\gamma (W_h^K)^T K^T \mathbb{P}_h \frac{\partial f(X)}{\partial A}^T Qd(W_h^Q) \right) \end{aligned}$$

$$\frac{\partial f(X)}{\partial W_h^Q} = \gamma Q^T \frac{\partial f(X)}{\partial A} \mathbb{P}_h^T K W_h^K$$

$$\begin{aligned} df(X) &= \text{tr} \left(\gamma K^T \mathbb{P}_h \frac{\partial f(X)}{\partial A}^T Qd(W_h^Q(W_h^K)^T) \right) \\ &= \text{tr} \left(\gamma K^T \mathbb{P}_h \frac{\partial f(X)}{\partial A}^T QW_h^Q d((W_h^K)^T) \right) \end{aligned}$$

$$\frac{\partial f(X)}{\partial W_h^K} = \gamma K^T \mathbb{P}_h \frac{\partial f(X)}{\partial A}^T QW_h^Q$$

REFERENCES

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