

First we show that, $\nabla_A \text{tr}(AB) = B^\top$ for two matrices A and B . By definition,

$$\begin{aligned} \text{tr}(AB) &= \sum_{j=1}^n \sum_{i=1}^m a_{ji} b_{ij} \\ \implies \frac{\partial}{\partial a_{ij}} \text{tr}(AB) &= b_{ji} \\ \implies \nabla_A \text{tr}(AB) &= B^\top \end{aligned}$$

Also, $\text{tr}(AB) = \text{tr}(BA)$

Now, lets assume $\text{tr}(ABA^\top C) = \text{tr}(XY^\top C) = \mathbf{H}(X, Y)$ where $X = f(A) = AB$ and $Y = A$.

Using chain rule:

$$\begin{aligned} \nabla_A \mathbf{H}(X, Y) &= \nabla_X \mathbf{H}(X, Y) \nabla_A X + \nabla_Y \mathbf{H}(X, Y) \nabla_A Y \\ &= \nabla_X \mathbf{H}(X, Y) \nabla_A AB + \nabla_Y \mathbf{H}(X, Y) \nabla_A A \\ &= \nabla_X \mathbf{H}(X, Y) B^\top + \nabla_Y \mathbf{H}(X, Y) \end{aligned}$$

Now,

[The proof for the highlighted text is mentioned below](#)

$$\begin{aligned} \nabla_X \mathbf{H}(X, Y) &= \nabla_X \text{tr}(XY^\top C) \\ &= (Y^\top C)^\top \\ &= C^\top Y \\ &= C^\top A \end{aligned}$$

And,

$$\begin{aligned} \nabla_Y \mathbf{H}(X, Y) &= \nabla_Y \text{tr}(XY^\top C) \\ &= \nabla_Y \text{tr}(Y^\top CX) \\ &= (\nabla_Y^\top \text{tr}(Y^\top CX))^\top \\ &= CX \\ &= CAB \end{aligned}$$

Thus, $\nabla_A \text{tr}(ABA^\top C) = C^\top AB^\top + CAB$

Matrix-by-matrix derivative formula

Tanmoy Chakraborty

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Given:

- A is an $m \times n$ matrix.
- B is an $n \times p$ matrix.

We want to compute the derivative of $C = AB$ with respect to A , where C is an $m \times p$ matrix. We are interested in how each element of C changes with respect to each element of A .

Step-by-Step Derivation

Let $C = AB$, where:

$$C_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

i.e., the (i, j) -th element of C is the dot product of the i -th row of A and the j -th column of B .

Now, we compute the derivative of C_{ij} with respect to A_{pq} . There are two cases:

1. **When $p = i$:** In this case, A_{ik} is directly involved in the sum, so the derivative will be non-zero when $q = k$.

2. **When $p \neq i$:** The derivative will be zero because the row of A corresponding to p does not influence the row corresponding to i .

The derivative of C_{ij} with respect to A_{pq} is:

$$\frac{\partial C_{ij}}{\partial A_{pq}} = B_{qj} \delta_{ip}$$

where δ_{ip} is the *Kronecker delta*, which is 1 if $i = p$, and 0 otherwise.

Derivative of the Matrix Product AB

Now we can express the full derivative of the matrix product AB with respect to A :

$$\frac{\partial(AB)}{\partial A} = B^T$$

This derivative can be interpreted as follows: - The derivative of the matrix product $C = AB$ with respect to A results in the matrix B^T , the transpose of B .

Tensor Formulation

In tensor form, the derivative of the matrix $C = AB$ (i.e., treating the derivative of each element of C with respect to each element of A as a separate partial derivative) is a third-order tensor, where:

$$\frac{\partial(C_{ij})}{\partial A_{pq}} = B_{qj}\delta_{ip}$$

This means that for every element of C , the derivative with respect to A depends only on the corresponding elements of B , and is zero whenever the row index p does not match the row index i of C .

Matrix Notation Summary

To summarize the key points:

$$\frac{\partial(AB)}{\partial A} = B^T$$

This is a fundamental result in matrix calculus, often used in optimization problems involving matrix products.