

## Derivative through matrix multiplication

### ① Formal derivation

Let's say we have 2 matrices  $a_{(m \times p)}$ ,  $b_{(p \times n)}$

$$Z_{(m \times n)} = a \cdot b$$

$$z_{ij} = \sum_{k=1}^p a_{ik} b_{kj} \dots \dots (i)$$

Let say we perform certain operations on  $z$  such that

$$y = f(z) \quad \{ f \rightarrow \text{combination of } + \text{ and mult} \}$$

we perform  $y$ . backward  
(some index  $k, l$ )

$$\frac{\partial y}{\partial a_{kl}} = \sum_{i,j} \frac{\partial y}{\partial z_{ij}} \frac{\partial z_{ij}}{\partial a_{kl}} = \sum_{i,j} \frac{\partial y}{\partial z_{ij}} \frac{\partial z_{ij}}{\partial a_{kl}}$$

$$\frac{\partial z_{ij}}{\partial a_{kl}} = 0 \quad \{ i \neq k \} \text{ from (i)}$$

$$\Rightarrow \frac{\partial y}{\partial a_{kl}} = \sum_j \frac{\partial y}{\partial z_{kj}} \frac{\partial z_{kj}}{\partial a_{kl}} \dots \dots (ii)$$

$$z_{kj} = \sum_{i=1}^p a_{ki} b_{ij} \quad \{ \text{same as (i) with different index} \}$$

$$\frac{\partial z_{kj}}{\partial a_{kl}} = b_{lj} \dots \dots \text{substituting in (ii)}$$

$$\frac{\partial y}{\partial a_{KL}} = \sum_j \frac{\partial y}{\partial z_{Kj}} \cdot b_{Lj}$$

$$= \sum_j \frac{\partial y}{\partial z_{kj}} b^T_{jl}$$

$$\{ (b_{ij}) = (b^T)_{ji} \}$$

↓

↓  
This equation is similar to (i) in the sense that it represents matrix multiplication ~~term~~ term.

so for all ~~2~~ terms in a

$$\frac{\partial y}{\partial a} = G \cdot b^T \quad \left[ \begin{array}{l} \text{Where } G \text{ is } \frac{\partial y}{\partial z} \\ G_{(m \times n)} \end{array} \right]$$

$$\frac{\partial y}{\partial b} = a^T \cdot G \quad \left[ \text{where } G \text{ is } \frac{\partial y}{\partial z} \right]$$

In our problem  $\alpha$  is represented by 'gradients' parameter. (that is passed back from succeeding operation)

parameter - C

Then what about 'gradients == None' case. ~~This~~ This occurs when there are no succeeding operations

i.e. ~~the~~ ~~the~~ ~~the~~

i.e.  ~~$y = b(z) = z$~~   
 $y = b(z) = z \Rightarrow \left[ \frac{\partial y}{\partial z} = G = [1]_{m \times n} \right]$   
 ↳ ones matrix of  $(m \times n)$

## Derivative through Matrix Multiplication

② Simple informal derivation (with assumptions) [A formal derivation is also present in this document]

Let us assume we are performing matrix multiplication ~~through~~ between 2 matrices ( $2 \times 2$ )

$$a = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad b = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

~~we need~~

~~here~~

$$z = a \cdot b$$

[ $\cdot$  represents matmul here]

$$z = \begin{bmatrix} z_{11} & z_{12} \\ z_{21} & z_{22} \end{bmatrix} \quad \{ z \text{ represented as } \}$$

$$z_{11} = a_{11} \times b_{11} + a_{12} \times b_{21}$$

$$z_{12} = a_{11} \times b_{12} + a_{12} \times b_{22}$$

... So on

$$\frac{\partial z}{\partial a} = \begin{bmatrix} \partial z / \partial a_{11} & \partial z / \partial a_{12} \\ \partial z / \partial a_{21} & \partial z / \partial a_{22} \end{bmatrix} \quad \{ z.backward() \}$$

$$\frac{\partial z}{\partial a_{11}} = \frac{\partial z_{11}}{\partial a_{11}} + \frac{\partial z_{12}}{\partial a_{11}} + \frac{\partial z_{21}}{\partial a_{11}} + \frac{\partial z_{22}}{\partial a_{11}}$$

$$= b_{11} + b_{12} + 0 + 0$$

$$\frac{\partial z}{\partial a_{11}} = b_{11} + b_{12}$$

similarly performing for all  $a$

$$\frac{\partial z}{\partial a} = \begin{bmatrix} b_{11} + b_{12} & b_{12} + b_{22} \\ b_{11} + b_{12} & b_{12} + b_{22} \end{bmatrix}$$

$$\frac{\partial z}{\partial a} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} b_{11} & b_{12} \\ b_{12} & b_{22} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}_{2 \times 2} \cdot b^T$$



similarly performing for b

$$\frac{\partial z}{\partial b} = a^T \cdot [1]_{2 \times 2}$$

This ~~proof~~ doesn't generalize very well, if we were to try some other examples like  $(a_{2 \times 3} \cdot b_{3 \times 2})$  you will start noticing the pattern. so for a general we will get

$$z_{(m \times n)} = a_{(m \times p)} \cdot b_{(p \times n)}$$

$$\frac{\partial z}{\partial a} = [1]_{(m \times n)} \cdot b^T$$

$$\frac{\partial z}{\partial b} = a^T \cdot [1]_{m \times n}$$

You might be wondering where these random ones show up from. It comes from the fact that we are finding the derivative w.r.t to z itself rather than a function of z.

To generalize this further

$y = f(z)$  {  $f$  is a function made up of matmul and addition }

`y.backward()`

$$\frac{\partial y}{\partial a} = G \cdot b^T$$

$$\left\{ \bullet \frac{\partial y}{\partial z} = G \right\} \quad G \rightarrow m \times n$$

$$\frac{\partial y}{\partial b} = a^T \cdot G$$

{  $G$  is represented by 'gradients' parameter }

If gradients in 'None' representing that there is no succeeding operation.

This implies  $f(z) = z$  (you can use eqn we derived with ones)