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Vector, Matrix and Tensor Derivatives



Always has been.


$$\frac{\partial h_i}{\partial W_{jk}}$$

Wait its all scalar derivatives?

Overview

- Matrix derivatives as scalar derivatives
- Writing everything in terms of scalars
- Chain rule
- Kronocker delta function
- Examples
- Easy implementation (einstein notation + np.einsum)

Scalar derivatives

$$L = f(W)$$

$$\frac{\partial L}{\partial W} = \begin{bmatrix} \frac{\partial L}{\partial W_{11}} & \cdots & \frac{\partial L}{\partial W_{1m}} \\ \vdots & & \vdots \\ \frac{\partial L}{\partial W_{n1}} & \cdots & \frac{\partial L}{\partial W_{nm}} \end{bmatrix} = \left[\frac{\partial L}{\partial W_{ij}} \right]$$

Scalar derivatives

$$h = \sigma(Wx + b)$$

$$\frac{\partial h}{\partial W} \implies \frac{\partial h_i}{\partial W_{jk}}$$

$dh/dW, ndim = 3$

Rewriting matrix equations in scalars

$$\frac{\partial y_i}{\partial W_{jk}}$$

$$y = Wx + b$$

$$y_i = (Wx)_i + b_i = \sum_j W_{ij}x_j + b_i$$

Chain rule

$$f(x_1, \dots, x_n) = f(x) \quad x \in \mathbb{R}^n$$

$$\frac{\partial f}{\partial W_{ij}}$$

$$f(x_1(W), \dots, x_n(W))$$

$$\frac{\partial f}{\partial W_{ij}} = \frac{\partial f}{\partial x_1} \frac{x_1}{W_{ij}} + \dots + \frac{\partial f}{\partial x_n} \frac{x_n}{W_{ij}} = \sum_{k=1}^n \frac{\partial f}{\partial x_k} \frac{x_k}{W_{ij}}$$

$$\boxed{F(x, y)}, x, y \in \mathbb{R}^n \implies f(\underline{x_1}, \dots, \underline{x_n}, \underline{y_1}, \dots, \underline{y_n})$$

Remove nested equations

$$f(x) = \|Ax\|^2$$

$$f(y) = \|y\|^2, \quad y = Ax$$

$$\frac{\partial f}{\partial x_i} = \sum_k \frac{\partial f}{\partial y_k} \frac{\partial y_k}{\partial x_i}$$

Chain rule example

$$f(x) = \|Ax\|^2$$

$$\frac{\partial f}{\partial x_l} = \sum_a 2y_a A_{al}$$

$$= 2 \sum_a A_{al} y_a$$

$$= 2 A^T y$$

l_1, l_2, \dots, l_k

$$f(y) = \|y\|^2 \longrightarrow f = \sum_k y_k^2$$

$$y = Ax \longrightarrow y_i = \sum_j A_{ij} x_j$$

$$f = \sum_k y_k^2$$

$$y_i = \sum_j A_{ij} x_j$$

$$\frac{\partial f}{\partial x_l} = \sum_a \left[\frac{\partial f}{\partial y_a} \frac{\partial y_a}{\partial x_l} \right]$$

$$\frac{\partial f}{\partial y_a} = \frac{\partial}{\partial y_a} \sum_k y_k^2 = \sum_k \frac{\partial y_k^2}{\partial y_a} = \sum_k 2y_k \frac{\partial y_k}{\partial y_a}$$

$$= \underline{2y_a} \delta_{ka}$$

$$\frac{\partial y_a}{\partial x_l} = \frac{\partial}{\partial x_l} \left[\sum_j A_{aj} x_j \right] = \sum_j A_{aj} \frac{\partial x_j}{\partial x_l} = \sum_j A_{aj} \delta_{jl} = A_{al}$$

Caution

Don't repeat indices!

$$y = Wx + b$$

$$\frac{\partial y_i}{\partial W_{jk}} = \frac{\partial}{\partial W_{jk}} \left(\sum_l W_{il} x_l + b_i \right)$$

Kronecker delta function

Just a fancy way to write the identity matrix

$$\frac{\partial y_a}{\partial y_k}$$

$$= \begin{cases} 1 & \text{when } a=k \\ 0 & \text{otherwise} \end{cases} = \delta_{ak}$$

$$\delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$$

$$I = [\delta_{ij}] = \begin{bmatrix} \delta_{11} & \delta_{12} & \dots & \delta_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ \delta_{i1} & \dots & \dots & \delta_{in} \\ \vdots & \vdots & \ddots & \vdots \\ \delta_{n1} & \dots & \dots & \delta_{nn} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ \vdots & \vdots & \vdots \\ 0 & 0 & 1 \end{bmatrix}$$

$$\frac{\partial W_{ij}}{\partial W_{ab}} = \delta_{ia} \delta_{jb}$$

Kronecker delta function

Just a fancy way to write the identity matrix

$$\sum_k \gamma_k \delta_{ak} = \gamma_1 \delta_{a1} + \dots + \gamma_a \delta_{aa} + \dots + \gamma_n \delta_{an}$$

$\delta_{ij} = \begin{cases} 1 & \text{if } i = j \\ 0 & \text{otherwise} \end{cases}$

$$\begin{aligned} \sum_i \sum_j \gamma_i \gamma_j \delta_{ij} &\rightarrow \sum_i \gamma_i \sum_j \gamma_j \delta_{ij} \\ &= \sum_i \gamma_i \gamma_i \end{aligned}$$

Overview

1. Modularize your equations
2. Chain rule
3. Rewrite everything in terms of scalars

Example 1

$$\sum_r \delta_{il} \delta_{rj} = \delta_{ij}$$

$$h(W) = \sigma(Wx + b)$$

$$\frac{\partial h}{\partial W} = \frac{\partial h_i}{\partial W_{jk}}$$

$$h(y) = \sigma(y) \longrightarrow h_i = \sigma(y_i)$$

$$y = Wx + b \longrightarrow y_i = \sum_j W_{ij} X_j + b_i$$

$$\frac{\partial h_i}{\partial W_{jk}} = \sum_e \frac{\partial h_i}{\partial y_e} \frac{\partial y_e}{\partial W_{jk}}$$

$$= \sum_e \frac{\partial \sigma}{\partial y_i} \delta_{ie} \delta_{ej} X_k$$

$$= \frac{\partial \sigma}{\partial y_i} \delta_{ij} X_k$$

$$\frac{\partial h_i}{\partial y_e} = \frac{\partial \sigma(y_i)}{\partial y_e} = \frac{\partial \sigma}{\partial y_i} \frac{\partial y_i}{\partial y_e} = \frac{\partial \sigma}{\partial y_i} \delta_{ie}$$

$$\frac{\partial y_e}{\partial W_{jk}} = \frac{\partial}{\partial W_{jk}} \left(\sum_a W_{ea} X_a + b_e \right)$$

$$= \sum_a \frac{\partial W_{ea}}{\partial W_{jk}} X_a$$

$$= \sum_a \delta_{ej} \delta_{ak} X_a$$

$$= \delta_{ej} X_k$$

$$Y(p(\theta), q(\theta))$$

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Example 2

$$\frac{\partial Y}{\partial \theta} \rightarrow \frac{\partial Y_{ij}}{\partial \theta}$$

$$y(\theta) = (\underbrace{Ax(\theta)}_p) (\underbrace{Bx(\theta)}_q)^T$$

$$Y_{ij} = p_i q_j$$

$$p_i = \sum_u A_{iu} x_u$$

$$q_j = \sum_v B_{jv} x_v$$

$$\frac{\partial Y_{ij}}{\partial \theta} = \sum_k \left[\frac{\partial Y_{ij}}{\partial p_k} \frac{\partial p_k}{\partial \theta} + \frac{\partial Y_{ij}}{\partial q_k} \frac{\partial q_k}{\partial \theta} \right]$$

$$\frac{\partial Y_{ij}}{\partial p_k} = \frac{\partial p_i q_j}{\partial p_k} = \frac{\partial p_i}{\partial p_k} q_j = \delta_{ik} q_j \rightarrow \frac{\partial Y_{ij}}{\partial q_k} = \delta_{jk} p_i$$

$$\frac{\partial p_k}{\partial \theta} = \sum_u A_{ku} \frac{\partial x_u}{\partial \theta}$$

$$\frac{\partial q_k}{\partial \theta} = \sum_v B_{kv} \frac{\partial x_v}{\partial \theta}$$

$$\begin{aligned} &= \sum_k \delta_{ik} q_j \sum_u A_{ku} \frac{\partial x_u}{\partial \theta} + \delta_{jk} p_i \sum_v B_{kv} \frac{\partial x_v}{\partial \theta} \\ &= q_j \sum_u A_{iu} \frac{\partial x_u}{\partial \theta} + p_i \sum_v B_{jv} \frac{\partial x_v}{\partial \theta} \end{aligned}$$

Example 3

$$L = \text{tr}(W)$$

$$\text{tr}(W) = \sum_i W_{ii}$$

$$\frac{\partial L}{\partial W_{jk}} = \frac{\partial \sum_i W_{ii}}{\partial W_{jk}} = \sum_i \frac{\partial W_{ii}}{\partial W_{jk}} = \sum_i \delta_{ij} \delta_{ik} = \delta_{jk}$$

Implementation with einsum

- Einstein notation -> remove the sum symbol

$$y = ABx$$

$$y_i = \sum_j A_{ij} \sum_k B_{jk} x_k \implies A_{ij} B_{jk} x_k = y_i$$

$$A_{ij} B_{jk} x_k = y_i$$

$$y = \text{np.einsum}("ij, jk, k \rightarrow i", A, B, x)$$

Contraction indices

einops

einops rearrange("(h 2)(w 2) c -> h w (c 4)", X)

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Einsum example 2

$$L = \sum_i \sum_j \text{tr}(X^{(i)T} X^{(j)})$$

$X^{(i)} \in \mathbb{R}^{n \times m}$
 $i = 1, \dots, k$

$$L = \sum_i \sum_j \sum_k (X^{(i)T} X^{(j)})_{kk} = \sum_i \sum_j \sum_k \sum_l X_{lk}^{(i)} X_{lk}^{(j)}$$

```
L = np.einsum("ilk,jlk->", X, X)
```

↑ ↑
 Depends on X.shape