

Differentiation

What is Differentiation?

- Differentiation is the process of finding the **derivative** of a function.
- The **derivative** tells us **how fast something is changing**.

Example: Speed of a Car

Imagine you're driving a car, and the **speedometer** shows your speed at every moment.

- If at **time = 5 seconds**, your speed is **30 km/h**,
- And at **time = 6 seconds**, your speed is **35 km/h**,
- Then your **rate of change of speed** is $(35 - 30) / (6 - 5) = 5 \text{ km/h per second}$.

This rate of change is what **differentiation** calculates.

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Key Idea: Calculus (specifically differentiation) is used in ML to optimize models.

- **Derivative:** Measures the **instantaneous rate of change** of a function.
 - *Example:* Speed is the derivative of distance with respect to time.

Why "Instantaneous"?

- It's the rate of change at an **exact point** (not an average over time).
- *Analogy:* Your car's speedometer shows your speed **right now**, not your **average speed for the trip**.

$$\lim_{\Delta x \rightarrow 0} \frac{f(x + \Delta x) - f(x)}{\Delta x}$$

Relation to Slope:



The derivative is the slope of the tangent line to the function's curve at a point.

- *Example: For $y = x^2$, the slope at $x=2$ is 4.*

Example:

- If a hill has a steep **upward** slope, the derivative is **positive**.
- If the hill is going **downward**, the derivative is **negative**.
- If the hill is **flat**, the derivative is **zero**.



So, differentiation helps us understand whether a function is increasing, decreasing, or flat at a particular point.

Derivative of a Constant

- **Rule:** The derivative of any constant (e.g., $y=5$) is **0**.
- **Why?** Constants don't change, so their rate of change is zero.

$$\circ \text{ Example: If } y = 7, \frac{dy}{dx} = 0.$$

Maxima and Minima

- **Maxima:** The highest point of a curve (like the top of a hill).
- **Minima:** The lowest point of a curve (like the bottom of a valley).

At these points, the **derivative is zero** because the slope is flat.

Mathematically,

$$f'(x) = 0$$

means a function has reached a maximum or minimum.

To confirm whether it's a maximum or minimum, we use the **second derivative**:

- If $f''(x) > 0$, it's a **minimum** (valley).
- If $f''(x) < 0$, it's a **maximum** (peak).

Common Derivatives Cheatsheet

- **Power Rule:**
 - Example: $\frac{d}{dx}(x^3) = 3x^2$.
- **Trigonometric Functions:**
 - $\frac{d}{dx}(\sin x) = \cos x$.
 - $\frac{d}{dx}(\cos x) = -\sin x$.
- **Exponentials/Logarithms:**
 - $\frac{d}{dx}(e^x) = e^x$.
 - $\frac{d}{dx}(\ln x) = \frac{1}{x}$.

Power Rule

$$\frac{d}{dx}(x^n) = nx^{n-1}$$

Example:

$$\frac{d}{dx}(x^3) = 3x^2$$

- Bring the exponent down in front.
- Subtract 1 from the exponent.

Sum Rule

- The derivative of a sum is just the sum of the derivatives.

$$\frac{d}{dx}(f(x) + g(x)) = f'(x) + g'(x)$$

$$\frac{d}{dx}(x^3 + x^2) = 3x^2 + 2x$$

Product Rule

$$\frac{d}{dx}[f(x)g(x)] = \underbrace{f'(x)}_{\text{as is}} \underbrace{g(x)}_{\text{as is}} + \underbrace{f(x)}_{\text{as is}} \underbrace{g'(x)}_{\text{as is}}$$

Example:

If $f(x) = x^2$ and $g(x) = \sin x$ then:

$$\frac{d}{dx}(x^2 \sin x) = \underbrace{2x}_{\text{as is}} \underbrace{\sin x}_{\text{as is}} + \underbrace{x^2}_{\text{as is}} \underbrace{\cos x}_{\text{as is}}$$

Quotient Rule

For division of two functions:

$$\frac{d}{dx} \left(\frac{f(x)}{g(x)} \right) = \frac{f'(x)g(x) - f(x)g'(x)}{[g(x)]^2}$$

$$\frac{d}{dx} \left(\frac{x}{x+1} \right) = \frac{(1)(x+1) - x(1)}{(x+1)^2} = \frac{1}{(x+1)^2}$$

Chain Rule

Used when differentiating a function inside another function.

$$h(x) = f(g(x))$$

$$\frac{d}{dx} f(g(x)) = f'(g(x)) \cdot g'(x)$$

Example:

If $f(x) = \sin(x^2)$ then:

$$\frac{d}{dx}(\sin x^2) = \cos(x^2) \cdot 2x$$

Partial Differentiation

- When a function has **multiple variables**, we differentiate **one variable at a time** while keeping others constant.

If $f(x,y) = x^2 + y^3$, then:

$$\frac{\partial f}{\partial x} = 2x, \quad \frac{\partial f}{\partial y} = 3y^2$$

- This is useful in **machine learning** and **physics**, where functions often depend on multiple variables.

Example: Temperature in a Room

Imagine the temperature **T** inside a room depends on:

- x = Distance from the heater
- y = Distance from the window

The temperature function might be:

$$T(x, y) = 5x^2 + 3y + 10$$

- Here, x and y both affect T .

Now, what if we only want to know how T changes with x while ignoring y ? → That's where partial differentiation comes in!

Example: Let $f(x, y) = 3x^2y + 2xy^3$.

- **Partial derivative with respect to x :**

$$\frac{\partial f}{\partial x} = 6xy + 2y^3 \quad (\text{since } y \text{ is constant}).$$

- **Partial derivative with respect to y :**

$$\frac{\partial f}{\partial y} = 3x^2 + 6xy^2 \quad (\text{since } x \text{ is constant}).$$

Higher-Order Derivatives

- **First derivative** $f'(x)$ → tells us slope.
- **Second derivative** $f''(x)$ → tells us if the function is concave up or down.
- **Third derivative** $f'''(x)$ → used in physics (jerk in motion).

Example:

$$f(x) = x^4, \quad f'(x) = 4x^3, \quad f''(x) = 12x^2, \quad f'''(x) = 24x$$

If $f(x, y) = x^3 + 2xy$, then:

$$\frac{\partial f}{\partial x} = 3x^2 + 2y$$

$$\frac{\partial^2 f}{\partial x^2} = 6x$$

Matrix Differentiation

If \mathbf{X} is a matrix and \mathbf{w} is a vector, differentiation is done using special rules:

$$\frac{d}{d\mathbf{w}}(\mathbf{X}\mathbf{w}) = \mathbf{X}$$

- Fundamental tool in machine learning, optimization, and deep learning

Handwritten mathematical derivation:

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$$
$$AX = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} =$$
$$f_1(x_1, x_2) = x_1 + 2x_2$$
$$f_2(x_1, x_2) = 3x_1 + 4x_2$$

$$\frac{d}{dx} A x = \begin{bmatrix} \frac{df_1}{dx_1} & \frac{df_1}{dx_2} \\ \frac{df_2}{dx_1} & \frac{df_2}{dx_2} \end{bmatrix}$$

- We know f_1 & f_2
- Find out the derivatives of them wrt x_1 & x_2

$$\frac{d}{dx} A x = \begin{bmatrix} \frac{df_1}{dx_1} = 1 & \frac{df_1}{dx_2} = 2 \\ \frac{df_2}{dx_1} = 3 & \frac{df_2}{dx_2} = 4 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

- 🙌 This is the original matrix itself.

$$\frac{d}{dx} Ax = A$$

Example:

If \mathbf{X} is a matrix and \mathbf{w} is a vector, differentiation is done using special rules:

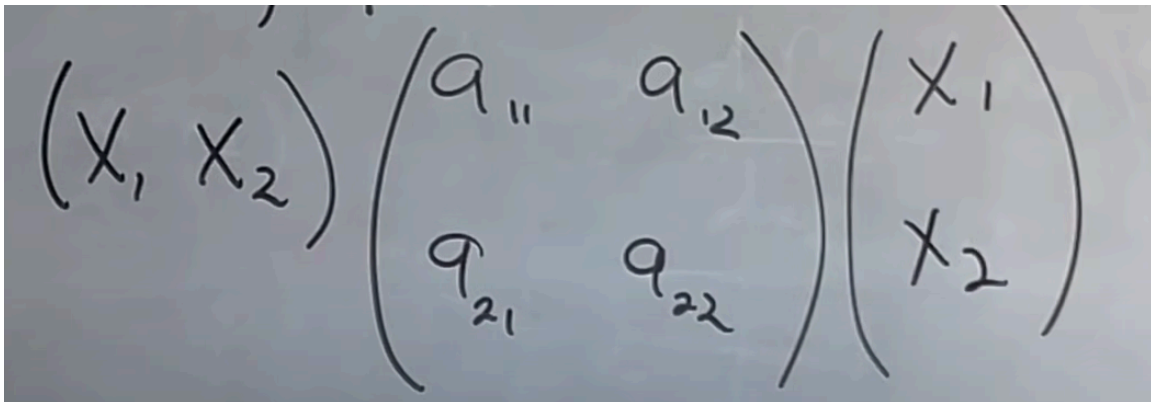
$$\frac{d}{d\mathbf{w}} (\mathbf{X}\mathbf{w}) = \mathbf{X}$$

This is heavily used in
gradient descent optimization.

Little tougher example →

Derivative of:

$$y = \underbrace{X^T A X}$$



A photograph of a handwritten equation on a chalkboard. The equation is $(x_1, x_2) \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$. The matrix A is symmetric, with a_{12} and a_{21} in the off-diagonal positions. The vectors X and X^T are written as (x_1, x_2) and $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ respectively.

- First let's multiply A & X

$$\begin{pmatrix} x_1 & x_2 \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\begin{pmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{pmatrix}$$

The diagram shows the multiplication of a row vector (x_1, x_2) by a 2x2 matrix $\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$ and then by a column vector $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$. A green bracket under the matrix and a green arrow pointing down to the resulting vector indicate the intermediate step of multiplying the row vector by the matrix.

We called a_{12} & a_{21} as $\rightarrow a$

- Now multiply this with (x_1, x_2)

$$a_{11}x_1^2 + a_{12}x_1x_2 + a_{21}x_1x_2 + a_{22}x_2^2$$

$$a_{11}x_1^2 + \underbrace{2a_{12}x_1x_2}_{\substack{2a_{12} \\ a_{21}}} + a_{22}x_2^2$$

- We'll call this 🙌 some function $f(x_1, x_2)$

$$= f(x_1, x_2)$$

$$\frac{d}{dx} x^T A x \quad \begin{pmatrix} \frac{df}{dx_1} \\ \frac{df}{dx_2} \end{pmatrix}$$

$$\begin{pmatrix} 2a_{11}x_1 + 2a_{12}x_2 \\ 2a_{21}x_1 + 2a_{22}x_2 \end{pmatrix}$$

$$2 \begin{pmatrix} 7x_1 + 9x_2 \\ 9x_1 + 29x_2 \end{pmatrix}$$

$$2 \begin{pmatrix} 7 & 9 \\ 9 & 29 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= 2A_x$$

Therefore,

$$\frac{d}{dx} x^T A x = 2A_x$$

$$\boxed{2x^T A}$$