Gradient Descent

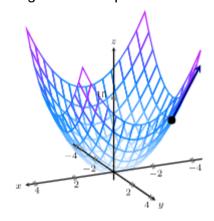
- What is a Gradient in math
- What is Gradient Descent
- Gradient Descent Algorithm
- Why Gradient Descent

What is Gradient in math

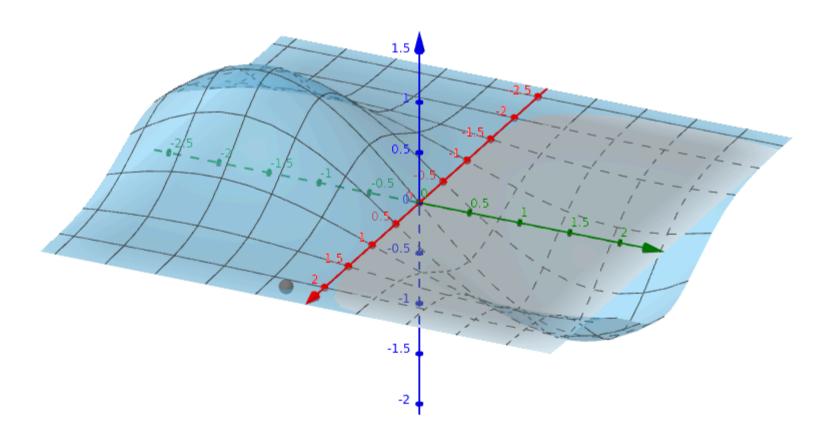
A gradient is a **vector** of partial derivatives for a **multivariate** function $f(x_1, x_2, \dots, x_n)$ with respect for each variable x_i

$$abla f = (rac{\partial f}{\partial x_1}, rac{\partial f}{\partial x_2}, \dots rac{\partial f}{\partial x_n})$$

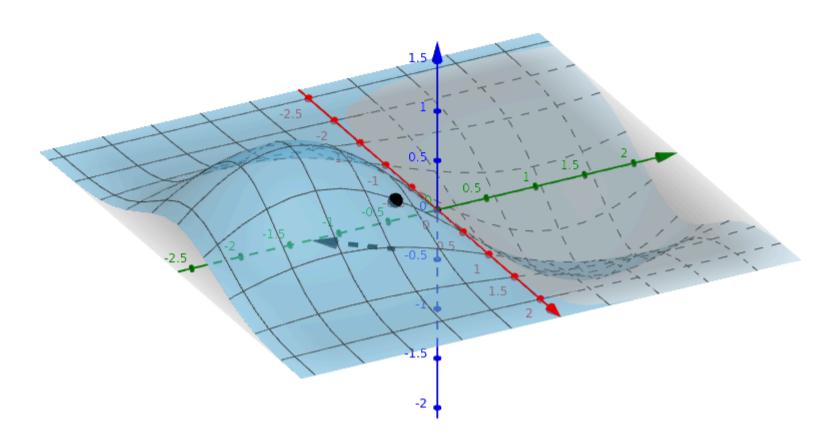
- The Gradient vector points in the direction of the maximum change (increase), formally steepest ascent
- Its magnitude is equal to the **maximum rate** of change



• A zero gradient abla f=0 indicates a **critical point**



• The Gradient measures the "slop" in all directions our point here is on a *flat surface* the gradient vector is zero



• Unlike here where there is a slight slop the gradient vector ∇f points to the direction of the **steepest ascent**

Gradient Rules

1. Product Rule

$$abla(f.\,g) = f
abla g + g
abla f$$

2. Quotient Rule

$$abla \left(rac{f}{g}
ight) = rac{g
abla f - f
abla g}{g^2}$$

3. Gradient of a Norm

$$abla ||x|| = rac{x}{||x||}$$

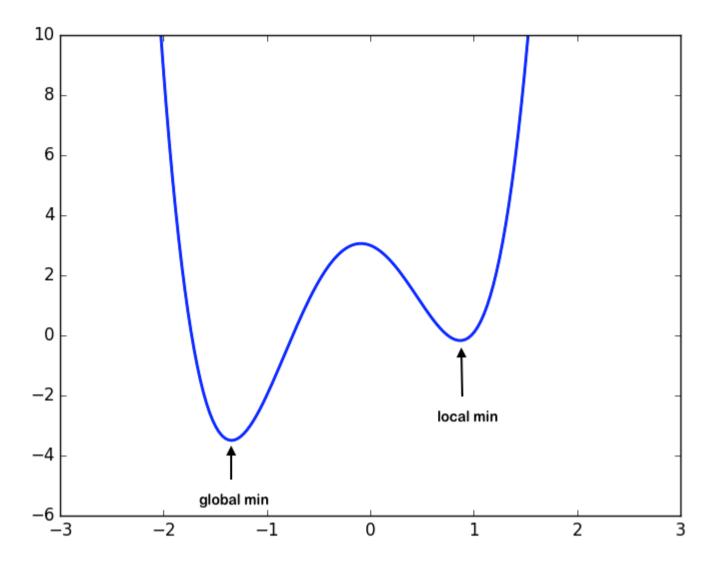
4. Directional Derivative Connection

$$D_{\hat{u}}G=
abla G.\,\hat{u}$$

What is Gradient Descent?

Simply its an **optimization** algorithm used to **train machine learning** models, by optimizing the **parameters** of the model **iteratively** to find the **global minimum** of a loss function curve

More formally the Gradient Descent is an algorithm that essentially computes the gradient ∇ of the cost loss function (MSE in Linear Regression) *h*



• That is the lowest point in a function curve

Intuition:

Imagine a person(gradient descent algorithm) is stuck in a foggy mountain(loss function curve) and he is trying to get down (finding the global minimum). Therefore the person need to use local information and calculations and what's visible to descent down, using the gradient descent which says look at your current position and goes into the direction of the steepest descent

- $\mathbf{Fog} \rightarrow \mathbf{Limited}$, local information
- **Slop** \rightarrow Gradient
- $\bullet \ \ \, \textbf{Step size} \rightarrow \textbf{Learning rate}$

Gradient Descent Algorithm

Generally the Gradient Descent follows These steps :

- 1. Initialize θ (randomly)
- 2. While not converged:
 - 1. Computer gradient : $\nabla J(\theta)$
 - 2. Update parameters : $heta^+ = heta^- lpha
 abla_ heta J(heta)$
 - 3. Check convergence optional
- 3. Return optimized θ

Note:

- $J(\theta)$ is a the **Loss Function**
- α is the **Learning Rate** which is the step size
- Batch size \rightarrow its the trade off between speed and stability