

# Gradient Descent:

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Main task to update the parameter of neural network by minimizing the loss function respect to parameters

## Optimization algorithms:

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the main task of optimization algorithms is to reduce the loss by adjusting weights and biases.

### Batch Gradient Descent:

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$\theta := \theta - \eta \frac{1}{N} \lim_{N \rightarrow \infty} \left( \sum_{i=1}^N \Delta L \right)$

#### Advantages:

stable convergence

#### Disadvantages:

Computationally expensive

### SGD:

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$\theta := \theta - \eta \Delta L$

#### Advantages:

1. Faster convergence
2. Due to randomness can escape local minima.

## Disadvantages:

1. as this is updating single single parameters that's why found high variances is updating parameters nad creating high fluctuating .

## Mini-Batch Gradient Descent:

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$\theta := \theta - \eta (\sum_{i=1}^M \Delta L)$

## Adantage:

it reduces variance of the paramter as it is wokring like a batch.

## Advance Optimization Alogorithms:

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to improve basic gradient descent problem, adding momentum to improve learning rate and accelerating convergence.

## Momentum:

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by adding previous value to current value make a stable and smoothing fluctuating and accelerate the convergence and prevent noise.

first means at initial stages all are 0,

like velocity  $v = 0$

parameter,  $\theta = 0$

update ,

velocity,  $V = \text{lemda } v + \eta \Delta L$  here add lemnda as a momentum

parameters update,  $\theta = \text{old } \theta - V(\text{update velocity})$