

# Shri Ramdeobaba College of Engineering and Management

## Nagpur, 440013

### Department of Computer Science Engineering (AIML)

#### Deep Learning Lab

**Name :** *Shantanu Mane*

**Roll No. :** *E63*

**Batch :** *CSE-AIML*

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**AIM - To implement Vanilla and Momentum based Gradient Descent algorithm to find the best fit curve for the points**

$x = \{1, 3.5, 6\}$

$y = \{4, 6.5, 9\}$

## Importing Dependencies

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

## Gradient Descent Class

```
class GradientDescentFamily:
    w = 0.0
    b = 0.0
    velocity, velocity_b = 0.0, 0.0

    X = np.array([])
    Y = np.array([])

    lr = 0.0
    gamma = 0.0
```

```
N = 0
```

```
self.colors = ["#9b5de5", "#f15bb5", "#fee440", "#00bbf9", "#00f5d4"]
```

```
def __init__(self, w, b, lr, gamma):
```

```
    self.w = w
```

```
    self.b = b
```

```
    self.lr = lr
```

```
    self.gamma = gamma
```

```
    self.X = np.array([1, 3.5, 6])
```

```
    self.Y = np.array([4, 6.5, 9])
```

```
    self.N = self.X.shape[0]
```

```
def Get_Gradient(self):
```

```
    dldw = 0.0
```

```
    dlbd = 0.0
```

```
    # N = self.X.shape[0]
```

```
    dldw += -2 * self.X * (self.Y - (self.w * self.X + self.b))
```

```
    dlbd += -2 * (self.Y - (self.w * self.X + self.b))
```

```
    return dldw, dlbd
```

```
def Vanilla_Gradient_Descent(self):
```

```
    loss = 0.0
```

```
    self.w = 0.0
```

```
    self.b = 0.0
```

```
    for epoch in range(301):
```

```
        dldw, dlbd = self.Get_Gradient()
```

```
        # print(dldw, dlbd, self.w, self.b)
```

```
        self.w -= self.lr * np.sum(dldw) / self.N
```

```
        self.b -= self.lr * np.sum(dlbd) / self.N
```

```
        y_pred = self.w * self.X + self.b
```

```
        loss = np.sum((self.Y - y_pred) ** 2) / self.N
```

```
        axs[0, 0].scatter(epoch, self.w, color=self.colors[0], s=5, alpha=0.7)
```

```
        axs[0, 0].set_title("epochs vs weight (vanilla)", fontsize=24)
```

```
        axs[0, 1].scatter(epoch, self.b, color=self.colors[1], s=5, alpha=0.7)
```

```
        axs[0, 1].set_title("epochs vs bias (vanilla)", fontsize=24)
```

```
        axs[0, 2].plot(epoch, loss, color=self.colors[2], marker='.', alpha=0.7)
```

```
        axs[0, 2].set_title("epochs vs loss (vanilla)", fontsize=24)
```

```
        # plt.plot(epoch, loss, color='g')
```

```
    # plt.show()
```

```
    print(f'w: {self.w}, b : {self.b}, final loss : {loss}')
```

```
def Momentum_Gradient_Descent(self):
```

```
    self.w = 0.0
```

```
    self.b = 0.0
```

```
    loss = 0.0
```

```
    for epoch in range(301):
```

```
        dldw, dlbd = self.Get_Gradient()
```

```

        # print(self.w, self.b)

        self.velocity = self.gamma * self.velocity + self.lr * np.sum(dldw) / self.N
        self.w -= self.velocity

        # self.velocity_b = self.gamma * self.velocity_b + self.lr * np.sum(dldb) / self.N
        # self.b -= self.velocity_b

        # self.w -= self.lr * np.sum(dldw) / self.X.shape[0]
        self.b -= self.lr * np.sum(dldb) / self.N

        y_pred = self.w * self.X + self.b

        loss = np.sum((self.Y - y_pred) ** 2) / self.N

        axs[1, 0].scatter(epoch, self.w, color=self.colors[4], s=5, alpha=0.7)
        axs[1, 0].set_title("epochs vs weight (momentum)", fontsize=24)
        axs[1, 1].scatter(epoch, self.b, color=self.colors[3], s=5, alpha=0.7)
        axs[1, 1].set_title("epochs vs bias (momentum)", fontsize=24)
        axs[1, 2].plot(epoch, loss, color=self.colors[2], marker='.', alpha=0.7)
        axs[1, 2].set_title("epochs vs loss (momentum)", fontsize=24)

    print(f'w: {self.w}, b : {self.b}, final loss : {loss}')

    @staticmethod
    def Show_Plot(self):
        plt.show()

```

---

## Main Function

```

from Gradient_Descent import GradientDescentFamily

if __name__ == '__main__':
    gd: GradientDescentFamily = GradientDescentFamily(0.0, 0.0, 0.05, 1.009)

    gd.Vanilla_Gradient_Descent()
    gd.Momentum_Gradient_Descent()
    gd.Show_Plot()

```

## Output

Vanilla Gradient Descent → w: 1.000354290958917, b : 2.998362765540911, final loss : 6.807  
 Momentum Gradient Descent → w: 1.000229682240711, b : 2.9986983624105608, final loss : 4.6

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## Plot

## Weights, Bias & Loss

