Exercise 1

November 26, 2021

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Machine Learning Lab
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Lab 03

Exercise 1

Importing Packages

```
[1]: import matplotlib.pyplot as plt #Importing Matplotlib import numpy as np #Importing Numpy
```

Defining Rosenbrock Function

```
[2]: #Defining Constant values of Rosenbrock Function
a = 1
b = 100
```

```
[3]: rosenbrock_func = lambda x , y : (a - x)**2 + b * (y - x**2)**2
```

Plotting a 3D Plot of Rosenbrock Function

```
[4]: #Creating x and y vectors for plotting Rosenbrock function within a specific

→range

x = np.linspace(-10,10)

y = np.linspace(-10,10)

X, Y = np.meshgrid(x, y)

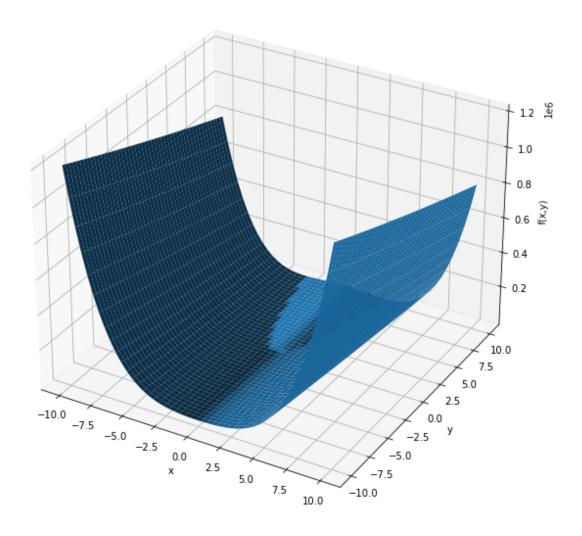
#Applying Rosenbrock function to all values of vector x and y to get values for

→f(x,y)

fxy = np.array(rosenbrock_func(X,Y))
```

```
[5]: #Plotting Rosenbrock Function as a 3D Surface
fig = plt.figure(figsize=(10,10))
ax = fig.add_subplot(111, projection='3d')
ax.plot_surface(X, Y, fxy)
ax.set_xlabel('x')
ax.set_ylabel('y')
ax.set_zlabel('f(x,y)')
ax.set_title('Rosenbrock Function')
plt.show()
```

Rosenbrock Function



Derive the partial gradients. The Partial Derivative of Rosenbrock Function with Respect to \mathbf{x} .

$$\frac{\partial f}{\partial x} = 2(a-x)(-1) + 2b(y-x^2)(-2x)$$
$$\frac{\partial f}{\partial x} = -2(a-x) - 4bx(y-x^2)$$

The Partial Derivative of Rosenbrock Function with Respect to Y:

$$\frac{\partial f}{\partial y} = 2b(y - x^2)(1)$$
$$\frac{\partial f}{\partial y} = 2b(y - x^2)$$

Convert the function and gradient of this function into equivalent code representation.

Partial Derivative of Rosenbrock Function with respect to X

```
[6]: dx = lambda x , y : -2 * (a - x) -4*x*b * (y - x**2)
```

Partial Derivative of Rosenbrock Function with respect to Y

```
[7]: dy = lambda x , y : 2*b * (y - x**2)
```

Optimize the function with Gradient Descent. Set the appropriate hyperparameters like initial value of (x,y) and the steplength—through trial and error.

```
[8]: #Defining Hyperparameters for Gradient Descent

alpha = 0.00001

epsilon = 10**-4

total_iteration = 3000
```

```
[9]: #Initializing x and y vector for storing all intermediate values of x and y so⊔

→ that to use for plotting

x_arr = []
y_arr = []
```

Function to perform Gradient Descent on X and Y

```
[10]: def gradient_descent(x,y,total_iter):
          global x_arr, y_arr
          for i in range(total iter):
               \#Updating the value of x and y based on gradient direction
              x_{-} = x - alpha * dx(x,y)
              y_ = y - alpha * dy(x,y)
               #Checking the stopping Condition
               if rosenbrock_func(x,y) - rosenbrock_func(x_,y_) < epsilon:</pre>
                   x_arr.append(x_)
                   y_arr.append(y_)
                   return (x_,y_)
               \#Swapping the new values of x and y with previous values
              x , y = x_{-} , y_{-}
               \#Appending the new x and y values in x and y arrays for plotting
              x_arr.append(x_)
              y_arr.append(y_)
          #Raising Exception if after total Iterations, our Gradient Descent didn't_{\sf L}
       \hookrightarrow Converge
          raise Exception('Not Converged in {} many Iterations'.format(total_iter))
```

The Minima Point after 3000 many Iteration is: (0.031151356204668527,-0.012073605080919696)

Visualize the trajectory on the same 3D plot. This trajectory should ideally lead to the function minimum, starting off with (x = 10, y = 10) for example.

```
[12]: #Preparing x and y array for plotting
X_arr, Y_arr = np.meshgrid(np.array(x_arr), np.array(y_arr))
fxy_ = np.array(rosenbrock_func(X_arr,Y_arr))
```

Rosenbrock Function with Gradient Descent

