PDL Lab1: Python Functions and Numpy

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In [1]: # PART-I Write a method for sigmoid function
In [2]: #1. Write a function, mysigmoid(x), that takes the real number x and returns the sigmoid value using math.exp
        import math
        def mysigmoid(x):
            return 1 / (1 + math.exp(-x))
In [3]: #2. Call mysigmoid() with x=4 and print the sigmoid value of 4
        x = 4
        sig= mysigmoid(x)
        print("Sigmoid value of 4:", sig)
        Sigmoid value of 4: 0.9820137900379085
In [4]: #3. Now, find the sigmoid values for x=[1, 2, 3]. Observe the results.
        x = [1, 2, 3]
        sig = [mysigmoid(val) for val in x]
        print("Sigmoid values for x=[1, 2, 3]:", sig)
        Sigmoid values for x=[1, 2, 3]: [0.7310585786300049, 0.8807970779778823, 0.9525741268224334]
In [5]: #4. Rewrite mysigmoid() using np.exp() function
        import numpy as np
        def mysigmoid(x):
            return 1 / (1 + np.exp(-x))
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In [6]: #5. Now call your function with x=[1, 2, 3] and observe the results
    x = [1, 2, 3]
    sig = [mysigmoid(val) for val in x]
    print("Sigmoid values for x=[1, 2, 3] ", sig)

Sigmoid values for x=[1, 2, 3] [0.7310585786300049, 0.8807970779778823, 0.9525741268224334]

In [7]: #PART-II: Gradient or derivative of sigmoid function

In [8]: def sig_derivative(s):
        return s * (1 - s)
        x = 5
        sigm=mysigmoid(x)
        gradient = sig_derivative(sigm)
        print("Gradient of sigmoid(4):", gradient)

Gradient of sigmoid(4): 0.006648056670790033

In [9]: #Part-III: Write a method image_to_vector()
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In [10]: def image_to_vector(image):
             length, height, channels = image.shape
             vector = image.reshape((length * height * channels, 1))
             return vector
         image = np.array([
             [[1, 2, 3], [4, 5, 6]],
             [[7, 8, 9], [10, 11, 12]],
             [[13, 14, 15], [16, 17, 18]]])
         vector = image_to_vector(image)
         print("Vector shape:", vector.shape)
         print("Vector:")
         print(vector)
         Vector shape: (18, 1)
         Vector:
         [[ 1]
          [ 2]
          [ 3]
          [ 4]
          [ 5]
          [ 6]
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[7] [8] [9] [10] [11] [12] [13] [14] [15] [16] [17] [18]]

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In [11]: def image to vector(image):
             length, height, channels = image.shape
             vector = image.reshape((length * height * channels, 1))
              return vector
         image = np.array([
              [[255, 0, 0], [0, 255, 0], [0, 0, 255]],
             [[255, 255, 0], [255, 0, 255], [0, 255, 255]],
             [[128, 128, 128], [64, 64, 64], [192, 192, 192]]
         ])
         vector = image_to_vector(image)
         print("Vector shape:", vector.shape)
         print("Vector:")
         print(vector)
           [ V]
           [255]
           [255]
           [255]
           [ 0]
           [255]
           [ 0]
           [255]
           [ 0]
           [255]
           [255]
           [128]
           [128]
           [128]
           [ 64]
           [ 64]
           [ 64]
           [192]
           [192]
           [192]]
In [12]: #Part-IV: Write a method normalizeRows()
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In [13]: def normalizeRows(x):
             norms = np.linalg.norm(x, axis=1, keepdims=True)
             return x / norms
         x = np.array([
             [1, 2, 3],
             [4, 5, 6],
             [7, 8, 9]
         ])
         normalized x = normalizeRows(x)
         print("Normalized matrix:")
         print(normalized x)
         Normalized matrix:
         [[0.26726124 0.53452248 0.80178373]
          [0.45584231 0.56980288 0.68376346]
          [0.50257071 0.57436653 0.64616234]]
In [14]: #Part-V: Multiplication and Vectorization Operations
In [15]: #1)A)
         x1 = np.array([9, 2, 5])
         x2 = np.array([7, 2, 2])
         mul result = np.multiply(x1, x2)
         print("Multiplication:", mul result)
         dot result = np.dot(x1, x2)
         print("Dot product:", dot result)
         Multiplication: [63 4 10]
         Dot product: 77
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In [16]: #B)
         x1 = np.array([9, 2, 5, 0, 0, 7, 5, 0, 0, 0, 9, 2, 5, 0, 0, 4, 5, 7])
         x2 = np.array([7, 2, 2, 9, 0, 9, 2, 5, 0, 0, 9, 2, 5, 0, 0, 8, 5, 3])
         mul fun = np.multiply(x1, x2)
         print("Multiplication:", mul_fun)
         dot fun = np.dot(x1, x2)
         print("Dot product:", dot fun)
         Multiplication: [63 4 10 0 0 63 10 0 0 0 81 4 25 0 0 32 25 21]
         Dot product: 338
In [17]:
         #2)
         import time
         N = 1000000
         x1 = np.random.random(N)
         x2 = np.random.random(N)
         start time = time.time()
         mul result = np.multiply(x1, x2)
         end time = time.time()
         mul time = end time - start time
         start time = time.time()
         dot result = np.dot(x1, x2)
         end time = time.time()
         dot time = end time - start time
         print("Multiplication time:", mul time)
         print("Vectorization (dot product) time:", dot time)
         Multiplication time: 0.0
         Vectorization (dot product) time: 0.20307421684265137
In [18]: #Part-VI: Implement L1 and L2 loss functions
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In [19]: #1)
    def loss_l1(y, ypred):
        return np.sum(np.abs(y - ypred))
    y = np.array([1, 0, 0, 1, 1])
    ypred = np.array([0.9, 0.2, 0.1, 0.4, 0.9])
    l1_loss = loss_l1(y, ypred)
    print("L1 Loss:", l1_loss)

L1 Loss: 1.1

In [20]: #2)
    def loss_l2(y, ypred):
        return np.sum(np.square(y - ypred))
    y = np.array([1, 0, 0, 1, 1])
    ypred = np.array([0.9, 0.2, 0.1, 0.4, 0.9])
    l2_loss = loss_l2(y, ypred)
    print("L2 Loss:", l2_loss)
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L2 Loss: 0.43