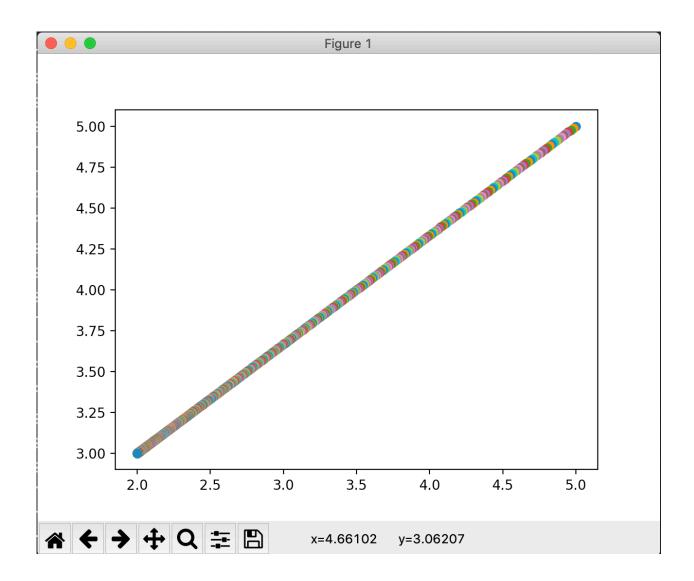
Week #4 Assignment – Satish Ramachandran

Problem 1

```
111
Week4 - Assignment problem #1
111
import math
import matplotlib.pyplot as plt
x = 5
y = 5
learning_rate = 0.01
epsilon = 0.000001
iteration = 0
111
Function is:
f(x,y) = z = -1 * math.sqrt(25 - (x - 2) ** 2 - (y - 3) **2)
def func z(x, y):
  return (-1 * math.sqrt(25 - (x - 2) ** 2 - (y - 3) **2))
111
Partial derivative w.r.t x
def dz_dx(x, y):
  return ((2 * (x - 2)) / (math.sqrt((25 - (x - 2) ** 2 - (y - 3) ** 2))))
111
Partial derivative w.r.t y
def dz dy(x, y):
  return ((2 * (y - 3)) / (math.sqrt((25 - (x - 2) ** 2 - (y - 3) ** 2))))
```

```
while True:
  iteration = iteration + 1
  plt.plot(x, y, 'o')
  new_x = x - learning_rate * dz_dx(x,y)
  new y = y - learning rate * dz dy(x,y)
  if (abs(x - new x) < epsilon) and (abs(y - new y) < epsilon):
    print("Solution reached..")
    x = new x
    y = new_y
    plt.plot(x, y, 'o')
    plt.waitforbuttonpress()
    print('New x and y less than epsilon')
    break
  # More improvements could be made
  x = new x
  y = new_y
print('Solution reached after ' + str(iteration) + ' adjustments')
print('Value of x is: ' + str(x))
print('Value of y is: ' + str(y))
print('Value of z is: ' + str(func z(x, y)))
(base) satishramac-a01:Week4 satishramach$ python Problem1.py
Solution reached..
New x and y less than epsilon
Solution reached after 2310 adjustments
Value of x is: 2.000248177318635
Value of y is: 3.0001654515457563
Value of z is: -4.999999991103381
```



```
Problem 2
Week 4 - Problem 2
Solution using Scikit-learn and hand-coded Gradient descent method
111
import numpy as np
import matplotlib.pyplot as plt
from sklearn import linear model
from sklearn import preprocessing
### Generate the data ###
def generate data(random seed, n samples):
  train_x = np.linspace(0,20,n_samples)
  train y = 3.7 * train x + 14 + 4 * np.random.randn(n samples)
  print("X data")
  print("----")
  print("Size: " + str(np.shape(train x)))
  print(train x)
  print("Y data")
  print("----")
  print("Size: " + str(np.shape(train_y)))
  print(train y)
  return(train x, train y)
### SciKit Learn method
def scikit method(x data, y data):
  print("Using SciKit learn..")
  linear reg = linear model.LinearRegression()
  print("Dimensions: X: " + str(x data.ndim) + ", Y: " + str(y data.ndim))
  # IMPORTANT: LinearRegression expects a 2-D array. So, add a dimension using
  # reshape()
  linear_reg.fit(x_data.reshape(-1,1), y_data.reshape(-1,1))
  print("Slope : " + str(linear_reg.coef_))
  print("Intercept: " + str(linear reg.intercept ))
  return (linear reg.coef , linear reg.intercept )
```

```
### Handcoded Gradient descent method
Partial differentiation w.r.t slope
def dy dslope(slope, intercept, x data, y data):
  return (-2 * sum((y data - slope * x data - intercept) * x data))
111
Partial differentiation w.r.t intercept
def dy dintercept(slope, intercept, x data, y data):
  return (-2 * sum((y data - slope * x data - intercept)))
111
Gradient Descent method
def gradient_descent(x_data, y_data, learn_rate, epochs):
  print("Gradient Descent method..")
  slope = 0
  intercept = 0
  for iteration in range(epochs):
    #print(iteration)
    slope = slope - learn_rate * dy_dslope(slope, intercept, x_data, y_data)
    intercept = intercept - learn rate * dy dintercept(slope, intercept, x data,
y_data)
  return (slope, intercept)
111
Function to descale a min max scaled data
def deScale_y(y_orig, scaled_y):
  result y = []
  min_y = min(y_orig)
  max_y = max(y_orig)
  for each_scaled_y in scaled_y:
```

```
result_y.append(each_scaled_y * (max_y - min_y) + min_y)
  return result y
train x actual, train y actual = generate data(42, 30)
train x = preprocessing.minmax scale(train x actual)
train_y = preprocessing.minmax_scale(train_y_actual)
s_slope, s_intercept = scikit_method(train_x, train_y)
gd slope, gd intercept = gradient descent(train x, train y, 0.001, 4000)
print("RESULTS" + "\n" + "-----")
print("SciKit Learn : slope: " + str(s_slope) + " intercept: " + str(s_intercept))
sci slope = s slope[0][0]
sci intercept = s intercept[0]
result_scaled_y = train_x * sci_slope + sci_intercept
sci_descaled_computed_y = deScale_y(train_y_actual, result_scaled_y)
print("Gradient Descent : slope: " + str(gd_slope) + " intercept: " +
str(gd intercept))
result_scaled_y = train_x * gd_slope + gd_intercept
gd_descaled_computed_y = deScale_y(train_y_actual, result_scaled_y)
# Plot the results
plt.plot(train x actual, train y actual, 'o')
plt.plot(train_x_actual, sci_descaled_computed_y, '-o')
plt.plot(train_x_actual, gd_descaled_computed_y, '-o')
plt.show()
plt.waitforbuttonpress()
```

```
(base) satishramac-a01:Week4 satishramach$ python Problem2.py
X data
_____
Size: (30,)
[ 0.
             0.68965517 1.37931034 2.06896552 2.75862069 3.44827586
 4.13793103 4.82758621 5.51724138 6.20689655 6.89655172 7.5862069
  8.27586207 8.96551724 9.65517241 10.34482759 11.03448276 11.72413793
 12.4137931 13.10344828 13.79310345 14.48275862 15.17241379 15.86206897
 16.55172414 17.24137931 17.93103448 18.62068966 19.31034483 20.
                                                                       ٦
Y data
Size: (30,)
10.1324881 20.27247899 19.56040352 19.8758327 27.9244679 24.10339172
 29.2153935 33.98646013 31.00014387 30.36080068 39.15754639 46.14557143
 47.74767228 49.16766352 47.77319689 53.89202668 55.98438066 47.71154659
 62.14662761 60.55171094 64.61373962 66.50456287 77.37409401 66.81942334
 74.68099732 82.00284451 76.03000291 87.25920073 82.03237915 81.202397257
Using SciKit learn..
Dimensions: X: 1, Y: 1
Slope: [[0.9426086]]
Intercept: [0.05218696]
Gradient Descent method...
RESULTS
SciKit Learn : slope: [[0.9426086]] intercept: [0.05218696]
Gradient Descent: slope: 0.9426085769850121 intercept: 0.05218697401882757
```

As can be seen, the SciKit derived values and Gradient Descent values are almost the same.

This is the plot of the deScaled values of Y based on the computed slope and intercept. Since both the values are extremely close, the plot line seems like just one.

