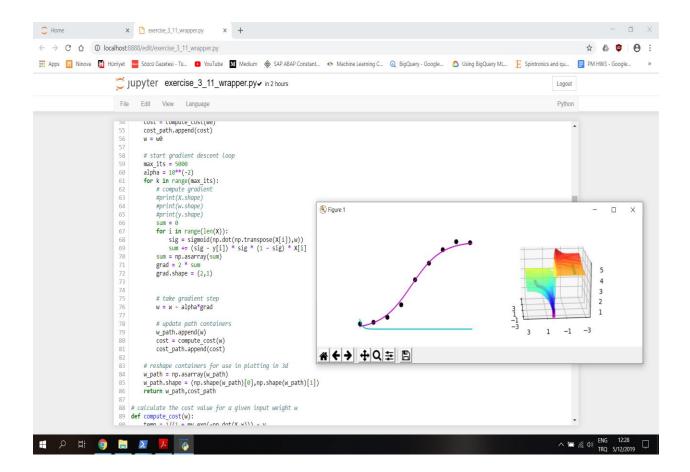
## **Learning from Data**

## **Assignment 2**

## Question 1:

The Added/Changed Code Statement	Explanation
def sigmoid(x): return 1 / (1 + math.exp(-x))	This is a function that returns the sigmoidal value of a parameter passed to it.
<pre>sum = 0 for i in range(len(X)):     sig = sigmoid(np.dot(np.transpose(X[i]),w))     sum += (sig - y[i]) * sig * (1 - sig) * X[i] sum = np.asarray(sum) grad = 2 * sum grad.shape = (2,1)</pre>	This is where the gradient is calculated by using the logistic regression gradient descent calculation formula. For each subarrays of X and y, calculate gradient, make the result a numpy array, reshape gradient and use it for the iterations.



## Question 2:

The Added/Changed Code Statement	Explanation
def sigmoid(x): return 1 / (1 + math.exp(-x))	This is a function that returns the sigmoidal value of a parameter passed to it.
sum = 0 for i in range(len(X)):     sig = sigmoid(np.dot(np.transpose(X[i]),w))     sum += (sig - y[i]) * sig * (1 - sig) * X[i]  ar = [] ar.append(0) ar.append(w[1][0]) ar = np.asarray(ar) ar = np.reshape(ar, (2,1)) sum = np.reshape(sum, (2,1)) # regularize it sum += 2 * lam * ar	This is where the gradient is calculated by using the logistic regression gradient descent calculation formula. The difference between the above one is the regularization term.

