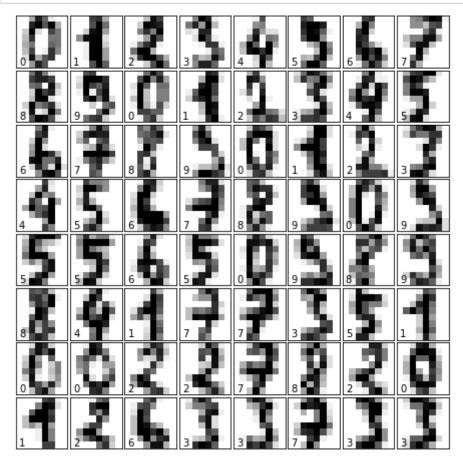
SETTING UP DEPENDENCIES NEEDED

```
In [2]: #Import data and dependencies
   import numpy as np
   import pandas as pd
   import matplotlib.pyplot as plt

from sklearn.datasets import load_digits
   from sklearn.model_selection import train_test_split
   from sklearn import preprocessing
```

CREATING A VISUALIZATION OF THE DATASET

```
In [3]: #taken from https://www.datacamp.com/community/tutorials/machine-learnin
        g-python
        digits = load_digits()
        # Import matplotlib
        import matplotlib.pyplot as plt
        # Figure size (width, height) in inches
        fig = plt.figure(figsize=(6, 6))
        # Adjust the subplots
        fig.subplots_adjust(left=0, right=1, bottom=0, top=1, hspace=0.05, wspac
        e = 0.05)
        # For each of the 64 images
        for i in range(64):
            # Initialize the subplots: add a subplot in the grid of 8 by 8, at t
        he i+1-th position
            ax = fig.add_subplot(8, 8, i + 1, xticks=[], yticks=[])
            # Display an image at the i-th position
            ax.imshow(digits.images[i], cmap=plt.cm.binary, interpolation='neare
        st')
            # label the image with the target value
            ax.text(0, 7, str(digits.target[i]))
        # Show the plot
        plt.show()
```



CREATING PRE-PROCESSING DIGITS DATA

```
In [6]: X = digits['data']
        Χ
Out[6]: array([[ 0.,
                     0., 5., ..., 0.,
                                        0.,
                                            0.1,
              [ 0., 0., 0., ..., 10., 0., 0.],
              [ 0.,
                     0.,
                         0., ..., 16., 9.,
                                            0.],
              . . . ,
              [ 0., 0., 1., ..., 6.,
                                        0., 0.1,
              [ 0., 0., 2., ..., 12., 0., 0.],
              [0., 0., 10., ..., 12., 1., 0.]])
In [7]: X = np.c [np.ones(digits['data'].shape[0]),X]
```

SPLITTING DATA SET INTO TRAINING AND TESTING DATA SETS

```
In [633]: #Splitting Data Set into Training and Testing Data Sets
X_train, X_test, y_train, y_test = train_test_split(X, digits['target'],
    test_size = .80, random_state = 1)
```

CREATING THE SIGMOID FUNCTION, COST, AND GRADIENT DESCENT FUNCTIONS

```
In [10]: #creating the sigmoid function for the prediction hypothesis

def sigmoid(z):
    return 1 / (1 + np.exp(-z))
```

```
In [114]: #Calculating the regularized cost function

def cost_function(X_train, y_train, thetas_array, n, reg):
    prediction = sigmoid(np.dot(X_train,np.transpose(thetas_array)))
    prediction[prediction == 1] = 0.999
    cost = -(np.sum((y_train * np.log(prediction)) + ((1 - y_train) * np.log(1-prediction)))/n) + ((reg/(2*n))*np.sum(thetas_array[1:]**2))
    return cost
```

CREATING THE ONE-VS-ALL FUNCTION FOR

```
In [607]: def training(X train, y train, alpha, iters, reg):
              n = len(X train)
              thetas_array = np.zeros(65)
              thetas_j = 0
              theta_0 = 0
              cost = []
              #final thetas = []
              for i in range(iters):
                  theta_0 = thetas_array[0] - ((alpha/n)*(np.sum(np.dot(X_train,th
          etas_array) - y_train) * 1))
                  thetas j = gradient descent(X train, y train, alpha, thetas arra
          y,n, reg)
                  thetas_array = np.insert(thetas_j, 0, theta_0)
                  cost.append(cost function(X train, y train, thetas array, n, req
          ))
              #final thetas.append(thetas array)
              #Plot cost function error per iteration
              """x = np.arange(0, len(cost), step=1)
              plt.plot(x, cost, "-b", label="Cost Function Curve")
              plt.title("Learning Curve")
              plt.xlabel("Number Of Iterations")
              plt.ylabel("Cost Function Value")
              plt.legend()
              plt.show()"""
              return thetas array, cost
```

```
In [624]: def one_vs_all(X_train, y_train, alpha, iters, reg):
    cost = []
    thetas_i = []
    for i in range(len(digits['target_names'])):
        thetas_i.append(training(X_train, np.where(y_train==i,1,0), alph
    a, iters, reg)[0])
        cost.append(training(X_train, np.where(y_train==i,1,0), alpha, i
    ters, reg)[1])
    return np.array(thetas_i), np.array(cost)
```

In [625]: one_vs_all(X_train, y_train, .001, 1000, .08)

```
Out[625]: (array([[ 4.73722309e+00, 0.00000000e+00, -5.03012002e-03,
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                   -9.99696592e-05, -2.38004912e-02, -9.07233286e-03,
                   -2.65147094e-02, 1.16204230e-02, 3.98905142e-02,
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         1.41021763e-01, 2.93056563e-02, 0.00000000e+00,
         0.000000000e+00, -7.90698608e-02, -8.45924803e-02,
          3.15301727e-02, -1.10086972e-01, -1.08496963e-02,
         2.44155282e-02, 0.00000000e+00, 0.00000000e+00,
         -2.69255207e-02, -1.68336073e-01, -1.33552549e-01,
        -1.28204324e-01, -3.31004469e-02, 9.95908327e-04,
        -6.15246856e-04, 0.00000000e+00, -5.84926944e-03,
         -2.38770588e-02, -8.67004826e-02, -9.99042582e-02,
        -7.30864267e-02, 1.76451747e-02, -3.12369353e-03,
         0.00000000e+00, -7.86374229e-03, -4.81988679e-02,
        -6.01906578e-02, -1.12998313e-02, -1.92764864e-03,
        -2.53774864e-02, -5.00346649e-03]]),
array([[0.43484911, 0.38524676, 0.36393725, ..., 0.02068795, 0.0206784
8,
        0.020669021,
        [0.39396703, 0.3406495, 0.32121488, ..., 0.08884629, 0.0888460
1,
        0.088845781,
        [0.39419662, 0.33923507, 0.31791476, ..., 0.027304, 0.0272837]
7,
```

```
0.02726357],
...,
[0.3995314 , 0.34496806, 0.32324576, ..., 0.01944164, 0.0194258

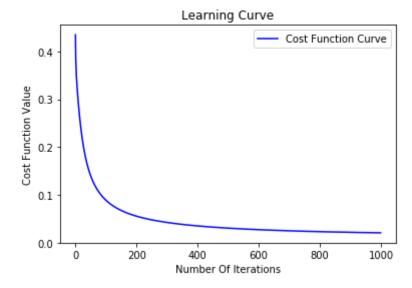
1,
0.01941 ],
[0.43774161, 0.39519345, 0.3813745 , ..., 0.12302737, 0.1229978

2,
0.12296833],
[0.36665805, 0.30788297, 0.28634522, ..., 0.06140076, 0.0613822

1,
0.06136372]]))
```

CHECKING IF GRADIENT DESCENT WORKS AS INTENDED

```
In [626]: cost = one_vs_all(X_train, y_train, .001, 1000, .08)[1]
    x = np.arange(0, len(cost[0]), step=1)
    plt.plot(x, cost[0], "-b", label="Cost Function Curve")
    plt.title("Learning Curve")
    plt.xlabel("Number Of Iterations")
    plt.ylabel("Cost Function Value")
    plt.legend()
    plt.show()
```



CREATING ARRAY OF PREDICTED VALUES

```
In [628]: def predict_one_vs_all(X_train, all_thetas):
    prediction = sigmoid(np.dot(X_train,np.transpose(all_thetas)))
    pred_argmax = np.argmax(prediction,axis = 1)
    return pred_argmax
```

```
predict one vs all(X train, all thetas)
Out[629]: array([4, 2, 4, 6, 0, 7, 8, 1, 3, 3, 8, 9, 4, 9, 7, 1, 0, 0, 6, 7, 4,
          9,
                 7, 6, 7, 5, 8, 8, 7, 0, 0, 8, 8, 0, 7, 2, 8, 3, 5, 0, 6, 9, 2,
          2,
                 2, 9, 5, 7, 7, 6, 9, 8, 8, 3, 5, 7, 7, 2, 6, 0, 8, 7, 0, 9, 7,
          6,
                 7, 4, 3, 0, 5, 0, 0, 6, 5, 8, 0, 1, 4, 1, 2, 4, 7, 5, 0, 7, 0,
          4,
                 0, 0, 9, 7, 7, 4, 6, 6, 5, 8, 0, 1, 5, 7, 9, 0, 7, 7, 0, 3, 8,
          7,
                 2, 0, 4, 4, 0, 1, 0, 2, 5, 5, 8, 7, 5, 3, 6, 7, 3, 4, 8, 0, 6,
          8,
                 8, 7, 2, 0, 9, 3, 7, 8, 6, 3, 7, 9, 8, 0, 3, 7, 6, 8, 0, 0, 8,
          5,
                 5, 1, 8, 2, 7, 3, 1, 5, 1, 8, 1, 6, 3, 9, 4, 4, 7, 5, 6, 2, 5,
          0,
                 7, 5, 6, 0, 9, 2, 3, 9, 0, 7, 8, 1, 0, 1, 4, 1, 2, 8, 0, 8, 0,
          9,
                 0, 1, 5, 5, 3, 6, 2, 4, 0, 1, 0, 0, 3, 3, 2, 5, 3, 8, 1, 3, 7,
          0,
                 1, 2, 2, 9, 3, 5, 3, 8, 1, 5, 8, 8, 6, 9, 3, 6, 6, 9, 0, 6, 2,
          7,
                 0, 9, 6, 8, 5, 3, 0, 2, 5, 5, 6, 6, 0, 8, 3, 9, 2, 1, 1, 2, 2,
          9,
                 2, 1, 7, 5, 2, 5, 0, 2, 8, 5, 1, 7, 6, 8, 3, 7, 1, 2, 9, 9, 6,
          5,
                 6, 7, 5, 2, 6, 1, 9, 9, 1, 9, 8, 4, 6, 3, 8, 6, 1, 3, 4, 4, 6,
          1,
                 5, 3, 4, 0, 6, 1, 0, 3, 8, 2, 3, 8, 3, 2, 2, 1, 3, 4, 7, 8, 4,
          6,
                 4, 5, 2, 2, 1, 0, 2, 6, 9, 0, 4, 3, 7, 8, 8, 8, 3, 4, 1, 4, 6,
          5,
                 2, 9, 1, 4, 9, 1, 51)
```

CALCULATING THE TRAINING ACCURACY

```
In [632]: training_accuracy(X_train, y_train, all_thetas)
```

Training Accuracy: 96.94%

CALCULATIN THE TESTING ACCURACY

```
In [636]: test_accuracy(X_test, y_test, all_thetas)
```

Training Accuracy: 92.84%

```
In [ ]:
```