```
In [230]:
          import matplotlib.pyplot as plt
          import numpy as np
          import pandas as pd
          from scipy import stats, io
          from scipy.special import expit
In [231]: class LogisticRegressionModel(object):
              def init (self, name, seed = 42):
                  if name != 'data':
                      raise ValueError('Incorrect Dataset')
                  np.random.seed(seed)
                  self.name = name
                  self.data = None
                  self.trainOrig = None
                  self.trLabelsOrig = None
                  self.train = None
                  self.trLabels = None
                  self.val = None
                  self.valLabels = None
                  self.test = None
                  self.dim = None
                  self.weight = None
                  self.pred = None
                  self.load data()
                  self.split(1000)
              def load data(self):
                  self.data = io.loadmat(self.name + '.mat')
                  print('Loaded: ' + self.name)
                  self.normalize()
                  self.trainOrig = self.data['X']
                  self.trLabelsOrig = self.data['y'].reshape(-1)
                  self.test = self.data['X test']
                  self.test = np.apply along axis(self.add1, 1, self.test)
                  print('Training size (Before Split): ' + str(len(self.trainOri
```

```
def split(self,valSize):
   totalLen = len(self.trainOrig)
   trainSize = totalLen - valSize
```

print('Test size: ' + str(len(self.test)))

print('Training labels (Before Split):' + str(len(self.trLabel

q)))

sOriq)))

```
randIdx = np.random.permutation(totalLen)
        self.train = self.trainOrig[randIdx][:trainSize]
        self.train = np.apply along axis(self.add1, 1, self.train)
        self.trLabels = self.trLabelsOrig[randIdx][:trainSize]
        self.val = self.trainOrig[randIdx][trainSize:]
        self.val = np.apply along axis(self.add1, 1, self.val)
        self.valLabels = self.trLabelsOriq[randIdx][trainSize:]
        self.dim = self.train.shape
        print('Training Data Len: ' + str(len(self.train)))
        print('Training Labels Len: ' + str(len(self.trLabels)))
        print('Validation Data Len: ' + str(len(self.val)))
        print('Validation Labels Len: ' + str(len(self.valLabels)))
    def normalize(self):
        for idx in ['X', 'X test']:
            mean = np.mean(self.data[idx], axis = 0)
            std = np.std(self.data[idx], axis = 0)
            z score = lambda val: (val-mean)/std
            self.data[idx] = np.apply along axis(z score,1,self.data[i
dx])
    def add1(self, arr):
        return np.append(arr, [[1]])
    def cost(self, X, y, w, l):
        s = expit(X @ w)
        s = np.maximum(s, 1e-7)
        s = np.minimum(s, 1-1e-7)
        return 1*w@w - (1/len(X))*(y.dot(np.log(s)) + (1-y).dot(np.log
(1-s))
    def grad penalty(self, X, y, w, l):
        s = expit(X @ w)
        return -X.T.dot(y - s) + 2*1*w
    def gradient descent(self,typeGrad,e,l,changingE,tol=1e-7,graph=Tr
ue, numIter=1000, maxLoops = 5):
        costLst, valCost = [], []
        converged = False
        w = np.zeros(self.dim[1])
        J o, J new = float("Inf"), self.cost(self.train, self.trLabels
, w, 1)
        i = 0
        if typeGrad == 'batch':
            while i < numIter and not converged:</pre>
                w = w - e * self.grad penalty(self.train, self.trLabel
s, w, 1)
                J o, J new = J new, self.cost(self.train, self.trLabel
s, w, 1)
```

```
costLst.append(J o)
                valCost.append(self.cost(self.val, self.valLabels, w,
1))
                i+=1
                if abs(J o - J new) < tol:</pre>
                    converged = True
                    break
        if typeGrad == 'stochastic':
            numLoops = 0
            while numLoops < maxLoops and not converged:</pre>
                numLoops += 1
                randIdx = np.random.permutation(self.dim[0])
                X = self.train[randIdx]
                y = self.trLabels[randIdx]
                for x i, y i in zip(X, y):
                     if changingE:
                        w = w - e/(.03*(i+10)) * self.grad penalty(x i)
, y i, w, 1)
                    else:
                        w = w - e*self.grad penalty(x i, y i, w, l)
                    J \circ, J new = J new, self.cost(X, y, w, 1)
                    costLst.append(J o)
                    valCost.append(self.cost(self.val, self.valLabels,
w, 1))
                    i+=1
                     if abs(J o - J new) < tol:</pre>
                         converged = True
                         break
        print("Cost Final: " + str(J o))
        print("Converged:" + str(i) + "iterations")
        if graph:
            plt.plot(costLst, label = "Training")
            plt.plot(valCost, label = "Validation")
            plt.xlabel("Number of Iterations")
            plt.ylabel("Cost")
            plt.title("Cost Over Iterations")
        self.weight = w
    def experiment(self,typeGrad,testData,e,l,changingE):
        self.gradient descent(typeGrad,e,l,changingE,graph=False)
        y val = expit(testData @ self.weight)
        y_pred = (y_val > 0.5).astype(np.int)
        self.pred = y pred
        return y pred
    def accuracy(self, labels):
        return np.sum(self.pred.reshape(-1) == labels.reshape(-1))/len
(self.pred)
```

## In [238]: m = LogisticRegressionModel('data', 42) m.gradient\_descent('batch', .01, 1e-6, changingE = False)

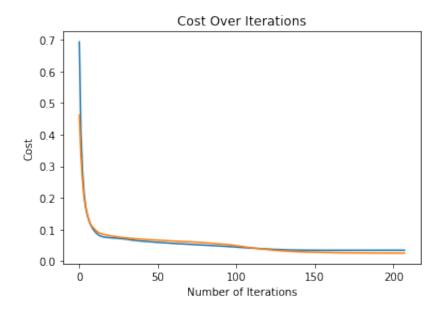
Loaded: data

Training size (Before Split): 6000 Training labels (Before Split):6000

Test size: 497

Training Data Len: 5000
Training Labels Len: 5000
Validation Data Len: 1000
Validation Labels Len: 1000
Cost Final: 0.03451128789693964

Converged: 208 iterations



```
In [234]: lrs = [10**(-7), 10**(-6), 10**(-5), 10**(-4), 10**(-3), 10**(-2)]
    regs = [.01,.001,.0001,.00001,.0000001]
    for lr in lrs:
        for reg in regs:
            mod = LogisticRegressionModel('data', 42)
            mod.experiment('batch', mod.val, lr, reg, changingE = False)
            print("LR: "+str(lr)+' RegParam: '+str(reg)+ ' Accuracy: '+
            str(mod.accuracy(mod.valLabels)))
```

Loaded: data

Training size (Before Split): 6000 Training labels (Before Split):6000

Test size: 497

Training Data Len: 5000 Training Labels Len: 5000

Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5129414392036031 Converged: 1000 iterations LR: 1e-07 RegParam: 0.01 Accuracy: 0.962 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5121343753396624 Converged: 1000 iterations LR: 1e-07 RegParam: 0.001 Accuracy: 0.962 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5120536688016132 Converged: 1000 iterations LR: 1e-07 RegParam: 0.0001 Accuracy: 0.962 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5120455981462916 Converged: 1000 iterations LR: 1e-07 RegParam: 1e-05 Accuracy: 0.962 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5120447910807444 Converged: 1000 iterations LR: 1e-07 RegParam: 1e-06 Accuracy: 0.962

Loaded: data

Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.5120447103741896 Converged: 1000 iterations LR: 1e-07 RegParam: 1e-07 Accuracy: 0.962 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.2038027086929647 Converged: 1000 iterations LR: 1e-06 RegParam: 0.01 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.18553734429978758 Converged: 1000 iterations LR: 1e-06 RegParam: 0.001 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.18371077663228988 Converged: 1000 iterations LR: 1e-06 RegParam: 0.0001 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000

Validation Labels Len: 1000 Cost Final: 0.18352811955325443 Converged: 1000 iterations LR: 1e-06 RegParam: 1e-05 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.1835098538422282 Converged: 1000 iterations LR: 1e-06 RegParam: 1e-06 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.1835080272710942 Converged: 1000 iterations LR: 1e-06 RegParam: 1e-07 Accuracy: 0.977 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.17324801893735775 Converged: 1000 iterations LR: 1e-05 RegParam: 0.01 Accuracy: 0.991 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.07048163099075341 Converged: 1000 iterations LR: 1e-05 RegParam: 0.001 Accuracy: 0.991 Loaded: data

Training size (Before Split): 6000

Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.06020321401375592 Converged: 1000 iterations LR: 1e-05 RegParam: 0.0001 Accuracy: 0.991 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.059175354531653845 Converged: 1000 iterations LR: 1e-05 RegParam: 1e-05 Accuracy: 0.991 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.05907256840559691 Converged: 1000 iterations LR: 1e-05 RegParam: 1e-06 Accuracy: 0.991 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.05906228979121318 Converged: 1000 iterations LR: 1e-05 RegParam: 1e-07 Accuracy: 0.991 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000

Cost Final: 0.357604435955906 Converged: 1000 iterations LR: 0.0001 RegParam: 0.01 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.06444357566896411 Converged: 268 iterations LR: 0.0001 RegParam: 0.001 Accuracy: 0.992 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.041655862093800636 Converged: 1000 iterations LR: 0.0001 RegParam: 0.0001 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.03877904273626833 Converged: 1000 iterations LR: 0.0001 RegParam: 1e-05 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.038491356274646026 Converged: 1000 iterations LR: 0.0001 RegParam: 1e-06 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000

Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.0384625875832243 Converged: 1000 iterations LR: 0.0001 RegParam: 1e-07 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.6568052834526016 Converged: 1000 iterations LR: 0.001 RegParam: 0.01 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.0973664924757946 Converged: 1000 iterations LR: 0.001 RegParam: 0.001 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.04034624649051768 Converged: 448iterations LR: 0.001 RegParam: 0.0001 Accuracy: 0.995 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000

Training Data Len: 5000
Training Labels Len: 5000
Validation Data Len: 1000
Validation Labels Len: 1000
Cost Final: 0.035192866625506684

Converged: 1000 iterations LR: 0.001 RegParam: 1e-05 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.03462706691031586 Converged: 1000 iterations LR: 0.001 RegParam: 1e-06 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.03457048635840218 Converged: 1000 iterations LR: 0.001 RegParam: 1e-07 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.8000429683031589 Converged: 963iterations LR: 0.01 RegParam: 0.01 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.11250010130894003 Converged: 809 iterations LR: 0.01 RegParam: 0.001 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497

Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.04219741338341202 Converged: 625iterations LR: 0.01 RegParam: 0.0001 Accuracy: 0.997 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000 Validation Data Len: 1000 Validation Labels Len: 1000 Cost Final: 0.03534560097392368 Converged: 172 iterations LR: 0.01 RegParam: 1e-05 Accuracy: 0.996 Loaded: data Training size (Before Split): 6000 Training labels (Before Split):6000 Test size: 497 Training Data Len: 5000 Training Labels Len: 5000

Training Data Len: 5000
Training Labels Len: 5000
Validation Data Len: 1000
Validation Labels Len: 1000
Cost Final: 0.03451128789693964

Converged:208iterations

LR: 0.01 RegParam: 1e-06 Accuracy: 0.997

Loaded: data

Training size (Before Split): 6000 Training labels (Before Split):6000

Test size: 497

Training Data Len: 5000
Training Labels Len: 5000
Validation Data Len: 1000
Validation Labels Len: 1000
Cost Final: 0.03442628911709663

Converged: 202 iterations

LR: 0.01 RegParam: 1e-07 Accuracy: 0.997

```
In [240]: #Cell for submission
    m = LogisticRegressionModel('data', 42)
    y_pred = m.experiment('batch', m.test, .01, 1e-6, changingE = False).r
    eshape(-1)

def results_to_csv(y_test):
        y_test = y_test.astype(int)
        df = pd.DataFrame({'Category': y_test})
        df.index += 1  # Ensures that the index starts at 1.
        df.to_csv('submission.csv', index_label='Id')

results_to_csv(y_pred)

Loaded: data
```

Training size (Before Split): 6000
Training labels (Before Split):6000
Test size: 497
Training Data Len: 5000
Training Labels Len: 5000
Validation Data Len: 1000
Validation Labels Len: 1000
Cost Final: 0.03451128789693964
Converged:208iterations

In [ ]: