# **EE5907 Programming Assignment Q3**

Lee Jianwei A0018867

```
In [1]: from __future__ import division
    import scipy.io
    from scipy.stats import norm
    import numpy as np
    import matplotlib.pyplot as plt

from multiprocessing import Pool
    np.set_printoptions(precision=2, suppress=True)
```

### **Load Data**

```
In [2]: d = scipy.io.loadmat('spamData.mat')
         ytest = d['ytest'].flatten()
ytrain = d['ytrain'].flatten()
         xtest = d['Xtest']
         xtrain = d['Xtrain']
In [3]: def binarize(array):
             """array: list of emails, each 57 features long"""
             return np.array([x>0 for x in array]).astype('uint8')
         xtrainBin = binarize(xtrain)
         xtestBin = binarize(xtest)
In [4]: #z-normalise features
         def znorm1D(array1D):
             m = np.mean(array1D)
             s = np.std(array1D)
             return np.array((array1D-m)/s)
         def znorm2D(array2D):
              """znorm along columns of 2D array"""
             znormed columns = np.array([znorm1D(c) for c in array2D.T]) # each column in array2D
         is a row (called c) in array2D.T
            return znormed_columns.T
         xtrainZ = znorm2D(xtrain)
         xtestZ = znorm2D(xtest)
In [3]: #log-transform features
         def log2D(array):
             """array: list of emails, each 57 features long"""
return np.array([np.log(x+0.1) for x in array]) #x is a 57 element array
         xtrainLog = log2D(xtrain)
         xtestLog = log2D(xtest)
In [4]: classMLE = np.average(ytrain)
In [5]: def sigmoid(x):
             return 1/(1 + np.exp(-x))
```

```
In [6]: def reg_params(x_array2D, y_array1D, w, lamb):
    """returns NLL, g, H"""
                      N_features = x_array2D.shape[1] # number of features per mail: default is 57
                      x_array2D = np.insert(x_array2D,0,1,axis=1) # modify feature list to include a 1 at t
               he start
                      g = np.dot(x_array2D.T,(mu-y_array1D))
                      S = np.diag(mu*(1-mu))
                      H = np.dot(x array2D.T, np.dot(S, x array2D))
                      NLL reg = NLL + 0.5*lamb*np.inner(w[1:],w[1:])
                      g_{eg} = g + np.insert(lamb*w[1:],0,0) # first element of g is unregularised
                      H_reg = H + np.diag(np.insert(lamb*np.ones(N_features),0,0)) # first diagonal element
               is unregularised
                         print x array2D
               #
               #
                         print mu
                      return NLL reg, g reg, H reg
In [9]: | NLL_reg, g_reg, H_reg = reg_params(xtrainZ,ytrain,np.ones(58),1)
               /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:7: RuntimeWarning: divide by
               zero encountered in log
               /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:7: RuntimeWarning: invalid va
               lue encountered in multiply
In [7]: def omega(x_array2D, y_array1D, lamb):
                      this function implements Newton's method for logistic regression to find w which desc
                ribes the dataset x array2D, yarray1D
                      """initialisation"""
                      N features = x array2D.shape[1]
                      wInit = np.zeros(N_features+1)
                      w = wInit
                      converge = False
                      counter = 0
                      """iterator to find optimal w"""
                      try:
                             while converge==False:
                                      _, g, H = reg_params(x_array2D, y_array1D, w, lamb)
                                     counter += 1
                                    w_new = w - np.dot(np.linalg.inv(np.dot(H.T,H)),np.dot(H.T,g)) # numerically
               more stable t\overline{h}an inv(H).g
                                       w_new = w - np.dot(np.linalg.inv(H),g)
                                     error = np.inner(w_new-w,w_new-w)/np.inner(w,w)
                                    if error < 0.001:
                                             """arbitrary error threshold set to 0.1%"""
                                           converge = True
                                    else:
                                           w = w_new
                             return w new
                             print ('lambda = {\n =
               rmat(lamb, counter, error*100, g, H))
In [8]: | def LogReg_ErrorRate(x_array2D, y_array1D, x_training_array2D, y_training_array1D, lamb):
                      uses omega to classify an array of emails x_array2D
                      compares results with y_array1D to check for error rate
                      w = omega(x_training_array2D,y_training_array1D,lamb)
                      category = sigmoid(w[0]+np.dot(x_array2D, w[1:]))>0.5
                      category_error = np.logical_xor(category,y_array1D)
                      error_rate = np.sum(category_error)/len(y_array1D)
                      return error rate
In [9]: lambdas = np.insert(np.arange(15,105,5),0,np.arange(1,11,1))
               print lambdas
                                                       6
                                                              7
                                                                      8
                                                                            9 10 15 20 25 30 35 40 45 50
                  55 60 65 70 75 80 85 90 95 100]
```

#### **Binarised Data**

```
In [27]: TrainingBin ErrorRates = map(lambda lamb: LogReg ErrorRate(xtrainBin,ytrain,xtrainBin,ytr
          ain, lamb), lambdas)
          /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:17: RuntimeWarning: divide by
         zero encountered in double scalars
In [34]: # np.savetxt('LogisticRegression TrainingBinErrorRates',zip(lambdas,TrainingBin ErrorRate
          s),header='lambdas\tTrainingBin ErrorRate')
In [16]: # lambdas,TrainingBin ErrorRates = np.loadtxt('LogisticRegression TrainingBinErrorRates')
In [20]:
         plt.figure()
          plt.title('error rates for binarised training data')
          plt.plot(lambdas, np.array(TrainingBin_ErrorRates)*100)
         plt.xlabel('lambda')
          plt.ylabel('error rate (%)')
         plt.show()
                     error rates for binarised training data
            9.5
            9.0
            85
          error rate (%)
            8.0
            7.5
            7.0
            6.5
            6.0
                                lambda
In [38]: # Error rates for training data for lambda = 1,10,100
         map(lambda lamb: LogReg ErrorRate(xtrainBin,ytrain,xtrainBin,ytrain,lamb), [1,10,100])
         /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
         zero encountered in double_scalars
Out[38]: [0.064274061990212072, 0.066884176182707991, 0.093964110929853184]
In [38]: TestingBin ErrorRates = map(lambda lamb: LogReg ErrorRate(xtestBin,ytest,xtrainBin,ytrain
          ,lamb), lambdas)
          /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
         zero encountered in double_scalars
In [39]: # np.savetxt('LogisticRegression TestingBinErrorRates',zip(lambdas,TestingBin ErrorRates)
          ,header='lambdas\tTestingBin_ErrorRate')
In [22]: # lambdas,TestingBin_ErrorRates = np.loadtxt('LogisticRegression_TestingBinErrorRates').T
In [23]:
         plt.figure()
          plt.title('error rates for binarised testing data')
          plt.plot(lambdas, np.array(TestingBin_ErrorRates)*100)
          plt.xlabel('lambda')
          plt.ylabel('error rate (%)')
         plt.show()
                      error rates for binarised testing data
            10.0
             9.5
             9.0
          8
          rate
             8.5
          error
             8.0
             75
             7.0
                                               80
```

```
In [39]: # Error rates for testing data for lambda = 1.10.100
         map(lambda lamb: LogReg_ErrorRate(xtestBin,ytest,xtrainBin,ytrain,lamb), [1,10,100])
         /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:17: RuntimeWarning: divide by
         zero encountered in double scalars
Out[39]: [0.073567708333333329, 0.076822916666666671, 0.09635416666666671]
```

### **Z-normed Data**

```
In [46]: TrainingZ ErrorRates = map(lambda lamb: LogReg ErrorRate(xtrainZ,ytrain,xtrainZ,ytrain,la
                 mb), lambdas)
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
                 zero encountered in double scalars
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:7: RuntimeWarning: divide by
                  zero encountered in log
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:7: RuntimeWarning: invalid va
                  lue encountered in multiply
In [47]: | # np.savetxt('LogisticRegression_TrainingZErrorRates',zip(lambdas,TrainingZ_ErrorRates),h
                  eader='lambdas\tTrainingZ_ErrorRate')
In [24]: | # lambdas,TrainingZ ErrorRates = np.loadtxt('LogisticRegression TrainingZErrorRates').T
In [25]: plt.figure()
                  plt.title('error rates for Z-normed Training data')
                  plt.plot(lambdas, np.array(TrainingZ_ErrorRates)*100)
                  plt.xlabel('lambda')
                  plt.ylabel('error rate (%)')
                 plt.show()
                                       error rates for Z-normed Training data
                      10.0
                        9.5
                   8
                       9.0
                   rate
                   error
                        8.5
                        8.0
                        7.5
                                         20
                                                        40
                                                                      60
                                                                                     80
                                                                                                   100
                                                            lambda
In [40]: # Error rates for training data for lambda = 1,10,100
                 map(lambda lamb: LogReg ErrorRate(xtrainZ,ytrain,xtrainZ,ytrain,lamb), [1,10,100])
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
                 zero encountered in double_scalars
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:7: RuntimeWarning: divide by
                 zero encountered in log
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:7: RuntimeWarning: invalid va
                 lue encountered in multiply
Out[40]: [0.076345840130505715, 0.081566068515497553, 0.096574225122349103]
In [43]: TestingZ_ErrorRates = map(lambda lamb: LogReg_ErrorRate(xtestZ,ytest,xtrainZ,ytrain,lamb)
                  , lambdas)
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
                  zero encountered in double scalars
                  /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:7: RuntimeWarning: divide by
                 zero encountered in log
                  /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:7: RuntimeWarning: invalid va
                  lue encountered in multiply
 In ~~[44]: \\ \# ~np.savetxt('LogisticRegression\_TestingZErrorRates', zip(lambdas, TestingZ\_ErrorRates), health is a superior of the context of the context
                  der='lambdas\tTestingZ_ErrorRate')
In [26]: | # lambdas,TestingZ_ErrorRates = np.loadtxt('LogisticRegression_TestingZErrorRates').T
```

```
In [27]: plt.figure()
    plt.title('error rates for Z-normed testing data')
    plt.plot(lambdas, np.array(TestingZ_ErrorRates)*100)
    plt.xlabel('lambda')
    plt.ylabel('error rate (%)')
    plt.show()
```



# Log Data

/usr/local/lib/python2.7/dist-packages/ipykernel/ $\_$ main $\_$ .py:17: RuntimeWarning: divide by zero encountered in double $\_$ scalars

```
In [32]: # np.savetxt('LogisticRegression_TrainingLogErrorRates',zip(valid_lambdas,TrainingLog_Err
orRates),header='lambdas\tTrainingLog_ErrorRate')
```

```
In [34]: plt.figure()
         plt.title('error rates for Log-normed Training data')
         plt.plot(valid_lambdas, np.array(TrainingLog_ErrorRates)*100)
         plt.xlabel('lambda')
         plt.ylabel('error rate (%)')
         plt.show()
                   error rates for Log-normed Training data
           6.6
           6.2
          8
          rate
           6.0
          error
           5.6
                     20
                            40
                                    60
                                           RΩ
                               lambda
In [39]: # Error rates for training data for lambda = 1,10,100
         map(lambda lamb: LogReg_ErrorRate(xtrainLog,ytrain,xtrainLog,ytrain,lamb), [1,10,100])
         /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
         zero encountered in double scalars
Out[39]: [0.055791190864600326, 0.056443719412724309, 0.064926590538336049]
In [13]: # TestingLog_ErrorRates = map(lambda lamb: LogReg_ErrorRate(xtrainLog,ytrain,xtrainLog,yt
         rain, lamb), lambdas)
         TestingLog_ErrorRates = []
         valid_lambdas = []
         invalid lambdas = []
         for lamb in lambdas:
                 b))
                 valid_lambdas.append(lamb)
             except:
                 invalid lambdas.append(lamb)
                 pass
         /usr/local/lib/python2.7/dist-packages/ipykernel/__main__.py:17: RuntimeWarning: divide by
         zero encountered in double_scalars
In [35]: # np.savetxt('LogisticRegression TestingLogErrorRates',zip(valid lambdas,TestingLog Error
         Rates), header='lambdas\tTestingLog_ErrorRate')
In [361:
         # valid lambdas, TestingLog ErrorRates = np.loadtxt('LogisticRegression TestingLogErrorRat
         es').T
In [37]:
         plt.figure()
         plt.title('error rates for Log-normed Testing data')
         plt.plot(valid_lambdas, np.array(TestingLog_ErrorRates)*100)
         plt.xlabel('lambda')
         plt.ylabel('error rate (%)')
         plt.show()
                   error rates for Log-normed Testing data
           6.4
           6.2
          € 6.0
          e 5.8
          5.6
```

5.4 5.2 5.0 0 20 40 60 80 100

```
In [38]: # Error rates for testing data for lambda = 1,10,100
                        map(lambda lamb: LogReg_ErrorRate(xtestLog,ytest,xtrainLog,ytrain,lamb), [1,10,100])
                        /usr/local/lib/python2.7/dist-packages/ipykernel/ main .py:17: RuntimeWarning: divide by
                        zero encountered in double scalars
Out[38]: [0.05078125, 0.05338541666666664, 0.06510416666666671]
In [40]: import seaborn
                        /usr/local/lib/python2.7/dist-packages/IPython/html.py:14: ShimWarning: The `IPython.html`
                        package has been deprecated since IPython 4.0. You should import from `notebook` instead. `IPython.html.widgets` has moved to `ipywidgets`.

"`IPython.html.widgets` has moved to `ipywidgets`.", ShimWarning)
In [52]: plt.figure(figsize=(10,5))
                        plt.plot(valid_lambdas, np.array(TestingLog_ErrorRates)*100, color = 'r', label='Log(test
                        plt.plot(lambdas, np.array(TrainingZ\_ErrorRates)*100, ls = `--', color = 'g', label='Z(trainingZ\_ErrorRates)*100, ls = '--', color = 'g', ls = '---', color = 'g', ls = '---', color = 'g', ls = '---', color = 'g', ls = 'g', ls = '---', color = 'g', ls = 'g', ls
                        ain)')
                        plt.plot(lambdas, np.array(TestingZ ErrorRates)*100, color = 'g', label='Z(test)')
                        plt.plot(lambdas, np.array(TrainingBin_ErrorRates)*100, ls = '--', color = 'b', label='Bi
                        n(training)')
                        plt.plot(lambdas, np.array(TestingBin_ErrorRates)*100, color = 'b', label='Bin(test)')
                        plt.xlabel('lambda')
                        plt.ylabel('error rate (%)')
                        plt.ylim(5,10)
                        plt.legend(loc='best')
                        plt.show()
                              10
                                                                                                                                                                                   Log(train)
                         8
                                                                                                                                                                                   Log(test)

    – Z(train)

                          rate
                                                                                                                                                                                    Z(test)
                                                                                                                                                                                    Bin(training)
                                                                                                                                                                                    Bin(test)
                                 6
                                                                   20
                                                                                                                                                                                                       100
  In [ ]:
```