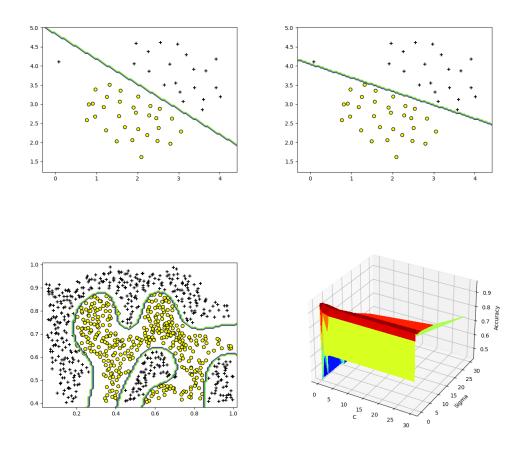
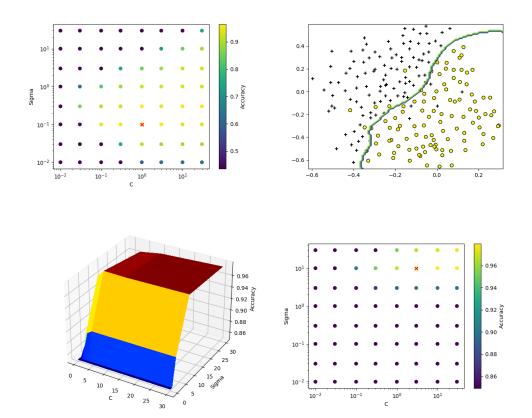
Práctica 5: Regresión lineal regularizada: sesgo y varianza

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1 Solución propuesta

1.1 Resultados obtenidos





1.2 Implementación

```
1000 import numpy as np
   import matplotlib.pyplot as plt
   import sklearn.svm as svm
   import codecs
   import os, os.path
1004
1005
   from matplotlib import cm
1006
   from mpl_toolkits.mplot3d import Axes3D
1007
   from scipy.io import loadmat
1008
   from sklearn.metrics import accuracy_score
1009
1010
1011
   from Data.Parte2.process_email import email2TokenList
1012
   from Data. Parte2.get_vocab_dict import getVocabDict
1013
   def scatter_mat(x,y):
1014
        pos = (y == 1).ravel()
1015
        neg = (y == 0).ravel()
1016
        plt.figure()
1017
        plt.scatter(x[pos, 0], x[pos, 1], color='black', marker='+')
1018
        plt.scatter(
1019
```

```
x[neg, 0], x[neg, 1], color='yellow', edgecolors='black', marker='o')
1020
1021
    def visualize_boundary(x, y, svm, zoom):
         x1 = np. linspace(x[:, 0].min() - zoom, x[:, 0].max() + zoom, 100)
1023
         x^2 = \text{np.linspace}(x[:, 1].\min() - \text{zoom}, x[:, 1].\max() + \text{zoom}, 100)
1024
1025
         x1, x2 = np.meshgrid(x1, x2)
1026
         yp = svm.\,predict\,(np.\,array\,(\,[\,x1\,.\,ravel\,(\,)\,\,,\,\,x2\,.\,ravel\,(\,)\,\,]\,)\,.T)\,.\,reshape\,(\,x1\,.\,shape\,)
         scatter_mat(x,y)
1030
         plt.contour(x1, x2, yp)
1031
         plt.show()
1033
    def parte1(x,y):
1034
         s1 = svm.SVC(kernel='linear', C=1.0)
         s1. fit(x,y)
1036
         s2 = svm.SVC(kernel='linear', C=100.0)
1038
1039
         s2. fit(x,y)
1040
         zoom = 0.4
1041
         visualize_boundary(x,y,s1,zoom)
1043
         visualize_boundary(x,y,s2,zoom)
1044
    def parte1_2(x, y, C, sigma):
1045
         s = svm.SVC(kernel='rbf', C=C, gamma= 1 / (2 * sigma ** 2))
1046
         s.fit(x,y)
         visualize_boundary(x,y,s, 0.02)
1049
         plt.show()
1051
1052
    def searchBestCandSigma(x, xVal, y, yVal):
         cs = np.array([0.01, 0.03, 0.1, 0.3, 1, 3, 10, 30])
         sigmas = np.array([0.01, 0.03, 0.1, 0.3, 1, 3, 10, 30])
1054
         numCs = cs.shape[0]
1056
         numSigmas = cs.shape[0]
1057
1058
         res = np.zeros(numCs * numSigmas).reshape(numCs, numSigmas)
1059
1060
         bestAcc = 0
1061
         bestC = -1
1062
         bestSigma = -1
1063
         \mathrm{bestSVM} \ = \ \{\}
1064
1065
1066
         for i in np.arange(numCs):
              for j in np.arange(numSigmas):
1067
                  s = svm.SVC(\,kernel=\,'rbf\,',\;C\!\!=\!\!cs\,[\,i\,]\,,\;gamma\!\!=\,1\,\;/\,\;(2\,\,*\,\,sigmas\,[\,j\,]\,\,**
1068
          2))
                  s. fit(x,y)
1069
1070
```

```
res[i,j] = accuracy_score(yVal, s.predict(xVal))
1071
                  if bestAcc < res[i,j]:
1072
1073
                      bestAcc = res[i,j]
                       bestC = cs[i]
1074
                      bestSigma = sigmas[j]
1075
                      bestSVM \, = \, s
                  print('Tested sigma ' + str(sigmas[j]))
1077
1078
             print('Tested c ' + str(cs[i]))
1080
         print("Best accuracy: " + str(bestAcc))
1081
         print("C: " + str(bestC))
1082
         print("Sigma: " + str(bestSigma))
1083
1084
        fig = plt.figure()
1085
1086
        cc, ss = np.meshgrid(sigmas, cs)
1087
1088
        ax = Axes3D(fig , auto_add_to_figure=False)
1089
        ax.set_xlabel('C')
1090
        ax.set_ylabel('Sigma')
1091
        ax.set_zlabel('Accuracy')
1092
1093
1094
         fig.add_axes(ax)
1095
        \verb|surf| = \verb|ax.plot_surface| (ss,cc,res, cmap=cm.jet, linewidth=0, antialiased)|
1096
        =False)
        plt.show()
         {\tt plt.scatter(ss, cc, c=res)}
1099
        plt.xscale('log')
plt.yscale('log')
1100
         plt.xlabel('C')
1102
         plt.ylabel('Sigma')
1103
         plt.clim(np.min(res), np.max(res))
1104
         plt.colorbar().set_label('Accuracy')
1105
        plt.scatter(bestC, bestSigma, marker='x', color='r')
1106
        plt.show()
1107
1108
        return bestSVM, bestAcc, bestC, bestSigma
1110
    def parte1_3(x, xVal, y, yVal):
1111
        s, acc, c, sigma = searchBestCandSigma(x, xVal, y, yVal)
1112
1113
1114
         visualize\_boundary(x,y,s,0.02)
1115
         plt.show()
1116
1117
1118
1119
1120
1121
1122
```

```
1123
   def filterEmailWithDictionary(dict, email):
1124
        emailDict = np.zeros([len(dict)])
        for word in email:
1126
            if word in dict:
1127
                emailDict[dict[word]-1] = 1
1128
1129
        return emailDict
1130
   def readEmails(dict, route, folder, format, size):
1132
        emails = np.empty((size, len(dict)))
        for i in np.arange(size):
1134
            email_contents = codecs.open(route + folder + '{0:04d}'.format(i+1)
1135
        + format, 'r', encoding='utf 8', errors='ignore').read()
            emails[i] = filterEmailWithDictionary(dict, email2TokenList(
1136
        email_contents))
        print('Done reading and preparing ' + folder)
        return emails
1138
1139
   def createXandY(spam, easyHam, hardHam, ySpam, yEasy, yHard):
1140
        return np.append(np.append(spam, easyHam, axis=0), hardHam, axis=0), np
1141
        .append(np.append(ySpam, yEasy, axis=0), yHard, axis=0)
1142
   def parte2():
1143
        dict = getVocabDict()
1144
1145
        route = 'Data/Parte2/
1146
        folders = ['spam/', 'easy_ham/', 'hard_ham/']
1147
        format = '.txt'
1148
1149
        print ('Comencing to read data')
1150
        spam = readEmails(dict, route, folders[0], format, len([name for name
1152
       in os. listdir ('./' + route + folders [0])))
        spamSize = len(spam)
1154
        ySpam = np.ones(spamSize)
        easy = readEmails(dict, route, folders[1], format, len([name for name
1156
       in os.listdir('./' + route + folders[1])]))
        easySize = len(easy)
        yEasy = np.zeros(easySize)
1158
1159
        hard = readEmails(dict, route, folders[2], format, len([name for name
1160
       in os.listdir('./' + route + folders[2])))
        hardSize = len(hard)
1161
        yHard = np. zeros (hardSize)
1162
1163
1164
        trainPerc = 0.6
        valPerc = 0.3 + trainPerc
1165
        # rest for test
1166
        trainSpam = int(trainPerc * spamSize)
1168
        trainEasy = int(trainPerc * easySize)
1169
```

```
trainHard = int(trainPerc * hardSize)
1170
1171
        valSpam = int(valPerc * spamSize)
1172
        valEasy = int(valPerc * easySize)
1173
        valHard = int(valPerc * hardSize)
1174
1175
        x, y = createXandY(spam[:trainSpam],
1176
                              easy [: trainEasy],
                              hard [: trainHard],
1178
                             ySpam[:trainSpam],
1179
                             yEasy [: trainEasy],
1180
                             yHard [: trainHard])
1181
1182
        xVal, yVal = createXandY(spam[trainSpam:valSpam],
1183
                                     easy[trainEasy:valEasy],
1184
                                     hard [trainHard:valHard],
1185
                                   ySpam[trainSpam:valSpam],
1186
                                   yEasy [trainEasy:valEasy],
1187
                                   yHard [trainHard:valHard])
1188
1189
        xTest, yTest = createXandY(spam[valSpam:],
1190
                              easy[valEasy:],
1191
                              hard [valHard:],
1192
                             ySpam [valSpam:],
1193
                             yEasy [valEasy:],
1194
                             yHard [valHard:])
1195
1196
        s, acc, c, sigma = searchBestCandSigma(x, xVal, y, yVal)
1197
1198
        print('Accuracy over Test sample: ' + str(accuracy_score(yTest, s.
1199
        predict(xTest)) * 100) + '%'
1200
1201
    def main():
        data1 = loadmat("Data/Parte1/ex6data1.mat")
1202
1203
        x1 = data1['X']
1204
        y1 = data1[,y,]
1205
        y1R = np.ravel(y1)
1206
1207
        data2 = loadmat("Data/Parte1/ex6data2.mat")
1208
1209
        x2 = data2['X']
1210
        y2 = data2['y']
        y2R = np.ravel(y2)
1212
1213
        data3 = loadmat("Data/Parte1/ex6data3.mat")
1214
1215
        x3 = data3['X']
1216
        y3 = data3['y']
1217
        y3R = np.ravel(y3)
1218
1219
        x3Val = data3['Xval']
        y3Val = data3['yval']
1221
```

```
1222 y3ValR = np.ravel(y3Val)

1223

1224 partel(x1, y1R)

1225 partel_2(x2, y2R, 1.0, 0.1)

1226 partel_3(x3, x3Val, y3R, y3ValR)

1227 parte2()

1228

1229

1230 if __name__ = "__main__":

main()
```

main.py