# United Way MLPP: MultinomialNB Optical Digits Example

```
0  # -*- coding: utf-8 -*-
1
2
3  """
4  Title:                  Sample MNBayes1
5  Project Descriptor:     Machine Learning Pilot Project (United Way)
6  Project ID:             2016SoE013 (STAT_570_Consulting)
7  Record:
8  Author:                 bmarron
9  Origin Date:            28 May 2016
10 """
11
12 """
13 This script provides an example of the use of the Multinomial Bayes
14 algorithm as applied to the  Optical Recognition of Handwritten Digits
15 dataset. The original source of the dataset:
16         E. Alpaydin, C. Kaynak
17         Department of Computer Engineering
18         Bogazici University, 80815 Istanbul Turkey
19         alpaydin@boun.edu.tr
20         July 1998
21
22 The dataset is available at UC Irivine Machine Learning Repository.
23 """
24
25 """
26 Important websites:
27     #UCI Machine Learning Repository
28 http://archive.ics.uci.edu/ml/
29
30     #Python machine learning tools
31 http://scikit-learn.org/
32 """
33
34 #%% Import packages
35
36
37 import numpy as np
38 import cPickle
39 import urllib
40 import csv
41 import operator
42 import pylab as pl
43 import pandas as pd
44 from time import time
45 from sklearn import cross_validation
46 from sklearn.utils.fixes import bincount
47 from sklearn.metrics import accuracy_score
48 from sklearn.metrics import confusion_matrix
49 from sklearn.metrics import classification_report
50 from sklearn.naive_bayes import MultinomialNB
51 from sklearn.svm import SVC
52 from sklearn.svm import LinearSVC
53
54 #%%
55
56 #Download and save the raw, UCI opttical digits data and data info sheet
57
58 '''
59 http://archive.ics.uci.edu/ml/datasets/Optical+Recognition+of+Handwritten+Digits
60 http://archive.ics.uci.edu/ml/machine-learning-databases/optdigits/optdigits.tra
61 http://archive.ics.uci.edu/ml/machine-learning-databases/optdigits/optdigits.tes
62 '''
63
```

```
80
81  X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=.5,
        random_state=random_state)
82
83  #%% Confirm data content with info sheet about training dataset
84
85  #Data sheet: 0=376, 1=389, 2=380, 9=382
86
87
88  test0 = np.where(raw_optdigits_tr[:, 64] == 0)
89  test0 = list(test0[0])
90
91  test1 = np.where(raw_optdigits_tr[:, 64] == 1)
92  test1 = list(test1[0])
93
94  test2 = np.where(raw_optdigits_tr[:, 64] == 2)
95  test2 = list(test2[0])
96
97  test9 = np.where(raw_optdigits_tr[:, 64] == 9)
98  test9 = list(test9[0])
99
100 print "zeros: %s" %(len(test0))
101 print "ones: %s" %(len(test1))
102 print "twos: %s" %(len(test2))
103 print "nines: %s" %(len(test9))
104
105 #%% Split training dataset(tr_d) and test dataset(te_d) into features (X)
106
107 #and class (y)
108
109
110 tr_d_X, tr_d_y = raw_optdigits_tr[:,0:64], raw_optdigits_tr[:, 64]
111 te_d_X, te_d_y = raw_optdigits_te[:,0:64], raw_optdigits_te[:, 64]
112
113
114 #%% Save the split (processed) data
115
116
117 # Save tr_d_X
118
119 with open("tr_d_X.pkl", 'wb') as f:
120     cPickle.dump(tr_d_X, f, protocol=2)
121
122 # Save tr_d_y
123
124 with open("tr_d_y.pkl", 'wb') as f:
125     cPickle.dump(tr_d_y, f, protocol=2)
126
127 # Save te_d_X
128
129 with open("te_d_X.pkl", 'wb') as f:
130     cPickle.dump(te_d_X, f, protocol=2)
131
132 # Save te_d_y
133
134 with open("te_d_y.pkl", 'wb') as f:
135     cPickle.dump(te_d_y, f, protocol=2)
136
137 #%%
138
```

```
#Reload the processed data into Python, if needed

tr_d_X = cPickle.load(open("/home/bmarron/Desktop/tr_d_X.pkl","rb"))

#Reload the processed data

tr_d_y = cPickle.load(open("/home/bmarron/Desktop/tr_d_y.pkl","rb"))

#Reload the processed data

te_d_X = cPickle.load(open("/home/bmarron/Desktop/te_d_X.pkl","rb"))

#Reload the processed data

te_d_y = cPickle.load(open("/home/bmarron/Desktop/te_d_y.pkl","rb"))


#%% Based on the method  at

#http://scikit-learn.org/stable/auto_examples/text/
   mlcomp_sparse_document_classification.html#example-text-mlcomp-sparse-
   document-classification-py

# Benchmark classifiers


def benchmark(clf_class, params, name):
    print("parameters:", params)
    t0 = time()
    clf = clf_class(**params).fit(tr_d_X, tr_d_y)
    print("done in %fs" % (time() - t0))

    if hasattr(clf, 'coef_'):
        print("Percentage of non zeros coef: %f"
              % (np.mean(clf.coef_ != 0) * 100))
        # .coef_ output is a read-only file

        # Python does not define a read-write variable!

        # Export the wgt vector data from .coef_

        wgts = clf.coef_
        with open('wgts_MNBayes.csv', 'w+') as f:
            a= csv.writer(f, delimiter=',')
            a.writerows(wgts)


    print("Predicting the outcomes of the testing set")
    t0 = time()
    pred = clf.predict(te_d_X)
    print("done in %fs" % (time() - t0))

    print("Classification report on test set for classifier:")
    print(clf)
    print()
    print(classification_report(te_d_y, pred))

    cm = confusion_matrix(te_d_y, pred)
    df = pd.DataFrame(cm)
    print("Confusion matrix:")
```

```
197     print(cm)
198     print df.to_latex()
199
200
201     print ("Accuracy:")
202     print accuracy_score(te_d_y, pred)
203
204     # Show confusion matrix
205
206     pl.matshow(cm)
207     pl.title('Confusion matrix of the\n %s classifier \n (alpha=0.03)' % name,
            fontsize=12, y=1.10)
208     pl.colorbar()
209     pl.savefig("MNBayes1.pdf")
210
211 #%%
212
213 print("Testbenching a MultinomialNB classifier...")
214 parameters = {'alpha': 0.03}
215
216 benchmark(MultinomialNB, parameters, 'MultinomialNB')
217
218 pl.show()
219
220 #%%
221
222 # coef_ is empirical log probability of features given a class, P(x_i|y)
223
224 # coef_ in MultinomialNB mirrors the feature_log_prob_ ; useful
225
226 #for interpreting MultinomialNB as a linear model
227
228
229 # Import 'wgts_MNBayes.csv' with Spyder (green arrow); import as 'array'
230
231 # Take the absolute value of the wgts
232
233 # Define a list of feature numbers (0 to 63); actual features run i=1,2,...63
234
235
236 ############
237
238 #import 'wgts_MNBayes.csv' with Spyder (green arrow)
239
240 ############
241
242 abs_csv = abs(wgts_MNBayescsv)
243 features =np.arange(64)
244
245 # Create tuples (feature #, feature value)
246
247 wgts_features=[]
248 for i in range(10):
249     x=zip(features, abs_csv[i])
250     wgts_features.append(x)
251
252 # Reverse sort on the the second element in each tuple
253
254 for i in range(10):
255     wgts_features[i]=sorted(wgts_features[i], key=operator.itemgetter(1),
```

```
        reverse=True)
256
257 df2 = pd.DataFrame(wgts_features)
258 df2.to_csv('sorted_wgts_features_MNBayes.csv')
259
260 #%% Ouput w/o panda dataframe
261
262
263 with open('sorted_wgts_features_MNBayes.csv', 'w+') as f:
264             a= csv.writer(f, delimiter=',')
265             a.writerows(wgts_features)
```