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PA3
CSE 151

Q1.

Passes 2, 3, 4 in perceptron:

training errors : 0.04

test errors:0.06

training errors : 0.02

test errors:0.05

training errors : 0.02

test errors:0.05

Passes 2, 3, 4 in voted perceptron:

training errors : 0.04

test errors:0.06

training errors : 0.03

test errors:0.04

training errors : 0.02

test errors:0.05

Passes 2, 3, 4 in averaged perceptron:

training errors : 0.05

test errors:0.08

training errors : 0.03

test errors:0.06

training errors : 0.03

test errors:0.05

Q2.

('The three highest:', ['file'], ['program'], ['line'])

('The three lowest:', ['he'], ['team'], ['game'])

Q3.

(a)

0.80128205 ($i = 5, j = 5$). Because it means the true label is 5 and the predict label is 5 as well, the rate is 80%. So the classifier 5 has the highest accuracy.

(b)

0.37142857 ($i = 3, j = 3$). Because it means the true label is 3 and the predict label is 3 as well, but the rate is 37%. So the classifier 3 has the least accuracy.

(c)

$i = 5, j = 1, 2, 3, 4, 6$. It means its predict label is 5, but the error rate is 25%. So the classifier 5 most often mistakenly classifies.

* output:

```
[[0.23243243 0.27604167 0.45142857 0.25543478 0.11538462 0.34259259]
 [0.71891892 0.00520833 0.03428571 0.02173913 0.      0.      ]
 [0.01081081 0.65625   0.03428571 0.02717391 0.01282051 0.01851852]
 [0.      0.015625  0.37142857 0.      0.      0.02777778]
 [0.01621622 0.00520833 0.      0.69021739 0.      0.      ]
 [0.01621622 0.03125   0.07428571 0.00543478 0.80128205 0.12037037]
 [0.00540541 0.01041667 0.03428571 0.      0.07051282 0.49074074]]
```

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```
from __future__ import division
import numpy
from heapq import nlargest, nsmallest
from sklearn.metrics import confusion_matrix
from operator import itemgetter
```

```
# read point from .txt
train = numpy.loadtxt("pa3train.txt")
test = numpy.loadtxt("pa3test.txt")
len_feature = len(train[0])-1
```

```
# get the subset that label is 1 or 2
sub_train = []
sub_test = []
for point in train:
    if point[len_feature] == 1 or point[len_feature] == 2:
        sub_train.append(point)
for point in test:
    if point[len_feature] == 1 or point[len_feature] == 2:
        sub_test.append(point)
```

```
# function for perceptron
def perceptron(train, passes, label):
    # initialize the w
    w = numpy.zeros(len_feature)
    # as the passes passed in
    for p in range(0, passes):
        # check each point
        for feature in train:
            # if the label is what we want, set as 1
            if label == feature[len_feature]:
                new_label = 1
            # else set as -1
            else:
                new_label = -1
            #label = point[feature_len]
            # print("Now the point line is: ", point)
            # check do we need to update
            if label * numpy.dot(w, point[:len_feature]) <= 0:
                w = w + (label * point[:len_feature])
            #print("Now the w is: ", w)
    return w
```

```
# check the sign of number
def sign(number):
    if number >= 0 :
        return 1
    else :
        return -1
```

```
# function to calculate the error of perceptron
def perceptron_error(test, w, label):
    # set the wrong number
    err = 0
    # for loop to check each loop
    for feature in test:
        # check if the sign match up what we expect (1 or -1)
        sign = numpy.dot(w, feature[:len_feature])
```

```

    # when the sign is - but the label should be 1
    if sign < 0 and (feature[len_feature] == label):
        err += 1
    # when the sign is + but the label should be 2(-1)
    elif sign >= 0 and (feature[len_feature] != label):
        err += 1
    #print("Now the wrong is: ", wrong)
    #print("len(test) is: ", len(test))
    return err / len(test)

# call function to get results
print("pass 2 to pass 4 in perceptron")
for passes in range(1,5):
    w = perceptron(sub_train, passes, 1)
    #print("The final w is: ", w)
    print(perceptron_error(sub_train, w, 1))
    print(perceptron_error(sub_test, w, 1))

# function for voted perceptron
def voted_perceptron(train, passes, label):
    w = numpy.zeros(len_feature)
    cm = 1
    output = []
    # for passes time
    for p in range(0, passes):
        # for each point
        for feature in train:
            # if label match up set it as 1 (label 1)
            if label == feature[len_feature]:
                new_label = 1
            # label 2
            else:
                new_label = -1
            # check if need to update w
            if new_label * numpy.dot(w, feature[:len_feature]) <= 0:
                w = w + (label * feature[:len_feature])
                cm = 1
            # save the (w, c)
            output.append((w, cm))
        else:
            # update cm
            cm += 1
    output.append((w, cm))
    return output

# function to calculate the error
def voted_perceptron_error(test, classifiers, label):
    err = 0
    # for each point in test file
    for feature in test:
        # set test_feature and predict
        test_feature = feature[:len_feature]
        predict = 0
        # for each (w, c) calculate the predict
        for c in classifiers:
            predict += c[1] * sign(numpy.dot(c[0], test_feature))
        # if doesn't match up then increment the wrong number
        if sign(predict) < 0 and feature[len_feature] == label:
            err += 1
        elif sign(predict) >= 0 and feature[len_feature] != label:
            err += 1
    return err / len(test)

# call function to calculate
print("pass 2 to pass 4 in voted perceptron")
for passes in range(1,5):
    classis = voted_perceptron(sub_train, passes, 1)

```

```

print(voted_perceptron_error(sub_train, classis, 1))
print(voted_perceptron_error(sub_test, classis, 1))

# function for averaged perceptron error
def average_perceptron_error(test, classifiers, label):
    err = 0
    w = sum([classi[0] * classi[1] for classi in classifiers])
    for feature in test:
        sign = numpy.dot(w, feature[:len_feature])
        if sign < 0 and feature[len_feature] == label:
            err += 1
        elif sign >= 0 and feature[len_feature] != label:
            err += 1
    return err / len(test)

# call function to calculate
print("pass 2 to pass 4 in averaged perceptron")
for passes in range(1,5):
    classis = voted_perceptron(sub_train, passes, 1)
    print(average_perceptron_error(sub_train, classis, 1))
    print(average_perceptron_error(sub_test, classis, 1))

# wavg from averaged perceptron and three passes
# part 2
# call function to get 3 passes voted perceptron
avg_all = train_voted_perceptron(one_two_subset, 3, 1)
#print("Now the classis is: ", classis)
w = sum([avg_each[0] * avg_each[1] for avg_each in avg_all])

# use nlargest and nsmallest to get 3 highest and the three lowest
# get the index of 3 max and min value in pair
largest = nlargest(3, enumerate(w), key=lambda x: x[1])
lowest = nsmallest(3, enumerate(w), key=lambda x: x[1])
#print("largest is: ", top)
#print("lowest is: ", bot)

# set to save diction
diction = []

# read ling by line
for line in open("pa3dictionary.txt", 'r'):
    diction.append(line.strip().split('/n'))

# get the answer
print("The three highest:", diction[largest[0][0]], diction[largest[1][0]], diction[largest[2][0]])
#print(max(classis, key=itemgetter(1))[0])
#print(diction[max(classis, key=itemgetter(1))[0]])
print("The three lowest:", diction[lowest[0][0]], diction[lowest[1][0]], diction[lowest[2][0]])

# part 3: one vs all
# for each class i : 1 to 6 run 1 pass

class_one = train_perceptron(train, 1, 1)
class_two = train_perceptron(train, 1, 2)
class_three = train_perceptron(train, 1, 3)
class_four = train_perceptron(train, 1, 4)
class_five = train_perceptron(train, 1, 5)
class_six = train_perceptron(train, 1, 6)

# if ci(x) = i, then predict label i ( exactly one i )
# if ci(x) = i ( more than one i ) or ci(x) = i ( no i , Don't know )
CX = []
CX.append(class_one)
CX.append(class_two)
CX.append(class_three)

```

```

CX.append(class_four)
CX.append(class_five)
CX.append(class_six)
# print("CX is: ", CX)

'''
For each test data, you will have {predicted label, true label}.
The predicted label can be {1, 2, 3, 4, 5, 6, Don't know} => 7 values
The true label can be {1, 2, 3, 4, 5, 6} => 6 values
'''

# function to predict the test example
def test_predict( test, CX ):
    # print("CX: ", CX)
    pred_result = []
    # for each line of feathre in testData
    for feature in test:
        # print("Now the feature is: ", feature)
        # set label as
        tmp_label = 0
        # loop to check each feature
        for i in range(0, len(CX)):
            if numpy.dot(CX[i], feature[0:-1]) > 0:
                print("legal")
                ''' if has more than one label -> Don't know '''
                if tmp_label != 0:
                    print("Dont know")
                    tmp_label = 0
                    break
                else:
                    ''' set the label '''
                    print("Get label")
                    tmp_label = i + 1
                ''' add in to result '''
            print("tmp_label: ", tmp_label)
            pred_result.append(tmp_label)
    return pred_result

# build confusion matrix
pred_l = test_predict( test, CX )
''' set the true label'''
print("pred_l:", pred_l)
true_l = []
for tf in test:
    true_l.append(tf[feature_len])
print("true_l: ", true_l)
my_matrix = confusion_matrix( pred_l, true_l)

print("my_martixL ")
print(my_matrix)

''' set the number of label '''
test_label = [0,0,0,0,0,0]
for tf in test:
    tl_idx = int(tf[feature_len])
    test_label[tl_idx] += 1
res_matrix = numpy.zeros((7,6))

''' calculate C/N for each entry '''
for r in range(0, 7):
    for c in range(1, 7):
        print("My mat: ", my_matrix[r][c] )
        print("test_m: ", test_label[c] )
        res_matrix[r][c-1] = my_matrix[r][c] / test_label[c]

```

```
print("Now con_matrix is ")  
print(res_matrix)
```