import numpy
import urllib
import scipy.optimize
import random
from sklearn import svm
import os
import yaml
import random
import operator

```
def parseData(fname):
    for I in urllib.request.urlopen(fname):
        yield eval(I)
data = list(parseData("http://jmcauley.ucsd.edu/cse190/data/beer/beer_50000.json"))
```

```
print('QUESTION 1')
ratings = {'1-star':0, '2-star':0, '3-star':0, '4-star':0, '5-star':0}
for d in data:
    ratings[str(int(d['review/taste']))+'-star'] += 1
print('Rounded down ratings: {}'.format(ratings))
```

```
print('QUESTION 2')
listOfBeers = set()
ratings = dict()
averageRatings = dict()
for d in data:
  listOfBeers.add(d['beer/name'])
for beerName in listOfBeers:
  ratings[beerName] = 0
  averageRatings[beerName] = 0
for d in data:
  ratings[d['beer/name']] += 1
  averageRatings[d['beer/name']] += d['review/taste']
for key in ratings.keys():
  averageRatings[key] = -1 if ratings[key] < 5 else averageRatings[key]/ratings[key]</pre>
maximum = max(averageRatings.values())
result = []
for item in averageRatings.items():
  if item[1] == maximum:
    result.append(item)
print(result)
```

```
print('QUESTION 3')
data2 = [d for d in data if 'beer/ABV' in d.keys() and 'beer/style' in d.keys()]
def feature(datum):
  feat = [1]
  if datum['beer/style'] == 'Hefeweizen':
    feat.append(1)
  else:
    feat.append(0)
  feat.append(datum['beer/ABV'])
  return feat
X = [feature(d) for d in data2]
y = [d['review/taste'] for d in data2]
theta,residuals,rank,s = numpy.linalg.lstsq(X, y, rcond=None)
print(theta)
#theta[0] shows us the offset. Theta[1] is the weighting of the beer being a
#Hefeweizen on the taste score, in this case it has negative impact on the score.
#Theta[2] shows the impact of the beer alcohol content on the taste score, in this case
#higher is better
```

```
print('QUESTION 4')
train = data2[:len(data2)//2]
X_train = [feature(d) for d in train]
y_train = [d['review/taste'] for d in train]
theta,residuals,rank,s = numpy.linalg.lstsq(X_train, y_train, rcond=-1)
predictions = numpy.matmul(theta, numpy.transpose(X_train))
print('Error for train data =', numpy.sum(numpy.square(numpy.subtract(predictions, y_train))))
test = data2[len(data2)//2:]
X_test = [feature(d) for d in test]
y_test = [d['review/taste'] for d in test]
predictions = numpy.matmul(theta, numpy.transpose(X_test))
print('Error for test data =', numpy.sum(numpy.square(numpy.subtract(predictions, y_test))))
```

```
#Question 5
```

#might result in skewed data

```
print('QUESTION 5')
dataCopy = data2.copy()
for i in range(10):
  random.shuffle(dataCopy)
train = dataCopy[:len(dataCopy)//2]
X train = [feature(d) for d in train]
y_train = [d['review/taste'] for d in train]
theta,residuals,rank,s = numpy.linalg.lstsq(X train, y train, rcond=-1)
predictions = numpy.matmul(theta, numpy.transpose(X_train))
print('Error for train data =', numpy.sum(numpy.square(numpy.subtract(predictions, y_train))))
test = dataCopy[len(dataCopy)//2:]
X_test = [feature(d) for d in test]
y test = [d['review/taste'] for d in test]
predictions = numpy.matmul(theta, numpy.transpose(X_test))
print('Error for test data =', numpy.sum(numpy.square(numpy.subtract(predictions, y_test))))
#The data that is pulled from the website might be in some certain order and dividing it before
shuffling
```

```
#Question 6
print('QUESTION 6')
def feature(datum):
  #feat = [1]
  feat = []
  feat.append(datum['review/taste'])
  feat.append(datum['review/appearance'])
  feat.append(datum['review/aroma'])
  feat.append(datum['review/palate'])
  feat.append(datum['review/overall'])
  return feat
dataCopy = data2.copy()
for i in range(10):
  random.shuffle(dataCopy)
train = dataCopy[:len(dataCopy)//2]
X train = [feature(d) for d in train]
y train = [1 if d['beer/style'] == 'Hefeweizen' else 0 for d in train]
clf = svm.SVC(C=1000, kernel='rbf', gamma='scale', degree=2)
clf.fit(X_train, y_train)
train predictions = clf.predict(X_train)
print('Accuracy for train data =', 100-
(100*numpy.sum(numpy.square(numpy.subtract(train_predictions, y_train)))/len(y_train)), '%')
test = dataCopy[len(dataCopy)//2:]
X test = [feature(d) for d in test]
y test = [1 if d['beer/style'] == 'Hefeweizen' else 0 for d in test]
test predictions = clf.predict(X test)
print('Accuracy for test data =', 100-
```

(100\*numpy.sum(numpy.square(numpy.subtract(test\_predictions, y\_test)))/len(y\_test)), '%')

```
#Question 7
print('QUESTION 7')
def feature(datum):
  feat = [1]
  feat.append(datum['beer/ABV'])
  feat.append(1 if 'malt' in datum['review/text'] or 'malty' in datum['review/text'] else 0)
  feat.append(datum['review/timeUnix'])
  return feat
train = dataCopy[:len(dataCopy)//2]
X train = [feature(d) for d in train]
y train = [1 if d['beer/style'] == 'Hefeweizen' else 0 for d in train]
clf2 = svm.SVC(C=1000, kernel='rbf', gamma='scale')
clf2.fit(X train, y train)
train_predictions = clf2.predict(X_train)
print('Accuracy for train data =', 100-
(100*numpy.sum(numpy.square(numpy.subtract(train predictions, y train)))/len(y train)), '%')
test = dataCopy[len(dataCopy)//2:]
X test = [feature(d) for d in test]
y_test = [1 if d['beer/style'] == 'Hefeweizen' else 0 for d in test]
test predictions = clf2.predict(X test)
print('Accuracy for test data =', 100-
(100*numpy.sum(numpy.square(numpy.subtract(test_predictions, y_test)))/len(y_test)), '%')
#The features used are:
#[1, beer/ABV, Boolean indicating if review text contains the word 'malt' or 'malty',
review/timeUnix]
```

```
print('QUESTION 8')
c = [0.1, 10, 1000, 100000]

for reg in c:
    clf = svm.SVC(C=reg, kernel='rbf', gamma='scale')
    clf.fit(X_train, y_train)
    train_predictions = clf.predict(X_train)
    print('------For regularization factor =', reg, '------')
    print('Accuracy for train data =', 100-
(100*numpy.sum(numpy.square(numpy.subtract(train_predictions, y_train)))/len(y_train)), '%')
    test_predictions = clf.predict(X_test)
    print('Accuracy for test data =', 100-
(100*numpy.sum(numpy.square(numpy.subtract(test_predictions, y_test)))/len(y_test)), '%')
```

# **Printed Results**

### **QUESTION 1**

Rounded down ratings: {'1-star': 554, '2-star': 2723, '3-star': 12934, '4-star': 29458, '5-star': 4331}

#### **QUESTION 2**

[('Founders CBS Imperial Stout', 4.6970172684458396)]

#### QUESTION 3

[3.11795084 -0.05637406 0.10877902]

#### **QUESTION 4**

Error for train data = 12099.201400335607 Error for test data = 10592.663029965459

#### **QUESTION 5**

Error for train data = 11435.60803063151 Error for test data = 11053.513210952766

#### **QUESTION 6**

Accuracy for train data = 98.776 % Accuracy for test data = 98.712 %

### **QUESTION 7**

Accuracy for train data = 99.172 % Accuracy for test data = 97.968 %

#### **QUESTION 8**

------For regularization factor = 0.1 ------
Accuracy for train data = 98.748 %

Accuracy for test data = 98.78 %

-------For regularization factor = 10 ------
Accuracy for train data = 98.872 %

Accuracy for test data = 98.56 %

------For regularization factor = 1000 -----
Accuracy for train data = 99.172 %

Accuracy for test data = 97.968 %

------For regularization factor = 100000 -----
Accuracy for train data = 99.592 %

Accuracy for test data = 97.508 %