q1

November 1, 2019

In [9]:
```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import string
from sklearn.model_selection import KFold
from sklearn.metrics import classification_report
from sklearn.metrics import f1_score, precision_score, recall_score, accuracy_score, c
from statistics import mean
from collections import Counter
from nltk import ngrams
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize, sent_tokenize
from copy import deepcopy
import operator
from math import log2
```

In [10]:
```
import nltk
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')
```

```
[nltk_data] Downloading package stopwords to
[nltk_data]     /Users/manishkumar/nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
[nltk_data] Downloading package punkt to
[nltk_data]     /Users/manishkumar/nltk_data...
[nltk_data]   Package punkt is already up-to-date!
[nltk_data] Downloading package averaged_perceptron_tagger to
[nltk_data]     /Users/manishkumar/nltk_data...
[nltk_data]   Package averaged_perceptron_tagger is already up-to-
[nltk_data]       date!
```

Out[10]: True

In [11]:
```
def get_ngrams(data, n):
    tokens = [token for token in data.split(" ") if token != ""]
    output = list(ngrams(tokens, n))
    return output
```

In [12]:
```
stop_words = set(stopwords.words('english'))

def get_postag(txt):
    tokenized = sent_tokenize(txt)

    wordsList = nltk.word_tokenize(tokenized[0])
    wordsList = [w for w in wordsList if not w in stop_words]
    tagged = nltk.pos_tag(wordsList)

    return tagged
```

In [13]:
```
def buildData(filePath):
    data = []
    uni = []
    bi = []
    tri = []
    pos = []
    file = open(filePath)

    for line in file:
        line = line.split(':')
        row = []
        row.append(line[0])
        row.append(' '.join(line[1].split(' ')[1:]).translate(str.maketrans('', '', st

        length = len(row[1].split(' '))
        unigram = get_ngrams(row[1], 1)
        bigram = get_ngrams(row[1], 2)
        trigram = get_ngrams(row[1], 3)
        postag = get_postag(row[1])

        row.append(length)

        row.append(unigram)
        uni.extend(unigram)

        row.append(bigram)
        bi.extend(bigram)

        row.append(trigram)
        tri.extend(trigram)

        row.append(postag)
        pos.extend(postag)

        data.append(row)

    return data, uni, bi, tri, pos
```

```
In [14]: data, uni, bi, tri, pos = buildData('./traindata.txt')

In [15]: def top_grams(grams, top_n):
             return Counter(grams).most_common(top_n)

In [16]: unigram_counts = top_grams(uni, 500)
         bigram_counts = top_grams(bi, 300)
         trigram_counts = top_grams(tri, 200)
         pos_counts = top_grams(pos, 500)
         lengthAvg = mean([row[2] for row in data])
         print(lengthAvg)
```

```
9.031548055759353
```

```
In [17]: def is_numeric(value):
             return isinstance(value, int) or isinstance(value, float)

In [18]: header = ['Label', 'Text', 'Length', 'Unigram', 'Bigram', 'Trigram']
         class Question:
             def __init__(self, column, value):
                 self.column = column
                 self.value = value

             def match(self, example):
                 val = example[self.column]

                 if is_numeric(val):
                     return val <= self.value

                 return self.value in val

             def __repr__(self):
                 condition = "contains"
                 return "Does %s %s %s?" % (
                     header[self.column], condition, str(self.value))

In [19]: def class_counts(rows):
             counts = {}
             for row in rows:
                 label = row[0]
                 if label not in counts:
                     counts[label] = 0
                 counts[label] += 1
             return counts

In [20]: def gini(rows):
             counts = class_counts(rows)
             impurity = 1
```

```
            for lbl in counts:
                prob_of_lbl = counts[lbl] / float(len(rows))
                impurity -= prob_of_lbl**2
            return impurity
```

In [21]:
```
def misclassifcation_error(rows):
    counts = class_counts(rows)
    max_prob = 0
    for lbl in counts:
        prob_of_lbl = counts[lbl] / float(len(rows))
        if prob_of_lbl > max_prob:
            max_prob = prob_of_lbl
    return 1 - max_prob
```

In [22]:
```
def entropy(rows):
    counts = class_counts(rows)
    impurity = 0
    for lbl in counts:
        prob_of_lbl = counts[lbl] / float(len(rows))
        impurity -= prob_of_lbl*log2(prob_of_lbl)
    return impurity
```

In [23]:
```
def info_gain(left, right, current_uncertainty, func):
    p = float(len(left)) / (len(left) + len(right))
    return current_uncertainty - p * func(left) - (1 - p) * func(right)
```

In [24]:
```
class Leaf:
    def __init__(self, rows):
        self.predictions = class_counts(rows)
```

In [25]:
```
class Decision_Node:
    def __init__(self,
                 question,
                 true_branch,
                 false_branch):
        self.question = question
        self.true_branch = true_branch
        self.false_branch = false_branch
```

In [26]:
```
questions = []

for x in unigram_counts:
    questions.append(Question(3, x[0]))

for x in bigram_counts:
    questions.append(Question(4, x[0]))

for x in trigram_counts:
    questions.append(Question(5, x[0]))
```

```
        for x in pos_counts:
            questions.append(Question(6, x[0]))

        questions.append(Question(2, lengthAvg))

        print(len(questions))
        # print(questions[1500])
```

```
1501
```

In [27]:
```
def partition(rows, question):
    trueRows = []
    falseRows = []

    for r in rows:
        if question.match(r):
            trueRows.append(r)
        else:
            falseRows.append(r)

    return trueRows, falseRows
```

In [28]:
```
def findBestSplit(rows, questions, func):
    best_gain = 0
    best_question = None
    current_uncertainty = func(rows)

    for q in questions:
        trueRows, falseRows = partition(rows, q)
        if len(trueRows) == 0 or len(falseRows) == 0:
            continue

        gain = info_gain(trueRows, falseRows, current_uncertainty, func)

        if gain >= best_gain:
            best_gain, best_question = gain, q

    return best_gain, best_question
```

In [29]:
```
def formTree(rows, questions, func):
    gain, question = findBestSplit(rows, questions, func)

    if gain == 0:
        return Leaf(rows)

    trueRows, falseRows = partition(rows, question)
    questions.remove(question)
```

```
            trueBranch = formTree(trueRows, questions, func)
            falseBranch = formTree(falseRows, questions, func)

            return Decision_Node(question, trueBranch, falseBranch)
```

```
In [30]: def classifyRow(node, row):
            if isinstance(node, Leaf):
                return node.predictions

            if node.question.match(row):
                return classifyRow(node.true_branch, row)
            else:
                return classifyRow(node.false_branch, row)
```

```
In [31]: def train(data, questions, func):
            return formTree(data, deepcopy(questions), func)
```

```
In [32]: def classify(root, rows):
            predictions = []
            for r in rows:
                predictions.append(max(classifyRow(root, r).items(), key=operator.itemgetter(
            return predictions
```

```
In [33]: def getDataInIndex(data, index):
            l = []
            for i in range(len(data)):
                if i in index:
                    l.append(data[i])
            return l
```

```
In [34]: def getActualLabels(act_data):
            act_labels = []
            for d in act_data:
                act_labels.append(d[0])
            return act_labels
```

```
In [24]: kfold = KFold(10, True, 1)
         precision = []
         recall = []
         f_score = []
         i = 0

         for trainInd,testInd in kfold.split(data):
             train_data = getDataInIndex(data, trainInd)
             test_data = getDataInIndex(data, testInd)

             root = train(train_data, questions, gini)
```

```
            prediction = classify(root, test_data)

            actual = getActualLabels(test_data)
            predicted = prediction

        #     print(classification_report(actual, predicted))
            precision.append(precision_score(actual, predicted, average='macro'))
            recall.append(recall_score(actual, predicted, average='macro'))
            f_score.append(f1_score(actual, predicted, average='macro'))

            print("Training...")

        print("Precision Score = " + str(mean(precision)))
        print("Recall Score = " + str(mean(recall)))
        print("F Score = " + str(mean(f_score)))
```

```
Training...
Training...
Training...
Training...
Training...
Training...
Training...
Training...
Training...
Training...
Precision Score = 0.8028907609403665
Recall Score = 0.7574562257221622
F Score = 0.7722368576621397
```

0.1 Part 2

- All
- Unigram, Bigram, Trigram, POS
- Unigram, Bigram, Trigram

```
In [35]: classes = ['ABBR', 'DESC', 'ENTY', 'HUM', 'LOC', 'NUM']
```

```
In [54]: def getReport(traindata, testdata, uniFlag=True, biFlag=True, triFlag=True, posFlag=T
             allQuestions = []

             if uniFlag:
                 for x in unigram_counts:
                     allQuestions.append(Question(3, x[0]))

             if biFlag:
                 for x in bigram_counts:
```

```
                    allQuestions.append(Question(4, x[0]))

            if triFlag:
                for x in trigram_counts:
                    allQuestions.append(Question(5, x[0]))

            if posFlag:
                for x in pos_counts:
                    allQuestions.append(Question(6, x[0]))

            if lenFlag:
                allQuestions.append(Question(2, lengthAvg))

            print("No of questions = " + str(len(allQuestions)))
            print("Training...")
            root = train(traindata, allQuestions, func)
            print("Predicting...")
            prediction = classify(root, testdata)
            actual = getActualLabels(testdata)
            print("Prediction done...")
            matrix = confusion_matrix(actual, prediction)
            acc = matrix.diagonal()/matrix.sum(axis=1)
            accuracy_report = dict(zip(classes, acc))

            return accuracy_report, root, prediction, actual
```

```
In [55]: testdata = buildData('./testdata.txt')[0]
         len(testdata)
```

```
Out[55]: 500
```

```
In [38]: print(getReport(traindata=data, testdata=testdata)[0])
```

```
No of questions = 1501
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.9710144927536232, 'ENTY': 0.723404255319149, 'HUM': 0.84
```

```
In [39]: print(getReport(traindata=data, testdata=testdata, func=entropy)[0])
```

```
No of questions = 1501
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.9710144927536232, 'ENTY': 0.5, 'HUM': 0.8615384615384616
```

```
In [40]: print(getReport(traindata=data, testdata=testdata, func=misclassifcation_error)[0])
```

```
No of questions = 1501
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.8260869565217391, 'ENTY': 0.7978723404255319, 'HUM': 0.8
```

```
In [41]: print(getReport(traindata=data, testdata=testdata, lenFlag=False)[0])
```

```
No of questions = 1500
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.9710144927536232, 'ENTY': 0.723404255319149, 'HUM': 0.84
```

```
In [44]: print(getReport(traindata=data, testdata=testdata, lenFlag=False, func=entropy)[0])
```

```
No of questions = 1500
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.9710144927536232, 'ENTY': 0.5, 'HUM': 0.8615384615384616
```

```
In [45]: print(getReport(traindata=data, testdata=testdata, lenFlag=False, func=misclassifcatio
```

```
No of questions = 1500
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.8260869565217391, 'ENTY': 0.7978723404255319, 'HUM': 0.8
```

```
In [46]: print(getReport(traindata=data, testdata=testdata, lenFlag=False, posFlag=False)[0])
```

```
No of questions = 1000
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.9782608695652174, 'ENTY': 0.6276595744680851, 'HUM': 0.8
```

```
In [47]: print(getReport(traindata=data, testdata=testdata, lenFlag=False, posFlag=False, func=
```

```
No of questions = 1000
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.427536231884058, 'ENTY': 0.648936170212766, 'HUM': 0.876
```

```
In [48]: print(getReport(traindata=data, testdata=testdata, lenFlag=False, posFlag=False, func=

No of questions = 1000
Training...
Predicting...
Prediction done...
{'ABBR': 0.6666666666666666, 'DESC': 0.8188405797101449, 'ENTY': 0.7340425531914894, 'HUM': 0.8
```

1 Error Analysis

```
In [51]: def getWrongPrediction(prediction, actual, dataset):
             data_list = []

             for i in range(len(prediction)):
                 if prediction[i] !=actual[i] :
                     data_list.append(dataset[i])

             return data_list

In [56]: _ , root_gini, prediction_gini, actual_gini  = getReport(traindata=data, testdata=test
         wrong_data = getWrongPrediction(prediction_gini, actual_gini, testdata)

No of questions = 1501
Training...
Predicting...
Prediction done...


In [58]: len(wrong_data)

Out[58]: 88

In [61]: _ , root_entropy, prediction_entropy, actual_entropy  = getReport(traindata=data, test
         wrong_data_en = getWrongPrediction(prediction_entropy, actual_entropy, wrong_data)
         len(wrong_data_en)

No of questions = 1501
Training...
Predicting...
Prediction done...


Out[61]: 78

In [63]: _ , root_mis, prediction_mis, actual_mis  = getReport(traindata=data, testdata=wrong_
         wrong_data_mis = getWrongPrediction(prediction_entropy, actual_entropy, wrong_data)
         len(wrong_data_mis)
```

```
No of questions = 1501
Training...
Predicting...
Prediction done...
```

Out[63]: 78