

q1

October 24, 2019

```
In [1]: 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
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
```

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In [2]: import nltk
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('averaged_perceptron_tagger')
```

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

Out[2]: True

```
In [3]: def get_ngrams(data, n, common_n):
tokens = [token for token in data.split(" ") if token != ""]
output = list(ngrams(tokens, n))

ngram_counts = Counter(output)
ng_counts = ngram_counts.most_common(common_n)
```

```
    return output
```

```
In [4]: stop_words = set(stopwords.words('english'))
```

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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
```

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In [5]: data = []
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```
    one_grams = []
```

```
    uni = []
```

```
    bi = []
```

```
    tri = []
```

```
    pos = []
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```
    file = open('./traindata.txt')
```

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    for line in file:
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```
        line = line.split(':')
```

```
        row = []
```

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        row.append(line[0])
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```
        row.append(' '.join(line[1].split(' ')[1:]).translate(str.maketrans('', '', string
```

```
        length = len(row[1].split(' '))
```

```
        unigram = get_ngrams(row[1], 1, 500)
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        bigram = get_ngrams(row[1], 2, 300)
```

```
        trigram = get_ngrams(row[1], 3, 200)
```

```
        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)
```

```
len(data)
```

```
print(data[0])
```

```
['DESC', 'How did serfdom develop in and then leave Russia', 9, [('How',), ('did',), ('serfdom
```

```
In [6]: def top_grams(grams, top_n):  
        return Counter(grams).most_common(top_n)
```

```
In [7]: 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 [8]: def is_numeric(value):  
        return isinstance(value, int) or isinstance(value, float)
```

```
In [9]: 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 [10]: 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 [11]: def gini(rows):  
        counts = class_counts(rows)
```

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        impurity = 1
        for lbl in counts:
            prob_of_lbl = counts[lbl] / float(len(rows))
            impurity -= prob_of_lbl**2
        return impurity

In [12]: def info_gain(left, right, current_uncertainty):
        p = float(len(left)) / (len(left) + len(right))
        return current_uncertainty - p * gini(left) - (1 - p) * gini(right)

In [13]: class Leaf:
        def __init__(self, rows):
            self.predictions = class_counts(rows)

In [14]: 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 [15]: 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 [16]: def partition(rows, question):
        trueRows = []
        falseRows = []

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

return trueRows, falseRows

```

```

In [17]: def findBestSplit(rows, questions):
    best_gain = 0
    best_question = None
    current_uncertainty = gini(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)

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

    return best_gain, best_question

```

```

In [18]: def formTree(rows, questions):
    gain, question = findBestSplit(rows, questions)

    if gain == 0:
        return Leaf(rows)

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

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

    return Decision_Node(question, trueBranch, falseBranch)

```

```

In [19]: 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)

```

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In [20]: def train(data, questions):
          return formTree(data, deepcopy(questions))

In [21]: def classify(root, rows):
          predictions = []
          for r in rows:
              predictions.append(max(classifyRow(root, r).items(), key=operator.itemgetter(0)))
          return predictions

In [22]: def getDataInIndex(data, index):
          l = []
          for i in range(len(data)):
              if i in index:
                  l.append(data[i])
          return l

In [23]: 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)

              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