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# 4 Question Classification

Write a python script to take file input which will contain a sentence per line. For each sentence you have to identify if it's a question or not by detecting question phrases like what, how etc. and tag it yes or no.

Write Below link tagged data into another output file is to a file containing sample sentences. You can test your script on this:https://drive.google.com/file/d/1LAB9zMwDRZgTEVubTcp\_Ctpa2ErKZ3Pj/view?usp=sharing

Format of input file:- Hello how are you doing

My name is Joe

I am going home

Format of output file:- Hello how are you doing Yes

My name is Joe No

I am going home No

- Download nps\_chat from nltk which includes the chat details available in nltk library.
- Download punkt from nltk which helps in tokenization of text into sentences.

#### 4.1 Importing the Libraries required

```
[1]: import pandas as pd import nltk import matplotlib.pyplot as plt
```

```
[]: # run only if not downloaded
nltk.download('nps_chat')
nltk.download('punkt')
```

#### 4.1.1 Reading all the data from the file.

```
[2]: filepath = 'data.txt'
      text = []
      with open(filepath) as fp:
          line = fp.readline()
          while line:
              text.append(line.replace('\n',''))
              line = fp.readline()
 [3]: text[:10]
 [3]: ['hindi movies for adults',
       'are panda dogs real',
       'asuedraw winning numbers',
       'sentry replacement keys',
       'rebuilding nicad battery packs',
       'can you overdose on melatonin',
       'list of browsers',
       'que son numeros naturales',
       'mini motos honda',
       'mulatto']
     4.1.2 Extarcting data from the posts.
 [4]: posts = nltk.corpus.nps_chat.xml_posts()
 [5]: # we create a function which extracts features from the posts
      def dialogue_features(post):
          # a dictionary which will take all the details of the post
          features = {}
          # we tokenize the content of the post
          for word in nltk.word_tokenize(post):
              features['contains({})'.format(word.lower())] = True
          # return the dictionary with the data
          return features
 [6]: # we create a list of tuple with the text and the class of the text
      featureSets = [(dialogue_features(post.text), post.get('class')) for post in_
       →posts]
[43]: # check number of posts
      len(featureSets)
[43]: 10567
 [7]: # we make take 20% of the data for the testing purpose and 80% for training
      size = int(len(featureSets) * 0.2)
```

```
train_set, test_set = featureSets[size:], featureSets[:size]
```

# 4.1.3 Testing Maxent and Naive Bayes Algorithm

# Maxent Classifier

- [8]: # we create a Maxent Classifier which trains the model for 15 iterations classifier1 = nltk.MaxentClassifier.train(train\_set, max\_iter=15)
  - ==> Training (15 iterations)

Iteration	Log Likelihood	Accuracy
1	-2.70805	0.051
2	-1.24488	0.857
3	-0.91412	0.891
4	-0.74260	0.905
5	-0.62950	0.917
6	-0.54642	0.924
7	-0.48235	0.927
8	-0.43205	0.933
9	-0.39230	0.937
10	-0.36064	0.941
11	-0.33505	0.945
12	-0.31394	0.949
13	-0.29617	0.952
14	-0.28095	0.954
Final	-0.26773	0.955

[9]: # we print the accuracy attainded by the model on test data print(nltk.classify.accuracy(classifier1, test\_set))

#### 0.7179365830572646

# Naive Bayes Classifier

- [10]: # we create a Naive Bayes classifer from the training set, where we pass text

  → and label

  classifier2 = nltk.NaiveBayesClassifier.train(train\_set)
- [11]: # we print the accuracy achieved by Naive bayes on the test data print(nltk.classify.accuracy(classifier2, test\_set))
  - 0.6630383341221012

#### Decision Tree Classifier

[12]: # we create a Decision Tree classifer from the training set, where we pass text

→ and label

classifier3 = nltk.classify.decisiontree.DecisionTreeClassifier.train(train\_set)

```
[13]: # we print the accuracy achieved by Decision Tree on the test data print(nltk.classify.accuracy(classifier3, test_set))
```

0.7666824420255561

#### 4.1.4 Providing selected question types.

The function will return true if NLTK model classifies the given text into whquestion or ynquestion category, else will return false.

```
[14]: # we specify the types of text which is required for our problem

questionTypes = ["whQuestion", "ynQuestion"]

# we create a function of text classification using nltk model

def isQuestionNltk(ques):

# we extract the text part and classify the data using the model

questionType = classifier3.classify(dialogue_features(ques))

# if the label or category returned by model is a question then we return

→ true, else false

return questionType in questionTypes
```

# 4.1.5 Adding an extra layer of questions for attaining better accuracy.

4.1.6 It will take care of the cases which are missed by the NLTK model.

```
[16]: # hepling verbs which works in the supporting role.
helpingVerbs = ["is", "am", "can", "are", "do", "does"]
```

#### 4.1.7 Creating a final function, which will classify that the text is a question or not

```
[17]: # creating a function which will take the text and tell that it is a question
or not

def isQuestion(data):
    # we convert the data to lowercase and remove leading and trailing spaces
    data = data.lower().strip()
    # we classify the data using the NLTK model, if it is classified as a
    →question by nltk we
    # return True else we go for in depth check of the text
    if not isQuestionNltk(data):
        isQues = False
    # check if any of pattern exist in sentence
```

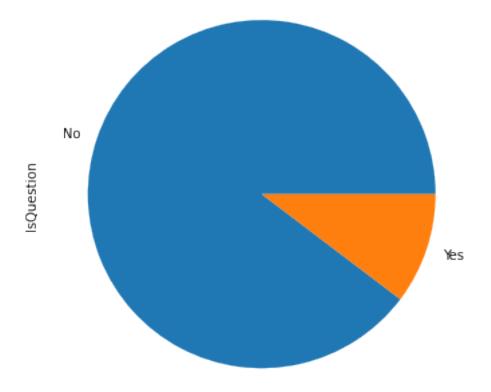
```
for pattern in questionPattern:
            # check that each pattern is their in the question or not.
            isQues = pattern in data
            # if pattern is found in question, change the value to true and
\rightarrow break
            if isQues:
                break
       \# We divide multiple sentences using . as a delimeter
       sentences = data.split(".")
       # we iteratre through all the sentences
       for sentence in sentences:
            # we check that the sentence is not empty
            if len(sentence.strip()):
                # we tokenize the the sentence using the nltk and extarct the
\hookrightarrow first work
                firstWord = nltk.word_tokenize(sentence)[0]
                # if question ends with ? or start with any helping verb, then_
\rightarrow it is question
                if sentence.endswith("?") or firstWord in helpingVerbs:
                    isQues = True
                    # we then break from the loop
                    break
       # we return the value true or false which is obtained by in depth search
       return isQues
   else:
       # we return true as it has been classified as a question by nltk \ model_{\sqcup}
\hookrightarrow only
       return True
```

# 4.1.8 Testing with some values

```
[18]: isQuestion('am i male')
[18]: True
[19]: isQuestion('How are You')
[19]: True
[20]: isQuestion('I am Fine')
[20]: False
```

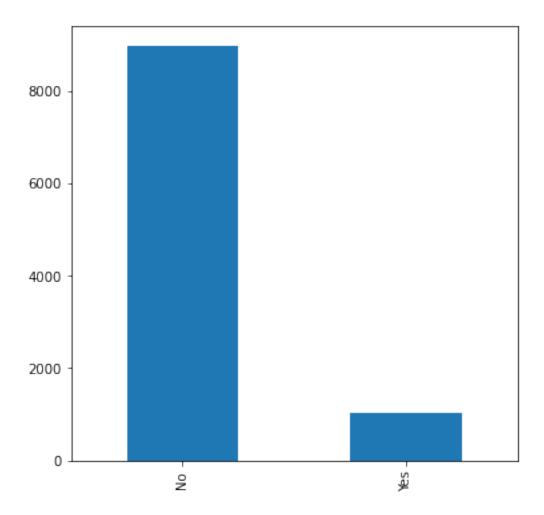
# 4.1.9 Classifying all the text

```
[21]: # we create a dataframe which will have all the text and label
      data = pd.DataFrame({'Sentence':text})
[32]: # we create an anonymous function using isQuestion function which will return
      → Yes or no according to the text
      data['IsQuestion'] = data['Sentence'].apply(lambda x: 'Yes' if isQuestion(x)
       →else 'No')
[33]: data.head()
[33]:
                               Sentence IsQuestion
                hindi movies for adults
     0
                                                No
      1
                    are panda dogs real
                                               Yes
      2
               asuedraw winning numbers
                                                No
                sentry replacement keys
                                                No
      4 rebuilding nicad battery packs
                                                No
[34]: data.shape
[34]: (9993, 2)
[35]: # we find out the distribution of the labels
      data.IsQuestion.value_counts()
[35]: No
             8962
     Yes
             1031
      Name: IsQuestion, dtype: int64
[38]: # we visualize the results
      plt.figure(figsize=(6,6))
      data.IsQuestion.value_counts().plot(kind='pie')
[38]: <AxesSubplot:ylabel='IsQuestion'>
```



```
[39]: # we visualize the results
plt.figure(figsize=(6,6))
data.IsQuestion.value_counts().plot(kind='bar')
```

[39]: <AxesSubplot:>



```
[41]: # we see the dataframe with the text and label data.head()
```

```
[41]:
                                Sentence IsQuestion
                hindi movies for adults
      0
                                                  No
                     are panda dogs real
      1
                                                 Yes
               asuedraw winning numbers
      2
                                                  No
      3
                sentry replacement keys
                                                  No
         rebuilding nicad battery packs
                                                  No
```

```
[42]: # we save the data to a csv file which will contain the sentence and answer⊔

→ that it is a question or not.

data.to_csv('Solution.csv')
```

#### 4.2 Inference

We have created a model which will classify that the given text is a question or not. We use NLTK Library for creating the model. We use the NPS chat data for training as it contains the

text classified into proper labels. We extract the text and labels from the posts. We use different algorithms such as Decision Tree, Naive Bayes, and Maxent and identified that Decision Tree gives us the best accuracy. We attained an accuracy of around 76% using the Decision Tree Model. We use the Decision Tree classfier of nltk and created a function which will predict that the text is a question or not. We optimized the function by adding some use cases which are not covered by the algorithm to improve the performance. In the end, we applied the function to all the rows using lamda and apply functions and saved the result in a dataframe. We visualized the results and saved it in an csv file for future reference.