Project

Text Classification

```
In [171...
          from sklearn.datasets import fetch_20newsgroups
          import string
          from nltk.corpus import stopwords
          from nltk.tokenize import word_tokenize
          from sklearn.model_selection import train_test_split
          import numpy as np
          from sklearn.naive_bayes import MultinomialNB
In [172...
          #fetching the dataset
          #instead of using the data from the website i used the inbuilt dataset which is same as
          newsgroups=fetch_20newsgroups()
          #importing stopwords and punctuations
          stops=set(stopwords.words('english'))
          punctuations=list(string.punctuation)
          stops.update(punctuations)
In [173...
          len(newsgroups.data)
         11314
Out[173...
In [174...
          all_documents=newsgroups.data
          all_categories=newsgroups.target
          #dividing into words as we have to work with words not with sentences
          all_documents_modified=[word_tokenize(doc) for doc in all_documents]
In [175...
          #splitting into training and testing
          x_train, x_test, y_train, y_test=train_test_split(all_documents_modified, all_categories
In [176...
          all_words=[]
          #this list is going to contain all the words from all our tokenized documents which we ec{v}
          #and unnecessary stopswords and punctuations are removed as they dont make sense
          for doc in x_train:
              for word in doc:
                  # removing unncecessary words
                  if (word.lower() not in stops) and len(word)!=1 and len(word)!=2 and word[0]!="
                      #appending necessary words or features
                      all_words.append(word)
In [177...
          # this functions returns the frequency of all the words from all_words which we will use
          def freq_dict(all_words):
              dic=dict()
              #it iterates through all the elements in the list and increases the frequency by one
              for word in all_words:
                  if word in dic.keys():
                      dic[word]+=1
                  else:
                      dic[word]=1
```

```
dic=freq_dict(all_words)
In [178...
          #diving the freq and words into two lists and sorting them into deccreasing order of fre
          #to get the maximum frequency words.
          import numpy as np
          freq=np.array([i for i in dic.values()])
          words=np.array([i for i in dic.keys()])
          words=words[np.argsort(freq)][::-1]
In [179...
          for i in range(50):
              print(words[i])
         Subject
         Lines
         Organization
         would
         writes
         one
         article
         people
         like
         University
         know
         get
         think
         use
         time
         also
         MAX
         could
         good
         may
         way
         even
         much
         make
         see
         two
         say
         Distribution
         God
         many
         right
         new
         want
         Nntp-Posting-Host
         said
         first
         used
         NNTP-Posting-Host
         system
         work
         need
         something
         well
         world
         anyone
         problem
```

return dic

going

```
believe
In [180...
          # taking only the releavent words as features since the most common are useless so that
          # upto 10,000 top words
          features=words[20:10000]
In [181...
          # It takes the patameters x_train or x_test and the list of all features and converts it
          # in that particular document, where rows are the documents and columns are the features
          def data_modifier(x_data, features):
              modified_data=np.zeros((len(x_data), len(features)))
              #creating the empty 2d array
              for i in range(len(x_data)):
                  #looping over each and every row in the x_data
                  current_doc=x_data[i]
                  #current_doc contains the current document on which we are iterating.(As the nam
                  d=dict()
                  #this dictionary contains the frequency of all the elements in our current_doc.
                  for word in current_doc:
                      if word in d.keys():
                          d[word] += 1
                      else:
                          d[word]=1
                  #dictionary created
                  for j in range(len(features)):
                      #now for each feature in features we will insert the value of the dictionary
                      #the frequency of each feature in that current document.
                      if features[j] in d.keys():
                          modified_data[i][j]=d[features[j]]
                      else:
                          continue
              #finally I have returned the modified array.
              return modified_data
In [182...
          len(x_train)
         8485
Out[182...
In [183...
          #converting the training data into 2d form
          x_train_modified = data_modifier(x_train, features)
In [184...
          #converting the testing data into 2d form
          x_test_modified= data_modifier(x_test, features)
```

Inbuilt Naive Bayes

still really

```
#first trying out the inbuilt Multinomial naive bayes classifier.

clf=MultinomialNB()

clf.fit(x_train_modified, y_train)

print(clf.score(x_test_modified, y_test)*100, "%")
```

Building our own Naive Bayes classifier from scratch

```
In [187...
          #this function takes our xtrain and ytrain and combine them into a dictionary with featl
          #in them and then returns a dictionary
          def fit(x_train , y_train):
              d = \{\}
              #defining a dictionary
              for i in range(20):
                  docs = x_train[y_train == i]
                  #taking the classes one by one from x_train
                  #making a dictionary on the ith class to save the features and their total value
                  d[i]['total'] = 0
                  #this holds the value of the total words present in the class to be used in the
                  for j in range(len(features)):
                      d[i][features[j]] = docs[:, j].sum()
                      #how many times jth feature is coming corresponding to class i
                      d[i]['total']+=d[i][features[j]]
                      #stores the sum of all the values of ith key
              return d
In [188...
          # finding probabilty of each word in document for the current class
          def probability(dictionary , x , current_class):
              prob\_word = []
              #it will save all the probabs
              for i in range(len(x)):
                  if x[i]!=0:
                      #we dont want to consider words which are not present
                      num = dictionary[current_class][features[i]]
                      #finding numerator
                      denom = dictionary[current_class]['total']
                      #finding denominator
                      prob = np.log((num + 1)/(denom + len(x)))
                      #finding probability with laplace correction
                      prob_word.append(prob)
                      # appending in the list
              return sum(prob_word)
In [189...
          #finding the best class using the above function by comparing all the probabilities
          def predictSinglePoint(dictionary, x):
              classes = dictionary.keys()
              # finding all classes
              bestp = -20
              #taking best probability negative
              bestc = -20
              #taking the best class negative
              firstrun = True
              #firstrun is created to update with the first probability no matter the case so nega
              for clas in classes:
                  #iterating through each class
                  prob_class = probability(dictionary, x, clas)
                  #finding the probab of current class using the probabilty function as given above
                  if(firstrun == True or bestp < prob_class):</pre>
                      #updating the values in our variables to get the maximum probab class
```

bestp = prob_class

```
#making firstrun as false as we dont want to use it anymore
              return bestc
In [190...
          #this function return the predicted classes by using the above functions
          def predict(x_test, dictionary):
              y_pred = []
              #creating the empty list for predicted values
              for doc in x_test:
                  #iterating through every doc and predicting values and appending to the predict
                  y_pred.append(predictSinglePoint(dictionary , doc))
              return y_pred
In [191...
          #dictionary created through fit function contains classes and their features list
          dictionary=fit(x_train_modified, y_train)
In [192...
          #example of class 2 in dictionary which contains 20 classes
          dictionary[2]
         {'total': 19433.0,
Out[192...
           'way': 66.0,
           'even': 51.0,
           'much': 71.0,
           'make': 80.0,
           'see': 49.0,
           'two': 42.0,
           'say': 34.0,
           'Distribution': 101.0,
           'God': 0.0,
           'many': 47.0,
           'right': 53.0,
           'new': 79.0,
           'want': 69.0,
           'Nntp-Posting-Host': 105.0,
           'said': 25.0,
           'first': 48.0,
           'used': 74.0,
           'NNTP-Posting-Host': 77.0,
           'system': 87.0,
           'work': 96.0,
           'need': 88.0,
           'something': 76.0,
           'well': 79.0,
           'world': 41.0,
           'anyone': 105.0,
           'problem': 163.0,
           'going': 36.0,
           'still': 47.0,
           'really': 51.0,
           'believe': 25.0,
           'back': 36.0,
           'years': 12.0,
           'must': 22.0,
           'find': 64.0,
           'year': 15.0,
```

bestc = clas
firstrun = False

'using': 168.0,

```
'point': 24.0,
'take': 28.0,
'better': 40.0,
'things': 33.0,
'Reply-To': 79.0,
'information': 56.0,
'might': 27.0,
'file': 232.0,
'program': 127.0,
'last': 24.0,
'question': 33.0,
'got': 53.0,
'government': 0.0,
'never': 36.0,
'help': 87.0,
'made': 17.0,
'available': 71.0,
'sure': 49.0,
'since': 43.0,
'number': 36.0,
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'without': 43.0,
'New': 16.0,
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'someone': 39.0,
'another': 30.0,
'read': 43.0,
'David': 29.0,
'Computer': 44.0,
'little': 24.0,
'come': 20.0,
'etc': 48.0,
'version': 125.0,
'give': 29.0,
'part': 28.0,
'John': 21.0,
'around': 36.0,
'case': 17.0,
'fact': 12.0,
'drive': 44.0,
'different': 29.0,
'anything': 19.0,
'long': 12.0,
'course': 9.0,
'1993': 20.0,
'least': 26.0,
'set': 99.0,
'says': 21.0,
'data': 26.0,
'look': 26.0,
'power': 2.0,
'best': 30.0,
'lot': 27.0,
'probably': 30.0,
'tell': 41.0,
'day': 7.0,
'possible': 38.0,
'enough': 14.0,
'seems': 42.0,
'car': 1.0,
'every': 24.0,
'put': 18.0,
'true': 24.0,
'name': 31.0,
```

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'key': 19.0,
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'Jesus': 0.0,
'far': 22.0,
'please': 44.0,
'law': 0.0,
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'try': 56.0,
'Q,3': 821.0,
'card': 170.0,
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'line': 30.0,
'files': 212.0,
'Windows': 501.0,
'else': 32.0,
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'hard': 34.0,
'team': 0.0,
'let': 24.0,
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'called': 45.0,
'problems': 73.0,
'great': 23.0,
'Well': 22.0,
'support': 45.0,
'mean': 20.0,
'life': 6.0,
'bit': 22.0,
'example': 15.0,
'wrong': 23.0,
'rather': 27.0,
'reason': 13.0,
'found': 38.0,
'done': 12.0,
'person': 9.0,
'keep': 15.0,
'send': 23.0,
'Inc.': 39.0,
'Please': 42.0,
'old': 19.0,
'Center': 23.0,
'USA': 34.0,
'thought': 20.0,
'nothing': 15.0,
'software': 65.0,
'post': 30.0,
'end': 22.0,
'able': 31.0,
'One': 9.0,
'message': 37.0,
'Keywords': 50.0,
'real': 21.0,
'order': 20.0,
'next': 22.0,
'always': 26.0,
'looking': 41.0,
'public': 12.0,
'means': 21.0,
'bad': 17.0,
'place': 21.0,
'less': 22.0,
'seen': 47.0,
'others': 5.0,
'state': 4.0,
```

```
'group': 30.0,
'State': 27.0,
'trying': 18.0,
'actually': 21.0,
'Science': 18.0,
'Mark': 8.0,
'following': 17.0,
'Israel': 0.0,
'away': 12.0,
'quite': 26.0,
'free': 33.0,
'Research': 28.0,
'wrote': 30.0,
'high': 24.0,
'window': 55.0,
'several': 30.0,
'ever': 11.0,
'heard': 22.0,
'left': 12.0,
'second': 15.0,
'already': 16.0,
'Also': 30.0,
'play': 15.0,
'start': 25.0,
'call': 19.0,
'opinions': 27.0,
'evidence': 0.0,
'However': 21.0,
'getting': 18.0,
'idea': 15.0,
'control': 33.0,
'Jews': 0.0,
'kind': 14.0,
'man': 3.0,
'seem': 22.0,
'netcom.com': 8.0,
'makes': 13.0,
'info': 30.0,
'three': 10.0,
'money': 8.0,
'space': 38.0,
'chip': 13.0,
'current': 23.0,
'human': 1.0,
'based': 21.0,
'Steve': 21.0,
'American': 1.0,
'Christian': 0.0,
'given': 4.0,
'Apr': 17.0,
'ago': 23.0,
'games': 9.0,
'times': 13.0,
'today': 7.0,
'whether': 34.0,
'change': 52.0,
'small': 41.0,
'Michael': 21.0,
'yet': 21.0,
'came': 17.0,
'code': 22.0,
'encryption': 0.0,
'local': 14.0,
'book': 8.0,
```

```
'email': 33.0,
'source': 14.0,
'April': 9.0,
'Internet': 46.0,
'answer': 15.0,
'interested': 16.0,
'usa': 48.0,
'Institute': 16.0,
'running': 71.0,
'told': 7.0,
'ask': 8.0,
'saying': 3.0,
'standard': 21.0,
'home': 14.0,
'B8F': 543.0,
'gun': 0.0,
'large': 26.0,
'Technology': 33.0,
'whole': 9.0,
'mail': 38.0,
'Bill': 5.0,
'National': 19.0,
'questions': 24.0,
'issue': 8.0,
'children': 3.0,
'buy': 28.0,
'Paul': 1.0,
'important': 9.0,
'disk': 71.0,
'Department': 31.0,
'works': 35.0,
'matter': 4.0,
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'speed': 22.0,
'Canada': 19.0,
'address': 24.0,
'Robert': 25.0,
'President': 3.0,
'show': 8.0,
'days': 9.0,
'live': 3.0,
'machine': 55.0,
'Article-I.D': 26.0,
'agree': 2.0,
'pretty': 17.0,
'stuff': 24.0,
'server': 16.0,
'Systems': 48.0,
'feel': 11.0,
'big': 13.0,
'word': 10.0,
'access': 61.0,
'went': 9.0,
'Mike': 26.0,
'comes': 28.0,
'memory': 69.0,
'side': 4.0,
'claim': 5.0,
'including': 12.0,
'California': 12.0,
'computer': 53.0,
'Mac': 14.0,
'general': 11.0,
'DOS': 157.0,
```

```
'package': 30.0,
'rights': 0.0,
'started': 12.0,
'working': 14.0,
'provide': 12.0,
'price': 17.0,
'Bible': 0.0,
'include': 11.0,
'programs': 54.0,
'understand': 4.0,
'simply': 8.0,
'often': 7.0,
'output': 15.0,
'X-Newsreader': 27.0,
'Jim': 2.0,
'Yes': 17.0,
'everything': 21.0,
'remember': 29.0,
'1.1': 29.0,
'care': 2.0,
'Armenian': 0.0,
'SCSI': 2.0,
'systems': 15.0,
'Turkish': 0.0,
'Space': 0.0,
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'making': 1.0,
'hope': 15.0,
'tried': 45.0,
'maybe': 19.0,
'Christians': 0.0,
'similar': 14.0,
'cost': 8.0,
'phone': 14.0,
'San': 11.0,
'Israeli': 0.0,
'couple': 20.0,
'country': 0.0,
'full': 22.0,
'hand': 6.0,
'e-mail': 26.0,
'took': 10.0,
'known': 3.0,
'image': 14.0,
'area': 6.0,
'mind': 2.0,
'argument': 1.0,
'consider': 5.0,
'difference': 18.0,
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'certain': 11.0,
'likely': 10.0,
'news': 18.0,
'Armenians': 0.0,
'almost': 11.0,
'later': 11.0,
'within': 15.0,
'however': 12.0,
'Christ': 0.0,
'season': 0.0,
'size': 58.0,
'York': 3.0,
'guess': 9.0,
'men': 0.0,
```

```
'write': 19.0,
'sort': 8.0,
'type': 16.0,
'per': 6.0,
'Dept': 22.0,
'A86': 413.0,
'words': 1.0,
'numbers': 10.0,
'talking': 1.0,
'cause': 5.0,
'145': 388.0,
'faith': 0.0,
'pay': 6.0,
'love': 14.0,
'religion': 0.0,
'truth': 0.0,
'color': 21.0,
'Mr.': 2.0,
'usually': 17.0,
'FAQ': 8.0,
'IBM': 20.0,
'deal': 7.0,
'College': 24.0,
'DoD': 0.0,
'single': 8.0,
'GMT': 14.0,
'Pittsburgh': 0.0,
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'sense': 5.0,
'wanted': 7.0,
'display': 25.0,
'experience': 15.0,
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'saw': 4.0,
'particular': 14.0,
'subject': 7.0,
'keys': 1.0,
'opinion': 3.0,
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'Clinton': 3.0,
'certainly': 8.0,
'Washington': 10.0,
'death': 0.0,
'100': 11.0,
'correct': 18.0,
'Engineering': 28.0,
'win': 18.0,
'Sun': 7.0,
'via': 17.0,
'goes': 16.0,
'uses': 30.0,
'groups': 19.0,
'body': 2.0,
'video': 49.0,
'points': 1.0,
'reading': 23.0,
'entry': 12.0,
'advance': 39.0,
'U.S.': 2.0,
'taken': 3.0,
'killed': 0.0,
'bike': 1.0,
'nice': 19.0,
```

```
'driver': 146.0,
'Dave': 18.0,
'screen': 58.0,
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'Software': 20.0,
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'top': 21.0,
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'become': 1.0,
'written': 15.0,
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'May': 6.0,
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'longer': 8.0,
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'Chicago': 10.0,
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'level': 9.0,
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'Brian': 17.0,
'head': 6.0,
'States': 0.0,
'mine': 15.0,
'view': 6.0,
'behind': 3.0,
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'business': 4.0,
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'sell': 5.0,
'drives': 16.0,
'major': 3.0,
'Smith': 5.0,
'week': 6.0,
'1D9': 327.0,
'interesting': 3.0,
'dead': 3.0,
'due': 11.0,
'board': 17.0,
'weapons': 0.0,
'accept': 8.0,
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'allow': 27.0,
'previous': 16.0,
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'contact': 8.0,
'company': 7.0,
'whatever': 7.0,
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'TIN': 17.0,
'mode': 52.0,
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'sound': 19.0,
'stop': 3.0,
'section': 15.0,
'Peter': 19.0,
```

```
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'move': 18.0,
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'night': 1.0,
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'short': 12.0,
'coming': 4.0,
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'moral': 1.0,
'early': 4.0,
'test': 5.0,
'parts': 5.0,
'future': 4.0,
'rules': 3.0,
'mentioned': 8.0,
'File': 17.0,
'expect': 18.0,
'upon': 2.0,
'crime': 0.0,
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'Division': 16.0,
'books': 11.0,
'cases': 10.0,
'PL+': 309.0,
'technology': 2.0,
'study': 2.0,
'directory': 48.0,
'Good': 9.0,
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'taking': 5.0,
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'offer': 9.0,
'necessary': 6.0,
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'force': 3.0,
'guy': 2.0,
'Note': 12.0,
'sent': 9.0,
'soon': 15.0,
'special': 10.0,
'includes': 9.0,
'result': 7.0,
'thinking': 5.0,
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```
. . . }
In [193...
          #predicted values from the predict function
          y_predicted=predict(x_test_modified, dictionary)
In [194...
          #comparing our predcited values with the y_test
          from sklearn.metrics import classification_report
          print(classification_report(y_true=y_test, y_pred=y_predicted))
                        precision
                                      recall f1-score
                                                          support
                     0
                             0.87
                                        0.89
                                                  0.88
                                                              114
                              0.64
                     1
                                        0.82
                                                  0.72
                                                              152
                     2
                             0.95
                                        0.67
                                                  0.78
                                                              139
                     3
                             0.61
                                        0.82
                                                  0.70
                                                              152
                     4
                             0.76
                                        0.88
                                                  0.82
                                                              138
                     5
                             0.85
                                        0.83
                                                  0.84
                                                              153
                             0.83
                                                              147
                     6
                                        0.69
                                                  0.75
                     7
                             0.81
                                        0.91
                                                  0.85
                                                              137
                     8
                             0.95
                                                  0.93
                                        0.90
                                                              131
                     9
                             0.91
                                        0.94
                                                  0.92
                                                              135
                             0.98
                    10
                                        0.95
                                                  0.97
                                                              136
                             0.93
                                                  0.95
                                                              145
                    11
                                        0.97
                    12
                             0.91
                                        0.70
                                                  0.79
                                                              157
                    13
                             0.98
                                                  0.96
                                                              151
                                        0.94
```

0.92

0.88

0.90

0.94

0.86

0.69

0.85

0.85

0.85

155

159

140

149 138

101

2829

2829

2829

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14

15

16

17

18

19

accuracy

macro avg

weighted avg

0.96

0.84

0.88

0.96

0.87

0.80

0.87

0.87

The result of our Mutlinomial Naive Bayes is almost same as that of our inbuilt naive bayes

0.88

0.94

0.91

0.91

0.85

0.60

0.85

0.85

```
In [195... np.savetxt('predicted_data.csv',y_predicted, delimiter=',')

In []:
```