```
In [173]: from tqdm import tqdm # tqdm is a library that helps us to visualize the runtime
          #https://tqdm.github.io/
          # it accepts only list of sentances
          def fit(dataset):
              unique_words = set() # at first we will initialize an empty set
              # check if its list type or not
              if isinstance(dataset, (list,)):
                   for row in dataset: # for each review in the dataset
                      for word in row.split(" "): # for each word in the review. #split met
                          if len(word) < 2:</pre>
                               continue
                          unique_words.add(word)
                  unique_words = sorted(list(unique_words))
                  vocab = {j:i for i,j in enumerate(unique words)}
                  #vocab = dict.fromkeys(unique_words, 0)
                  return vocab
              else:
                  print("you need to pass list of sentance")
```

```
In [174]: unique_words=[]
vocab = fit(corpus)
print(vocab)
for i in vocab.keys():
    unique_words.append(i)
print(unique_words)

{'and': 0, 'document': 1, 'first': 2, 'is': 3, 'one': 4, 'second': 5, 'the': 6,
```

```
{'and': 0, 'document': 1, 'first': 2, 'is': 3, 'one': 4, 'second': 5, 'the': 6
'third': 7, 'this': 8}
['and', 'document', 'first', 'is', 'one', 'second', 'the', 'third', 'this']
```

```
In [179]: #Returs a sparse matrix of the all non-zero values along with their row and col
          import math
          from collections import Counter
          from tqdm import tqdm
          from scipy.sparse import csr_matrix
          import math
          import operator
          from sklearn.preprocessing import normalize
          import numpy
          def transform(dataset, vocab):
                  row = []
                  col = []
                  values = []
                  for ibx,document in enumerate(dataset):
                       #print(dataset)
                      word_freq = dict(Counter(document.split()))
                       print(word_freq)
                       for word, freq in word freq.items():
                           col index = vocab.get(word,-1)
                           if col_index != -1:
                               if len(word)<2:</pre>
                                   continue
                           col.append(col_index)
                           row.append(ibx)
                           tf=freq/float(len(word freq))
                           #print(word, tf)
                           #td = freq/float(len(document)) # the number of times a word occi
                           idf=IDF1(corpus,unique words,word)
                           #print(word, freq, td, idf)
                           values.append((tf)*(idf))
                   print(len(values),len(row),len(col),len(vocab),len(dataset))
                  return normalize(csr matrix( ((values),(row,col)), shape=(len(dataset),le
          print(transform(corpus, vocab))
              (0, 8) 0.38408524091481483
              (0, 6) 0.38408524091481483
              (0, 3) 0.38408524091481483
              (0, 2) 0.5802858236844359
              (0, 1) 0.46979138557992045
          {'this': 1, 'is': 1, 'the': 1, 'first': 1, 'document': 1}
          {'this': 1, 'is': 1, 'the': 1, 'second': 1, 'document': 1}
          {'and': 1, 'this': 1, 'is': 1, 'the': 1, 'third': 1, 'one': 1}
          {'is': 1, 'this': 1, 'the': 1, 'first': 1, 'document': 1}
          21 21 21 9 4
            (0, 1)
                           0.4697913855799205
            (0, 2)
                          0.580285823684436
            (0, 3)
                          0.3840852409148149
            (0, 6)
                          0.3840852409148149
            (0, 8)
                          0.3840852409148149
            (1, 1)
                          0.4279695901493821
            (1, 3)
                          0.34989318276628206
            (1, 5)
                           0.6704970632809761
```

```
(1, 6)
              0.34989318276628206
              0.34989318276628206
(1, 8)
(2, 0)
              0.511848512707169
(2, 3)
              0.267103787642168
(2, 4)
              0.511848512707169
(2, 6)
              0.267103787642168
(2, 7)
              0.511848512707169
(2, 8)
              0.267103787642168
(3, 1)
              0.4697913855799205
(3, 2)
              0.580285823684436
(3, 3)
              0.3840852409148149
(3, 6)
              0.3840852409148149
(3, 8)
              0.3840852409148149
```

```
In [ ]:
```