## Implementing Bag of Words

#### Fit method:

- 1. With this function, we will find all unique words in the data and we will assign a dimension-number to each unique word.
- 2. We will create a python dictionary to save all the unique words, such that the key of dictionary represents a unique word and the corresponding value represent it's dimension-number.
- 3. For example, if you have a review, \_\_\_'very bad pizza'\_\_\_ then you can represent each unique word with a dimension\_number as,

```
dict = { 'very' : 1, 'bad' : 2, 'pizza' : 3}
```

In [3]:

```
import warnings
warnings.filterwarnings("ignore")
import pandas as pd
from tqdm import tqdm
import os
```

In [2]:

```
from tqdm import tqdm # tqdm is a library that helps us to visualize the runtime of for loop. refe
r this to know more about tqdm
#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 method converts a sti
ing into list of words
                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)}
       return vocab
   else:
        print("you need to pass list of sentance")
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                                                                                                 F
```

In [4]:

```
vocab = fit(["abc def aaa prq", "lmn pqr aaaaaaa aaa abbb baaa"])
print(vocab)

{'aaa': 0, 'aaaaaaaa': 1, 'abbb': 2, 'abc': 3, 'baaa': 4, 'def': 5, 'lmn': 6, 'pqr': 7, 'prq': 8}
```

### What is a Sparse Matrix?

- 1. Before going further into details about Transform method, we will understand what sparse matrix is.
- 2. Sparse matrix stores only non-zero elements and they occupy less amount of RAM comapre to a dense matrix. You can refer to this <a href="link">link</a>.
- 3. For example, assume you have a matrix,

```
[[1, 0, 0, 0, 0],
[0, 0, 0, 1, 0],
```

```
[0, 0, 4, 0, 0]]
```

In [0]:

```
from sys import getsizeof
import numpy as np
# we store every element here
a = np.array([[1, 0, 0, 0, 0], [0, 0, 0, 1, 0], [0, 0, 4, 0, 0]])
print(getsizeof(a))

# here we are storing only non zero elements here (row, col, value)
a = [ (0, 0, 1), (1, 3, 1), (2,2,4)]
# with this way of storing we are saving alomost 50% memory for this example
print(getsizeof(a))
```

172 88

#### How to write a Sparse Matrix?:

- 1. You can use csr\_matrix() method of scipy.sparse to write a sparse matrix.
- 2. You need to pass indices of non-zero elements into csr\_matrix() for creating a sparse matrix.
- 3. You also need to pass element value of each pair of indices.
- 4. You can use lists to save the indices of non-zero elements and their corresponding element values.
- 5. For example,
  - Assume you have a matrix,

```
[[1, 0, 0], [0, 0, 1], [4, 0, 6]]
```

- Then you can save the indices using a list as, list\_of\_indices = [(0,0), (1,2), (2,0), (2,2)]
- And you can save the corresponding element values as, **element\_values** = [1, 1, 4, 6]
- 6. Further you can refer to the documentation here.

#### **Transform method:**

1. With this function, we will write a feature matrix using sprase matrix.

```
In [5]:
```

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def transform(dataset, vocab):

```
from collections import Counter
from scipy.sparse import csr_matrix
test = 'abc def abc def zzz zzz pqr'
a = dict(Counter(test.split()))
for i,j in a.items():
    print(i, j)

abc 2
def 2
zzz 2
pqr 1

In [10]:
# https://stackoverflow.com/questions/9919604/efficiently-calculate-word-frequency-in-a-string
# https://docs.scipy.org/doc/scipy-0.19.0/reference/generated/scipy.sparse.csr_matrix.html
# note that we are we need to send the preprocessing text here, we have not inlouded the processin
```

```
rows = []
    columns = []
    values = []
    if isinstance(dataset, (list,)):
        for idx, row in enumerate(tqdm(dataset)): # for each document in the dataset
            # it will return a dict type object where key is the word and values is its frequency,
{word:frequency}
            word freq = dict(Counter(row.split()))
            # for every unique word in the document
            for word, freq in word_freq.items(): # for each unique word in the review.
                if len(word) < 2:</pre>
                    continue
                # we will check if its there in the vocabulary that we build in fit() function
                # dict.get() function will return the values, if the key doesn't exits it will retu
rn -1
                    col index = vocab.get(word, -1) # retreving the dimension number of a word
                # if the word exists
                print(col index)
                if col index !=-1:
                    # we are storing the index of the document
                    rows.append(idx)
                    # we are storing the dimensions of the word
                    columns.append(col index)
                    # we are storing the frequency of the word
                    values.append(freq)
        return csr_matrix((values, (rows,columns)), shape=(len(dataset),len(vocab)))
    else:
        print("you need to pass list of strings")
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In [11]:
strings = ["the method of lagrange multipliers is the economists workhorse for solving
optimization problems",
           "the technique is a centerpiece of economic theory but unfortunately its usually taught
poorly"]
vocab = fit(strings)
print(list(vocab.keys()))
print(transform(strings, vocab).toarray())
['but', 'centerpiece', 'economic', 'economists', 'for', 'is', 'its', 'lagrange', 'method',
'multipliers', 'of', 'optimization', 'poorly', 'problems', 'solving', 'taught', 'technique',
'the', 'theory', 'unfortunately', 'usually', 'workhorse']
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```

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# Comparing results with countvectorizer