Wow Loved this place. 1 Crust is not good. 0 Not tasty and the texture was just nasty. 0 Stopped by during the late May bank holiday of 1 The selection on the menu was great and so wer 1 The selection on the menu was great and so wer 1 The selection on the menu was great and so wer 0 Appetite instantly gone. 0 Por Overall I was not impressed and would not go b 0 The whole experience was underwhelming, and I 0 Then, as if I hadn't wasted enough of my life 0			
2 Not tasty and the texture was just nasty. 0 3 Stopped by during the late May bank holiday of 1 4 The selection on the menu was great and so wer 1	0	Wow Loved this place.	1
3 Stopped by during the late May bank holiday of 1 4 The selection on the menu was great and so wer 1	1	Crust is not good.	0
4 The selection on the menu was great and so wer 1	2	Not tasty and the texture was just nasty.	0
	3	Stopped by during the late May bank holiday of	1
995 I think food should have flavor and texture an 0 996 Appetite instantly gone. 0 997 Overall I was not impressed and would not go b 0 998 The whole experience was underwhelming, and I 0	4	The selection on the menu was great and so wer	1
 996 Appetite instantly gone. 0 997 Overall I was not impressed and would not go b 0 998 The whole experience was underwhelming, and I 0 	•••		
 997 Overall I was not impressed and would not go b 998 The whole experience was underwhelming, and I 0 	995	I think food should have flavor and texture an	0
998 The whole experience was underwhelming, and I 0	996	Appetite instantly gone.	0
	997	Overall I was not impressed and would not go b	0
999 Then, as if I hadn't wasted enough of my life 0	998	The whole experience was underwhelming, and I	0
	999	Then, as if I hadn't wasted enough of my life	0

1000 rows × 2 columns

review sentiment f5o@od is # good & good! good food is tasty good Quality is Good good food is not good not good servi89ce is poor not good it is to_o costly not good che^ap quality not good

```
In [30]: doc1='f5o@od is # good & good!'
    doc2='& Food # is * tasty'
    doc3='Quality is Good'
    doc4='food is not good'
    doc5='servi89ce is Poor poor means very poor'
    doc6='it is to_o costly'
    doc7='che^ap quality'

    corpus=[doc1,doc2,doc3,doc4,doc5,doc6,doc7]
    target=['pos','pos','pos','neg','neg','neg']
    print(corpus)

['f5o@od is # good & good!', '& Food # is * tasty', 'Quality is Good', 'food is not good', 'servi89ce is Poor poor means very poor', 'it is to_o costly', 'che^ap quality']
```

```
In [7]: #Text Preprocessing

#step-1 convert corpus to lowercase
corpus1=list(map(str.lower,corpus))
print(corpus1)

#step-2 remove punctutations(symblos like _%^&()!@#$)
import re
def removePunc(doc):
    newdoc=re.sub("[^a-z ]","",doc)
    return newdoc

print(removePunc('f5o@od is # good & good!'))

corpus2=list(map(removePunc,corpus1))
```

```
print(corpus2)
#step-3 remove stopwords(words having no sentiment) like it,is,was,did,has,have,
```

['f5o@od is # good & good!', '& food # is * tasty', 'quality is good', 'food is not good', 'servi89ce is poor poor means very poor', 'it is to_o costly', 'che^ap quality'] food is good good

['food is good', 'food is tasty', 'quality is good', 'food is not good', 'serv ice is poor poor means very poor', 'it is too costly', 'cheap quality']

In [9]: from sklearn.feature_extraction.text import ENGLISH_STOP_WORDS

```
In [12]: print(len(ENGLISH_STOP_WORDS))
   print(ENGLISH_STOP_WORDS)
```

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frozenset({'besides', 'everywhere', 'seems', 'ourselves', 'anyhow', 'your', 'thru', 'int erest', 'below', 'now', 'an', 'wherein', 'whole', 'very', 'anyone', 'hereafter', 'unde r', 'describe', 'cannot', 'might', 'somewhere', 'mostly', 'along', 'being', 'most', 'per haps', 'nine', 'least', 'whether', 'off', 'everyone', 'without', 'never', 'twenty', 'thi s', 'whereby', 'yourselves', 'beforehand', 'would', 'yours', 'could', 'whose', 'sincer e', 'it', 'whereupon', 'become', 'though', 'done', 'other', 'we', 'anyway', 'and', 'exce pt', 'upon', 'up', 'meanwhile', 'find', 'six', 'be', 'noone', 'mine', 'part', 'well', 'i ndeed', 'too', 'both', 'with', 'several', 'alone', 'yet', 'when', 'am', 'even', 'its', 'thick', 'nowhere', 'get', 'is', 'former', 'than', 'whereafter', 'before', 'there', 'sam e', 'may', 'fire', 'somehow', 'third', 'during', 'about', 'please', 'every', 'couldnt', 'through', 'more', 'back', 'hasnt', 'himself', 'yourself', 'itself', 'keep', 'each', 'so metime', 'afterwards', 'although', 'nothing', 'latter', 'or', 'themselves', 'per', 'ther eby', 'beyond', 'none', 'us', 'therein', 'formerly', 'empty', 'con', 'de', 'see', 'whil e', 'no', 'next', 'hers', 'if', 'those', 'by', 'few', 'these', 'together', 'serious', 'm ill', 'down', 'so', 'otherwise', 'will', 'whither', 'herein', 'ours', 'has', 'latterly', 'are', 'behind', 'wherever', 'amongst', 'seemed', 'in', 'our', 'then', 'less', 'rather', 'many', 'move', 'first', 'anywhere', 'into', 'from', 'herself', 'made', 'should', 'amoun t', 'her', 'until', 'nobody', 'were', 'others', 'front', 'which', 'becoming', 'where', 'either', 'have', 'seeming', 'three', 'against', 'hereupon', 'two', 'how', 'because', 'n or', 'the', 'toward', 'me', 'can', 'etc', 'fifty', 'ten', 'thin', 'thus', 'to', 'cry', 'fill', 'she', 'not', 'as', 'around', 'for', 'sixty', 'i', 'hereby', 'cant', 'one', 'm y', 'since', 'go', 'already', 'towards', 'all', 'whoever', 'any', 'top', 'but', 'amoungs t', 'what', 'was', 'show', 'detail', 'over', 'you', 'on', 'else', 'further', 'eg', 'elev en', 'four', 'whenever', 'system', 'another', 'un', 'always', 'ie', 'twelve', 'almost', 'hundred', 'that', 'thereafter', 'him', 'eight', 're', 'last', 'he', 'who', 'found', 'na mely', 'often', 'nevertheless', 'own', 'been', 'neither', 'hence', 'whereas', 'ever', 'v ia', 'due', 'also', 'they', 'do', 'some', 'whatever', 'whence', 'only', 'had', 'onto', 'must', 'again', 'a', 'something', 'thereupon', 'moreover', 'becomes', 'at', 'seem', 'be side', 'once', 'after', 'side', 'five', 'why', 'put', 'bill', 'call', 'however', 'ltd', 'whom', 'within', 'elsewhere', 'inc', 'of', 'give', 'here', 'still', 'enough', 'th eir', 'full', 'throughout', 'anything', 'became', 'fifteen', 'them', 'name', 'myself', 'above', 'someone', 'forty', 'out', 'bottom', 'thence', 'much', 'sometimes', 'across', 'such', 'his', 'among', 'therefore', 'everything', 'between', 'take'})

```
In [18]: spwords=list(ENGLISH_STOP_WORDS)
    spwords.remove('not')
```

```
In [23]: def removespwords(doc):
    wordslist=doc.split()
    newdoc=""
    for word in wordslist:
        if word not in spwords:
            newdoc=newdoc+word+" "
    return newdoc.strip()

removespwords('he and she likes quality of the food')
```

'likes quality food'

Out[23]:

```
print(corpus3)
         ['food good good', 'food tasty', 'quality good', 'food not good', 'service poor poor mea
         ns poor', 'costly', 'cheap quality']
In [29]:
         #step-4 extract features (each unique is a feature)
         #food, good, tasty, quality, not, service, poor, means, costly, cheap
         #cheap,costly,food,good,means,not,poor,quality,service,tasty
         #step-5 obtain vector representation of each document and get feature matrix of docs
         from sklearn.feature extraction.text import CountVectorizer
         cv=CountVectorizer()
         X=cv.fit transform(corpus3) #first extract features then returns sparse matrix for docs
         print(cv.get feature names out())
         print(X)
         X1=X.toarray()
         print(X1)
         ['cheap' 'costly' 'food' 'good' 'means' 'not' 'poor' 'quality' 'service'
          'tasty']
          (0, 2)
                         1
           (0, 3)
           (1, 2)
                        1
          (1, 9)
                        1
           (2, 3)
                        1
           (2, 7)
                        1
           (3, 2)
                       1
           (3, 3)
                       1
           (3, 5)
                        1
                       1
           (4, 8)
          (4, 6)
                       1
           (4, 4)
           (5, 1)
                        1
           (6, 7)
          (6, 0)
                        1
         [[0 0 1 2 0 0 0 0 0 0]
          [0 0 1 0 0 0 0 0 0 1]
          [0 0 0 1 0 0 0 1 0 0]
          [0 0 1 1 0 1 0 0 0 0]
          [0 0 0 0 1 0 3 0 1 0]
          [0 1 0 0 0 0 0 0 0 0]
          [1 0 0 0 0 0 0 1 0 0]]
In [49]: #model training
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.linear model import LogisticRegression
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.svm import SVC
         from sklearn.ensemble import AdaBoostClassifier
         from sklearn.ensemble import RandomForestClassifier
         model=KNeighborsClassifier()
         model=LogisticRegression()
         model=DecisionTreeClassifier()
         model=SVC()
         model=AdaBoostClassifier()
         model=RandomForestClassifier()
         model.fit(X, target)
         print(model.predict(cv.transform(['good quality is not food'])))
         ['neq']
        sample='Food quality is not good$'
```

In [24]: | corpus3=list(map(removespwords,corpus2))

sample1=sample.lower()

```
sample2=removePunc(sample1)
sample3=removespwords(sample2)
print(sample3)
sample4=cv.transform([sample3])
print(model.predict(sample4))
```

food quality not good
['neg']

In []: #text cleaning
 #feature extraction
 #vectorization
 #model training