

1. Importing Packages

In [2]:

```
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")

import sqlite3
import pandas as pd
import numpy as np
import nltk
import string
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import TfidfVectorizer

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.metrics import confusion_matrix
from sklearn import metrics
from sklearn.metrics import roc_curve, auc
from nltk.stem.porter import PorterStemmer
import re
# Tutorial about Python regular expressions: https://pymotw.com/2/re/
import string
from nltk.corpus import stopwords
from nltk.stem import PorterStemmer
from nltk.stem.wordnet import WordNetLemmatizer

from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle

from tqdm import tqdm
import os

from plotly import plotly
import plotly.offline as offline
import plotly.graph_objs as go
offline.init_notebook_mode()
from collections import Counter
```

2. Loading Data

In [3]:

```
project_data = pd.read_csv('D:\\train_data.csv')
resource_data = pd.read_csv('D:\\resources.csv')
```

In [4]:

```
print("Number of data points in train data", project_data.shape)
print('*'*50)
print("The attributes of data :", project_data.columns.values)
```

Number of data points in train data (109248, 17)

```
=====
The attributes of data : ['Unnamed: 0' 'id' 'teacher_id' 'teacher_prefix' 'school_state'
 'project_submitted_datetime' 'project_grade_category'
 'project_subject_categories' 'project_subject_subcategories'
 'project_title' 'project_essay_1' 'project_essay_2' 'project_essay_3'
 'project_essay_4' 'project_resource_summary'
 'teacher_number_of_previously_posted_projects' 'project_is_approved']
```

In [5]:

```
print("Number of data points in resources data", resource_data.shape)
print(resource_data.columns.values)
```

```
Number of data points in resources data (1541272, 4)
['id' 'description' 'quantity' 'price']
```

In [6]:

```
resource_data.head()
```

Out [6]:

	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95
2	p069063	Cory Stories: A Kid's Book About Living With Adhd	1	8.45
3	p069063	Dixon Ticonderoga Wood-Cased #2 HB Pencils, Bo...	2	13.59
4	p069063	EDUCATIONAL INSIGHTS FLUORESCENT LIGHT FILTERS...	3	24.95

In [7]:

```
project_data.head(2)
```

Out [7]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_categ
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

In [8]:

```
# we get the cost of the project using resource.csv file
resource_data.head(2)
```

Out [8]:

	id	description	quantity	price
0	p233245	LC652 - Lakeshore Double-Space Mobile Drying Rack	1	149.00
1	p069063	Bouncy Bands for Desks (Blue support pipes)	3	14.95

In [9]:

```
# https://stackoverflow.com/questions/22407798/how-to-reset-a-dataframes-indexes-for-all-groups-in-one-step
price_data = resource_data.groupby('id').agg({'price':'sum', 'quantity':'sum'}).reset_index()
price_data.head(2)
```

Out [9]:

	id	price	quantity
0	p000001	459.56	7
1	p000002	515.89	21

```
In [10]:
```

```
# join two dataframes in python:  
project_data = pd.merge(project_data, price_data, on='id', how='left')
```

```
In [11]:
```

```
project_data['teacher_prefix'] = project_data['teacher_prefix'].replace(np.NaN, 'Mrs.')
```

3. Text Preprocessing

3.1. Concatenating all essay text

```
In [12]:
```

```
# merge two column text dataframe:  
project_data["essay"] = project_data["project_essay_1"].map(str) + \  
    project_data["project_essay_2"].map(str) + \  
    project_data["project_essay_3"].map(str) + \  
    project_data["project_essay_4"].map(str)
```

3.2. Preprocessing Essay text

```
In [13]:
```

```
# printing some random essays.  
print(project_data['essay'].values[0])  
print("=="*50)  
print(project_data['essay'].values[150])  
print("=="*50)  
print(project_data['essay'].values[1000])  
print("=="*50)  
print(project_data['essay'].values[20000])  
print("=="*50)  
print(project_data['essay'].values[49999])  
print("=="*50)
```

My students are English learners that are working on English as their second or third languages. We are a melting pot of refugees, immigrants, and native-born Americans bringing the gift of language to our school. \r\n\r\n We have over 24 languages represented in our English Learner program with students at every level of mastery. We also have over 40 countries represented with the families within our school. Each student brings a wealth of knowledge and experiences to us that open our eyes to new cultures, beliefs, and respect.\\"The limits of your language are the limits of your world.\\"--Ludwig Wittgenstein Our English learner's have a strong support system at home that begs for more resources. Many times our parents are learning to read and speak English along side of their children. Sometimes this creates barriers for parents to be able to help their child learn phonetics, letter recognition, and other reading skills.\r\n\r\nBy providing these dvd's and players, students are able to continue their mastery of the English language even if no one at home is able to assist. All families with students within the Level 1 proficiency status, will be offered to be a part of this program. These educational videos will be specially chosen by the English Learner Teacher and will be sent home regularly to watch. The videos are to help the child develop early reading skills.\r\n\r\nParents that do not have access to a dvd player will have the opportunity to check out a dvd player to use for the year. The plan is to use these videos and educational dvd's for the years to come for other EL students.\r\nnnannan
=====

The 51 fifth grade students that will cycle through my classroom this year all love learning, at least most of the time. At our school, 97.3% of the students receive free or reduced price lunch. Of the 560 students, 97.3% are minority students. \r\nThe school has a vibrant community that loves to get together and celebrate. Around Halloween there is a whole school parade to show off the beautiful costumes that students wear. On Cinco de Mayo we put on a big festival with crafts made by the students, dances, and games. At the end of the year the school hosts a carnival to celebrate the hard work put in during the school year, with a dunk tank being the most popular activity. My students will use these five brightly colored Hokki stools in place of regular, stationary, 4-legged chairs. As I will only have a total of ten in the classroom and not enough for each student to have an individual one, they will be used in a variety of ways. During independent reading time they will be used as special chairs students will each use on occasion. T will utilize them in place of

will be used as special chairs students will each use on occasion. I will utilize them in place of chairs at my small group tables during math and reading times. The rest of the day they will be used by the students who need the highest amount of movement in their life in order to stay focused on school.\r\n\r\nWhenever asked what the classroom is missing, my students always say more Hokki Stools. They can't get their fill of the 5 stools we already have. When the students are sitting in group with me on the Hokki Stools, they are always moving, but at the same time doing their work. Anytime the students get to pick where they can sit, the Hokki Stools are the first to be taken. There are always students who head over to the kidney table to get one of the stools who are disappointed as there are not enough of them. \r\n\r\nWe ask a lot of students to sit for 7 hours a day. The Hokki stools will be a compromise that allow my students to do desk work and move at the same time. These stools will help students to meet their 60 minutes a day of movement by allowing them to activate their core muscles for balance while they sit. For many of my students, these chairs will take away the barrier that exists in schools for a child who can't sit still.nannan

=====

How do you remember your days of school? Was it in a sterile environment with plain walls, rows of desks, and a teacher in front of the room? A typical day in our room is nothing like that. I work hard to create a warm inviting themed room for my students look forward to coming to each day.\r\n\r\nMy class is made up of 28 wonderfully unique boys and girls of mixed races in Arkansas.\r\nThey attend a Title I school, which means there is a high enough percentage of free and reduced-price lunch to qualify. Our school is an "open classroom" concept, which is very unique as there are no walls separating the classrooms. These 9 and 10 year-old students are very eager learners; they are like sponges, absorbing all the information and experiences and keep on wanting more. With these resources such as the comfy red throw pillows and the whimsical nautical hanging decor and the blue fish nets, I will be able to help create the mood in our classroom setting to be one of a themed nautical environment. Creating a classroom environment is very important in the success in each and every child's education. The nautical photo props will be used with each child as they step foot into our classroom for the first time on Meet the Teacher evening. I'll take pictures of each child with them, have them developed, and then hung in our classroom ready for their first day of 4th grade. This kind gesture will set the tone before even the first day of school! The nautical thank you cards will be used throughout the year by the students as they create thank you cards to their team groups.\r\n\r\nYour generous donations will help me to help make our classroom a fun, inviting, learning environment from day one.\r\n\r\nIt costs lost of money out of my own pocket on resources to get our classroom ready. Please consider helping with this project to make our new school year a very successful one. Thank you!nannan

=====

My kindergarten students have varied disabilities ranging from speech and language delays, cognitive delays, gross/fine motor delays, to autism. They are eager beavers and always strive to work their hardest working past their limitations. \r\n\r\nThe materials we have are the ones I seek out for my students. I teach in a Title I school where most of the students receive free or reduced price lunch. Despite their disabilities and limitations, my students love coming to school and come eager to learn and explore. Have you ever felt like you had ants in your pants and you needed to groove and move as you were in a meeting? This is how my kids feel all the time. They want to be able to move as they learn or so they say. Wobble chairs are the answer and I love them because they develop their core, which enhances gross motor and in Turn fine motor skills. \r\n\r\nThey also want to learn through games, my kids don't want to sit and do worksheets. They want to learn to count by jumping and playing. Physical engagement is the key to our success. The number toss and color and shape mats can make that happen. My students will forget they are doing work and just have the fun a 6 year old deserves.nannan

=====

We have GRIT! If you want to meet tenacious, respectful seven year olds with growth mindsets, you need to come to our classroom. We give hugs, high-fives, and compliments! We Begin with the End in Mind and work hard everyday to reach our goals.\r\n\r\nWe don't believe in making excuses, but there are times in life when you just need to ask for help. As a classroom teacher in a low-income/high poverty school district, my 2nd grade students face real-life struggles both in and out of the classroom. Even though, as a visitor to my classroom, you wouldn't know the daily struggle for some of them. I ask you. How can you learn with your belly growling? How can I provide the absolute best learning environment when we do not have the money to buy research-based materials? \r\n\r\n"Education is not the filling of a pail, but the lighting of a fire," William Butler Yeats. We are not asking you to fill our pail with "things," but to help provide resources to light the fire in young minds. Receiving books written by the same author will teach students how to develop their own Writer's Craft. It will inspire them to think about different ways established authors have developed successful text that appeal to various audiences. \r\n\r\nWe never forget our first love. My mother read the Berenstain Bears series to me when I was five and I fell in love with the Berenstain family. She took me to the public library every week and I would hunt for books written by Stan and Jan Berenstain. Next, was the curious monkey and the man in the yellow hat, Curious George! Thank you Margaret and H.A. Rey for creating a series that captured my heart and attention. \r\n\r\nAs a teacher, it is my hope and dream to inspire the students in my classroom to find their first love in reading. Help me help them to discover writer's craft, go on adventures in their minds, and develop a tenacious love for reading.nannan

=====

In [14]:

```
# https://stackoverflow.com/a/47091490/4084039
import re
```

```

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can't", "can not", phrase)

    # general
    phrase = re.sub(r"\n't", " not", phrase)
    phrase = re.sub(r"\'re", " are", phrase)
    phrase = re.sub(r"\s", " is", phrase)
    phrase = re.sub(r"\d", " would", phrase)
    phrase = re.sub(r"\ll", " will", phrase)
    phrase = re.sub(r"\t", " not", phrase)
    phrase = re.sub(r"\ve", " have", phrase)
    phrase = re.sub(r"\m", " am", phrase)
    return phrase

```

In [15]:

```

sent = decontracted(project_data['essay'].values[16499])
print(sent)
print("=="*50)

```

I teach an amazing, energetic, engaged, and kind group of 5th grade students in an inner city high poverty public school in Indianapolis. Many of my students have parents who work odd hours and have limited time to spend with their wonderfully talented children. My students work hard in class giving 110% with everything that they do. They persevere through difficult topics, enjoy being engaged in their hands-on activities, and they love to laugh while learning. I set high expectation for my students. They understand that true, authentic learning takes hard work, dedication, and requires them to take ownership over their education. My goal for my students is to leave my class as life long learners. The students work hard to overcome all obstacles in their path to meet and grow past my expectations. My students love being active while they are learning and wiggling while they are working. I am lucky enough to have one Hokki stools in my classroom. Sadly, one is not enough to reach all my students. My students love to use the Hokki stools while they learn and want more! One of my students suggested that I write a project since I \"only have one, and we need more.\" These stools help my amazing kiddos get focused while engaging their core to keep them happy and healthy.\r\n\r\nMy students love to wiggle so they can not only focus on their work, but engage in a healthy lifestyle.\r\n\r\nThe Hokki stools would allow my students to continue to be active throughout the day whether they are in small groups or working at their own seat.\r\nnnannan

=====

In [16]:

```

# \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
sent = sent.replace('\\r', ' ')
sent = sent.replace('\\n', ' ')
sent = sent.replace('\\t', ' ')
print(sent)

```

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In [17]:

```

#remove spacial character: https://stackoverflow.com/a/5843547/4084039
sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
print(sent)

```

I teach an amazing energetic engaged and kind group of 5th grade students in an inner city high poverty public school in Indianapolis. Many of my students have parents who work odd hours and have limited time to spend with their wonderfully talented children. My students work hard in class giving 110 with everything that they do. They persevere through difficult topics, enjoy being engaged in their hands-on activities and they love to laugh while learning. I set high expectation for my students. They understand that true authentic learning takes hard work, dedication and requires them to take ownership over their education. My goal for my students is to leave my class as life long learners. The students work hard to overcome all obstacles in their path to meet and grow past my expectations. My students love being active while they are learning and wiggle while they are working. I am lucky enough to have one Hokki stools in my classroom. Sadly, one is not enough to reach all my students. My students love to use the Hokki stools while they learn and want more. One of my students suggested that I write a project since I only have one and we need more. These stools help my amazing kiddos get focused while engaging their core to keep them happy and healthy. My students love to wiggle so they can not only focus on their work but engage in a healthy lifestyle. The Hokki stools would allow my students to continue to be active throughout the day whether they are in small groups or working at their own seat.

In [18]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
\          "you'll", "you'd", "your", 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
           'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their', \
           'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', 'that'll',
'these', 'those', \
           'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
           'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until',
'while', 'of', \
           'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after', \
           'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under',
'again', 'further', \
           'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both',
'each', 'few', 'more', \
           'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very',
's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll',
'm', 'o', 're', \
           've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn',
"doesn't", 'hadn', \
           'hadn't', 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
"mightn't", 'mustn', \
           'mustn't', 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
           'won', "won't", 'wouldn', "wouldn't"]
```

In [19]:

```
# Combining all the above statements
from tqdm import tqdm
preprocessed_essays = []
# tqdm is for printing the status bar
for sentance in tqdm(project_data['essay'].values):
    sent = decontracted(sentance)
    sent = sent.replace('\r', ' ')
    sent = sent.replace('\n', ' ')
    sent = sent.replace('\n', ' ')
    sent = re.sub('[^A-Za-z0-9]+', ' ', sent)
    sent = sent.lower()
    # https://gist.github.com/sebleier/554280
    sent = ' '.join(e for e in sent.split() if e not in stopwords)
    preprocessed_essays.append(sent.strip())
```

100% | [01:36<00:00, 1131.48it/s] | 109248/109248

In [20]:

```
# after preprocessing
preprocessed_essays[20000]
```

Out [20]:

'kindergarten students varied disabilities ranging speech language delays cognitive delays gross fine motor delays autism eager beavers always strive work hardest working past limitations materials ones seek students teach title school students receive free reduced price lunch despite disabilities limitations students love coming school come eager learn explore ever felt like ants pants needed groove move meeting kids feel time want able move learn say wobble chairs answer love develop core enhances gross motor turn fine motor skills also want learn games kids not want sit worksheets want learn count jumping playing physical engagement key success number toss color shape mats make happen students forget work fun 6 year old deserves nannan'

In [21]:

```
project_data['preprocessed_essays'] = preprocessed_essays
project_data.drop(['essay'], axis=1, inplace=True)
project_data.head(2)
```

Out [21]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945 p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

3.2.1. Counting number of words in the combine essays

In [22]:

```
essay_count = []
for word in project_data['preprocessed_essays']:
    a = len(word.split())
    b = str(a)
    essay_count.append(b)
```

In [23]:

```
essay_count[20000]
```

Out [23]:

```
'112'
```

In [24]:

```
project_data['number_of_words_in_essays'] = essay_count
project_data.head(2)
```

Out [24]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945 p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_category
2 rows × 21 columns						

3.3. Preprocessing Title text

In [25]:

```
# printing some random essays.
print(project_data['project_title'].values[0])
print("=="*50)
print(project_data['project_title'].values[150])
print("=="*50)
print(project_data['project_title'].values[1000])
print("=="*50)
print(project_data['project_title'].values[20000])
print("=="*50)
print(project_data['project_title'].values[49999])
print("=="*50)
```

Educational Support for English Learners at Home
=====

More Movement with Hokki Stools
=====

Sailing Into a Super 4th Grade Year
=====

We Need To Move It While We Input It!
=====

Inspiring Young Authors Through Reading
=====

In [26]:

```
# https://stackoverflow.com/a/47091490/4084039
import re

def decontracted(phrase):
    # specific
    phrase = re.sub(r"won't", "will not", phrase)
    phrase = re.sub(r"can't", "can not", phrase)

    # general
    phrase = re.sub(r"\n\t", " not", phrase)
    phrase = re.sub(r"\re", " are", phrase)
    phrase = re.sub(r"\s", " is", phrase)
    phrase = re.sub(r"\d", " would", phrase)
    phrase = re.sub(r"\ll", " will", phrase)
    phrase = re.sub(r"\t", " not", phrase)
    phrase = re.sub(r"\ve", " have", phrase)
    phrase = re.sub(r"\m", " am", phrase)
    return phrase
```

In [27]:

```
title = decontracted(project_data['project_title'].values[20000])
print(title)
print("=="*50)
```

We Need To Move It While We Input It!
=====

In [28]:

```
# \r \n \t remove from string python: http://texthandler.com/info/remove-line-breaks-python/
title = title.replace('\r', ' ')
title = title.replace('\t', ' ')
title = title.replace('\n', ' ')
print(title)
```

We Need To Move It While We Input It!

In [29]:

```
#remove spacial character: https://stackoverflow.com/a/5843547/4084039
title = re.sub('[^A-Za-z0-9]+', ' ', title)
print(title)
```

We Need To Move It While We Input It

In [30]:

```
# https://gist.github.com/sebleier/554280
# we are removing the words from the stop words list: 'no', 'nor', 'not'
stopwords= ['i', 'me', 'my', 'myself', 'we', 'our', 'ours', 'ourselves', 'you', "you're", "you've",
\\
    "you'll", "you'd", "your", 'yours', 'yourself', 'yourselves', 'he', 'him', 'his',
'himself', \
    'she', "she's", 'her', 'hers', 'herself', 'it', "it's", 'its', 'itself', 'they', 'them',
'their', \
    'theirs', 'themselves', 'what', 'which', 'who', 'whom', 'this', 'that', "that'll",
'these', 'those', \
    'am', 'is', 'are', 'was', 'were', 'be', 'been', 'being', 'have', 'has', 'had', 'having',
'do', 'does', \
    'did', 'doing', 'a', 'an', 'the', 'and', 'but', 'if', 'or', 'because', 'as', 'until',
'while', 'of', \
    'at', 'by', 'for', 'with', 'about', 'against', 'between', 'into', 'through', 'during',
'before', 'after', \
    'above', 'below', 'to', 'from', 'up', 'down', 'in', 'out', 'on', 'off', 'over', 'under',
', 'again', 'further', \
    'then', 'once', 'here', 'there', 'when', 'where', 'why', 'how', 'all', 'any', 'both',
'each', 'few', 'more', \
    'most', 'other', 'some', 'such', 'only', 'own', 'same', 'so', 'than', 'too', 'very',
's', 't', 'can', 'will', 'just', 'don', "don't", 'should', "should've", 'now', 'd', 'll',
', 'm', 'o', 're', \
    've', 'y', 'ain', 'aren', "aren't", 'couldn', "couldn't", 'didn', "didn't", 'doesn",
"doesn't", 'hadn', \
    'hadn't', 'hasn', "hasn't", 'haven', "haven't", 'isn', "isn't", 'ma', 'mightn',
'mightn't', 'mustn', \
    'mustn't', 'needn', "needn't", 'shan', "shan't", 'shouldn', "shouldn't", 'wasn',
"wasn't", 'weren', "weren't", \
    'won', "won't", 'wouldn', "wouldn't"]
```

In [31]:

```
# Combining all the above statements
from tqdm import tqdm
preprocessed_titles = []
# tqdm is for printing the status bar
for t in tqdm(project_data['project_title'].values):
    title = decontracted(t)
    title = title.replace('\\r', ' ')
    title = title.replace('\\n', ' ')
    title = re.sub('[^A-Za-z0-9]+', ' ', title)
    # https://gist.github.com/sebleier/554280
    title = title.lower()
    title = ' '.join(e for e in title.split() if e not in stopwords)
    preprocessed_titles.append(title.strip())
```

100% | 109248/109248
[00:04<00:00, 24758.84it/s]

In [32]:

```
# after preprocessing
preprocessed_titles[20000]
```

Out[32]:

'need move input'

In [33]:

```
project_data['preprocessed_titles'] = preprocessed_titles  
project_data.drop(['project_title'], axis=1, inplace=True)  
project_data.head(2)
```

Out[33]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945 p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

2 rows × 21 columns

3.3.1. Counting number of words in the title

In [34]:

```
title_count = []  
for word in project_data['preprocessed_titles']:  
    a = len(word.split())  
    b = str(a)  
    title_count.append(b)
```

In [35]:

```
title_count[20000]
```

Out[35]:

```
'3'
```

In [36]:

```
project_data['number_of_words_in_the_title'] = title_count  
project_data.head(2)
```

Out[36]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_grade_cate
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Grades P
1	140945 p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Grade

2 rows × 22 columns

3.4. Preprocessing project_grade_category

In [37]:

```
project_grade_clean_category = []  
  
for i in range(len(project_data)):  
    a = project_data["project_grade_category"][i].replace(" ", "_").replace("-", "_")  
    project_grade_clean_category.append(a)
```

```
project_grade_clean_category.append(a)
```

In [38]:

```
project_grade_clean_category[0:5]
```

Out[38]:

```
['Grades_PreK_2', 'Grades_6_8', 'Grades_6_8', 'Grades_PreK_2', 'Grades_PreK_2']
```

In [39]:

```
project_data['project_grade_clean_category'] = project_grade_clean_category
project_data.drop(['project_grade_category'], axis=1, inplace=True)
project_data.head(2)
```

Out[39]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_subject_ca
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	Literacy & L
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	History & Civics,

2 rows × 22 columns

3.5. Preprocessing project_subject_categories

In [40]:

```
catogories = list(project_data['project_subject_categories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python
cat_list = []
for i in catogories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','):# it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing 'The')
            j = j.replace(' ', '') # we are placeing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Science"
            temp+=j.strip()+" "# abc ".strip() will return "abc", remove the trailing spaces
            temp = temp.replace('&','_') # we are replacing the & value into
    cat_list.append(temp.strip())
```

In [41]:

```
cat_list[0:5]
```

Out[41]:

```
['Literacy_Language',
 'History_Civics_Health_Sports',
 'Health_Sports',
 'Literacy_Language_Math_Science',
 'Math_Science']
```

In [42]:

```
project_data['clean_categories'] = cat_list
project_data.drop(['project_subject_categories'], axis=1, inplace=True)
project_data.head(2)
```

Out[42]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_subject_su
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	
1	140945 p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Civics & Gover

2 rows × 22 columns

3.6. Preprocessing project_subject_subcategories

In [43]:

```
sub_categories = list(project_data['project_subject_subcategories'].values)
# remove special characters from list of strings python:
https://stackoverflow.com/a/47301924/4084039

# https://www.geeksforgeeks.org/removing-stop-words-nltk-python/
# https://stackoverflow.com/questions/23669024/how-to-strip-a-specific-word-from-a-string
# https://stackoverflow.com/questions/8270092/remove-all-whitespace-in-a-string-in-python

sub_cat_list = []
for i in sub_categories:
    temp = ""
    # consider we have text like this "Math & Science, Warmth, Care & Hunger"
    for j in i.split(','):# it will split it in three parts ["Math & Science", "Warmth", "Care & Hunger"]
        if 'The' in j.split(): # this will split each of the category based on space "Math & Science"=> "Math", "&", "Science"
            j=j.replace('The','') # if we have the words "The" we are going to replace it with ''(i.e removing 'The')
            j = j.replace(' ','') # we are placeing all the ' '(space) with ''(empty) ex:"Math & Science"=>"Math&Science"
            temp +=j.strip()+" "# abc ".strip() will return "abc", remove the trailing spaces
            temp = temp.replace('&','_')
        sub_cat_list.append(temp.strip())

```

In [44]:

```
sub_cat_list[0:5]
```

Out[44]:

```
['ESL Literacy',
'Civics_Government TeamSports',
'Health_Wellness TeamSports',
'Literacy Mathematics',
'Mathematics']
```

In [45]:

```
project_data['clean_subcategories'] = sub_cat_list
project_data.drop(['project_subject_subcategories'], axis=1, inplace=True)
project_data.head(2)
```

Out[45]:

Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_subject_su	project_clean_subcategory
0	160221 p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57		ESL Literacy

Unnamed: 0				teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay_1
				teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay_1
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.		IN	2016-12-05 13:43:57	My students are English learners that are work...
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.		FL	2016-10-25 09:22:10	Our students arrive to our school eager to lea...

2 rows × 22 columns

In [46]:

```
project_data.head()
```

Out [46]:

Unnamed: 0				teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay_1
				teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay_1
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.		IN	2016-12-05 13:43:57	My students are English learners that are work...
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.		FL	2016-10-25 09:22:10	Our students arrive to our school eager to lea...
2	21895	p182444	3465aa82da834c0582ebd0ef8040ca0	Ms.		AZ	2016-08-31 12:03:56	\r\n\"True champions aren't always the ones th...
3	45	p246581	f3cb9bffbba169bef1a77b243e620b60	Mrs.		KY	2016-10-06 21:16:17	I work at a unique school filled with both ESL...
4	172407	p104768	be1f7507a41f8479dc06f047086a39ec	Mrs.		TX	2016-07-11 01:10:09	Our second grade classroom next year will be m...

5 rows × 22 columns

4. Sentiment score's of each of the essay

In [47]:

```
import nltk
from nltk.sentiment.vader import SentimentIntensityAnalyzer
```

In [48]:

```
analyser = SentimentIntensityAnalyzer()
```

In [49]:

```
neg = []
pos = []
neu = []
compound = []

for a in tqdm(project_data["preprocessed_essays"]):
    b = analyser.polarity_scores(a) ['neg']
    c = analyser.polarity_scores(a) ['pos']
    d = analyser.polarity_scores(a) ['neu']
    e = analyser.polarity_scores(a) ['compound']
    neg.append(b)
    pos.append(c)
    neu.append(d)
    compound.append(e)
```

```
100%|██████████| 109248/109248  
[20:08<00:00, 90.38it/s]
```

In [50]:

```
project_data["pos"] = pos
```

In [51]:

```
project_data["neg"] = neg
```

In [52]:

```
project_data["neu"] = neu
```

In [53]:

```
project_data["compound"] = compound
```

In [54]:

```
project_data.head(2)
```

Out[54]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay_1	...
0	160221	p253737	c90749f5d961ff158d4b4d1e7dc665fc	Mrs.	IN	2016-12-05 13:43:57	My students are English learners that are work...	...
1	140945	p258326	897464ce9ddc600bcfd1151f324dd63a	Mr.	FL	2016-10-25 09:22:10	Our students arrive to our school eager to lea...	:

2 rows × 26 columns

In []:

5. Splitting data into Train and cross validation(or test): Stratified Sampling

In [55]:

```
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(project_data,
project_data['project_is_approved'], test_size=0.33, stratify=project_data['project_is_approved'])
X_train, X_cv, y_train, y_cv = train_test_split(X_train, y_train, test_size=0.33, stratify=y_train)
```

6. Dropping Target values from Train, Test and CV set

In [56]:

```
X_train.drop(['project_is_approved'], axis=1, inplace=True)
X_test.drop(['project_is_approved'], axis=1, inplace=True)
X_cv.drop(['project_is_approved'], axis=1, inplace=True)
```

In [57]:

```
print(X_train.shape)
print(X_test.shape)
print(X_cv.shape)
```

```
(49041, 25)
(36052, 25)
(24155, 25)
```

In [58]:

```
X_train.head(2)
```

Out[58]:

	Unnamed: 0	id	teacher_id	teacher_prefix	school_state	project_submitted_datetime	project_essay
74256	167247	p157147	9e2e9b636c10d12e91e2daa9715b989f	Ms.	NJ	2017-02-21 08:02:30	Our 4th grade inclusiv classroom filled w
18907	119402	p001440	344b01f27068c2047f3facfb2a8b58b2	Mrs.	IL	2017-01-08 17:43:27	Jasper Cour Junior Hi students love do

2 rows × 25 columns

In [59]:

```
y_train.head()
```

Out[59]:

```
74256    1
18907    1
90972    1
61993    0
59817    1
Name: project_is_approved, dtype: int64
```

7. Encoding Categorical Data

7.1. One Hot Encoding of clean_categories

In [60]:

```
# # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# from collections import Counter
# my_counter = Counter()
# for word in project_data['clean_categories'].values:
#     my_counter.update(word.split())

# # dict sort by value python: https://stackoverflow.com/a/613218/4084039
# cat_dict = dict(my_counter)
# sorted_cat_dict = dict(sorted(cat_dict.items(), key=lambda kv: kv[1]))
```

In [61]:

```
# we use count vectorizer to convert the values into one hot encoded features
from sklearn.feature_extraction.text import CountVectorizer
vectorizer1 = CountVectorizer(lowercase=False, binary=True)
vectorizer1.fit(X_train['clean_categories'].values)
print(vectorizer1.get_feature_names())

categories_one_hot_Xtrain = vectorizer1.transform(X_train['clean_categories'].values)
```

```

categories_one_hot_Xtest = vectorizer1.transform(X_test['clean_categories'].values)
categories_one_hot_Xcv = vectorizer1.transform(X_cv['clean_categories'].values)
print("Shape of matrix after one hot encoding ", categories_one_hot_Xtrain.shape)
print("Shape of matrix after one hot encoding ", categories_one_hot_Xtest.shape)
print("Shape of matrix after one hot encoding ", categories_one_hot_Xcv.shape)

```

```

['AppliedLearning', 'Care_Hunger', 'Health_Sports', 'History_Civics', 'Literacy_Language',
'Math_Science', 'Music_Arts', 'SpecialNeeds', 'Warmth']
Shape of matrix after one hot encoding  (49041, 9)
Shape of matrix after one hot encoding  (36052, 9)
Shape of matrix after one hot encoding  (24155, 9)

```

7.2. One Hot Encoding of clean_subcategories

In [62]:

```

# # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# from collections import Counter
# my_counter = Counter()
# for word in project_data['clean_subcategories'].values:
#     my_counter.update(word.split())

# # dict sort by value python: https://stackoverflow.com/a/613218/4084039
# sub_cat_dict = dict(my_counter)
# sorted_sub_cat_dict = dict(sorted(sub_cat_dict.items(), key=lambda kv: kv[1]))

```

In [63]:

```

# we use count vectorizer to convert the values into one hot encoded features
vectorizer2 = CountVectorizer(lowercase=False, binary=True)
vectorizer2.fit(project_data['clean_subcategories'].values)
print(vectorizer2.get_feature_names())

sub_categories_one_hot_Xtrain = vectorizer2.transform(X_train['clean_subcategories'].values)
sub_categories_one_hot_Xtest = vectorizer2.transform(X_test['clean_subcategories'].values)
sub_categories_one_hot_Xcv = vectorizer2.transform(X_cv['clean_subcategories'].values)

print("Shape of matrix after one hot encoding ", sub_categories_one_hot_Xtrain.shape)
print("Shape of matrix after one hot encoding ", sub_categories_one_hot_Xtest.shape)
print("Shape of matrix after one hot encoding ", sub_categories_one_hot_Xcv.shape)

['AppliedSciences', 'Care_Hunger', 'CharacterEducation', 'Civics_Government',
'College_CareerPrep', 'CommunityService', 'ESL', 'EarlyDevelopment', 'Economics',
'EnvironmentalScience', 'Extracurricular', 'FinancialLiteracy', 'ForeignLanguages', 'Gym_Fitness',
'Health_LifeScience', 'Health_Wellness', 'History_Geography', 'Literacy', 'Literature_Writing', 'Mathematics',
'Music', 'NutritionEducation', 'Other', 'ParentInvolvement', 'PerformingArts', 'SocialSciences',
'SpecialNeeds', 'TeamSports', 'VisualArts', 'Warmth']
Shape of matrix after one hot encoding  (49041, 30)
Shape of matrix after one hot encoding  (36052, 30)
Shape of matrix after one hot encoding  (24155, 30)

```

7.3. One Hot Encoding of school_state

In [64]:

```

# # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# from collections import Counter
# my_counter = Counter()
# for word in project_data['school_state'].values:
#     my_counter.update(word.split())

# # dict sort by value python: https://stackoverflow.com/a/613218/4084039
# school_state_dict = dict(my_counter)
# sorted_school_state_dict = dict(sorted(school_state_dict.items(), key=lambda kv: kv[1]))

```

In [65]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer3 = CountVectorizer(lowercase=False, binary=True)
vectorizer3.fit(project_data['school_state'].values)
print(vectorizer3.get_feature_names())

school_state_one_hot_Xtrain = vectorizer3.transform(X_train['school_state'].values)
school_state_one_hot_Xtest = vectorizer3.transform(X_test['school_state'].values)
school_state_one_hot_Xcv = vectorizer3.transform(X_cv['school_state'].values)

print("Shape of matrix after one hot encoding ", school_state_one_hot_Xtrain.shape)
print("Shape of matrix after one hot encoding ", school_state_one_hot_Xtest.shape)
print("Shape of matrix after one hot encoding ", school_state_one_hot_Xcv.shape)
```

```
['AK', 'AL', 'AR', 'AZ', 'CA', 'CO', 'CT', 'DC', 'DE', 'FL', 'GA', 'HI', 'IA', 'ID', 'IL', 'IN', 'KS',
 'KY', 'LA', 'MA', 'MD', 'ME', 'MI', 'MN', 'MO', 'MS', 'MT', 'NC', 'ND', 'NE', 'NH', 'NJ', 'NM',
 'NV', 'NY', 'OH', 'OK', 'OR', 'PA', 'RI', 'SC', 'SD', 'TN', 'TX', 'UT', 'VA', 'VT', 'WA', 'WI', 'WV',
 'WY']
Shape of matrix after one hot encoding (49041, 51)
Shape of matrix after one hot encoding (36052, 51)
Shape of matrix after one hot encoding (24155, 51)
```

7.4. One Hot Encoding of teacher_prefix

In [66]:

```
# # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# from collections import Counter
# my_counter = Counter()
# for word in project_data['teacher_prefix'].values:
#     my_counter.update(word.split())

# # dict sort by value python: https://stackoverflow.com/a/613218/4084039
# teacher_prefix_dict = dict(my_counter)
# sorted_teacher_prefix_dict = dict(sorted(teacher_prefix_dict.items(), key=lambda kv: kv[1]))
```

In [67]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer4 = CountVectorizer(lowercase=False, binary=True)
vectorizer4.fit(project_data['teacher_prefix'].values)
print(vectorizer4.get_feature_names())

teacher_prefix_one_hot_Xtrain = vectorizer4.transform(X_train['teacher_prefix'].values)
teacher_prefix_one_hot_Xtest = vectorizer4.transform(X_test['teacher_prefix'].values)
teacher_prefix_one_hot_Xcv = vectorizer4.transform(X_cv['teacher_prefix'].values)

print("Shape of matrix after one hot encoding ", teacher_prefix_one_hot_Xtrain.shape)
print("Shape of matrix after one hot encoding ", teacher_prefix_one_hot_Xtest.shape)
print("Shape of matrix after one hot encoding ", teacher_prefix_one_hot_Xcv.shape)

['Dr', 'Mr', 'Mrs', 'Ms', 'Teacher']
Shape of matrix after one hot encoding (49041, 5)
Shape of matrix after one hot encoding (36052, 5)
Shape of matrix after one hot encoding (24155, 5)
```

7.5. One Hot Encoding of project_grade_clean_category

In [68]:

```
# # count of all the words in corpus python: https://stackoverflow.com/a/22898595/4084039
# from collections import Counter
# my_counter = Counter()
# for word in project_data['project_grade_clean_category'].values:
#     my_counter.update(word.split())

# # dict sort by value python: https://stackoverflow.com/a/613218/4084039
# grade_dict = dict(my_counter)
```

```
# sorted_grade_dict = dict(sorted(grade_dict.items(), key=lambda kv: kv[1]))
```

In [69]:

```
# we use count vectorizer to convert the values into one hot encoded features
vectorizer5 = CountVectorizer(lowercase=False, binary=True)
vectorizer5.fit(project_data['project_grade_clean_category'].values)
print(vectorizer5.get_feature_names())

grade_one_hot_Xtrain = vectorizer5.transform(X_train['project_grade_clean_category'].values)
grade_one_hot_Xtest = vectorizer5.transform(X_test['project_grade_clean_category'].values)
grade_one_hot_Xcv = vectorizer5.transform(X_cv['project_grade_clean_category'].values)

print("Shape of matrix after one hot encoding ", grade_one_hot_Xtrain.shape)
print("Shape of matrix after one hot encoding ", grade_one_hot_Xtest.shape)
print("Shape of matrix after one hot encoding ", grade_one_hot_Xcv.shape)

['Grades_3_5', 'Grades_6_8', 'Grades_9_12', 'Grades_PreK_2']
Shape of matrix after one hot encoding (49041, 4)
Shape of matrix after one hot encoding (36052, 4)
Shape of matrix after one hot encoding (24155, 4)
```

8. Encoding of Text Data

8.1.1. BOW encoding of preprocessed_essays

In [70]:

```
# We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer6 = CountVectorizer(min_df=10)
text_bow_Xtrain = vectorizer6.fit_transform(X_train['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ", text_bow_Xtrain.shape)
text_bow_Xtest = vectorizer6.transform(X_test['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ", text_bow_Xtest.shape)
text_bow_Xcv = vectorizer6.transform(X_cv['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ", text_bow_Xcv.shape)

Shape of matrix after one hot encoding (49041, 12030)
Shape of matrix after one hot encoding (36052, 12030)
Shape of matrix after one hot encoding (24155, 12030)
```

8.1.2. BOW encoding of preprocessed_titles

In [71]:

```
# We are considering only the words which appeared in at least 10 documents(rows or projects).
vectorizer7 = CountVectorizer(min_df=10)
title_bow_Xtrain = vectorizer7.fit_transform(X_train['preprocessed_titles'].values)
print("Shape of matrix after one hot encoding ", title_bow_Xtrain.shape)
title_bow_Xtest = vectorizer7.transform(X_test['preprocessed_titles'].values)
print("Shape of matrix after one hot encoding ", title_bow_Xtest.shape)
title_bow_Xcv = vectorizer7.transform(X_cv['preprocessed_titles'].values)
print("Shape of matrix after one hot encoding ", title_bow_Xcv.shape)

Shape of matrix after one hot encoding (49041, 2016)
Shape of matrix after one hot encoding (36052, 2016)
Shape of matrix after one hot encoding (24155, 2016)
```

8.2.1. TFIDF encoding of preprocessed_essays

In [72]:

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

```

vectorizer8 = TfidfVectorizer(min_df=10)
text_tfidf_Xtrain = vectorizer8.fit_transform(X_train['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ",text_tfidf_Xtrain.shape)
text_tfidf_Xtest = vectorizer8.transform(X_test['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ",text_tfidf_Xtest.shape)
text_tfidf_Xcv = vectorizer8.transform(X_cv['preprocessed_essays'].values)
print("Shape of matrix after one hot encoding ",text_tfidf_Xcv.shape)

```

Shape of matrix after one hot encoding (49041, 12030)
 Shape of matrix after one hot encoding (36052, 12030)
 Shape of matrix after one hot encoding (24155, 12030)

Using elbow method to find n_components for truncated SVD

In [83]:

```
X = text_tfidf_Xtrain[:,0:3000]      # Due to memory issue
Y= text_tfidf_Xtest[:,0:3000]
Z= text_tfidf_Xcv[:,0:3000]
```

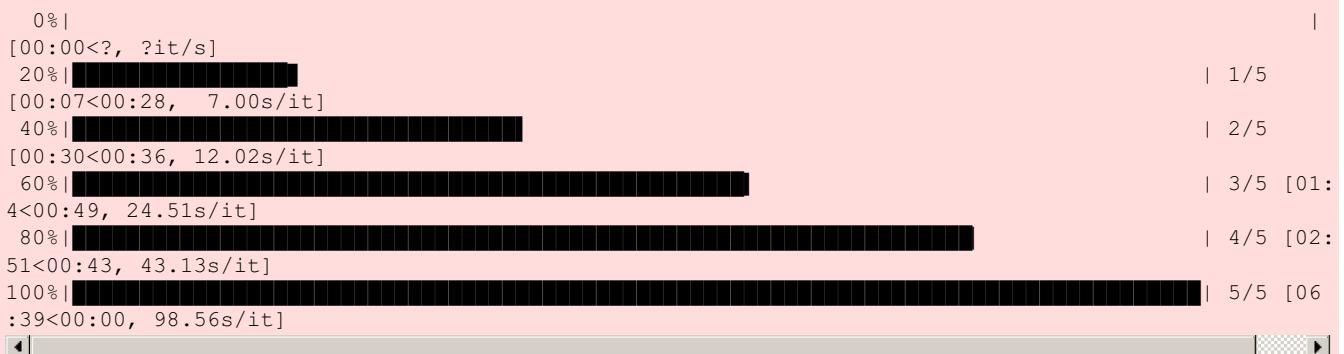
In [84]:

```
print(X.shape)
print(Y.shape)
print(Z.shape)
```

(49041, 3000)
 (36052, 3000)
 (24155, 3000)

In [75]:

```
from sklearn.decomposition import TruncatedSVD
feature_number = [100,400,800,1200,2000]
Variance_sum = []
for i in tqdm(feature_number):
    svd = TruncatedSVD(n_components = i , random_state = 42)
    svd.fit(X)
    Variance_sum.append(svd.explained_variance_ratio_.sum())
```



In []:

```
# from sklearn.decomposition import TruncatedSVD
# https://scikit-learn.org/stable/modules/generated/sklearn.decomposition.TruncatedSVD.html
# declaring index as Dimensions in train_text_tfidf
# svd = TruncatedSVD(n_components = 2500, random_state = 42)
# svd.fit(X)
# Variance_sum.append(svd.explained_variance_ratio_.sum())
```

In [76]:

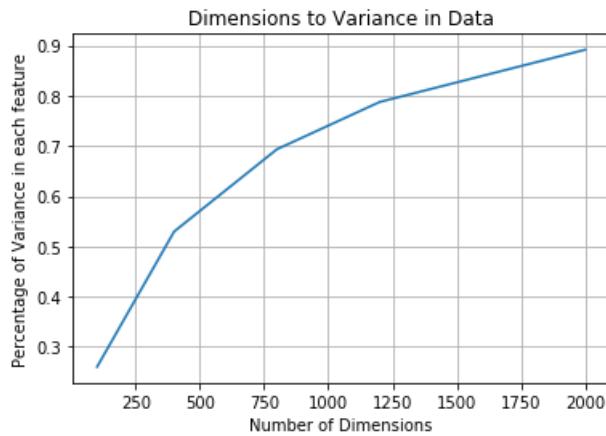
Variance_sum

Out[76]:

```
[0.26027893183455386,  
 0.5296091192140049,  
 0.6933967121237589,  
 0.7875106167036654,  
 0.8915278107419001]
```

In [79]:

```
plt.xlabel("Number of Dimensions")  
plt.ylabel("Percentage of Variance in each feature")  
plt.title("Dimensions to Variance in Data")  
plt.plot(feature_number, Variance_sum)  
plt.grid(True)  
plt.show()
```



In [85]:

```
svd = TruncatedSVD(n_components = 2000)  
#Train SVD  
svdtrain = svd.fit_transform(X)  
#Test SVD  
svdtest = svd.transform(Y)  
#CV SVD  
svdcv = svd.transform(Z)
```

In [86]:

```
print(svdtrain.shape)  
print(svdtest.shape)  
print(svdcv.shape)
```

```
(49041, 2000)  
(36052, 2000)  
(24155, 2000)
```

In []:

In []:

8.2.2. TFIDF encoding of preprocessed_titles

In [88]:

```
from sklearn.feature_extraction.text import TfidfVectorizer  
vectorizer9 = TfidfVectorizer(min_df=10)  
title_tfidf_Xtrain = vectorizer9.fit_transform(X_train['preprocessed_titles'].values)  
print("Shape of matrix after one hot encoding " + title_tfidf_Xtrain.shape)
```

```

print("Shape of matrix after one hot encoding ",title_cvec_Xtrain.shape)
title_tfidf_Xtest = vectorizer9.transform(X_test['preprocessed_titles'].values)
print("Shape of matrix after one hot encoding ",title_tfidf_Xtest.shape)
title_tfidf_Xcv = vectorizer9.transform(X_cv['preprocessed_titles'].values)
print("Shape of matrix after one hot encoding ",title_tfidf_Xcv.shape)

```

```

Shape of matrix after one hot encoding  (49041, 2016)
Shape of matrix after one hot encoding  (36052, 2016)
Shape of matrix after one hot encoding  (24155, 2016)

```

8.3.1. Average Word2Vec encoding of preprocessed_essays on Train Data

In [89]:

```

# stronging variables into pickle files python: http://www.jessicayung.com/how-to-use-pickle-to-save-and-load-variables-in-python/
# make sure you have the glove_vectors file
with open('D:\glove_vectors', 'rb') as f:
    model = pickle.load(f)
    glove_words = set(model.keys())

```

In [217]:

```

# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_essays_Xtrain = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['preprocessed_essays']): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_essays_Xtrain.append(vector)

print(len(avg_w2v_vectors_essays_Xtrain))
print(len(avg_w2v_vectors_essays_Xtrain[2]))

```

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3%|██████████
09:15, 86.03it/s] | 1266/49041 [00:

3%|██████████
09:29, 83.88it/s] | 1276/49041 [00:

3%|██████████
09:51, 80.74it/s] | 1285/49041 [00:

3%|██████████
08:04, 98.48it/s] | 1306/49041 [00:

3%|██████████
7:19, 108.49it/s] | 1321/49041 [00:4

3%|██████████
6:39, 119.26it/s] | 1338/49041 [00:4

3%|██████████
6:43, 118.23it/s] | 1353/49041 [00:4

3%|██████████
6:37, 119.87it/s] | 1366/49041 [00:4

3%|██████████
6:43, 118.08it/s] | 1379/49041 [00:4

3%|██████████
7:07, 111.44it/s] | 1392/49041 [00:4

3%|██████████
6:32, 121.42it/s] | 1408/49041 [00:4

3%|██████████
6:41, 118.74it/s] | 1422/49041 [00:4

3%|██████████
7:51, 100.90it/s] | 1435/49041 [00:4

3%|██████████
7:02, 112.62it/s] | 1453/49041 [00:4

3%|██████████
7:41, 103.02it/s] | 1466/49041 [00:4

3%|██████████

7:26, 106.50it/s] | 1491/49041 [00:4
3%|██████ 7:51, 100.84it/s] | 1505/49041 [00:4
3%|██████ 7:49, 101.26it/s] | 1527/49041 [00:4
3%|██████ 6:33, 120.63it/s] | 1541/49041 [00:4
3%|██████ 7:12, 109.95it/s] | 1554/49041 [00:4
3%|██████ 7:13, 109.40it/s] | 1567/49041 [00:4
3%|██████ 6:33, 120.49it/s] | 1585/49041 [00:4
3%|██████ 08:49, 89.66it/s] | 1598/49041 [00:
3%|██████ 08:14, 95.99it/s] | 1610/49041 [00:
3%|██████ 7:52, 100.34it/s] | 1622/49041 [00:5
3%|██████ 6:48, 116.04it/s] | 1643/49041 [00:5
3%|██████ 6:59, 113.07it/s] | 1657/49041 [00:5
3%|██████ 08:43, 90.57it/s] | 1670/49041 [00:
3%|██████ 10:06, 78.10it/s] | 1681/49041 [00:
3%|██████ 09:34, 82.47it/s] | 1691/49041 [00:
3%|██████ 07:53, 99.98it/s] | 1711/49041 [00:
4%|██████ 7:07, 110.73it/s] | 1726/49041 [00:5
4%|██████ 6:47, 116.06it/s] | 1740/49041 [00:5
4%|██████ 6:27, 122.05it/s] | 1754/49041 [00:5

4%|██████
6:15, 126.00it/s] | 1768/49041 [00:5

4%|██████
5:47, 136.04it/s] | 1788/49041 [00:5

4%|██████
6:48, 115.58it/s] | 1803/49041 [00:5

4%|██████
6:19, 124.33it/s] | 1820/49041 [00:5

4%|██████
08:06, 97.00it/s] | 1834/49041 [00:5

4%|██████
6:52, 114.47it/s] | 1858/49041 [00:5

4%|██████
6:31, 120.46it/s] | 1872/49041 [00:5

4%|██████
6:30, 120.76it/s] | 1886/49041 [00:5

4%|██████
5:55, 132.56it/s] | 1904/49041 [00:5

4%|██████
5:50, 134.27it/s] | 1919/49041 [00:5

4%|██████
6:45, 116.26it/s] | 1934/49041 [00:5

4%|██████
7:26, 105.42it/s] | 1947/49041 [00:5

4%|██████
6:28, 121.32it/s] | 1967/49041 [00:5

4%|██████
7:26, 105.49it/s] | 1981/49041 [00:5

4%|██████
7:42, 101.76it/s] | 1993/49041 [00:5

4%|██████
7:22, 106.36it/s] | 2008/49041 [00:5

4%|██████
09:03, 86.53it/s] | 2020/49041 [00:5

4%|██████
09:32, 82.17it/s] | 2030/49041 [00:5

4%|██████
09:17, 84.34it/s] | 2040/49041 [00:5

4%|██████| 2050/49041 [00:
09:42, 80.64it/s]

4%|██████| 2064/49041 [00:
08:40, 90.28it/s]

4%|██████| 2074/49041 [00:
09:08, 85.60it/s]

4%|██████| 2089/49041 [00:
08:20, 93.73it/s]

4%|██████| 2099/49041 [00:
08:39, 90.42it/s]

4%|██████| 2109/49041 [00:
08:49, 88.64it/s]

4%|██████| 2119/49041 [00:
09:53, 79.07it/s]

4%|██████| 2128/49041 [00:
10:54, 71.70it/s]

4%|██████| 2140/49041 [00:
09:38, 81.03it/s]

4%|██████| 2160/49041 [00:
08:04, 96.70it/s]

4%|██████| 2175/49041
[00:55<07:15, 107.62it/s]

4%|██████| 2188/49041
[00:55<07:19, 106.53it/s]

4%|██████| 2200/49041 [00:
08:45, 89.16it/s]

5%|█████| 2211/49041 [00:
08:34, 90.97it/s]

5%|█████| 2223/49041 [00:
07:59, 97.63it/s]

5%|█████| 2234/49041 [00:
08:28, 92.08it/s]

5%|█████| 2244/49041 [00:
09:35, 81.31it/s]

5%|█████| 2259/49041 [00:
08:17, 94.08it/s]

5%|█████| 2270/49041 [00:
08:02, 96.98it/s]

5% | [██████] | 2283/49041 [00:56<07:37, 102.27it/s]

5% | [██████] | 2297/49041 [00:56<07:14, 107.57it/s]

5% | [██████] | 2309/49041 [00:08:29, 91.68it/s]

5% | [██████] | 2321/49041 [00:08:05, 96.17it/s]

5% | [██████] | 2338/49041 [00:08:36, 90.47it/s]

5% | [██████] | 2350/49041 [00:09:09, 84.89it/s]

5% | [██████] | 2361/49041 [00:08:06, 95.93it/s]

5% | [██████] | 2376/49041 [00:08:32, 91.07it/s]

5% | [██████] | 2387/49041 [00:00:57<07:30, 103.56it/s]

5% | [██████] | 2406/49041 [00:00:57<07:24, 104.85it/s]

5% | [██████] | 2421/49041 [00:08:30, 91.32it/s]

5% | [██████] | 2433/49041 [00:09:35, 80.95it/s]

5% | [██████] | 2444/49041 [00:07:55, 98.05it/s]

5% | [██████] | 2464/49041 [00:08:21, 92.76it/s]

5% | [██████] | 2477/49041 [00:09:48, 79.10it/s]

5% | [██████] | 2488/49041 [00:09:49, 78.94it/s]

5% | [██████] | 2498/49041 [00:08:59, 86.27it/s]

5% | [██████] | 2510/49041 [00:07:54, 98.04it/s]

5% | [██████] | 2529/49041 [00:00:54<07:54, 98.04it/s]

5% | [██████] | 2541/49041 [00:00:54<07:54, 98.04it/s]

08:39, 89.45it/s] | 2556/49041
5%|██████ [00:59<07:42, 100.56it/s]
5%|██████ 07:49, 98.88it/s] | 2568/49041 [00:
08:47, 88.13it/s] | 2579/49041 [00:
08:44, 88.58it/s] | 2590/49041 [00:
08:30, 91.02it/s] | 2601/49041 [00:
07:57, 97.20it/s] | 2613/49041 [00:
07:44, 99.96it/s] | 2624/49041 [01:
[01:00<07:35, 101.87it/s] | 2635/49041
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11:41, 66.12it/s] | 2655/49041 [01:
09:47, 78.90it/s] | 2672/49041 [01:
09:19, 82.80it/s] | 2682/49041 [01:
09:54, 77.96it/s] | 2692/49041 [01:
08:53, 86.85it/s] | 2705/49041 [01:
08:13, 93.89it/s] | 2717/49041 [01:
[01:01<06:59, 110.51it/s] | 2737/49041
6%|██████ 09:46, 78.93it/s] | 2750/49041 [01:
08:51, 87.06it/s] | 2765/49041 [01:
[01:01<07:25, 103.79it/s] | 2784/49041

6% ██████	2798/49041 [01: 07:46, 99.09it/s]	2798/49041 [01:
6% ██████	2810/49041 [01: 08:08, 94.68it/s]	2810/49041 [01:
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6% ██████	08:54, 86.43it/s]	2856/49041 [01:
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6% ██████	[01:02<07:17, 105.58it/s]	2897/49041
6% ██████	[01:02<06:25, 119.63it/s]	2917/49041
6% ██████	[01:03<06:18, 121.78it/s]	2931/49041
6% ██████	[01:03<05:49, 131.80it/s]	2950/49041
6% ██████	[01:03<06:31, 117.69it/s]	2965/49041
6% ██████	[01:03<06:46, 113.38it/s]	2978/49041
6% ██████	[01:03<07:11, 106.73it/s]	2991/49041
6% ██████	[01:03<07:28, 102.66it/s]	3003/49041
6% ██████	09:04, 84.45it/s]	3014/49041 [01:
6% ██████	10:33, 72.60it/s]	3024/49041 [01:
6% ██████	12:27, 61.55it/s]	3033/49041 [01:
6% ██████	10:11, 75.15it/s]	3053/49041 [01:

6%|██████
09:13, 83.07it/s] | 3065/49041 [01:

6%|██████
08:46, 87.32it/s] | 3076/49041 [01:

6%|██████
09:37, 79.53it/s] | 3086/49041 [01:

6%|██████
09:53, 77.37it/s] | 3095/49041 [01:

6%|██████
08:29, 90.21it/s] | 3110/49041 [01:

6%|██████
07:56, 96.40it/s] | 3122/49041 [01:

6%|██████
[01:05<07:21, 103.88it/s] | 3136/49041

6%|██████
08:11, 93.35it/s] | 3148/49041 [01:

6%|██████
07:40, 99.71it/s] | 3160/49041 [01:

6%|██████
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7%|███████
[01:05<07:08, 107.05it/s] | 3190/49041

7%|███████
[01:05<06:36, 115.66it/s] | 3206/49041

7%|███████
[01:05<06:32, 116.63it/s] | 3219/49041

7%|███████
[01:06<06:25, 118.73it/s] | 3233/49041

7%|███████
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7%|███████
08:24, 90.78it/s] | 3258/49041 [01:

7%|███████
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7%|███████
[01:06<07:16, 104.71it/s] | 3301/49041

7%|███████
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7% | [██████] | 3327/49041 [01:
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7% | [██████] | 3342/49041 [01:
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7% | [██████] | 3388/49041
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7% | [██████] | 3408/49041
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7% | [██████] | 3423/49041
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7% | [██████] | 3437/49041
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7% | [██████] | 3451/49041
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7% | [██████] | 3477/49041
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7% | [██████] | 3497/49041
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7% | [██████] | 3511/49041 [01:
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7% | [██████] | 3525/49041
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7% | [██████] | 3546/49041
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7% | [██████] | 3561/49041
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7% | [██████] | 3575/49041 [01:
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7% | [██████] | 3586/49041 [01:
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7% | [██████] | 3603/49041 [01:
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7% | [██████] | 3614/49041 [01:*

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[01:12<06:01, 124.00it/s]

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| 3963/49041

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| 3980/49041

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| 3993/49041

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| 4006/49041

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| 4087/49041

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| 4109/49041

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| 4124/49041

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| 4137/49041

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| 4150/49041

8%|███████| [01:14<09:08, 81.80it/s]

| 4162/49041

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| 4175/49041

9%|███████| [01:14<08:01, 93.08it/s]

| 4187/49041

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9%|███████| 4482/49041
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[01:17<05:34, 133.19it/s]	
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[01:17<06:05, 121.77it/s]	
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[01:17<06:35, 112.52it/s]	
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[01:17<06:59, 106.04it/s]	
9% [██████]	4582/49041
[01:17<07:10, 103.30it/s]	
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[01:18<07:52, 94.10it/s]	
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[01:18<08:26, 87.77it/s]	
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[01:18<09:36, 77.00it/s]	
9% [██████]	4624/49041
[01:18<09:15, 79.93it/s]	
9% [██████]	4633/49041
[01:18<11:15, 65.71it/s]	
9% [██████]	4641/49041
[01:18<12:23, 59.69it/s]	
9% [██████]	4653/49041
[01:18<10:34, 69.91it/s]	
10% [██████]	4666/49041
[01:18<09:17, 79.59it/s]	
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[01:19<08:16, 89.37it/s]	
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[01:19<07:38, 96.82it/s]	
10% [██████]	4705/49041
[01:19<07:52, 93.82it/s]	
10% [██████]	4716/49041
[01:19<09:53, 74.74it/s]	
10% [██████]	4730/49041
[01:19<09:14, 79.98it/s]	
10% [██████]	4736/49041
[01:19<09:14, 79.98it/s]	

10%|██████████| 4 / 39 / 49041
[01:19<09:10, 80.48it/s]

10%|██████████| 4748 / 49041
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10%|██████████| 4757 / 49041
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10%|██████████| 4785 / 49041
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10%|██████████| 4797 / 49041
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10%|██████████| 4808 / 49041
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10%|██████████| 4818 / 49041
[01:20<09:45, 75.52it/s]

10%|██████████| 4827 / 49041
[01:20<10:29, 70.25it/s]

10%|██████████| 4835 / 49041
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10%|██████████| 4843 / 49041
[01:21<12:17, 59.92it/s]

10%|██████████| 4850 / 49041
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10%|██████████| 4857 / 49041
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10%|██████████| 4870 / 49041
[01:21<09:59, 73.72it/s]

10%|██████████| 4879 / 49041
[01:21<10:28, 70.29it/s]

10%|██████████| 4899 / 49041
[01:21<08:27, 86.93it/s]

10%|██████████| 4911 / 49041
[01:21<07:56, 92.52it/s]

10%|██████████| 4923 / 49041
[01:22<08:19, 88.27it/s]

10%|██████████| 4934 / 49041
[01:22<08:57, 81.98it/s]

10%|██████████| 4944 / 49041
[01:22<09:34, 85.44it/s]

[01:22<08:32, 86.11it/s]

10%|██████████ [01:22<07:39, 95.84it/s]

| 4960/49041

10%|██████████ [01:22<07:03, 104.00it/s]

| 4976/49041

10%|██████████ [01:22<06:51, 106.93it/s]

| 4989/49041

10%|██████████ [01:22<06:01, 121.75it/s]

| 5007/49041

10%|██████████ [01:22<06:11, 118.36it/s]

| 5021/49041

10%|██████████ [01:23<06:28, 113.37it/s]

| 5034/49041

10%|██████████ [01:23<06:35, 111.32it/s]

| 5046/49041

10%|██████████ [01:23<08:27, 86.66it/s]

| 5058/49041

10%|██████████ [01:23<08:17, 88.33it/s]

| 5068/49041

10%|██████████ [01:23<07:45, 94.35it/s]

| 5080/49041

10%|██████████ [01:23<06:49, 107.20it/s]

| 5100/49041

10%|██████████ [01:23<07:43, 94.71it/s]

| 5112/49041

10%|██████████ [01:24<08:17, 88.20it/s]

| 5123/49041

10%|██████████ [01:24<10:18, 70.95it/s]

| 5133/49041

10%|██████████ [01:24<11:04, 66.09it/s]

| 5142/49041

11%|██████████ [01:24<09:38, 75.83it/s]

| 5154/49041

11%|██████████ [01:24<07:59, 91.50it/s]

| 5174/49041

11%|██████████ [01:24<06:58, 104.78it/s]

| 5190/49041

11%|██████████ [01:24<06:47, 107.67it/s]

| 5203/49041

11% [███████]	[01:24<06:38, 110.09it/s]	5216/49041
11% [███████]	[01:25<06:55, 105.37it/s]	5229/49041
11% [███████]	[01:25<07:33, 96.58it/s]	5241/49041
11% [███████]	[01:25<07:32, 96.76it/s]	5252/49041
11% [███████]	[01:25<08:21, 87.21it/s]	5263/49041
11% [███████]	[01:25<06:48, 107.16it/s]	5295/49041
11% [███████]	[01:25<06:27, 112.84it/s]	5308/49041
11% [███████]	[01:25<05:27, 133.57it/s]	5333/49041
11% [███████]	[01:26<05:50, 124.68it/s]	5350/49041
11% [███████]	[01:26<06:28, 112.29it/s]	5365/49041
11% [███████]	[01:26<05:49, 124.76it/s]	5382/49041
11% [███████]	[01:26<06:07, 118.89it/s]	5397/49041
11% [███████]	[01:26<06:18, 115.15it/s]	5411/49041
11% [███████]	[01:26<07:12, 100.86it/s]	5424/49041
11% [███████]	[01:26<07:39, 94.85it/s]	5435/49041
11% [███████]	[01:27<08:03, 90.20it/s]	5446/49041
11% [███████]	[01:27<06:52, 105.64it/s]	5465/49041
11% [███████]	[01:27<05:52, 123.49it/s]	5487/49041

11% [███████]	5502/49041
[01:27<06:58, 103.96it/s]	
11% [███████]	5515/49041
[01:27<06:38, 109.27it/s]	
11% [███████]	5528/49041
[01:27<06:34, 110.38it/s]	
11% [███████]	5547/49041
[01:27<05:57, 121.83it/s]	
11% [███████]	5561/49041
[01:27<05:51, 123.53it/s]	
11% [███████]	5575/49041
[01:28<06:15, 115.66it/s]	
11% [███████]	5588/49041
[01:28<07:07, 101.68it/s]	
11% [███████]	5599/49041
[01:28<06:58, 103.76it/s]	
11% [███████]	5610/49041
[01:28<07:15, 99.73it/s]	
11% [███████]	5637/49041
[01:28<05:58, 121.22it/s]	
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[01:28<05:47, 124.88it/s]	
12% [███████]	5667/49041
[01:28<06:20, 113.99it/s]	
12% [███████]	5686/49041
[01:28<05:45, 125.54it/s]	
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[01:29<06:20, 113.89it/s]	
12% [███████]	5713/49041
[01:29<06:08, 117.67it/s]	
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[01:29<05:30, 131.11it/s]	
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[01:29<07:32, 95.66it/s]	
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[01:29<06:18, 114.13it/s]	

12% [██████████]	5801/49041
[01:29<06:27, 111.60it/s]	
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[01:30<06:31, 110.33it/s]	
12% [██████████]	5829/49041
[01:30<06:09, 117.03it/s]	
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[01:30<06:12, 116.05it/s]	
12% [██████████]	5860/49041
[01:30<05:33, 129.64it/s]	
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[01:30<05:42, 126.14it/s]	
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[01:30<05:52, 122.50it/s]	
12% [██████████]	5918/49041
[01:30<04:49, 148.72it/s]	
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[01:30<04:55, 145.70it/s]	
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[01:30<05:39, 126.96it/s]	
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[01:31<08:00, 89.56it/s]	
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[01:31<07:53, 90.85it/s]	
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[01:31<06:54, 103.75it/s]	
12% [██████████]	6018/49041
[01:31<06:08, 116.62it/s]	
12% [██████████]	6036/49041
[01:31<05:29, 130.33it/s]	
12% [██████████]	6053/49041
[01:31<05:09, 138.96it/s]	
12% [██████████]	6069/49041
[01:31<05:41, 125.98it/s]	
12% [██████████]	6085/49041
[01:32<05:27, 131.14it/s]	
12% [██████████]	6109/49041
[01:32<04:43, 151.52it/s]	
12% [██████████]	6126/49041

[01:32<04:56, 144.67it/s]

13%|███████████| [01:32<05:48, 123.18it/s]

| 6142/49041

13%|███████████| [01:32<07:04, 101.03it/s]

| 6156/49041

13%|███████████| [01:32<07:53, 90.49it/s]

| 6168/49041

13%|███████████| [01:33<07:59, 89.32it/s]

| 6181/49041

13%|███████████| [01:33<07:46, 91.78it/s]

| 6191/49041

13%|███████████| [01:33<07:21, 97.09it/s]

| 6203/49041

13%|███████████| [01:33<06:53, 103.61it/s]

| 6217/49041

13%|███████████| [01:33<07:17, 97.87it/s]

| 6228/49041

13%|███████████| [01:33<07:03, 100.96it/s]

| 6239/49041

13%|███████████| [01:33<06:58, 102.33it/s]

| 6254/49041

13%|███████████| [01:33<06:52, 103.72it/s]

| 6265/49041

13%|███████████| [01:33<08:05, 88.06it/s]

| 6276/49041

13%|███████████| [01:34<06:33, 108.57it/s]

| 6300/49041

13%|███████████| [01:34<06:40, 106.59it/s]

| 6314/49041

13%|███████████| [01:34<06:20, 112.16it/s]

| 6327/49041

13%|███████████| [01:34<05:28, 130.14it/s]

| 6348/49041

13%|███████████| [01:34<05:14, 135.81it/s]

| 6364/49041

13%|███████████| [01:34<06:01, 117.94it/s]

| 6380/49041

13%|███████████| [01:34<05:47, 122.88it/s]

| 6394/49041

13% [██████████]	[01:34<05:42, 124.57it/s]	6408/49041
13% [██████████]	[01:35<05:35, 127.09it/s]	6422/49041
13% [██████████]	[01:35<05:26, 130.62it/s]	6441/49041
13% [██████████]	[01:35<06:49, 104.09it/s]	6455/49041
13% [██████████]	[01:35<06:11, 114.49it/s]	6471/49041
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13% [██████████]	[01:35<07:16, 97.41it/s]	6499/49041
13% [██████████]	[01:35<07:43, 91.82it/s]	6510/49041
13% [██████████]	[01:36<07:19, 96.83it/s]	6523/49041
13% [██████████]	[01:36<06:24, 110.43it/s]	6540/49041
13% [██████████]	[01:36<07:06, 99.55it/s]	6553/49041
13% [██████████]	[01:36<06:16, 112.71it/s]	6570/49041
13% [██████████]	[01:36<06:05, 116.16it/s]	6583/49041
13% [██████████]	[01:36<06:00, 117.80it/s]	6597/49041
13% [██████████]	[01:36<05:58, 118.47it/s]	6612/49041
14% [██████████]	[01:36<04:58, 141.95it/s]	6644/49041
14% [██████████]	[01:36<04:21, 162.12it/s]	6669/49041
14% [██████████]	[01:37<03:49, 184.28it/s]	6697/49041
14% [██████████]	[01:37<03:34, 197.43it/s]	6721/49041

14% | ██████████ | 6744/49041
[01:37<03:57, 178.38it/s]

14% | ██████████ | 6764/49041
[01:37<05:11, 135.51it/s]

14% | ██████████ | 6786/49041
[01:37<04:36, 152.90it/s]

14% | ██████████ | 6805/49041
[01:37<04:55, 143.01it/s]

14% | ██████████ | 6836/49041
[01:37<04:21, 161.68it/s]

14% | ██████████ | 6864/49041
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14% | ██████████ | 6888/49041
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14% | ██████████ | 6910/49041
[01:38<03:50, 183.14it/s]

14% | ██████████ | 6930/49041
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14% | ██████████ | 6950/49041
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14% | ██████████ | 6968/49041
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14% | ██████████ | 6986/49041
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14% | ██████████ | 7003/49041
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14% | ██████████ | 7039/49041
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14% | ██████████ | 7061/49041
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14% | ██████████ | 7081/49041
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14% | ██████████ | 7099/49041
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15% | ██████████ | 7140/49041
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[01:39<04:59, 140.01it/s]

15% [██████████] | 7174/49041
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15% [██████████] | 7190/49041
[01:40<04:49, 144.80it/s]

15% [██████████] | 7206/49041
[01:40<04:56, 141.23it/s]

15% [██████████] | 7221/49041
[01:40<05:00, 139.06it/s]

15% [██████████] | 7248/49041
[01:40<04:17, 162.47it/s]

15% [██████████] | 7280/49041
[01:40<03:39, 190.30it/s]

15% [██████████] | 7303/49041
[01:40<03:47, 183.28it/s]

15% [██████████] | 7324/49041
[01:40<03:53, 178.68it/s]

15% [██████████] | 7344/49041
[01:40<04:20, 160.19it/s]

15% [██████████] | 7362/49041
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15% [██████████] | 7379/49041
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15% [██████████] | 7398/49041
[01:41<05:01, 138.26it/s]

15% [██████████] | 7419/49041
[01:41<04:39, 149.15it/s]

15% [██████████] | 7443/49041
[01:41<04:23, 157.91it/s]

15% [██████████] | 7460/49041
[01:41<04:29, 154.46it/s]

15% [██████████] | 7483/49041
[01:41<04:03, 170.99it/s]

15% [██████████] | 7508/49041
[01:41<03:45, 183.92it/s]

15% [██████████] | 7528/49041
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15% [██████████] | 7548/49041

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15% | ██████████
[01:42<05:47, 119.33it/s]

| 7565/49041

15% | ██████████
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| 7579/49041

16% | ██████████
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| 7608/49041

16% | ██████████
[01:42<04:14, 162.45it/s]

| 7635/49041

16% | ██████████
[01:42<04:04, 169.61it/s]

| 7655/49041

16% | ██████████
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| 7681/49041

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| 7702/49041

16% | ██████████
[01:43<03:36, 190.88it/s]

| 7729/49041

16% | ██████████
[01:43<03:16, 210.50it/s]

| 7760/49041

16% | ██████████
[01:43<03:38, 188.81it/s]

| 7783/49041

16% | ██████████
[01:43<03:37, 189.37it/s]

| 7804/49041

16% | ██████████
[01:43<04:00, 171.13it/s]

| 7825/49041

16% | ██████████
[01:43<03:59, 172.17it/s]

| 7844/49041

16% | ██████████
[01:43<03:33, 193.07it/s]

| 7872/49041

16% | ██████████
[01:44<04:18, 159.28it/s]

| 7893/49041

16% | ██████████
[01:44<04:19, 158.58it/s]

| 7911/49041

16% | ██████████
[01:44<05:51, 116.92it/s]

| 7929/49041

16% | ██████████
[01:44<05:12, 131.30it/s]

| 7950/49041

16% | ██████████
[01:44<04:57, 137.97it/s]

| 7968/49041

16% [01:44<04:17, 159.41it/s]	7994/49041
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16% [01:45<03:57, 172.88it/s]	8034/49041
16% [01:45<03:15, 209.02it/s]	8076/49041
17% [01:45<04:14, 160.55it/s]	8102/49041
17% [01:45<04:39, 146.45it/s]	8123/49041
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17% [01:45<03:59, 170.26it/s]	8193/49041
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17% [01:46<04:09, 163.38it/s]	8278/49041
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17% [01:46<03:33, 191.11it/s]	8330/49041
17% [01:46<03:37, 186.90it/s]	8352/49041
17% [01:46<03:38, 186.23it/s]	8373/49041
17% [01:47<03:26, 196.47it/s]	8399/49041
17% [01:47<03:47, 178.85it/s]	8420/49041

17% [██████████]	[01:47<03:17, 205.27it/s]	8452/49041
17% [██████████]	[03:58, 169.79it/s]	8475/49041 [01:4
17% [██████████]	[03:43, 181.34it/s]	8499/49041 [01:4
17% [██████████]	[04:09, 162.51it/s]	8520/49041 [01:4
17% [██████████]	[03:45, 179.28it/s]	8547/49041 [01:4
17% [██████████]	[03:37, 185.89it/s]	8572/49041 [01:4
18% [██████████]	[03:35, 187.71it/s]	8592/49041 [01:4
18% [██████████]	[03:17, 205.15it/s]	8620/49041 [01:4
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18% [██████████]	[04:32, 148.28it/s]	8662/49041 [01:4
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18% [██████████]	[04:31, 148.55it/s]	8697/49041 [01:4
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18% [01:49<03:55, 170.81it/s]		8740/49041
18% [01:49<04:17, 156.18it/s]		8758/49041
18% [01:49<04:14, 158.36it/s]		8775/49041
18% [03:24, 196.70it/s]		8822/49041 [01:4
18% [03:29, 192.02it/s]		8848/49041 [01:4
18% [03:25, 195.81it/s]		8872/49041 [01:4

18% [REDACTED]	8895/49041 [01:4
03:32, 189.21it/s]	
18% [REDACTED]	8916/49041 [01:4
03:29, 191.88it/s]	
18% [REDACTED]	8937/49041 [01:4
03:27, 192.83it/s]	
18% [REDACTED]	8958/49041 [01:5
03:34, 186.47it/s]	
18% [REDACTED]	8980/49041 [01:5
03:41, 180.77it/s]	
18% [REDACTED]	8999/49041 [01:5
04:01, 165.54it/s]	
18% [REDACTED]	9024/49041 [01:5
03:41, 180.91it/s]	
18% [REDACTED]	9043/49041
[01:50<03:45, 177.41it/s]	
19% [REDACTED]	9076/49041
[01:50<03:14, 205.62it/s]	
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[01:50<02:49, 235.81it/s]	
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02:41, 246.51it/s]	
19% [REDACTED]	9173/49041 [01:5
03:11, 208.55it/s]	
19% [REDACTED]	9197/49041 [01:5
03:25, 193.50it/s]	
19% [REDACTED]	9219/49041 [01:5
03:37, 183.47it/s]	
19% [REDACTED]	9239/49041 [01:5
03:34, 185.24it/s]	
19% [REDACTED]	9259/49041 [01:5
03:34, 185.77it/s]	
19% [REDACTED]	9282/49041 [01:5
03:22, 196.17it/s]	
19% [REDACTED]	9303/49041 [01:5
03:56, 167.74it/s]	
19% [REDACTED]	9326/49041 [01:5
03:47, 174.95it/s]	
10% [REDACTED]	9356/49041 [01:5

19%|███████████| 03:21, 197.23it/s] | 9393/49041 [01:5

19%|███████████| 02:55, 226.02it/s] | 9419/49041 [01:5

19%|███████████| 03:04, 214.24it/s] | 9452/49041 [01:5

19%|███████████| 02:45, 238.92it/s] | 9479/49041 [01:5

19%|███████████| 03:15, 202.44it/s] | 9502/49041 [01:5

19%|███████████| 03:08, 209.76it/s] | 9527/49041 [01:5

19%|███████████| 03:07, 210.95it/s] | 9550/49041 [01:5

20%|███████████| 03:14, 203.18it/s] | 9572/49041 [01:5

20%|███████████| 04:14, 154.87it/s] | 9604/49041 [01:5

20%|███████████| 03:39, 179.93it/s] | 9626/49041 [01:5

20%|███████████| 03:37, 181.21it/s] | 9661/49041 [01:5

20%|███████████| 03:15, 200.99it/s] | 9685/49041 [01:5

20%|███████████| 03:09, 207.42it/s] | 9708/49041 [01:5

20%|███████████| 02:59, 218.62it/s] | 9734/49041 [01:5

20%|███████████| 03:04, 213.27it/s] | 9757/49041 [01:5

20%|███████████| 02:40, 244.31it/s] | 9794/49041 [01:5

20%|███████████| 03:16, 199.66it/s] | 9821/49041 [01:5

20%|███████████| 03:29, 186.70it/s] | 9844/49041 [01:5

20%|███████████| 03:25, 190.92it/s] | 9872/49041 [01:5

03:29, 185.29it/s]

| 9893/49041 [01:5

20%|███████████| 03:31, 185.29it/s]

| 9917/49041 [01:5

20%|███████████| 03:24, 191.37it/s]

| 9937/49041 [01:5

20%|███████████| 03:57, 164.33it/s]

| 9970/49041 [01:5

03:21, 193.45it/s]

| 10000/49041 [01:5

20%|███████████| 03:01, 215.63it/s]

| 10033/49041 [01:5

02:42, 240.16it/s]

| 10060/49041 [01:5

21%|███████████| 02:57, 219.44it/s]

| 10085/49041 [01:5

02:54, 223.88it/s]

| 10115/49041 [01:5

02:44, 235.92it/s]

| 10140/49041 [01:5

21%|███████████| 02:45, 235.45it/s]

| 10165/49041 [01:5

03:09, 205.11it/s]

| 10197/49041 [01:5

02:52, 224.58it/s]

| 10227/49041 [01:5

02:49, 228.36it/s]

| 10251/49041 [01:5

02:56, 220.13it/s]

| 10279/49041 [01:5

02:45, 234.60it/s]

| 10304/49041 [01:5

03:12, 200.87it/s]

| 10326/49041 [01:5

03:19, 193.86it/s]

| 10351/49041 [01:5

03:06, 207.38it/s]

| 10373/49041 [01:5

03:19, 193.88it/s]

21%|███████████| 10399/49041 [01:5
03:07, 206.59it/s]

21%|███████████| 10424/49041 [01:5
02:57, 217.79it/s]

21%|███████████| 10458/49041 [01:5
02:44, 234.29it/s]

21%|███████████| 10495/49041 [01:5
02:30, 256.94it/s]

21%|███████████| 10522/49041 [01:5
02:37, 243.85it/s]

22%|███████████| 10548/49041 [01:5
02:41, 238.69it/s]

22%|███████████| 10573/49041 [01:5
03:01, 212.41it/s]

22%|███████████| 10596/49041 [01:5
03:20, 191.40it/s]

22%|███████████| 10617/49041 [01:5
03:33, 179.98it/s]

22%|███████████| 10643/49041 [01:5
03:52, 165.18it/s]

22%|███████████| 10661/49041 [01:5
03:55, 162.77it/s]

22%|███████████| 10678/49041 [01:5
05:36, 114.15it/s]

22%|███████████| 10692/49041 [01:5
05:31, 115.82it/s]

22%|███████████| 10714/49041 [01:5
04:44, 134.52it/s]

22%|███████████| 10734/49041 [01:5
04:17, 148.85it/s]

22%|███████████| 10752/49041 [01:5
04:17, 148.85it/s]

22%|███████████| 10773/49041 [01:5
04:00, 159.05it/s]

22%|███████████| 10793/49041 [01:5
03:50, 166.11it/s]

22%|███████████| 10811/49041 [01:5
04:10, 152.71it/s]

22% [██████████] 04:09, 153.28it/s]	10828/49041 [01:5
22% [██████████] 04:23, 145.13it/s]	10844/49041 [01:5
22% [██████████] 03:55, 162.14it/s]	10869/49041 [01:5
22% [██████████] 04:01, 158.29it/s]	10887/49041 [01:5
22% [██████████] 03:32, 179.69it/s]	10916/49041 [01:5
22% [██████████] 03:13, 196.89it/s]	10948/49041 [01:5
22% [██████████] 03:18, 191.97it/s]	10970/49041 [02:0
22% [██████████] 03:14, 195.80it/s]	10993/49041 [02:0
22% [██████████] 02:58, 213.21it/s]	11020/49041 [02:0
23% [██████████] 03:11, 198.20it/s]	11043/49041 [02:0
23% [██████████] 02:54, 217.12it/s]	11071/49041 [02:0
23% [██████████] 02:52, 219.51it/s]	11096/49041 [02:0
23% [██████████] 02:56, 214.63it/s]	11119/49041 [02:0
23% [██████████] 03:28, 181.76it/s]	11142/49041 [02:0
23% [██████████] 03:28, 181.73it/s]	11162/49041 [02:0
23% [██████████] 03:15, 193.38it/s]	11187/49041 [02:0
23% [██████████] 02:59, 210.42it/s]	11214/49041 [02:0
23% [██████████] 02:46, 226.85it/s]	11242/49041 [02:0
23% [██████████] 02:56, 214.45it/s]	11266/49041 [02:0
23% [██████████]	11291/49041 [02:0

23% [██████████]	03:08, 200.69it/s]	11289/49041 [02:c]
23% [██████████]	03:04, 204.61it/s]	11313/49041 [02:c]
23% [██████████]	03:30, 178.83it/s]	11335/49041 [02:c]
23% [██████████]	03:47, 165.62it/s]	11354/49041 [02:c]
23% [██████████]	03:46, 165.96it/s]	11372/49041 [02:c]
23% [██████████]	03:45, 167.09it/s]	11390/49041 [02:c]
23% [██████████]	03:50, 162.92it/s]	11408/49041 [02:c]
23% [██████████]	03:50, 163.39it/s]	11425/49041 [02:c]
23% [██████████]	03:24, 183.68it/s]	11452/49041 [02:c]
23% [██████████]	02:46, 225.39it/s]	11500/49041 [02:c]
24% [██████████]	02:33, 245.16it/s]	11531/49041 [02:c]
24% [██████████]	02:35, 241.43it/s]	11560/49041 [02:c]
24% [██████████]	02:41, 231.31it/s]	11588/49041 [02:c]
24% [██████████]	02:44, 227.37it/s]	11614/49041 [02:c]
24% [██████████]	02:46, 224.82it/s]	11639/49041 [02:c]
24% [██████████]	02:43, 228.10it/s]	11663/49041 [02:c]
24% [██████████]	02:50, 218.87it/s]	11687/49041 [02:c]
24% [██████████]	03:11, 195.32it/s]	11710/49041 [02:c]
24% [██████████]	03:07, 198.90it/s]	11733/49041 [02:c]
24% [██████████]	03:06, 198.90it/s]	11754/49041 [02:c]

03:26, 180.60it/s]

24%|███████████| 03:08, 197.14it/s]

| 11781/49041 [02:C

24%|███████████| 03:30, 176.65it/s]

| 11802/49041 [02:C

24%|███████████| 03:42, 167.40it/s]

| 11821/49041 [02:C

24%|███████████| 03:26, 180.10it/s]

| 11844/49041 [02:C

24%|███████████| 03:40, 168.28it/s]

| 11863/49041 [02:C

24%|███████████| 03:50, 161.51it/s]

| 11881/49041 [02:C

24%|███████████| 03:25, 180.69it/s]

| 11906/49041 [02:C

24%|███████████| 03:33, 173.54it/s]

| 11926/49041 [02:C

24%|███████████| 03:54, 157.97it/s]

| 11945/49041 [02:C

24%|███████████| 03:38, 169.85it/s]

| 11966/49041 [02:C

24%|███████████| 04:11, 147.50it/s]

| 11984/49041 [02:C

24%|███████████| 03:51, 159.86it/s]

| 12006/49041 [02:C

25%|███████████| 03:44, 164.86it/s]

| 12028/49041 [02:C

25%|███████████| 03:49, 161.38it/s]

| 12046/49041 [02:C

25%|███████████| 04:14, 145.41it/s]

| 12063/49041 [02:C

25%|███████████| 03:39, 168.48it/s]

| 12090/49041 [02:C

25%|███████████| 03:25, 179.92it/s]

| 12115/49041 [02:C

25%|███████████| 03:12, 192.11it/s]

| 12141/49041 [02:C

25%|███████████| 03:05, 198.79it/s]

| 12173/49041 [02:C

25% [███████████]	12198/49041 [02:00:55, 209.70it/s]
25% [███████████]	12220/49041 [02:00:00, 203.76it/s]
25% [███████████]	12241/49041 [02:00:01, 202.31it/s]
25% [███████████]	12264/49041 [02:00:56, 208.92it/s]
25% [███████████]	12286/49041 [02:00:52, 158.05it/s]
25% [███████████]	12304/49041 [02:00:05, 149.88it/s]
25% [███████████]	12329/49041 [02:00:38, 167.96it/s]
25% [███████████]	12367/49041 [02:00:04, 198.59it/s]
25% [███████████]	12395/49041 [02:00:50, 215.08it/s]
25% [███████████]	12421/49041 [02:00:44, 222.20it/s]
25% [███████████]	12446/49041 [02:00:04, 198.74it/s]
25% [███████████]	12468/49041 [02:00:03, 199.23it/s]
25% [███████████]	12490/49041 [02:00:00, 202.81it/s]
26% [███████████]	12512/49041 [02:00:07, 194.49it/s]
26% [███████████]	12533/49041 [02:00:08, 193.44it/s]
26% [███████████]	12555/49041 [02:00:02, 200.19it/s]
26% [███████████]	12576/49041 [02:00:40, 165.70it/s]
26% [███████████]	12594/49041 [02:00:47, 160.50it/s]
26% [███████████]	12633/49041 [02:00:07, 194.63it/s]

26% [██████████] 03:21, 180.33it/s]	12657/49041 [02:C]
26% [██████████] 03:18, 183.34it/s]	12679/49041 [02:C]
26% [██████████] 02:52, 211.16it/s]	12715/49041 [02:C]
26% [██████████] 02:49, 213.58it/s]	12740/49041 [02:C]
26% [██████████] 02:25, 248.36it/s]	12782/49041 [02:C]
26% [██████████] 02:17, 263.16it/s]	12815/49041 [02:C]
26% [██████████] 02:23, 253.02it/s]	12845/49041 [02:C]
26% [██████████] 02:12, 273.87it/s]	12880/49041 [02:C]
26% [██████████] 02:18, 261.42it/s]	12910/49041 [02:C]
26% [██████████] 02:41, 223.75it/s]	12938/49041 [02:C]
26% [██████████] 02:38, 227.63it/s]	12963/49041 [02:C]
26% [██████████] 02:44, 219.46it/s]	12988/49041 [02:1]
27% [██████████] 02:31, 237.58it/s]	13018/49041 [02:1]
27% [██████████] 02:45, 217.63it/s]	13043/49041 [02:1]
27% [██████████] 02:34, 233.31it/s]	13073/49041 [02:1]
27% [██████████] 02:26, 244.99it/s]	13101/49041 [02:1]
27% [██████████] 02:11, 272.25it/s]	13144/49041 [02:1]
27% [██████████] 02:12, 270.33it/s]	13173/49041 [02:1]
27% [██████████] 02:34, 232.49it/s]	13202/49041 [02:1]

27% [REDACTED]	13237/49041 [02:1
02:18, 258.04it/s]	
27% [REDACTED]	13265/49041 [02:1
02:26, 244.26it/s]	
27% [REDACTED]	13292/49041 [02:1
02:30, 238.18it/s]	
27% [REDACTED]	13328/49041 [02:1
02:14, 264.55it/s]	
27% [REDACTED]	13357/49041 [02:1
02:16, 260.50it/s]	
27% [REDACTED]	13389/49041 [02:1
02:20, 254.62it/s]	
27% [REDACTED]	13429/49041 [02:1
02:23, 248.76it/s]	
27% [REDACTED]	13455/49041 [02:1
02:32, 232.86it/s]	
27% [REDACTED]	13479/49041 [02:1
03:16, 180.88it/s]	
28% [REDACTED]	13500/49041 [02:1
03:30, 168.77it/s]	
28% [REDACTED]	13537/49041 [02:1
02:57, 199.68it/s]	
28% [REDACTED]	13572/49041 [02:1
02:34, 228.85it/s]	
28% [REDACTED]	13599/49041 [02:1
03:25, 172.77it/s]	
28% [REDACTED]	13622/49041 [02:1
03:13, 183.38it/s]	
28% [REDACTED]	13644/49041 [02:1
03:19, 177.20it/s]	
28% [REDACTED]	13665/49041 [02:1
03:15, 180.67it/s]	
28% [REDACTED]	13692/49041 [02:1
02:59, 197.08it/s]	
28% [REDACTED]	13714/49041 [02:1
02:58, 197.45it/s]	
28% [REDACTED]	13735/49041 [02:1
03:01, 194.00it/s]	
28% [REDACTED]	13756/49041 [02:1

03:06, 189.33it/s] | 13776/49041 [02:1
28%|███████████ | 03:11, 184.03it/s]
28%|███████████ | 03:15, 180.51it/s] | 13795/49041 [02:1
28%|███████████ | 02:41, 217.94it/s] | 13841/49041 [02:1
28%|███████████ | 02:31, 232.91it/s] | 13869/49041 [02:1
28%|███████████ | 02:25, 242.28it/s] | 13900/49041 [02:1
28%|███████████ | 02:19, 251.85it/s] | 13928/49041 [02:1
28%|███████████ | 02:40, 218.79it/s] | 13955/49041 [02:1
29%|███████████ | 03:30, 166.85it/s] | 13979/49041 [02:1
29%|███████████ | 03:20, 174.44it/s] | 13999/49041 [02:1
29%|███████████ | 02:49, 206.64it/s] | 14036/49041 [02:1
29%|███████████ | 02:37, 222.17it/s] | 14064/49041 [02:1
29%|███████████ | 02:29, 233.44it/s] | 14091/49041 [02:1
29%|███████████ | 02:47, 209.09it/s] | 14117/49041 [02:1
29%|███████████ | 02:57, 196.68it/s] | 14141/49041 [02:1
29%|███████████ | 02:44, 211.41it/s] | 14169/49041 [02:1
29%|███████████ | 02:33, 227.65it/s] | 14197/49041 [02:1
29%|███████████ | 02:40, 216.36it/s] | 14222/49041 [02:1
29%|███████████ | 02:51, 203.42it/s] | 14245/49041 [02:1
29%|███████████ | 03:01, 191.45it/s] | 14267/49041 [02:1

29% [███████████]	02:37, 220.37it/s]	14305/49041 [02:1
29% [███████████]	02:23, 241.68it/s]	14339/49041 [02:1
29% [███████████]	02:12, 262.18it/s]	14372/49041 [02:1
29% [███████████]	02:47, 207.35it/s]	14401/49041 [02:1
29% [███████████]	02:46, 207.51it/s]	14425/49041 [02:1
29% [███████████]	03:18, 174.43it/s]	14448/49041 [02:1
30% [███████████]	02:40, 215.65it/s]	14497/49041 [02:1
30% [███████████]	02:38, 218.35it/s]	14526/49041 [02:1
30% [███████████]	02:50, 202.57it/s]	14553/49041 [02:1
30% [███████████]	02:27, 232.88it/s]	14589/49041 [02:1
30% [███████████]	02:15, 254.88it/s]	14622/49041 [02:1
30% [███████████]	02:15, 253.34it/s]	14651/49041 [02:1
30% [███████████]	02:30, 228.86it/s]	14679/49041 [02:1
30% [███████████]	02:11, 260.82it/s]	14718/49041 [02:1
30% [███████████]	02:09, 265.73it/s]	14748/49041 [02:1
30% [███████████]	02:04, 274.14it/s]	14780/49041 [02:1
30% [███████████]	02:26, 234.33it/s]	14809/49041 [02:1
30% [███████████]	02:25, 235.14it/s]	14835/49041 [02:1
30% [███████████]	02:26, 234.07it/s]	14860/49041 [02:1

30% [██████████] | 14890/49041 [02:1
02:19, 245.64it/s]

30% [██████████] | 14919/49041 [02:1
02:15, 251.69it/s]

30% [██████████] | 14945/49041 [02:1
02:26, 232.26it/s]

31% [██████████] | 14977/49041 [02:1
02:18, 246.22it/s]

31% [██████████] | 15003/49041 [02:1
02:19, 244.38it/s]

31% [██████████] | 15059/49041 [02:1
01:56, 291.77it/s]

31% [██████████] | 15093/49041 [02:1
02:13, 253.59it/s]

31% [██████████] | 15123/49041 [02:1
02:37, 215.37it/s]

31% [██████████] | 15174/49041 [02:1
02:11, 256.77it/s]

31% [██████████] | 15217/49041 [02:1
01:59, 282.87it/s]

31% [██████████] | 15251/49041 [02:1
02:41, 208.72it/s]

31% [██████████] | 15279/49041 [02:1
02:50, 198.24it/s]

31% [██████████] | 15319/49041 [02:1
02:26, 230.01it/s]

31% [██████████] | 15348/49041 [02:2
02:18, 242.60it/s]

31% [██████████] | 15388/49041 [02:2
02:04, 269.54it/s]

31% [██████████] | 15419/49041 [02:2
02:14, 249.17it/s]

31% [██████████] | 15447/49041 [02:2
02:40, 209.80it/s]

32% [██████████] | 15471/49041 [02:2
02:50, 197.11it/s]

32% [██████████] | 15497/49041 [02:2
02:38, 211.52it/s]

32% [REDACTED]	15527/49041 [02:2
02:24, 231.54it/s]	
32% [REDACTED]	15574/49041 [02:2
02:02, 273.03it/s]	
32% [REDACTED]	15606/49041 [02:2
02:20, 237.86it/s]	
32% [REDACTED]	15649/49041 [02:2
02:01, 273.76it/s]	
32% [REDACTED]	15682/49041 [02:2
02:10, 256.12it/s]	
32% [REDACTED]	15712/49041 [02:2
02:07, 260.94it/s]	
32% [REDACTED]	15760/49041 [02:2
01:51, 299.56it/s]	
32% [REDACTED]	15794/49041
[02:21<01:49, 302.41it/s]	
32% [REDACTED]	15827/49041
[02:21<01:49, 302.56it/s]	
32% [REDACTED]	15860/49041
[02:21<02:01, 272.00it/s]	
32% [REDACTED]	15900/49041
[02:22<01:52, 293.48it/s]	
32% [REDACTED]	15932/49041
[02:22<01:59, 276.09it/s]	
33% [REDACTED]	15971/49041
[02:22<01:52, 294.38it/s]	
33% [REDACTED]	16002/49041
[02:22<01:54, 289.57it/s]	
33% [REDACTED]	16032/49041
[02:22<01:54, 287.19it/s]	
33% [REDACTED]	16064/49041
[02:22<01:51, 294.78it/s]	
33% [REDACTED]	16095/49041
[02:22<02:00, 273.36it/s]	
33% [REDACTED]	16124/49041
[02:22<02:01, 271.71it/s]	
33% [REDACTED]	16152/49041
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33% [02:23<02:02, 268.93it/s]

| 16229/49041

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| 16285/49041

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| 16324/49041

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| 16394/49041

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| 16426/49041

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| 16456/49041

34% [02:24<01:57, 276.23it/s]

| 16486/49041

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| 16529/49041

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| 16606/49041

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| 16640/49041

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| 16822/49041

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| 16864/49041

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| 16900/49041

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| 16932/49041

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[02:35<01:49, 201.11it/s]

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| 19785/49041

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| 19816/49041

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| 19868/49041

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| 20062/49041

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| 20164/49041

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| 20325/49041

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| 23962/49041 [02:4

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| 24106/49041 [02:4

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| 24353/49041 [02:4

| 24447/49041 [02:4

| 24524/49041 [02:4

| 24594/49041 [02:4

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| 24917/49041 [02:4

| 25016/49041 [02:4

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[02:54<00:29, 616.46it/s]	62% [REDACTED]	30640/49041
[02:54<00:30, 603.41it/s]	63% [REDACTED]	30704/49041
[02:54<00:27, 658.07it/s]	63% [REDACTED]	30788/49041
[02:55<00:32, 560.21it/s]	63% [REDACTED]	30857/49041
[02:55<00:39, 457.99it/s]	63% [REDACTED]	30918/49041
[02:55<00:35, 503.21it/s]	63% [REDACTED]	30984/49041

63% [02:55<00:35, 507.80it/s]	31040/49041
63% [02:55<00:37, 484.17it/s]	31095/49041
64% [02:55<00:34, 512.70it/s]	31155/49041
64% [02:55<00:38, 463.11it/s]	31209/49041
64% [02:55<00:37, 474.72it/s]	31273/49041
64% [02:56<00:44, 398.10it/s]	31323/49041
64% [02:56<00:42, 411.24it/s]	31380/49041
64% [02:56<00:41, 420.16it/s]	31426/49041
64% [02:56<00:42, 409.36it/s]	31470/49041
64% [02:56<00:39, 439.41it/s]	31532/49041
64% [02:56<00:41, 423.34it/s]	31578/49041
64% [02:56<00:47, 368.61it/s]	31622/49041
65% [02:57<00:46, 376.49it/s]	31662/49041
65% [02:57<00:42, 404.85it/s]	31716/49041
65% [02:57<00:43, 400.64it/s]	31759/49041
65% [02:57<00:43, 400.52it/s]	31801/49041
65% [02:57<00:43, 392.95it/s]	31842/49041
65% [02:57<00:44, 388.21it/s]	31882/49041
65% [02:57<00:44, 381.65it/s]	31922/49041

65% [02:57<00:44, 383.05it/s]	31961/49041
65% [02:57<00:45, 374.10it/s]	32000/49041
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65% [02:58<00:41, 408.16it/s]	32090/49041
66% [02:58<00:39, 433.13it/s]	32141/49041
66% [02:58<00:34, 481.93it/s]	32207/49041
66% [02:58<00:31, 526.85it/s]	32279/49041
66% [02:58<00:33, 501.87it/s]	32335/49041
66% [02:58<00:36, 452.61it/s]	32388/49041
66% [02:58<00:36, 450.49it/s]	32441/49041
66% [02:58<00:33, 500.21it/s]	32509/49041
66% [02:59<00:36, 451.34it/s]	32563/49041
66% [02:59<00:35, 457.18it/s]	32612/49041
67% [02:59<00:31, 514.19it/s]	32691/49041
67% [02:59<00:36, 448.68it/s]	32747/49041
67% [02:59<00:33, 483.25it/s]	32810/49041
67% [02:59<00:30, 533.14it/s]	32883/49041
67% [02:59<00:31, 503.78it/s]	32941/49041
67% [02:59<00:36, 435.82it/s]	32995/49041

67% [02:59<00:34, 459.23it/s]	33048/49041
67% [03:00<00:34, 465.43it/s]	33097/49041
68% [03:00<00:29, 542.20it/s]	33187/49041
68% [03:00<00:30, 524.82it/s]	33248/49041
68% [03:00<00:32, 478.87it/s]	33306/49041
68% [03:00<00:31, 501.74it/s]	33363/49041
68% [03:00<00:29, 523.99it/s]	33422/49041
68% [03:00<00:30, 506.65it/s]	33477/49041
68% [03:00<00:30, 509.10it/s]	33530/49041
68% [03:01<00:32, 481.59it/s]	33583/49041
69% [03:01<00:29, 518.90it/s]	33652/49041
69% [03:01<00:31, 480.32it/s]	33706/49041
69% [03:01<00:33, 459.36it/s]	33756/49041
69% [03:01<00:28, 525.65it/s]	33836/49041
69% [03:01<00:33, 452.47it/s]	33893/49041
69% [03:01<00:33, 457.06it/s]	33944/49041
69% [03:01<00:31, 479.03it/s]	33999/49041
69% [03:01<00:29, 500.27it/s]	34057/49041
70% [03:02<00:29, 498.85it/s]	34110/49041 [03:02<00:29, 498.85it/s]
70%	34162/49041 [03:02<00:29, 498.85it/s]

2<00:31, 469.55it/s]		
70% ███████████	34219/49041 [03:0	
2<00:29, 494.55it/s]		
70% ███████████	34270/49041 [03:0	
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70% ███████████	34325/49041 [03:0	
2<00:29, 499.42it/s]		
70% ███████████	34378/49041 [03:0	
2<00:29, 494.10it/s]		
70% ███████████	34448/49041 [03:0	
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70% ███████████	34528/49041 [03:0	
2<00:25, 578.52it/s]		
71% ███████████	34588/49041 [03:0	
2<00:24, 578.15it/s]		
71% ███████████	34648/49041 [03:0	
3<00:25, 572.91it/s]		
71% ███████████	34739/49041 [03:0	
3<00:22, 631.21it/s]		
71% ███████████	34805/49041 [03:0	
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71% ███████████	34870/49041 [03:0	
3<00:26, 526.90it/s]		
71% ███████████	34927/49041 [03:0	
3<00:29, 479.40it/s]		
71% ███████████	34990/49041 [03:0	
3<00:27, 502.91it/s]		
71% ███████████	35044/49041 [03:0	
3<00:29, 466.97it/s]		
72% ███████████	35152/49041 [03:0	
3<00:24, 561.11it/s]		
72% ███████████	35219/49041 [03:0	
4<00:28, 488.11it/s]		
72% ███████████	35277/49041 [03:0	
4<00:30, 457.91it/s]		
72% ███████████	35337/49041 [03:0	
4<00:28, 481.14it/s]		
72% ███████████	35393/49041 [03:0	
4<00:27, 490.57it/s]		

72% [REDACTED]	35488/49041 [03:0
73% [REDACTED]	35561/49041 [03:0
73% [REDACTED]	35628/49041 [03:0
73% [REDACTED]	35686/49041 [03:0
73% [REDACTED]	35743/49041 [03:0
73% [REDACTED]	35800/49041 [03:0
73% [REDACTED]	35905/49041 [03:0
73% [REDACTED]	35988/49041 [03:0
74% [REDACTED]	36098/49041 [03:0
74% [REDACTED]	36181/49041 [03:0
74% [REDACTED]	36262/49041 [03:0
74% [REDACTED]	36338/49041 [03:0
74% [REDACTED]	36431/49041 [03:0
74% [REDACTED]	36507/49041 [03:0
75% [REDACTED]	36582/49041 [03:0
75% [REDACTED]	36648/49041 [03:0
75% [REDACTED]	36712/49041 [03:0
75% [REDACTED]	36780/49041 [03:0
75% [REDACTED]	36842/49041 [03:0

75% ███████████	6<00:22, 550.58it/s]	36907/49041 [03:C]
75% ███████████	6<00:21, 574.09it/s]	36972/49041 [03:C]
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76% ███████████	7<00:18, 638.20it/s]	37118/49041 [03:C]
76% ███████████	7<00:20, 589.46it/s]	37184/49041 [03:C]
76% ███████████	7<00:20, 579.93it/s]	37246/49041 [03:C]
76% ███████████	7<00:19, 609.89it/s]	37316/49041 [03:C]
76% ███████████	7<00:17, 651.04it/s]	37394/49041 [03:C]
76% ███████████	7<00:17, 649.11it/s]	37461/49041 [03:C]
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77% ███████████	8<00:20, 546.39it/s]	37658/49041 [03:C]
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78% ███████████	8<00:15, 723.24it/s]	38121/49041 [03:C]
78% ███████████	8<00:14, 724.83it/s]	38196/49041 [03:C]

78% [8<00:14, 743.90it/s]	38276/49041 [03:0]
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78% [9<00:16, 631.11it/s]	38421/49041 [03:0]
78% [9<00:16, 623.32it/s]	38487/49041 [03:0]
79% [9<00:16, 626.48it/s]	38551/49041 [03:0]
79% [9<00:17, 582.42it/s]	38615/49041 [03:0]
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79% [9<00:19, 535.58it/s]	38733/49041 [03:0]
79% [9<00:19, 527.43it/s]	38789/49041 [03:0]
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79% [0<00:17, 572.89it/s]	38924/49041 [03:1]
80% [0<00:16, 613.08it/s]	38998/49041 [03:1]
80% [0<00:14, 667.54it/s]	39083/49041 [03:1]
80% [0<00:14, 680.27it/s]	39157/49041 [03:1]
80% [0<00:14, 687.04it/s]	39228/49041 [03:1]
80% [0<00:14, 661.49it/s]	39301/49041 [03:1]
80% [0<00:14, 688.94it/s]	39378/49041 [03:1]
80% [0<00:16, 592.60it/s]	39449/49041 [03:1]
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83%		3<00:11, 679.01it/s]	40921/49041 [03:1]
84%		3<00:13, 603.20it/s]	40998/49041 [03:1]
84%		3<00:11, 700.44it/s]	41112/49041 [03:1]
84%	672 52it/s	3<00:11	41194/49041 [03:1]

3<00:11, 072.32it/s]		
84% ███████████	41342/49041 [03:1	
3<00:09, 792.17it/s]		
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3<00:09, 797.74it/s]		
85% ███████████	41536/49041 [03:1	
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85% ███████████	41919/49041 [03:1	
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86% ███████████	42114/49041 [03:1	
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86% ███████████	42213/49041 [03:1	
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86% ███████████	42307/49041 [03:1	
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4<00:07, 945.28it/s]		
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<00:05, 1047.49it/s]		
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5<00:06, 955.68it/s]		
88% ███████████	43080/49041 [03:1	
5<00:07, 837.50it/s]		
88% ███████████	43169/49041 [03:1	
5<00:06, 843.06it/s]		

88% [██████████] 5<00:06, 943.64it/s]	43301/49041 [03:1
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89% [██████████] 5<00:06, 833.97it/s]	43502/49041 [03:1
89% [██████████] 6<00:07, 774.80it/s]	43592/49041 [03:1
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90% [██████████] [03:16<00:05, 955.12it/s]	44040/49041
90% [██████████] [03:16<00:04, 1012.38it/s]	44160/49041
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90% [██████████] [03:16<00:05, 793.64it/s]	44365/49041
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92% [██████████] [03:17<00:04, 865.90it/s]	45164/49041

92% | ██████████ | 45279/49041
[03:17<00:04, 933.07it/s]

93% | ██████████ | 45394/49041
[03:18<00:03, 986.64it/s]

93% | ██████████ | 45499/49041
[03:18<00:03, 974.25it/s]

93% | ██████████ | 45601/49041
[03:18<00:03, 860.31it/s]

93% | ██████████ | 45693/49041
[03:18<00:04, 714.47it/s]

93% | ██████████ | 45776/49041
[03:18<00:04, 745.47it/s]

94% | ██████████ | 45857/49041
[03:18<00:04, 755.31it/s]

94% | ██████████ | 45937/49041
[03:18<00:04, 732.46it/s]

94% | ██████████ | 46014/49041
[03:18<00:04, 714.50it/s]

94% | ██████████ | 46134/49041
[03:19<00:03, 811.64it/s]

94% | ██████████ | 46266/49041
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95% | ██████████ | 46366/49041
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95% | ██████████ | 46460/49041
[03:19<00:02, 873.32it/s]

95% | ██████████ | 46552/49041
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95% | ██████████ | 46637/49041
[03:19<00:03, 774.54it/s]

95% | ██████████ | 46718/49041
[03:19<00:03, 752.20it/s]

95% | ██████████ | 46827/49041
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96% | ██████████ | 46914/49041
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[03:20<00:02, 847.73it/s]

96% | ██████████ | 47100/49041

96% | [03:20<00:02, 755.76it/s] | 47102/49041 [03:
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21<00:01, 822.25it/s] | 47850/49041 [03:
21<00:01, 766.73it/s] | 47938/49041 [03:
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21<00:00, 957.46it/s] | 48297/49041 [03:
1<00:00, 1041.12it/s] | 48429/49041 [03:2
[03:21<00:00, 1017.24it/s] | 48539/49041
99% | [03:21<00:00, 828.88it/s] | 48645/49041 [03:
21<00:00, 687.06it/s] | 48737/49041 [03:
22<00:00, 656.47it/s] | 48816/49041 [03:
22<00:00, 688.24it/s] | 48910/49041 [03:
22<00:00, 678.81it/s] | 48985/49041 [03:
22<00:00, 688.24it/s] | 49041/49041 [03:
22<00:00, 688.24it/s]

```
22<00:00, 242.18it/s]
```

```
49041  
300
```

```
In [92]:
```

```
average_w2v_on_essay_Xtrain = np.vstack(avg_w2v_vectors_essays_Xtrain)  
print(average_w2v_on_essay_Xtrain.shape)
```

```
(49041, 300)
```

8.3.2. Average Word2Vec encoding of preprocessed_essays on Test Data

```
In [93]:
```

```
# average Word2Vec  
# compute average word2vec for each review.  
avg_w2v_vectors_essays_Xtest = [] # the avg-w2v for each sentence/review is stored in this list  
for sentence in tqdm(X_test['preprocessed_essays'].values): # for each review/sentence  
    vector = np.zeros(300) # as word vectors are of zero length  
    cnt_words = 0; # num of words with a valid vector in the sentence/review  
    for word in sentence.split(): # for each word in a review/sentence  
        if word in glove_words:  
            vector += model[word]  
            cnt_words += 1  
    if cnt_words != 0:  
        vector /= cnt_words  
    avg_w2v_vectors_essays_Xtest.append(vector)  
  
print(len(avg_w2v_vectors_essays_Xtest))  
print(len(avg_w2v_vectors_essays_Xtest[2]))
```

```
0%| [00:00<?, ?it/s] | 0/36  
0%| [0:39, 912.81it/s] | 96/36052 [00:00  
1%| [::33, 1065.84it/s] | 313/36052 [00:00  
1%| [::30, 1165.52it/s] | 468/36052 [00:00  
2%| [::25, 1367.80it/s] | 707/36052 [00:00  
3%| [::22, 1568.51it/s] | 955/36052 [00:00  
3%| [::24, 1448.15it/s] | 1122/36052 [00:00  
4%| [::23, 1504.39it/s] | 1294/36052 [00:00  
4%| [::25, 1350.62it/s] | 1451/36052 [00:00  
4%| [00:01<00:24, 1418.32it/s] | 1618/36052  
5%| [00:01<00:21, 1583.39it/s] | 1844/36052  
6%| [00:01<00:23, 1433.39it/s] | 2013/36052
```

6% [██████]	2167/36052
[00:01<00:23, 1415.45it/s]	
7% [██████]	2351/36052
[00:01<00:22, 1505.90it/s]	
7% [██████]	2510/36052
[00:01<00:22, 1512.81it/s]	
7% [██████]	2698/36052
[00:01<00:20, 1590.77it/s]	
8% [██████]	2869/36052
[00:01<00:20, 1588.66it/s]	
8% [██████]	3031/36052
[00:01<00:20, 1579.36it/s]	
9% [██████]	3192/36052
[00:02<00:25, 1296.82it/s]	
9% [██████]	3352/36052
[00:02<00:24, 1361.09it/s]	
10% [██████]	3540/36052
[00:02<00:22, 1470.15it/s]	
10% [██████]	3695/36052
[00:02<00:21, 1476.18it/s]	
11% [██████]	3849/36052
[00:02<00:22, 1444.51it/s]	
11% [██████]	4038/36052
[00:02<00:20, 1539.33it/s]	
12% [██████]	4262/36052
[00:02<00:18, 1683.51it/s]	
12% [██████]	4438/36052
[00:02<00:19, 1594.62it/s]	
13% [██████]	4604/36052
[00:02<00:20, 1541.88it/s]	
13% [██████]	4781/36052
[00:03<00:19, 1586.69it/s]	
14% [██████]	4944/36052
[00:03<00:21, 1446.37it/s]	
14% [██████]	5117/36052
[00:03<00:20, 1489.85it/s]	
15% [██████]	5340/36052
[00:03<00:19, 1570.73it/s]	
15% [██████]	5536/36052
[00:03<00:18, 1653.53it/s]	
16% [██████]	5723/36052
[00:03<00:17, 1694.53it/s]	
16% [██████]	5909/36052
[00:03<00:17, 1721.77it/s]	
17% [██████]	6084/36052
[00:03<00:18, 1597.55it/s]	
17% [██████]	6248/36052
[00:03<00:18, 1591.52it/s]	
18% [██████]	6420/36052 [00:04
0:19, 1540.63it/s]	
18% [██████]	6588/36052 [00:04
0:18, 1562.35it/s]	

19% [00:04<00:18, 1616.84it/s]	6771/36052
19% [0:18, 1604.87it/s]	6935/36052 [00:04
20% [0:18, 1572.02it/s]	7097/36052 [00:04
20% [0:20, 1394.87it/s]	7256/36052 [00:04
21% [0:19, 1437.65it/s]	7417/36052 [00:04
21% [0:20, 1416.88it/s]	7565/36052 [00:04
22% [0:18, 1557.94it/s]	7776/36052 [00:04
22% [0:18, 1522.87it/s]	7938/36052 [00:05
22% [0:18, 1501.56it/s]	8095/36052 [00:05
23% [0:18, 1505.99it/s]	8277/36052 [00:05
23% [0:17, 1547.79it/s]	8449/36052 [00:05
24% [0:18, 1467.36it/s]	8606/36052 [00:05
24% [0:18, 1456.90it/s]	8755/36052 [00:05
25% [0:16, 1620.02it/s]	8983/36052 [00:05
25% [0:16, 1621.63it/s]	9152/36052 [00:05
26% [0:15, 1717.18it/s]	9359/36052 [00:05
26% [0:16, 1567.08it/s]	9536/36052 [00:06
27% [0:16, 1619.85it/s]	9768/36052 [00:06
28% [0:21, 1203.31it/s]	9935/36052 [00:06
28% [0:20, 1261.84it/s]	10083/36052 [00:06
28% [0:19, 1298.95it/s]	10228/36052 [00:06
29% [00:06<00:18, 1373.74it/s]	10393/36052
29% [00:06<00:17, 1466.50it/s]	10574/36052
30% [00:06<00:16, 1579.01it/s]	10774/36052
30% [00:07<00:16, 1531.70it/s]	10940/36052
31%	11127/36052

[00:07<00:15, 1602.96it/s]		
31% [REDACTED]	11293/36052	
[00:07<00:16, 1547.42it/s]		
32% [REDACTED]	11452/36052	
[00:07<00:15, 1541.92it/s]		
32% [REDACTED]	11609/36052	
[00:07<00:15, 1532.19it/s]		
33% [REDACTED]	11770/36052	
[00:07<00:15, 1536.94it/s]		
33% [REDACTED]	11929/36052	
[00:07<00:15, 1534.72it/s]		
34% [REDACTED]	12084/36052	
[00:07<00:15, 1521.28it/s]		
34% [REDACTED]	12240/36052	
[00:08<00:15, 1498.62it/s]		
35% [REDACTED]	12392/36052	
[00:08<00:16, 1400.75it/s]		
35% [REDACTED]	12543/36052	
[00:08<00:15, 1465.35it/s]		
36% [REDACTED]	12710/36052	
[00:08<00:14, 1557.93it/s]		
36% [REDACTED]	12900/36052	
[00:08<00:13, 1691.80it/s]		
37% [REDACTED]	13120/36052	
[00:08<00:13, 1700.63it/s]		
37% [REDACTED]	13299/36052	
[00:08<00:13, 1695.33it/s]		
38% [REDACTED]	13474/36052	
[00:08<00:12, 1811.11it/s]		
39% [REDACTED]	13698/36052	
[00:08<00:11, 1959.71it/s]		
39% [REDACTED]	13950/36052	
[00:08<00:11, 1988.22it/s]		
40% [REDACTED]	14164/36052	
[00:09<00:12, 1702.70it/s]		
40% [REDACTED]	14368/36052	
[00:09<00:13, 1639.38it/s]		
41% [REDACTED]	14549/36052	
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41% [REDACTED]	14721/36052	
[00:09<00:13, 1568.20it/s]		
42% [REDACTED]	14891/36052	
[00:09<00:13, 1540.53it/s]		
42% [REDACTED]	15053/36052	[00:09:00:13, 1534.20it/s]
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43% [REDACTED]	15410/36052	[00:09:00:11, 1739.49it/s]
43% [REDACTED]	15624/36052	[00:09:00:11, 1739.49it/s]

44% [00:11, 1694.77it/s]	15803/36052 [00:10]
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45% [00:12, 1577.50it/s]	16171/36052 [00:10]
45% [00:12, 1619.02it/s]	16347/36052 [00:10]
46% [00:12, 1612.22it/s]	16513/36052 [00:10]
46% [00:12, 1601.68it/s]	16677/36052 [00:10]
47% [00:12, 1588.34it/s]	16839/36052 [00:10]
47% [00:10, 1755.62it/s]	17081/36052 [00:10]
48% [00:10, 1829.65it/s]	17292/36052 [00:10]
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49% [00:11, 1645.06it/s]	17668/36052 [00:11]
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51% [00:10, 1755.51it/s]	18241/36052 [00:11]
51% [00:10, 1724.94it/s]	18420/36052 [00:11]
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52% [00:09, 1840.30it/s]	18893/36052 [00:11]
53% [00:09, 1777.69it/s]	19084/36052 [00:11]
53% [00:10, 1571.48it/s]	19267/36052 [00:12]
54% [00:10, 1611.18it/s]	19445/36052 [00:12]
54% [00:09, 1684.61it/s]	19641/36052 [00:12]
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56% [00:09, 1575.86it/s]	20336/36052 [00:12]
57% [00:09, 1621.97it/s]	20517/36052 [00:12]

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60% [00:13<00:09, 1576.59it/s]	21528/36052
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66% [00:14<00:07, 1664.96it/s]	23846/36052
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67% [00:15<00:07, 1675.45it/s]	24229/36052
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68% [00:15<00:06, 1642.07it/s]	24610/36052
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70% [00:15<00:06, 1717.99it/s]	25185/36052

[00:15<00:06, 1779.75it/s]	25100/36052
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71% <00:06, 1708.44it/s]	25727/36052 [00:15
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73% <00:06, 1569.81it/s]	26443/36052 [00:16
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75% <00:05, 1665.36it/s]	27162/36052 [00:16
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79% <00:04, 1749.32it/s]	28559/36052 [00:17
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80% <00:04, 1719.52it/s]	28921/36052 [00:17
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81% <00:03, 1805.29it/s]	29320/36052 [00:17
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84% [<00:03, 1827.48it/s]	30355/36052 [00:18
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86% [<00:02, 1697.22it/s]	31138/36052 [00:18
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88% [<00:02, 1780.37it/s]	31723/36052 [00:19
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92% [00:20<00:01, 1631.62it/s]	33254/36052
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95% [00:20<00:00, 1825.40it/s]	34403/36052
96% [00:20<00:00, 1840.11it/s]	34598/36052
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97% [00:21<00:00, 1789.32it/s]	34979/36052

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[00:21<00:00, 1709.52it/s]
98%|███████████| 35185/36052 [00:2
1<00:00, 1842.88it/s]

98%|███████████| 35392/36052 [00:2
1<00:00, 1884.96it/s]

99%|███████████| 35582/36052 [00:2
1<00:00, 1803.51it/s]

99%|███████████| 35793/36052
[00:21<00:00, 1865.82it/s]

100%|███████████| 36011/36052
[00:21<00:00, 1929.56it/s]

100%|███████████| 36052/36052
[00:21<00:00, 1662.93it/s]
```

36052

300

In [94]:

```
average_w2v_on_essay_Xtest = np.vstack(avg_w2v_vectors_essays_Xtest)
print(average_w2v_on_essay_Xtest.shape)
```

(36052, 300)

8.3.3. Average Word2Vec encoding of preprocessed_essays on CV Data

In [95]:

```
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_essays_Xcv = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_cv['preprocessed_essays'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_essays_Xcv.append(vector)

print(len(avg_w2v_vectors_essays_Xcv))
print(len(avg_w2v_vectors_essays_Xcv[2]))
```

```
0%| 0/24
[00:00<?, ?it/s]

1%| :15, 1558.07it/s| 162/24155 [00:00

1%| :14, 1639.15it/s| 356/24155 [00:00

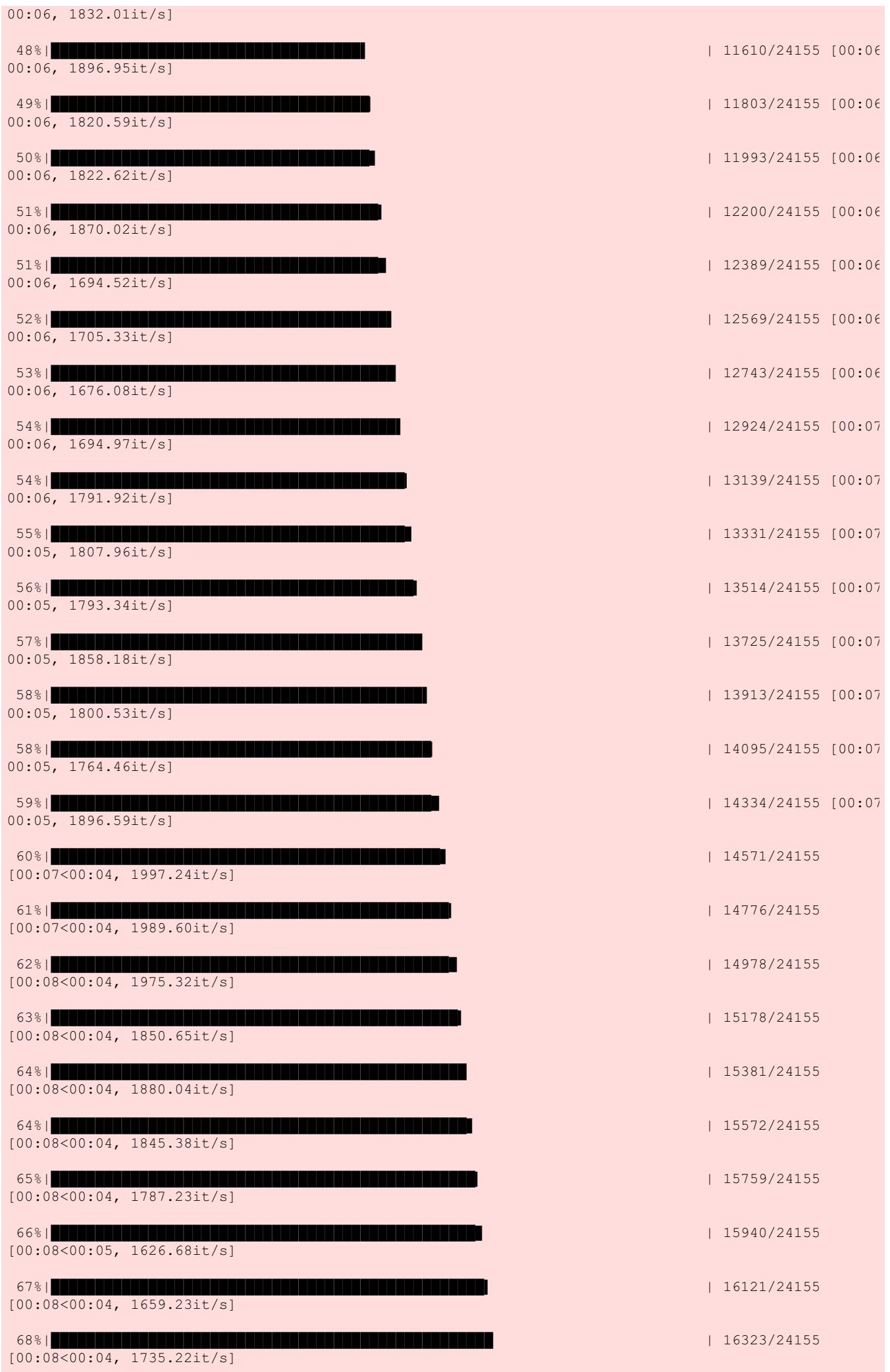
2%| :15, 1567.92it/s| 504/24155 [00:00

3%| :13, 1719.76it/s| 735/24155 [00:00

4%| :12, 1828.45it/s| 958/24155 [00:00
```

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5% [■■■■■]	1307/24155
[00:00<00:13, 1665.14it/s]	
6% [■■■■■■■■■]	1484/24155
[00:00<00:13, 1676.12it/s]	
7% [■■■■■■■■■■]	1698/24155
[00:00<00:12, 1757.45it/s]	
8% [■■■■■■■■■■■]	1872/24155
[00:01<00:13, 1711.00it/s]	
9% [■■■■■■■■■■■■]	2080/24155
[00:01<00:12, 1788.67it/s]	
9% [■■■■■■■■■■■■]	2259/24155
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10% [■■■■■■■■■■■■■]	2443/24155
[00:01<00:12, 1753.95it/s]	
11% [■■■■■■■■■■■■■■]	2638/24155
[00:01<00:12, 1788.61it/s]	
12% [■■■■■■■■■■■■■■■]	2836/24155
[00:01<00:11, 1821.93it/s]	
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[00:01<00:12, 1702.33it/s]	
13% [■■■■■■■■■■■■■■■■■]	3205/24155
[00:01<00:12, 1727.40it/s]	
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[00:01<00:11, 1866.58it/s]	
15% [■■■■■■■■■■■■■■■■■■■]	3647/24155
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16% [■■■■■■■■■■■■■■■■■■■■]	3859/24155
[00:02<00:10, 1933.46it/s]	
17% [■■■■■■■■■■■■■■■■■■■■■]	4055/24155
[00:02<00:10, 1853.41it/s]	
18% [■■■■■■■■■■■■■■■■■■■■■■]	4272/24155 [00:02
0:10, 1917.82it/s]	
19% [■■■■■■■■■■■■■■■■■■■■■■■]	4482/24155 [00:02
0:10, 1947.28it/s]	
19% [■■■■■■■■■■■■■■■■■■■■■■■■]	4706/24155 [00:02
0:09, 2005.11it/s]	
20% [■■■■■■■■■■■■■■■■■■■■■■■■■]	4909/24155 [00:02
0:09, 1965.89it/s]	
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0:09, 1987.26it/s]	
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0:09, 1944.65it/s]	
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0:09, 1995.71it/s]	
24% [■■■■■■■■■■■■■■■■■■■■■■■■■■■■■]	5743/24155 [00:03
0:09, 1887.45it/s]	
25% [■■■■■■■■■■■■■■■■■■■■■■■■■■■■■■]	5934/24155 [00:03
0:09, 1850.29it/s]	
25% [■■■■■■■■■■■■■■■■■■■■■■■■■■■■■■■]	6121/24155 [00:03
0:09, 1834.43it/s]	

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28% [0:09, 1865.25it/s]	6763/24155 [00:03
29% [0:08, 1948.16it/s]	6989/24155 [00:03
30% [0:08, 1979.94it/s]	7203/24155 [00:03
31% [0:08, 1898.84it/s]	7404/24155 [00:03
31% [0:08, 1864.01it/s]	7597/24155 [00:04
32% [00:04<00:08, 1849.89it/s]	7786/24155
33% [00:04<00:09, 1751.72it/s]	7973/24155
34% [00:04<00:09, 1680.39it/s]	8151/24155
34% [00:04<00:09, 1650.28it/s]	8322/24155
35% [00:04<00:08, 1770.84it/s]	8544/24155
36% [00:04<00:08, 1741.47it/s]	8725/24155
37% [00:04<00:08, 1729.47it/s]	8902/24155
38% [00:04<00:08, 1798.13it/s]	9108/24155
39% [00:05<00:07, 1915.61it/s]	9343/24155
40% [00:05<00:07, 1937.56it/s]	9550/24155
40% [00:05<00:07, 1924.43it/s]	9747/24155
41% [00:05<00:07, 1921.33it/s]	9946/24155
42% [00:05<00:08, 1685.98it/s]	10140/24155
43% [00:07, 1771.67it/s]	10349/24155 [00:05
44% [00:07, 1908.41it/s]	10591/24155 [00:05
45% [00:07, 1822.77it/s]	10789/24155 [00:05
45% [00:07, 1827.07it/s]	10980/24155 [00:05
46% [00:06, 1905.19it/s]	11200/24155 [00:06
47% [00:06, 1905.19it/s]	11395/24155 [00:06



68% [00:08<00:04, 1704.34it/s]	16500/24155
69% [00:09<00:04, 1836.15it/s]	16733/24155
70% [00:09<00:03, 1880.10it/s]	16940/24155
71% [<00:03, 1766.65it/s]	17132/24155 [00:09
72% [<00:03, 1803.18it/s]	17329/24155 [00:09
73% [<00:03, 1848.07it/s]	17533/24155 [00:09
73% [<00:03, 1793.97it/s]	17721/24155 [00:09
74% [<00:03, 1760.02it/s]	17903/24155 [00:09
75% [<00:03, 1745.33it/s]	18081/24155 [00:09
76% [<00:03, 1760.88it/s]	18268/24155 [00:09
77% [<00:03, 1873.56it/s]	18497/24155 [00:10
78% [<00:02, 1992.09it/s]	18740/24155 [00:10
78% [<00:02, 1994.64it/s]	18948/24155 [00:10
79% [<00:02, 1936.52it/s]	19151/24155 [00:10
80% [<00:02, 1834.43it/s]	19347/24155 [00:10
81% [<00:02, 1823.56it/s]	19534/24155 [00:10
82% [<00:02, 1880.73it/s]	19745/24155 [00:10
83% [<00:02, 1901.69it/s]	19948/24155 [00:10
83% [<00:02, 1862.94it/s]	20140/24155 [00:10
84% [<00:02, 1896.98it/s]	20346/24155 [00:11
85% [<00:01, 1890.47it/s]	20541/24155 [00:11
86% [<00:01, 1849.15it/s]	20731/24155 [00:11
87% [<00:01, 1833.64it/s]	20918/24155 [00:11
87% [<00:01, 1822.97it/s]	21105/24155 [00:11
88% [00:11<00:01, 1762.07it/s]	21288/24155
89% [00:11<00:01, 1825.00it/s]	21495/24155

90% | [00:11<00:01, 1831.44it/s] | 21687/24155

91% | [00:11<00:01, 1730.65it/s] | 21871/24155

91% | [00:12<00:01, 1676.66it/s] | 22046/24155

92% | [00:12<00:01, 1695.41it/s] | 22227/24155

93% | [00:12<00:01, 1660.28it/s] | 22398/24155

94% | [00:12<00:00, 1729.03it/s] | 22597/24155

94% | [00:12<00:00, 1732.58it/s] | 22778/24155

95% | [00:12<00:00, 1740.73it/s] | 22961/24155

96% | [00:12<00:00, 1702.90it/s] | 23136/24155

97% | [00:12<00:00, 1748.71it/s] | 23330/24155

97% | [00:12<00:00, 1757.80it/s] | 23515/24155

98% | [00:12<00:00, 1720.43it/s] | 23692/24155 [00:12<00:00, 1720.43it/s]

99% | [00:13<00:00, 1732.09it/s] | 23875/24155

100% | [00:13<00:00, 1764.91it/s] | 24067/24155

100% | [00:13<00:00, 1827.18it/s] | 24155/24155

24155
300

In [96]:

```
average_w2v_on_essay_Xcv = np.vstack(avg_w2v_vectors_essays_Xcv)
print(average_w2v_on_essay_Xcv.shape)
```

(24155, 300)

8.4.1. Average Word2Vec encoding of preprocessed_titles on Train Data

In [97]:

```
#t-title
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_titles_Xtrain = [] # the avg-w2v for each sentence/review is stored in this list
for t in tqdm(X_train['preprocessed_titles'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in t.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
```

```

        cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_Xtrain.append(vector)

print(len(avg_w2v_vectors_titles_Xtrain))
print(len(avg_w2v_vectors_titles_Xtrain[0]))
```

0%| [00:00<?, ?it/s] | 0/49

7%| [██████] | 3336/49041 [00:00<00:01, 32079.32it/s]

16%| [███] | 7661/49041 [00:00<00:01, 34444.76it/s]

26%| [██████] | 12754/49041 [00:00<00, 37810.21it/s]

37%| [██████] | 17911/49041 [00:00<00:00, 40713.35it/s]

44%| [██████] | 21370/49041 [00:00<0:00, 38154.45it/s]

51%| [██████] | 24789/49041 [00:00<0:00, 36401.21it/s]

61%| [██████] | 29791/49041 [00:00<00:00, 39270.08it/s]

70%| [██████] | 34372/49041 [00:00<00:00, 40593.75it/s]

78%| [██████] | 38361/49041 [00:00<00:00, 37218.83it/s]

86%| [██████] | 42143/49041 [00:01<00:00, 36961.50it/s]

93%| [██████] | 45844/49041 [00:01<00:00, 35695.22it/s]

100%| [██████] | 49041/49041 [00:01<00:00, 39055.33it/s]

49041
300

In [98]:

```
average_w2v_on_titles_Xtrain = np.vstack(avg_w2v_vectors_titles_Xtrain)
print(average_w2v_on_titles_Xtrain.shape)
```

(49041, 300)

8.4.2. Average Word2Vec encoding of preprocessed_titles on Test Data

In [99]:

```
#t-title
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_titles_Xtest = [] # the avg-w2v for each sentence/review is stored in this list
for t in tqdm(X_test['preprocessed_titles'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in t.split(): # for each word in a review/sentence
```

```

    if word in glove_words:
        vector += model[word]
        cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_Xtest.append(vector)

print(len(avg_w2v_vectors_titles_Xtest))
print(len(avg_w2v_vectors_titles_Xtest[0]))

```

0% | 0/36052
[00:00<?, ?it/s]

13% | ██████████ | 4754/36052
[00:00<00:00, 45723.34it/s]

23% | ██████████ | 8351/36052 [00:00<
:00, 41699.57it/s]

29% | ██████████ | 10350/36052 [00:00<
:00, 27784.32it/s]

34% | ██████████ | 12218/36052
[00:00<00:01, 23160.72it/s]

39% | ██████████ | 14026/36052
[00:00<00:01, 20773.04it/s]

44% | ██████████ | 15794/36052 [00:00<
:01, 18976.16it/s]

49% | ██████████ | 17506/36052 [00:00<
:01, 18146.30it/s]

53% | ██████████ | 19198/36052 [00:00<
:00, 17540.67it/s]

59% | ██████████ | 21140/36052
[00:00<00:00, 17867.05it/s]

65% | ██████████ | 23471/36052
[00:01<00:00, 19025.99it/s]

72% | ██████████ | 25994/36052 [00:01<
00:00, 20343.85it/s]

78% | ██████████ | 28041/36052 [00:01<
00:00, 19907.14it/s]

83% | ██████████ | 30043/36052 [00:01<
00:00, 15497.60it/s]

88% | ██████████ | 31751/36052
[00:01<00:00, 8615.10it/s]

92% | ██████████ | 33338/36052
[00:01<00:00, 9909.87it/s]

100% | ██████████ | 36052/36052
[00:02<00:00, 17206.28it/s]

36052
300

In [100]:

```

average_w2v_on_titles_Xtest = np.vstack(avg_w2v_vectors_titles_Xtest)
print(average_w2v_on_titles_Xtest.shape)

```

(36052, 300)

8.4.3. Average Word2Vec encoding of preprocessed_titles on CV Data

In [101]:

```
#t-title
# average Word2Vec
# compute average word2vec for each review.
avg_w2v_vectors_titles_Xcv = [] # the avg-w2v for each sentence/review is stored in this list
for t in tqdm(X_cv['preprocessed_titles'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    cnt_words = 0; # num of words with a valid vector in the sentence/review
    for word in t.split(): # for each word in a review/sentence
        if word in glove_words:
            vector += model[word]
            cnt_words += 1
    if cnt_words != 0:
        vector /= cnt_words
    avg_w2v_vectors_titles_Xcv.append(vector)

print(len(avg_w2v_vectors_titles_Xcv))
print(len(avg_w2v_vectors_titles_Xcv[0]))
```

0% | 00:00<?, ?it/s] | 0/24
20% | :00, 47397.72it/s] | 4928/24155 [00:00<
38% | [00:00<00:00, 45043.78it/s] | 9125/24155
54% | 0:00, 42438.72it/s] | 13013/24155 [00:00<
70% | [00:00<00:00, 40702.89it/s] | 16877/24155
87% | 00:00, 40285.76it/s] | 20967/24155 [00:00<
100% | [00:00<00:00, 35952.99it/s] | 24155/24155
24155
300

In [102]:

```
average_w2v_on_titles_Xcv = np.vstack(avg_w2v_vectors_titles_Xcv)
print(average_w2v_on_titles_Xcv.shape)
```

(24155, 300)

8.5.1. TFIDF weighted Word2Vec encoding of preprocessed_essays on Train Data

In [103]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]
tfidf_model = TfidfVectorizer()
tfidf_model.fit(X_train['preprocessed_essays'].values)
# we are converting a dictionary with word as a key, and the idf as a value
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
tfidf_words = set(tfidf_model.get_feature_names())
```

In [104]:

```

# average Word2Vec
# compute average word2vec for each review.
tfidf_weighted_w2v_vectors_eassays_Xtrain = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_train['preprocessed_essays'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
            value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
            idf value for each word
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_weighted_w2v_vectors_eassays_Xtrain.append(vector)

print(len(tfidf_weighted_w2v_vectors_eassays_Xtrain))
print(len(tfidf_weighted_w2v_vectors_eassays_Xtrain[0]))

```

0% | 0/49
[00:00<?, ?it/s]

0% | 24/49041 [00:
12:13, 66.78it/s]

0% | 57/49041 [00:
09:19, 87.55it/s]

0% | 89/49041 [00:
7:32, 108.24it/s]

0% | 104/49041 [00:
7:49, 104.31it/s]

0% | 141/49041 [00:
6:09, 132.39it/s]

0% | 167/49041 [00:
5:17, 154.15it/s]

0% | 199/49041 [00:
4:29, 181.30it/s]

0% | 228/49041 [00:
4:00, 202.57it/s]

1% | 254/49041 [00:
3:56, 206.62it/s]

1% | 279/49041 [00:
4:14, 191.89it/s]

1% | 301/49041 [00:
4:09, 195.30it/s]

1% | 327/49041 [00:
3:53, 209.03it/s]

1% | 350/49041 [00:
3:49, 212.54it/s]

1% | 375/49041 [00:
3:43, 217.91it/s]

1% | 398/49041 [00:
3:44, 216.42it/s]

1% | 429/49041 [00:
3:26, 235.81it/s]

1% | 454/49041 [00:
1:21, 240.11it/s]

1%|██████████
3:24, 237.17it/s] | 454/49041 [00:00

1%|██████████
3:10, 254.71it/s] | 486/49041 [00:00

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3:07, 258.90it/s] | 514/49041 [00:00

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2:48, 287.07it/s] | 554/49041 [00:00

1%|██████████
2:42, 298.00it/s] | 588/49041 [00:00

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2:40, 300.87it/s] | 620/49041 [00:00

1%|██████████
2:50, 283.58it/s] | 651/49041 [00:00

1%|██████████
3:01, 266.80it/s] | 681/49041 [00:00

1%|██████████
2:52, 280.20it/s] | 714/49041 [00:00

2%|██████████
2:51, 282.24it/s] | 745/49041 [00:00

2%|██████████
2:43, 294.33it/s] | 779/49041 [00:00

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2:48, 285.87it/s] | 809/49041 [00:00

2%|██████████
2:49, 283.75it/s] | 838/49041 [00:00

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2:47, 287.43it/s] | 870/49041 [00:00

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2:37, 305.02it/s] | 907/49041 [00:00

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2:23, 335.46it/s] | 984/49041 [00:00

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2:29, 321.05it/s] | 1019/49041 [00:00

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2:33, 311.58it/s] | 1085/49041 [00:00

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2:34, 310.43it/s] | 1117/49041 [00:00

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2:43, 292.64it/s] | 1149/49041 [00:00

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2:53, 275.35it/s] | 1179/49041 [00:00

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2:33, 308.69it/s] | 1532/49041 [00:C

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3%|██████
2:39, 298.07it/s] | 1607/49041 [00:C

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2:40, 294.70it/s] | 1638/49041 [00:C

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3:04, 257.42it/s] | 1669/49041 [00:C

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2:54, 270.70it/s] | 1701/49041 [00:C

4%|██████
3:12, 245.40it/s] | 1730/49041 [00:C

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3:22, 233.49it/s] | 1756/49041 [00:C

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3:12, 245.47it/s] | 1785/49041 [00:C

4%|██████
3:02, 258.80it/s] | 1817/49041 [00:C

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3:06, 253.25it/s] | 1844/49041 [00:C

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2:54, 269.58it/s] | 1877/49041 [00:C

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2:46, 282.34it/s] | 1910/49041 [00:C

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2:45, 284.19it/s] | 1940/49041 [00:C

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3:13, 242.84it/s] | 1969/49041 [00:C

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3:09, 247.66it/s] | 1996/49041 [00:C

4%|██████
[00:07<03:26, 227.52it/s] | 2022/49041

4%|██████
[00:07<03:23, 231.33it/s] | 2048/49041

4%|██████
[00:07<03:21, 231.33it/s] | 2073/49041

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4%|██████ [00:08<03:02, 257.56it/s] | 2135/49041
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4%|██████ [00:08<03:04, 253.97it/s] | 2192/49041
5%|█████ [00:08<03:10, 246.33it/s] | 2218/49041
5%|█████ [00:08<03:24, 229.13it/s] | 2247/49041
5%|█████ [00:08<03:13, 242.08it/s] | 2275/49041
5%|█████ [00:08<03:07, 249.67it/s] | 2305/49041
5%|█████ [00:08<02:59, 260.17it/s] | 2339/49041
5%|█████ [00:08<02:48, 277.16it/s] | 2380/49041
5%|█████ [00:09<02:33, 304.29it/s] | 2419/49041
5%|█████ [00:09<02:25, 319.40it/s] | 2458/49041
5%|█████ [00:09<02:19, 334.27it/s] | 2496/49041
5%|█████ [00:09<02:15, 343.06it/s] | 2532/49041
5%|█████ [00:09<02:22, 325.36it/s] | 2566/49041
5%|█████ [00:09<02:32, 304.84it/s] | 2598/49041
5%|█████ [00:09<02:45, 280.04it/s] | 2627/49041
5%|█████ [00:09<02:51, 270.31it/s] | 2655/49041
5%|█████ [00:09<03:05, 249.77it/s] | 2684/49041
6%|█████ [00:10<02:59, 257.87it/s] | 2712/49041
6%|█████ [00:10<02:57, 261.19it/s] | 2744/49041
6%|█████ [00:10<02:49, 273.62it/s] | 2772/49041
6%|█████ [00:10<02:53, 266.08it/s] | 2799/49041
6%|█████ [00:10<03:05, 249.50it/s] | 2829/49041
6%|█████ [00:10<02:57, 260.04it/s] | 2856/49041
6%|█████ [00:10<03:03, 251.23it/s] | 2883/49041

6% [██████]	2885/49041
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6% [██████]	2943/49041
[00:11<02:56, 261.80it/s]	
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[00:11<02:55, 261.87it/s]	
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[00:11<03:04, 249.63it/s]	
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7% [█████]	3392/49041
[00:12<02:16, 334.55it/s]	
7% [█████]	3436/49041
[00:12<02:07, 356.98it/s]	
7% [█████]	3483/49041
[00:12<01:59, 381.02it/s]	
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[00:13<01:50, 410.60it/s]	
8% [█████]	3785/49041
[00:13<01:52, 403.87it/s]	

8% [██████]	3829/49041
[00:13<01:50, 409.46it/s]	
8% [██████]	3875/49041
[00:13<01:47, 418.82it/s]	
8% [██████]	3919/49041
[00:13<01:47, 420.14it/s]	
8% [██████]	3962/49041
[00:14<01:49, 413.33it/s]	
8% [██████]	4004/49041
[00:14<01:49, 410.44it/s]	
8% [██████]	4048/49041
[00:14<01:48, 414.18it/s]	
8% [██████]	4091/49041
[00:14<01:48, 414.02it/s]	
8% [██████]	4133/49041
[00:14<02:03, 363.93it/s]	
9% [██████]	4171/49041
[00:14<02:11, 340.87it/s]	
9% [██████]	4210/49041
[00:14<02:07, 350.48it/s]	
9% [██████]	4250/49041
[00:14<02:04, 360.07it/s]	
9% [██████]	4288/49041
[00:14<02:03, 361.68it/s]	
9% [██████]	4328/49041
[00:14<02:01, 368.31it/s]	
9% [██████]	4369/49041
[00:15<01:58, 375.75it/s]	
9% [██████]	4410/49041
[00:15<01:57, 381.14it/s]	
9% [██████]	4449/49041
[00:15<01:57, 379.29it/s]	
9% [██████]	4493/49041
[00:15<01:53, 391.49it/s]	
9% [██████]	4533/49041
[00:15<01:55, 384.91it/s]	
9% [██████]	4575/49041
[00:15<01:53, 390.46it/s]	
9% [██████]	4615/49041
[00:15<01:59, 371.39it/s]	
9% [██████]	4655/49041
[00:15<01:58, 375.27it/s]	
10% [██████]	4702/49041
[00:15<01:52, 395.44it/s]	
10% [██████]	4748/49041
[00:15<01:49, 404.14it/s]	
10% [██████]	4791/49041
[00:16<01:48, 406.90it/s]	
10% [██████]	4832/49041
[00:16<01:53, 389.28it/s]	
10% [██████]	4872/49041

[00:16<01:56, 379.09it/s] | 4911/49041
10% [] | 4952/49041
[00:16<01:59, 369.28it/s] | 4994/49041
10% [] | 5039/49041
[00:16<01:50, 397.69it/s] | 5083/49041
10% [] | 5124/49041
[00:16<01:54, 383.68it/s] | 5163/49041
11% [] | 5204/49041
[00:17<01:55, 381.07it/s] | 5249/49041
11% [] | 5292/49041
[00:17<01:48, 402.67it/s] | 5333/49041
11% [] | 5373/49041
[00:17<01:54, 380.21it/s] | 5413/49041
11% [] | 5456/49041
[00:17<01:51, 390.60it/s] | 5496/49041
11% [] | 5541/49041
[00:18<01:53, 384.86it/s] | 5580/49041
11% [] | 5621/49041
[00:18<01:53, 382.35it/s] | 5660/49041
12% [] | 5699/49041
[00:18<01:54, 380.13it/s] | 5742/49041
12% [] | 5781/49041
[00:18<01:55, 375.06it/s] | 5822/49041
12% [] | 5865/49041
[00:18<01:50, 389.94it/s] | 5905/49041
12% [] | 5905/49041
[00:19<01:56, 371.07it/s]

12% [███████]	5947/49041
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[00:19<01:54, 374.41it/s]	
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[00:19<02:02, 351.09it/s]	
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[00:19<01:59, 360.54it/s]	
13% [███████]	6147/49041
[00:19<01:55, 371.08it/s]	
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[00:19<01:53, 377.76it/s]	
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[00:19<01:51, 382.59it/s]	
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[00:19<01:47, 396.38it/s]	
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[00:20<01:51, 383.77it/s]	
13% [███████]	6355/49041
[00:20<01:50, 386.87it/s]	
13% [███████]	6395/49041
[00:20<01:50, 386.22it/s]	
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[00:20<01:46, 399.11it/s]	
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[00:20<01:48, 393.09it/s]	
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[00:20<01:56, 364.89it/s]	
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[00:20<01:59, 356.84it/s]	
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[00:20<02:01, 348.47it/s]	
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[00:20<01:59, 356.07it/s]	
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[00:21<01:52, 378.20it/s]	
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[00:21<01:47, 395.42it/s]	
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[00:21<01:48, 390.60it/s]	
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[00:21<01:46, 397.22it/s]	
14% [███████]	6857/49041
[00:21<01:42, 409.76it/s]	
14% [███████]	6899/49041
[00:21<01:44, 403.31it/s]	
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[00:21<01:54, 366.24it/s]	
14% [███████]	6984/49041
[00:21<01:50, 381.65it/s]	

14% [██████████]	7028/49041
[00:21<01:46, 393.20it/s]	
14% [██████████]	7072/49041
[00:22<01:44, 401.76it/s]	
15% [██████████]	7113/49041
[00:22<01:44, 399.48it/s]	
15% [██████████]	7154/49041
[00:22<01:48, 384.51it/s]	
15% [██████████]	7193/49041
[00:22<01:53, 368.65it/s]	
15% [██████████]	7234/49041
[00:22<01:52, 371.89it/s]	
15% [██████████]	7272/49041
[00:22<01:54, 365.69it/s]	
15% [██████████]	7315/49041
[00:22<01:50, 378.83it/s]	
15% [██████████]	7359/49041
[00:22<01:46, 391.13it/s]	
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[00:22<01:48, 384.71it/s]	
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[00:23<01:51, 371.20it/s]	
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[00:23<01:52, 369.44it/s]	
15% [██████████]	7594/49041
[00:23<01:49, 379.16it/s]	
16% [██████████]	7639/49041
[00:23<01:45, 393.82it/s]	
16% [██████████]	7682/49041
[00:23<01:43, 399.54it/s]	
16% [██████████]	7723/49041
[00:23<01:49, 376.07it/s]	
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[00:23<01:53, 363.16it/s]	
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[00:23<01:54, 360.96it/s]	
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16% [██████████]	7920/49041
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16% [██████████]	7961/49041
[00:24<01:48, 379.01it/s]	
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16% [██████████]	8040/49041

[00:24<01:51, 366.21it/s]	
16% [██████████]	8079/49041
[00:24<01:51, 368.81it/s]	
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[00:25<01:53, 358.54it/s]	
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[00:25<01:48, 375.55it/s]	
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[00:25<01:47, 378.25it/s]	
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01:46, 381.53it/s]	
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18% [██████████]	8596/49041 [00:2
01:45, 385.10it/s]	
18% [██████████]	8640/49041 [00:2
01:42, 395.78it/s]	
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[00:26<01:44, 383.80it/s]	
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01:42, 391.71it/s]	
18% [██████████]	8885/49041 [00:2
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18% [██████████]	8927/49041 [00:2
01:41, 395.88it/s]	
18% [██████████]	8967/49041 [00:2
01:44, 383.41it/s]	
18% [██████████]	9006/49041 [00:2
01:45, 380.88it/s]	
18% [██████████]	9045/49041
[00:27<01:57, 339.54it/s]	

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19% [00:27<02:18, 289.13it/s]	9113/49041
19% [02:26, 273.17it/s]	9143/49041 [00:2
19% [02:25, 274.87it/s]	9172/49041 [00:2
19% [02:30, 264.01it/s]	9201/49041 [00:2
19% [02:33, 259.67it/s]	9228/49041 [00:2
19% [02:36, 253.82it/s]	9255/49041 [00:2
19% [02:37, 252.68it/s]	9282/49041 [00:2
19% [02:43, 243.40it/s]	9308/49041 [00:2
19% [02:43, 242.51it/s]	9333/49041 [00:2
19% [02:51, 231.15it/s]	9358/49041 [00:2
19% [03:01, 218.44it/s]	9382/49041 [00:2
19% [03:06, 211.99it/s]	9405/49041 [00:2
19% [03:17, 200.30it/s]	9427/49041 [00:2
19% [03:23, 194.12it/s]	9448/49041 [00:2
19% [03:10, 208.08it/s]	9474/49041 [00:2
19% [02:52, 228.82it/s]	9505/49041 [00:2
19% [05:12, 126.46it/s]	9529/49041 [00:2
19% [04:24, 149.39it/s]	9558/49041 [00:2
20% [03:39, 179.31it/s]	9593/49041 [00:2
20% [03:16, 200.83it/s]	9622/49041 [00:3
20% [03:10, 207.31it/s]	9648/49041 [00:3
20% [03:03, 214.03it/s]	9673/49041 [00:3
20% [02:54, 225.95it/s]	9700/49041 [00:3
20% [02:42, 241.67it/s]	9730/49041 [00:3
20% [02:49, 231.11it/s]	9756/49041 [00:3

20% ███████████	9783/49041 [00:3
02:44, 239.00it/s]	
20% ███████████	9808/49041 [00:3
02:43, 239.43it/s]	
20% ███████████	9839/49041 [00:3
02:34, 254.47it/s]	
20% ███████████	9871/49041 [00:3
02:25, 268.41it/s]	
20% ███████████	9899/49041 [00:3
02:30, 259.71it/s]	
20% ███████████	9926/49041 [00:3
02:30, 259.70it/s]	
20% ███████████	9953/49041 [00:3
02:33, 253.84it/s]	
20% ███████████	9994/49041 [00:3
02:19, 280.61it/s]	
20% ███████████	10029/49041 [00:3
02:13, 291.38it/s]	
21% ███████████	10061/49041 [00:3
02:10, 297.81it/s]	
21% ███████████	10101/49041 [00:3
02:01, 319.46it/s]	
21% ███████████	10138/49041 [00:3
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01:54, 339.58it/s]	
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01:53, 341.55it/s]	
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21% ███████████	10279/49041 [00:3
02:07, 303.02it/s]	
21% ███████████	10311/49041 [00:3
02:07, 304.43it/s]	
21% ███████████	10342/49041 [00:3
02:09, 299.01it/s]	
21% ███████████	10373/49041 [00:3
02:21, 273.46it/s]	
21% ███████████	10402/49041 [00:3
02:20, 275.08it/s]	
21% ███████████	10430/49041 [00:3
02:34, 249.91it/s]	
21% ███████████	10456/49041 [00:3
02:41, 238.92it/s]	
21% ███████████	10481/49041 [00:3
02:42, 236.66it/s]	
21% ███████████	10506/49041 [00:3
02:43, 235.11it/s]	
21% ███████████	10541/49041 [00:3
02:28, 258.50it/s]	
22% ███████████	10568/49041 [00:3

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02:16, 281.31it/s]	22% [██████████]	10691/49041 [00:3
02:24, 265.00it/s]	22% [██████████]	10724/49041 [00:3
02:17, 278.81it/s]	22% [██████████]	10758/49041 [00:3
02:11, 291.70it/s]	22% [██████████]	10788/49041 [00:3
02:11, 290.74it/s]	22% [██████████]	10818/49041 [00:3
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02:25, 261.54it/s]	22% [██████████]	11050/49041 [00:3
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02:37, 240.33it/s]	23% [██████████]	11107/49041 [00:3
02:30, 252.67it/s]	23% [██████████]	11133/49041 [00:3
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02:28, 254.28it/s]	23% [██████████]	11251/49041 [00:3
02:18, 272.47it/s]	23% [██████████]	11295/49041 [00:3
02:03, 305.06it/s]	23% [██████████]	11328/49041 [00:3
02:03, 305.22it/s]	23% [██████████]	

23% [██████████] | 11368/49041 [00:3
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23% [██████████] | 11408/49041 [00:3
01:50, 341.17it/s]

23% [██████████] | 11451/49041 [00:3
01:44, 360.08it/s]

23% [██████████] | 11489/49041 [00:3
01:43, 361.70it/s]

24% [██████████] | 11528/49041 [00:3
01:42, 365.62it/s]

24% [██████████] | 11567/49041 [00:3
01:41, 368.41it/s]

24% [██████████] | 11605/49041 [00:3
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24% [██████████] | 11643/49041 [00:3
01:53, 328.81it/s]

24% [██████████] | 11677/49041 [00:3
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24% [██████████] | 11796/49041 [00:3
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24% [██████████] | 11834/49041 [00:3
01:42, 361.95it/s]

24% [██████████] | 11873/49041 [00:3
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24% [██████████] | 11918/49041 [00:3
01:36, 383.61it/s]

24% [██████████] | 11957/49041 [00:3
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24% [██████████] | 11994/49041 [00:3
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01:45, 352.24it/s]

25% [██████████] | 12070/49041 [00:3
02:36, 236.95it/s]

25% [██████████] | 12103/49041 [00:3
02:24, 256.46it/s]

25% [██████████] | 12140/49041 [00:3
02:11, 279.92it/s]

25% [██████████] | 12172/49041 [00:3
02:45, 222.28it/s]

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03:17, 186.20it/s]

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03:02, 201.65it/s]

25% [██████████] | 12262/49041 [00:3
02:38, 231.78it/s]

25% [██████████] | 12294/49041 [00:3
02:26, 250.31it/s]

22:20, 200.01it/s]		
25% [██████████]	12333/49041 [00:3	
02:13, 275.72it/s]		
25% [██████████]	12364/49041 [00:3	
02:17, 267.47it/s]		
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02:19, 262.00it/s]		
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02:18, 264.15it/s]		
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02:25, 251.38it/s]		
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02:25, 250.98it/s]		
25% [██████████]	12502/49041 [00:4	
02:25, 250.69it/s]		
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02:18, 263.27it/s]		
26% [██████████]	12574/49041 [00:4	
02:05, 289.95it/s]		
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02:11, 276.67it/s]		
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02:10, 279.43it/s]		
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02:05, 289.83it/s]		
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02:15, 267.74it/s]		
26% [██████████]	12725/49041 [00:4	
02:21, 256.41it/s]		
26% [██████████]	12756/49041 [00:4	
02:15, 267.67it/s]		
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02:18, 262.13it/s]		
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02:18, 261.39it/s]		
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02:18, 260.69it/s]		
26% [██████████]	12927/49041 [00:4	
02:17, 263.21it/s]		
26% [██████████]	12954/49041 [00:4	
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02:16, 265.09it/s]		
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02:11, 274.21it/s]		
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02:15, 266.48it/s]		
27% [██████████]	13076/49041 [00:4	

27% [REDACTED]	13073/49041 [00:4
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27% [REDACTED]	13105/49041 [00:4
02:10, 274.51it/s]	
27% [REDACTED]	13137/49041 [00:4
02:06, 283.71it/s]	
27% [REDACTED]	13166/49041 [00:4
02:08, 279.01it/s]	
27% [REDACTED]	13195/49041 [00:4
02:11, 272.69it/s]	
27% [REDACTED]	13223/49041 [00:4
02:14, 265.47it/s]	
27% [REDACTED]	13251/49041 [00:4
02:15, 263.60it/s]	
27% [REDACTED]	13280/49041 [00:4
02:13, 268.02it/s]	
27% [REDACTED]	13309/49041 [00:4
02:11, 271.20it/s]	
27% [REDACTED]	13337/49041 [00:4
02:22, 250.30it/s]	
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02:19, 255.72it/s]	
27% [REDACTED]	13392/49041 [00:4
02:18, 256.90it/s]	
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02:16, 260.50it/s]	
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02:12, 268.32it/s]	
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02:13, 265.64it/s]	
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02:21, 252.03it/s]	
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02:21, 251.44it/s]	
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02:26, 242.14it/s]	
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02:21, 249.69it/s]	
28% [REDACTED]	13668/49041 [00:4
02:15, 260.20it/s]	
28% [REDACTED]	13698/49041 [00:4
02:11, 268.10it/s]	
28% [REDACTED]	13726/49041 [00:4
02:11, 268.46it/s]	
28% [REDACTED]	13753/49041 [00:4
02:12, 265.76it/s]	
28% [REDACTED]	13782/49041 [00:4
02:10, 269.57it/s]	

28% ███████████	13811/49041 [00:4
02:09, 272.31it/s]	
28% ███████████	13839/49041 [00:4
02:11, 268.28it/s]	
28% ███████████	13866/49041 [00:4
02:13, 262.53it/s]	
28% ███████████	13893/49041 [00:4
03:05, 189.11it/s]	
28% ███████████	13925/49041 [00:4
02:43, 215.28it/s]	
28% ███████████	13951/49041 [00:4
02:36, 224.51it/s]	
29% ███████████	13979/49041 [00:4
02:28, 236.30it/s]	
29% ███████████	14009/49041 [00:4
02:20, 249.87it/s]	
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02:35, 224.95it/s]	
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02:21, 247.54it/s]	
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02:21, 247.13it/s]	
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02:20, 248.00it/s]	
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02:17, 253.84it/s]	
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02:24, 241.83it/s]	
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02:17, 252.23it/s]	
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02:16, 254.43it/s]	
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02:16, 254.19it/s]	
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02:16, 252.94it/s]	
29% ███████████	14428/49041 [00:4
02:13, 260.20it/s]	
29% ███████████	14461/49041 [00:4
02:05, 275.07it/s]	
30% ███████████	14498/49041 [00:4
01:57, 295.18it/s]	
30% ███████████	14529/49041 [00:4
02:25, 237.83it/s]	

02.20, 237.85it/s]

30% [REDACTED]	14556/49041 [00:4
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02:23, 239.31it/s]	
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02:10, 262.31it/s]	
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02:58, 189.59it/s]	
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02:48, 200.48it/s]	
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02:44, 204.86it/s]	
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02:57, 190.25it/s]	
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[00:53<06:53, 80.52it/s]	
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42% <01:57, 242.49it/s]	20669/49041 [01:1]
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42% <03:04, 153.84it/s]	20714/49041 [01:1]
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42% <02:14, 209.26it/s]	20812/49041 [01:1]
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43%	20870/49041 [01:1]

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44% ███████████	<02:13, 205.84it/s]	21636/49041 [01:1
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48% ███████████	<02:01, 209.56it/s]	23565/49041 [01:2
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<01:33, 256.79it/s]	
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53% ███████████	<01:41, 227.62it/s]	25963/49041 [01:4
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~01:33, 250.01it/s		
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54% ███████████	<01:30, 249.98it/s]	26427/49041 [01:4
54% ███████████	<01:29, 252.82it/s]	26454/49041 [01:4
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54% ███████████	<01:46, 210.33it/s]	26588/49041 [01:4
54% ███████████	<01:42, 218.54it/s]	26613/49041 [01:4
54% ███████████	<01:40, 222.08it/s]	26637/49041 [01:4
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55% ███████████		27065/49041 [01:4

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57% [REDACTED]	27718/49041 [01:4
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57% ███████████	<01:30, 235.27it/s]	27743/49041 [01:4
57% ███████████	<01:29, 236.81it/s]	27768/49041 [01:4
57% ███████████	<01:27, 243.25it/s]	27795/49041 [01:4
57% ███████████	<01:20, 263.50it/s]	27829/49041 [01:4
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71% [REDACTED]	34804/49041 [02:1
6<00:50, 279.30it/s]	
71% [REDACTED]	34834/49041 [02:1
6<00:50, 282.00it/s]	
71% [REDACTED]	34863/49041 [02:1
6<00:51, 277.84it/s]	
71% [REDACTED]	34891/49041 [02:1
6<00:51, 275.22it/s]	
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6<00:51, 276.00it/s]	
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6<00:51, 273.95it/s]	
71% [REDACTED]	34980/49041 [02:1
6<00:50, 280.78it/s]	
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6<00:49, 283.08it/s]	
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6<00:51, 272.29it/s]	
72% [REDACTED]	35067/49041 [02:1
7<00:53, 262.23it/s]	
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7<00:53, 258.47it/s]	
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7<00:52, 264.28it/s]	72% [REDACTED]	35151/49041 [02:1
7<00:52, 265.77it/s]	72% [REDACTED]	35182/49041 [02:1
7<00:50, 274.71it/s]	72% [REDACTED]	35215/49041 [02:1
7<00:48, 286.27it/s]	72% [REDACTED]	35244/49041 [02:1
7<00:49, 277.50it/s]	72% [REDACTED]	35274/49041 [02:1
7<00:49, 280.72it/s]	72% [REDACTED]	35303/49041 [02:1
7<00:50, 273.82it/s]	72% [REDACTED]	35332/49041 [02:1
7<00:49, 275.34it/s]	72% [REDACTED]	35362/49041 [02:1
8<00:48, 279.17it/s]	72% [REDACTED]	35391/49041 [02:1
8<00:48, 279.09it/s]	72% [REDACTED]	35419/49041 [02:1
8<00:52, 257.79it/s]	72% [REDACTED]	35450/49041 [02:1
8<00:50, 268.69it/s]	72% [REDACTED]	35483/49041 [02:1
8<00:48, 281.66it/s]	72% [REDACTED]	35513/49041 [02:1
8<00:47, 283.69it/s]	72% [REDACTED]	35542/49041 [02:1
8<00:47, 282.24it/s]	72% [REDACTED]	35571/49041 [02:1
8<00:48, 277.99it/s]	73% [REDACTED]	35600/49041 [02:1
8<00:48, 278.28it/s]	73% [REDACTED]	35636/49041 [02:1
9<00:45, 295.68it/s]	73% [REDACTED]	35666/49041 [02:1
9<00:48, 274.20it/s]	73% [REDACTED]	35696/49041 [02:1
9<00:47, 278.34it/s]	73% [REDACTED]	35734/49041 [02:1
9<00:44, 299.79it/s]	73% [REDACTED]	35765/49041 [02:1
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9<00:43, 303.06it/s]	73% [REDACTED]	35841/49041 [02:1
9<00:41, 318.08it/s]	73% [REDACTED]	35874/49041 [02:1
9<00:41, 314.24it/s]	73% [REDACTED]	35913/49041 [02:1
9<00:40, 327.85it/s]	73% [REDACTED]	

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0<00:49, 263.70it/s]	
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0<00:45, 283.98it/s]	
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0<00:43, 295.71it/s]	
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0<00:56, 229.49it/s]	
74% [REDACTED]	36183/49041 [02:2
0<00:50, 252.41it/s]	
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1<00:45, 278.67it/s]	
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1<00:51, 246.72it/s]	
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2<01:01, 203.77it/s]	
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2<00:53, 232.68it/s]	
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2<00:47, 261.18it/s]	
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2<00:44, 276.88it/s]	
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75% [REDACTED]	36804/49041 [02:2
3<00:37, 327.66it/s]	

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3<00:43, 278.21it/s]	75% [██████████]	36907/49041 [02:2
3<00:45, 268.41it/s]	75% [██████████]	36938/49041 [02:2
3<00:41, 289.90it/s]	75% [██████████]	36977/49041 [02:2
3<00:38, 312.92it/s]	75% [██████████]	37019/49041 [02:2
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4<00:40, 291.78it/s]	76% [██████████]	37090/49041 [02:2
4<00:43, 275.37it/s]	76% [██████████]	37121/49041 [02:2
4<00:47, 248.05it/s]	76% [██████████]	37150/49041 [02:2
4<00:53, 222.75it/s]	76% [██████████]	37177/49041 [02:2
4<00:47, 249.50it/s]	76% [██████████]	37213/49041 [02:2
4<00:47, 249.68it/s]	76% [██████████]	37240/49041 [02:2
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5<00:46, 250.71it/s]	76% [██████████]	37294/49041 [02:2
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77% [██████████]		37610/49041 [02:2

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9<00:43, 248.24it/s]	
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9<00:41, 260.86it/s]	

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82%		40362/49041 [02:3]
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83% [███████████] 0<00:39, 204.36it/s]	40936/49041 [02:4
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87% [REDACTED]	42895/49041 [02:4
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```
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100%|██████████| 48931/49041 [03:  
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100%|██████████| 48989/49041 [03:  
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100%|██████████| 49020/49041 [03:  
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100%|██████████| 49041/49041 [03:  
10<00:00, 257.97it/s]  
|██████████| 49041  
49041  
300
```

In [108]:

```
tfidf_weighted_w2v_on_essay_matrix_Xtrain = np.vstack(tfidf_weighted_w2v_vectors_eassays_Xtrain)  
print(tfidf_weighted_w2v_on_essay_matrix_Xtrain.shape)
```

(49041, 300)

8.5.2. TFIDF weighted Word2Vec encoding of preprocessed_essays on Test Data

In [109]:

```
# # S = ["abc def pqr", "def def def abc", "pqr pqr def"]  
# tfidf_model = TfidfVectorizer()  
# tfidf_model.fit(X_test['preprocessed_essays'].values)  
# # we are converting a dictionary with word as a key, and the idf as a value  
# dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))  
# tfidf_words = set(tfidf_model.get_feature_names())
```

In [110]:

```
# average Word2Vec  
# compute average word2vec for each review.  
tfidf_weighted_w2v_vectors_eassays_Xtest = [] # the avg-w2v for each sentence/review is stored in  
this list  
for sentence in tqdm(X_test['preprocessed_essays'].values): # for each review/sentence  
    vector = np.zeros(300) # as word vectors are of zero length  
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review  
    for word in sentence.split(): # for each word in a review/sentence  
        if (word in glove_words) and (word in tfidf_words):  
            vec = model[word] # getting the vector for each word  
            # here we are multiplying idf value(dictionary[word]) and the tf  
            value((sentence.count(word)/len(sentence.split())))  
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
```

```

    idf value for each word
        vector += (vec * tf_idf) # calculating tfidf weighted w2v
        tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_weighted_w2v_vectors_eassays_Xtest.append(vector)

print(len(tfidf_weighted_w2v_vectors_eassays_Xtest))
print(len(tfidf_weighted_w2v_vectors_eassays_Xtest[0]))

```

0%| [00:00<?, ?it/s] | 0/36
 0%| 1:45, 342.69it/s] | 37/36052 [00:0
 0%|| 1:44, 343.73it/s] | 73/36052 [00:0
 0%|| 1:47, 335.40it/s] | 106/36052 [00:0
 0%|| 1:47, 332.83it/s] | 140/36052 [00:0
 0%|| 2:00, 298.60it/s] | 166/36052 [00:0
 1%|| 1:56, 306.59it/s] | 200/36052 [00:0
 1%|| 2:13, 267.38it/s] | 228/36052 [00:0
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 1%|| 2:00, 295.40it/s] | 329/36052 [00:0
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 1%|| 2:19, 255.19it/s] | 518/36052 [00:0
 2%|| 2:09, 273.18it/s] | 552/36052 [00:0
 2%|| 2:16, 260.11it/s] | 581/36052 [00:0
 2%|| 2:06, 279.13it/s] | 616/36052 [00:0
 2%|| 2:08, 275.90it/s] | 645/36052 [00:0
 2%|| 2:02, 289.47it/s] | 679/36052 [00:0

2% ██████████	2:05, 282.64it/s]	709/36052 [00:C]
2% ██████████	2:04, 284.39it/s]	739/36052 [00:C]
2% ██████████	2:25, 242.94it/s]	768/36052 [00:C]
2% ██████████	2:40, 220.15it/s]	794/36052 [00:C]
2% ██████████	2:53, 202.87it/s]	818/36052 [00:C]
2% ██████████	3:01, 194.51it/s]	840/36052 [00:C]
2% ██████████	3:06, 188.24it/s]	861/36052 [00:C]
2% ██████████	3:05, 189.45it/s]	881/36052 [00:C]
2% ██████████	3:06, 188.17it/s]	901/36052 [00:C]
3% ██████████	3:00, 194.63it/s]	923/36052 [00:C]
3% ██████████	3:09, 185.32it/s]	943/36052 [00:C]
3% ██████████	3:16, 178.30it/s]	962/36052 [00:C]
3% ██████████	3:07, 187.14it/s]	984/36052 [00:C]
3% ██████████	3:22, 173.01it/s]	1004/36052 [00:C]
3% ██████████	3:22, 173.03it/s]	1022/36052 [00:C]
3% ██████████	3:19, 175.84it/s]	1041/36052 [00:C]
3% ██████████	3:24, 171.03it/s]	1059/36052 [00:C]
3% ██████████	3:26, 169.71it/s]	1077/36052 [00:C]
3% ██████████	3:23, 171.54it/s]	1096/36052 [00:C]
3% ██████████	3:17, 177.30it/s]	1116/36052 [00:C]
3% ██████████	3:07, 186.36it/s]	1138/36052 [00:C]
3% ██████████	2:56, 197.95it/s]	1163/36052 [00:C]
3% ██████████	2:48, 206.78it/s]	1187/36052 [00:C]
3% ██████████	2:36, 222.26it/s]	1215/36052 [00:C]
3% ██████████	2:14, 258.07it/s]	1258/36052 [00:C]
4% ██████████	2:00, 261.70it/s]	1297/36052 [00:C]

2:02, 284.73it/s]		
4% ██████████	1331/36052 [00:c]	
1:57, 296.22it/s]		
4% ██████████	1363/36052 [00:c]	
2:03, 280.68it/s]		
4% ██████████	1393/36052 [00:c]	
2:05, 276.72it/s]		
4% ██████████	1431/36052 [00:c]	
1:55, 298.47it/s]		
4% ██████████	1463/36052 [00:c]	
1:57, 294.55it/s]		
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[00:06<01:52, 307.02it/s]		
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[00:07<02:12, 256.41it/s]		
6% ██████████	2053/36052	
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[00:08<02:03, 275.25it/s]		

6% [██████]	2112/36052
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7% [██████]	2407/36052
[00:09<01:58, 284.68it/s]	
7% [██████]	2437/36052
[00:09<01:58, 282.58it/s]	
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[00:09<02:00, 278.19it/s]	
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[00:09<02:06, 265.07it/s]	
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[00:09<02:04, 269.08it/s]	
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[00:09<02:04, 269.13it/s]	
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[00:10<02:05, 265.15it/s]	
8% [██████]	2860/36052
[00:10<02:05, 263.49it/s]	
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10% ███████	3702/36052

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10% [REDACTED]
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11% [REDACTED]
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01:41, 287.46it/s]	

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01:33, 303.29it/s]

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01:34, 298.06it/s]

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22% [██████████] | 7829/36052 [00:2
01:37, 288.47it/s]

22% [██████████] | 7858/36052 [00:2
01:42, 275.77it/s]

22% [██████████] | 7886/36052 [00:2
01:42, 273.80it/s]

22% [██████████] | 7914/36052 [00:2
01:45, 266.22it/s]

22% [██████████] | 7947/36052 [00:2
01:40, 279.73it/s]

22% [██████████] | 7981/36052 [00:2
01:35, 292.42it/s]

22% [██████████] | 8011/36052 [00:2
01:37, 287.89it/s]

22% [██████████] | 8041/36052 [00:2
01:38, 284.81it/s]

22% [██████████] | 8071/36052 [00:2
01:37, 285.92it/s]

22% [██████████] | 8100/36052 [00:2
01:43, 271.04it/s]

23% [██████████] | 8131/36052 [00:2
01:41, 275.67it/s]

23% [██████████] | 8166/36052 [00:2
01:35, 291.52it/s]

23% [██████████] | 8198/36052 [00:2
01:34, 296.21it/s]

23% [██████████] | 8228/36052 [00:2
01:41, 274.51it/s]

23% [██████████] | 8257/36052 [00:2
01:40, 275.82it/s]

23% [██████████] | 8288/36052 [00:2
01:38, 282.16it/s]

23% [██████████] | 8317/36052 [00:2
01:39, 277.95it/s]

23% [██████████] | 8352/36052 [00:2
01:34, 293.28it/s]

23% [██████████] | 8383/36052 [00:3
01:32, 297.76it/s]

23% [██████████] | 8416/36052 [00:3
01:32, 300.09it/s]

23% [██████████] | 8448/36052 [00:3
01:31, 302.34it/s]

24% [██████████] | 8480/36052 [00:3

01:29, 307.26it/s]	24% [██████████]	8512/36052 [00:3
01:29, 307.42it/s]	24% [██████████]	8548/36052 [00:3
01:26, 318.10it/s]	24% [██████████]	8583/36052 [00:3
01:24, 323.46it/s]	24% [██████████]	8616/36052 [00:3
01:27, 314.26it/s]	24% [██████████]	8648/36052 [00:3
01:28, 308.67it/s]	24% [██████████]	8679/36052 [00:3
01:38, 279.05it/s]	24% [██████████]	8708/36052 [00:3
01:39, 275.81it/s]	24% [██████████]	8742/36052 [00:3
01:34, 289.42it/s]	24% [██████████]	8775/36052 [00:3
01:31, 297.28it/s]	24% [██████████]	8809/36052 [00:3
01:29, 305.61it/s]	25% [██████████]	8840/36052 [00:3
01:29, 303.32it/s]	25% [██████████]	8876/36052 [00:3
01:26, 315.05it/s]	25% [██████████]	8909/36052 [00:3
01:25, 315.75it/s]	25% [██████████]	8941/36052 [00:3
01:28, 306.10it/s]	25% [██████████]	8972/36052 [00:3
01:33, 290.06it/s]	25% [██████████]	9003/36052 [00:3
01:33, 289.16it/s]	25% [██████████]	9033/36052 [00:3
01:35, 282.45it/s]	25% [██████████]	9064/36052 [00:3
01:34, 286.99it/s]	25% [██████████]	9099/36052 [00:3
01:29, 300.27it/s]	25% [██████████]	9130/36052 [00:3
01:29, 299.63it/s]	25% [██████████]	9161/36052 [00:3
01:33, 289.14it/s]	25% [██████████]	9192/36052 [00:3
01:32, 291.79it/s]	26% [██████████]	9222/36052 [00:3
01:36, 277.88it/s]	26% [██████████]	9251/36052 [00:3
01:42, 260.22it/s]	26% [██████████]	9278/36052 [00:3
01:44, 257.08it/s]	26% [██████████]	

26% [███████████] 01:41, 263.26it/s]	9307/36052 [00:3
26% [███████████] 01:53, 234.83it/s]	9334/36052 [00:3
26% [███████████] 01:51, 239.20it/s]	9360/36052 [00:3
26% [███████████] 01:46, 249.88it/s]	9389/36052 [00:3
26% [███████████] 01:43, 257.93it/s]	9418/36052 [00:3
26% [███████████] 01:42, 258.42it/s]	9445/36052 [00:3
26% [███████████] 01:38, 269.21it/s]	9476/36052 [00:3
26% [███████████] 02:01, 218.76it/s]	9504/36052 [00:3
26% [███████████] 01:59, 222.25it/s]	9528/36052 [00:3
26% [███████████] 02:03, 215.09it/s]	9552/36052 [00:3
27% [███████████] 02:01, 217.19it/s]	9576/36052 [00:3
27% [███████████] 01:59, 221.12it/s]	9600/36052 [00:3
27% [███████████] 02:03, 213.75it/s]	9623/36052 [00:3
27% [███████████] 02:13, 197.07it/s]	9645/36052 [00:3
27% [███████████] 02:23, 183.91it/s]	9666/36052 [00:3
27% [███████████] 02:16, 193.71it/s]	9689/36052 [00:3
27% [███████████] 02:32, 173.21it/s]	9709/36052 [00:3
27% [███████████] 02:32, 172.14it/s]	9728/36052 [00:3
27% [███████████] 02:26, 180.12it/s]	9749/36052 [00:3
27% [███████████] 02:23, 183.63it/s]	9769/36052 [00:3
27% [███████████] 02:10, 201.32it/s]	9796/36052 [00:3
27% [███████████] 01:56, 224.63it/s]	9828/36052 [00:3
27% [███████████] 01:54, 229.24it/s]	9854/36052 [00:3
27% [███████████] 01:53, 229.72it/s]	9878/36052 [00:3
27% [███████████] 01:46, 244.68it/s]	9908/36052 [00:3
28% [███████████] 01:44, 249.00it/s]	9935/36052 [00:3

28% [REDACTED]	9961/36052 [00:3
01:47, 243.71it/s]	
28% [REDACTED]	9998/36052 [00:3
01:37, 266.83it/s]	
28% [REDACTED]	10026/36052 [00:3
01:38, 264.54it/s]	
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01:39, 260.02it/s]	
28% [REDACTED]	10083/36052 [00:3
01:37, 265.41it/s]	
28% [REDACTED]	10119/36052 [00:3
01:30, 285.40it/s]	
28% [REDACTED]	10149/36052 [00:3
01:37, 265.09it/s]	
28% [REDACTED]	10177/36052 [00:3
01:48, 239.06it/s]	
28% [REDACTED]	10202/36052 [00:3
01:56, 221.63it/s]	
28% [REDACTED]	10226/36052 [00:3
01:57, 219.40it/s]	
28% [REDACTED]	10249/36052 [00:3
02:02, 210.29it/s]	
28% [REDACTED]	10271/36052 [00:3
02:09, 199.24it/s]	
29% [REDACTED]	10292/36052 [00:3
02:19, 185.23it/s]	
29% [REDACTED]	10312/36052 [00:3
02:18, 185.23it/s]	
29% [REDACTED]	10335/36052 [00:3
02:12, 194.73it/s]	
29% [REDACTED]	10355/36052 [00:3
02:13, 191.78it/s]	
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01:56, 220.21it/s]	
29% [REDACTED]	10422/36052 [00:3
01:47, 238.60it/s]	
29% [REDACTED]	10448/36052 [00:3
02:04, 205.27it/s]	
29% [REDACTED]	10471/36052 [00:3
02:04, 205.31it/s]	
29% [REDACTED]	10493/36052 [00:3
02:04, 204.83it/s]	
29% [REDACTED]	10515/36052 [00:3
02:10, 195.78it/s]	
29% [REDACTED]	10539/36052 [00:3
02:04, 205.12it/s]	
29% [REDACTED]	10561/36052 [00:3
02:05, 202.45it/s]	
29% [REDACTED]	10584/36052 [00:3
02:02, 207.74it/s]	
29% [REDACTED]	10606/36052 [00:3

29% ███████████	02:05, 201.98it/s]	10000/36052 [00:3
29% ███████████	01:58, 214.34it/s]	10632/36052 [00:3
30% ███████████	01:51, 228.17it/s]	10661/36052 [00:3
30% ███████████	01:46, 239.12it/s]	10689/36052 [00:3
30% ███████████	01:55, 219.35it/s]	10714/36052 [00:3
30% ███████████	01:51, 227.74it/s]	10740/36052 [00:3
30% ███████████	01:53, 223.55it/s]	10764/36052 [00:3
30% ███████████	01:48, 233.29it/s]	10791/36052 [00:3
30% ███████████	01:44, 240.63it/s]	10818/36052 [00:3
30% ███████████	01:47, 235.14it/s]	10843/36052 [00:4
30% ███████████	01:48, 231.12it/s]	10867/36052 [00:4
30% ███████████	02:00, 209.27it/s]	10891/36052 [00:4
30% ███████████	01:59, 209.98it/s]	10913/36052 [00:4
30% ███████████	02:06, 199.04it/s]	10935/36052 [00:4
30% ███████████	02:00, 207.62it/s]	10959/36052 [00:4
30% ███████████	02:02, 204.15it/s]	10981/36052 [00:4
31% ███████████	02:07, 196.63it/s]	11002/36052 [00:4
31% ███████████	02:09, 193.07it/s]	11022/36052 [00:4
31% ███████████	02:06, 198.28it/s]	11044/36052 [00:4
31% ███████████	01:54, 218.81it/s]	11074/36052 [00:4
31% ███████████	01:54, 217.04it/s]	11097/36052 [00:4
31% ███████████	01:49, 228.28it/s]	11124/36052 [00:4
31% ███████████	01:51, 223.91it/s]	11148/36052 [00:4
31% ███████████	01:44, 237.99it/s]	11177/36052 [00:4
31% ███████████	01:42, 241.49it/s]	11203/36052 [00:4
31% ███████████	01:37, 253.91it/s]	11233/36052 [00:4

31% [REDACTED]	11259/36052 [00:4
01:44, 236.21it/s]	
31% [REDACTED]	11284/36052 [00:4
01:44, 237.47it/s]	
31% [REDACTED]	11315/36052 [00:4
01:38, 250.46it/s]	
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01:34, 260.42it/s]	
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01:31, 270.71it/s]	
32% [REDACTED]	11406/36052 [00:4
01:30, 273.12it/s]	
32% [REDACTED]	11434/36052 [00:4
01:33, 262.78it/s]	
32% [REDACTED]	11461/36052 [00:4
01:33, 261.84it/s]	
32% [REDACTED]	11488/36052 [00:4
01:34, 261.19it/s]	
32% [REDACTED]	11517/36052 [00:4
01:32, 266.27it/s]	
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01:33, 261.19it/s]	
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01:36, 254.83it/s]	
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[00:43<01:34, 259.01it/s]	
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[00:43<01:39, 243.93it/s]	
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[00:44<01:38, 245.36it/s]	
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[00:44<01:35, 254.55it/s]	
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[00:44<01:29, 270.62it/s]	
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[00:44<01:24, 286.06it/s]	
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[00:44<01:20, 264.62it/s]	

[00:44<01:50, 204.02it/s]		
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[00:53<01:18, 276.55it/s]	40% [REDACTED]	14331/36052
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[00:53<01:19, 271.46it/s]	40% [REDACTED]	14418/36052
[00:53<01:19, 270.82it/s]	40% [REDACTED]	14450/36052
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[00:54<01:12, 293.21it/s]	41% [REDACTED]	14759/36052
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[00:55<01:10, 302.07it/s]	41% [REDACTED]	14919/36052
[00:55<01:09, 303.77it/s]	41% [REDACTED]	14950/36052
[00:55<01:12, 291.83it/s]	42% [REDACTED]	14980/36052 [00:5<01:16, 274.85it/s]
[00:55<01:16, 276.06it/s]	42% [REDACTED]	15009/36052 [00:5<01:16, 276.06it/s]

42% [REDACTED]	15039/36052 [00:5
<01:15, 279.69it/s]	
42% [REDACTED]	15068/36052 [00:5
<01:21, 258.53it/s]	
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<01:25, 244.79it/s]	
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<01:24, 246.36it/s]	
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<01:23, 250.21it/s]	
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<01:24, 247.31it/s]	
42% [REDACTED]	15206/36052 [00:5
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67% [01:29<00:43, 271.93it/s]	24274/36052
67% [01:29<00:42, 276.71it/s]	24304/36052
68% [01:29<00:40, 287.78it/s]	24337/36052

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68% [███████████] [01:29<00:38, 300.03it/s]	24403/36052
68% [███████████] [01:29<00:38, 299.48it/s]	24434/36052
68% [███████████] [01:29<00:37, 307.25it/s]	24468/36052
68% [███████████] [01:29<00:36, 312.92it/s]	24502/36052
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68% [███████████] [01:30<00:37, 308.26it/s]	24570/36052
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68% [███████████] [01:30<00:35, 318.65it/s]	24638/36052
68% [███████████] [01:30<00:36, 311.07it/s]	24671/36052
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69% [███████████] [01:30<00:35, 318.09it/s]	24805/36052
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69% [███████████] [01:31<00:37, 296.42it/s]	24901/36052
69% [███████████] [01:31<00:36, 307.43it/s]	24936/36052
69% [███████████] [01:31<00:37, 294.18it/s]	24967/36052
69% [███████████] [01:31<00:38, 285.78it/s]	24997/36052
69% [███████████] [01:31<00:37, 294.57it/s]	25030/36052
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70% [███████████] 1<00:36, 299.87it/s]	25094/36052 [01:3]
70% [███████████] 1<00:35, 310.04it/s]	25129/36052 [01:3]
70% [███████████] 2<00:36, 295.62it/s]	25161/36052 [01:3]
70% [███████████] 2<00:36, 295.30it/s]	25192/36052 [01:3]

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2<00:38, 280.49it/s]		
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```

In [111]:

```
tfidf_weighted_w2v_on_essay_matrix_Xtest = np.vstack(tfidf_weighted_w2v_vectors_eassays_Xtest)  
print(tfidf_weighted_w2v_on_essay_matrix_Xtest.shape)
```

(36052, 300)

8.5.3. TFIDF weighted Word2Vec encoding of preprocessed_essays on CV Data

In [112]:

```
# # S = ["abc def pqr", "def def def abc", "pqr pqr def"]  
# tfidf_model = TfidfVectorizer()  
# tfidf_model.fit(X_cv['preprocessed_essays'].values)  
# # we are converting a dictionary with word as a key, and the idf as a value  
# dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))  
# tfidf_words = set(tfidf_model.get_feature_names())
```

In [113]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_weighted_w2v_vectors_eassays_Xcv = [] # the avg-w2v for each sentence/review is stored in this list
for sentence in tqdm(X_cv['preprocessed_essays'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in sentence.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
            value((sentence.count(word)/len(sentence.split())))
            tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split())) # getting the tf
            idf value for each word
            vector += (vec * tf_idf) # calculating tfidf weighted w2v
            tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_weighted_w2v_vectors_eassays_Xcv.append(vector)

print(len(tfidf_weighted_w2v_vectors_eassays_Xcv))
print(len(tfidf_weighted_w2v_vectors_eassays_Xcv[0]))
```

```
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0%|| 1:09, 346.01it/s] | 70/24155 [00:C
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1%|| 1:16, 315.48it/s] | 139/24155 [00:C
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1%|| 1:22, 287.73it/s] | 352/24155 [00:C
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2%|| 1:23, 284.88it/s] | 475/24155 [00:C
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```

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[00:10<01:23, 255.35it/s]	
12% [██████]	2831/24155
[00:10<01:23, 256.65it/s]	

12% [███████]	2857/24155
[00:10<01:27, 243.21it/s]	
12% [███████]	2882/24155
[00:10<01:28, 239.59it/s]	
12% [███████]	2908/24155
[00:10<01:27, 242.64it/s]	
12% [███████]	2934/24155
[00:10<01:26, 244.82it/s]	
12% [███████]	2959/24155
[00:10<01:28, 240.67it/s]	
12% [███████]	2986/24155
[00:10<01:27, 240.81it/s]	
12% [███████]	3011/24155
[00:10<01:31, 230.07it/s]	
13% [███████]	3035/24155
[00:11<01:32, 227.68it/s]	
13% [███████]	3058/24155
[00:11<01:37, 215.54it/s]	
13% [███████]	3099/24155
[00:11<01:23, 250.70it/s]	
13% [███████]	3127/24155
[00:11<01:24, 247.85it/s]	
13% [███████]	3154/24155
[00:11<01:24, 248.51it/s]	
13% [███████]	3188/24155
[00:11<01:18, 267.78it/s]	
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[00:11<01:15, 276.24it/s]	
13% [███████]	3248/24155
[00:11<01:16, 273.90it/s]	
14% [███████]	3277/24155
[00:11<01:16, 272.28it/s]	
14% [███████]	3305/24155
[00:12<01:17, 268.26it/s]	
14% [███████]	3333/24155
[00:12<01:19, 262.53it/s]	
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[00:12<01:17, 267.24it/s]	
14% [███████]	3390/24155
[00:12<01:18, 264.81it/s]	
14% [███████]	3417/24155
[00:12<01:25, 243.33it/s]	
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[00:12<01:24, 245.31it/s]	
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[00:12<01:20, 256.85it/s]	
14% [███████]	3500/24155
[00:12<01:32, 224.31it/s]	
15% [███████]	3524/24155
[00:13<01:36, 214.11it/s]	
15% [███████]	3547/24155

[00:13<01:46, 194.29it/s]	
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[00:13<01:44, 196.03it/s]	
15% [██████████]	3613/24155
[00:13<01:56, 175.91it/s]	
15% [██████████]	3632/24155
[00:13<01:55, 177.90it/s]	
15% [██████████]	3651/24155
[00:13<01:54, 179.33it/s]	
15% [██████████]	3670/24155
[00:13<02:01, 168.81it/s]	
15% [██████████]	3689/24155
[00:13<01:58, 172.76it/s]	
15% [██████████]	3707/24155
[00:14<01:59, 170.90it/s]	
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[00:14<01:57, 174.29it/s]	
15% [██████████]	3744/24155
[00:14<02:02, 166.23it/s]	
16% [██████████]	3765/24155
[00:14<01:58, 172.10it/s]	
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[00:14<01:48, 188.14it/s]	
16% [██████████]	3823/24155
[00:14<01:34, 214.32it/s]	
16% [██████████]	3853/24155
[00:14<01:27, 232.24it/s]	
16% [██████████]	3883/24155
[00:14<01:22, 246.68it/s]	
16% [██████████]	3910/24155
[00:14<01:24, 239.77it/s]	
16% [██████████]	3940/24155
[00:15<01:20, 252.58it/s]	
16% [██████████]	3967/24155
[00:15<01:20, 251.81it/s]	
17% [██████████]	3993/24155
[00:15<01:20, 251.29it/s]	
17% [██████████]	4019/24155
[00:15<01:25, 234.62it/s]	
17% [██████████]	4044/24155
[00:15<01:25, 236.34it/s]	
17% [██████████]	4071/24155
[00:15<01:22, 242.89it/s]	
17% [██████████]	4098/24155
[00:15<01:20, 247.65it/s]	
17% [██████████]	4131/24155
[00:15<01:15, 264.71it/s]	
17% [██████████]	4159/24155
[00:15<01:15, 266.09it/s]	

17% ███████████	4193/24155 [00:1 01:12, 276.34it/s]
17% ███████████	4221/24155 [00:1 01:12, 274.19it/s]
18% ███████████	4249/24155 [00:1 01:14, 266.48it/s]
18% ███████████	4277/24155 [00:1 01:13, 269.93it/s]
18% ███████████	4310/24155 [00:16<01:10, 279.72it/s]
18% ███████████	4339/24155 [00:1 01:14, 264.22it/s]
18% ███████████	4375/24155 [00:1 01:09, 284.43it/s]
18% ███████████	4405/24155 [00:1 01:13, 270.21it/s]
18% ███████████	4438/24155 [00:1 01:09, 282.83it/s]
18% ███████████	4467/24155 [00:17<01:13, 266.14it/s]
19% ███████████	4495/24155 [00:1 01:17, 252.62it/s]
19% ███████████	4522/24155 [00:1 01:17, 254.70it/s]
19% ███████████	4548/24155 [00:1 01:17, 253.29it/s]
19% ███████████	4574/24155 [00:1 01:18, 249.41it/s]
19% ███████████	4600/24155 [00:1 01:19, 246.77it/s]
19% ███████████	4625/24155 [00:1 01:22, 236.50it/s]
19% ███████████	4649/24155 [00:1 01:25, 229.38it/s]
19% ███████████	4674/24155 [00:1 01:23, 232.59it/s]
19% ███████████	4701/24155 [00:1 01:21, 240.11it/s]
20% ███████████	4730/24155 [00:1 01:17, 250.57it/s]
20% ███████████	4756/24155 [00:1 01:17, 250.42it/s]
20% ███████████	4787/24155 [00:1 01:13, 263.05it/s]
20% ███████████	4814/24155 [00:1 01:14, 259.01it/s]
20% ███████████	4841/24155 [00:1 01:15, 256.26it/s]
20% ███████████	4867/24155 [00:1 01:15, 254.37it/s]
20% ███████████	4898/24155 [00:1 01:13, 263.37it/s]

20% [██████████] | 4925/24155 [00:1
01:13, 262.26it/s]

21% [██████████] | 4952/24155 [00:1
01:19, 241.82it/s]

21% [██████████] | 4977/24155 [00:1
01:20, 238.64it/s]

21% [██████████] | 5003/24155 [00:1
01:20, 239.29it/s]

21% [██████████] | 5028/24155 [00:1
01:19, 239.63it/s]

21% [██████████] | 5053/24155 [00:1
01:19, 239.87it/s]

21% [██████████] | 5078/24155 [00:1
01:22, 232.03it/s]

21% [██████████] | 5106/24155 [00:1
01:19, 239.59it/s]

21% [██████████] | 5132/24155 [00:1
01:18, 242.64it/s]

21% [██████████] | 5163/24155 [00:1
01:13, 257.00it/s]

22% [██████████] | 5195/24155 [00:1
01:10, 270.38it/s]

22% [██████████] | 5225/24155 [00:2
01:08, 275.58it/s]

22% [██████████] | 5253/24155 [00:2
01:15, 250.19it/s]

22% [██████████] | 5280/24155 [00:2
01:14, 252.97it/s]

22% [██████████] | 5312/24155 [00:2
01:11, 264.59it/s]

22% [██████████] | 5344/24155 [00:2
01:08, 276.22it/s]

22% [██████████] | 5381/24155 [00:2
01:04, 293.28it/s]

22% [██████████] | 5412/24155 [00:2
01:04, 291.40it/s]

23% [██████████] | 5442/24155 [00:2
01:08, 271.61it/s]

23% [██████████] | 5470/24155 [00:2
01:09, 267.79it/s]

23% [██████████] | 5498/24155 [00:2
01:09, 268.25it/s]

23% [██████████] | 5526/24155 [00:2
01:09, 268.56it/s]

23% [██████████] | 5554/24155 [00:2
01:10, 262.74it/s]

23% [██████████] | 5593/24155 [00:2
01:04, 288.67it/s]

23% [██████████] | 5623/24155 [00:2
01:05, 282.12it/s]

23% [██████████] | 5656/24155 [00:2

01:03, 291.85it/s]		
24% [██████████]	5686/24155	[00:2
01:10, 263.29it/s]		
24% [██████████]	5714/24155	[00:2
01:12, 253.54it/s]		
24% [██████████]	5744/24155	[00:2
01:09, 263.12it/s]		
24% [██████████]	5774/24155	[00:2
01:08, 270.26it/s]		
24% [██████████]	5802/24155	[00:2
01:13, 249.75it/s]		
24% [██████████]	5833/24155	[00:2
01:09, 262.54it/s]		
24% [██████████]	5861/24155	[00:2
01:09, 264.53it/s]		
24% [██████████]	5890/24155	[00:2
01:07, 268.69it/s]		
25% [██████████]	5920/24155	[00:2
01:07, 271.36it/s]		
25% [██████████]	5948/24155	[00:2
01:08, 267.64it/s]		
25% [██████████]	5975/24155	[00:2
01:10, 256.15it/s]		
25% [██████████]	6002/24155	[00:2
01:10, 257.20it/s]		
25% [██████████]	6031/24155	[00:2
01:08, 263.35it/s]		
25% [██████████]	6068/24155	[00:2
01:03, 285.61it/s]		
25% [██████████]	6098/24155	[00:2
01:03, 283.24it/s]		
25% [██████████]	6127/24155	[00:2
01:07, 266.41it/s]		
25% [██████████]	6155/24155	[00:2
01:08, 261.28it/s]		
26% [██████████]	6185/24155	[00:2
01:06, 268.91it/s]		
26% [██████████]	6214/24155	[00:2
01:06, 271.83it/s]		
26% [██████████]	6243/24155	[00:2
01:05, 273.92it/s]		
26% [██████████]	6276/24155	[00:2
01:02, 285.66it/s]		
26% [██████████]	6305/24155	[00:2
01:08, 259.27it/s]		
26% [██████████]	6332/24155	[00:2
01:14, 240.02it/s]		
26% [██████████]	6357/24155	[00:2
01:20, 222.22it/s]		
26% [██████████]	6380/24155	[00:2
01:22, 214.47it/s]		

27% [███████████]	6403/24155 [00:2
01:24, 209.36it/s]	
27% [███████████]	6427/24155 [00:2
01:22, 215.37it/s]	
27% [███████████]	6457/24155 [00:2
01:15, 233.10it/s]	
27% [███████████]	6481/24155 [00:2
01:16, 232.42it/s]	
27% [███████████]	6509/24155 [00:2
01:12, 242.38it/s]	
27% [███████████]	6534/24155 [00:2
01:12, 241.79it/s]	
27% [███████████]	6560/24155 [00:2
01:12, 244.22it/s]	
27% [███████████]	6590/24155 [00:2
01:08, 256.01it/s]	
27% [███████████]	6623/24155 [00:2
01:04, 271.77it/s]	
28% [███████████]	6654/24155 [00:2
01:02, 279.19it/s]	
28% [███████████]	6683/24155 [00:2
01:04, 269.77it/s]	
28% [███████████]	6711/24155 [00:2
01:06, 260.59it/s]	
28% [███████████]	6746/24155 [00:2
01:02, 279.53it/s]	
28% [███████████]	6779/24155 [00:2
00:59, 289.89it/s]	
28% [███████████]	6811/24155 [00:2
00:58, 295.05it/s]	
28% [███████████]	6841/24155 [00:2
00:59, 289.66it/s]	
28% [███████████]	6874/24155 [00:2
00:57, 299.89it/s]	
29% [███████████]	6905/24155 [00:2
00:57, 299.39it/s]	
29% [███████████]	6936/24155 [00:2
00:59, 288.97it/s]	
29% [███████████]	6966/24155 [00:2
01:04, 267.24it/s]	
29% [███████████]	6997/24155 [00:2
01:02, 272.89it/s]	
29% [███████████]	7032/24155 [00:2
00:59, 289.33it/s]	
29% [███████████]	7062/24155 [00:2
01:02, 273.30it/s]	
29% [███████████]	7090/24155 [00:2
01:04, 262.90it/s]	
29% [███████████]	7122/24155 [00:2
01:01, 274.92it/s]	
30% [███████████]	7150/24155 [00:2
01:05, 258.11it/s]	

30% [REDACTED]	7177/24155 [00:2
01:06, 255.64it/s]	
30% [REDACTED]	7203/24155 [00:2
01:06, 253.94it/s]	
30% [REDACTED]	7231/24155 [00:2
01:05, 258.36it/s]	
30% [REDACTED]	7258/24155 [00:2
01:09, 242.06it/s]	
30% [REDACTED]	7286/24155 [00:2
01:07, 249.63it/s]	
30% [REDACTED]	7312/24155 [00:2
01:08, 244.13it/s]	
30% [REDACTED]	7337/24155 [00:2
01:10, 240.22it/s]	
31% [REDACTED]	7368/24155 [00:2
01:05, 255.09it/s]	
31% [REDACTED]	7394/24155 [00:2
01:06, 253.56it/s]	
31% [REDACTED]	7421/24155 [00:2
01:05, 255.37it/s]	
31% [REDACTED]	7453/24155 [00:2
01:02, 269.11it/s]	
31% [REDACTED]	7481/24155 [00:2
01:03, 263.10it/s]	
31% [REDACTED]	7508/24155 [00:2
01:04, 259.05it/s]	
31% [REDACTED]	7535/24155 [00:2
01:04, 259.24it/s]	
31% [REDACTED]	7562/24155 [00:2
01:06, 250.70it/s]	
31% [REDACTED]	7592/24155 [00:2
01:03, 260.97it/s]	
32% [REDACTED]	7621/24155 [00:2
01:02, 266.10it/s]	
32% [REDACTED]	7648/24155 [00:2
01:03, 261.08it/s]	
32% [REDACTED]	7679/24155 [00:2
01:00, 271.20it/s]	
32% [REDACTED]	7710/24155 [00:2
00:58, 278.76it/s]	
32% [REDACTED]	7741/24155 [00:2
00:57, 284.31it/s]	
32% [REDACTED]	7770/24155 [00:2
00:57, 282.67it/s]	
32% [REDACTED]	7799/24155 [00:2
00:58, 281.53it/s]	
32% [REDACTED]	7828/24155 [00:2
00:59, 274.37it/s]	
33% [REDACTED]	7858/24155 [00:2
00:58, 278.47it/s]	
33% [REDACTED]	7888/24155 [00:2
00:58, 278.47it/s]	

00:57, 281.42it/s]	7917/24155 [00:3
33% [REDACTED]	7951/24155 [00:3
00:56, 285.85it/s]	7980/24155 [00:3
33% [REDACTED]	8009/24155 [00:3
00:57, 280.44it/s]	8036/24155 [00:3
01:01, 264.64it/s]	8063/24155 [00:3
33% [REDACTED]	8089/24155 [00:3
01:04, 249.97it/s]	8120/24155 [00:3
01:01, 262.70it/s]	8147/24155 [00:3
01:01, 261.79it/s]	8174/24155 [00:3
34% [REDACTED]	8202/24155 [00:3
01:01, 258.13it/s]	8230/24155 [00:3
34% [REDACTED]	8257/24155 [00:3
01:01, 261.41it/s]	8291/24155 [00:3
34% [REDACTED]	8321/24155
[00:31<00:56, 280.08it/s]	8350/24155
35% [REDACTED]	8378/24155
[00:31<00:59, 265.96it/s]	8413/24155
35% [REDACTED]	8442/24155
[00:32<00:55, 282.34it/s]	8471/24155
35% [REDACTED]	8500/24155
[00:32<00:57, 274.22it/s]	8535/24155
35% [REDACTED]	8565/24155
[00:32<00:55, 280.08it/s]	8596/24155
36% [REDACTED]	8624/24155
[00:32<00:58, 267.75it/s]	

36% [00:32<00:56, 273.67it/s]	8654/24155
36% [00:32<00:54, 285.46it/s]	8687/24155
36% [00:33<00:56, 273.83it/s]	8716/24155
36% [00:33<00:58, 263.24it/s]	8744/24155
36% [00:33<00:57, 267.46it/s]	8774/24155
36% [00:33<00:55, 275.98it/s]	8805/24155
37% [00:33<01:01, 247.77it/s]	8833/24155
37% [00:33<01:01, 248.45it/s]	8859/24155
37% [00:33<00:57, 265.77it/s]	8892/24155
37% [00:33<00:56, 269.58it/s]	8921/24155
37% [00:33<00:55, 275.00it/s]	8951/24155
37% [00:34<00:56, 267.01it/s]	8979/24155
37% [00:34<00:57, 261.69it/s]	9007/24155
37% [00:34<00:57, 263.93it/s]	9035/24155
38% [00:34<00:55, 270.86it/s]	9065/24155
38% [00:34<00:54, 278.51it/s]	9096/24155
38% [00:34<00:55, 272.35it/s]	9125/24155
38% [00:34<00:53, 282.09it/s]	9157/24155
38% [00:34<00:51, 288.75it/s]	9190/24155
38% [00:34<00:55, 270.00it/s]	9220/24155
38% [00:35<00:53, 280.32it/s]	9252/24155
38% [00:35<00:51, 290.50it/s]	9285/24155
39% [00:35<00:53, 277.06it/s]	9315/24155
39% [00:35<00:51, 288.04it/s]	9348/24155
39% [00:35<00:53, 278.56it/s]	9378/24155
39% [00:35<00:53, 275.49it/s]	9407/24155

[00:35<00:53, 276.19it/s]	39% [██████████]	9437/24155
[00:35<00:52, 279.38it/s]	39% [██████████]	9468/24155
[00:35<00:52, 281.65it/s]	39% [██████████]	9499/24155
[00:36<00:55, 262.53it/s]	39% [██████████]	9528/24155
[00:36<00:56, 258.65it/s]	40% [██████████]	9555/24155
[00:36<00:53, 273.30it/s]	40% [██████████]	9589/24155
[00:36<00:53, 272.09it/s]	40% [██████████]	9617/24155
[00:36<00:53, 271.03it/s]	40% [██████████]	9646/24155
[00:36<00:51, 281.10it/s]	40% [██████████]	9678/24155
[00:36<00:51, 280.44it/s]	40% [██████████]	9707/24155
[00:36<00:54, 264.65it/s]	40% [██████████]	9736/24155
[00:36<00:54, 263.14it/s]	40% [██████████]	9763/24155
[00:37<00:54, 262.09it/s]	41% [██████████]	9790/24155
[00:37<00:56, 252.57it/s]	41% [██████████]	9817/24155
[00:37<00:53, 269.05it/s]	41% [██████████]	9850/24155
[00:37<00:49, 286.29it/s]	41% [██████████]	9885/24155
[00:37<00:49, 286.96it/s]	41% [██████████]	9915/24155
[00:37<00:49, 287.43it/s]	41% [██████████]	9945/24155
[00:37<00:51, 277.61it/s]	41% [██████████]	9975/24155
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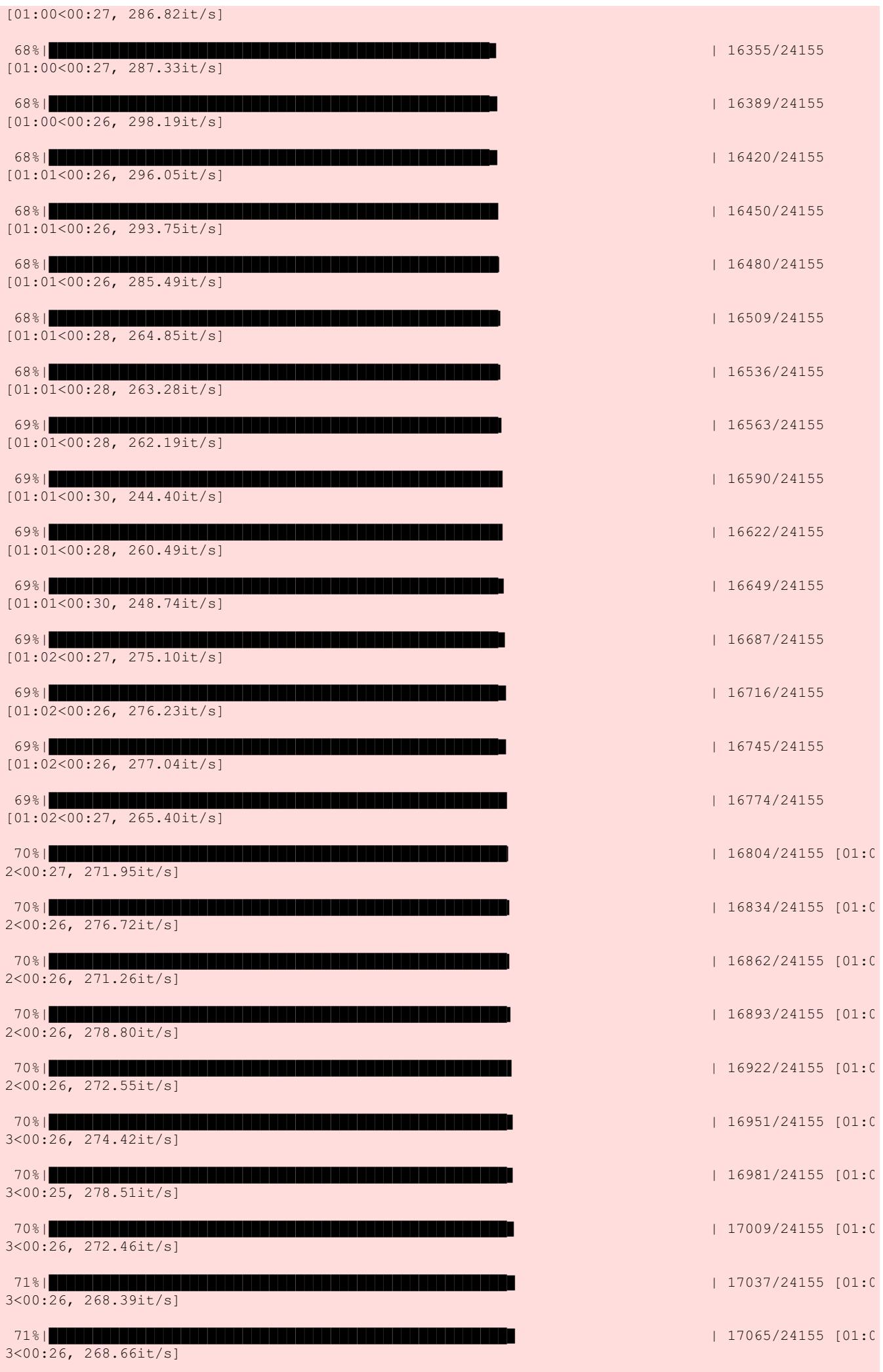
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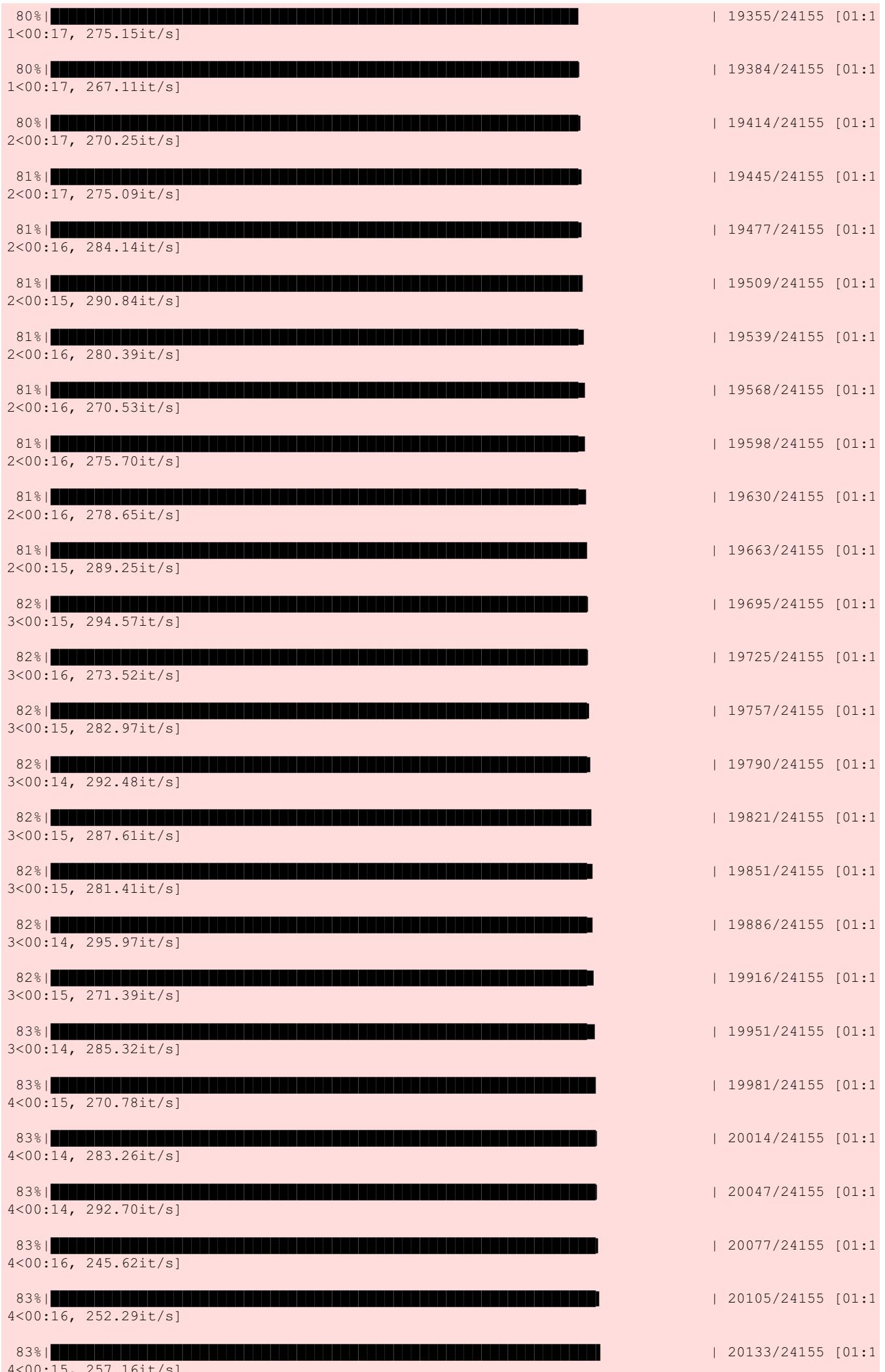
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76% [██████████] 8<00:20, 281.32it/s]	18434/24155 [01:C]
76% [██████████] 8<00:20, 280.60it/s]	18463/24155 [01:C]
77% [██████████] 8<00:21, 267.69it/s]	18492/24155 [01:C]
77% [██████████] 8<00:19, 285.21it/s]	18527/24155 [01:C]
77% [██████████] 8<00:19, 280.01it/s]	18556/24155 [01:C]
77% [██████████]	18587/24155 [01:C]

9<00:19, 285.22it/s]	77% [REDACTED]	18618/24155 [01:0
9<00:19, 288.98it/s]	77% [REDACTED]	18651/24155 [01:0
9<00:18, 296.95it/s]	77% [REDACTED]	18681/24155 [01:0
9<00:19, 287.60it/s]	77% [REDACTED]	18710/24155 [01:0
9<00:19, 284.94it/s]	78% [REDACTED]	18745/24155 [01:0
9<00:18, 298.70it/s]	78% [REDACTED]	18776/24155 [01:0
9<00:18, 295.13it/s]	78% [REDACTED]	18807/24155 [01:0
9<00:18, 296.03it/s]	78% [REDACTED]	18837/24155 [01:0
9<00:18, 283.74it/s]	78% [REDACTED]	18866/24155 [01:0
0<00:19, 275.83it/s]	78% [REDACTED]	18894/24155 [01:0
0<00:19, 273.84it/s]	78% [REDACTED]	18922/24155 [01:0
0<00:19, 266.24it/s]	78% [REDACTED]	18951/24155 [01:0
0<00:19, 269.92it/s]	79% [REDACTED]	18982/24155 [01:0
0<00:18, 277.81it/s]	79% [REDACTED]	19010/24155 [01:0
0<00:19, 268.87it/s]	79% [REDACTED]	19038/24155 [01:0
0<00:20, 254.34it/s]	79% [REDACTED]	19067/24155 [01:0
0<00:19, 261.24it/s]	79% [REDACTED]	19096/24155 [01:0
0<00:18, 266.31it/s]	79% [REDACTED]	19123/24155 [01:0
1<00:19, 261.22it/s]	79% [REDACTED]	19150/24155 [01:0
1<00:19, 260.76it/s]	79% [REDACTED]	19177/24155 [01:0
1<00:19, 260.43it/s]	80% [REDACTED]	19204/24155 [01:0
1<00:19, 248.71it/s]	80% [REDACTED]	19235/24155 [01:0
1<00:18, 261.73it/s]	80% [REDACTED]	19264/24155 [01:0
1<00:18, 266.66it/s]	80% [REDACTED]	19291/24155 [01:0
1<00:19, 249.84it/s]	80% [REDACTED]	19320/24155 [01:0
1<00:18, 255.19it/s]	80% [REDACTED]	



83% [REDACTED]	20160/24155 [01:1
4<00:15, 252.13it/s]	
84% [REDACTED]	20191/24155 [01:1
4<00:14, 264.38it/s]	
84% [REDACTED]	20230/24155 [01:1
5<00:13, 290.06it/s]	
84% [REDACTED]	20261/24155 [01:1
5<00:13, 285.97it/s]	
84% [REDACTED]	20293/24155 [01:1
5<00:13, 292.18it/s]	
84% [REDACTED]	20325/24155 [01:1
5<00:12, 296.67it/s]	
84% [REDACTED]	20356/24155 [01:1
5<00:13, 280.98it/s]	
84% [REDACTED]	20386/24155 [01:1
5<00:13, 283.20it/s]	
85% [REDACTED]	20416/24155 [01:1
5<00:13, 281.58it/s]	
85% [REDACTED]	20446/24155 [01:1
5<00:13, 283.60it/s]	
85% [REDACTED]	20475/24155 [01:1
5<00:13, 282.21it/s]	
85% [REDACTED]	20504/24155 [01:1
5<00:13, 277.97it/s]	
85% [REDACTED]	20532/24155 [01:1
6<00:13, 262.91it/s]	
85% [REDACTED]	20563/24155 [01:1
6<00:13, 269.73it/s]	
85% [REDACTED]	20591/24155 [01:1
6<00:13, 269.60it/s]	
85% [REDACTED]	20620/24155 [01:1
6<00:12, 272.33it/s]	
85% [REDACTED]	20648/24155 [01:1
6<00:13, 262.26it/s]	
86% [REDACTED]	20675/24155 [01:1
6<00:13, 258.46it/s]	
86% [REDACTED]	20701/24155 [01:1
6<00:13, 250.00it/s]	
86% [REDACTED]	20727/24155 [01:1
6<00:13, 250.02it/s]	
86% [REDACTED]	20754/24155 [01:1
6<00:13, 252.84it/s]	
86% [REDACTED]	20780/24155 [01:1
7<00:13, 254.94it/s]	
86% [REDACTED]	20806/24155 [01:1
7<00:13, 247.69it/s]	
86% [REDACTED]	20835/24155 [01:1
7<00:12, 256.30it/s]	
86% [REDACTED]	20861/24155 [01:1
7<00:12, 254.39it/s]	
86% [REDACTED]	20894/24155 [01:1

7<00:12, 270.50it/s]	20991/24155 [01:1]
87% [REDACTED]	20926/24155 [01:1]
7<00:11, 277.78it/s]	
87% [REDACTED]	20955/24155 [01:1]
7<00:11, 274.95it/s]	
87% [REDACTED]	20985/24155 [01:1]
7<00:11, 275.82it/s]	
87% [REDACTED]	21013/24155 [01:1]
7<00:11, 264.52it/s]	
87% [REDACTED]	21040/24155 [01:1]
8<00:12, 251.30it/s]	
87% [REDACTED]	21073/24155 [01:1]
8<00:11, 268.04it/s]	
87% [REDACTED]	21104/24155 [01:1]
8<00:11, 276.42it/s]	
87% [REDACTED]	21134/24155 [01:1]
8<00:10, 279.94it/s]	
88% [REDACTED]	21164/24155 [01:1]
8<00:10, 282.47it/s]	
88% [REDACTED]	21194/24155 [01:1]
8<00:10, 281.07it/s]	
88% [REDACTED]	21223/24155 [01:1]
8<00:10, 280.41it/s]	
88% [REDACTED]	21252/24155 [01:1]
8<00:10, 267.57it/s]	
88% [REDACTED]	21282/24155 [01:1]
8<00:10, 273.54it/s]	
88% [REDACTED]	21310/24155 [01:1]
8<00:10, 272.25it/s]	
88% [REDACTED]	21342/24155 [01:1]
9<00:09, 282.01it/s]	
88% [REDACTED]	21371/24155 [01:1]
9<00:10, 271.60it/s]	
89% [REDACTED]	21403/24155 [01:1]
9<00:09, 281.52it/s]	
89% [REDACTED]	21438/24155 [01:1]
9<00:09, 296.06it/s]	
89% [REDACTED]	21468/24155 [01:1]
9<00:09, 290.36it/s]	
89% [REDACTED]	21498/24155 [01:1]
9<00:09, 283.24it/s]	
89% [REDACTED]	21527/24155 [01:1]
9<00:09, 278.68it/s]	
89% [REDACTED]	21556/24155 [01:1]
9<00:09, 275.57it/s]	
89% [REDACTED]	21584/24155 [01:1]
9<00:09, 264.36it/s]	
89% [REDACTED]	21611/24155 [01:2]
0<00:09, 262.94it/s]	
90% [REDACTED]	21638/24155
[01:20<00:09, 261.95it/s]	

90% [███████████] [01:20<00:09, 255.34it/s]	21665/24155
90% [███████████] [01:20<00:09, 256.62it/s]	21692/24155
90% [███████████] [01:20<00:08, 272.26it/s]	21725/24155
90% [███████████] [01:20<00:08, 284.39it/s]	21758/24155
90% [███████████] [01:20<00:07, 300.50it/s]	21794/24155
90% [███████████] [01:20<00:08, 271.45it/s]	21825/24155
90% [███████████] [01:20<00:08, 258.99it/s]	21854/24155
91% [███████████] [01:21<00:08, 261.99it/s]	21882/24155
91% [███████████] [01:21<00:08, 261.30it/s]	21909/24155
91% [███████████] [01:21<00:08, 268.91it/s]	21939/24155
91% [███████████] [01:21<00:07, 277.06it/s]	21970/24155
91% [███████████] [01:21<00:07, 277.62it/s]	21999/24155
91% [███████████] [01:21<00:07, 275.05it/s]	22027/24155
91% [███████████] [01:21<00:07, 281.62it/s]	22058/24155
91% [███████████] [01:21<00:07, 285.88it/s]	22090/24155
92% [███████████] [01:21<00:07, 280.46it/s]	22119/24155
92% [███████████] [01:22<00:07, 267.59it/s]	22148/24155
92% [███████████] [01:22<00:07, 273.56it/s]	22178/24155
92% [███████████] [01:22<00:07, 269.13it/s]	22206/24155
92% [███████████] [01:22<00:07, 249.07it/s]	22234/24155
92% [███████████] [01:22<00:07, 241.05it/s]	22260/24155
92% [███████████] [01:22<00:07, 238.11it/s]	22285/24155
92% [███████████] [01:22<00:07, 233.46it/s]	22310/24155
92% [███████████] [01:22<00:07, 235.51it/s]	22335/24155
93% [███████████] [01:22<00:07, 247.05it/s]	22364/24155
93% [███████████] [01:23<00:06, 260.44it/s]	22395/24155

[01:23<00:06, 260.21it/s]		22422/24155
[01:23<00:06, 275.08it/s]		22455/24155
[01:23<00:06, 273.32it/s]		22483/24155
[01:23<00:05, 277.31it/s]		22514/24155
[01:23<00:05, 271.66it/s]		22542/24155
[01:23<00:05, 276.51it/s]		22572/24155
[01:23<00:05, 271.12it/s]		22600/24155
[01:23<00:05, 283.52it/s]		22633/24155
[01:23<00:05, 287.76it/s]		22664/24155
[01:24<00:05, 275.31it/s]		22693/24155
[01:24<00:05, 264.19it/s]		22721/24155
[01:24<00:05, 262.82it/s]		22748/24155
[01:24<00:05, 267.45it/s]		22777/24155
[01:24<00:04, 273.45it/s]		22807/24155
[01:24<00:04, 265.98it/s]		22835/24155
[01:24<00:04, 274.88it/s]		22866/24155
[01:24<00:04, 286.39it/s]		22899/24155
[01:24<00:04, 277.58it/s]		22928/24155
[01:25<00:04, 265.65it/s]		22956/24155
[01:25<00:04, 272.13it/s]		22986/24155
[01:25<00:04, 276.85it/s]		23016/24155
[01:25<00:04, 274.54it/s]		23044/24155
[01:25<00:03, 278.59it/s]		23074/24155
[01:25<00:03, 272.52it/s]		23102/24155
[01:25<00:03, 271.33it/s]		23131/24155
[01:25<00:03, 270.11it/s]		23159/24155

90% [██████████	[01:25<00:03, 258.71it/s]	23159/24155
96% [███████████	[01:25<00:03, 259.00it/s]	23186/24155
96% [███████████	[01:26<00:03, 247.79it/s]	23213/24155
96% [███████████	[01:26<00:03, 265.24it/s]	23246/24155 [01:
96% [███████████	[01:26<00:03, 273.45it/s]	23279/24155 [01:
97% [███████████	[01:26<00:02, 282.37it/s]	23312/24155 [01:
97% [███████████	[01:26<00:02, 289.54it/s]	23344/24155 [01:
97% [███████████	[01:26<00:02, 298.44it/s]	23380/24155 [01:
97% [███████████	[01:26<00:02, 294.95it/s]	23411/24155 [01:
97% [███████████	[01:26<00:02, 289.60it/s]	23441/24155 [01:
97% [███████████	[01:26<00:02, 276.49it/s]	23471/24155 [01:
97% [███████████	[01:27<00:02, 279.99it/s]	23501/24155 [01:
97% [███████████	[01:27<00:02, 270.29it/s]	23530/24155 [01:
98% [███████████	[01:27<00:02, 275.52it/s]	23560/24155 [01:
98% [███████████	[01:27<00:02, 281.92it/s]	23591/24155 [01:
98% [███████████	[01:27<00:01, 268.54it/s]	23620/24155 [01:
98% [███████████	[01:27<00:01, 254.14it/s]	23648/24155 [01:
98% [███████████	[01:27<00:01, 247.13it/s]	23674/24155 [01:
98% [███████████	[01:27<00:01, 268.55it/s]	23709/24155 [01:
98% [███████████	[01:27<00:01, 268.77it/s]	23737/24155 [01:
98% [███████████	[01:28<00:01, 278.37it/s]	23771/24155 [01:
99% [███████████	[01:28<00:01, 272.26it/s]	23800/24155 [01:
99% [███████████	[01:28<00:01, 268.24it/s]	23828/24155 [01:
99% [███████████	[01:28<00:01, 256.74it/s]	23856/24155 [01:
99% [███████████	[01:28<00:01, 267.91it/s]	23887/24155 [01:
99% [███████████	[01:28<00:00, 256.53it/s]	23915/24155 [01:

```
99%|██████████| 23942/24155 [01:  
28<00:00, 254.55it/s]  
  
99%|██████████| 23973/24155 [01:  
28<00:00, 266.23it/s]  
  
99%|██████████| 24005/24155 [01:  
28<00:00, 277.47it/s]  
  
100%|██████████| 24035/24155 [01:  
29<00:00, 280.70it/s]  
  
100%|██████████| 24064/24155 [01:  
29<00:00, 273.81it/s]  
  
100%|██████████| 24092/24155 [01:  
29<00:00, 254.61it/s]  
  
100%|██████████| 24118/24155 [01:  
29<00:00, 253.23it/s]  
  
100%|██████████| 24146/24155 [01:  
29<00:00, 257.84it/s]  
  
100%|██████████| 24155/24155 [01:  
29<00:00, 269.85it/s]  
█
```

24155

300

In [114]:

```
tfidf_weighted_w2v_on_essay_matrix_Xcv = np.vstack(tfidf_weighted_w2v_vectors_eassays_Xcv)  
print(tfidf_weighted_w2v_on_essay_matrix_Xcv.shape)
```

(24155, 300)

8.6.1. TFIDF Weighted Word2Vec encoding of preprocessed_titles on Train Data

In [115]:

```
# S = ["abc def pqr", "def def def abc", "pqr pqr def"]  
tfidf_model = TfidfVectorizer()  
tfidf_model.fit(X_train['preprocessed_titles'].values)  
# we are converting a dictionary with word as a key, and the idf as a value  
dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))  
tfidf_words = set(tfidf_model.get_feature_names())
```

In [116]:

```
# average Word2Vec  
# compute average word2vec for each review.  
tfidf_weighted_w2v_vectors_title_Xtrain = [] # the avg-w2v for each sentence/review is stored in  
this list  
for t in tqdm(X_train['preprocessed_titles'].values): # for each review/sentence  
    vector = np.zeros(300) # as word vectors are of zero length  
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review  
    for word in t.split(): # for each word in a review/sentence  
        if (word in glove_words) and (word in tfidf_words):  
            vec = model[word] # getting the vector for each word  
            # here we are multiplying idf value(dictionary[word]) and the tf  
            value((sentence.count(word)/len(sentence.split())))  
            tf_idf = dictionary[word]*(t.count(word)/len(t.split())) # getting the tfidf value for  
each word  
            vector += (vec * tf_idf) # calculating tfidf weighted w2v  
            tf_idf_weight += tf_idf  
    if tf_idf_weight != 0:  
        vector /= tf_idf_weight  
    tfidf_weighted_w2v_vectors_title_Xtrain.append(vector)
```

```
print(len(tfidf_weighted_w2v_vectors_title_Xtrain))
print(len(tfidf_weighted_w2v_vectors_title_Xtrain[0]))
```

0%| [00:00<?, ?it/s] | 0/49

3%| [03, 14743.51it/s] | 1533/49041 [00:00<

6%| [00:00<00:03, 13958.95it/s] | 2824/49041

8%| [00:00<00:03, 12433.02it/s] | 3854/49041

11%| [00:00<00:03, 13062.37it/s] | 5394/49041

14%| [00:00<00:03, 13191.62it/s] | 6798/49041

18%| [00:00<00:02, 14327.07it/s] | 8662/49041

20%| [03, 12927.06it/s] | 9967/49041 [00:00<

23%| [02, 12903.64it/s] | 11303/49041 [00:00<

26%| [02, 13115.51it/s] | 12721/49041 [00:00<

29%| [02, 12278.06it/s] | 14003/49041 [00:01<

32%| [00:01<00:02, 13382.43it/s] | 15764/49041

35%| [00:01<00:02, 13778.63it/s] | 17303/49041

38%| [00:01<00:02, 13208.80it/s] | 18700/49041

41%| [00:01<00:02, 13812.43it/s] | 20308/49041

44%| [0:01, 13709.73it/s] | 21709/49041 [00:01<

47%| [0:01, 13987.02it/s] | 23235/49041 [00:01<

51%| [0:01, 14790.69it/s] | 25011/49041 [00:01<

54%| [0:01, 14509.48it/s] | 26511/49041 [00:01<

57%| [0:01, 14556.34it/s] | 28036/49041 [00:02<

60%| [00:02<00:01, 14775.02it/s] | 29628/49041

63%| [00:02<00:01, 14626.77it/s] | 31114/49041

67%| [00:02<00:01, 14589.06it/s] | 32622/49041

70%| [00:02<00:01, 14121.40it/s] | 34093/49041

72%|███████████| 35512/49041 [00:02<
00:00, 13975.85it/s]

75%|███████████| 36990/49041 [00:02<
00:00, 14046.93it/s]

78%|███████████| 38399/49041 [00:02<
00:00, 13732.09it/s]

81%|███████████| 39777/49041 [00:02<
00:00, 12971.13it/s]

84%|███████████| 41104/49041 [00:03<
00:00, 12195.73it/s]

86%|███████████| 42342/49041 [00:03<
00:00, 11967.82it/s]

89%|███████████| 43635/49041 [00:03<
00:00, 12104.58it/s]

91%|███████████| 44856/49041
[00:03<00:00, 11853.57it/s]

94%|███████████| 46264/49041
[00:03<00:00, 12314.63it/s]

98%|███████████| 47838/49041
[00:03<00:00, 13044.63it/s]

100%|███████████| 49041/49041
[00:03<00:00, 13520.80it/s]

49041
300

In [117]:

```
tfidf_weighted_w2v_on_title_matrix_Xtrain = np.vstack(tfidf_weighted_w2v_vectors_title_Xtrain)
print(tfidf_weighted_w2v_on_title_matrix_Xtrain.shape)
```

(49041, 300)

8.6.2. TFIDF Weighted Word2Vec encoding of preprocessed_titles on Test Data

In [118]:

```
# # S = ["abc def pqr", "def def def abc", "pqr pqr def"]
# tfidf_model = TfidfVectorizer()
# tfidf_model.fit(X_test['preprocessed_titles'].values)
# # we are converting a dictionary with word as a key, and the idf as a value
# dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
# tfidf_words = set(tfidf_model.get_feature_names())
```

In [119]:

```
# average Word2Vec
# compute average word2vec for each review.
tfidf_weighted_w2v_vectors_title_Xtest = [] # the avg-w2v for each sentence/review is stored in this list
for t in tqdm(X_test['preprocessed_titles'].values): # for each review/sentence
    vector = np.zeros(300) # as word vectors are of zero length
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review
    for word in t.split(): # for each word in a review/sentence
        if (word in glove_words) and (word in tfidf_words):
            vec = model[word] # getting the vector for each word
            # here we are multiplying idf value(dictionary[word]) and the tf
            value((sentence.count(word)/len(sentence.split())))
    tfidf_weighted_w2v_vectors_title_Xtest.append(vec)
```

```

        tf_idf = dictionary[word]*t.count(word)/len(t.split()) # getting the tfidf value for
each word
        vector += (vec * tf_idf) # calculating tfidf weighted w2v
        tf_idf_weight += tf_idf
    if tf_idf_weight != 0:
        vector /= tf_idf_weight
    tfidf_weighted_w2v_vectors_title_Xtest.append(vector)

print(len(tfidf_weighted_w2v_vectors_title_Xtest))
print(len(tfidf_weighted_w2v_vectors_title_Xtest[0]))

```

0% | 0/36 [00:00<?, ?it/s]

3% | 1098/36052 [00:00<03, 10559.93it/s]

6% | 2274/36052 [00:00<00:03, 10774.58it/s]

13% | 4587/36052 [00:00<00:02, 12746.41it/s]

18% | 6520/36052 [00:00<02, 14073.62it/s]

22% | 7820/36052 [00:00<02, 13395.20it/s]

27% | 9745/36052 [00:00<01, 14606.74it/s]

32% | 11490/36052 [00:00<00:01, 15198.06it/s]

36% | 13006/36052 [00:00<00:01, 13558.25it/s]

40% | 14410/36052 [00:00<00:01, 13541.75it/s]

45% | 16324/36052 [00:01<0:01, 14708.23it/s]

49% | 17839/36052 [00:01<0:01, 14333.86it/s]

54% | 19306/36052 [00:01<0:01, 13940.55it/s]

58% | 20897/36052 [00:01<00:01, 14322.79it/s]

62% | 22350/36052 [00:01<00:00, 14051.48it/s]

66% | 23771/36052 [00:01<00:00, 13772.02it/s]

70% | 25161/36052 [00:01<00:00, 13489.58it/s]

75% | 26964/36052 [00:00<00:00, 14452.50it/s]

79% | 28436/36052 [00:01<00:00, 14196.47it/s]

83% | 29875/36052 [00:02<00:00, 13764.39it/s]

87% | 31384/36052 [00:02<00:00, 13979.77it/s]

91% | 32795/36052 [00:02<00:00, 13084.91it/s]

```
[00:02<00:00, 10001.51it/s] 95%|██████████| 34124/36052  
[00:02<00:00, 12410.26it/s]  
  
98%|██████████| 35387/36052  
[00:02<00:00, 12187.47it/s]  
  
100%|██████████| 36052/36052  
[00:02<00:00, 13977.19it/s]
```

36052
300

In [121]:

```
tfidf_weighted_w2v_on_title_matrix_Xtest = np.vstack(tfidf_weighted_w2v_vectors_title_Xtest)  
print(tfidf_weighted_w2v_on_title_matrix_Xtest.shape)
```

(36052, 300)

8.6.3. TFIDF Weighted Word2Vec encoding of preprocessed_titles on CV Data

In [122]:

```
# # S = ["abc def pqr", "def def def abc", "pqr pqr def"]  
# tfidf_model = TfidfVectorizer()  
# tfidf_model.fit(X_cv['preprocessed_titles'].values)  
# # we are converting a dictionary with word as a key, and the idf as a value  
# dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))  
# tfidf_words = set(tfidf_model.get_feature_names())
```

In [123]:

```
# average Word2Vec  
# compute average word2vec for each review.  
tfidf_weighted_w2v_vectors_title_Xcv = [] # the avg-w2v for each sentence/review is stored in this list  
for t in tqdm(X_cv['preprocessed_titles'].values): # for each review/sentence  
    vector = np.zeros(300) # as word vectors are of zero length  
    tf_idf_weight = 0; # num of words with a valid vector in the sentence/review  
    for word in t.split(): # for each word in a review/sentence  
        if (word in glove_words) and (word in tfidf_words):  
            vec = model[word] # getting the vector for each word  
            # here we are multiplying idf value(dictionary[word]) and the tf  
            value((sentence.count(word)/len(sentence.split())))  
            tf_idf = dictionary[word]*(t.count(word)/len(t.split())) # getting the tfidf value for  
            each word  
            vector += (vec * tf_idf) # calculating tfidf weighted w2v  
            tf_idf_weight += tf_idf  
    if tf_idf_weight != 0:  
        vector /= tf_idf_weight  
    tfidf_weighted_w2v_vectors_title_Xcv.append(vector)  
  
print(len(tfidf_weighted_w2v_vectors_title_Xcv))  
print(len(tfidf_weighted_w2v_vectors_title_Xcv[0]))
```

```
0%| 0/24  
[00:00<?, ?it/s]  
  
6%|████| 1526/24155  
[00:00<00:01, 14676.70it/s]  
  
12%|██████| 2934/24155  
[00:00<00:01, 14316.69it/s]  
  
21%|███████| 4952/24155 [00:00<  
:01, 15539.95it/s]
```



In [125]:

```
tfidf_weighted_w2v_on_title_matrix_Xcv = np.vstack(tfidf_weighted_w2v_vectors_title_Xcv)
print(tfidf_weighted_w2v_on_title_matrix_Xcv.shape)
```

(24155, 300)

9. Encoding of Numerical Data

9.1.1. Encoding of price on Train Data

In [130]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html
from sklearn.preprocessing import MinMaxScaler

scalar = MinMaxScaler()

price_standardized_Xtrain = scalar.fit_transform(X_train['price'].values.reshape(-1, 1))
price_standardized_Xtest = scalar.transform(X_test['price'].values.reshape(-1, 1))
price_standardized_Xcv = scalar.transform(X_cv['price'].values.reshape(-1, 1))
```

In [131]:

```
price_standardized_Xtrain
```

```
Out[131]:  
array([[0.01045574],  
       [0.        ],  
       [0.09994959],  
       ...,  
       [0.01355325],  
       [0.01542656],  
       [0.03463175]])
```

In [133]:

```
print(price_standardized_Xtrain.shape)  
print(price_standardized_Xtest.shape)  
print(price_standardized_Xcv.shape)
```

```
(49041, 1)  
(36052, 1)  
(24155, 1)
```

9.2.1. Encoding of quantity on Train,Test and CV data

In [132]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s  
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html  
from sklearn.preprocessing import MinMaxScaler  
  
scalar = MinMaxScaler()  
  
quantity_standardized_Xtrain = scalar.fit_transform(X_train['quantity'].values.reshape(-1, 1))  
quantity_standardized_Xtest = scalar.transform(X_test['quantity'].values.reshape(-1, 1))  
quantity_standardized_Xcv = scalar.transform(X_cv['quantity'].values.reshape(-1, 1))
```

In [134]:

```
quantity_standardized_Xtrain
```

Out[134]:

```
array([[0.00375469],  
       [0.40550688],  
       [0.        ],  
       ...,  
       [0.0175219 ],  
       [0.01001252],  
       [0.00876095]])
```

In [135]:

```
print(quantity_standardized_Xtrain.shape)  
print(quantity_standardized_Xtest.shape)  
print(quantity_standardized_Xcv.shape)
```

```
(49041, 1)  
(36052, 1)  
(24155, 1)
```

9.3.1. Encoding of teacher_number_of_previously_posted_projects on Train,Test and CV data

In [136]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
```

```
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html
from sklearn.preprocessing import MinMaxScaler

scalar = MinMaxScaler()

# Now standardize the data with above mean and variance.
teacher_number_of_previously_posted_projects_standardized_Xtrain = scalar.fit_transform(X_train['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1))
teacher_number_of_previously_posted_projects_standardized_Xtest =
scalar.transform(X_test['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1))
teacher_number_of_previously_posted_projects_standardized_Xcv =
scalar.transform(X_cv['teacher_number_of_previously_posted_projects'].values.reshape(-1, 1))
```

In [137]:

```
teacher_number_of_previously_posted_projects_standardized_Xtrain
```

Out[137]:

```
array([[0.00457666],
       [0.00457666],
       [0.01372998],
       ...,
       [0.00457666],
       [0.        ],
       [0.00228833]])
```

In [138]:

```
print(teacher_number_of_previously_posted_projects_standardized_Xtrain.shape)
print(teacher_number_of_previously_posted_projects_standardized_Xtest.shape)
print(teacher_number_of_previously_posted_projects_standardized_Xcv.shape)
```

```
(49041, 1)
(36052, 1)
(24155, 1)
```

9.4.1. Encoding of pos on Train,Test and CV data

In [139]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html
from sklearn.preprocessing import MinMaxScaler

scalar = MinMaxScaler()

essay_pos_standardized_Xtrain = scalar.fit_transform(X_train['pos'].values.reshape(-1, 1))
essay_pos_standardized_Xtest = scalar.transform(X_test['pos'].values.reshape(-1, 1))
essay_pos_standardized_Xcv = scalar.transform(X_cv['pos'].values.reshape(-1, 1))
```

In [140]:

```
essay_pos_standardized_Xtrain
```

Out[140]:

```
array([[0.38539898],
       [0.5942275 ],
       [0.39219015],
       ...,
       [0.44991511],
       [0.32767402],
       [0.38879457]])
```

In [141]:

```
print(essay_pos_standardized_Xtrain.shape)
print(essay_pos_standardized_Xtest.shape)
```

```
print(essay_pos_standardized_Xcv.shape)
```

```
(49041, 1)  
(36052, 1)  
(24155, 1)
```

9.5.1. Encoding of neg on Train,Test and CV data

In [142]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s  
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html  
from sklearn.preprocessing import MinMaxScaler  
  
scalar = MinMaxScaler()  
  
essay_neg_standardized_Xtrain = scalar.fit_transform(X_train['neg'].values.reshape(-1, 1))  
essay_neg_standardized_Xtest = scalar.transform(X_test['neg'].values.reshape(-1, 1))  
essay_neg_standardized_Xcv = scalar.transform(X_cv['neg'].values.reshape(-1, 1))
```

In [143]:

```
essay_neg_standardized_Xtrain
```

Out[143]:

```
array([[0.05722071],  
       [0.17711172],  
       [0.07084469],  
       ...,  
       [0.02452316],  
       [0.19891008],  
       [0.05722071]])
```

In [144]:

```
print(essay_neg_standardized_Xtrain.shape)  
print(essay_neg_standardized_Xtest.shape)  
print(essay_neg_standardized_Xcv.shape)
```

```
(49041, 1)  
(36052, 1)  
(24155, 1)
```

9.6.1. Encoding of neu on Train,Test and CV data

In [145]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s  
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html  
from sklearn.preprocessing import MinMaxScaler  
  
scalar = MinMaxScaler()  
  
essay_neu_standardized_Xtrain = scalar.fit_transform(X_train['neu'].values.reshape(-1, 1))  
essay_neu_standardized_Xtest = scalar.transform(X_test['neu'].values.reshape(-1, 1))  
essay_neu_standardized_Xcv = scalar.transform(X_cv['neu'].values.reshape(-1, 1))
```

In [146]:

```
essay_neu_standardized_Xtrain
```

Out[146]:

```
array([[0.61426256],  
       [0.34359806],  
       ...])
```

```
[0.6012966 ],
...,
[0.57050243],
[0.58508914],
[0.61102107]])
```

In [147]:

```
print(essay_neu_standardized_Xtrain.shape)
print(essay_neu_standardized_Xtest.shape)
print(essay_neu_standardized_Xcv.shape)
```

```
(49041, 1)
(36052, 1)
(24155, 1)
```

9.7.1. Encoding of compound on Train,Test and CV data

In [148]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html
from sklearn.preprocessing import MinMaxScaler

scalar = MinMaxScaler()

essay_compound_standardized_Xtrain = scalar.fit_transform(X_train['compound'].values.reshape(-1, 1))
essay_compound_standardized_Xtest = scalar.transform(X_test['compound'].values.reshape(-1, 1))
essay_compound_standardized_Xcv = scalar.transform(X_cv['compound'].values.reshape(-1, 1))
```

In [149]:

```
essay_compound_standardized_Xtrain
```

Out[149]:

```
array([[0.99618799],
       [0.99824447],
       [0.99538546],
       ...,
       [0.99538546],
       [0.98600592],
       [0.99337914]])
```

In [150]:

```
print(essay_compound_standardized_Xtrain.shape)
print(essay_compound_standardized_Xtest.shape)
print(essay_compound_standardized_Xcv.shape)
```

```
(49041, 1)
(36052, 1)
(24155, 1)
```

9.8.1. Encoding of number_of_words_in_essays on Train,Test and CV data

In [151]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html
from sklearn.preprocessing import MinMaxScaler

scalar = MinMaxScaler()

number_of_words_in_essays_standardized_Xtrain =
```

```
number_of_words_in_essays_standardized_Xtrain =  
scalar.fit_transform(X_train['number_of_words_in_essays'].values.reshape(-1, 1))  
number_of_words_in_essays_standardized_Xtest = scalar.transform(X_test['number_of_words_in_essays'].values.reshape(-1, 1))  
number_of_words_in_essays_standardized_Xcv = scalar.transform(X_cv['number_of_words_in_essays'].values.reshape(-1, 1))
```

In [152]:

```
number_of_words_in_essays_standardized_Xtrain
```

Out[152]:

```
array([[0.49003984],  
[0.32270916],  
[0.35458167],  
...,  
[0.24701195],  
[0.31075697],  
[0.23904382]])
```

In [153]:

```
print(number_of_words_in_essays_standardized_Xtrain.shape)  
print(number_of_words_in_essays_standardized_Xtest.shape)  
print(number_of_words_in_essays_standardized_Xcv.shape)
```

```
(49041, 1)  
(36052, 1)  
(24155, 1)
```

9.9.1. Encoding of number_of_words_in_the_title on Train,Test and CV data

In [154]:

```
# check this one: https://www.youtube.com/watch?v=0HOqOcln3Z4&t=530s  
# https://scikit-learn.org/stable/modules/generated/sklearn.preprocessing.MinMaxScaler.html  
from sklearn.preprocessing import MinMaxScaler  
  
scalar = MinMaxScaler()  
  
number_of_words_in_the_title_standardized_Xtrain =  
scalar.fit_transform(X_train['number_of_words_in_the_title'].values.reshape(-1, 1))  
number_of_words_in_the_title_standardized_Xtest =  
scalar.transform(X_test['number_of_words_in_the_title'].values.reshape(-1, 1))  
number_of_words_in_the_title_standardized_Xcv =  
scalar.transform(X_cv['number_of_words_in_the_title'].values.reshape(-1, 1))
```

In [155]:

```
number_of_words_in_the_title_standardized_Xtrain
```

Out[155]:

```
array([[0.45454545],  
[0.72727273],  
[0.63636364],  
...,  
[0.36363636],  
[0.27272727],  
[0.27272727]])
```

In [156]:

```
print(number_of_words_in_the_title_standardized_Xtrain.shape)  
print(number_of_words_in_the_title_standardized_Xtest.shape)  
print(number_of_words_in_the_title_standardized_Xcv.shape)
```

```
(49041, 1)
(36052, 1)
(24155, 1)
```

In []:

In []:

In []:

10. Printing Dimensions of all Preprocessed Data

In [157]:

```
print(categories_one_hot_Xtrain.shape)
print(categories_one_hot_Xtest.shape)
print(categories_one_hot_Xcv.shape)
print(sub_categories_one_hot_Xtrain.shape)
print(sub_categories_one_hot_Xtest.shape)
print(sub_categories_one_hot_Xcv.shape)
print(school_state_one_hot_Xtrain.shape)
print(school_state_one_hot_Xtest.shape)
print(school_state_one_hot_Xcv.shape)
print(teacher_prefix_one_hot_Xtrain.shape)
print(teacher_prefix_one_hot_Xtest.shape)
print(teacher_prefix_one_hot_Xcv.shape)
print(grade_one_hot_Xtrain.shape)
print(grade_one_hot_Xtest.shape)
print(grade_one_hot_Xcv.shape)
print(text_bow_Xtrain.shape)
print(text_bow_Xtest.shape)
print(text_bow_Xcv.shape)
print(title_bow_Xtrain.shape)
print(title_bow_Xtest.shape)
print(title_bow_Xcv.shape)
print(text_tfidf_Xtrain.shape)
print(text_tfidf_Xtest.shape)
print(text_tfidf_Xcv.shape)
print(title_tfidf_Xtrain.shape)
print(title_tfidf_Xtest.shape)
print(title_tfidf_Xcv.shape)
print(average_w2v_on_essay_Xtrain.shape)
print(average_w2v_on_essay_Xtest.shape)
print(average_w2v_on_essay_Xcv.shape)
print(average_w2v_on_titles_Xtrain.shape)
print(average_w2v_on_titles_Xtest.shape)
print(average_w2v_on_titles_Xcv.shape)
print(tfidf_weighted_w2v_on_essay_matrix_Xtrain.shape)
print(tfidf_weighted_w2v_on_essay_matrix_Xtest.shape)
print(tfidf_weighted_w2v_on_essay_matrix_Xcv.shape)
print(tfidf_weighted_w2v_on_title_matrix_Xtrain.shape)
print(tfidf_weighted_w2v_on_title_matrix_Xtest.shape)
print(tfidf_weighted_w2v_on_title_matrix_Xcv.shape)
print(price_standardized_Xtrain.shape)
print(price_standardized_Xtest.shape)
print(price_standardized_Xcv.shape)
print(quantity_standardized_Xtrain.shape)
print(quantity_standardized_Xtest.shape)
print(quantity_standardized_Xcv.shape)
print(teacher_number_of_previously_posted_projects_standardized_Xtrain.shape)
print(teacher_number_of_previously_posted_projects_standardized_Xtest.shape)
print(teacher_number_of_previously_posted_projects_standardized_Xcv .shape)
print(essay_pos_standardized_Xtrain.shape)
print(essay_pos_standardized_Xtest.shape)
print(essay_pos_standardized_Xcv .shape)
```

```
print(essay_pos_standardized_Xcv.shape)
print(essay_neg_standardized_Xtrain.shape)
print(essay_neg_standardized_Xtest.shape)
print(essay_neu_standardized_Xcv.shape)
print(essay_neu_standardized_Xtrain.shape)
print(essay_neu_standardized_Xtest.shape)
print(essay_neu_standardized_Xcv.shape)
print(essay_compound_standardized_Xtrain.shape)
print(essay_compound_standardized_Xtest.shape)
print(essay_compound_standardized_Xcv.shape)
print(number_of_words_in_essays_standardized_Xtrain.shape)
print(number_of_words_in_essays_standardized_Xtest.shape)
print(number_of_words_in_essays_standardized_Xcv.shape)
print(number_of_words_in_the_title_standardized_Xtrain.shape)
print(number_of_words_in_the_title_standardized_Xtest.shape)
print(number_of_words_in_the_title_standardized_Xcv.shape)
```

```
(49041, 9)
(36052, 9)
(24155, 9)
(49041, 30)
(36052, 30)
(24155, 30)
(49041, 51)
(36052, 51)
(24155, 51)
(49041, 5)
(36052, 5)
(24155, 5)
(49041, 4)
(36052, 4)
(24155, 4)
(49041, 12030)
(36052, 12030)
(24155, 12030)
(49041, 2016)
(36052, 2016)
(24155, 2016)
(49041, 12030)
(36052, 12030)
(24155, 12030)
(49041, 2016)
(36052, 2016)
(24155, 2016)
(49041, 300)
(36052, 300)
(24155, 300)
(49041, 300)
(36052, 300)
(24155, 300)
(49041, 300)
(36052, 300)
(24155, 300)
(49041, 300)
(36052, 300)
(24155, 300)
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
```

```
(49041, 1)
(36052, 1)
(24155, 1)
(49041, 1)
(36052, 1)
(24155, 1)
```

11. Creating Different Sets of Data for Training Model

Set 1: categorical, numerical features + project_title(BOW) + preprocessed_eassay (BOW)

In [158]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matirx :
Xtrain1 =
hstack((categories_one_hot_Xtrain,sub_categories_one_hot_Xtrain,school_state_one_hot_Xtrain,teacher_prefix_one_hot_Xtrain,grade_one_hot_Xtrain,price_standardized_Xtrain,quantity_standardized_Xtrain,teacher_number_of_previously_posted_projects_standardized_Xtrain,text_bow_Xtrain,title_bow_Xtrain)).tocsr()
Xtest1 = hstack((categories_one_hot_Xtest,sub_categories_one_hot_Xtest,school_state_one_hot_Xtest,teacher_prefix_one_hot_Xtest,grade_one_hot_Xtest,price_standardized_Xtest,quantity_standardized_Xtest,teacher_number_of_previously_posted_projects_standardized_Xtest,text_bow_Xtest,title_bow_Xtest)).tocsr()
Xcv1 =
hstack((categories_one_hot_Xcv,sub_categories_one_hot_Xcv,school_state_one_hot_Xcv,teacher_prefix_one_hot_Xcv,grade_one_hot_Xcv,price_standardized_Xcv,quantity_standardized_Xcv,teacher_number_of_previously_posted_projects_standardized_Xcv,text_bow_Xcv,title_bow_Xcv)).tocsr()

print(Xtrain1.shape,y_train.shape)
print(Xtest1.shape,y_test.shape)
print(Xcv1.shape,y_cv.shape)

```

(49041, 14148) (49041,)
(36052, 14148) (36052,)
(24155, 14148) (24155,)

Set 2: categorical, numerical features + project_title(TFIDF)+ preprocessed_eassay (TFIDF)

In [159]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matirx :
Xtrain2 =
hstack((categories_one_hot_Xtrain,sub_categories_one_hot_Xtrain,school_state_one_hot_Xtrain,teacher_prefix_one_hot_Xtrain,grade_one_hot_Xtrain,price_standardized_Xtrain,quantity_standardized_Xtrain,teacher_number_of_previously_posted_projects_standardized_Xtrain,text_tfidf_Xtrain,title_tfidf_Xtrain)).tocsr()
Xtest2 = hstack((categories_one_hot_Xtest,sub_categories_one_hot_Xtest,school_state_one_hot_Xtest,teacher_prefix_one_hot_Xtest,grade_one_hot_Xtest,price_standardized_Xtest,quantity_standardized_Xtest,teacher_number_of_previously_posted_projects_standardized_Xtest,text_tfidf_Xtest,title_tfidf_Xtest)).tocsr()
Xcv2 =
hstack((categories_one_hot_Xcv,sub_categories_one_hot_Xcv,school_state_one_hot_Xcv,teacher_prefix_one_hot_Xcv,grade_one_hot_Xcv,price_standardized_Xcv,quantity_standardized_Xcv,teacher_number_of_previously_posted_projects_standardized_Xcv,text_tfidf_Xcv,title_tfidf_Xcv)).tocsr()

print(Xtrain2.shape,y_train.shape)
print(Xtest2.shape,y_test.shape)
print(Xcv2.shape,y_cv.shape)

```

(49041, 14148) (49041,)
(36052, 14148) (36052,)
(24155, 14148) (24155,)

Set 3: categorical, numerical features + project_title(AVG W2V)+ preprocessed_eassay (AVG W2V)

In [160]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matirx :
Xtrain3 =
hstack((categories_one_hot_Xtrain,sub_categories_one_hot_Xtrain,school_state_one_hot_Xtrain,teacher_prefix_one_hot_Xtrain,grade_one_hot_Xtrain,price_standardized_Xtrain,quantity_standardized_Xtrain,teacher_number_of_previously_posted_projects_standardized_Xtrain,average_w2v_on_essay_Xtrain,average_w2v_on_titles_Xtrain)).tocsr()
Xtest3 = hstack((categories_one_hot_Xtest,sub_categories_one_hot_Xtest,school_state_one_hot_Xtest,teacher_prefix_one_hot_Xtest,grade_one_hot_Xtest,price_standardized_Xtest,quantity_standardized_Xtest,teacher_number_of_previously_posted_projects_standardized_Xtest,average_w2v_on_essay_Xtest,average_w2v_on_titles_Xtest)).tocsr()
Xcv3 =
hstack((categories_one_hot_Xcv,sub_categories_one_hot_Xcv,school_state_one_hot_Xcv,teacher_prefix_one_hot_Xcv,grade_one_hot_Xcv,price_standardized_Xcv,quantity_standardized_Xcv,teacher_number_of_previously_posted_projects_standardized_Xcv,average_w2v_on_essay_Xcv,average_w2v_on_titles_Xcv)).tocsr()

print(Xtrain3.shape,y_train.shape)
print(Xtest3.shape,y_test.shape)
print(Xcv3.shape,y_cv.shape)
```

(49041, 702) (49041,)
(36052, 702) (36052,)
(24155, 702) (24155,)

Set 4: categorical, numerical features + project_title(TFIDF W2V)+ preprocessed_eassay (TFIDF W2V)

In [161]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matirx :
Xtrain4 =
hstack((categories_one_hot_Xtrain,sub_categories_one_hot_Xtrain,school_state_one_hot_Xtrain,teacher_prefix_one_hot_Xtrain,grade_one_hot_Xtrain,price_standardized_Xtrain,quantity_standardized_Xtrain,teacher_number_of_previously_posted_projects_standardized_Xtrain,tfidf_weighted_w2v_on_essay_matrix_Xtrain,tfidf_weighted_w2v_on_title_matrix_Xtrain)).tocsr()
Xtest4 = hstack((categories_one_hot_Xtest,sub_categories_one_hot_Xtest,school_state_one_hot_Xtest,teacher_prefix_one_hot_Xtest,grade_one_hot_Xtest,price_standardized_Xtest,quantity_standardized_Xtest,teacher_number_of_previously_posted_projects_standardized_Xtest,tfidf_weighted_w2v_on_essay_matrix_Xtest,tfidf_weighted_w2v_on_title_matrix_Xtest)).tocsr()
Xcv4 =
hstack((categories_one_hot_Xcv,sub_categories_one_hot_Xcv,school_state_one_hot_Xcv,teacher_prefix_one_hot_Xcv,grade_one_hot_Xcv,price_standardized_Xcv,quantity_standardized_Xcv,teacher_number_of_previously_posted_projects_standardized_Xcv,tfidf_weighted_w2v_on_essay_matrix_Xcv,tfidf_weighted_w2v_on_title_matrix_Xcv)).tocsr()

print(Xtrain4.shape,y_train.shape)
print(Xtest4.shape,y_test.shape)
print(Xcv4.shape,y_cv.shape)
```

(49041, 702) (49041,)
(36052, 702) (36052,)
(24155, 702) (24155,)

Set 5: categorical + numerical features + TruncatedSVD on TFIDF vectorizer of essay text

In [163]:

```
from scipy.sparse import hstack
# with the same hstack function we are concatenating a sparse matrix and a dense matirx :
Xtrain5 =
hstack((categories_one_hot_Xtrain,sub_categories_one_hot_Xtrain,school_state_one_hot_Xtrain,teacher_prefix_one_hot_Xtrain,grade_one_hot_Xtrain,price_standardized_Xtrain,quantity_standardized_Xtrain,teacher_number_of_previously_posted_projects_standardized_Xtrain,essay_pos_standardized_Xtrain,essay_neg_standardized_Xtrain,essay_neu_standardized_Xtrain,essay_compound_standardized_Xtrain,number_of_words_in_essays_standardized_Xtrain,number_of_words_in_the_title_standardized_Xtrain,svdtrain)).tocsr()
Xtest5 = hstack(( categories_one_hot_Xtest,sub_categories_one_hot_Xtest,school_state_one_hot_Xtest,teacher_prefix_one_hot_Xtest,grade_one_hot_Xtest,price_standardized_Xtest,quantity_standardized_Xtest,teacher_number_of_previously_posted_projects_standardized_Xtest,essay_pos_standardized_Xtest,essay_neg_standardized_Xtest,essay_neu_standardized_Xtest,essay_compound_standardized_Xtest,number_of_words_in_essays_standardized_Xtest,number_of_words_in_the_title_standardized_Xtest,svdtest)).tocsr()
Xcv5 =
hstack((categories_one_hot_Xcv,sub_categories_one_hot_Xcv,school_state_one_hot_Xcv,teacher_prefix_one_hot_Xcv,grade_one_hot_Xcv,price_standardized_Xcv,quantity_standardized_Xcv,teacher_number_of_previously_posted_projects_standardized_Xcv,essay_pos_standardized_Xcv,essay_neg_standardized_Xcv,essay_neu_standardized_Xcv,essay_compound_standardized_Xcv,number_of_words_in_essays_standardized_Xcv,number_of_words_in_the_title_standardized_Xcv,svdcv)).tocsr()

print(Xtrain5.shape,y_train.shape)
print(Xtest5.shape,y_test.shape)
print(Xcv5.shape,y_cv.shape)
```

```
(49041, 2108) (49041,)
(36052, 2108) (36052,)
(24155, 2108) (24155,)
```

12. Applying Support Vector Machines on different kind of featurization

12.1. Applying Support Vector Machines on BOW, SET 1

In [2]:

```
# def batch_predict(clf, data):
#     # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive class
#     # not the predicted outputs

#     y_data_pred = []
#     tr_loop = data.shape[0] - data.shape[0]%1000
#     # consider you X_tr shape is 49041, then your tr_loop will be 49041 - 49041%1000 = 49000
#     # in this for loop we will iterate until the last 1000 multiplier
#     for i in range(0, tr_loop, 1000):
#         y_data_pred.extend(clf.predict_proba(data[i:i+1000])[:,1])
#         # we will be predicting for the last data points
#     if data.shape[0]%1000 !=0:
#         y_data_pred.extend(clf.predict_proba(data[tr_loop:])[:,1])

#     return y_data_pred
```

12.1.1. Finding Hyperparameter 'alpha' using 'l2' regularization

As SVM's Do not predict probability values so we use calibratedClassifier() for predicting probability values

In []:

```
# import matplotlib.pyplot as plt
# from sklearn.calibration import CalibratedClassifierCV
# from sklearn.linear_model import SGDClassifier
# from sklearn.metrics import roc_auc_score
```

```

# train_auc = []
# cv_auc = []
# alpha = [10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]
# for i in tqdm(alpha):
#     SVM = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = i , max_iter = 1000 )
#     calibrated_svm = CalibratedClassifierCV(SVM, cv=5, method='sigmoid')
#     calibrated_svm.fit(Xtrain1, y_train)

#     y_train_pred = batch_predict(calibrated_svm, Xtrain1)
#     y_cv_pred = batch_predict(calibrated_svm, Xcv1)

#     # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive class
#     # not the predicted outputs
#     train_auc.append(roc_auc_score(y_train,y_train_pred))[:,1]
#     cv_auc.append(roc_auc_score(y_cv, y_cv_pred))[:,1]

# plt.figure(figsize=(20,10))
# plt.plot(alpha, train_auc, label='Train AUC')
# plt.plot(alpha, cv_auc, label='CV AUC')

# plt.scatter(alpha, train_auc, label='Train AUC points')
# plt.scatter(alpha, cv_auc, label='CV AUC points')

# plt.legend()
# plt.xlabel("alpha: hyperparameter")
# plt.ylabel("AUC")
# plt.title("ERROR PLOTS")
# plt.grid(True)
# plt.show()

```

12.1.3. Finding Hyperparameter 'alpha' using 'l2' regularization

In [165]:

```

from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

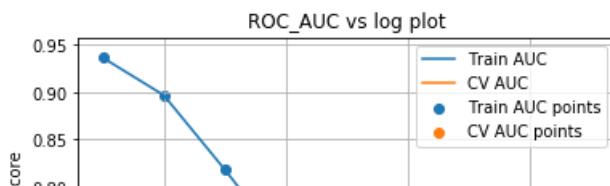
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

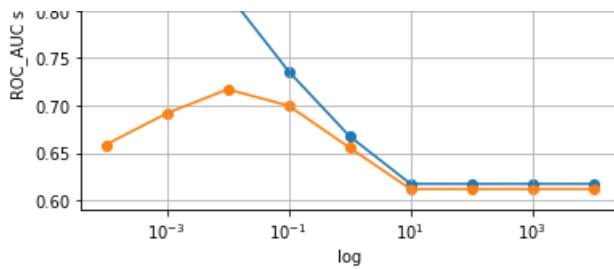
parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM = SGDClassifier(loss = 'hinge', penalty = 'l2', class_weight = 'balanced')
clf = GridSearchCV(SVM, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain1, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()

```





In [166]:

```
cv_auc
```

Out[166]:

```
array([0.65844201, 0.69172293, 0.71705073, 0.69977688, 0.65544748,
       0.61167609, 0.61163012, 0.61163785, 0.61163786])
```

In [167]:

```
score_cv = [x for x in cv_auc]
optimal_alpha_cv = alpha[score_cv.index(max(score_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_cv)))
print("Corresponding alpha value of cv is:",optimal_alpha_cv, '\n')
best_alpha_bow = optimal_alpha_cv
print(best_alpha_bow)
```

```
Maximum AUC score of cv is: 0.7170507253164411
Corresponding alpha value of cv is: 0.01
```

```
0.01
```

12.1.4. Finding Hyperparameter 'alpha' using 'l1' regularization

In [168]:

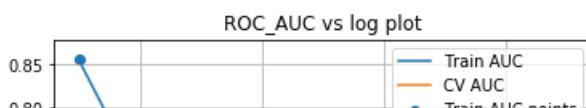
```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

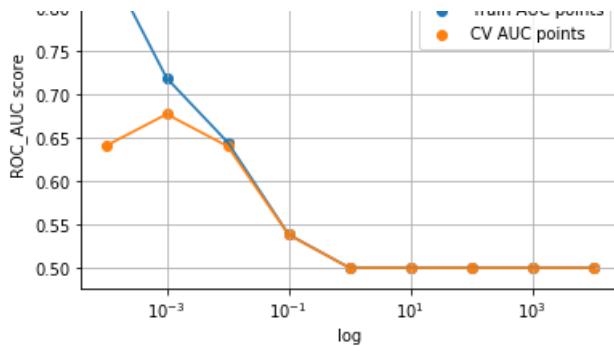
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM = SGDClassifier(loss = 'hinge', penalty = 'l1', class_weight = 'balanced')
clf = GridSearchCV(SVM, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain1, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()
```





12.1.5. Testing the performance of the model on test data, plotting ROC Curves

In [168]:

```
# best_alpha_bow = 0.1 #we choose l2 regularization hyperparameter
```

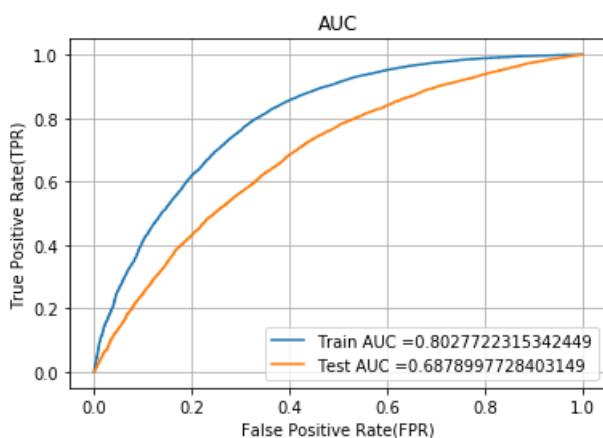
In [169]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
SVM = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_bow)#we choose l2 regularization hyperparameter
SVM.fit(Xtrain1 ,y_train)

y_train_pred_bow = SVM.decision_function(Xtrain1)
y_test_pred_bow = SVM.decision_function(Xtest1)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_bow)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_bow)

plt.plot(train_fpr, train_tpr, label="Train AUC =" +str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC =" +str(auc(test_fpr, test_tpr)))
plt.legend()
plt.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid()
plt.show()
```



In []:

```
# # https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
# from sklearn.metrics import roc_curve, auc
# from sklearn.calibration import CalibratedClassifierCV
```

```

# SVM =SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_tfidf , max_iter = 1000)
# calibrated_svm = CalibratedClassifierCV(SVM, cv=5, method='sigmoid')
# calibrated_svm.fit(Xtrain1, y_train)
# # roc_auc_score(y_true, y_score) the 2nd parameter should be probability estimates of the positive class
# # not the predicted outputs

# y_train_pred_bow = batch_predict(calibrated_svm, Xtrain1)
# y_test_pred_tfidf = batch_predict(calibrated_svm, Xtest1)

# train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_bow)
# test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_bow)

# plt.plot(train_fpr, train_tpr, label="train AUC =" + str(auc(train_fpr, train_tpr)))
# plt.plot(test_fpr, test_tpr, label="test AUC =" + str(auc(test_fpr, test_tpr)))
# plt.legend()
# plt.xlabel("TPR")
# plt.ylabel("FPR")
# plt.title("ERROR PLOTS")
# plt.grid()
# plt.show()

```

12.1.6. Building Confusion Matrix

In [172]:

```

# we are writing our own function for predict, with defined threshold
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshold, fpr, tpr):
    t = threshold[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
    return t

def predict_with_best_t(proba, threshold):
    predictions = []
    for i in proba:
        if i>=threshold:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions

```

In [173]:

```

print("=="*100)
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred_bow, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred_bow, best_t)))

```

```

=====
the maximum value of tpr*(1-fpr) 0.5354053623854204 for threshold 1.048
Train confusion matrix
[[ 5120  2306]
 [ 9299 32316]]
Test confusion matrix
[[ 2961  2498]
 [ 7878 22715]]

```

In [174]:

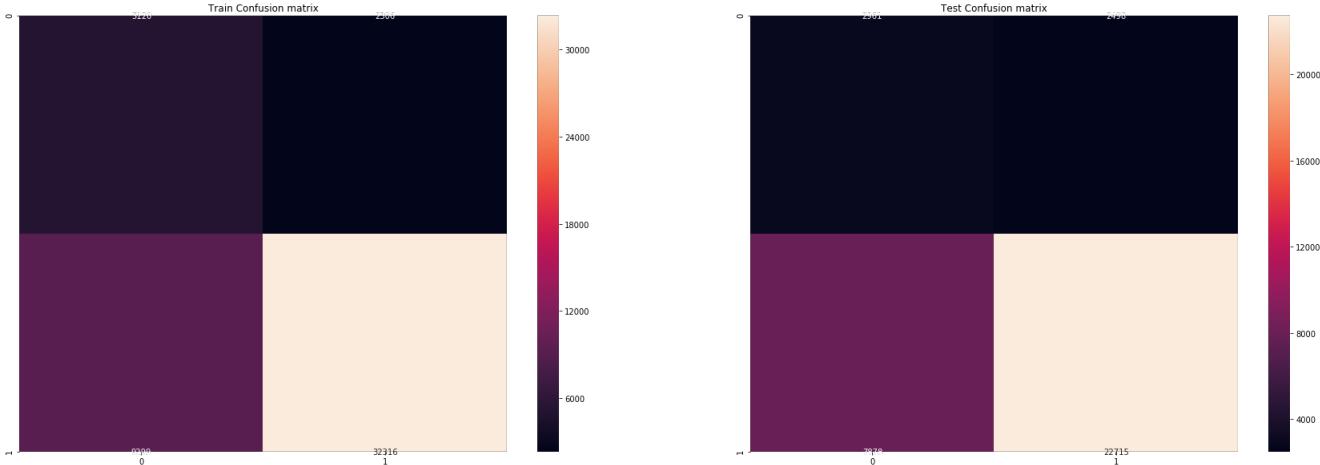
```

confusion_matrix_train_bow = pd.DataFrame(confusion_matrix(y_train,
predict_with_best_t(y_train_pred_bow, best_t)))
confusion_matrix_test_bow = pd.DataFrame(confusion_matrix(y_test,
predict_with_best_t(y_test_pred_bow, best_t)))

```

In [175]:

```
import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(30,10))
# sns.set(font_scale = 1.4)
sns.heatmap(confusion_matrix_train_bow, annot = True ,ax = axes[0],fmt='g')
sns.heatmap(confusion_matrix_test_bow,annot = True , ax = axes[1],fmt = 'g')
axes[0].set_title('Train Confusion matrix')
axes[1].set_title('Test Confusion matrix')
plt.show()
```



12.2. Applying Support Vector Machines on TFIDF, SET 2

12.2.1. Finding The Best Hyperparameter "alpha" with 'l2' regularizer

In [176]:

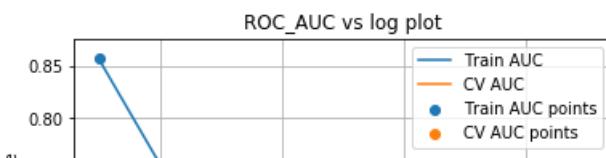
```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

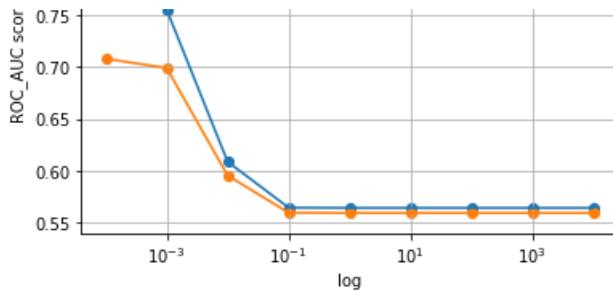
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_tfidf = SGDClassifier(loss = 'hinge', penalty = 'l2', class_weight = 'balanced')
clf = GridSearchCV(SVM_tfidf, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain2, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()
```





In [177]:

```
cv_auc
```

Out[177]:

```
array([0.70808345, 0.69912552, 0.59534224, 0.55942737, 0.55924127,
       0.55924046, 0.55923993, 0.55924046, 0.55924046])
```

In [178]:

```
score_cv = [x for x in cv_auc]
optimal_alpha_cv = alpha[score_cv.index(max(score_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_cv)))
print("Corresponding alpha value of cv is:",optimal_alpha_cv, '\n')
best_alpha_tfidf = optimal_alpha_cv
print(best_alpha_tfidf)
```

Maximum AUC score of cv is: 0.7080834505322461
Corresponding alpha value of cv is: 0.0001

0.0001

12.2.2. Finding The Best Hyperparameter "alpha" using L1 regularizer

In [179]:

```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

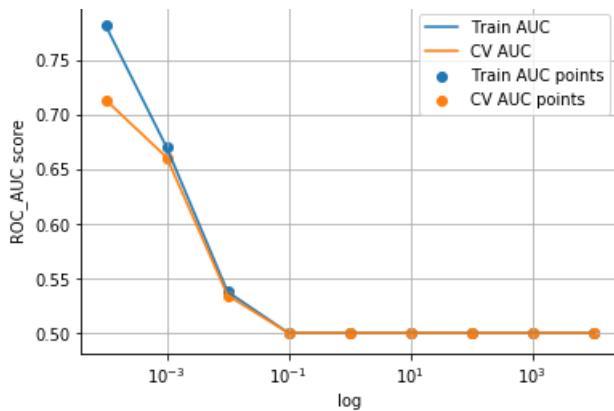
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_tfidf = SGDClassifier(loss = 'hinge', penalty = 'l1', class_weight = 'balanced')
clf = GridSearchCV(SVM_tfidf, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain2, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()
```

ROC_AUC vs log plot



12.2.3. Testing the performance of the model on test data, plotting ROC Curves

In [186]:

```
# best_alpha_tfidf = 0.001
```

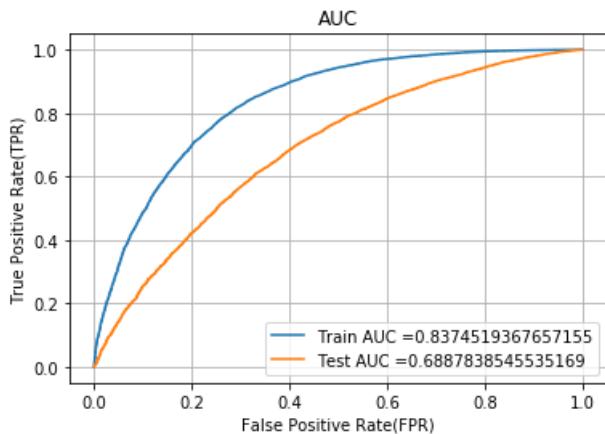
In [180]:

```
# https://scikitlearn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
SVM_tfidf = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_tfidf) #we choose l2 regularization hyperparameter
SVM_tfidf.fit(Xtrain2 ,y_train)

y_train_pred_tfidf = SVM_tfidf.decision_function(Xtrain2)
y_test_pred_tfidf = SVM_tfidf.decision_function(Xtest2)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_tfidf)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_tfidf)

plt.plot(train_fpr, train_tpr, label="Train AUC =" + str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC =" + str(auc(test_fpr, test_tpr)))
plt.legend()
plt.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid()
plt.show()
```



12.2.4. Building Confusion Matrix

In [181]:

```

# we are writing our own function for predict, with defined threshold
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshold, fpr, tpr):
    t = threshold[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
    return t

def predict_with_best_t(proba, threshold):
    predictions = []
    for i in proba:
        if i>=threshold:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions

```

In [182]:

```

print("=="*100)
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred_tfidf, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred_tfidf, best_t)))

```

=====

the maximum value of tpr*(1-fpr) 0.5788616289801294 for threshold 0.962
 Train confusion matrix
 [[5260 2166]
 [7606 34009]]
 Test confusion matrix
 [[2784 2675]
 [7183 23410]]

In [183]:

```

confusion_matrix_train_tfidf = pd.DataFrame(confusion_matrix(y_train,
predict_with_best_t(y_train_pred_tfidf, best_t)))
confusion_matrix_test_tfidf = pd.DataFrame(confusion_matrix(y_test,
predict_with_best_t(y_test_pred_tfidf, best_t)))

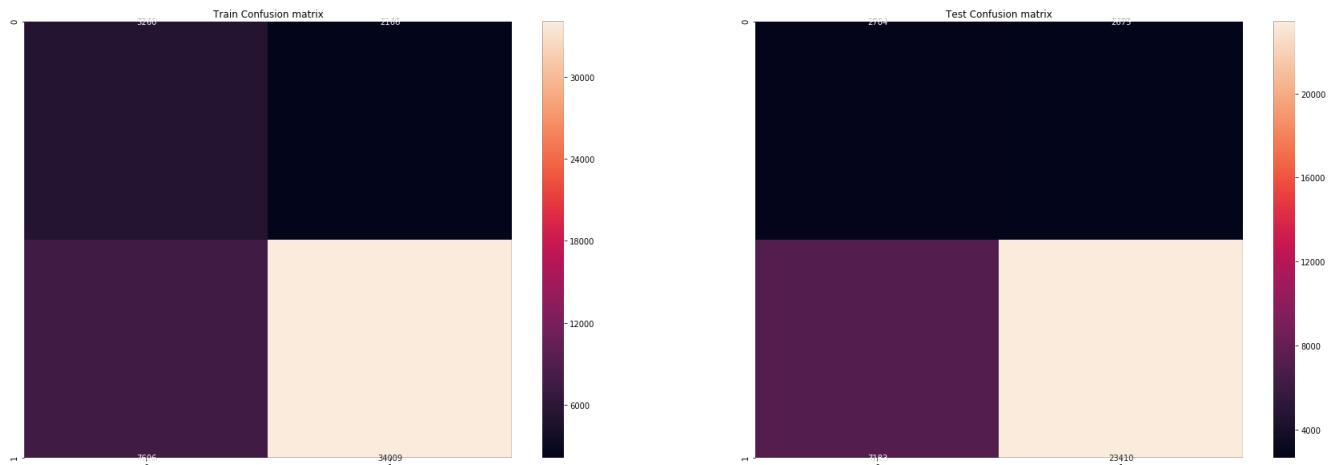
```

In [184]:

```

import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(30,10))
# sns.set(font_scale = 1.4)
sns.heatmap(confusion_matrix_train_tfidf, annot = True, ax = axes[0], fmt='g')
sns.heatmap(confusion_matrix_test_tfidf, annot = True, ax = axes[1], fmt='g')
axes[0].set_title('Train Confusion matrix')
axes[1].set_title('Test Confusion matrix')
plt.show()

```



12.3. Applying Support Vector Machines on Average Word2Vec, SET3

12.3.1. Finding The Best Hyperparameter "alpha" using 'l2' regularizer

In [185]:

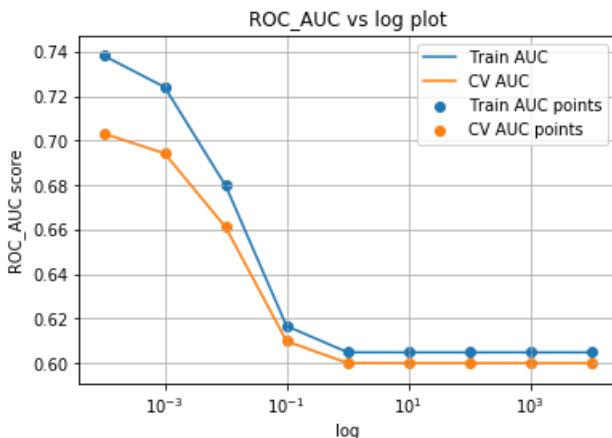
```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_avgw2v = SGDClassifier(loss = 'hinge', penalty = 'l2', class_weight = 'balanced')
clf = GridSearchCV(SVM_avgw2v, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain3, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()
```



In [186]:

```
cv_auc
```

Out[186]:

```
array([0.70313035, 0.69421254, 0.66103837, 0.60978999, 0.59993896,
       0.59991882, 0.59992194, 0.59992235, 0.59992177])
```

In [187]:

```
score_cv = [x for x in cv_auc]
optimal_alpha_cv = alpha[score_cv.index(max(score_cv))]
```

```

print("Maximum AUC score of cv is:" + ' ' + str(max(score_cv)))
print("Corresponding alpha value of cv is:",optimal_alpha_cv, '\n')
best_alpha_avgw2v = optimal_alpha_cv
print(best_alpha_avgw2v)

```

Maximum AUC score of cv is: 0.7031303486477328
Corresponding alpha value of cv is: 0.0001

0.0001

12.3.2. Finding The Best Hyperparameter "alpha" using L1 regularizer

In [188]:

```

from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

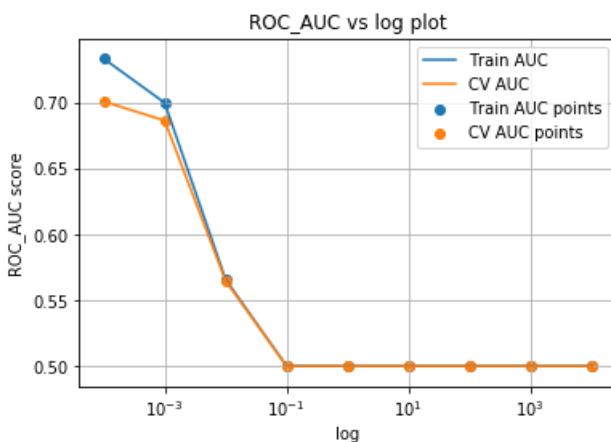
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_avgw2v = SGDClassifier(loss = 'hinge', penalty = 'l1', class_weight = 'balanced')
clf = GridSearchCV(SVM_avgw2v, parameters, cv = 5, scoring='roc_auc',return_train_score=True)
clf.fit(Xtrain3, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()

```



12.3.3. Testing the performance of the model on test data, plotting ROC Curves

In [192]:

```
# best_alpha_avgw2v = 0.01
```

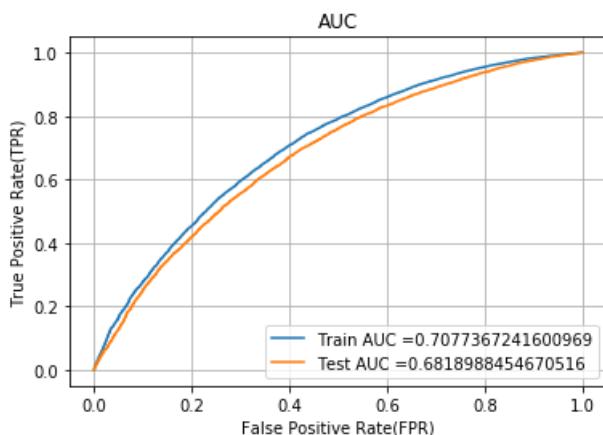
In [189]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
SVM_avgw2v = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_avgw2v) #we choose l2 regularization hyperparameter
SVM_avgw2v.fit(Xtrain3 ,y_train)

y_train_pred_avgw2v = SVM_avgw2v.decision_function(Xtrain3)
y_test_pred_avgw2v = SVM_avgw2v.decision_function(Xtest3)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_avgw2v)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_avgw2v)

plt.plot(train_fpr, train_tpr, label="Train AUC =" + str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC =" + str(auc(test_fpr, test_tpr)))
plt.legend()
plt.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid()
plt.show()
```



12.3.4. Building Confusion Matrix

In [190]:

```
# we are writing our own function for predict, with defined threshold
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshold, fpr, tpr):
    t = threshold[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
    return t

def predict_with_best_t(proba, threshold):
    predictions = []
    for i in proba:
        if i>=threshold:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [191]:

```
print("=="*100)
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred_avgw2v, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred_avgw2v, best_t)))
```

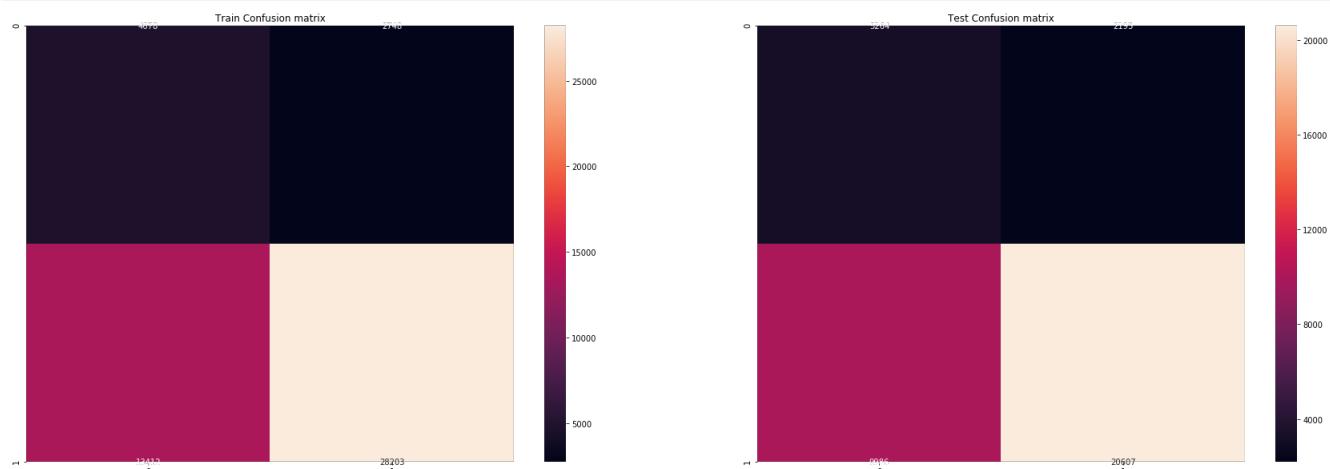
```
=====
the maximum value of tpr*(1-fpr) 0.4269241092997871 for threshold 1.238
Train confusion matrix
[[ 4678 2748]
 [13412 28203]]
Test confusion matrix
[[ 3264 2195]
 [ 9986 20607]]
```

In [192]:

```
confusion_matrix_train_avgw2v = pd.DataFrame(confusion_matrix(y_train,
predict_with_best_t(y_train_pred_avgw2v, best_t)))
confusion_matrix_test_avgw2v = pd.DataFrame(confusion_matrix(y_test,
predict_with_best_t(y_test_pred_avgw2v, best_t)))
```

In [193]:

```
import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(30,10))
# sns.set(font_scale = 1.4)
sns.heatmap(confusion_matrix_train_avgw2v, annot = True ,ax = axes[0],fmt='g')
sns.heatmap(confusion_matrix_test_avgw2v, annot = True , ax = axes[1],fmt = 'g')
axes[0].set_title('Train Confusion matrix')
axes[1].set_title('Test Confusion matrix')
plt.show()
```



12.4. Applying Support Vector Machines on TFIDF Word2Vec, SET4

12.4.1. Finding The Best Hyperparameter "alpha" with l2 regularizer

In [194]:

```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

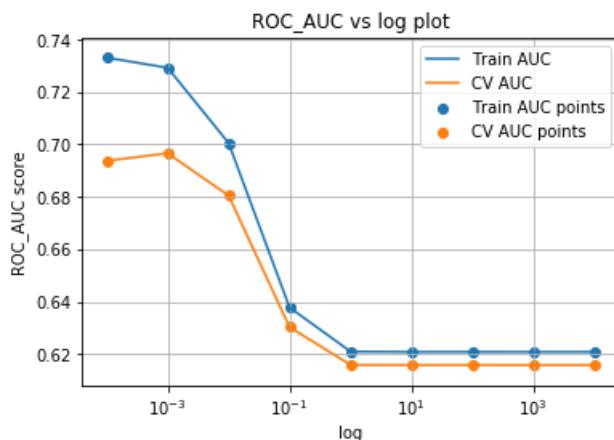
parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_tfidfW2V = SGDClassifier(loss = 'hinge', penalty = 'l2', class_weight = 'balanced')
clf = GridSearchCV(SVM_tfidfW2V, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain4, v_train)
```

```

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()

```



In [195]:

```
cv_auc
```

Out[195]:

```
array([0.69367568, 0.69655527, 0.68042026, 0.6302879 , 0.61587482,
       0.61581933, 0.615814 , 0.6158073 , 0.61580876])
```

In [196]:

```

score_cv = [x for x in cv_auc]
optimal_alpha_cv = alpha[score_cv.index(max(score_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_cv)))
print("Corresponding alpha value of cv is:",optimal_alpha_cv, '\n')
best_alpha_tfidf2v = optimal_alpha_cv
print(best_alpha_tfidf2v)

```

```
Maximum AUC score of cv is: 0.6965552704468035
Corresponding alpha value of cv is: 0.001
```

```
0.001
```

12.4.2. Finding The Best Hyperparameter "alpha" using l1 regularizer

In [197]:

```

from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

```

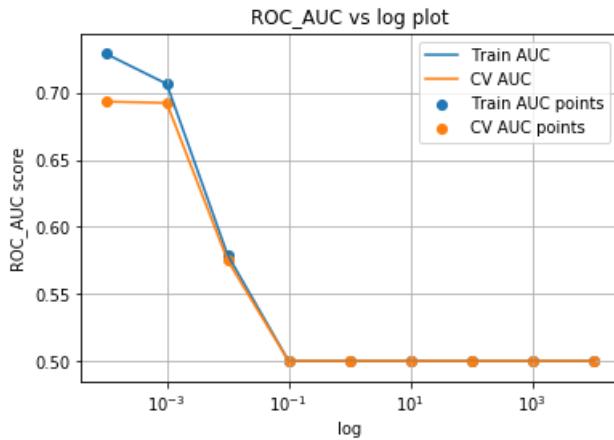
```

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_tfidf2v = SGDClassifier(loss = 'hinge', penalty = 'l1', class_weight = 'balanced')
clf = GridSearchCV(SVM_tfidf2v, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain4, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()

```



12.4.3. Testing the performance of the model on test data, plotting ROC Curves

In [198]:

```
# best_alpha_tfidf2v = 0.01
```

In [198]:

```

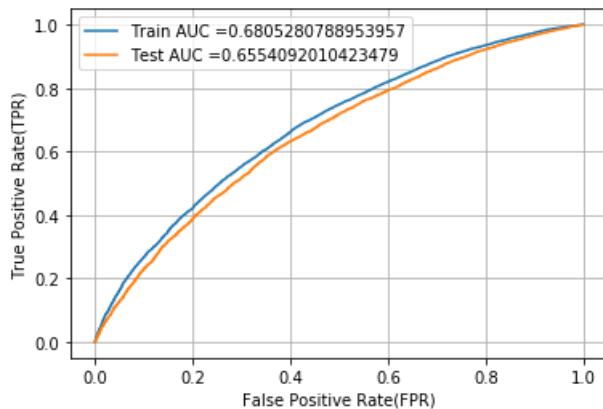
#
https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
SVM_tfidf2v = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_tfidf2v)#we choose l2 regularization hyperparameter
SVM_tfidf2v.fit(Xtrain4 ,y_train)

y_train_pred_tfidf2v = SVM_tfidf2v.decision_function(Xtrain4)
y_test_pred_tfidf2v = SVM_tfidf2v.decision_function(Xtest4)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_tfidf2v)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_tfidf2v)

plt.plot(train_fpr, train_tpr, label="Train AUC =" + str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC =" + str(auc(test_fpr, test_tpr)))
plt.legend()
plt.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(True)
plt.show()

```



12.4.4. Building Confusion Matrix

In [199]:

```
# we are writing our own function for predict, with defined threshold
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshold, fpr, tpr):
    t = threshold[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
    return t

def predict_with_best_t(proba, threshold):
    predictions = []
    for i in proba:
        if i>=threshold:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions
```

In [200]:

```
print("=="*100)
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred_tfidf2v, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred_tfidf2v, best_t)))
```

```
=====
the maximum value of tpr*(1-fpr) 0.3982780673351412 for threshold 1.271
Train confusion matrix
[[ 4418  3008]
 [13756 27859]]
Test confusion matrix
[[ 3019  2440]
 [10098 20495]]
```

In [201]:

```
confusion_matrix_train_tfidf2v = pd.DataFrame(confusion_matrix(y_train,
predict_with_best_t(y_train_pred_tfidf2v, best_t)))
confusion_matrix_test_tfidf2v = pd.DataFrame(confusion_matrix(y_test,
predict_with_best_t(y_test_pred_tfidf2v, best_t)))
```

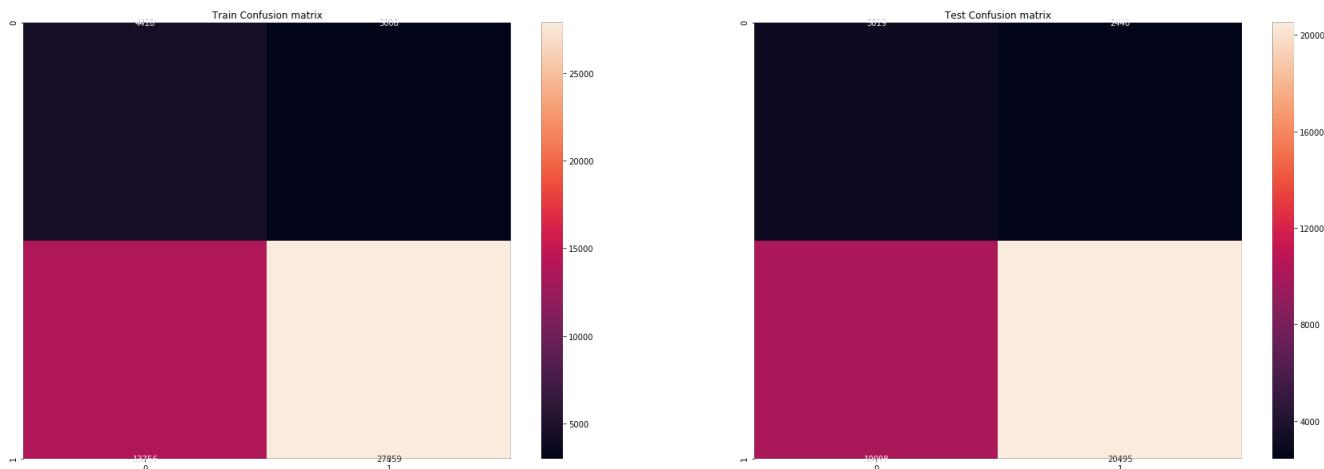
In [202]:

```
import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(30,10))
# sns.set(font_scale = 1.4)
sns.heatmap(confusion_matrix_train_tfidf2v, annot = True, ax = axes[0], fmt='g')
```

```

sns.heatmap(confusion_matrix_test_tfidfwzv, annot = True , ax = axes[1], fmt = 'g')
axes[0].set_title('Train Confusion matrix')
axes[1].set_title('Test Confusion matrix')
plt.show()

```



12.5. Applying Support Vector Machines on SET5

12.5.1. Finding The Best Hyperparameter "alpha" using l2 regularizer

In [203]:

```

from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

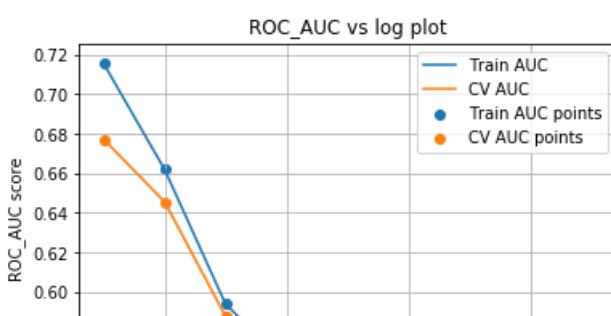
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

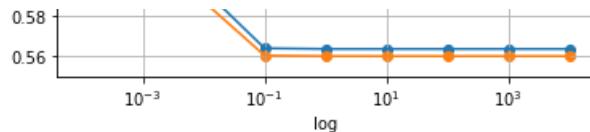
parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_set5 = SGDClassifier(loss = 'hinge', penalty = 'l2', class_weight = 'balanced')
clf = GridSearchCV(SVM_set5, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain5, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()

```





In [204]:

```
cv_auc
```

Out[204]:

```
array([0.67707643, 0.64543614, 0.58763918, 0.56029835, 0.56000054,
       0.56000089, 0.55999905, 0.56000122, 0.56000122])
```

In [208]:

```
score_cv = [x for x in cv_auc]
optimal_alpha_cv = alpha[score_cv.index(max(score_cv))]
print("Maximum AUC score of cv is:" + ' ' + str(max(score_cv)))
print("Corresponding alpha value of cv is:",optimal_alpha_cv, '\n')
best_alpha_set5 = optimal_alpha_cv
print(best_alpha_set5)
```

Maximum AUC score of cv is: 0.6770764278517651

Corresponding alpha value of cv is: 0.0001

0.0001

12.5.2. Finding The Best Hyperparameter "alpha" using l1 regularizer

In [209]:

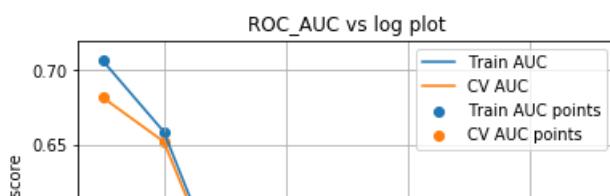
```
from sklearn.metrics import roc_auc_score
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV

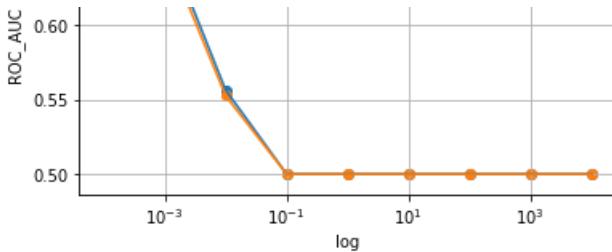
from sklearn import linear_model
from sklearn.linear_model import SGDClassifier
from sklearn import svm

parameters = {'alpha':[10**-4, 10**-3, 10**-2, 10**-1, 10**0, 10**1, 10**2, 10**3, 10**4]}
SVM_set5 = SGDClassifier(loss = 'hinge', penalty = 'l1', class_weight = 'balanced')
clf = GridSearchCV(SVM_set5, parameters, cv = 5, scoring='roc_auc', return_train_score=True)
clf.fit(Xtrain5, y_train)

train_auc = clf.cv_results_['mean_train_score']
cv_auc= clf.cv_results_['mean_test_score']

plt.plot(parameters['alpha'], train_auc, label='Train AUC')
plt.plot(parameters['alpha'], cv_auc, label='CV AUC')
plt.scatter(parameters['alpha'], train_auc, label='Train AUC points')
plt.scatter(parameters['alpha'], cv_auc, label='CV AUC points')
plt.legend()
plt.xlabel("log")
plt.xscale('log')
plt.ylabel("ROC_AUC score")
plt.title("ROC_AUC vs log plot")
plt.grid(True)
plt.show()
```





12.5.3. Testing the performance of the model on test data, plotting ROC Curves

In [243]:

```
# best_alpha_set5 = 0.001
```

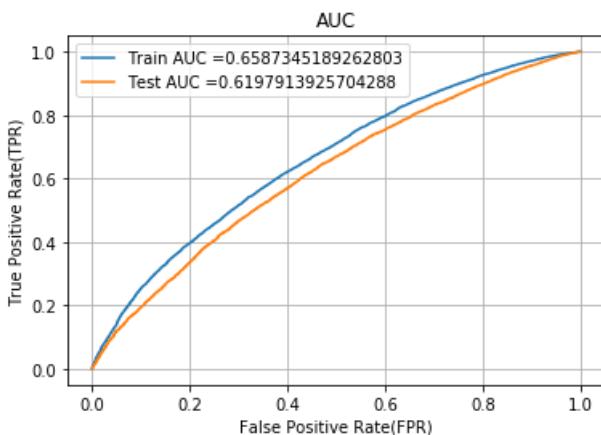
In [210]:

```
# https://scikit-learn.org/stable/modules/generated/sklearn.metrics.roc_curve.html#sklearn.metrics.roc_curve
from sklearn.metrics import roc_curve, auc
SVM_set5 = SGDClassifier(loss = 'hinge', penalty = 'l2', alpha = best_alpha_set5) #we choose l2 regularization hyperparameter
SVM_set5.fit(Xtrain5,y_train)

y_train_pred_set5 = SVM_set5.decision_function(Xtrain5)
y_test_pred_set5 = SVM_set5.decision_function(Xtest5)

train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_pred_set5)
test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_pred_set5)

plt.plot(train_fpr, train_tpr, label="Train AUC =" + str(auc(train_fpr, train_tpr)))
plt.plot(test_fpr, test_tpr, label="Test AUC =" + str(auc(test_fpr, test_tpr)))
plt.legend()
plt.ylabel("True Positive Rate(TPR)")
plt.xlabel("False Positive Rate(FPR)")
plt.title("AUC")
plt.grid(True)
plt.show()
```



12.5.4. Building Confusion Matrix

In [211]:

```
# we are writing our own function for predict, with defined threshold
# we will pick a threshold that will give the least fpr
def find_best_threshold(threshold, fpr, tpr):
    t = threshold[np.argmax(tpr*(1-fpr))]
    # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
    print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t, 3))
```

```

    print('the maximum value of tpr*(1-fpr) = %f for threshold %f' % (tpr*(1-fpr), t))
    return t

def predict_with_best_t(proba, threshold):
    predictions = []
    for i in proba:
        if i>=threshold:
            predictions.append(1)
        else:
            predictions.append(0)
    return predictions

```

In [212]:

```

print("=="*100)
from sklearn.metrics import confusion_matrix
best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
print("Train confusion matrix")
print(confusion_matrix(y_train, predict_with_best_t(y_train_pred_set5, best_t)))
print("Test confusion matrix")
print(confusion_matrix(y_test, predict_with_best_t(y_test_pred_set5, best_t)))

```

=====

the maximum value of tpr*(1-fpr) 0.3738922210214515 for threshold 1.074

Train confusion matrix

```

[[ 4531  2895]
 [16114 25501]]

```

Test confusion matrix

```

[[ 3116  2343]
 [12162 18431]]

```

In [213]:

```

confusion_matrix_train_set5 = pd.DataFrame(confusion_matrix(y_train,
predict_with_best_t(y_train_pred_set5, best_t)))
confusion_matrix_test_set5 = pd.DataFrame(confusion_matrix(y_test,
predict_with_best_t(y_test_pred_set5, best_t)))

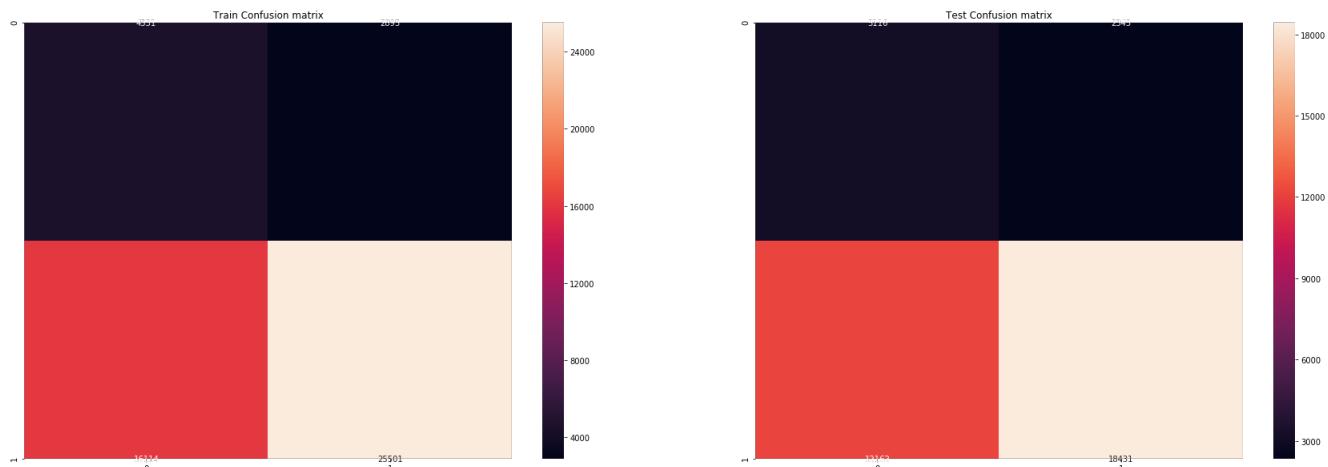
```

In [214]:

```

import seaborn as sns
fig, axes = plt.subplots(nrows=1, ncols=2, figsize=(30,10))
#sns.set(font_scale = 1.4)
sns.heatmap(confusion_matrix_train_set5, annot = True, ax = axes[0], fmt='g')
sns.heatmap(confusion_matrix_test_set5, annot = True, ax = axes[1], fmt = 'g')
axes[0].set_title('Train Confusion matrix')
axes[1].set_title('Test Confusion matrix')
plt.show()

```



13. Conclusion

In [215]:

```
# http://zetcode.com/python/prettytable/
from prettytable import PrettyTable

#If you get a ModuleNotFoundError error , install prettytable using: pip3 install prettytable

x = PrettyTable()

x.field_names = ["Vectorizer", "Model", "Hyper parameter", "Train AUC","Test AUC"]

x.add_row(["BOW", "SVM using l2 regularization", 0.01, 0.8027,0.6878])
x.add_row(["TFIDF", "SVM using l2 regularization", 0.0001, 0.8374,0.6887])
x.add_row(["AverageW2V", "SVM using l2 regularization", 0.0001, 0.7077,0.6818])
x.add_row(["TFIDFW2V", "SVM using l2 regularization", 0.0001, 0.6805,0.6554])
x.add_row(["SET5", "SVM using l2 regularization", 0.0001, 0.6587,0.6197])
```

In [216]:

```
print(x)
```

Vectorizer	Model	Hyper parameter	Train AUC	Test AUC
BOW	SVM using l2 regularization	0.01	0.8027	0.6878
TFIDF	SVM using l2 regularization	0.0001	0.8374	0.6887
AverageW2V	SVM using l2 regularization	0.0001	0.7077	0.6818
TFIDFW2V	SVM using l2 regularization	0.0001	0.6805	0.6554
SET5	SVM using l2 regularization	0.0001	0.6587	0.6197

In []: