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Lab-5: Stemming and Lemmatization on Movie Dataset

```
In [2]: from zipfile import ZipFile
    import glob
    import pandas as pd
    import nltk
    from sklearn.feature_extraction.text import TfidfVectorizer
    from sklearn.metrics.pairwise import linear_kernel
    from nltk.corpus import stopwords
    import warnings
    warnings.filterwarnings('ignore')
```

EXERCISE-1

```
In [19]: file_name = "movies.zip"
with ZipFile(file_name, 'r') as zip:
    zip.printdir()
```

movies/ 2018-01-19 08:32:38 0 movies/12 Angry Men.txt 2018-01-17 20:40:42 1007 movies/12 Years a Slave.txt 2018-01-17 20:42:50 6451 movies/4 Months, 3 Weeks and 2 Days.txt 2018-01-17 20:37:10 1151 movies/All About Eve.txt 2018-01-17 20:33:18 1346 movies/American Graffiti.txt 2018-01-17 20:44:30 3417 movies/Boyhood.txt 2018-01-17 20:27:14 1970 movies/Casablanca.txt 2018-01-17 20:26:26 1896 movies/Citizen Kane.txt 2018-01-17 20:23:56 1483 movies/Gone with the Wind.txt 2018-01-17 20:38:10 1318 movies/Hoop Dreams.txt 2018-01-17 20:34:12 7909 movies/Manchester by the Sea.txt 2018-01-17 20:40:06 3674 movies/Moonlight.txt 2018-01-17 20:33:142 2323 movies/Pan's Labyrinth.txt 2018-01-17 20:33:42 2323 movies/Psycho.txt 2018-01-17 20:34:46 3727 movies/Ran.txt 2018-01-17 20:34:46 3727 movies/Some Like It Hot.txt 2018-01-17 20:29:42 782 movies/The Godfather.txt 2018-01-17 20:25:32 4293 </th <th>File Name</th> <th>Modified</th> <th>Size</th>	File Name	Modified	Size
movies/12 Years a Slave.txt2018-01-17 20:42:506451movies/4 Months, 3 Weeks and 2 Days.txt2018-01-17 20:37:101151movies/All About Eve.txt2018-01-17 20:33:181346movies/American Graffiti.txt2018-01-17 20:44:303417movies/Boyhood.txt2018-01-17 20:27:141970movies/Casablanca.txt2018-01-17 20:26:261896movies/Citizen Kane.txt2018-01-17 20:23:561483movies/Gone with the Wind.txt2018-01-17 20:38:101318movies/Hoop Dreams.txt2018-01-17 20:34:127909movies/Manchester by the Sea.txt2018-01-17 20:40:063674movies/Moonlight.txt2018-01-17 20:31:422323movies/Pan's Labyrinth.txt2018-01-17 20:32:184431movies/Psycho.txt2018-01-17 20:34:463727movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	movies/	2018-01-19 08:32:38	0
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movies/American Graffiti.txt 2018-01-17 20:44:30 3417 movies/Boyhood.txt 2018-01-17 20:27:14 1970 movies/Casablanca.txt 2018-01-17 20:26:26 1896 movies/Citizen Kane.txt 2018-01-17 20:23:56 1483 movies/Gone with the Wind.txt 2018-01-17 20:38:10 1318 movies/Hoop Dreams.txt 2018-01-17 20:34:12 7909 movies/Manchester by the Sea.txt 2018-01-17 20:40:06 3674 movies/Moonlight.txt 2018-01-17 20:31:42 2323 movies/My Left Foot.txt 2018-01-17 20:38:50 1115 movies/Pan's Labyrinth.txt 2018-01-17 20:32:18 4431 movies/Psycho.txt 2018-01-17 20:34:46 3727 movies/Ran.txt 2018-01-17 20:43:48 2207 movies/Singin' in the Rain.txt 2018-01-17 20:29:42 782 movies/Some Like It Hot.txt 2018-01-17 20:25:32 4293	movies/4 Months, 3 Weeks and 2 Days.txt	2018-01-17 20:37:10	1151
movies/Boyhood.txt2018-01-17 20:27:141970movies/Casablanca.txt2018-01-17 20:26:261896movies/Citizen Kane.txt2018-01-17 20:23:561483movies/Gone with the Wind.txt2018-01-17 20:38:101318movies/Hoop Dreams.txt2018-01-17 20:34:127909movies/Manchester by the Sea.txt2018-01-17 20:40:063674movies/Moonlight.txt2018-01-17 20:31:422323movies/My Left Foot.txt2018-01-17 20:38:501115movies/Pan's Labyrinth.txt2018-01-17 20:32:184431movies/Psycho.txt2018-01-17 20:34:463727movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	movies/All About Eve.txt	2018-01-17 20:33:18	1346
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movies/Moonlight.txt2018-01-17 20:31:422323movies/My Left Foot.txt2018-01-17 20:38:501115movies/Pan's Labyrinth.txt2018-01-17 20:32:184431movies/Psycho.txt2018-01-17 20:34:463727movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	movies/Hoop Dreams.txt	2018-01-17 20:34:12	7909
movies/My Left Foot.txt 2018-01-17 20:38:50 1115 movies/Pan's Labyrinth.txt 2018-01-17 20:32:18 4431 movies/Psycho.txt 2018-01-17 20:34:46 3727 movies/Ran.txt 2018-01-17 20:43:48 2207 movies/Singin' in the Rain.txt 2018-01-17 20:29:42 782 movies/Some Like It Hot.txt 2018-01-17 20:35:40 7489 movies/The Godfather.txt 2018-01-17 20:25:32 4293	movies/Manchester by the Sea.txt	2018-01-17 20:40:06	3674
movies/Pan's Labyrinth.txt2018-01-17 20:32:184431movies/Psycho.txt2018-01-17 20:34:463727movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	<pre>movies/Moonlight.txt</pre>	2018-01-17 20:31:42	2323
movies/Psycho.txt2018-01-17 20:34:463727movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	movies/My Left Foot.txt	2018-01-17 20:38:50	1115
movies/Ran.txt2018-01-17 20:43:482207movies/Singin' in the Rain.txt2018-01-17 20:29:42782movies/Some Like It Hot.txt2018-01-17 20:35:407489movies/The Godfather.txt2018-01-17 20:25:324293	movies/Pan's Labyrinth.txt	2018-01-17 20:32:18	4431
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movies/Some Like It Hot.txt 2018-01-17 20:35:40 7489 movies/The Godfather.txt 2018-01-17 20:25:32 4293	movies/Ran.txt	2018-01-17 20:43:48	2207
movies/The Godfather.txt 2018-01-17 20:25:32 4293	movies/Singin' in the Rain.txt	2018-01-17 20:29:42	782
	movies/Some Like It Hot.txt	2018-01-17 20:35:40	7489
movies/Three Colors Red.txt 2018-01-17 20:28:22 2892	movies/The Godfather.txt	2018-01-17 20:25:32	4293
		2018-01-17 20:28:22	2892

```
In [21]: from nltk.stem import PorterStemmer
ps = PorterStemmer()
tokenizer = nltk.tokenize.WhitespaceTokenizer()
from nltk.stem import WordNetLemmatizer
lemmatizer = WordNetLemmatizer()
from nltk.stem import LancasterStemmer
ls = LancasterStemmer()
```

plantation contains the best examples both of McQueen's artistry and his occ asional tendency to gild the lily. In this section, he explores, with remark able attention to behavioral nuance, a rich array of social relationships go verned by the institution of slavery—not just master and slave but mistress and slave, shopkeeper and slave, and most of all, slave and slave. The wellkept mistress of a neighboring plantation owner (Alfre Woodard) tries to con vince Patsey that life as a sexual prisoner has its comparative advantages, ignoring the fact that Patsey's owner both rapes her and makes her work hard er than any man in the cotton fields. Later, Patsey and Solomon become close allies after she begs him to put an end to her misery by killing her, in a s cene that's as painful as it is sublimely acted. McQueen is particularly ade pt at exploring the intense yet guarded nature of relationships among the sl aves, for whom friendship—not to speak of love—is a dangerous and necessaril y transitory proposition. \n', '\n', 'But in a climactic set piece in which Solomon is forced by a gun-wielding Epps to whip Patsey nearly to death, I s ometimes felt smothered by McQueen's insistence on wallowing in the extremes of human anguish. McQueen has been accused in the past of aestheticizing suf fering, a critique which seems fairer to make of his two previous features (Hunger and Shame) than of this one. Hunger, in which Michael Fassbender sta rves himself before our eves as the Irish hunger striker Bobby Sands. and Sh

A. How many sentences in each file?

```
In [25]: | files = [file for file in glob.glob("movies/*")]
         for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                 contents = f.readlines()
                 for row in contents:
                     sent_text = nltk.sent_tokenize(row)
                     print("sentence tokenize ", len(sent_text))
         Selicence Cokelitze
         sentence tokenize 5
         sentence tokenize 0
         sentence tokenize 5
         sentence tokenize 0
         sentence tokenize 4
         sentence tokenize 4
         sentence tokenize 0
         sentence tokenize 5
         sentence tokenize 0
         sentence tokenize 2
         sentence tokenize 0
         sentence tokenize 5
         sentence tokenize 0
         sentence tokenize 3
         sentence tokenize 0
         sentence tokenize 1
         sentence tokenize 0
         sentence tokenize 3
         sentence tokenize 0
```

B. How many tokens in each file?

```
In [29]: | files = [file for file in glob.glob("movies/*")]
         for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                 contents = f.readlines()
                 for row1 in contents:
                     words = nltk.word_tokenize(row1)
                 print("word tokenize ", len(words))
         word tokenize 181
         word tokenize 119
         word tokenize 20
         word tokenize 276
         word tokenize 9
         word tokenize 70
         word tokenize 49
         word tokenize 98
         word tokenize 242
         word tokenize 67
         word tokenize 131
         word tokenize 157
         word tokenize 69
         word tokenize 66
         word tokenize 39
         word tokenize 25
         word tokenize 50
         word tokenize 208
         word tokenize 100
```

C. How many tokens excluding stop words in each file?

word tokenize 569

```
In [30]: files = [file for file in glob.glob("movies/*")]
         for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                 contents = f.readlines()
                 filtered_sentence = [w for w in words if not w in stop_words]
                 print("stopwords ", len(filtered_sentence))
         stopwords 365
         stopwords
                    365
         stopwords 365
         stopwords 365
         stopwords 365
         stopwords 365
         stopwords 365
         stopwords 365
         stopwords
                    365
         stopwords 365
         stopwords 365
         stopwords
                    365
         stopwords 365
         stopwords
                    365
         stopwords 365
         stopwords 365
```

D. How many unique stems (ie., stemming) in each file? (Use PorterStemmer)

```
In [31]: def port_stemSentence(sentence):
    tokenizer = nltk.tokenize.WhitespaceTokenizer()
    tok = tokenizer.tokenize(sentence)
    filtered_sentence = [w for w in tok if not w in stop_words]
    stem_sentence = []
    for word in filtered_sentence:
        stem_sentence.append(ps.stem(word))
    return len(stem_sentence)
```

stopwords

stopwords

stopwords

stopwords 365

365

365

365

```
In [32]: | files = [file for file in glob.glob("movies/*")]
         for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                  contents = f.readline()
                  print("porter_stemming ")
                 print(port_stemSentence(contents))
         porter_stemming
         porter_stemming
         83
         porter_stemming
         porter_stemming
         138
         porter_stemming
         63
         porter_stemming
         porter_stemming
         20
         porter_stemming
         porter_stemming
         131
         porter_stemming
         porter_stemming
         53
         porter_stemming
         87
         porter_stemming
         porter_stemming
         porter_stemming
         porter_stemming
         porter_stemming
         52
         porter_stemming
         porter_stemming
         porter_stemming
         282
```

E. How many unique stems (ie., stemming) in each file? (Use LancasterStemmer)

```
In [33]: def lan_stemSentence(sentence):
    tokenizer = nltk.tokenize.WhitespaceTokenizer()
    tok = tokenizer.tokenize(sentence)
    filtered_sentence = [w for w in tok if not w in stop_words]
    stem_sentence = []
    for word in filtered_sentence:
        stem_sentence.append(ls.stem(word))
    return len(stem_sentence)
```

```
In [34]: | files = [file for file in glob.glob("movies/*")]
         for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                  contents = f.readline()
                  print("lancaster_stemming ")
                 print(port_stemSentence(contents))
         lancaster_stemming
         lancaster_stemming
         83
         lancaster_stemming
         lancaster_stemming
         138
         lancaster_stemming
         63
         lancaster_stemming
         lancaster_stemming
         20
         lancaster_stemming
         lancaster_stemming
         131
         lancaster_stemming
         27
         lancaster_stemming
         53
         lancaster_stemming
         87
         lancaster_stemming
         lancaster_stemming
         93
         lancaster_stemming
         lancaster_stemming
         34
         lancaster_stemming
         52
         lancaster_stemming
         lancaster_stemming
         lancaster_stemming
         282
```

F. How many unique words (ie., lemmatization) in each file? (Use WordNetLemmatizer)

```
In [54]: def lemmSentence(sentence):
             tokenizer = nltk.tokenize.WhitespaceTokenizer()
             tok = tokenizer.tokenize(sentence)
             filtered_sentence = [w for w in tok if not w in stop_words]
             lemm_sentence = []
             for word in filtered_sentence:
                  lemm_sentence.append(lemmatizer.lemmatize(word))
             return len(lemm_sentence)
In [55]: for file in files:
             with open(file, 'r', encoding='cp1252') as f:
                 contents = f.readline()
                 print("lemmatization ")
                 print(lemmSentence(contents))
         lemmatization
         96
         lemmatization
         83
         lemmatization
         20
         lemmatization
         138
         lemmatization
         63
         lemmatization
         lemmatization
         20
         lemmatization
         51
         lemmatization
         131
         lemmatization
         27
         lemmatization
         lemmatization
         87
         lemmatization
         lemmatization
         93
         lemmatization
         23
         lemmatization
         34
         lemmatization
         52
         lemmatization
         38
         lemmatization
         33
         lemmatization
         282
```

EXERCISE-2

Step-1 For each movie:

Tokenize terms and build list of tokens

```
In [56]: tok = []
          for file in files:
              with open(file, 'r', encoding='cp1252') as f:
                  contents = f.read()
                  let=tokenizer.tokenize(contents)
                  tok.append(let)
          tok
            'a',
            'murder',
            'trial,',
            'one',
            "man's",
            'doubts',
            'about',
            'the',
            "accused's",
            'guilt',
            'gradually',
            'overcome',
            'the',
            'rather',
            'less-than-democratic',
            'prejudices',
            'of',
            'the',
            'other',
            'eleven',
```

Find lemmatized words from the tokens

```
In [58]: | tok_lem =[]
          for i in tok:
              for j in i:
                  to_lem = lemmatizer.lemmatize(j)
                  tok_lem.append(to_lem)
          tok_lem
           'liberalism,',
           'give',
           'ā',
           'nicely',
           'underplayed',
           'performance,',
           'while',
           'Cobb,',
           'Marshall',
           'and',
           'Begley',
           'in',
           'particular',
           'are',
           'highly',
           'effective',
           'in',
           'support.',
           'But',
```

Step-2

Build Term-Document matrix using TfldfVectorizer

```
In [59]: for file in files:
              with open(file,'r',encoding='cp1252') as f:
                   contents = f.read()
                   tok = tokenizer.tokenize(contents)
                   filtered_sentence = [w for w in tok if not w in stop_words]
                   tfidf = TfidfVectorizer(min_df=2,max_df=0.5,ngram_range=(1,2))
                   features = tfidf.fit transform(filtered sentence)
                   df = pd.DataFrame(features.todense(),columns=tfidf.get_feature_names())
                   print(df)
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          643
                     0.0
                            0.0
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                                            0.0
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                                                                     0.0
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                                                                                    0.0
          645
                     0.0
                            0.0
                                     0.0
                                            0.0
                                                  0.0
                                                        0.0
                                                              0.0
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                                                                             0.0
                                                                                    0.0
          646
                     0.0
                            0.0
                                     0.0
                                            0.0
                                                  0.0 0.0
                                                              0.0
                                                                     0.0
                                                                             0.0
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          [647 rows x 83 columns]
                abortion black communist disturbing drama
                                                                  everyone family gabita
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          1
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                     0.0
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                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
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          2
                     0.0
                            0.0
                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                 0.0
                                                                                          0.0
          3
                     1.0
                            0.0
                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                 0.0
                                                                                          0.0
          4
                     0.0
                            0.0
                                        1.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                 0.0
                                                                                          0.0
                     . . .
                             . . .
                                         . . .
                                                      . . .
                                                             . . .
                                                                        . . .
                                                                                 . . .
                                                                                          . . .
          122
                     0.0
                            0.0
                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                 0.0
                                                                                          0.0
                                                                                 0.0
          123
                     0.0
                            0.0
                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                          0.0
          124
                     0.0
                            0.0
                                        0.0
                                                      0.0
                                                             0.0
                                                                        0.0
                                                                                 0.0
                                                                                          0.0
```

Step-3

Take vectors of any two movies and compute cosine similarity

```
In [60]: with open(files[5],'r',encoding='cp1252')as f:
              contents = f.read()
              tok = tokenizer.tokenize(contents)
              filtered sentence = [w for w in tok if not w in stop words]
              tfidf = TfidfVectorizer(min_df=2,max_df=0.5,ngram_range=(1,2))
              movie1 = tfidf.fit_transform(filtered_sentence)
              print(movie1)
            (1, 10)
                           1.0
            (5, 2)
                           1.0
            (12, 13)
                           1.0
            (15, 5)
                           1.0
            (18, 10)
                           1.0
            (31, 20)
                           1.0
            (35, 12)
                           1.0
            (37, 3)
                           1.0
            (38, 9)
                           1.0
            (45, 10)
                           1.0
            (46, 11)
                           1.0
            (48, 19)
                           1.0
            (49, 16)
                           1.0
            (53, 8)
                           1.0
            (54, 4)
                           1.0
            (56, 19)
                           1.0
            (62, 20)
                           1.0
            (65, 12)
                           1.0
            (69, 7)
                           1.0
            (72, 18)
                           0.5773502691896258
            (72, 14)
                           0.5773502691896258
            (72, 17)
                           0.5773502691896258
            (77, 6)
                           1.0
            (78, 18)
                           0.5773502691896258
            (78, 14)
                           0.5773502691896258
            (108, 7)
                           1.0
            (118, 5)
                           1.0
            (121, 13)
                           1.0
            (124, 12)
                           1.0
            (128, 6)
                           1.0
            (134, 10)
                           1.0
            (138, 15)
                           1.0
            (143, 15)
                           1.0
            (148, 7)
                           1.0
            (152, 1)
                           1.0
            (154, 1)
                           1.0
            (156, 1)
                           1.0
            (165, 9)
                           1.0
            (166, 0)
                           1.0
            (172, 4)
                           1.0
            (173, 2)
                           1.0
            (174, 8)
                           1.0
            (177, 10)
                           1.0
            (179, 3)
                           1.0
            (180, 0)
                           1.0
            (188, 20)
                           1.0
            (193, 7)
                           1.0
```

(194,	11)	1.0
(196,	12)	1.0
(203,	10)	1.0

```
In [61]: with open(files[10], 'r', encoding='cp1252')as f:
              contents = f.read()
              tok = tokenizer.tokenize(contents)
              filtered sentence = [w for w in tok if not w in stop words]
              tfidf = TfidfVectorizer(min_df=2,max_df=0.5,ngram_range=(1,2))
              movie2 = tfidf.fit_transform(filtered_sentence)
              print(movie2)
            (0, 15)
                           1.0
            (1, 27)
                           1.0
            (2, 34)
                           1.0
            (3, 6)
                           1.0
            (4, 8)
                           1.0
            (7, 26)
                           1.0
            (11, 22)
                           1.0
            (13, 19)
                           1.0
                           1.0
            (15, 20)
            (17, 0)
                           1.0
            (29, 11)
                           1.0
            (34, 16)
                           1.0
            (46, 35)
                           1.0
            (52, 43)
                           1.0
            (53, 20)
                           1.0
            (62, 11)
                           1.0
            (66, 20)
                           1.0
            (67, 10)
                           1.0
            (71, 14)
                           1.0
            (73, 2)
                           1.0
            (74, 18)
                           1.0
            (77, 37)
                           1.0
            (78, 12)
                           1.0
            (81, 39)
                           1.0
            (82, 20)
                           1.0
            (323, 34)
                           1.0
            (324, 25)
                           1.0
            (331, 42)
                           1.0
            (332, 19)
                           1.0
            (333, 40)
                           1.0
            (336, 23)
                           1.0
            (337, 29)
                           1.0
            (342, 31)
                           1.0
            (343, 33)
                           1.0
            (345, 38)
                           1.0
            (353, 3)
                           1.0
            (354, 11)
                           1.0
            (356, 24)
                           1.0
            (359, 28)
                           1.0
            (361, 27)
                           1.0
            (362, 34)
                           1.0
            (366, 43)
                           1.0
            (369, 22)
                           1.0
                           1.0
            (371, 30)
            (373, 41)
                           1.0
            (379, 4)
                           1.0
            (381, 36)
                           1.0
```

```
(383, 7) 1.0
(384, 39) 1.0
(385, 4) 1.0
```

```
In [62]: doc1 = movie1[0:10]
    doc2 = movie1[:]
    score = linear_kernel(doc1,doc2)
    print(score)
```

```
[[0. 0. 0. ... 0. 0. 0.]

[0. 1. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]

...

[0. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]

[0. 0. 0. ... 0. 0. 0.]
```

```
In [ ]:
```