

```
from google.colab import files
upload=files.upload()
```

movies.csv

- **movies.csv**(application/vnd.ms-excel) - 494431 bytes, last modified: 12/21/2021 - 100% done
Saving movies.csv to movies (1).csv

```
import pandas as pd
```

```
movies = pd.read_csv('movies.csv', usecols=['movieId', 'title'])
movies.head()
```

	movieId	title	
0	1	Toy Story (1995)	
1	2	Jumanji (1995)	
2	3	Grumpier Old Men (1995)	
3	4	Waiting to Exhale (1995)	
4	5	Father of the Bride Part II (1995)	

```
from google.colab import files
upload=files.upload()
```

ratings.csv

- **ratings.csv**(application/vnd.ms-excel) - 2483723 bytes, last modified: 12/21/2021 - 100% done
Saving... ratings.csv

```
ratings = pd.read_csv('ratings.csv', usecols=['userId', 'movieId', 'rating'])
ratings.head()
```

	userId	movieId	rating	
0	1	1	4.0	
1	1	3	4.0	
2	1	6	4.0	
3	1	47	5.0	
4	1	50	5.0	

```
movies.shape
```

```
(9742, 2)
```

```
ratings.shape
```

(100836, 3)

```
movies_users=ratings.pivot(index='movieId',columns='userId',values='rating').fillna(0)
movies_users.head()
```

userId	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	1
movieId																	
1	4.0	0.0	0.0	0.0	4.0	0.0	4.5	0.0	0.0	0.0	0.0	0.0	0.0	0.0	2.5	0.0	4.
2	0.0	0.0	0.0	0.0	0.0	4.0	0.0	4.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.
3	4.0	0.0	0.0	0.0	0.0	5.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.
4	0.0	0.0	0.0	0.0	0.0	3.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	3.0	0.0	0.0	0.
5	0.0	0.0	0.0	0.0	0.0	5.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.

5 rows × 610 columns



```
from scipy.sparse import csr_matrix
```

```
mat_movies=csr_matrix(movies_users.values)
```

```
<9724x610 sparse matrix of type '<class 'numpy.float64'>'
  with 100836 stored elements in Compressed Sparse Row format>
```

Saving...



NearestNeighbors

```
model=NearestNeighbors(metric='cosine',algorithm='brute',n_neighbors=20)
model.fit(mat_movies)
```

```
from sklearn.neighbors import NearestNeighbors
model=NearestNeighbors(metric='cosine',algorithm='brute',n_neighbors=20)
model.fit(mat_movies)
```

```
NearestNeighbors(algorithm='brute', metric='cosine', n_neighbors=20)
```

```
!pip install fuzzywuzzy
```

Collecting fuzzywuzzy

Downloading fuzzywuzzy-0.18.0-py2.py3-none-any.whl (18 kB)

Installing collected packages: fuzzywuzzy

Successfully installed fuzzywuzzy-0.18.0

```
from fuzzywuzzy import process
```

```
/usr/local/lib/python3.7/dist-packages/fuzzywuzzy/fuzz.py:11: UserWarning: Using slow
warnings.warn('Using slow pure-python SequenceMatcher. Install python-Levenshtein t
```

```
def recommender(movie_name,data,n):
    idx= process.extractOne(movie_name, movies['title'])[2]
    print("Movies Selected:", movies[ 'title' ][idx], 'Index: ',idx)
    print('Searching for recomondation....')
    distance, indices= model.kneighbors (data[idx],n_neighbors=n)

    for i in indices :
        print (movies['title'][i].where(i!=idx))
```

```
recommender('toy story',mat_movies,10)
```

```
☞ Movies Selected: Toy Story (1995) Index:  0
   Searching for recomondation....
   0                                     NaN
2353                                'night Mother (1986)
418                                Jurassic Park (1993)
615                Independence Day (a.k.a. ID4) (1996)
224                Star Wars: Episode IV - A New Hope (1977)
314                                Forrest Gump (1994)
322                                Lion King, The (1994)
910    Once Upon a Time in the West (C'era una volta ...
546                                Mission: Impossible (1996)
655                                Diva (1981)
```

Saving...



✓ 5s completed at 10:05 PM



Saving... ✕