Consider the ratings dataset below containing data on UserID, MovieID, Rating, and Timestamp. Each line of this file represents one rating of one movie by one user and has the following format: UserID::MovieID::Rating::Timestamp

Ratings are made on a 5 star scale with half star increments.

UserID: represents the ID of the user

MovieID: represents the ID of the movie

Timestamp: represents seconds from midnight Coordinated Universal Time (UTC) of January 1, 1970.

Objective: Predict a movie-movie recommendation model.

```
import pandas as pd
In [1]:
          import numpy as np
         df = pd.read csv('Recommend.csv',names=['user id', 'movie id', 'rating','timestamp'])
          df
Out[2]:
                user_id movie_id rating timestamp
             0
                   196
                            242
                                        881250949
             1
                   186
                            302
                                        891717742
             2
                    22
                            377
                                        878887116
             3
                   244
                             51
                                        880606923
                            346
                                        886397596
                   166
         99995
                   880
                            476
                                        880175444
         99996
                   716
                            204
                                        879795543
                           1090
                                     1 874795795
         99997
                   276
```

	user_id	movie_id	rating	timestamp
99998	13	225	2	882399156
99999	12	203	3	879959583

100000 rows × 4 columns

Create a train test split of 75/25 by declaring 75/25 declaring number of users and movies

Populate the train matrix(user_id x movie_id) with ratings such that [user_id index, movie_id index] = given rating

Populate the test matrix(user_id x movie_id) with ratings such that [user-id index, movie_id index] = given rating

```
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.],
[0., 0., 0., ..., 0., 0., 0.]])
```

Create a cosine similarity matrices for movies and predict a movie-movie recommendation model

```
from sklearn.metrics import pairwise distances
In [10]:
          movie similarity = pairwise distances(train data matrix.T,metric='cosine')
          movie pred = train data matrix.dot(movie similarity)
          np.array([np.abs(movie similarity).sum(axis=1)])
          movie pred
Out[10]: array([[529.34336941, 563.79740673, 596.20312653, ..., 735.60471439,
                 715.68515867, 704.51924654],
                 [119.26521569, 142.02490873, 140.16706549, ..., 161.12038293,
                 162.18067772, 162.90120315],
                 [102.65652807, 108.25259751, 107.03939949, ..., 112.1286667,
                 118.29210216, 119.51285059],
                 [ 43.43428632, 57.07284422, 56.19586081, ..., 72.74203747,
                  70.25985418, 70.23903141],
                 [199.53548344, 220.14902596, 236.23080468, ..., 267.37143781,
                 262.77536636, 265.69586961],
                [292.02641008, 297.84766669, 329.29925712, ..., 415.
                 397.21555288, 399.70734371]])
 In [ ]:
```