

```
import pandas as pd
import numpy as np
import warnings
warnings.filterwarnings("ignore")
```

```
movie = pd.read_csv("movies.csv")
rating = pd.read_csv("ratings.csv")
```

```
movie.head()
```

	movieId	title	genres
0	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy
1	2	Jumanji (1995)	Adventure Children Fantasy
2	3	Grumpier Old Men (1995)	Comedy Romance
3	4	Waiting to Exhale (1995)	Comedy Drama Romance
4	5	Father of the Bride Part II (1995)	Comedy

```
rating.head()
```

	userId	movieId	rating	timestamp
0	1	31	2.5	1260759144
1	1	1029	3.0	1260759179
2	1	1061	3.0	1260759182
3	1	1129	2.0	1260759185
4	1	1172	4.0	1260759205

```
movie_final = pd.merge(movie, rating, on="movieId").drop(["genres", "timestamp", "movieId"], ax
```

```
movie_final["userId"] = "User"+movie_final["userId"].astype("str")
```

```
movie_final.head()
```

```

    title  userId  rating
from surprise import Dataset, SVD, accuracy, Reader
from surprise.model_selection import train_test_split

1 Toy Story (1995)    User9    4.0

movie_final1 = movie_final

2 Toy Story (1995)    User45    3.0

reader = Reader(rating_scale=(1,5))

data = Dataset.load_from_df(movie_final,reader)

trainset,testset = train_test_split(data,test_size=0.3,random_state=1)

svd = SVD(n_factors=100)

svd.fit(trainset)

<surprise.prediction_algorithms.matrix_factorization.SVD at 0x2b35fca7c40>

predictions = svd.test(testset)
accuracy.rmse(predictions)


RMSE: 0.9023
0.9023039191919443

movie_final.iloc[200]
user = "User534"
movie = "Toy Story (1995)"

svd.predict(user,movie)

Prediction(uid='User534', iid='Toy Story (1995)', r_ui=None, est=3.5402059941144537,

```



```

movie_user_matrix = movie_final1.pivot_table(index="userId",columns="title",values="rating"

movie_user_matrix.head()

```

```

title    "Great Performances" Cats (1998)    $9.99 (2008)    'Hellboy': The Seeds of Creation (2004)    'Neath the Arizona Skies (1934)    'Round Midnight (1986)    'Salem's Lot (2004)    'Til There Was You (1997)    'bui (19
userId
User1    NaN    NaN    NaN    NaN    NaN    NaN    NaN
movie_user_matrix.fillna(0,inplace=True)
User100    NaN    NaN    NaN    NaN    NaN    NaN    NaN
movie_user_matrix.head()
```

	title	"Great Performances" Cats (1998)	\$9.99 (2008)	'Hellboy': The Seeds of Creation (2004)	'Neath the Arizona Skies (1934)	'Round Midnight (1986)	'Salem's Lot (2004)	'Til There Was You (1997)	'bui (19
userId									
User1		0.0	0.0	0.0	0.0	0.0	0.0	0.0	
User10		0.0	0.0	0.0	0.0	0.0	0.0	0.0	
User100		0.0	0.0	0.0	0.0	0.0	0.0	0.0	
User101		0.0	0.0	0.0	0.0	0.0	0.0	0.0	
User102		0.0	0.0	0.0	0.0	0.0	0.0	0.0	

5 rows × 9064 columns

```

movie_watched = movie_user_matrix["Toy Story (1995)"]
movie_watched

userId
User1    0.0
User10   0.0
User100  4.0
User101  0.0
User102  0.0
...
User95   0.0
User96   0.0
User97   1.0
User98   0.0
User99   4.0
Name: Toy Story (1995), Length: 671, dtype: float64

similarity_scores = movie_user_matrix.corrwith(movie_watched)
similarity_scores
```

```

title
"Great Performances" Cats (1998)          -0.032411
$9.99 (2008)                               -0.018494
'Hellboy': The Seeds of Creation (2004)     0.040978
'Neath the Arizona Skies (1934)             -0.028157
'Round Midnight (1986)                     0.037176
...
xXx (2002)                                 0.112633
xXx: State of the Union (2005)              0.060730
¡Three Amigos! (1986)                      0.149133
À nous la liberté (Freedom for Us) (1931)  0.050854
İtirazım Var (2014)                        -0.028157
Length: 9064, dtype: float64

```

```
similarity_scores.sort_values(ascending=False).head(10)
```

```

title
Toy Story (1995)          1.000000
Toy Story 2 (1999)        0.474141
Bug's Life, A (1998)      0.393799
Groundhog Day (1993)      0.372371
Monsters, Inc. (2001)     0.366277
Independence Day (a.k.a. ID4) (1996)  0.356876
Finding Nemo (2003)       0.348941
Incredibles, The (2004)   0.346066
Shrek (2001)              0.344068
Willy Wonka & the Chocolate Factory (1971)  0.320002
dtype: float64

```