Amazon

November 21, 2022

1 import required libraries

```
[60]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import re
import seaborn as sns
%matplotlib inline
```

2 Import the Data

```
[61]: df = pd.read_csv('Amazon - Movies and TV Ratings.csv')
      df.head()
[61]:
                 user_id
                           Movie1
                                    Movie2
                                             Movie3
                                                      Movie4
                                                               Movie5
                                                                        Movie6
                                                                                  Movie7
          A3R50BKS70M2IR
                               5.0
                                        5.0
                                                 NaN
                                                          NaN
                                                                   NaN
                                                                            NaN
                                                                                     NaN
      1
           AH3QC2PC1VTGP
                               NaN
                                        NaN
                                                 2.0
                                                          NaN
                                                                   NaN
                                                                            NaN
                                                                                     NaN
      2
        A3LKP6WPMP9UKX
                                                          5.0
                                                                                     NaN
                               {\tt NaN}
                                        NaN
                                                 NaN
                                                                   NaN
                                                                            NaN
      3
           AVIY68KEPQ5ZD
                               NaN
                                        NaN
                                                 NaN
                                                          5.0
                                                                   NaN
                                                                            NaN
                                                                                     NaN
        A1CV1WROP5KTTW
                               NaN
                                        {\tt NaN}
                                                 NaN
                                                          NaN
                                                                   5.0
                                                                                     NaN
                                                                            NaN
          Movie8
                  Movie9
                               Movie197
                                          Movie198
                                                     Movie199
                                                                 Movie200
                                                                            Movie201
      0
             NaN
                      NaN
                                     NaN
                                                NaN
                                                           NaN
                                                                      NaN
                                                                                  NaN
      1
             NaN
                      NaN
                                    NaN
                                                NaN
                                                           NaN
                                                                      NaN
                                                                                  NaN
      2
             NaN
                      NaN
                                    NaN
                                                NaN
                                                           NaN
                                                                      NaN
                                                                                  NaN
      3
             NaN
                                                           NaN
                                                                                  NaN
                      NaN
                                    NaN
                                                NaN
                                                                      NaN
      4
             NaN
                      NaN
                                    NaN
                                                NaN
                                                           NaN
                                                                      NaN
                                                                                  NaN
          Movie202
                     Movie203
                                Movie204
                                           Movie205
                                                      Movie206
      0
               NaN
                          NaN
                                      NaN
                                                 NaN
                                                            NaN
      1
               NaN
                          NaN
                                      NaN
                                                 NaN
                                                            NaN
      2
               NaN
                          NaN
                                      NaN
                                                 NaN
                                                            NaN
      3
                          NaN
               NaN
                                      NaN
                                                 NaN
                                                            NaN
               NaN
                          NaN
                                      NaN
                                                 NaN
                                                            NaN
```

[5 rows x 207 columns]

	[0 10		0 0 1 0111111	-									
62]:	df.in	fo											
52]:		d method							user_id	Movie1	Movie2	Movie3	
		4 Movie		96 I			^	N - N	NT - NT	NT - NT	NT - NT	N - N	
	0		S70M2IR		5.0		.0	NaN	NaN	NaN	NaN	NaN	
	1	-	PC1VTGP		NaN		aN	2.0	NaN	NaN	NaN	NaN	
	2		PMP9UKX		NaN		aN	NaN	5.0	NaN	NaN	NaN	
	3	AVIY68			NaN		aN	NaN	5.0	NaN	NaN	NaN	
	4	A1CV1WR	UP5KTTW		NaN		aN	NaN	NaN	5.0	NaN	NaN	
	 4843	A1IMQ9W	 MEVKWH5	•••	 NaN	 N:	aN	 NaN	 NaN	 NaN	NaN	NaN	
	4844	A1KLIKP			NaN		aN aN	NaN	NaN	NaN	NaN	NaN	
	4845		FZL010D		NaN		aN aN	NaN	NaN	NaN	NaN	NaN	
	4846				NaN								
		AI4J76			NaN		aN aN	NaN NaN	NaN NaN	NaN NaN	NaN NaN	NaN NaN	
	4041	A14J/0	2110500		Ivaiv	IVe	alv	NaN	IValV	NaN	NaN	NaN	
		Movie8	Movie9	•••	Movi	e197	Mov	ie198	Movie199	Movie2	00 Movi	.e201 \	
	0	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	1	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	2	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	3	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	4	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	•••	•••				•••		•••	•••				
	4843	NaN	NaN	•••		${\tt NaN}$		NaN	NaN	N	aN	NaN	
	4844	NaN	NaN	•••		${\tt NaN}$		NaN	NaN	N	aN	NaN	
	4845	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	4846	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
	4847	NaN	NaN	•••		NaN		NaN	NaN	N	aN	NaN	
		Movie20	2 Movie	203	Mov	ie204	Мо	vie205	Movie206	3			
	0	Na		NaN		NaN		NaN	NaN				
	1	Na		NaN		NaN		NaN	NaN				
	2	Na		NaN		NaN		NaN	NaN				
	3	Na		NaN		NaN		NaN	NaN				
	4	Na		NaN		NaN		NaN	NaN				
		•••			•••								
	4843	Na		NaN		NaN		NaN	5.0				
	4844	Na		NaN		NaN		NaN	5.0				
	4845	Na	N	NaN		NaN		NaN	5.0)			

[4848 rows x 207 columns]>

NaN

NaN

NaN

NaN

NaN

NaN

[63]: df.shape

4846

4847

NaN

NaN

5.0

5.0

```
[63]: (4848, 207)
[64]: df.size
[64]: 1003536
[65]: df.dtypes
[65]: user_id
                   object
      Movie1
                  float64
      Movie2
                  float64
     Movie3
                  float64
      Movie4
                  float64
      Movie202
                  float64
                  float64
      Movie203
                  float64
      Movie204
      Movie205
                  float64
      Movie206
                  float64
      Length: 207, dtype: object
[66]: df.count()
[66]: user_id
                  4848
      Movie1
                     1
      Movie2
                     1
      Movie3
                     1
     Movie4
                     2
     Movie202
                     6
     Movie203
                     1
     Movie204
                     8
     Movie205
                    35
      Movie206
                    13
     Length: 207, dtype: int64
         maximum number of views
[67]: df.describe().T["count"].sort_values(ascending=False)[0:6]
[67]: Movie127
                  2313.0
      Movie140
                   578.0
      Movie16
                   320.0
      Movie103
                   272.0
```

Movie29

243.0

Movie91 128.0

Name: count, dtype: float64

```
[68]: df.index
```

[68]: RangeIndex(start=0, stop=4848, step=1)

```
[69]: df.columns
```

'Movie197', 'Movie198', 'Movie199', 'Movie200', 'Movie201', 'Movie202', 'Movie203', 'Movie204', 'Movie205', 'Movie206'], dtype='object', length=207)

[70]: df_filtered = df.fillna(value=0) df_filtered

[70]:			user_id	Mov	rie1	Movie	e2	Movie3	Movi	ie4	Movie5	Movi	e6	Movie7	\
	0	A3R50BK	S70M2IR		5.0	5.	. 0	0.0	(0.0	0.0	0	.0	0.0	
	1	AH3QC2	PC1VTGP		0.0	0.	. 0	2.0	(0.0	0.0	0	.0	0.0	
	2	A3LKP6W	PMP9UKX		0.0	0.	. 0	0.0	5	5.0	0.0	0	.0	0.0	
	3	AVIY68	KEPQ5ZD		0.0	0.	. 0	0.0	5	5.0	0.0	0	.0	0.0	
	4	A1CV1WR	OP5KTTW		0.0	0.	. 0	0.0	(0.0	5.0	0	.0	0.0	
	•••			•••	•••				•••	•••					
	4843	A1IMQ9W	MFYKWH5		0.0	0.	. 0	0.0	(0.0	0.0	0	.0	0.0	
	4844	A1KLIKP	UF5E88I		0.0	0.	. 0	0.0	(0.0	0.0	0	.0	0.0	
	4845	A5HG6W	FZL010D		0.0	0.	. 0	0.0	(0.0	0.0	0	.0	0.0	
	4846	A3UU690	TWXCG1X		0.0	0.	. 0	0.0	(0.0	0.0	0	.0	0.0	
	4847	AI4J76	2YI6S06		0.0	0.	. 0	0.0	(0.0	0.0	0	.0	0.0	
		Movie8	Movie9	•••	Movi	.e197	Моч	vie198	Movie	199	Movie2	200 M	ovi	e201 \	
	0	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	1	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	2	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	3	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	4	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	•••	•••				•••					•••				
	4843	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	4844	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	4845	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	4846	0.0	0.0	•••		0.0		0.0		0.0	C	0.0		0.0	
	4847	0.0	0.0			0.0		0.0		0.0	C	0.0		0.0	

Movie202 Movie203 Movie204 Movie205 Movie206 0 0.0 0.0 0.0 0.0 0.0

1	0.0	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0	0.0
•••	•••		•••	•••	
4843	0.0	0.0	0.0	0.0	5.0
4844	0.0	0.0	0.0	0.0	5.0
4845	0.0	0.0	0.0	0.0	5.0
4846	0.0	0.0	0.0	0.0	5.0
4847	0.0	0.0	0.0	0.0	5.0

[4848 rows x 207 columns]

3

0.0

0.0

```
[71]: df_filtered1 = df_filtered.drop(columns='user_id')
df_filtered1
```

[71]:		Movie1	Movie2	Movie3	Movie4	Movie5	Movie6	Movie7	Movie8	Movie9	\
	0			0.0				0.0		0.0	
	1	0.0	0.0	2.0	0.0	0.0	0.0	0.0	0.0	0.0	
	2	0.0	0.0	0.0	5.0	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	0.0	
	4	0.0	0.0	0.0	0.0	5.0	0.0	0.0	0.0	0.0	
				•••	•••			•••			
	4843	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	4844	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	4845	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	4846	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
	4847	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	
						Movie19					
	0	0.0	•••		0.0			0.0	0.0		
	1	0.0		0.0	0.0			0.0	0.0		
	2	0.0		0.0	0.0			0.0	0.0		
	3	0.0		0.0	0.0			0.0	0.0		
	4	0.0	•••	0.0	0.0		0 (0.0	0.0		
	4843	0.0		0.0					0.0		
	4844	0.0	•••	0.0	0.0			0.0	0.0		
	4845	0.0	•••	0.0	0.0	0.		0.0	0.0		
	4846	0.0	•••	0.0	0.0				0.0		
	4847	0.0	•••	0.0	0.0	0.	0 (0.0	0.0		
		Movie202) Movie	203 Ma	wie20/ 1	Movie205	Movie 206	3			
	0	0.0					0.0				
	1	0.0					0.0				
	2	0.0				0.0	0.0				
	4	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
			•••		
4843	0.0	0.0	0.0	0.0	5.0
4844	0.0	0.0	0.0	0.0	5.0
4845	0.0	0.0	0.0	0.0	5.0
4846	0.0	0.0	0.0	0.0	5.0
4847	0.0	0.0	0.0	0.0	5.0

[4848 rows x 206 columns]

```
[72]: df_max_views = df_filtered1.sum()
      df_max_views
```

```
[72]: Movie1
                    5.0
     Movie2
                    5.0
     Movie3
                    2.0
     Movie4
                   10.0
     Movie5
                  119.0
     Movie202
                   26.0
     Movie203
                    3.0
     Movie204
                   35.0
     Movie205
                  162.0
     Movie206
                   64.0
     Length: 206, dtype: float64
```

4 Which movies have maximum views/ratings?

```
[73]: max_views = df_max_views.argmax()
      max_views
```

[73]: 126

[74]: len(df_max_views.index)

[74]: 206

5 What is the average rating for each movie? Define the top 5 movies with the maximum ratings

```
[75]: Average_rating_of_each_movie = sum(df_max_views)/len(df_max_views.index)
      Average rating of each movie
[75]: 106.44660194174757
[76]: amazon_df = pd.DataFrame(df_max_views)
      amazon_df.head()
[76]:
                  0
      Movie1
                5.0
      Movie2
                5.0
      Movie3
                2.0
     Movie4
             10.0
     Movie5 119.0
[77]: amazon_df.columns=['rating']
[78]: amazon_df.index
[78]: Index(['Movie1', 'Movie2', 'Movie3', 'Movie4', 'Movie5', 'Movie6', 'Movie7',
             'Movie8', 'Movie9', 'Movie10',
             'Movie197', 'Movie198', 'Movie199', 'Movie200', 'Movie201', 'Movie202',
             'Movie203', 'Movie204', 'Movie205', 'Movie206'],
            dtype='object', length=206)
[79]: amazon_df.nsmallest(5, 'rating')
[79]:
               rating
     Movie45
                  1.0
     Movie58
                  1.0
     Movie60
                  1.0
     Movie67
                  1.0
      Movie69
                  1.0
```

6 Divide the data into training and test data

```
[80]: import sklearn
[81]: from sklearn.model_selection import train_test_split
```

```
[82]: trainset, testset = train_test_split(df, test_size=0.25)
```

7 Build a recommendation model on training data

```
[83]: import surprise
     from surprise import Reader
     from surprise import accuracy
     from surprise import Dataset
     from surprise.model_selection import train_test_split
     from surprise import SVD
     from surprise.model_selection import cross_validate
[84]: from surprise import SVD
[85]: df_melt = df.melt(id_vars = df.columns[0],value_vars=df.columns[1:],
                                 var_name="Movies", value_name="Rating")
     df melt
                    user_id Movies Rating
[85]:
             A3R50BKS70M2IR Movie1
                                          5.0
     1
              AH3QC2PC1VTGP Movie1
                                          NaN
     2
             A3LKP6WPMP9UKX Movie1
                                          NaN
     3
              AVIY68KEPQ5ZD Movie1
                                          NaN
             A1CV1WROP5KTTW Movie1
                                          NaN
                                          5.0
     998683 A1IMQ9WMFYKWH5 Movie206
                                          5.0
     998684 A1KLIKPUF5E88I Movie206
     998685
             A5HG6WFZL010D
                             Movie206
                                          5.0
     998686 A3UU690TWXCG1X Movie206
                                          5.0
     998687 AI4J762YI6S06 Movie206
                                          5.0
     [998688 rows x 3 columns]
[86]: rdr = Reader()
     df = Dataset.load_from_df(df_melt.fillna(0),reader = rdr)
[86]: <surprise.dataset.DatasetAutoFolds at 0x7f54d5390090>
[87]: trainset, testset = train_test_split(df,test_size = 0.25)
[88]:
      from sklearn.decomposition import TruncatedSVD
[89]: svd = SVD()
     svd.fit(trainset)
```

[89]: <surprise.prediction_algorithms.matrix_factorization.SVD at 0x7f54cb1f1850>

8 Make predictions on the test data

```
[90]: pred = svd.test(testset)
[91]: accuracy.rmse(pred)
     RMSE: 1.0266
[91]: 1.0265592501512613
[92]: accuracy.mae(pred)
     MAE: 1.0122
[92]: 1.0122054491299677
[93]: cross_validate(svd, df, measures = ['RMSE', 'MAE'], cv = 3, verbose = True)
     Evaluating RMSE, MAE of algorithm SVD on 3 split(s).
                       Fold 1 Fold 2 Fold 3 Mean
     RMSE (testset)
                       1.0256 1.0265 1.0262 1.0261
                                                       0.0004
     MAE (testset)
                       1.0118 1.0122 1.0122 1.0121
                                                       0.0002
     Fit time
                       36.51
                               38.08
                                       40.78
                                               38.46
                                                       1.76
     Test time
                       4.00
                               3.44
                                       3.85
                                               3.76
                                                       0.24
[93]: {'test_rmse': array([1.02564356, 1.02650914, 1.02624722]),
       'test_mae': array([1.01183495, 1.01220878, 1.01224944]),
       'fit_time': (36.509862184524536, 38.084420680999756, 40.77697539329529),
       'test time': (4.00281548500061, 3.4420347213745117, 3.8467113971710205)}
 []:
```