Covid-19 (Personalized Treatment Recommendations) In [1]: import pandas as pd import numpy as np import warnings warnings.filterwarnings("ignore") In [2]: treatment=pd.read_csv(r"C:\Users\reshma_koduri\Downloads\treatment ratings.csv") In [3]: treatment USER_ID AGE SEX SOURCE HAEMOGLOBINS ERYTHROCYTE LEUCOCYTE HAEMATOCRIT Out[3]: 0 1 68 F 11.8 4.65 6.3 35.1 out 1 2 70 F out 14.8 5.39 12.7 43.5 2 3 64 F out 11.3 4.74 13.2 33.5 3 4 73 F 13.7 4.98 10.5 39.1 out 5 66 9.9 22.1 30.9 [4 Μ out 4.23 1184 1185 90 F out 12.5 4.41 10.9 37.1 1185 F 9.4 33.7 1186 90 10.6 4.44

1189 rows × 11 columns

1187

1188

1189

90

25

31

F

Μ

Μ

1186

1187

1188

In [4]: treatment.describe()

11.2

13.8

13.8

4.06

4.88

4.86

8.4

3.5

2.3

35.4

40.0

39.4

out

in

in

in

)ut[4]:		USER_ID	AGE	HAEMOGLOBINS	ERYTHROCYTE	LEUCOCYTE	HAEMATOCRIT	R.
	count	1189.000000	1189.000000	1189.000000	1189.000000	1189.000000	1189.000000	1189
	mean	595.000000	47.915055	13.158621	4.890294	7.475610	39.529521	2
	std	343.379042	20.320298	1.869245	0.672515	3.994819	5.194292	0
	min	1.000000	19.000000	3.800000	2.150000	1.100000	14.600000	1
	25%	298.000000	31.000000	12.000000	4.480000	4.900000	36.300000	1
	50%	595.000000	38.000000	13.200000	4.880000	6.900000	39.500000	2
	75%	892.000000	69.000000	14.400000	5.300000	9.200000	43.200000	3
	max	1189.000000	96.000000	18.100000	7.610000	32.000000	54.000000	3
	4							•

```
In [5]:
         treatment.info()
         <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1189 entries, 0 to 1188
        Data columns (total 11 columns):
         #
              Column
                            Non-Null Count Dtype
         0
              USER_ID
                            1189 non-null
                                             int64
         1
                            1189 non-null
                                             int64
         2
                                             object
              SEX
                            1189 non-null
         3
              SOURCE
                            1189 non-null
                                             object
              HAEMOGLOBINS 1189 non-null
         4
                                             float64
         5
              ERYTHROCYTE
                            1189 non-null
                                             float64
         6
              LEUCOCYTE
                            1189 non-null
                                            float64
         7
              HAEMATOCRIT
                            1189 non-null
                                             float64
         8
              TREATMENT
                            1189 non-null
                                             object
         9
              CONDITION
                            1189 non-null
                                             object
         10 RATINGS
                            1189 non-null
                                             int64
        dtypes: float64(4), int64(3), object(4)
        memory usage: 102.3+ KB
In [6]:
         treatment.isna().sum()
                         0
        USER ID
Out[6]:
                         0
         SEX
                         0
        SOURCE
                         0
        HAEMOGLOBINS
        ERYTHROCYTE
                         0
        LEUCOCYTE
                         0
        HAEMATOCRIT
                         0
        TREATMENT
                         0
        CONDITION
                         0
        RATINGS
        dtype: int64
In [7]:
         treatment.groupby(['CONDITION']).count()
                    USER_ID AGE SEX SOURCE HAEMOGLOBINS ERYTHROCYTE LEUCOCYTE HAEMATO
Out[7]:
         CONDITION
           improved
                        414
                             414 414
                                          414
                                                          414
                                                                        414
                                                                                   414
              stable
                             418 418
                                                          418
                                                                        418
                                                                                   418
                        418
                                          418
              worse
                        357
                             357
                                  357
                                           357
                                                          357
                                                                        357
                                                                                   357
In [8]:
         treatment['CONDITION'].unique()
         array(['worse', 'improved', 'stable'], dtype=object)
Out[8]:
In [9]:
         treatment['SEX'].unique()
        array(['F', 'M'], dtype=object)
Out[9]:
```

```
In [10]:
           treatment['SOURCE'].unique()
          array(['out', 'in'], dtype=object)
Out[10]:
In [11]:
           treatment['TREATMENT'].unique()
          array(['Oxygen Therapy', 'Convalescent Plasma', 'COVID-19 Vaccines',
Out[11]:
                  'Tocilizumab', 'Dexamethasone',
                  'Non-invasive Ventilation (CPAP/BiPAP)',
                  'Casirivimab and imdevimab', 'Baricitinib', 'Remdesivir',
                  'Monoclonal Antibodies', 'Etesevimab', 'Mechanical Ventilation',
                  'Molnupiravir', 'Bamlanivimab', 'Sotrovimab', 'Anticoagulants',
                  'Methylprednisolone', 'Heparin'], dtype=object)
In [12]:
           treatment['SEX']=treatment['SEX'].map({'F':1,'M':0})
           treatment['CONDITION']=treatment['CONDITION'].map({'worse':1,'stable':2,'improved':3
           treatment['SOURCE']=treatment['SOURCE'].map({'in':1,'out':0})
In [13]:
           treatment
                USER ID AGE SEX SOURCE HAEMOGLOBINS ERYTHROCYTE LEUCOCYTE HAEMATOCRIT
Out[13]:
             0
                      1
                           68
                                 1
                                          0
                                                        11.8
                                                                      4.65
                                                                                   6.3
                                                                                                 35.1
                      2
                           70
                                          0
                                                                      5.39
                                                                                  12.7
                                                                                                 43.5
             1
                                 1
                                                        14.8
             2
                      3
                           64
                                 1
                                          0
                                                        11.3
                                                                      4.74
                                                                                  13.2
                                                                                                 33.5
                                                                                  10.5
             3
                      4
                           73
                                          0
                                                                      4.98
                                                                                                 39.1
                                 1
                                                        13.7
                      5
                           66
                                 0
                                          0
                                                         9.9
                                                                      4.23
                                                                                  22.1
                                                                                                 30.9 [
                      ...
                                                                                  10.9
          1184
                    1185
                           90
                                 1
                                          0
                                                        12.5
                                                                      4.41
                                                                                                 37.1
          1185
                    1186
                           90
                                 1
                                          0
                                                        10.6
                                                                      4.44
                                                                                   9.4
                                                                                                 33.7
          1186
                    1187
                           90
                                                        11.2
                                                                      4.06
                                                                                   8.4
                                                                                                 35.4
          1187
                    1188
                           25
                                 0
                                                        13.8
                                                                      4.88
                                                                                   3 5
                                                                                                 40.0
                                                                                   2.3
                                                                                                 39.4
          1188
                    1189
                           31
                                 0
                                          1
                                                        13.8
                                                                      4.86
         1189 rows × 11 columns
In [14]:
           treatment1=treatment.drop(['TREATMENT'],axis=1)
           treatment1
                USER_ID AGE SEX SOURCE HAEMOGLOBINS ERYTHROCYTE LEUCOCYTE HAEMATOCRIT (
Out[14]:
             0
                      1
                           68
                                 1
                                          0
                                                        11.8
                                                                                   6.3
                                                                                                 35.1
                                                                      4.65
                      2
                                          0
             1
                           70
                                 1
                                                        14.8
                                                                      5.39
                                                                                  12.7
                                                                                                 43.5
```

	USER_ID	AGE	SEX	SOURCE	HAEMOGLOBINS	ERYTHROCYTE	LEUCOCYTE	HAEMATOCRIT	(
2	3	64	1	0	11.3	4.74	13.2	33.5	
3	4	73	1	0	13.7	4.98	10.5	39.1	
4	5	66	0	0	9.9	4.23	22.1	30.9	
•••									
1184	1185	90	1	0	12.5	4.41	10.9	37.1	
1185	1186	90	1	0	10.6	4.44	9.4	33.7	
1186	1187	90	1	1	11.2	4.06	8.4	35.4	
1187	1188	25	0	1	13.8	4.88	3.5	40.0	
1188	1189	31	0	1	13.8	4.86	2.3	39.4	

1189 rows × 10 columns

```
In [15]:
    import numpy as np
    import seaborn as sns
    from matplotlib import pyplot as plt
    import warnings
    import importlib
    warnings.filterwarnings('ignore')
    sns.set_theme(color_codes=True)

In [16]:
    def check_and_install_library(library_name):
        try:
        importlib.import_module(library_name)
```

```
def check_and_install_library(library_name):
    try:
        importlib.import_module(library_name)
        print(f"{library_name} is already installed.")

except ImportError:
    print(f"{library_name} is not installed. Installing...")

    try:
        import pip
        pip.main(['install', library_name])
    except:
        print("Error: Failed to install the library. Please install it manually.
```

```
In [17]: #!pip install scikit-surprise
```

In [18]: !pip show scikit-surprise

```
Name: scikit-surprise

Version: 1.1.3

Summary: An easy-to-use library for recommender systems.

Home-page: https://surpriselib.com

Author: Nicolas Hug

Author-email: contact@nicolas-hug.com

License: GPLv3+
```

Location: c:\users\reshma_koduri\anaconda3\lib\site-packages
Requires: joblib, numpy, scipy

Required-by:

```
In [19]:
          import sys
          print(sys.path)
          import svs
          sys.path.append('/path/to/directory')
         ['C:\\Users\\reshma_koduri', 'C:\\Users\\reshma_koduri\\anaconda3\\python39.zip',
         b', 'C:\\Users\\reshma_koduri\\anaconda3', '', 'C:\\Users\\reshma_koduri\\anaconda3
         \\lib\\site-packages', 'C:\\Users\\reshma_koduri\\anaconda3\\lib\\site-packages\\lock
         et-0.2.1-py3.9.egg', 'C:\\Users\\reshma_koduri\\anaconda3\\lib\\site-packages\\win3
         2', 'C:\\Users\\reshma_koduri\\anaconda3\\lib\\site-packages\\win32\\lib', 'C:\\Users
         \\reshma_koduri\\anaconda3\\lib\\site-packages\\Pythonwin', 'C:\\Users\\reshma_koduri
         \\anaconda3\\lib\\site-packages\\IPython\\extensions', 'C:\\Users\\reshma_koduri\\.ip
         ython']
In [20]:
          from surprise import Dataset, Reader
          from surprise import KNNBasic, accuracy,SVD
          from surprise.model_selection import train_test_split
In [21]:
          #df = pd.DataFrame(treatment)
In [22]:
          # Assuming you have a DataFrame named 'data' with columns 'User', 'Treatment', and '
          reader = Reader(rating_scale=(1,3)) # Define the rating scale
          surprise_data = Dataset.load_from_df(treatment[['USER_ID', 'TREATMENT', 'RATINGS']],
In [23]:
          trainset, testset = train_test_split(surprise_data, test_size=0.3,random_state=42)
In [24]:
          model = SVD()
          model.fit(trainset)
         <surprise.prediction algorithms.matrix factorization.SVD at 0x1bc0b61b520>
Out[24]:
In [25]:
          predictions =model.test(testset)
In [26]:
          from surprise.accuracy import rmse
In [27]:
          accuracy = rmse(predictions)
          print(f'RMSE: {accuracy}')
         RMSE: 0.1020
         RMSE: 0.1019821695834463
```