```
In [1]: #importing all required libraries
   import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   import seaborn as sns
   %matplotlib inline
```

In [2]: #Loading the Data movies =pd.read_csv('movies.dat',sep='::',header=None,names=['MovieID','Title' ,'Generes'],engine='python') ratings=pd.read_csv('ratings.dat',sep='::',header=None,names=['userID','MovieI D','Ratings','Timestamp']) users =pd.read_csv('users.dat',sep='::',header=None,names=['userID','Gender','Age','Occupation','Zip-code'])

E:\New folder (2)\lib\site-packages\ipykernel_launcher.py:3: ParserWarning: F alling back to the 'python' engine because the 'c' engine does not support re gex separators (separators > 1 char and different from '\s+' are interpreted as regex); you can avoid this warning by specifying engine='python'.

This is separate from the ipykernel package so we can avoid doing imports u ntil

E:\New folder (2)\lib\site-packages\ipykernel_launcher.py:4: ParserWarning: F alling back to the 'python' engine because the 'c' engine does not support re gex separators (separators > 1 char and different from '\s+' are interpreted as regex); you can avoid this warning by specifying engine='python'.

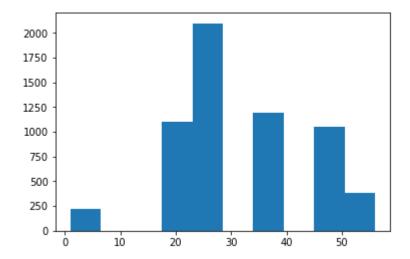
after removing the cwd from sys.path.

```
In [3]: #merging the dataset
    mergeDF=pd.merge(movies,ratings, on='MovieID')
    FinalDF=pd.merge(mergeDF,users, on='userID')
    FinalDF.head()
```

Out[3]:

	MovielD	Title	Generes	userID	Ratings	Timestamp	Gender
0	1	Toy Story (1995)	Animation Children's Comedy	1	5	978824268	F
1	48	Pocahontas (1995)	Animation Children's Musical Romance	1	5	978824351	F
2	150	Apollo 13 (1995)	Drama	1	5	978301777	F
3	260	Star Wars: Episode IV - A New Hope (1977)	Action Adventure Fantasy Sci-Fi	1	4	978300760	F
4	527	Schindler's List (1993)	Drama War	1	5	978824195	F
4							•

```
In [4]: #1.Visualisation on user age
plt.hist(users["Age"])
```

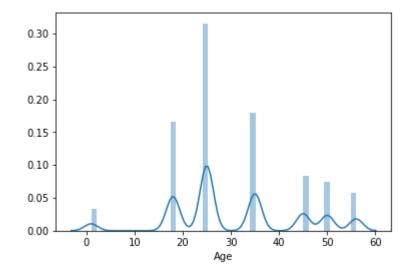


In [5]: #visualisation using seaborn
sns.distplot(users['Age'])

E:\New folder (2)\lib\site-packages\scipy\stats.py:1713: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interprete d as an array index, `arr[np.array(seq)]`, which will result either in an err or or a different result.

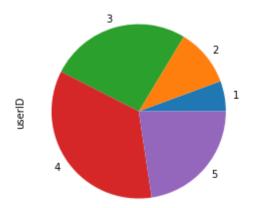
return np.add.reduce(sorted[indexer] * weights, axis=axis) / sumval

Out[5]: <matplotlib.axes. subplots.AxesSubplot at 0x1abbf8552e8>



```
In [6]: #2.Visualizing overall ratings by users
FinalDF.groupby('Ratings')['userID'].count().plot(kind='pie')
```

Out[6]: <matplotlib.axes._subplots.AxesSubplot at 0x1abbf81bc18>

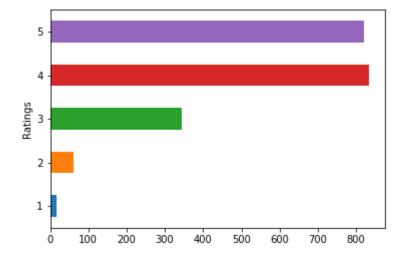


Ratings

- 1 16
- 2 61
- 3 345
- 4 835
- 5 820

Name: userID, dtype: int64

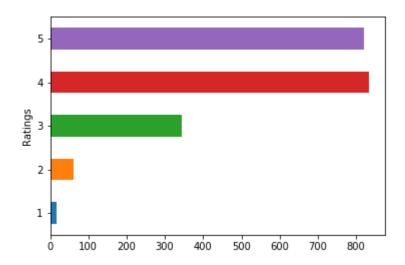
Out[7]: <matplotlib.axes._subplots.AxesSubplot at 0x1abbf8bba20>



```
In [35]: print(FinalDF[FinalDF.Title == "Toy Story (1995)"].groupby("Age")['userID'].co
unt())
FinalDF[FinalDF.Title == "Toy Story (1995)"].groupby("Ratings")['userID'].coun
t().plot(kind='barh')
```

```
Age
      112
1
18
      448
25
      790
35
      423
45
      143
50
      108
       53
56
Name: userID, dtype: int64
```

Out[35]: <matplotlib.axes._subplots.AxesSubplot at 0x185cd33bbe0>

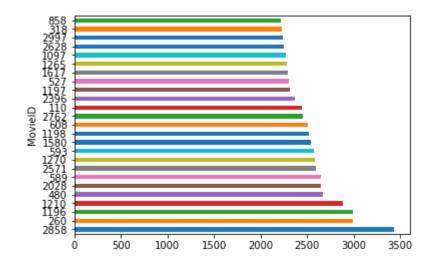


In [8]: #Top 25 movies by viewership rating.
print(FinalDF.groupby('MovieID').Ratings.count().nlargest(25))
FinalDF.groupby('MovieID').Ratings.count().nlargest(25).plot(kind='barh')

MovieI	.D		
2858	3428		
260	2991		
1196	2990		
1210	2883		
480	2672		
2028	2653		
589	2649		
2571	2590		
1270	2583		
593	2578		
1580	2538		
1198	2514		
608	2513		
2762	2459		
110	2443		
2396	2369		
1197	2318		
527	2304		
1617	2288		
1265	2278		
1097	2269		
2628	2250		
2997	2241		
318	2227		
858	2223		
Name:	Ratings.	dtvpe:	int64

Name: Ratings, dtype: int64

Out[8]: <matplotlib.axes._subplots.AxesSubplot at 0x1abbf90a438>



```
In [9]:
         #Find 2696 userid
         FinalDF[FinalDF.userID == 2696].groupby("Ratings")["MovieID"].count()
Out[9]: Ratings
         1
               2
         2
               3
               3
         3
         4
              11
               1
         Name: MovieID, dtype: int64
In [10]:
         FinalDF[FinalDF.userID == 2696].groupby("Ratings")["MovieID"].count().plot(kin
         d='barh')
Out[10]: <matplotlib.axes. subplots.AxesSubplot at 0x1abbf95d0b8>
            5
            4
          Ratings
w
            2
            1
                                                    10
In [11]:
         #500 ML
         mlData=FinalDF.head(500)
         mlData.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 500 entries, 0 to 499
         Data columns (total 10 columns):
         MovieID
                        500 non-null int64
                        500 non-null object
         Title
         Generes
                        500 non-null object
         userID
                        500 non-null int64
                        500 non-null int64
         Ratings
         Timestamp
                        500 non-null int64
         Gender
                        500 non-null object
                        500 non-null int64
         Age
         Occupation
                        500 non-null int64
         Zip-code
                        500 non-null object
         dtypes: int64(6), object(4)
         memory usage: 43.0+ KB
In [12]: features=mlData.iloc[:,[0,7,8]].values
         label = mlData.iloc[:,4].values
```

```
In [13]:
         #Multiple classification
         #K-NN
         mlData.Ratings.unique()
Out[13]: array([5, 4, 3, 2, 1], dtype=int64)
In [23]:
         #train test
         from sklearn.model selection import train test split
         X_train,X_test,Y_train,Y_test = train_test_split(features,label,test_size=0.2,
         random_state=15)
In [24]: | #creating the model
         from sklearn.neighbors import KNeighborsClassifier
         model = KNeighborsClassifier(n neighbors=15)
         model.fit(X_train,Y_train)
Out[24]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
                    metric params=None, n jobs=None, n neighbors=15, p=2,
                    weights='uniform')
In [25]:
         # generalization
         print(model.score(X_train,Y_train))
         print(model.score(X_test,Y_test))
         0.4275
         0.43
In [ ]:
```