```
In [1]: import numpy as np
         import matplotlib.pyplot as plt
         import warnings
         warnings.filterwarnings("ignore")
         %matplotlib inline
         import pandas as pd
         import seaborn as sns
         import matplotlib as mpl
         import seaborn as sns
         import plotly.express as px
In [2]:
         movies = pd.read_csv('Downloads/movies.csv')
           movield
                                        title
                                                                           genres Unnamed: 3
Out[2]:
                               Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy
                                                                                         NaN
                 2
                                                            Adventure|Children|Fantasy
                                                                                         NaN
         1
                                Jumanii (1995)
         2
                 3
                         Grumpier Old Men (1995)
                                                                   Comedy|Romance
                                                                                         NaN
                 4
                         Waiting to Exhale (1995)
                                                             Comedy|Drama|Romance
                                                                                         NaN
                 5 Father of the Bride Part II (1995)
                                                                          Comedy
                                                                                        NaN
In [3]: movies.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 9742 entries, 0 to 9741
         Data columns (total 4 columns):
            Column
                          Non-Null Count Dtype
              movieId
                           9742 non-null
          0
                                             int64
          1
              title
                           9742 non-null
                                             object
                           9742 non-null
                                             object
              genres
              Unnamed: 3 1 non-null
          3
                                             object
         dtypes: int64(1), object(3)
         memory usage: 304.6+ KB
In [4]: ratings = pd.read_csv('Downloads/ratings.csv')
         ratings.head()
           userld movield rating timestamp
Out[4]:
                                964982703
                       3
                            4.0 964981247
         2
                1
                        6
                            4.0 964982224
         3
                       47
                            5.0 964983815
                            5.0 964982931
                       50
In [5]: ratings.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 100836 entries, 0 to 100835
         Data columns (total 4 columns):
          # Column
                       Non-Null Count
                                             Dtype
                          100836 non-null int64
          0
             userTd
          1
              movieId
                          100836 non-null int64
              rating
                          100836 non-null
                                            float64
          3
              timestamp 100836 non-null int64
         dtypes: float64(1), int64(3)
         memory usage: 3.1 MB
         movies.drop(['Unnamed: 3'], axis = 1, inplace = True)
In [6]:
         #gereksiz sütun silme
         # MovieID üzerinden kişilerin yorumlarına film isimlerini ve genrelerini ekliyoruz.
In [7]:
         df = pd.merge(ratings, movies, how='left', on='movieId')
         df.head()
           userld movield rating timestamp
                                                             title
                                                                                               genres
         0
                                964982703
                                                   Toy Story (1995) Adventure|Animation|Children|Comedy|Fantasy
                        3
                            4.0 964981247
                                            Grumpier Old Men (1995)
                                                                                       Comedy|Romance
         2
                1
                        6
                            4.0 964982224
                                                       Heat (1995)
                                                                                      Action|Crime|Thriller
         3
                       47
                            5.0 964983815 Seven (a.k.a. Se7en) (1995)
                                                                                         Mystery|Thriller
                            5.0 964982931 Usual Suspects, The (1995)
                                                                                    Crime|Mystery|Thriller
                       50
        .. . .. ..
```

```
In [8]: | df.describe()
             #Ortalama puan 3.5 ve minimum ve maksimum puan sırasıyla 0.5 ve 5'tir.
             #193609 film için oy vermiş 610 kullanıcı vardır.
                             userld
                                            movield
                                                               rating
                                                                          timestamp
 Out[8]:
             count 100836.000000 100836.000000 100836.000000 1.008360e+05
             mean
                        326 127564
                                       19435 295718
                                                            3 501557 1 205946e+09
               std
                        182.618491
                                       35530.987199
                                                            1.042529 2.162610e+08
                           1.000000
                                           1.000000
                                                            0.500000 8.281246e+08
              min
              25%
                        177.000000
                                        1199 000000
                                                            3 000000 1 019124e+09
              50%
                        325.000000
                                        2991.000000
                                                            3.500000
                                                                       1.186087e+09
              75%
                        477.000000
                                        8122.000000
                                                                       1.435994e+09
                                                            4.000000
                        610.000000 193609.000000
              max
                                                            5.000000 1.537799e+09
             df.isnull().values.any()
                                                            #eksik değer yoktur.temiz
 In [9]:
             False
 Out[9]:
In [10]:
             df1=df.groupby(['title'])[['rating']].sum()
             df2 = pd.DataFrame(df.groupby('title')['rating'].mean())
             df2['rating_oy_sayisi'] = pd.DataFrame(df.groupby('title')['rating'].count())
             df2['total_rating']=df.groupby(['title'])[['rating']].sum()
             high_rated =df2.nlargest(20,'rating')
             high_rated_total=df2.nlargest(20,'total_rating')
             a = df2.nlargest(20, 'rating oy sayisi')
             df2.head()
Out[10]:
                                                      rating rating oy sayisi total rating
                                                title
                                          '71 (2014)
                                                         4.0
                                                                             1
                                                                                         4.0
             'Hellboy': The Seeds of Creation (2004)
                                                         4.0
                                                                                         4.0
                            'Round Midnight (1986)
                                                                             2
                                                                                         7.0
                                 'Salem's Lot (2004)
                                                         5.0
                                                                                         5.0
                          'Til There Was You (1997)
                                                         4.0
                                                                             2
                                                                                         8.0
In [11]:
             plt.figure(figsize=(40,15))
             plt.title('Rating toplamlar1 en çok olan 20 film',fontsize=70)
             colors=['#db7093']
             plt.ylabel('movies title',fontsize=35)
             plt.xticks(fontsize=25, rotation=90)
             plt.xlabel('toplam rating', fontsize=35)
             plt.yticks(fontsize=25)
             plt.barh(high_rated_total.index,high_rated_total['total_rating'],linewidth=3,edgecolor='green',color=colors)
             <BarContainer object of 20 artists>
Out[11]:
                                                                                   Rating toplamları en çok olan 20 film
                                           Seven (a.k.a. Se7en) (1995)
                               Star Wars: Episode VI - Return of the Jedi (1983)
                          Star Wars: Episode VI - Return of the Jean (1900)
Lord of the Rings: The Fellowship of the Ring, The (2001)
Godfather, The (1972)
              Raiders of the Lost Ark (Indiana Jones and the Raiders of the Lost Ark) (1981)

To y Story (1995)

Usual Suspepects, The (1995)

Terminator 2: Judgment Day (1991)

Star Wars: Episode V - The Empire Strikes Back (1980)

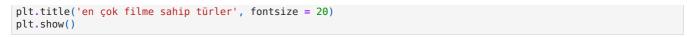
Jurassic Park (1993)

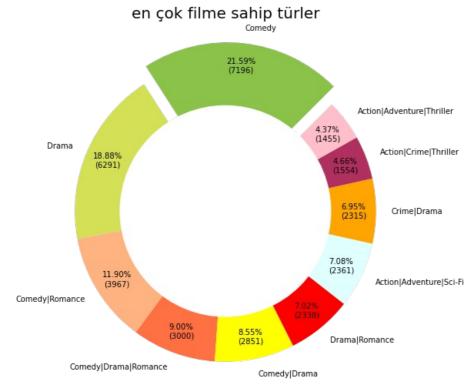
Schindler's List (1993)

Finito Club (1999)
            title
            movies
                                                  Fight Club (1999
                                                 Braveheart (1995
                                  Star Wars: Episode IV - A New Hope (1977
                                        Silence of the Lambs, The (1991
                                                 Pulp Fiction (1994
                                      Forrest Gump (1994)
Shawshank Redemption, The (1994)
                                                                          200
                                                                                                                                                                 1400
                                                                                                            toplam rating
In [12]: plt.figure(figsize=(40,15))
```

```
plt.title('ortalamas1 en yüksek 20 film',fontsize=70)
colors=['#db7093']
plt.ylabel('movies title',fontsize=35)
plt.xticks(fontsize=25,rotation=90)
plt.xlabel('ortalama rating',fontsize=35)
plt.yticks(fontsize=25)
```

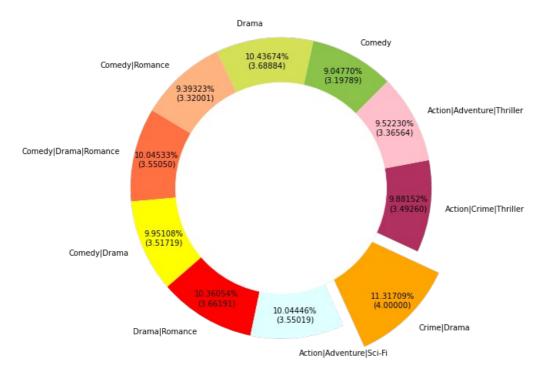
```
plt.barh(high rated.index,high rated['rating'],linewidth=3,edgecolor='white',color=colors)
               #filmlere kaç kullanıcının oy verdiği ile de ilgilenilmelidir. tek kişinin oyladığı filmler
               #sistemde sorun yaratacaktır.
               <BarContainer object of 20 artists>
                                                                                                            ortalaması en yüksek 20 film
                                      American Friend, The (Amerikanische Freund, Der) (1977)
All the Vermeers in New York (1990)
                                                                 All Yours (2016)
                                                          Alien Contamination (1980)
                                            Alesha Popovich and Tugarin the Dragon (2004)
                                                          Advise and Consent (1962)
              ⊕ Adventures Of Sherlock Holmes And Dr. Watson: The Twentieth Century Approaches (1986)
Act of Killing, The (2012)
              movies
                                                   A Perfect Day (2015)
A Filintstones Christmas Carol (1994)
A Detective Story (2003)
9/11 (2002)
7 Faces of Dr. Lao (1964)
61* (2001)
20 Million Miles to Earth (1957)
12 Chairs (1976)
12 Angry Men (1997)
'Salem's Lot (2004)
                                                              A Perfect Day (2015)
                                                                                                                                ortalama rating
In [13]: plt.figure(figsize=(40,15))
               plt.title('en çok oy alan filmlerin ortalaması',fontsize=70)
               colors=['#db7093']
               plt.ylabel('movies title',fontsize=35)
               plt.xticks(fontsize=25, rotation=90)
               plt.xlabel('ortalama rating',fontsize=35)
               plt.yticks(fontsize=25)
               plt.barh(a.index,a['rating'],linewidth=3,edgecolor='white',color=colors)
               <BarContainer object of 20 artists>
Out[13]:
                                                                                                  en çok oy alan filmlerin ortalaması
                Lord of the Rings: The Fellowship of the Ring, The (2001)
Raiders of the Lost Ark (Indiana Jones and the Raiders of the Lost Ark) (1981)
                               k (Indiana Jones and the Raiders of the Lost Ark) (1981)
Apollo 13 (1995)
Independence Day (a.k.a. ID4) (1996)
Seven (a.k.a. Se7en) (1995)
Usual Suspects, The (1995)
American Beauty (1999)
Star Wars: Episode V - The Empire Strikes Back (1980)
              title
                                                         Toy Story (1995)
Fight Club (1999)
                                                      Schindler's List (1993)
                                             Terminator 2: Judgment Day (1991)
                                                         Braveheart (1995)
                                        Jurassic Park (1993)
Star Wars: Episode IV - A New Hope (1977)
                                            Wars: Episode IV - A New Hope (1977)
Matrix, The (1999)
Silence of the Lambs, The (1991)
Pulp Fiction (1994)
Shawshank Redemption, The (1994)
Forrest Gump (1994)
                                                                                                                           ortalama rating
In [14]: df3=df.groupby(['genres'])[['rating']].mean()
               df3['genres_siklik'] = pd.DataFrame(df.groupby('genres')['rating'].count())
               df3.head()
Out[14]:
                                                                rating genres_sıklık
                                                   genres
                                      (no genres listed) 3.489362
                                                                                       47
                                                   Action 2.935484
                                                                                      186
                                      Action|Adventure 3.706306
                                                                                     555
                          Action|Adventure|Animation 3.583333
                                                                                       42
               Action|Adventure|Animation|Children 3.410256
                                                                                       39
In [15]: maks tür=df3.nlargest(10,'genres siklik')
               #En fazla filme sahip olan ilk 10 türün Donut Plot grafiği
In [16]:
               #Elimizdeki verilerde en çok 'Comedy' türünde film vardır.
               plt.figure(figsize = (9,9))
               area = [7196, 6291, 3967, 3000, 2851, 2338, 2361, 2315, 1554, 1455]
               labels = ['Comedy', 'Drama', 'Comedy|Romance', 'Comedy|Drama|Romance', 'Comedy|Drama', 'Drama|Romance', 'Action colors = ['#8BC34A', '#D4E157', '#FFB380', '#FF7043', 'yellow', 'red', '#e0ffff', 'orange', '#b03060', 'pink']
               total = np.sum(area)
               def val_per(x):
                     return '{:.2f}%\n({:.0f})'.format(x, total*x/100)
               plt.pie(area, labels = labels, colors = colors, startangle = 45, pctdistance = 0.85, autopct = val_per, explode
               my_circle = plt.Circle((0, 0), 0.7, color = 'white')
               p = plt.gcf()
               p.gca().add artist(my circle)
```





```
In [17]: #en çok filmi olan türlerden ortalama ratingi en fazla olan Crime|Drama kategorisidir.
            plt.figure(figsize = (9,9))
            area = [3.197888,3.688841, 3.320015,3.550500,3.517187,3.661910,3.550191,4.000000,3.492600,3.365636]
labels = ['Comedy', 'Drama', 'Comedy|Romance', 'Comedy|Drama|Romance', 'Comedy|Drama', 'Drama|Romance', 'Action colors = ['#8BC34A', '#D4E157', '#FFB380', '#FF7043', 'yellow', 'red', '#e0ffff', 'orange', '#b03060', 'pink']
            total = np.sum(area)
            def val_per(x):
                  return '{:.5f}%\n({:.5f})'.format(x, total*x/100)
            plt.pie(area, labels = labels, colors = colors, startangle = 45, pctdistance = 0.85, autopct = val_per, explode
            my_circle = plt.Circle((0, 0), 0.7, color = 'white')
            p = plt.gcf()
            p.gca().add_artist(my_circle)
            plt.title('En çok filme sahip türlerin ortalama ratingleri', fontsize=20)
            plt.show()
```

En çok filme sahip türlerin ortalama ratingleri



```
plt.figure(figsize=(16,8))
           sns.distplot(df2['rating'], kde=True, color='r')
          <AxesSubplot:xlabel='rating', ylabel='Density'>
Out[18]:
             1.0
             0.8
          Density
9.0
             0.4
             0.2
             0.0
                                                                                 ż
                                                                          rating
          movietable = df.pivot table(index='userId',columns='title',values='rating')
In [19]:
          movietable.head()
                                                                                                                                    anoha
Out[19]:
                         'Hellboy':
                                                      'Til
                                                           'Tis the
                                                                                    (500)
                                                                                          *batteries
                                                                                                                             [REC]3
                                                                                                                                       Flo
                             The
                                   'Round 'Salem's
                                                    There
                                                           Season
                                                                   'burbs,
                                                                           'night
                                                                                  Days of
                                                                                                              [REC] [REC]<sup>2</sup>
                                                                                                                                      We S
                                                                                               not
                                                                                                        Zulu
             title
                         Seeds of
                                  Midnight
                                               Lot
                                                     Was
                                                              for
                                                                     The
                                                                          Mother
                                                                                                                     (2009)
                 (2014)
                                                                                  Summer
                                                                                           included
                                                                                                       (2013)
                                                                                                              (2007)
                                                                                                                                     That I
                                             (2004)
                                                                   (1989)
                                    (1986)
                                                                           (1986)
                         Creation
                                                     You
                                                             Love
                                                                                   (2009)
                                                                                             (1987)
                                                                                                                              (2012)
                           (2004)
                                                    (1997)
                                                            (2015)
                                                                                                                                       Mc
                                                                                                                                       (20
          userld
                   NaN
                            NaN
                                      NaN
                                              NaN
                                                     NaN
                                                             NaN
                                                                     NaN
                                                                            NaN
                                                                                     NaN
                                                                                               NaN
                                                                                                         NaN
                                                                                                               NaN
                                                                                                                      NaN
                                                                                                                               NaN
               1
               2
                   NaN
                             NaN
                                      NaN
                                              NaN
                                                     NaN
                                                             NaN
                                                                     NaN
                                                                            NaN
                                                                                     NaN
                                                                                               NaN
                                                                                                         NaN
                                                                                                               NaN
                                                                                                                       NaN
                                                                                                                               NaN
                                                                                                                               NaN
               3
                   NaN
                             NaN
                                      NaN
                                              NaN
                                                     NaN
                                                             NaN
                                                                     NaN
                                                                            NaN
                                                                                     NaN
                                                                                               NaN
                                                                                                         NaN
                                                                                                               NaN
                                                                                                                       NaN
               4
                   NaN
                             NaN
                                      NaN
                                              NaN
                                                     NaN
                                                             NaN
                                                                     NaN
                                                                            NaN
                                                                                     NaN
                                                                                               NaN
                                                                                                         NaN
                                                                                                               NaN
                                                                                                                       NaN
                                                                                                                               NaN
               5
                   NaN
                             NaN
                                      NaN
                                              NaN
                                                     NaN
                                                             NaN
                                                                     NaN
                                                                            NaN
                                                                                     NaN
                                                                                               NaN
                                                                                                         NaN
                                                                                                               NaN
                                                                                                                       NaN
                                                                                                                               NaN
          5 rows × 9719 columns
          type(movietable)
In [20]:
          pandas.core.frame.DataFrame
Out[20]:
          #Forrest Gump filminin oy oranlarına bakıyoruz.
In [21]:
           rating_ = movietable["Forrest Gump (1994)"]
           rating_.head(10)
          userId
                 4.0
          1
          2
                 NaN
          3
                 NaN
          4
                 NaN
          5
                 NaN
          6
                 5.0
                 5.0
          8
                 3.0
          9
                 NaN
          10
                 3.5
          Name: Forrest Gump (1994), dtype: float64
          #Benzerliklerine bakmak için korelasyonlarını hesaplatıyoruz.Bazı kayıtlarda boşluklar olduğu için hata veriyor
In [22]:
          similar= movietable.corrwith(rating )
In [23]:
          # Korelasyonun 1 ve -1 arasında olduğunu görüyoruz, 1 uyumlu -1 uyumsuz olduğunu gösteriyor.
           similar
```

```
Out[23]: title
'71 (2014)
                                                                      NaN
           'Hellboy': The Seeds of Creation (2004)
                                                                      NaN
           'Round Midnight (1986)
                                                                      NaN
           'Salem's Lot (2004)
                                                                      NaN
           'Til There Was You (1997)
                                                                      NaN
                                                                0.011189
           eXistenZ (1999)
           xXx (2002)
                                                                0.306817
           xXx: State of the Union (2005)
                                                                0.383482
                                                                0.449692
           ¡Three Amigos! (1986)
           À nous la liberté (Freedom for Us) (1931)
                                                                      NaN
           Length: 9719, dtype: float64
In [24]: corr_ = pd.DataFrame(similar, columns=['Correlation'])
           corr_.dropna(inplace=True)
corr_.head()
                                      Correlation
Out[24]:
                                 title
                     'burbs, The (1989)
                                        0.197712
            (500) Days of Summer (2009)
                                        0.234095
           *batteries not included (1987)
                                        0.892710
                                        0.928571
             ...And Justice for All (1979)
                   10 Cent Pistol (2015)
                                        -1 000000
           corr_.sort_values('Correlation', ascending=False).head(10)
                                                Correlation
                                           title
                             Lost & Found (1999)
                                                        1.0
                    Century of the Self, The (2002)
                                                        1.0
                             The 5th Wave (2016)
                                                        1.0
                  Play Time (a.k.a. Playtime) (1967)
                                                        1.0
                      Memories (Memorîzu) (1995)
                                                        1.0
                              Playing God (1997)
                                                        1.0
                                   Killers (2010)
                                                        1.0
           Girl Walks Home Alone at Night, A (2014)
                                                        1.0
                                 Tampopo (1985)
                                                        1.0
           Cercle Rouge, Le (Red Circle, The) (1970)
                                                        1.0
In [26]:
           # corr_ dataframe'ine rating_oy_sayisi sütununu ekleyelim
           #oy sayısı düşük olanlar tahmin sistemini yanıltabileceğinden oy adetleri ile de ilgilenilmelidir.
           corr_ = corr_.join(df2['rating_oy_sayisi'])
           corr_.head()
                                      Correlation rating_oy_sayisi
Out[26]:
                                 title
                                                              17
                     'burbs, The (1989)
                                        0.197712
                                                              42
            (500) Days of Summer (2009)
                                        0.234095
           *batteries not included (1987)
                                        0.892710
                                                               7
             ...And Justice for All (1979)
                                        0.928571
                                                               3
                   10 Cent Pistol (2015)
                                        -1.000000
                                                               2
In [27]: corr_[corr_['rating_oy_sayisi']>100].sort_values('Correlation',ascending=False).head()
                                   Correlation rating_oy_sayisi
Out[27]:
                              title
                Forrest Gump (1994)
                                      1.000000
                                                          329
            Good Will Hunting (1997)
                                     0 484042
                                                          141
                     Aladdin (1992)
                                      0.464268
                                                          183
           American History X (1998)
                                      0.457287
                                                          129
                                      0.432556
                                                          125
            Truman Show, The (1998)
In [28]: from tkinter import *
           import tkinter as tk
```

```
from PIL import Image,ImageTk
          from PIL import Image
In [29]: #python için PIL yüklendi.
 In [ ]: from tkinter import *
          import tkinter as tk
          from PIL import Image,ImageTk
          import os
          window = Tk()
           window.title("Test")
          window.geometry("1200x900")
          window.config(bg = '\#b03060')
           lb baslik=Label(text="Film Tavsiye Sistemi",font="Times 22 bold",fg="#b03060",width=50,bg="white")
           lb baslik.place(x=180,y=10)
           lb_baslik=Label(text="Lütfen bir film seçiniz.",font="Times 10 italic", bg= 'white')
           lb_baslik.place(x=30,y=50)
           data=movies['title']
           list=Listbox(window,height="30",width="45",bg="#dddddd",bd="2px",font="SansSerif",fg="black")
           list.place(x=20,y=75)
           label=Label()
           label.place(x=450, y=110)
           resim = ImageTk.PhotoImage(Image.open('/home/ilke/Downloads/pngegg (2).png'))
          a=1
           for i in data:
               list.insert(a,i)
               a=a+1
           lb1=Label(text="")
           lb1.place(x=470,y=125)
           lb=Label(text="")
           lb.place(x=460, y=165)
           #Tavsiye işini yapacak olan metot
           def metot():
               eleman=list.curselection()
               if(eleman):
                    deger=list.get(eleman)
                    lb1.configure(text=deger,font="Times 10 italic")
                    label.configure(text="
                    ratings = movietable[deger]
                    similar = movietable.corrwith(ratings)
                    corr_ = pd.DataFrame(similar, columns=['Correlation'])
corr_.dropna(inplace=True)
                    corr_.sort_values('Correlation',ascending=False)
                    df2.sort_values('rating',ascending=False)
df2.sort_values('rating_oy_sayisi',ascending=False)
                    corr_.sort_values('Correlation',ascending=False)
                    corr_ = corr_.join(df2['rating_oy_sayisi'])
corr_ = corr_.join(df2['rating'])
                    text1=corr_[corr_['rating_oy_sayisi']>100].sort_values('Correlation',ascending=False).head()
text1=text1.rename(columns={\frac{\text{Title': 'Film', 'Correlation': 'Korelasyon'}})
lb.configure(text=text1, font="SansSerif 12",bg = "white",justify = RIGHT)
                    label.configure(text="Lütfen seçim yapınız.")
           b1=Button(text="Tavsiyeleri Göster",command=metot,bg="#F0681F",fg="pink")
           button = Button(window, image= resim)
          b1.place(x=450, y=75)
          button.place(x=80, y=600)
          window.mainloop()
```