```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 from wordcloud import WordCloud
4 from sklearn.feature_extraction.text import TfidfVectorizer
5 from sklearn.metrics.pairwise import linear kernel
1 movies=pd.read_csv('/content/movies.csv')
2 ratings=pd.read_csv('/content/ratings.csv')
1 movies.info()
2
   <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 10329 entries, 0 to 10328
    Data columns (total 3 columns):
         Column
                 Non-Null Count Dtype
         movieId 10329 non-null int64
     0
     1
         title 10329 non-null object
     2
                  10329 non-null object
         genres
    dtypes: int64(1), object(2)
    memory usage: 242.2+ KB
1 ratings.info()
→▼ <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 105339 entries, 0 to 105338
    Data columns (total 4 columns):
         Column
                    Non-Null Count
                                      Dtype
         ____
                    -----
    _ _ _
                                      ____
     0
         userId
                    105339 non-null int64
         movieId
                    105339 non-null int64
     1
     2
         rating
                    105339 non-null float64
         timestamp 105339 non-null int64
    dtypes: float64(1), int64(3)
    memory usage: 3.2 MB
1 movies.shape
\rightarrow \overline{\phantom{a}} (10329, 3)
1 genres=[]
2 for genre in movies.genres:
3
4
      x=genre.split('|')
5
      for i in x:
```

```
if i not in genres:
6
 7
               genres.append(str(i))
 8 genres=str(genres)
 9 movie title=[]
10 for title in movies.title:
      movie_title.append(title[0:-7])
12 movie_title=str(movie_title)
 1 wordcloud_genre=WordCloud(width=1500,height=800,background_color='black',min_font_size=2
 2
                       ,min_word_length=3).generate(genres)
 3 wordcloud_title=WordCloud(width=1500,height=800,background_color='cyan',min_font_size=2
                       ,min_word_length=3).generate(movie_title)
 1 plt.figure(figsize=(30,10))
 2 plt.axis('off')
 3 plt.title('WORDCLOUD for Movies Genre',fontsize=30)
4 plt.imshow(wordcloud_genre)
```

<matplotlib.image.AxesImage at 0x7c32b075c160>

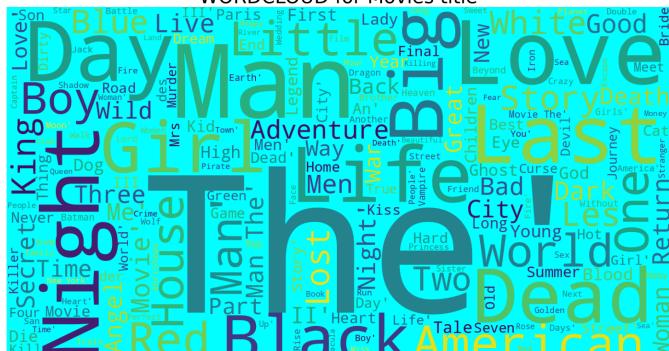


```
1 plt.figure(figsize=(30,10))
2 plt.axis('off')
3 plt.title('WORDCLOUD for Movies title',fontsize=30)
4 plt.imshow(wordcloud_title)
```



<matplotlib.image.AxesImage at 0x7c32af9aacb0>

## WORDCLOUD for Movies title



1 df=pd.merge(ratings,movies, how='left',on='movieId')
2 df.head()

<b>→</b>		userId	movieId	rating	timestamp	title	genres
	0	1	16	4.0	1217897793	Casino (1995)	Crime Drama
	1	1	24	1.5	1217895807	Powder (1995)	Drama Sci-Fi
	2	1	32	4.0	1217896246	Twelve Monkeys (a.k.a. 12 Monkeys) (1995)	Mystery Sci-Fi Thriller
	3	1	47	4.0	1217896556	Seven (a.k.a. Se7en) (1995)	Mystery Thriller
	4	1	50	4.0	1217896523	Usual Suspects, The (1995)	Crime Mystery Thriller

<sup>1</sup> df1=df.groupby(['title'])[['rating']].sum()

<sup>2</sup> high\_rated=df1.nlargest(20,'rating')

<sup>3</sup> high\_rated.head()



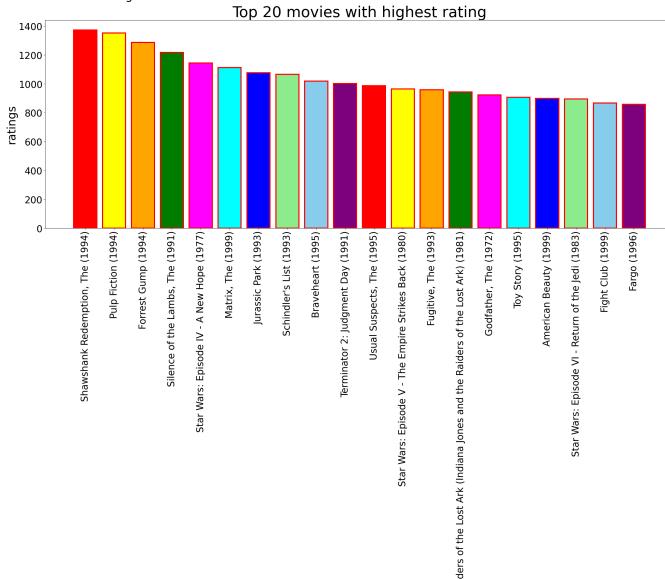
## rating

title		
Shawshank Redemption, The (1994)	1372.0	
Pulp Fiction (1994)	1352.0	
Forrest Gump (1994)	1287.0	
Silence of the Lambs, The (1991)	1216.5	
Star Wars: Episode IV - A New Hope (1977)	1143.5	

```
1 plt.figure(figsize=(30,10))
2 plt.title('Top 20 movies with highest rating',fontsize=40)
3 colors=['red','yellow','orange','green','magenta','cyan','blue','lightgreen','skyblue','r
4 plt.ylabel('ratings',fontsize=30)
5 plt.xticks(fontsize=25,rotation=90)
6 plt.xlabel('movies title',fontsize=30)
7 plt.yticks(fontsize=25)
8 plt.bar(high_rated.index,high_rated['rating'],linewidth=3,edgecolor='red',color=colors)
```

## $\rightarrow$

<BarContainer object of 20 artists>



1 df2=df.groupby('title')[['rating']].count()

2 rating\_count\_20=df2.nlargest(20,'rating')

3 rating\_count\_20.head()



	rating
title	
Pulp Fiction (1994)	325
Forrest Gump (1994)	311
Shawshank Redemption, The (1994)	308
Jurassic Park (1993)	294
Silence of the Lambs, The (1991)	290

<sup>1</sup> plt.figure(figsize=(30,10))

<sup>2</sup> nlt title/'Ton 20 movies with highest number of ratings' fontsize=30\ https://colab.research.google.com/drive/1LWw3T6YLOEQMr3UrtvutxNQhYM7gTkeP#scrollTo=L6Vd7ouvKUfC&printMode=true

```
MovieRecommendationSystem.ipynb - Colab
     creater top to morates water riabilities riamber of rattings profit
3 plt.xticks(fontsize=25,rotation=90)
4 plt.yticks(fontsize=25)
5 plt.xlabel('movies title',fontsize=30)
6 plt.ylabel('ratings',fontsize=30)
8 plt.bar
₹
      matplotlib.pyplot.bar
      def bar(x, height, width=0.8, bottom=None, *, align='center', data=None, **kwargs)
      Make a bar plot.
      The bars are positioned at *x* with the given *align*\ment. Their
      dimensions are given by *height* and *width*. The vertical baseline
      is *bottom* (default 0).
                                   Top 20 movies with highest number of ratings
       1.0
       0.8
     ratings
9.0
       0.2
                                                 movies title
1 cv=TfidfVectorizer()
2 tfidf_matrix=cv.fit_transform(movies['genres'])
1 movie_user = df.pivot_table(index='userId',columns='title',values='rating')
2 movie_user.head()
```

\*batteries

included

(1987)

not

Justi

for A

(197

6997

```
\overline{2}
                                                  'Til
                       'Hellboy':
                                                                           (500)
                        The Seeds
                                       'Round
                                                There
                                                        'burbs,
                                                                  'night
                                                                            Days
                  '71
                                of
       title
                                    Midnight
                                                            The
                                                                 Mother
                                                  Was
                                                                              of
              (2014)
                                       (1986)
                         Creation
                                                  You
                                                         (1989)
                                                                  (1986)
                                                                          Summer
                            (2004)
                                               (1997)
                                                                          (2009)
 1 cosine_sim = linear_kernel(tfidf_matrix, tfidf_matrix)
 1 indices=pd.Series(movies.index,index=movies['title'])
 2 titles=movies['title']
 3 def recommendations(title):
       idx = indices[title]
 4
       sim scores = list(enumerate(cosine_sim[idx]))
 5
       sim_scores = sorted(sim_scores, key=lambda x: x[1], reverse=True)
 6
       sim scores = sim scores[1:21]
 7
 8
       movie_indices = [i[0] for i in sim_scores]
       return titles.iloc[movie indices]
 9
 1 recommendations('Toy Story (1995)')
 2
\rightarrow
                                                      title
      1815
                                                 Antz (1998)
      2496
                                           Toy Story 2 (1999)
      2967
                Adventures of Rocky and Bullwinkle, The (2000)
      3166
                           Emperor's New Groove, The (2000)
      3811
                                        Monsters, Inc. (2001)
      6617
               DuckTales: The Movie - Treasure of the Lost La...
```

Wild, The (2006)