recommendation-system

November 25, 2024

[1]:
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

[2]:
```
data=pd.read_csv(r"/content/anime.csv")
```

[3]:
```
data
```

[3]:

	anime_id	name	genre	type	episodes	rating	members
0	32281	Kimi no Na wa.	Drama, Romance, School, Supernatural	Movie	1	9.37	200630
1	5114	Fullmetal Alchemist: Brotherhood	Action, Adventure, Drama, Fantasy, Magic, Mili…	TV	64		
2	28977	Gintama°	Action, Comedy, Historical, Parody, Samurai, S…	TV	51		
3	9253	Steins;Gate	Sci-Fi, Thriller	TV	24		
4	9969	Gintama'	Action, Comedy, Historical, Parody, Samurai, S…	TV	51		
…	…	…	…	…	…		
12289	9316	Toushindai My Lover: Minami tai Mecha-Minami	Hentai	OVA	1		
12290	5543	Under World	Hentai	OVA	1		
12291	5621	Violence Gekiga David no Hoshi	Hentai	OVA	4		
12292	6133	Violence Gekiga Shin David no Hoshi: Inma Dens…	Hentai	OVA	1		
12293	26081	Yasuji no Pornorama: Yacchimae!!	Hentai	Movie	1		

```
1         9.26   793665
2         9.25   114262
3         9.17   673572
4         9.16   151266
…          …       …
12289     4.15      211
12290     4.28      183
12291     4.88      219
12292     4.98      175
12293     5.46      142

[12294 rows x 7 columns]
```

[4]:
```
data.shape
```

[4]:
```
(12294, 7)
```

[5]:
```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12294 entries, 0 to 12293
Data columns (total 7 columns):
 #   Column    Non-Null Count  Dtype
---  ------    --------------  -----
 0   anime_id  12294 non-null  int64
 1   name      12294 non-null  object
 2   genre     12232 non-null  object
 3   type      12269 non-null  object
 4   episodes  12294 non-null  object
 5   rating    12064 non-null  float64
 6   members   12294 non-null  int64
dtypes: float64(1), int64(2), object(4)
memory usage: 672.5+ KB
```

[6]:
```
data.isnull().sum()
```

[6]:
```
anime_id      0
name          0
genre        62
type         25
episodes      0
rating      230
members       0
dtype: int64
```

[7]:
```
data.dropna(inplace=True)
```

[8]:
```
data.isnull().sum()
```

[8]:
```
anime_id    0
name        0
genre       0
type        0
episodes    0
rating      0
members     0
dtype: int64
```

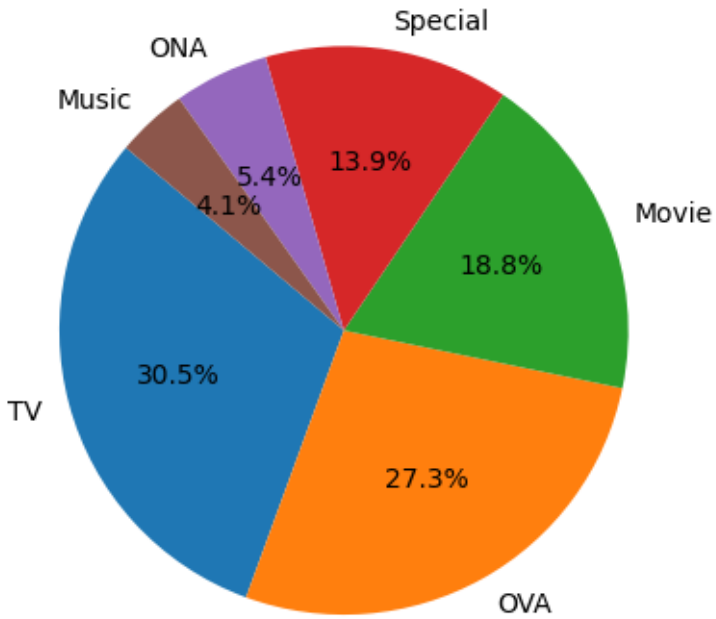
[9]:
```
data.duplicated().sum()
```

[9]:
```
0
```

[10]:
```
data.isnull().sum()
```

[10]:
```
anime_id    0
name        0
genre       0
type        0
episodes    0
rating      0
members     0
dtype: int64
```

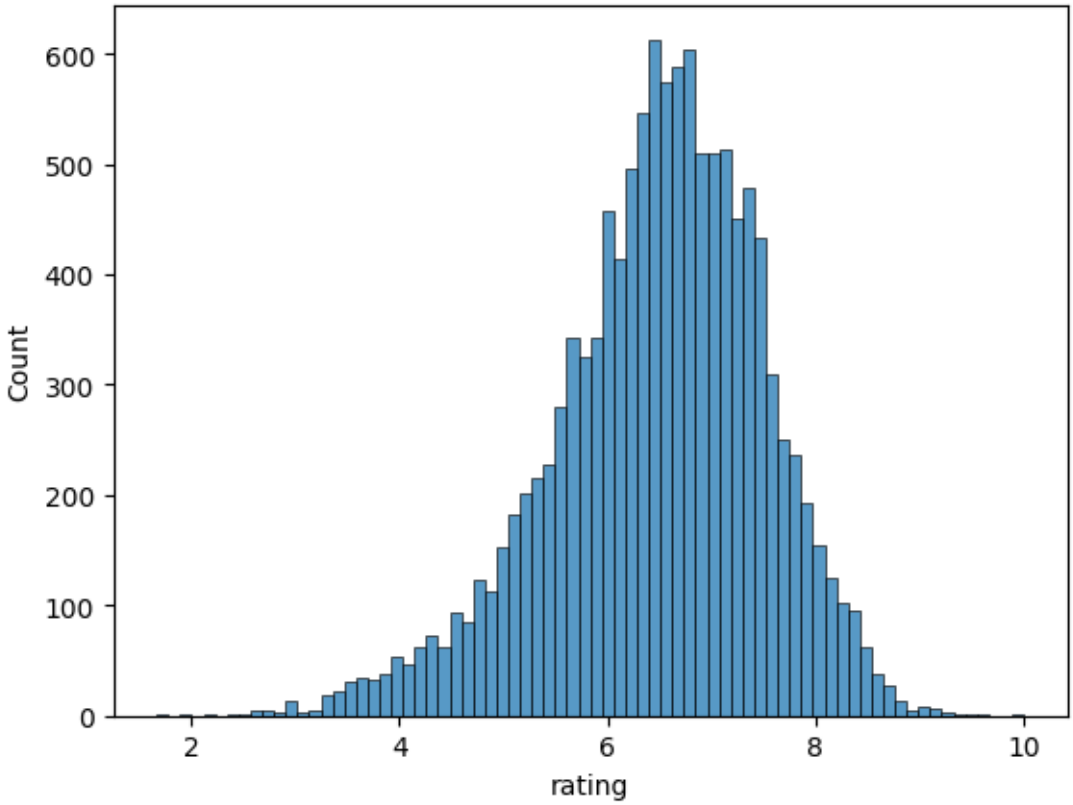
[11]:
```
b=data['type'].value_counts()
b

labels = ['TV', 'OVA', 'Movie', 'Special','ONA','Music']
plt.pie(b, labels=labels, autopct='%1.1f%%', startangle=140)

### checking for different genre

data['genre'] = data['genre'].apply(lambda x: x.split(', '))
```



[12]:
```
data

data['rating'].info()

### lets check the distribution of the rating

sns.histplot(data['rating'])

### lets check the skewness of the plot

data['rating'].skew()

### the data is moderately distributed
```

```
<class 'pandas.core.series.Series'>
Index: 12017 entries, 0 to 12293
Series name: rating
Non-Null Count  Dtype
--------------  -----
12017 non-null  float64
dtypes: float64(1)
memory usage: 187.8 KB
```

[12]: -0.5443140848094782



[13]:
```
data['genre']
exploded_series = data['genre'].explode()
unique_elements = exploded_series.unique()

print(unique_elements)
```

```
['Drama' 'Romance' 'School' 'Supernatural' 'Action' 'Adventure' 'Fantasy'
 'Magic' 'Military' 'Shounen' 'Comedy' 'Historical' 'Parody' 'Samurai'
 'Sci-Fi' 'Thriller' 'Sports' 'Super Power' 'Space' 'Slice of Life'
 'Mecha' 'Music' 'Mystery' 'Seinen' 'Martial Arts' 'Vampire' 'Shoujo'
 'Horror' 'Police' 'Psychological' 'Demons' 'Ecchi' 'Josei' 'Shounen Ai'
 'Game' 'Dementia' 'Harem' 'Cars' 'Kids' 'Shoujo Ai' 'Hentai' 'Yaoi'
 'Yuri']
```

[16]:
```
import pandas as pd
from wordcloud import WordCloud
import matplotlib.pyplot as plt
```

```
# Convert NaN values to empty strings
data['genre'] = data['genre'].fillna('')

# Concatenate all genres into a single string
all_genres = ' '.join(data['genre'].astype(str))

# Generate a word cloud
wordcloud = WordCloud(width=800, height=400, background_color='white').
 ↪generate(all_genres)

# Display the word cloud
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.show()
```

[18]:
```
import pandas as pd
from sklearn.metrics.pairwise import cosine_similarity

# Replace with the actual path to your anime.csv file
anime_data = pd.read_csv("anime.csv")  # If the file is in the same directory␣
 ↪as the script
# or
# anime_data = pd.read_csv("/home/user/data/anime.csv") # If the file is in a␣
 ↪different directory, provide the full path

anime_data['genre'] = anime_data['genre'].fillna('')
anime_data
```

[18]:
```
       anime_id                                              name  \
0         32281                                     Kimi no Na wa.
1          5114                  Fullmetal Alchemist: Brotherhood
2         28977                                          Gintama°
3          9253                                        Steins;Gate
4          9969                                      Gintama&#039;
…           …                                                 …
12289      9316       Toushindai My Lover: Minami tai Mecha-Minami
12290      5543                                         Under World
12291      5621                   Violence Gekiga David no Hoshi
12292      6133  Violence Gekiga Shin David no Hoshi: Inma Dens…
12293     26081                  Yasuji no Pornorama: Yacchimae!!

                                                  genre   type episodes  \
0                  Drama, Romance, School, Supernatural  Movie        1
1      Action, Adventure, Drama, Fantasy, Magic, Mili…     TV       64
2      Action, Comedy, Historical, Parody, Samurai, S…     TV       51
3                                  Sci-Fi, Thriller     TV       24
4      Action, Comedy, Historical, Parody, Samurai, S…     TV       51
…                                                  …    …       …
12289                                         Hentai    OVA        1
12290                                         Hentai    OVA        1
12291                                         Hentai    OVA        4
12292                                         Hentai    OVA        1
12293                                         Hentai  Movie        1

       rating  members
0        9.37   200630
1        9.26   793665
2        9.25   114262
3        9.17   673572
4        9.16   151266
…         …       …
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12291    4.88      219
12292    4.98      175
12293    5.46      142

[12294 rows x 7 columns]
```

[19]:
```
genres = set(genre for sublist in anime_data['genre'] for genre in sublist)
for genre in genres:
    anime_data[genre] = anime_data['genre'].apply(lambda x: 1 if genre in x␣
 ↪else 0)
```

```
 #Drop unnecessary columns
anime_data.drop(['anime_id', 'name', 'genre', 'type', 'episodes', 'rating',␣
 ↪'members'], axis=1, inplace=True)

# Calculate cosine similarity between items (anime)
item_similarity = cosine_similarity(anime_data)

# Convert the cosine similarity matrix into a DataFrame
item_similarity_df = pd.DataFrame(item_similarity, index=anime_data.index,␣
 ↪columns=anime_data.index)

def get_similar_anime(anime_id, top_n=5):
    # Get similarity scores for the given anime
    similar_anime = item_similarity_df.loc[anime_id].
 ↪sort_values(ascending=False)[1:top_n+1]
    return similar_anime

# Example usage:
similar_anime = get_similar_anime(anime_id=60, top_n=5)
print(similar_anime)
```

```
3089    0.973329
3544    0.973329
4418    0.973329
5805    0.971825
0       0.971825
Name: 60, dtype: float64
```

[ ]: