

Netflix Movie Analysis

```
In [1]: # importing lib.
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: df = pd.read_csv('mymoviedb.csv', lineterminator='\n')
df.head()
```

Out[2]:

		Release_Date	Title	Overview	Popularity	Vote_Count	Vote_Average	Original_
0	2021-12-15	Spider-Man: No Way Home	Peter Parker is unmasked and no longer able to...	5083.954	8940	8.3		
1	2022-03-01	The Batman	In his second year of fighting crime, Batman u...	3827.658	1151	8.1		
2	2022-02-25	No Exit	Stranded at a rest stop in the mountains durin...	2618.087	122	6.3		
3	2021-11-24	Encanto	The tale of an extraordinary family, the Madri...	2402.201	5076	7.7		
4	2021-12-22	The King's Man	As a collection of history's worst tyrants and...	1895.511	1793	7.0		



```
In [3]: # viewing dataset info
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9827 entries, 0 to 9826
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Release_Date     9827 non-null    object  
 1   Title            9827 non-null    object  
 2   Overview         9827 non-null    object  
 3   Popularity       9827 non-null    float64 
 4   Vote_Count       9827 non-null    int64  
 5   Vote_Average     9827 non-null    float64 
 6   Original_Language 9827 non-null    object  
 7   Genre            9827 non-null    object  
 8   Poster_Url        9827 non-null    object  
dtypes: float64(2), int64(1), object(6)
memory usage: 691.1+ KB
```

- looks like our dataset has no NaNs!
- Overview, Original_Language and Poster-Url wouldn't be so useful during analysis
- Release_Date column needs to be casted into date time and to extract only the year value

```
In [8]: # exploring genres column
df['Genre'].head()
```

```
Out[8]: 0    Action, Adventure, Science Fiction
        1          Crime, Mystery, Thriller
        2                  Thriller
        3    Animation, Comedy, Family, Fantasy
        4    Action, Adventure, Thriller, War
Name: Genre, dtype: object
```

- genres are separated by commas followed by whitespaces.

```
In [11]: # check for duplicated rows
df.duplicated().sum()
```

```
Out[11]: 0
```

- our dataset has no duplicated rows either.

```
In [15]: # exploring summary statistics
df.describe()
```

```
Out[15]:
```

	Popularity	Vote_Count	Vote_Average
count	9827.000000	9827.000000	9827.000000
mean	40.326088	1392.805536	6.439534
std	108.873998	2611.206907	1.129759
min	13.354000	0.000000	0.000000
25%	16.128500	146.000000	5.900000
50%	21.199000	444.000000	6.500000
75%	35.191500	1376.000000	7.100000
max	5083.954000	31077.000000	10.000000

```
In [ ]:
```

- Exploration Summary
- we have a dataframe consisting of 9827 rows and 9 columns.
- our dataset looks a bit tidy with no NaNs nor duplicated values.
- Release_Date column needs to be casted into date time and to extract only the year
- Overview, Original_Languege and Poster-Url wouldn't be so useful during analysis
- there is noticeable outliers in Popularity column
- Vote_Average better be categorised for proper analysis.
- Genre column has comma saperated values and white spaces that needs to be handled

```
In [18]: # Data Cleaning
```

Casting Release_Date column and extracting year values

```
In [21]: df.head()
```

Out[21]:

	Release_Date	Title	Overview	Popularity	Vote_Count	Vote_Average	Original_
0	2021-12-15	Spider-Man: No Way Home	Peter Parker is unmasked and no longer able to...	5083.954	8940	8.3	
1	2022-03-01	The Batman	In his second year of fighting crime, Batman u...	3827.658	1151	8.1	
2	2022-02-25	No Exit	Stranded at a rest stop in the mountains durin...	2618.087	122	6.3	
3	2021-11-24	Encanto	The tale of an extraordinary family, the Madri...	2402.201	5076	7.7	
4	2021-12-22	The King's Man	As a collection of history's worst tyrants and...	1895.511	1793	7.0	



In [23]:

```
# casting column a
df['Release_Date'] = pd.to_datetime(df['Release_Date'])

# confirming changes
print(df['Release_Date'].dtypes)
```

datetime64[ns]

In [25]:

```
df['Release_Date'] = df['Release_Date'].dt.year
df['Release_Date'].dtypes
```

Out[25]:

dtype('int32')

In [27]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9827 entries, 0 to 9826
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Release_Date     9827 non-null    int32  
 1   Title            9827 non-null    object  
 2   Overview          9827 non-null    object  
 3   Popularity        9827 non-null    float64 
 4   Vote_Count        9827 non-null    int64  
 5   Vote_Average      9827 non-null    float64 
 6   Original_Language 9827 non-null    object  
 7   Genre             9827 non-null    object  
 8   Poster_Url         9827 non-null    object  
dtypes: float64(2), int32(1), int64(1), object(5)
memory usage: 652.7+ KB
```

In [29]: `df.head()`

		Release_Date	Title	Overview	Popularity	Vote_Count	Vote_Average	Original_
0	2021	Spider-Man: No Way Home	Peter Parker is unmasked and no longer able to...	5083.954	8940	8.3		
1	2022	The Batman	In his second year of fighting crime, Batman u...	3827.658	1151	8.1		
2	2022	No Exit	Stranded at a rest stop in the mountains durin...	2618.087	122	6.3		
3	2021	Encanto	The tale of an extraordinary family, the Madri...	2402.201	5076	7.7		
4	2021	The King's Man	As a collection of history's worst tyrants and...	1895.511	1793	7.0		

In [32]: `# making list of column to be dropped
cols = ['Overview', 'Original_Language', 'Poster_Url']`

```
# dropping columns and confirming changes
df.drop(cols, axis = 1, inplace = True)
df.columns
```

```
Out[32]: Index(['Release_Date', 'Title', 'Popularity', 'Vote_Count', 'Vote_Average',
               'Genre'],
               dtype='object')
```

```
In [34]: df.head()
```

```
Out[34]:
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
0	2021	Spider-Man: No Way Home	5083.954	8940	8.3	Action, Adventure, Science Fiction
1	2022	The Batman	3827.658	1151	8.1	Crime, Mystery, Thriller
2	2022	No Exit	2618.087	122	6.3	Thriller
3	2021	Encanto	2402.201	5076	7.7	Animation, Comedy, Family, Fantasy
4	2021	The King's Man	1895.511	1793	7.0	Action, Adventure, Thriller, War

categorizing Vote_Average column

We would cut the `Vote_Average` values and make 4 categories: `popular` `average` `below_avg` `not_popular` to describe it more using `catgorize_col()` function provided above.

```
In [37]: def catgorize_col (df, col, labels):
    """
    catgorizes a certain column based on its quartiles

    Args:
        (df)      df      - dataframe we are proccesing
        (col)     str     - to be catgorized column's name
        (labels)  list   - list of labels from min to max

    Returns:
        (df)      df      - dataframe with the categorized col
    """

    # setting the edges to cut the column accordingly
    edges = [df[col].describe()['min'],
             df[col].describe()['25%'],
             df[col].describe()['50%'],
             df[col].describe()['75%'],
             df[col].describe()['max']]
```

```
df[col] = pd.cut(df[col], edges, labels = labels, duplicates='drop')
return df
```

```
In [39]: # define labels for edges
labels = ['not_popular', 'below_avg', 'average', 'popular']

# categorize column based on Labels and edges
catgorize_col(df, 'Vote_Average', labels)

# confirming changes
df['Vote_Average'].unique()
```

```
Out[39]: ['popular', 'below_avg', 'average', 'not_popular', NaN]
Categories (4, object): ['not_popular' < 'below_avg' < 'average' < 'popular']
```

```
In [41]: df.head()
```

```
Out[41]:
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
0	2021	Spider-Man: No Way Home	5083.954	8940	popular	Action, Adventure, Science Fiction
1	2022	The Batman	3827.658	1151	popular	Crime, Mystery, Thriller
2	2022	No Exit	2618.087	122	below_avg	Thriller
3	2021	Encanto	2402.201	5076	popular	Animation, Comedy, Family, Fantasy
4	2021	The King's Man	1895.511	1793	average	Action, Adventure, Thriller, War

```
In [43]: # exploring column
df['Vote_Average'].value_counts()
```

```
Out[43]:
```

Vote_Average	count
not_popular	2467
popular	2450
average	2412
below_avg	2398

Name: count, dtype: int64

```
In [45]: # dropping Nans
df.dropna(inplace = True)

# confirming
df.isna().sum()
```

```
Out[45]:
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
	0	0	0	0	0	0

dtype: int64

```
In [47]: df.head()
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
0	2021	Spider-Man: No Way Home	5083.954	8940	popular	Action, Adventure, Science Fiction
1	2022	The Batman	3827.658	1151	popular	Crime, Mystery, Thriller
2	2022	No Exit	2618.087	122	below_avg	Thriller
3	2021	Encanto	2402.201	5076	popular	Animation, Comedy, Family, Fantasy
4	2021	The King's Man	1895.511	1793	average	Action, Adventure, Thriller, War

we'd split genres into a list and then explode our dataframe to have only one genre per row for each movie

```
In [52]: # split the strings into lists
df['Genre'] = df['Genre'].str.split(', ')
# explode the lists
df = df.explode('Genre').reset_index(drop=True)
df.head()
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
0	2021	Spider-Man: No Way Home	5083.954	8940	popular	Action
1	2021	Spider-Man: No Way Home	5083.954	8940	popular	Adventure
2	2021	Spider-Man: No Way Home	5083.954	8940	popular	Science Fiction
3	2022	The Batman	3827.658	1151	popular	Crime
4	2022	The Batman	3827.658	1151	popular	Mystery

```
In [55]: # casting column into category
df['Genre'] = df['Genre'].astype('category')
# confirming changes
df['Genre'].dtypes
```

```
Out[55]: CategoricalDtype(categories=['Action', 'Adventure', 'Animation', 'Comedy', 'Crime',
                                         'Documentary', 'Drama', 'Family', 'Fantasy', 'History',
                                         'Horror', 'Music', 'Mystery', 'Romance', 'Science Fiction',
                                         'TV Movie', 'Thriller', 'War', 'Western'],
                                         , ordered=False, categories_dtype=object)
```

```
In [57]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 25552 entries, 0 to 25551
Data columns (total 6 columns):
 #   Column      Non-Null Count  Dtype  
 ---  --          -----          ----  
 0   Release_Date 25552 non-null   int32  
 1   Title        25552 non-null   object  
 2   Popularity   25552 non-null   float64 
 3   Vote_Count   25552 non-null   int64  
 4   Vote_Average 25552 non-null   category 
 5   Genre        25552 non-null   category  
dtypes: category(2), float64(1), int32(1), int64(1), object(1)
memory usage: 749.6+ KB
```

```
In [59]: df.unique()
```

```
Out[59]: Release_Date    100
          Title         9415
          Popularity    8088
          Vote_Count    3265
          Vote_Average   4
          Genre         19
          dtype: int64
```

Now that our dataset is clean and tidy, we are left with a total of 6 columns and 25551 rows to dig into during our analysis

Data Visualization

here, we'd use Matplotlib and seaborn for making some informative visuals to gain insights about our data.

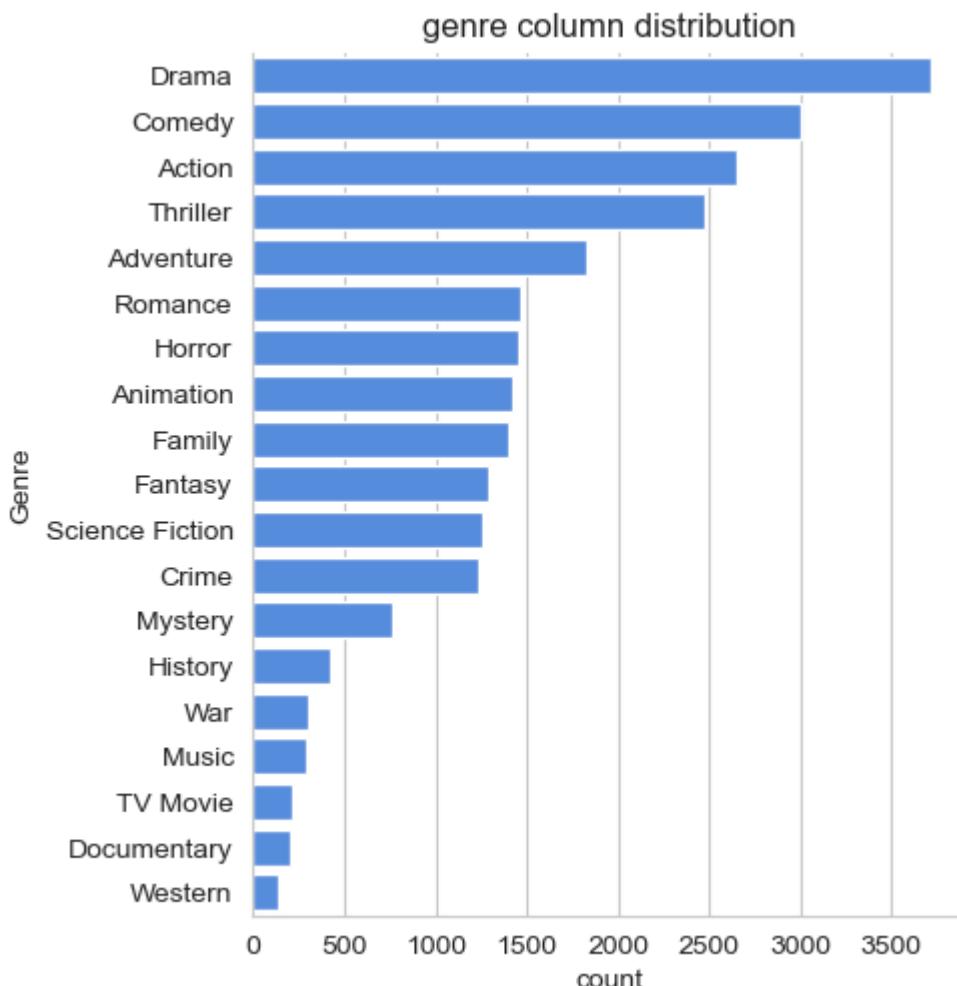
```
In [62]: # setting up seaborn configurations
sns.set_style('whitegrid')
```

Q1: What is the most frequent genre in the dataset?

```
In [65]: # showing stats. on genre column
df['Genre'].describe()
```

```
Out[65]: count    25552
unique      19
top        Drama
freq       3715
Name: Genre, dtype: object
```

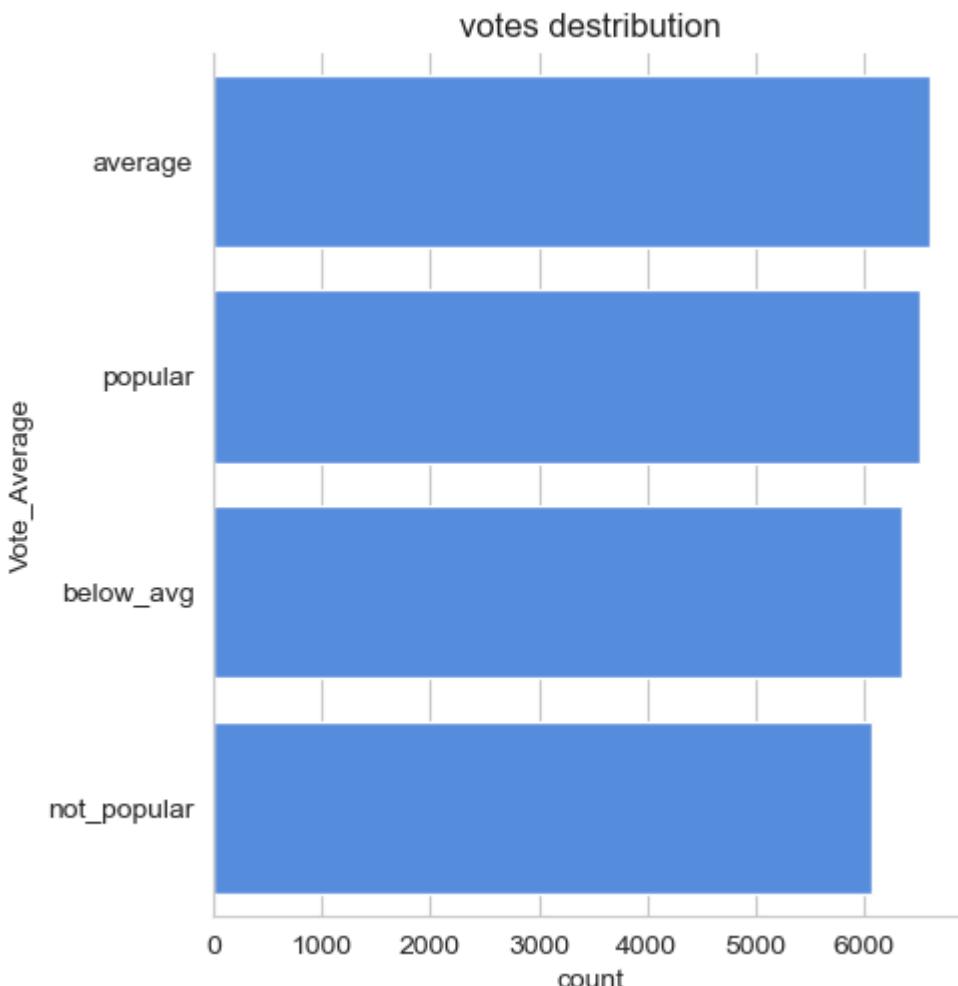
```
In [67]: # visualizing genre column
sns.catplot(y = 'Genre', data = df, kind = 'count',
            order = df['Genre'].value_counts().index,
            color = '#4287f5')
plt.title('genre column distribution')
plt.show()
```



- we can notice from the above visual that `Drama` genre is the most frequent genre in our dataset and has appeared more than 14% of the times among 19 other genres.

Q2: What genres has highest votes ?

```
In [71]: # visualizing vote_average column
sns.catplot(y = 'Vote_Average', data = df, kind = 'count',
            order = df['Vote_Average'].value_counts().index,
            color = '#4287f5')
plt.title('votes distribution')
plt.show()
```



Q3: What movie got the highest popularity? what's its genre?

```
In [74]: # checking max popularity in dataset
df[df['Popularity'] == df['Popularity'].max()]
```

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
0	2021	Spider-Man: No Way Home	5083.954	8940	popular	Action
1	2021	Spider-Man: No Way Home	5083.954	8940	popular	Adventure
2	2021	Spider-Man: No Way Home	5083.954	8940	popular	Science Fiction

Q4: What movie got the lowest popularity? what's its genre?

```
In [86]: # checking max popularity in dataset
df[df['Popularity'] == df['Popularity'].min()]
```

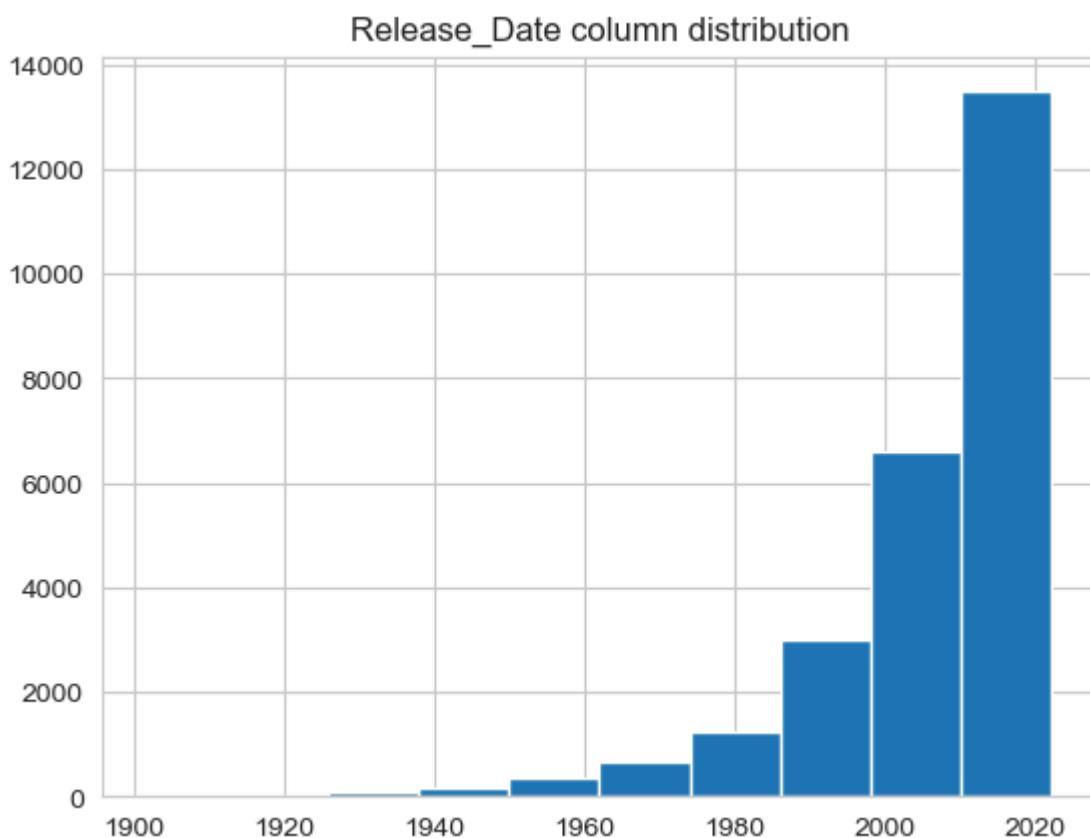
Out[86]:

	Release_Date	Title	Popularity	Vote_Count	Vote_Average	Genre
25546	2021	The United States vs. Billie Holiday	13.354	152	average	Music
25547	2021	The United States vs. Billie Holiday	13.354	152	average	Drama
25548	2021	The United States vs. Billie Holiday	13.354	152	average	History
25549	1984	Threads	13.354	186	popular	War
25550	1984	Threads	13.354	186	popular	Drama
25551	1984	Threads	13.354	186	popular	Science Fiction

Q5: Which year has the most filmed movies?

In [82]:

```
df['Release_Date'].hist()
plt.title('Release_Date column distribution')
plt.show()
```



Conclusion

Q1: What is the most frequent genre in the dataset?

Drama genre is the most frequent genre in our dataset and has appeared more than 14% of the times among 19 other genres.

Q2: What genres has highest votes ?

we have 25.5% of our dataset with popular vote (6520 rows). Drama again gets the highest popularity among fans by being having more than 18.5% of movies popularities.

Q3: What movie got the highest popularity ? what's its genre ?

Spider-Man: No Way Home has the highest popularity rate in our dataset and it has genres of Action , Adventure and Sience Fiction .

Q3: What movie got the lowest popularity ? what's its genre ?

The united states, thread' has the highest lowest rate in our dataset and it has genres of music , drama , 'war', 'sci-fi' and history'.

Q4: Which year has the most filammed movies?

year 2020 has the highest filmming rate in our dataset.

In []: