# **IMDB Movie Data Analysis**

```
In [ ]: |
        import numpy as np
        import matplotlib.pyplot as plt
         import seaborn as sns
         import pandas as pd
        import warnings
        warnings.filterwarnings("ignore")
In [ ]: | data=pd.read csv('datasets/movies.csv')
        data exploration
In [ ]:
        data.head()
        data.info()
In [ ]:
        remvoing columns we dont need
In [ ]: | data.drop(['Poster', 'Movie URL'], axis=1, inplace=True)
        checking for null values
In [ ]:
        data.isnull().sum()
In [ ]:
        data=data.dropna()
        checking for duplicates
        dupli=data.duplicated()
In [ ]:
        sum(dupli)
        there are no duplicates
In [ ]:
        data.info()
```

### removing outliers

## **Analysis**

#### Number of movies released over the years

```
In [ ]: sns.lineplot(x=g.index,y='Title',data=g)
   plt.ylabel('No. of movies released')
   plt.title('Number of movies released over the years')
```

#### IMDB rating over the years

```
In [ ]: sns.lineplot(x=g.index,y='IMDB Rating',data=g)
plt.title('IMDB Rating over the years')
```

#### rating count over the years

```
In [ ]: sns.lineplot(x=g.index,y='Rating Count',data=g)
   plt.title('number of ratings over the years')
   plt.ylim(0, g['Rating Count'].max() + 100)
   plt.ylabel('rating count in M')
```

#### length of movies over the years

```
In [ ]: sns.lineplot(x=g.index,y='Length in Min',data=g)
```

### correlation analysis

```
In [ ]: plt.figure(figsize=(3,3))
    corr=data.iloc[:,1:5].corr()
    sns.heatmap(corr,cmap='viridis')
```

there isnt much correlation among the variables

#### Genres with highest number of movies

```
In [ ]: genres=data.Genres.str.split("|",expand=True)
    genres.head(1)

In [ ]: melted_df = genres.melt(value_vars=[0, 1, 2], value_name='genre')
    melted_df = melted_df.dropna()
    melted_df
    genre_counts = melted_df['genre'].value_counts().head(10)
    genre_counts.plot(kind='bar',edgecolor='k')
```

#### director with most movies

```
In [ ]: directors=data.Directors.str.split("|",expand=True)
    directors.head(3)
```

creating single list

```
In [ ]: melted_df = directors.melt(value_vars=[0, 1, 2, 3], value_name='director')
    melted_df = melted_df.dropna()
    melted_df = melted_df[melted_df['director'] != '']
```

getting count of directors

```
In [ ]: director_counts = melted_df['director'].value_counts().head(10)
In [ ]: director_counts.plot(kind='bar',edgecolor='k')
    plt.ylim(0, 130)
    plt.title('director with highest number of movies')
```

#### directors with highest mean rating count

```
In [ ]: data['Directors'] = data['Directors'].str.split('|').apply(sorted)
data['Directors'] = data['Directors'].apply(', '.join)
```

```
In [ ]: # Filter directors with at least 10 movies and calculate mean rating count
    director_ratings = data.groupby('Directors').filter(lambda x: x['Title'].count
    director_ratings = director_ratings.groupby('Directors')['Rating Count'].mean(

# Select top 10 directors with highest mean rating count
    top_10_directors = director_ratings.sort_values(ascending=False).head(10)
    top_10_directors.plot(kind='bar',edgecolor='k')
    plt.title('top 10 directors with highest mean rating count')
```

```
top 10 highest rated movies
In [ ]:
        sorted_df = data.sort_values(by='IMDB Rating', ascending=False)
        sorted df=sorted df.head(10)
        sorted df[['Title','Release Year','IMDB Rating']]
In [ ]: data
In [ ]:
        df = data.copy()
        all_stars = df['Stars'].str.split('|',expand=True)
        x=all_stars.iloc[:,:3]
        df_comb= pd.concat([df[['Release Year', 'Length in Min']], x], axis=1)
        df comb.head(2)
In [ ]: | from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LinearRegression
        from sklearn.metrics import r2 score
        from sklearn.preprocessing import LabelEncoder
        label_encoder = LabelEncoder()
        df comb[0] = label encoder.fit transform(df comb[0])
        df comb[1] = label encoder.fit transform(df comb[1])
        df_comb[2] = label_encoder.fit_transform(df_comb[2])
        df comb
In [ ]: | y = pd.Series(df['IMDB Rating'])
        y.shape
In [ ]:
        X_train, X_test, y_train, y_test = train_test_split(df_comb, y, test_size=0.2,
        model = LinearRegression()
        model.fit(X_train, y_train)
        test_r2 = r2_score(y_test, model.predict(X_test))
```

In [ ]:	
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