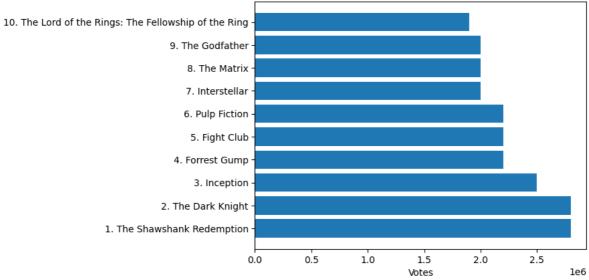
```
In [50]: import requests
                from bs4 import BeautifulSoup
In [51]: user_agent = 'Mozilla/5.0 (Windows NT 10.0; Win64; x64) AppleWebKit/537.36 (KHTML, like Gecko) Chrome/116.0.0.0 Safari/537.36'
                 Link_url = 'https://www.imdb.com/chart/top'
                 response = requests.get(Link_url, headers = {'User-Agent':user_agent,'Accept_language':'en-US, en;q=0.5'})
                print(response)
                 <Response [200]>
In [52]: soup = BeautifulSoup(response.content, 'html.parser')
                movie_data = soup.findAll('li',attrs = {'class':'ipc-metadata-list-summary-item sc-59b6048d-0 jemTre cli-parent'})
                movie_rank = []
                movie_name = []
released_year = []
                duration = []
rated_type = []
                 ratings = []
                 votes = []
                # To convert votes count from Millions and thousands to numbers
In [53]:
                def convert_to_int (vote):
                        if 'm' in vote:
                               return int(float(vote.replace('m','')) * 1000000)
                        elif 'k' in vote:
                              return int(float(vote.replace('k','')) * 1000)
                        else:
                               return int(vote)
                 # Convert duration into minutes
                 def convert_to_min(r_time):
                        total_minutes = 0
                        parts = r time.split()
                        for part in parts:
    if 'h' in part:
                                      total_minutes += int(part.replace('h','')) * 60
                               elif 'g' in part:
                                      total_minutes += int(part.replace('g',''))
                        return total minutes
                 for movie in movie data:
                        # Name and rank
                        rank_name = movie.find('div',class_="ipc-metadata-list-summary-item__c").div.a.text
                        rank = int(rank_name.split('. ')[0])
name = rank_name.split('. ')[1]
                        name = rank_name.split('.
                        movie rank.append(rank)
                        movie name.append(name)
                        year = int(movie.find('div',class_="ipc-metadata-list-summary-item__c").find('span').find_next('span').text)
                        released_year.append(year)
                        # Duration of movie
                        runtime = movie.find('div',class_="ipc-metadata-list-summary-item__c").findAll('span')[2].text
                        runtime = convert_to_minutes(runtime)
                        duration.append(runtime)
                        # Rated Type
                        r_type = movie.find('div',class_="ipc-metadata-list-summary-item_tc").find('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').find_next('span').f
                        rated_type.append(r_type)
                        # Ratings
                        rating = movie.find('div',class_="ipc-metadata-list-summary-item_c").find('span',class_='ipc-rating-star ipc-rating-star--base ipc-rati
                        rating = float(rating)
                        ratings.append(rating)
                        # votes
                        vote = movie.find('div',class_="ipc-metadata-list-summary-item_c").find('span',class_='ipc-rating-star ipc-rating-star--base ipc-rating vote = vote.replace('(','').replace(')','')
                        vote = convert_votes_to_int(vote)
                        votes.append(vote)
In [54]: import pandas as pd
                 import numpy as np
                df = pd.DataFrame({"Rank": movie_rank, 'Movie_Name' : movie_name, 'Duration(Minutes)' : duration, 'Released_Year' : released_year, 'Certificat
                df.head(10)
Out[541:
                     Rank
                                                                          Movie Name Duration(Minutes) Released Year Certification Ratings
                                                                                                                                                                                Votes
                          1
                                                        The Shawshank Redemption
                                                                                                                                    1994
                                                                                                                                                                      9.3 2800000
                                                                                                                                                           Α
                1
                          2
                                                                         The Godfather
                                                                                                                175
                                                                                                                                    1972
                                                                                                                                                           Α
                                                                                                                                                                      9.2 2000000
                2
                          3
                                                                       The Dark Knight
                                                                                                                152
                                                                                                                                    2008
                                                                                                                                                         UA
                                                                                                                                                                      9.0 2800000
                3
                                                                                                                                                                      9.0 1300000
                          4
                                                                The Godfather: Part II
                                                                                                                202
                                                                                                                                    1974
                                                                                                                                                           Α
                4
                          5
                                                                         12 Angry Men
                                                                                                                 96
                                                                                                                                    1957
                                                                                                                                                           U
                                                                                                                                                                      9.0 834000
                5
                          6
                                                                        Schindler's List
                                                                                                                195
                                                                                                                                    1993
                                                                                                                                                           Α
                                                                                                                                                                      9.0 1400000
                                    The Lord of the Rings: The Return of the King
                6
                          7
                                                                                                                201
                                                                                                                                    2003
                                                                                                                                                           U
                                                                                                                                                                      9.0 1900000
                7
                         8
                                                                            Pulp Fiction
                                                                                                                154
                                                                                                                                    1994
                                                                                                                                                           Α
                                                                                                                                                                      8.9 2200000
                          9 The Lord of the Rings: The Fellowship of the Ring
                                                                                                                178
                                                                                                                                    2001
                                                                                                                                                                      8.8 1900000
                9
                         10
                                                         Il Buono, Il Brutto, Il Cattivo
                                                                                                                161
                                                                                                                                    1966
                                                                                                                                                                      8.8 791000
```

```
<class 'pandas.core.frame.DataFrame'>
           RangeIndex: 250 entries, 0 to 249
           Data columns (total 7 columns):
            #
                Column
                                       Non-Null Count Dtype
                                       250 non-null
            0
                Rank
                                                          int64
                Movie Name
                                       250 non-null
                                                          object
                Duration(Minutes) 250 non-null
                                                          int64
                Released_Year
                                       250 non-null
                                                          int64
                Certification
                                       250 non-null
                                                          object
            5
                Ratings
                                       250 non-null
                                                          float64
            6
                Votes
                                      250 non-null
                                                          int64
           dtypes: float64(1), int64(4), object(2) memory usage: 13.8+ KB
In [56]: df.shape
Out[56]: (250, 7)
In [57]: df.size
Out[57]: 1750
In [58]: df.ndim
Out[58]: 2
In [59]: df.describe()
                      Rank Duration(Minutes) Released_Year
                                                                Ratings
                                                 250.000000 250.000000 2.500000e+02
           count 250.000000
                                   250.000000
           mean 125.500000
                                   129.012000
                                               1986.816000 8.307600 6.719320e+05
             std 72.312977
                                    29.756236
                                                  25.387086
                                                              0.232462 5.418230e+05
             min
                   1.000000
                                   45.000000
                                                1921.000000
                                                              8.000000 3.600000e+04
            25% 63 250000
                                   107 250000
                                                1966 250000
                                                              8 100000 2 295000e+05
            50% 125.500000
                                   126.500000
                                                1994.500000
                                                              8.200000 5.370000e+05
            75% 187.750000
                                   145.750000
                                                2007.000000
                                                               8.400000 9.910000e+05
            max 250.000000
                                   238.000000
                                                2023.000000 9.300000 2.800000e+06
In [60]: top_10_movies = df.sort_values(by='Votes', ascending=False).head(10)
           # Add a new column 'Rank' for displaying the rank of each movie
top_10_movies['Rank'] = np.arange(1, len(top_10_movies) + 1)
           # Display the top 10 movies with their rank, name, and votes
print(top_10_movies[['Rank', 'Movie_Name', 'Votes']])
           # Extract Movie Names and Votes
           Movie_Name = top_10_movies['Movie_Name']
           Votes = top_10_movies['Votes']
           Rank = top_10_movies['Rank']
           y_pos = np.arange(len(Movie_Name))
           plt.barh(y_pos[::-1], Votes[::-1], align='center') # Reverse the order
plt.yticks(y_pos[::-1], [f'{r}. {m}' for r, m in zip(Rank[::-1], Movie_Name[::-1])]) # Reverse the order
           plt.xlabel('Votes')
plt.title('Top 10 Movies by Votes')
           plt.show()
                                                                    Movie_Name
           0
                                                    The Shawshank Redemption 2800000
           2
                                                              The Dark Knight
                                                                                  2800000
           13
                   3
                                                                     Inception 2500000
           10
                   4
                                                                  Forrest Gump
                                                                                  2200000
                   5
                                                                    Fight Club
                                                                                  2200000
           11
                                                                  Pulp Fiction
                                                                                  2200000
           22
                                                                  Interstellar
                                                                                  2000000
           15
                                                                    The Matrix
                                                                                  2000000
           1
                                                                The Godfather
                                                                                  2000000
```

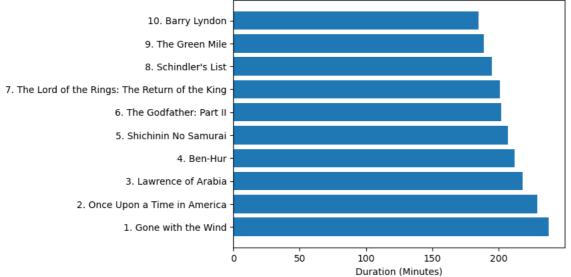
10 The Lord of the Rings: The Fellowship of the Ring 1900000

Top 10 Movies by Votes



```
Gone with the Wind
159
82
        2
                              Once Upon a Time in America
                                                                            229
97
        3
                                        Lawrence of Arabia
                                                                            218
184
        4
                                                   Ben-Hur
                                                                           212
                                     Shichinin No Samurai
21
        5
                                                                           207
3
                                    The Godfather: Part II
                                                                           202
        6
6
           The Lord of the Rings: The Return of the King
                                                                            201
5
                                          Schindler's List
                                                                           195
27
                                            The Green Mile
                                                                            189
186
       10
                                              Barry Lyndon
                                                                            185
```

Top 10 Movies by Duration



```
# Calculate the average ratings for each decade
average_ratings_by_decade = df.groupby('Decade')['Ratings'].mean()

# Plot the average ratings in a line chart
plt.plot(average_ratings_by_decade.index, average_ratings_by_decade.values, marker='o', linestyle='-')

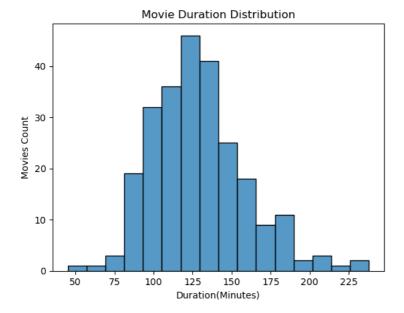
# Set labels and title
plt.xlabel('Decade')
plt.ylabel('Average Ratings')
plt.title('Average Ratings by Decade')

# Show the plot
plt.show()
```

```
8.45 - 8.35 - 8.25 - 8.20 - 1920 1940 1960 1980 2000 2020 Decade
```

```
In [69]: import seaborn as sns
    sns.histplot(x=df['Duration(Minutes)'])
    plt.title('Movie Duration Distribution')
    plt.ylabel('Movies Count')
```

Text(0, 0.5, 'Movies Count')



```
In [73]: from wordcloud import WordCloud
high_rated_movies = df[df['Ratings'] > 8.0]

# Create a string containing movie names and their ratings
movie_ratings_text = ' '.join(f"{name} {rating}" for name, rating in zip(high_rated_movies['Movie_Name'], high_rated_movies['Ratings']))

# Generate the word cloud
wordcloud = WordCloud(width=800, height=400, background_color='white').generate(movie_ratings_text)

# Display the word cloud
plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.show()
```

```
01deuboi
                 Beautiful
                                      Sunseta
Buono
Buono
Tengol
Visua
Nes,
Liem
ξΦ
eng
                  an
                                                  Usual
                                                  Nes
      Mind
   a)
                                                                                       en ⊆
                               Green
                                                                Godfather
          der
                 Amer
                                Das Ark
Inception
        Men
                                                                                           gm
Dar
         Cuckoo
                 olta
Hope
God
       Par
             t
                                                          Episod
                           Whiplash
                                              War
                                                     S
                                                                       le
                                   ather
                                                                                   Fiction
                  Paths
        Di
                                                                         llowship Knight
                                                            Rush
                    M
                                  Max
                                                            Time
                                                     Gladiator
                                                        Φ
                                                                       Dog
                                                             Angr
```

```
In [70]: sns.scatterplot(x=df['Votes'],y=df['Ratings'])
    plt.title('Correlation between Ratings and Votes')
```

Fit a linear regression model
model = LinearRegression()
model.fit(X, df['Votes'])

predicted_votes = model.predict(X)

Calculate the residuals
residuals = df['Votes'] - predicted_votes

Predict 'Votes' using the linear regression model

 $\texttt{Out}[70]\colon$ Text(0.5, 1.0, 'Correlation between Ratings and Votes')

Correlation between Ratings and Votes

```
9.2 - 9.0 - 8.8 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 - 9.0 -
```

```
In [75]: Q1 = df['Votes'].quantile(0.25)
Q3 = df['Votes'].quantile(0.75)
           IQR = Q3 - Q1
           # Define the lower and upper bounds for outliers
           lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR
            # Find outliers
           outliers = df[(df['Votes'] < lower_bound) | (df['Votes'] > upper_bound)]
           print("Outliers:")
           print(outliers)
           Outliers:
                Rank
                                        Movie Name
                                                      Duration(Minutes)
                                                                            Released Year
                       The Shawshank Redemption
                                                                      142
                                                                                       1994
           2
                    3
                                  The Dark Knight
                                                                      152
                                                                                       2008
                   8
                                     Pulp Fiction
                                                                      154
                                                                                       1994
           10
                  11
                                     Forrest Gump
                                                                      142
                                                                                       1994
           11
                  12
                                        Fight Club
                                                                      139
                                                                                       1999
                                                                                       2010
           13
                  14
                                         Inception
                                                                      148
               Certification
                                              Votes
                                Ratings
                                                      Decade
                                           2800000
           2
                            UA
                                     9.0
                                           2800000
                                                        2000
           7
                            Α
                                     8.9
                                           2200000
                                                         1990
           10
                            UΑ
                                     8.8
                                           2200000
                                                         1990
                                           2200000
                                                        1990
           11
                            Α
                                     8.8
           13
                            UA
                                     8.8
                                           2500000
                                                         2010
           \textbf{from} \ \text{sklearn.linear\_model} \ \textbf{import} \ \text{LinearRegression}
In [76]:
           X = df[['Votes']]
```

```
# Identify potential outliers (e.g., points with residuals > 2 standard deviations) outliers = df[np.abs(residuals) > 2 * np.std(residuals)]
                              # Print the potential outliers
                             print("Potential Outliers:")
                             print(outliers)
                             Potential Outliers:
                             Empty DataFrame
                             Columns: [Rank, Movie_Name, Duration(Minutes), Released_Year, Certification, Ratings, Votes, Decade]
In [77]: highest_rating_movie = df[df['Ratings'] == df['Ratings'].max()]
                             # Find the movie with the maximum votes
max_votes_movie = df[df['Votes'] == df['Votes'].max()]
                              # Find the movie with the maximum duration
                             max_duration_movie = df[df['Duration(Minutes)'] == df['Duration(Minutes)'].max()]
                              # Combine all the relevant information
                             combined\_info = pd.concat([highest\_rating\_movie, \ max\_votes\_movie, \ max\_duration\_movie], \ ignore\_index = \texttt{True})
                              # Create a figure to display the information
                              fig, ax = plt.subplots(figsize=(10, 5))
                              # Plot the information
                              for index, row in combined_info.iterrows():
                                         ax.text(0.1, 0.9 - index*0.2, f<sup>*</sup>Movie Name: {row['Movie_Name']}\nRatings: {row['Ratings']}\nVotes: {row['Votes']}\nDuration: {row['Duration: {row['Duration:
                             # Remove the axes for a cleaner look
ax.axis('off')
                             plt.show()
```

Movie Name: The Shawshank Redemption

Ratings: 9.3 Votes: 2800000 Duration: 142 minutes

Movie Name: The Shawshank Redemption

Ratings: 9.3 Votes: 2800000 Duration: 142 minutes

Movie Name: The Dark Knight

Ratings: 9.0 Votes: 2800000 Duration: 152 minutes

Movie Name: Gone with the Wind

Ratings: 8.2 Votes: 327000 Duration: 238 minutes

```
In [79]; highest_rating_movie = df[df['Ratings'] == df['Ratings'].max()]
            # Find the movie with the maximum votes
           max_votes_movie = df[df['Votes'] == df['Votes'].max()]
            # Find the movie with the maximum duration
           max_duration_movie = df[df['Duration(Minutes)'] == df['Duration(Minutes)'].max()]
            # Combine all the relevant information
           combined_info = pd.concat([highest_rating_movie, max_votes_movie, max_duration_movie], ignore_index=True)
            # Create a bar chart to display the information
           fig, ax = plt.subplots(figsize=(10, 5))
            # Plot the information
           ax.barh(combined_info['Movie_Name'], combined_info['Ratings'], color='blue', label='Ratings')
ax.barh(combined_info['Movie_Name'], combined_info['Votes'], color='green', label='Votes')
ax.barh(combined_info['Movie_Name'], combined_info['Duration(Minutes)'], color='red', label='Duration')
            # Add labels and legend
           ax.set_xlabel('Values')
           ax.set_title('Highest votes, rating, and duration movie')
            ax.legend()
           plt.show()
```

