

Scraping IMDB movie data using requests and beatifulsoup

1. Importing libraries/modules

```
In [48]: import numpy as np
import pandas as pd
import requests
from bs4 import BeautifulSoup
import openpyxl
import matplotlib.pyplot as pp
import seaborn as sb
%matplotlib inline
```

2. Scraping the data and saving it as an excel file

```
In [18]: excel = openpyxl.Workbook() #creatng an excel workbook where the data will be saved
sheet = excel.active
sheet.title = "Top Rated Movies" #naming the sheet
sheet.append(['Rank', 'Name', 'Year of Release', 'IMDB Rating']) #naming the column headers

try:
    source = requests.get("https://www.imdb.com/chart/top/")
    source.raise_for_status() #checking if the url is a valid one, and if it is not, it will return an exception

    soup = BeautifulSoup(source.text, 'html.parser') #this collect the html content of the url above and then par

    movies = soup.find('tbody', class_="lister-list").find_all('tr') #finds all the tags with tr. Each tr tag is

    for movie in movies: #creating a loop. for each movie in the movies body, the program should return thefollo

        name = movie.find('td', class_="titleColumn").a.text #this extracts the title of movies into the name var

        rank = movie.find('td', class_="titleColumn").get_text(strip=True).split('.')[0] #first value after the '

        year = movie.find('td', class_="titleColumn").span.text.strip('()') #strip() basically removes anything y

        rating = movie.find('td', class_="ratingColumn imdbRating").strong.text

        sheet.append([rank, name, year, rating]) #saves the values into the excel file

except Exception as e:
    print(e)

excel.save('IMDB Movie Ratings.xlsx')
```

3. Importing the dataset

```
In [120]: data=pd.read_excel('IMDB Movie Ratings.xlsx')
data.head()
```

Out[120]:

| | Rank | Name | Year of Release | IMDB Rating |
|---|------|--------------------------|-----------------|-------------|
| 0 | 1 | The Shawshank Redemption | 1994 | 9.2 |
| 1 | 2 | The Godfather | 1972 | 9.2 |
| 2 | 3 | The Dark Knight | 2008 | 9.0 |
| 3 | 4 | The Godfather Part II | 1974 | 9.0 |
| 4 | 5 | 12 Angry Men | 1957 | 8.9 |

4. Inspecting the data

```
In [121]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 250 entries, 0 to 249
Data columns (total 4 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Rank            250 non-null   int64
1   Name            250 non-null   object
2   Year of Release  250 non-null   int64
3   IMDB Rating     250 non-null   float64
dtypes: float64(1), int64(2), object(1)
memory usage: 7.9+ KB
```

Notes:

- * This data contains records of the top 250 movies on the IMDB website.
- * It has 4 columns and 250 rows
- * It has 2 columns of type (int64), 2 columns each of types (float64) and (object)
- * There are no missing values in the data

5. Performing EDA (Exploratory Data Analysis)

```
In [160]: #total number of release years  
data["Year of Release"].nunique()
```

Out[160]: 86

```
In [161]: #return the number of movies released in each year  
movies_per_releaseYear = pd.DataFrame(data.groupby(["Year of Release"])["Name"].count())  
movies_per_releaseYear.reset_index(inplace=True)  
movies_per_releaseYear.head()
```

Out[161]:

| | Year of Release | Name |
|---|-----------------|------|
| 0 | 1921 | 1 |
| 1 | 1924 | 1 |
| 2 | 1925 | 1 |
| 3 | 1926 | 1 |
| 4 | 1927 | 1 |

```
In [124]: #changing the data type of the year column to string  
data["Year of Release"]=data["Year of Release"].astype(str)  
movies_per_releaseYear["Year of Release"]=movies_per_releaseYear["Year of Release"].astype(str)
```

```
In [125]: #what is the maximum number of movies released?  
movies_per_releaseYear["Name"].max()
```

Out[125]: 8

```
In [126]: #distinct number of movies released  
movies_per_releaseYear["Name"].unique()
```

Out[126]: array([1, 2, 3, 4, 6, 5, 8, 7], dtype=int64)

```
In [127]: #in what year was the maximum number of movies released?
movies_per_releaseYear[movies_per_releaseYear["Name"]==8]
```

Out[127]:

| | Year of Release | Name |
|----|-----------------|------|
| 58 | 1995 | 8 |

```
In [158]: #filtering top 250 movies by release year
data[data["Year of Release"]=="1995"]
```

Out[158]:

| | Rank | Name | Year of Release | IMDB Rating |
|-----|------|--------------------|-----------------|-------------|
| 18 | 19 | Se7en | 1995 | 8.6 |
| 39 | 40 | The Usual Suspects | 1995 | 8.5 |
| 73 | 74 | Braveheart | 1995 | 8.3 |
| 74 | 75 | Toy Story | 1995 | 8.3 |
| 110 | 111 | Heat | 1995 | 8.2 |
| 136 | 137 | Casino | 1995 | 8.2 |
| 179 | 180 | Before Sunrise | 1995 | 8.1 |
| 237 | 238 | La haine | 1995 | 8.0 |

```
In [150]: #get the min and max rating values
print("Min Rating: ", data["IMDB Rating"].min())
print("Max Rating: ", data["IMDB Rating"].max())
```

Min Rating: 8.0
Max Rating: 9.2

In [155]: *#top 250 movies with the lowest rating, their ranks and their release years*
 data[data["IMDB Rating"]==8.0]

Out[155]:

| | Rank | Name | Year of Release | IMDB Rating |
|-----|------|-----------------|-----------------|-------------|
| 210 | 211 | Rocky | 1976 | 8.0 |
| 211 | 212 | Ford v Ferrari | 2019 | 8.0 |
| 212 | 213 | Platoon | 1986 | 8.0 |
| 213 | 214 | Pather Panchali | 1955 | 8.0 |
| 214 | 215 | Stand by Me | 1986 | 8.0 |
| 215 | 216 | The Terminator | 1984 | 8.0 |
| 216 | 217 | Spotlight | 2015 | 8.0 |
| 217 | 218 | Rush | 2013 | 8.0 |
| 218 | 219 | Logan | 2017 | 8.0 |
| 219 | 220 | Network | 1976 | 8.0 |
| 220 | 221 | Ratatouille | 2007 | 8.0 |

In [157]: *#top 250 movies with the highest rating and their release date*
 data[data["IMDB Rating"]==9.2]

Out[157]:

| | Rank | Name | Year of Release | IMDB Rating |
|---|------|--------------------------|-----------------|-------------|
| 0 | 1 | The Shawshank Redemption | 1994 | 9.2 |
| 1 | 2 | The Godfather | 1972 | 9.2 |

6. Summary

1. There are 86 years altogether, from 1921 to 2022
2. from the top 250 movies data, the highest number of movies released is 8, in the year 1995
3. rating ranged from 8.0 to 9.2

4. 40 movies were rated 8.0, while 2 movies were rated 9.2 (The Shawshank Redemption (1994) and The God father (1972), in the order of their ranking)

In []:

In []: