

Top 10000 Movies Analysis

In [34]:

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

In [35]:

```
data=pd.read_csv("popular_10000_movies_tmdb.csv")
data.head()
```

Out[35]:

	id	title	release_date	genres	original_language	vote_average	vote_count
0	758323	The Pope's Exorcist	2023-04-05	['Horror', 'Mystery', 'Thriller']	English	7.4	619
1	640146	Ant-Man and the Wasp: Quantumania	2023-02-15	['Action', 'Adventure', 'Science Fiction']	English	6.6	2294
2	502356	The Super Mario Bros. Movie	2023-04-05	['Animation', 'Adventure', 'Family', 'Fantasy']	English	7.5	1861
3	868759	Ghosted	2023-04-18	['Action', 'Comedy', 'Romance']	English	7.2	652
4	594767	Shazam! Fury of the Gods	2023-03-15	['Action', 'Comedy', 'Fantasy', 'Adventure']	English	6.8	1510

In [36]:

```
data.shape
```

Out[36]:

(10000, 14)

In [37]:

```
data.columns
```

```
Out[37]: Index(['id', 'title', 'release_date', 'genres', 'original_language',
       'vote_average', 'vote_count', 'popularity', 'overview', 'budget',
       'production_companies', 'revenue', 'runtime', 'tagline'],
      dtype='object')
```

Dropping useless columns from dataframe

```
In [38]: data.drop(['id', 'overview', 'tagline'], axis=1, inplace=True)
```

list of final columns

```
In [39]: data.columns
```

```
Out[39]: Index(['title', 'release_date', 'genres', 'original_language', 'vote_average',
       'vote_count', 'popularity', 'budget', 'production_companies', 'revenue',
       'runtime'],
      dtype='object')
```

Checking for null values

```
In [40]: data.isnull().sum()
```

```
Out[40]: title          0
release_date     21
genres           0
original_language 0
vote_average     0
vote_count       0
popularity        0
budget            0
production_companies 0
revenue           0
runtime           0
dtype: int64
```

Dropping rows having null values

```
In [41]: data.dropna(axis=0, inplace=True)
```

Unique lang list

```
In [42]: data['original_language'].unique()
```

```
Out[42]: array(['English', 'French', 'Dutch', 'Spanish', 'Korean', 'Japanese',
       'Finnish', 'Ukrainian', 'Norwegian', 'Estonian', 'cn', 'Polish',
       'Russian', 'German', 'Chinese', 'Italian', 'Basque', 'Thai',
       'Turkish', 'Swedish', 'Icelandic', 'Tagalog', 'Bengali', 'Arabic',
       'Tamil', 'Telugu', 'Romanian', 'Indonesian', 'Galician', 'Danish',
       'Macedonian', 'Portuguese', 'Vietnamese', 'Catalan', 'Hindi',
       'Persian', 'Hebrew', 'Serbian', 'Malayalam', 'Greek', 'Hungarian',
       'Czech', 'Norwegian Bokmal', 'xx', 'Kannada', 'Irish', 'Khmer',
       'sh', 'Dzongkha', 'Panjabi', 'Sundanese'], dtype=object)
```

Dropping rows with 'xx' as original language

```
In [43]: #Rows with 'xx' as Lang
data[data['original_language']=='xx']

#Dropping rows having 'xx'
data=data[~(data['original_language']=='xx')]
```

Dropping rows with 'sh' as original language

```
In [44]: #Rows with 'sh' as Lang
data[data['original_language']=='sh']

#Dropping rows having 'xx'
data=data[~(data['original_language']=='sh')]
```

Checking for duplicate entries

```
In [45]: data.duplicated().sum()
```

Out[45]: 0

No. of movies with unique title

```
In [46]: data['title'].nunique()
```

Out[46]: 9629

```
In [47]: data['title'].duplicated().sum()
```

Out[47]: 347

Rows with dupli movi titles

```
In [48]: data[data['title'].isin(data['title'][data['title'].duplicated()])].sort_values('id')
```

		title	release_date	genres	original_language	vote_average	vote_count	popularity
9364	3:10 to Yuma	1957-08-07	["Western", "Drama", "Thriller"]		English	7.2	291	10.9%
5116	3:10 to Yuma	2007-09-06	["Western"]		English	7.2	3131	18.9%
2684	A Nightmare on Elm Street	2010-04-30	["Horror", "Mystery", "Thriller"]		English	5.5	2415	25.6%

		title	release_date	genres	original_language	vote_average	vote_count	popularity
		A						
1637	Nightmare on Elm Street		1984-11-09	['Horror']	English	7.3	4430	36.60
7330	A Tale of Two Sisters		2023-05-06	[]	English	0.0	0	10.53
...
902	Wonder Woman		2017-05-30	['Action', 'Adventure', 'Fantasy']	English	7.2	18595	54.48
1133	Wrong Turn		2003-05-30	['Horror', 'Thriller']	English	6.3	2307	50.41
774	Wrong Turn		2021-01-26	['Horror', 'Thriller', 'Drama']	English	6.0	902	59.58
202	X		2022-03-17	['Horror', 'Mystery', 'Thriller']	English	6.8	1918	129.62
5515	X		2011-11-23	['Action', 'Thriller', 'Romance']	English	6.5	221	20.59

Rows with empty genres

In [49]:

```
data[data['genres'].str.len()==2]
```

Out[49]:

		title	release_date	genres	original_language	vote_average	vote_count	popularity
610	Snake Beauty		1994-03-26	[]	Chinese	0.0	0	61.292
684	Za gyakutai: Nyotai ikedori-hen		1987-07-18	[]	Japanese	6.5	1	62.096
816	Gabriel's Inferno: Part IV		2022-03-30	[]	English	5.0	2	37.269
943	Yu Pui Tsuen		1986-12-12	[]	cn	4.0	4	78.229
1178	Oppressive Torture		1978-01-14	[]	Japanese	4.7	3	45.046
...
8809	The Legend of Zhao Yun		2021-01-04	[]	Chinese	2.0	1	9.461
8982	Russian Nymphet: Temptation		2004-11-28	[]	Russian	5.3	3	16.728

		title	release_date	genres	original_language	vote_average	vote_count	popularity
9484		La Boheme: Breathe Umphefumlo	2015-02-05	[]	English	6.7	3	11.171
9775		野浪花	1987-05-26	[]	Chinese	0.0	0	8.040
9991		The Witcher Season One Recap: From the Beginning	2021-12-17	[]	English	5.6	8	9.045

In [50]:

```
#Dropping rows with empty genres
data=data[~(data['genres'].str.len()==2)]
```

Rows with empty production companies

In [51]:

```
data[data['production_companies'].str.len()==2]
```

Out[51]:

		title	release_date	genres	original_language	vote_average	vote_count	popularity
21		Adrenaline	2022-12-15	['Action']	English	5.9	39	717.8
27		The Elderly	2023-04-21	['Horror', 'Thriller', 'Fantasy']	Spanish	5.8	6	521.2
28		Ripper's Revenge	2023-04-03	['Horror']	English	4.9	10	523.1
36		Gangs of Lagos	2023-04-07	['Crime']	English	6.1	35	462.8
46		Prizefighter: The Life of Jem Belcher	2022-06-30	['Drama', 'History']	English	6.2	122	365.0
...	
9905		My Mother's Lovers	2020-05-29	['Drama']	Spanish	2.8	5	12.4
9924		My Brother's Wife 2	2016-08-31	['Romance', 'Family', 'Drama']	Korean	3.5	2	10.3
9930		Zombie Fight Club	2014-10-23	['Action', 'Horror']	Chinese	4.7	54	12.8
9975		Heart Shot	2022-02-17	['Romance', 'Crime']	English	5.6	69	9.1
9998		My Sister-in-law's Job	2017-08-31	['Drama', 'Romance']	Korean	5.0	5	10.4

519 rows × 11 columns

In [52]:

```
#Dropping rows with empty production
data=data[~(data['production_companies'].str.len()==2)]
```

Rows with zero vote_avg or vote_count or runtime values

In [53]:

```
data[(data['vote_average']==0) | (data['vote_count']==0) | (data['runtime']==0)]
```

Out[53]:

		title	release_date	genres	original_language	vote_average	vote_count	popularity
23		Fast X	2023-05-17	['Action', 'Crime', 'Thriller']	English	0.0	0	731
100		Kiss, Kiss!	2023-04-26	['Romance', 'Comedy']	Polish	6.9	15	361
101		The Little Mermaid	2023-05-18	['Adventure', 'Family', 'Fantasy', 'Romance']	English	0.0	0	220
128		Transformers: Rise of the Beasts	2023-06-07	['Action', 'Adventure', 'Science Fiction']	English	0.0	0	221
184		The Flash	2023-06-14	['Science Fiction', 'Action', 'Adventure']	English	0.0	0	121
...
9847		Divaldo: O Mensageiro da Paz	2019-09-12	['Drama']	Portuguese	8.3	59	11
9889		Influencer	2023-05-18	['Thriller', 'Horror', 'Mystery']	English	0.0	0	1
9922		Mental Finger	2023-05-06	['Comedy', 'Action']	Sundanese	0.0	0	1
9926		Patricia, A Hidden Passion	2020-01-21	['Drama', 'Comedy']	Spanish	6.0	33	11
9932		Godzilla x Kong: The New Empire	2024-03-13	['Action', 'Science Fiction', 'Adventure']	English	0.0	0	11

173 rows × 11 columns

In [54]:

```
#Dropping Rows with zero vote_avg or vote_count or runtime values

data= data[~((data['vote_average']==0) | (data['vote_count']==0) | (data['runtime']
```

Descriptive statistics of DF

In [55]:

```
data.describe()
```

Out[55]:

	vote_average	vote_count	popularity	budget	revenue	runtime
count	9223.000000	9223.000000	9223.000000	9.223000e+03	9.223000e+03	9223.000000
mean	6.542546	1652.788247	31.635078	2.081719e+07	6.479669e+07	103.287867
std	0.907423	2952.562540	114.867339	3.935068e+07	1.596469e+08	24.410714
min	1.000000	1.000000	7.411000	0.000000e+00	0.000000e+00	2.000000
25%	6.000000	170.000000	13.591000	0.000000e+00	0.000000e+00	91.000000
50%	6.600000	580.000000	17.619000	2.500000e+06	3.769990e+06	101.000000
75%	7.200000	1726.000000	27.182500	2.500000e+07	5.806524e+07	115.000000
max	10.000000	33633.000000	5089.969000	5.793304e+08	2.923706e+09	449.000000

Adding profit column

In [56]:

```
data['Profit']=data['revenue']-data['budget']
data.head()
```

Out[56]:

	title	release_date	genres	original_language	vote_average	vote_count	popularity
0	The Pope's Exorcist	2023-04-05	['Horror', 'Mystery', 'Thriller']	English	7.4	619	5089.96
1	Ant-Man and the Wasp: Quantumania	2023-02-15	['Action', 'Adventure', 'Science Fiction']	English	6.6	2294	4665.43
2	The Super Mario Bros. Movie	2023-04-05	['Animation', 'Adventure', 'Family', 'Fantasy']	English	7.5	1861	3935.55
3	Ghosted	2023-04-18	['Action', 'Comedy', 'Romance']	English	7.2	652	2791.53
4	Shazam! Fury of the Gods	2023-03-15	['Action', 'Comedy', 'Fantasy', 'Adventure']	English	6.8	1510	2702.59

Saving this DF as csv

```
In [57]: data.to_csv("topmovies.csv",index=False)
```

Top 10 Movies of every category

```
In [58]: max_budget=data.sort_values('budget',ascending=False).head(10)
max_revenue=data.sort_values('revenue',ascending=False).head(10)
max_votcnt=data.sort_values('vote_count',ascending=False).head(10)
max_votavg=data.sort_values('vote_average',ascending=False).head(10)
max_popularity=data.sort_values('popularity',ascending=False).head(10)
max_profit=data.sort_values('Profit',ascending=False).head(10)

#Loss
max_loss = data.sort_values('Profit', ascending=True).head(10)
max_loss['loss'] = -max_loss['Profit']
```

Visualization of them

In [90]:

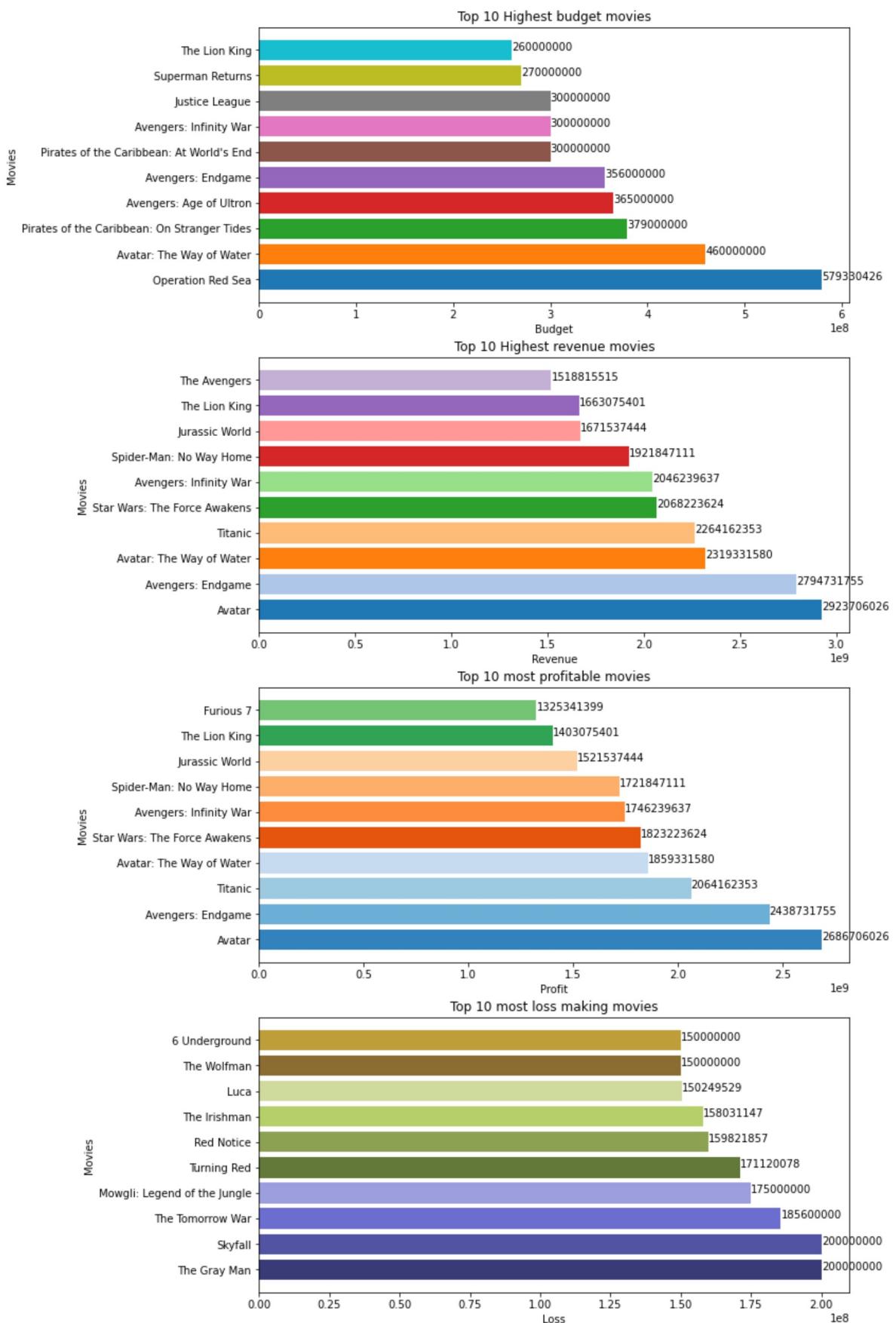
```
fig,axs = plt.subplots(4,1, figsize=(10,22))

#Plotting for Highest budget movies
col_map= plt.get_cmap('tab10')
axs[0].barh(max_budget['title'],max_budget['budget'],color=col_map.colors)
axs[0].set_title("Top 10 Highest budget movies")
axs[0].set_xlabel("Budget")
axs[0].set_ylabel("Movies")
for i in range(len(max_budget)):
    axs[0].text(max_budget['budget'].iloc[i],max_budget['title'].iloc[i],max_budget['title'].iloc[i],max_budget['budget'].iloc[i],color='black',va='bottom',ha='center')

#Plotting for Highest revenue movies
col_map= plt.get_cmap('tab20')
axs[1].barh(max_revenue['title'],max_revenue['revenue'],color=col_map.colors)
axs[1].set_title("Top 10 Highest revenue movies")
axs[1].set_xlabel("Revenue")
axs[1].set_ylabel("Movies")
for i in range(len(max_revenue)):
    axs[1].text(max_revenue['revenue'].iloc[i],max_revenue['title'].iloc[i],max_revenue['title'].iloc[i],max_revenue['revenue'].iloc[i],color='black',va='bottom',ha='center')

#Plotting for profitable movies
col_map= plt.get_cmap('tab20c')
axs[2].barh(max_profit['title'],max_profit['Profit'],color=col_map.colors)
axs[2].set_title("Top 10 most profitable movies")
axs[2].set_xlabel("Profit")
axs[2].set_ylabel("Movies")
for i in range(len(max_profit)):
    axs[2].text(max_profit['Profit'].iloc[i],max_profit['title'].iloc[i],max_profit['title'].iloc[i],max_profit['Profit'].iloc[i],color='black',va='bottom',ha='center')

#Plotting for Loss
col_map= plt.get_cmap('tab20b')
axs[3].barh(max_loss['title'],max_loss['loss'],color=col_map.colors)
axs[3].set_title("Top 10 most loss making movies")
axs[3].set_xlabel("Loss")
axs[3].set_ylabel("Movies")
for i in range(len(max_loss)):
    axs[3].text(max_loss['loss'].iloc[i],max_loss['title'].iloc[i],max_loss['title'].iloc[i],max_loss['loss'].iloc[i],color='black',va='bottom',ha='center')
```



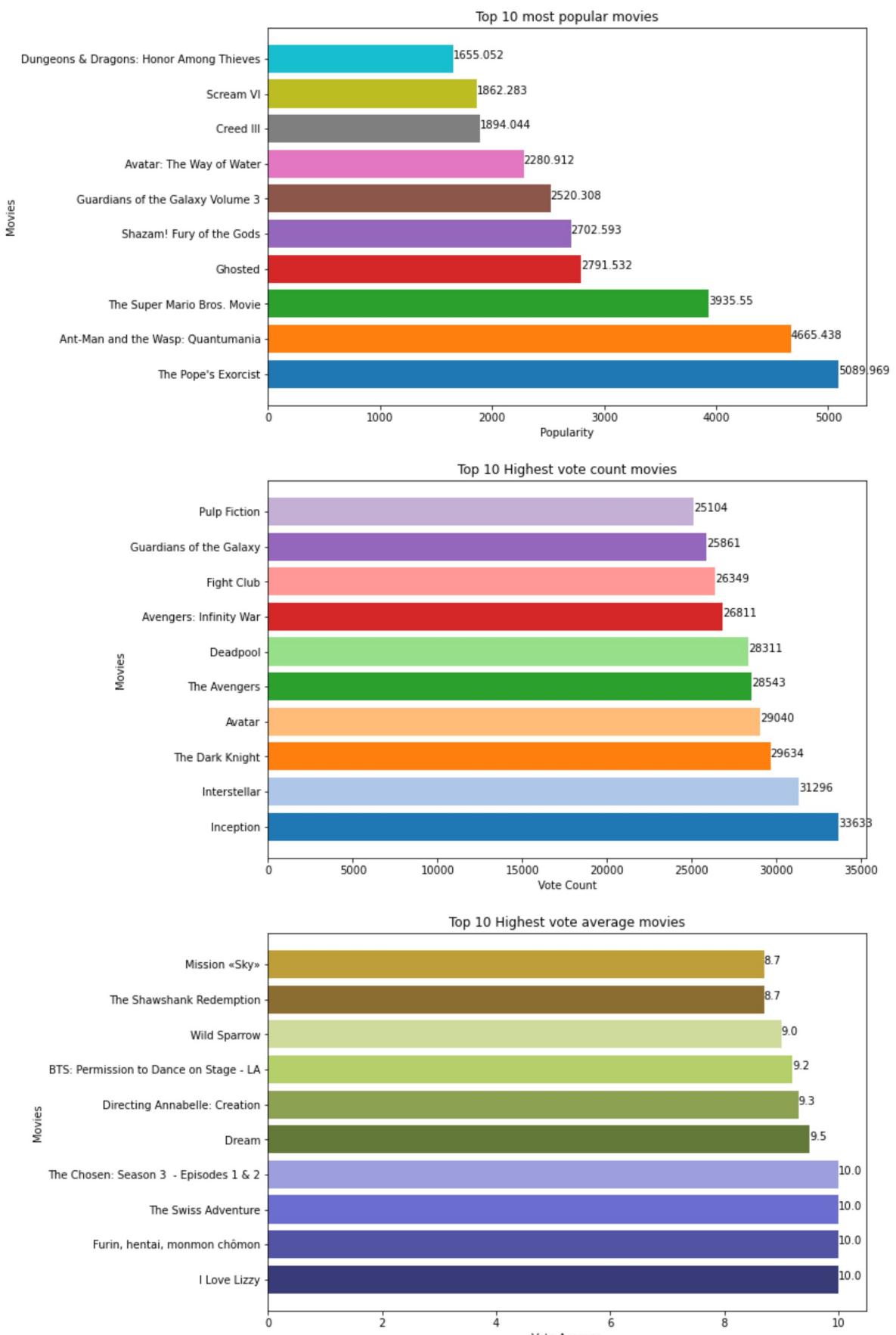
In [60]:

```
fig,axs = plt.subplots(3,1, figsize=(10,22))

#Most Popular movies
col_map= plt.get_cmap('tab10')
axs[0].barh(max_popularity['title'],max_popularity['popularity'],color=col_map.colors)
axs[0].set_title("Top 10 most popular movies")
axs[0].set_xlabel("Popularity")
axs[0].set_ylabel("Movies")
for i in range(len(max_popularity)):
    axs[0].text(max_popularity['popularity'].iloc[i],max_popularity['title'].iloc[i],max_popularity['title'].iloc[i],color=col_map.colors)

#Highest Voted movies
col_map= plt.get_cmap('tab20')
axs[1].barh(max_votcnt['title'],max_votcnt['vote_count'],color=col_map.colors)
axs[1].set_title("Top 10 Highest vote count movies")
axs[1].set_xlabel("Vote Count")
axs[1].set_ylabel("Movies")
for i in range(len(max_votcnt)):
    axs[1].text(max_votcnt['vote_count'].iloc[i],max_votcnt['title'].iloc[i],max_votcnt['title'].iloc[i],color=col_map.colors)

#Highest vote avg movies
col_map= plt.get_cmap('tab20b')
axs[2].barh(max_votavg['title'],max_votavg['vote_average'],color=col_map.colors)
axs[2].set_title("Top 10 Highest vote average movies")
axs[2].set_xlabel("Vote Average")
axs[2].set_ylabel("Movies")
for i in range(len(max_votavg)):
    axs[2].text(max_votavg['vote_average'].iloc[i],max_votavg['title'].iloc[i],max_votavg['title'].iloc[i],color=col_map.colors)
```



Counting of movies wrt languages

In [61]:

```
data['original_language'].value_counts()
```

```
Out[61]: English          6962
Japanese        648
Spanish          289
French           283
Korean           182
Italian          143
cn              119
Chinese          108
German            78
Russian           65
Tagalog           35
Hindi             33
Norwegian         31
Danish            30
Polish             26
Portuguese        24
Thai              23
Swedish           23
Dutch             22
Indonesian        16
Turkish           14
Tamil              8
Telugu            7
Finnish            6
Greek              4
Romanian          4
Ukrainian         4
Arabic             3
Basque             3
Hungarian          3
Persian            3
Galician           2
Khmer              2
Kannada            2
Czech              2
Malayalam          2
Serbian            2
Hebrew              2
Catalan            2
Icelandic          2
Estonian           1
Vietnamese         1
Norwegian Bokmal   1
Irish              1
Macedonian         1
Dzongkha           1
Name: original_language, dtype: int64
```

Telugu language movies data

```
In [62]: data = data.set_index('original_language')
df=data.loc['Telugu']
result=df[['title','production_companies']]
print(result)
```

original_language	title \
Telugu	RRR
Telugu	Bāhubali: The Beginning
Telugu	Pushpa: The Rise - Part 1
Telugu	Bāhubali 2: The Conclusion

Telugu	Ramabanam
Telugu	Eega
Telugu	Ala Vaikunthapurramuloo
	production_companies
original_language	
Telugu	['DVV Entertainment', 'Pen Studios', 'Lyca Pro...]
Telugu	['Arka Media Works']
Telugu	['Mythri Movie Makers', 'Muttamsetty Media', '...]
Telugu	['Arka Media Works']
Telugu	['People Media Factory']
Telugu	['Vaaraahi Chalana Chitram']
Telugu	['Geetha Arts', 'Haarika & Hassine Creations', ...]

Details of a particular movie

```
In [89]: data= data.set_index('title')

#Details of the movie 'RRR'
re = data.loc['RRR']
print(re)
```

release_date	2022-03-24
genres	['Action', 'Drama']
vote_average	7.8
vote_count	918
popularity	121.562
budget	69000000
production_companies	['DVV Entertainment', 'Pen Studios', 'Lyca Pro...']
revenue	160000000
runtime	186
Profit	91000000
Name:	RRR, dtype: object

Unique genres

```
In [70]: genres_col=data['genres']
uniq_genres=set()
for i in genres_col:
    genres=eval(i)
    uniq_genres.update(genres)
print(uniq_genres)
print()
#No. of unique genres
x=len(uniq_genres)
print("No. of unique genres: ",x)
```

```
{'Thriller', 'Music', 'Crime', 'Action', 'Family', 'Romance', 'Comedy', 'Animation',
 'Fantasy', 'Mystery', 'Documentary', 'Horror', 'Drama', 'War', 'History', 'TV Movie',
 'Adventure', 'Western', 'Science Fiction'}
```

No. of unique genres: 19

Top 5 Popular movies of every genre

In [81]:

```
for genre in uniq_genres:
    movi=data[data['genres'].str.contains(genre)].sort_values('popularity', ascending=False)
    mov_titl=movi['title']
    print(genre)
    for i,title in enumerate(mov_titl):
        print(i+1,title)
    print()
```

Thriller

- 1 The Pope's Exorcist
- 2 Scream VI
- 3 AKA
- 4 John Wick: Chapter 4
- 5 Cocaine Bear

Music

- 1 Encanto at the Hollywood Bowl
- 2 Guillermo del Toro's Pinocchio
- 3 Lyle, Lyle, Crocodile
- 4 Blue's Big City Adventure
- 5 Coco

Crime

- 1 AKA
- 2 John Wick: Chapter 4
- 3 Murder Mystery 2
- 4 Cocaine Bear
- 5 Kill Boksoon

Action

- 1 Ant-Man and the Wasp: Quantumania
- 2 Ghosted
- 3 Shazam! Fury of the Gods
- 4 Guardians of the Galaxy Volume 3
- 5 Avatar: The Way of Water

Family

- 1 The Super Mario Bros. Movie
- 2 Peter Pan & Wendy
- 3 Puss in Boots: The Last Wish
- 4 Pirates Down the Street II: The Ninjas from Across
- 5 Mummies

Romance

- 1 Ghosted
- 2 Shotgun Wedding
- 3 The Quintessential Quintuplets Movie
- 4 Unhappily Ever After
- 5 The Forbidden Legend: Sex & Chopsticks 2

Comedy

- 1 The Super Mario Bros. Movie
- 2 Ghosted
- 3 Shazam! Fury of the Gods
- 4 Dungeons & Dragons: Honor Among Thieves
- 5 Puss in Boots: The Last Wish

Animation

- 1 The Super Mario Bros. Movie
- 2 Puss in Boots: The Last Wish
- 3 Justice League x RWBY: Super Heroes & Huntsmen, Part One

4 Mummies

5 That Time I Got Reincarnated as a Slime the Movie: Scarlet Bond

Fantasy

- 1 The Super Mario Bros. Movie
- 2 Shazam! Fury of the Gods
- 3 Dungeons & Dragons: Honor Among Thieves
- 4 Peter Pan & Wendy
- 5 Puss in Boots: The Last Wish

Mystery

- 1 The Pope's Exorcist
- 2 Scream VI
- 3 Clock
- 4 Invitation to a Murder
- 5 Batman: The Doom That Came to Gotham

Documentary

- 1 Money Shot: The Pornhub Story
- 2 Orgasm Inc: The Story of OneTaste
- 3 Cocaine Bear: The True Story
- 4 Harry Potter 20th Anniversary: Return to Hogwarts
- 5 Melody Makers

Horror

- 1 The Pope's Exorcist
- 2 Scream VI
- 3 Evil Dead Rise
- 4 The Communion Girl
- 5 Winnie the Pooh: Blood and Honey

Drama

- 1 Creed III
- 2 The Last Kingdom: Seven Kings Must Die
- 3 Puss in Boots: The Last Wish
- 4 The Park
- 5 Marcel the Shell with Shoes On

War

- 1 The Last Kingdom: Seven Kings Must Die
- 2 Sisu
- 3 Sniper: The White Raven
- 4 Gold Run
- 5 Devotion

History

- 1 The Last Kingdom: Seven Kings Must Die
- 2 The Woman King
- 3 3-D Sex and Zen: Extreme Ecstasy
- 4 Medieval
- 5 Devotion

TV Movie

- 1 Girl in the Basement
- 2 Teen Wolf: The Movie
- 3 Dungeons & Dragons: Wrath of the Dragon God
- 4 Monster High: The Movie
- 5 Miraculous World: New York, United HeroeZ

Adventure

- 1 Ant-Man and the Wasp: Quantumania
- 2 The Super Mario Bros. Movie
- 3 Shazam! Fury of the Gods

4 Guardians of the Galaxy Volume 3
5 Avatar: The Way of Water

Western

1 Ghosts of the Ozarks
2 Django Unchained
3 Spirit: Stallion of the Cimarron
4 Paws of Fury: The Legend of Hank
5 Tom and Jerry Cowboy Up!

Science Fiction

1 Ant-Man and the Wasp: Quantumania
2 Guardians of the Galaxy Volume 3
3 Avatar: The Way of Water
4 65
5 Black Panther: Wakanda Forever

Top 5 Profitable movies of every genre

In [82]:

```
for genre in uniq_genres:  
    movi=data[data['genres'].str.contains(genre)].sort_values('Profit',ascending=False)  
    mov_titl=movi['title']  
    print(genre)  
    for i,title in enumerate(mov_titl):  
        print(i+1,title)  
    print()
```

Thriller

1 Jurassic World
2 Furious 7
3 Jurassic World: Fallen Kingdom
4 Joker
5 The Fate of the Furious

Music

1 Bohemian Rhapsody
2 Coco
3 Sing
4 La La Land
5 A Star Is Born

Crime

1 Furious 7
2 Joker
3 The Fate of the Furious
4 The Dark Knight Rises
5 The Dark Knight

Action

1 Avatar
2 Avengers: Endgame
3 Avatar: The Way of Water
4 Star Wars: The Force Awakens
5 Avengers: Infinity War

Family

1 The Lion King
2 Frozen II
3 Frozen
4 Beauty and the Beast
5 Minions

Romance

- 1 Titanic
- 2 Beauty and the Beast
- 3 Aladdin
- 4 Shrek 2
- 5 The Twilight Saga: Breaking Dawn - Part 2

Comedy

- 1 Frozen II
- 2 Minions
- 3 The Super Mario Bros. Movie
- 4 Despicable Me 3
- 5 Jumanji: Welcome to the Jungle

Animation

- 1 The Lion King
- 2 Frozen II
- 3 Frozen
- 4 Minions
- 5 Incredibles 2

Fantasy

- 1 Avatar
- 2 Star Wars: The Force Awakens
- 3 Frozen II
- 4 Harry Potter and the Deathly Hallows: Part 2
- 5 Frozen

Mystery

- 1 Harry Potter and the Order of the Phoenix
- 2 Full River Red
- 3 The Da Vinci Code
- 4 The Sixth Sense
- 5 The Batman

Documentary

- 1 Fahrenheit 9/11
- 2 This Is It
- 3 Jackass 3D
- 4 Jackass Forever
- 5 Jackass Number Two

Horror

- 1 It
- 2 Jaws
- 3 The Exorcist
- 4 It Chapter Two
- 5 The Meg

Drama

- 1 Titanic
- 2 The Lion King
- 3 Top Gun: Maverick
- 4 Joker
- 5 Bohemian Rhapsody

War

- 1 Wolf Warrior 2
- 2 Dunkirk
- 3 American Sniper
- 4 The Battle at Lake Changjin: Water Gate Bridge
- 5 Saving Private Ryan

History

- 1 Bohemian Rhapsody
- 2 Full River Red
- 3 The Battle at Lake Changjin: Water Gate Bridge
- 4 Saving Private Ryan
- 5 The King's Speech

TV Movie

- 1 High School Musical 2
- 2 A Year-End Medley
- 3 Stargate: The Ark of Truth
- 4 Teen Wolf: The Movie
- 5 Under the Sea: A Descendants Story

Adventure

- 1 Avatar
- 2 Avengers: Endgame
- 3 Avatar: The Way of Water
- 4 Star Wars: The Force Awakens
- 5 Avengers: Infinity War

Western

- 1 Dances with Wolves
- 2 The Revenant
- 3 Django Unchained
- 4 True Grit
- 5 City Slickers

Science Fiction

- 1 Avatar
- 2 Avengers: Endgame
- 3 Avatar: The Way of Water
- 4 Star Wars: The Force Awakens
- 5 Avengers: Infinity War

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