

# final\_project\_disney

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## 1 Data Analysis for Disney Movies

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### 1.2 Introduction:

This report analyzes a few datasets related to Disney movies and answers a question related to the movies by using some Python concepts, data structures and modules like Pandas, NumPy and altair for visualization.

### 1.3 Question(s) of interests

The question considered in this report is “Which adventure movie director had a blockbuster movie with gross earning more than \$400M in the 21st Century(the year 2000’s)?” I am interested in knowing the answer to this question to understand who is the skilled Disney director to work with and vote for the Academy awards and also vote for the best song. I am considering the total gross earnings for the movies as against the inflation adjusted gross.

### 1.4 Dataset description

The below descriptions were taken directly from the website where the datasets were obtained.

“Disney characters, box office success & annual gross income”

The Disney movie dataset is composed of 5 tables - disney\_movies\_total\_gross.csv, disney\_revenue\_1991-2016.csv, disney-characters.csv, disney-director.csv and disney-voice-actors.csv . Each table is stored in a .csv file and contains different information about Disney movies pieces including gross income, characters, songs, revenue, directors and voice actors. I will be using the disney\_movies\_total\_gross, disney-characters and disney-director tables formally described below:

disney\_movies\_total\_gross.csv This file contains information on movie title, including a release date, genre, MPAA rating, total gross income and inflation adjusted gross income. disney-characters.csv This file includes information on movie title, including a release date, hero, villain and song for every movie. disney-director.csv This file includes information on movie title and the director for each movie.

### 1.5 Methods & Results:

In this analysis, I plan to import the relevant modules like pandas, altair and numpy. Then read the 3 datasets - disney\_movies\_total\_gross.csv, disney-director.csv and disney-characters.csv. Then I

would do some wrangling and cleaning of the datasets to convert them into a format where I can merge them and extract the answer to my question. I am going to use a function for wrangling and formatting and also some test functions to test my function. I am going to format my function using flake8 and black formatting. Finally, I will derive the answer to my question. Let's begin.

### 1.5.1 Let us import the modules, read our tables and do some basic visualizations:

```
[1]: #import the modules to be used
import pandas as pd
import altair as alt
import numpy as np
```

```
[2]: #Reading the first dataset - disney_movies_total_gross:
disney_total_gross_df = pd.read_csv('data/disney_movies_total_gross.csv')
disney_total_gross_df
```

```
[2]:
```

	movie_title	release_date	genre	MPAA_rating	\
0	Snow White and the Seven Dwarfs	Dec 21, 1937	Musical	G	
1	Pinocchio	Feb 9, 1940	Adventure	G	
2	Fantasia	Nov 13, 1940	Musical	G	
3	Song of the South	Nov 12, 1946	Adventure	G	
4	Cinderella	Feb 15, 1950	Drama	G	
..	...	...	...	...	
574	The Light Between Oceans	Sep 2, 2016	Drama	PG-13	
575	Queen of Katwe	Sep 23, 2016	Drama	PG	
576	Doctor Strange	Nov 4, 2016	Adventure	PG-13	
577	Moana	Nov 23, 2016	Adventure	PG	
578	Rogue One: A Star Wars Story	Dec 16, 2016	Adventure	PG-13	

	total_gross	inflation_adjusted_gross
0	\$184,925,485	\$5,228,953,251
1	\$84,300,000	\$2,188,229,052
2	\$83,320,000	\$2,187,090,808
3	\$65,000,000	\$1,078,510,579
4	\$85,000,000	\$920,608,730
..	...	...
574	\$12,545,979	\$12,545,979
575	\$8,874,389	\$8,874,389
576	\$232,532,923	\$232,532,923
577	\$246,082,029	\$246,082,029
578	\$529,483,936	\$529,483,936

[579 rows x 6 columns]

```
[3]: #Reading the second dataset - disney_director:
disney_director_df = pd.read_csv('data/disney-director.csv')
#renaming the column to make merging easier
```

```
disney_director_df_new = disney_director_df.rename(columns={"name":  
↳"movie_title"})  
disney_director_df_new
```

```
[3]:
```

	movie_title	director
0	Snow White and the Seven Dwarfs	David Hand
1	Pinocchio	Ben Sharpsteen
2	Fantasia	full credits
3	Dumbo	Ben Sharpsteen
4	Bambi	David Hand
5	Saludos Amigos	Jack Kinney
6	The Three Caballeros	Norman Ferguson
7	Make Mine Music	Jack Kinney
8	Fun and Fancy Free	Jack Kinney
9	Melody Time	Clyde Geronimi
10	The Adventures of Ichabod and Mr. Toad	Jack Kinney
11	Cinderella	Wilfred Jackson
12	Alice in Wonderland	Clyde Geronimi
13	Peter Pan	Hamilton Luske
14	Lady and the Tramp	Hamilton Luske
15	Sleeping Beauty	Clyde Geronimi
16	101 Dalmatians	Wolfgang Reitherman
17	The Sword in the Stone	Wolfgang Reitherman
18	The Jungle Book	Wolfgang Reitherman
19	The Aristocats	Wolfgang Reitherman
20	Robin Hood	Wolfgang Reitherman
21	The Many Adventures of Winnie the Pooh	Wolfgang Reitherman
22	The Rescuers	Wolfgang Reitherman
23	The Fox and the Hound	Art Stevens
24	The Black Cauldron	Ted Berman
25	The Great Mouse Detective	Ron Clements
26	Oliver & Company	George Scribner
27	The Little Mermaid	Ron Clements
28	The Rescuers Down Under	Mike Gabriel
29	Beauty and the Beast	Gary Trousdale
30	Aladdin	Ron Clements
31	The Lion King	Roger Allers
32	Pocahontas	Mike Gabriel
33	The Hunchback of Notre Dame	Gary Trousdale
34	Hercules	Ron Clements
35	Mulan	Barry Cook
36	Tarzan	Chris Buck
37	Fantasia 2000	full credits
38	Dinosaur	Ralph Zondag
39	The Emperor's New Groove	Mark Dindal
40	Atlantis: The Lost Empire	Gary Trousdale
41	Lilo & Stitch	Chris Sanders

42	Treasure Planet	Ron Clements
43	Brother Bear	Robert Walker
44	Home on the Range	Will Finn
45	Chicken Little	Mark Dindal
46	Meet the Robinsons	Stephen J. Anderson
47	Bolt	Chris Williams
48	The Princess and the Frog	Ron Clements
49	Tangled	Nathan Greno
50	Winnie the Pooh	Stephen J. Anderson
51	Wreck-It Ralph	Rich Moore
52	Frozen	Chris Buck
53	Big Hero 6	Don Hall
54	Zootopia	Byron Howard
55	Moana	Ron Clements

```
[4]: #Reading the third dataset - disney_characters:

disney_characters_df = pd.read_csv('data/disney-characters.csv')
```

## 1.6 Summarizing the datasets

```
[5]: disney_total_gross_df.describe()
```

```
[5]:
```

	movie_title	release_date	genre	MPAA_rating	total_gross \
count	579	579	562	523	579
unique	573	553	12	5	576
top	The Jungle Book	Dec 25, 1997	Comedy	PG	\$0
freq	3	3	182	187	4

	inflation_adjusted_gross
count	579
unique	576
top	\$0
freq	4

```
[6]: disney_director_df_new.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 56 entries, 0 to 55
Data columns (total 2 columns):
#   Column      Non-Null Count  Dtype
---  -
0   movie_title  56 non-null    object
1   director    56 non-null    object
dtypes: object(2)
memory usage: 1.0+ KB
```

```
[7]: disney_characters_df.shape
```

```
[7]: (56, 5)
```

## 1.7 Data Wrangling and clean up

```
[8]: #Wrangling the disney_movies_total_gross dataset using the wrangle_disney_data
      ↪function

      from wrangle_data import wrangle_disney_data

      ch = "$,"
      col1 = 'total_gross'
      disney_gross_wrangled_df = wrangle_disney_data(disney_total_gross_df,col1,ch,
      ↪col2 = 'release_date')
      disney_gross_wrangled_df
```

```
[8]:
```

	movie_title	release_date	genre	MPAA_rating	\
0	Snow White and the Seven Dwarfs	1937-12-21	Musical	G	
1	Pinocchio	1940-02-09	Adventure	G	
2	Fantasia	1940-11-13	Musical	G	
3	Song of the South	1946-11-12	Adventure	G	
4	Cinderella	1950-02-15	Drama	G	
..	...	...	...	...	
574	The Light Between Oceans	2016-09-02	Drama	PG-13	
575	Queen of Katwe	2016-09-23	Drama	PG	
576	Doctor Strange	2016-11-04	Adventure	PG-13	
577	Moana	2016-11-23	Adventure	PG	
578	Rogue One: A Star Wars Story	2016-12-16	Adventure	PG-13	

	total_gross	inflation_adjusted_gross
0	184925485	\$5,228,953,251
1	84300000	\$2,188,229,052
2	83320000	\$2,187,090,808
3	65000000	\$1,078,510,579
4	85000000	\$920,608,730
..	...	...
574	12545979	\$12,545,979
575	8874389	\$8,874,389
576	232532923	\$232,532,923
577	246082029	\$246,082,029
578	529483936	\$529,483,936

```
[579 rows x 6 columns]
```

```
[9]: #Wrangling the disney-characters dataset using the wrangle_disney_data function
```

```
from wrangle_data import wrangle_disney_data
```

```
ch = "[\\n,]"
```

```
column1 = 'movie_title'
```

```
disney_characters_wrangled_df =
```

```
    ↪ wrangle_disney_data(disney_characters_df, column1, ch, col2 = 'release_date')
```

```
disney_characters_wrangled_df
```

```
[9]:
```

	movie_title	release_date	\
0	Snow White and the Seven Dwarfs	1937-12-21	
1	Pinocchio	1940-02-07	
2	Fantasia	1940-11-13	
3	Dumbo	1941-10-23	
4	Bambi	1942-08-13	
5	Saludos Amigos	1943-02-06	
6	The Three Caballeros	1945-02-03	
7	Make Mine Music	1946-04-20	
8	Fun and Fancy Free	1947-09-27	
9	Melody Time	1948-05-27	
10	The Adventures of Ichabod and Mr. Toad	1949-10-05	
11	Cinderella	1950-02-15	
12	Alice in Wonderland	1951-07-28	
13	Peter Pan	1953-02-05	
14	Lady and the Tramp	1955-06-22	
15	Sleeping Beauty	1959-01-29	
16	One Hundred and One Dalmatians	1961-01-25	
17	The Sword in the Stone	1963-12-25	
18	The Jungle Book	1967-10-18	
19	The Aristocats	1970-12-24	
20	Robin Hood	1973-11-08	
21	The Many Adventures of Winnie the Pooh	1977-03-11	
22	The Rescuers	1977-06-22	
23	The Fox and the Hound	1981-07-10	
24	The Black Cauldron	1985-07-24	
25	The Great Mouse Detective	1986-07-02	
26	Oliver & Company	1988-11-18	
27	The Little Mermaid	1989-11-17	
28	The Rescuers Down Under	1990-11-16	
29	Beauty and the Beast	1991-11-22	
30	Aladdin	1992-11-25	
31	The Lion King	1994-06-24	
32	Pocahontas	1995-06-23	
33	The Hunchback of Notre Dame	1996-06-21	
34	Hercules	1997-06-27	
35	Mulan	1998-06-19	

36	Tarzan	1999-06-18
37	Fantasia 2000	1999-12-17
38	Dinosaur	2000-05-19
39	The Emperor's New Groove	2000-12-15
40	Atlantis: The Lost Empire	2001-06-15
41	Lilo & Stitch	2002-06-21
42	Treasure Planet	2002-11-27
43	Brother Bear	2003-11-01
44	Home on the Range	2004-04-02
45	Chicken Little	2005-11-04
46	Meet the Robinsons	2007-03-30
47	Bolt	2008-11-21
48	The Princess and the Frog	2009-12-11
49	Tangled	2010-11-24
50	Winnie the Pooh	2011-07-15
51	Wreck-It Ralph	2012-11-02
52	Frozen	2013-11-27
53	Big Hero 6	2014-11-07
54	Zootopia	2016-03-04
55	Moana	2016-11-23

	hero	villian \
0	Snow White	Evil Queen
1	Pinocchio	Stromboli
2	NaN	Chernabog
3	Dumbo	Ringmaster
4	Bambi	Hunter
5	Donald Duck	NaN
6	Donald Duck	NaN
7	NaN	NaN
8	Mickey Mouse	Willie the Giant
9	NaN	NaN
10	Mr. Toad and Ichabod Crane	Mr. Winkie and The Headless Horseman
11	Cinderella	Lady Tremaine
12	Alice	Queen of Hearts
13	Peter Pan	Captain Hook
14	Lady and Tramp	Si and Am
15	Aurora	Maleficent
16	Pongo	Cruella de Vil
17	Arthur	Madam Mim
18	Mowgli	Kaa and Shere Khan
19	Thomas and Duchess	Edgar Balthazar
20	Robin Hood	Prince John
21	Winnie the Pooh	NaN
22	Bernard and Miss Bianca	Madame Medusa
23	Tod and Copper	Amos Slade
24	Taran	Horned King

25	Basil	Professor Ratigan
26	Oliver	Sykes
27	Ariel	Ursula
28	Bernard and Miss Bianca	Percival C. McLeach
29	Belle	Gaston
30	Aladdin	Jafar
31	Simba	Scar
32	Pocahontas	Governor Ratcliffe
33	Quasimodo	Claude Frollo
34	Hercules	Hades
35	Mulan	Shan Yu
36	Tarzan	Clayton
37	NaN	NaN
38	Aladar	Kron
39	Kuzco	Yzma
40	Milo Thatch	Commander Rourke
41	Lilo and Stitch	NaN
42	Jim Hawkins	John Silver
43	Kenai	Denahi
44	Maggie	Alameda Slim
45	Ace Cluck	Foxy Loxy
46	Lewis	Doris
47	Bolt	Dr. Calico
48	Tiana	Dr. Facilier
49	Rapunzel	Mother Gothel
50	Winnie the Pooh	NaN
51	Ralph	Turbo
52	Elsa	Prince Hans
53	Hiro Hamada	Professor Callaghan
54	Judy Hopps	NaN
55	Moana	NaN

	song
0	Some Day My Prince Will Come
1	When You Wish upon a Star
2	NaN
3	Baby Mine
4	Love Is a Song
5	Saludos Amigos\n
6	NaN
7	NaN
8	NaN
9	Little Toot
10	The Merrily Song
11	Bibbidi-Bobbidi-Boo
12	The Unbirthday Song
13	You Can Fly!



14 Bella Notte  
 15 Once Upon a Dream  
 16 Cruella De Vil  
 17 Higitus Figitus\n  
 18 The Bare Necessities\n  
 19 Ev'rybody Wants to Be a Cat  
 20 Oo De Lally  
 21 Winnie the Pooh  
 22 The Journey  
 23 Best of Friends  
 24 NaN  
 25 The World's Greatest Criminal Mind  
 26 Once Upon a Time in New York City  
 27 Under the Sea  
 28 \n  
 29 Be Our Guest  
 30 A Whole New World  
 31 Circle of Life  
 32 Colors of the Wind  
 33 God Help the Outcasts  
 34 Go the Distance  
 35 I'll Make a Man Out of You\n  
 36 You'll Be in My Heart  
 37 NaN  
 38 NaN  
 39 My Funny Friend and Me\n  
 40 Where the Dream Takes You  
 41 He Mele No Lilo\n  
 42 I'm Still Here\n  
 43 Look Through My Eyes  
 44 NaN  
 45 NaN  
 46 Little Wonders  
 47 I Thought I Lost You  
 48 Almost There  
 49 I See the Light  
 50 Winnie the Pooh\n  
 51 Sugar Rush  
 52 Let It Go\n  
 53 Immortals  
 54 Try Everything  
 55 How Far I'll Go\n

## 1.8 Filtering the first dataset for our analysis

```
[10]: #Finding the movies from first dataset(data/disney_movies_total_gross.csv)
      ↳where gross earnings is more than $400M

disney_movies_400M_df = disney_total_gross_df.
      ↳loc[disney_gross_wrangled_df['total_gross'].astype(int) > 400000000]
disney_movies_400M_df
```

```
[10]:
```

	movie_title	release_date	genre	MPAA_rating	\
179	The Lion King	1994-06-15	Adventure	G	
441	Pirates of the Caribbean: Dead Man's...	2006-07-07	Adventure	PG-13	
499	Toy Story 3	2010-06-18	Adventure	G	
524	The Avengers	2012-05-04	Action	PG-13	
532	Iron Man 3	2013-05-03	Action	PG-13	
539	Frozen	2013-11-22	Adventure	PG	
558	Avengers: Age of Ultron	2015-05-01	Action	PG-13	
564	Star Wars Ep. VII: The Force Awakens	2015-12-18	Adventure	PG-13	
569	Captain America: Civil War	2016-05-06	Action	PG-13	
571	Finding Dory	2016-06-17	Adventure	PG	
578	Rogue One: A Star Wars Story	2016-12-16	Adventure	PG-13	

	total_gross	inflation_adjusted_gross
179	422780140	\$761,640,898
441	423315812	\$544,817,142
499	415004880	\$443,408,255
524	623279547	\$660,081,224
532	408992272	\$424,084,233
539	400738009	\$414,997,174
558	459005868	\$459,005,868
564	936662225	\$936,662,225
569	408084349	\$408,084,349
571	486295561	\$486,295,561
578	529483936	\$529,483,936

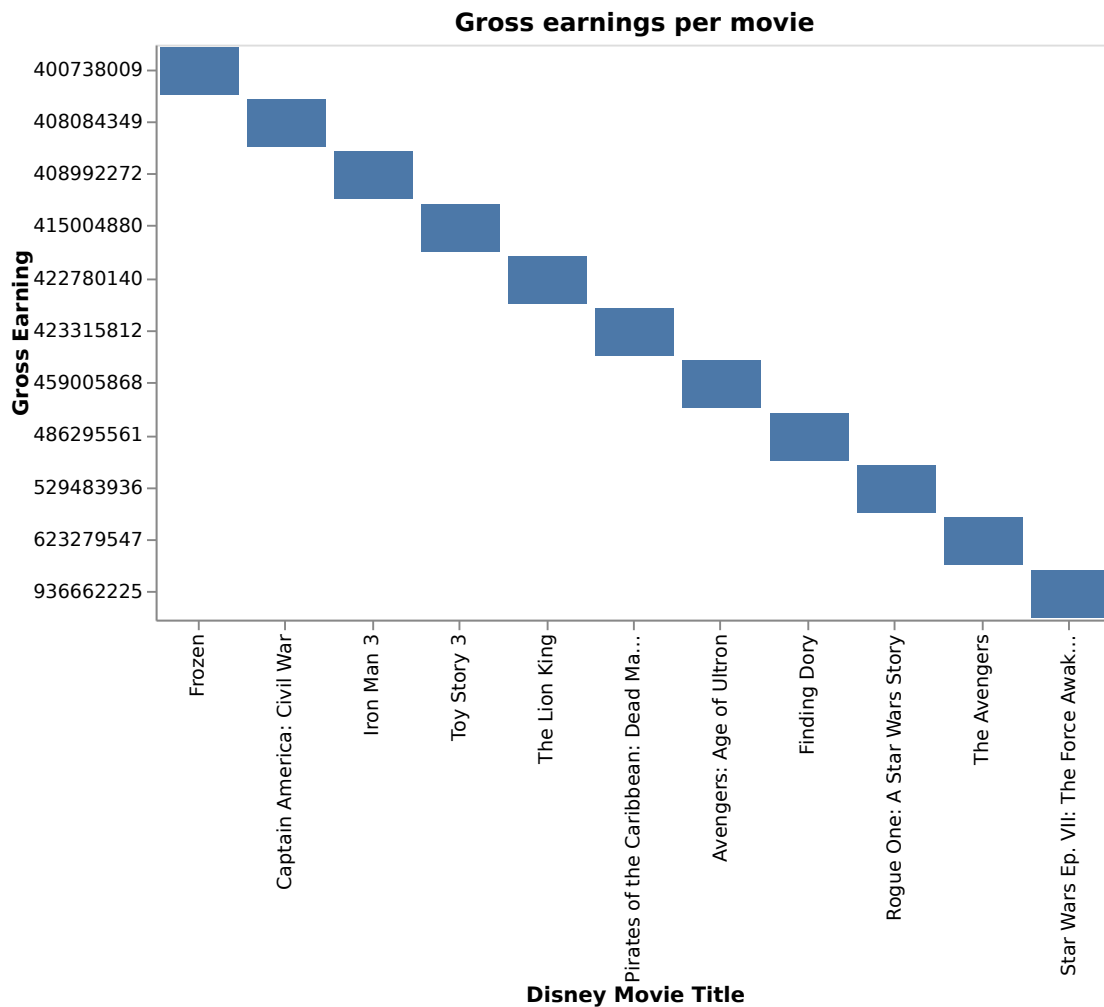
## 1.9 Visualization with Altair

```
[11]: #Visualizing the first dataset:
alt.renderers.enable('altair_saver', fmts=["html", "svg"])

disney_total_gross_plot = alt.Chart(disney_movies_400M_df, width=500,
      ↳height=300).mark_bar().encode (x=alt.X('movie_title', sort="y",
      ↳title='Disney Movie Title'),
      ↳y=alt.Y('total_gross', title= 'Gross Earning')).properties(title="Gross
      ↳earnings per movie")
```

disney\_total\_gross\_plot

[11]:

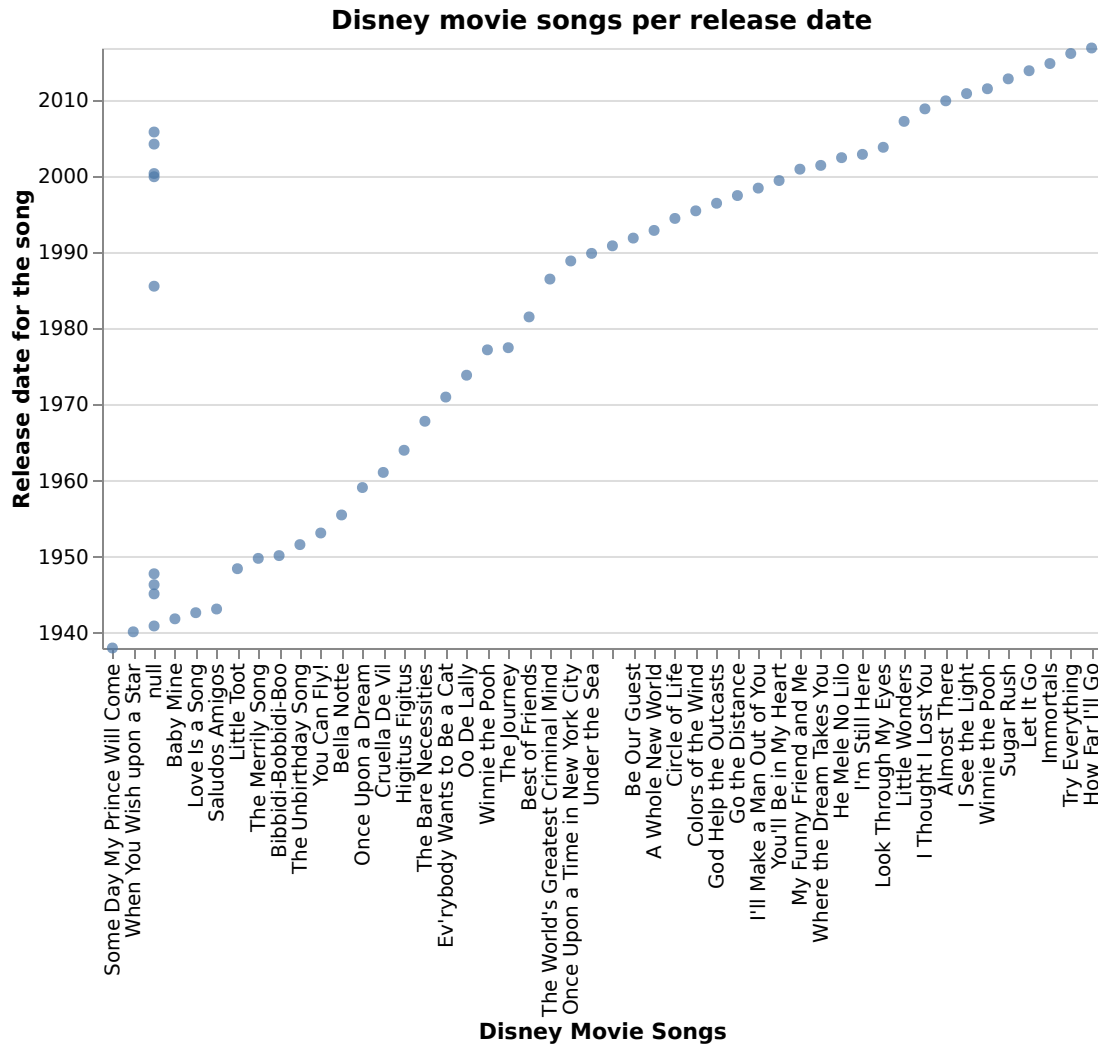


[12]:

```
#Visualizing the first dataset:
disney_characters_plot = alt.Chart(disney_characters_wrangled_df, width=500, height=300).mark_circle().encode(x=alt.X('song', sort="y", title='Disney Movie Songs'),
y=alt.Y('release_date', title= 'Release date for the song')).
properties(title="Disney movie songs per release date")

disney_characters_plot
```

[12]:



## 1.10 Code Improvement - black formatting

```
[13]: !flake8 wrangle_data.py
```

```
#Using black formatting to improve our function code
!black wrangle_data.py
```

```
wrangle_data.py:4:80: E501 line too long (81 > 79 characters)
wrangle_data.py:25:80: E501 line too long (86 > 79 characters)
All done!
1 file left unchanged.
```

## 1.11 Testing the wrangle\_data function

```
[14]: !pytest test_wrangle_data.py
```

```
===== test session starts
=====
platform linux -- Python 3.8.5, pytest-6.2.4, py-1.10.0, pluggy-0.13.1
rootdir: /home/jupyter/prog-python-ds-students/release/final_project
plugins: anyio-3.2.1, dash-1.20.0
collected 2 items

test_wrangle_data.py ..
[100%]

===== 2 passed in 0.81s
=====
```

```
[15]: #Summarizing the data - combining the 3 datasets into one consolidated
      ↳ dataframe to make our analysis easier
```

```
merged_gross_character_df = pd.merge(disney_movies_400M_df,
↳disney_director_df_new, how="left", on=["movie_title"])
merged_gross_character_df = pd.
↳merge(merged_gross_character_df,disney_characters_df, how="left",
↳on=["movie_title"] )
merged_gross_character_df = merged_gross_character_df.
↳rename(columns={'release_date_x':'release_date','release_date_y':
↳'release_date_song'})
merged_gross_character_df
```

```
[15]:
```

	movie_title	release_date	genre	MPAA_rating	\
0	The Lion King	1994-06-15	Adventure	G	
1	Pirates of the Caribbean: Dead Man's...	2006-07-07	Adventure	PG-13	
2	Toy Story 3	2010-06-18	Adventure	G	
3	The Avengers	2012-05-04	Action	PG-13	
4	Iron Man 3	2013-05-03	Action	PG-13	
5	Frozen	2013-11-22	Adventure	PG	
6	Avengers: Age of Ultron	2015-05-01	Action	PG-13	
7	Star Wars Ep. VII: The Force Awakens	2015-12-18	Adventure	PG-13	
8	Captain America: Civil War	2016-05-06	Action	PG-13	
9	Finding Dory	2016-06-17	Adventure	PG	
10	Rogue One: A Star Wars Story	2016-12-16	Adventure	PG-13	

	total_gross	inflation_adjusted_gross	director	release_date_song	\
0	422780140	\$761,640,898	Roger Allers	1994-06-24	

1	423315812	\$544,817,142	NaN	NaT
2	415004880	\$443,408,255	NaN	NaT
3	623279547	\$660,081,224	NaN	NaT
4	408992272	\$424,084,233	NaN	NaT
5	400738009	\$414,997,174	Chris Buck	2013-11-27
6	459005868	\$459,005,868	NaN	NaT
7	936662225	\$936,662,225	NaN	NaT
8	408084349	\$408,084,349	NaN	NaT
9	486295561	\$486,295,561	NaN	NaT
10	529483936	\$529,483,936	NaN	NaT

	hero	villian	song
0	Simba	Scar	Circle of Life
1	NaN	NaN	NaN
2	NaN	NaN	NaN
3	NaN	NaN	NaN
4	NaN	NaN	NaN
5	Elsa	Prince Hans	Let It Go\n
6	NaN	NaN	NaN
7	NaN	NaN	NaN
8	NaN	NaN	NaN
9	NaN	NaN	NaN
10	NaN	NaN	NaN

[16]: *#Dropping the rows with Nan values*

```
df_dropped = merged_gross_character_df.dropna()
df_dropped
```

```
[16]:      movie_title  release_date      genre MPAA_rating  total_gross \
0  The Lion King   1994-06-15  Adventure          G    422780140
5         Frozen   2013-11-22  Adventure          PG    400738009

inflation_adjusted_gross      director  release_date_song  hero \
0          $761,640,898  Roger Allers    1994-06-24  Simba
5          $414,997,174   Chris Buck    2013-11-27   Elsa

      villian      song
0         Scar  Circle of Life
5  Prince Hans    Let It Go\n
```

[17]: *# Filtering the movie which was made in the 2000's*

```
df_dropped_final = df_dropped.loc[df_dropped['release_date'].dt.year > 2000 ]
print("Highest Earning Adventure Movie Director")
df_dropped_final["director"]
```

Highest Earning Adventure Movie Director

```
[17]: 5    Chris Buck  
      Name: director, dtype: object
```

```
[18]: print("Award Winning Song")  
      df_dropped_final["song"].replace("\n,", "", regex=True)
```

Award Winning Song

```
[18]: 5    Let It Go  
      Name: song, dtype: object
```

## 1.12 Discussions

In this analysis, I found the director who directed the highest earning movie of the 2000's and most importantly the Academy Award winning song from the movie(Frozen). The song is "Let it go" which is definitely a very popular song. This answers my original question which was a part of this analysis. The other questions that I want to answer are the genre and the MPAA rating of this movie.

## 2 References

Not all the work in this notebook is original. Some parts were borrowed from online resources. I take no credit for parts that are not mine. They were solely used for illustration purposes

- [Data Source](#) All the tables(dataset) were derived from this website
- [Question Of Interest](#) I got my interest in the question after going through this website