

Introduction

This report contains an analysis of sales for different video games over time. We strive to show how the industry of video games has evolved from the year 1971 to 2024. The data we have used to carry out the research has various fields that enable us to conduct various investigations and get findings. The data has different game titles whose sales information is broken down to different regions and the ultimate total sales. The different attributes of each game is also given in the data, that is, the critic score of the game, the genre of the game, the console to which the game is made and more. These attributes are probable factors that affect how successful a game is, for instance the higher the critic score the higher the sales are expected to be. We are going to do a step-by-step analysis to ascertain or dispute the assumptions.

```
'''We import the necessary libraries to be used in the analysis'''
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
# this sets to display all the columns in our data
pd.set_option('display.max_columns',500)
```

```
# We import our data as a DataFrame and assign it to a variable named df.
df = pd.read_csv('vgchartz-2024.csv')
df.head() # displays the first five rows of our data
```

	img
title \	
0 /games/boxart/full_6510540AmericaFrontccc.jpg	Grand Theft Auto V
1 /games/boxart/full_5563178AmericaFrontccc.jpg	Grand Theft Auto V
2 /games/boxart/827563ccc.jpg	Grand Theft Auto: Vice City
3 /games/boxart/full_9218923AmericaFrontccc.jpg	Grand Theft Auto V
4 /games/boxart/full_4990510AmericaFrontccc.jpg	Call of Duty: Black Ops 3

	console	genre	publisher	developer	critic_score
total_sales \					
0	PS3	Action	Rockstar Games	Rockstar North	9.4
20.32					
1	PS4	Action	Rockstar Games	Rockstar North	9.7
19.39					
2	PS2	Action	Rockstar Games	Rockstar North	9.6
16.15					
3	X360	Action	Rockstar Games	Rockstar North	NaN
15.86					

4	PS4	Shooter	Activision	Treyarch	8.1	15.09
	na_sales	jp_sales	pal_sales	other_sales	release_date	last_update
0	6.37	0.99	9.85	3.12	2013-09-17	NaN
1	6.06	0.60	9.71	3.02	2014-11-18	2018-01-03
2	8.41	0.47	5.49	1.78	2002-10-28	NaN
3	9.06	0.06	5.33	1.42	2013-09-17	NaN
4	6.18	0.41	6.05	2.44	2015-11-06	2018-01-14

```
print(f'We have a data that contains {df.shape[0]} rows and {df.shape[1]} columns')
```

We have a data that contains 64016 rows and 14 columns

```
# To confirm the content of the data
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 64016 entries, 0 to 64015
Data columns (total 14 columns):
#   Column                Non-Null Count  Dtype
---  -
0   img                   64016 non-null  object
1   title                 64016 non-null  object
2   console               64016 non-null  object
3   genre                 64016 non-null  object
4   publisher             64016 non-null  object
5   developer             63999 non-null  object
6   critic_score          6678 non-null   float64
7   total_sales           18922 non-null  float64
8   na_sales              12637 non-null  float64
9   jp_sales              6726 non-null   float64
10  pal_sales             12824 non-null  float64
11  other_sales           15128 non-null  float64
12  release_date          56965 non-null  object
13  last_update           17879 non-null  object
dtypes: float64(6), object(8)
memory usage: 6.8+ MB
```

From the information provided above under the Non-Null count column, it can be seen that a number of columns have null entries. It is not practical to use null values in our research. Therefore, for this analysis we dropped the null values respective for the different analysis in order to improve on the accuracy of the results.

```
# A description of what all the columns contain
df2 = pd.read_csv('vg_data_dictionary.csv')
df2
```

	Field	Description
0	img	URL slug for the box art at vgchartz.com
1	title	Game title
2	console	Console the game was released for
3	genre	Genre of the game
4	publisher	Publisher of the game
5	developer	Developer of the game
6	critic_score	Metacritic score (out of 10)
7	total_sales	Global sales of copies in millions
8	na_sales	North American sales of copies in millions
9	jp_sales	Japanese sales of copies in millions
10	pal_sales	European & African sales of copies in millions
11	other_sales	Rest of world sales of copies in millions
12	release_date	Date the game was released on
13	last_update	Date the data was last updated

1. Top-Selling Titles Worldwide

The first analysis was to get top-selling video games and identify factors that led to their outstanding performance.

```
# drop the null values in the dataset
df_clean = df.dropna()
df_clean
```

	img	\
1	/games/boxart/full_5563178AmericaFrontccc.jpg	
4	/games/boxart/full_4990510AmericaFrontccc.jpg	
7	/games/boxart/full_4653215AmericaFrontccc.jpg	
8	/games/boxart/full_1977964AmericaFrontccc.jpg	
9	/games/boxart/full_4649679AmericaFrontccc.png	
...	...	
11794	/games/boxart/full_2874583AmericaFrontccc.jpg	
11920	/games/boxart/full_4489495AmericaFrontccc.jpg	
12216	/games/boxart/full_5524695AmericaFrontccc.jpg	
13198	/games/boxart/full_5198003AmericaFrontccc.jpg	
15044	/games/boxart/full_8777680PALFrontccc.png	

	title	console	genre	\
1	Grand Theft Auto V	PS4	Action	
4	Call of Duty: Black Ops 3	PS4	Shooter	
7	Red Dead Redemption 2	PS4	Action-Adventure	
8	Call of Duty: Black Ops II	X360	Shooter	
9	Call of Duty: Black Ops II	PS3	Shooter	
...	

11794	RPG Maker: Fes	3DS	Role-Playing
11920	BlazBlue: Cross Tag Battle	NS	Fighting
12216	Disgaea 1 Complete	NS	Role-Playing
13198	GrimGrimoire	PS2	Strategy
15044	Tenkai Knights: Brave Battle	3DS	Action

	publisher	developer	critic_score
total_sales \			
1	Rockstar Games	Rockstar North	9.7
19.39			
4	Activision	Treyarch	8.1
15.09			
7	Rockstar Games	Rockstar Games	9.8
13.94			
8	Activision	Treyarch	8.4
13.86			
9	Activision	Treyarch	8.0
13.80			
...
...			
11794	NIS America	Kadokawa Games	7.0
0.07			
11920	Arc System Works	Arc System Works	8.0
0.07			
12216	NIS America	Nippon Ichi Software	8.0
0.06			
13198	NIS America	Vanillaware	7.7
0.05			
15044	Namco Bandai Games	Delta Factory	4.5
0.02			

	na_sales	jp_sales	pal_sales	other_sales	release_date	
last_update						
1	6.06	0.60	9.71	3.02	2014-11-18	2018-
01-03						
4	6.18	0.41	6.05	2.44	2015-11-06	2018-
01-14						
7	5.26	0.21	6.21	2.26	2018-10-26	2018-
11-02						
8	8.27	0.07	4.32	1.20	2012-11-13	2018-
04-07						
9	4.99	0.65	5.88	2.28	2012-11-13	2018-
04-07						
...
...						
11794	0.03	0.02	0.00	0.00	2017-06-27	2018-
04-22						
11920	0.04	0.01	0.01	0.01	2018-06-05	2018-
06-06						

12216	0.03	0.01	0.01	0.00	2018-10-09	2019-02-19
13198	0.01	0.03	0.01	0.00	2007-06-26	2019-01-10
15044	0.01	0.00	0.01	0.00	2014-10-07	2019-01-24

[282 rows x 14 columns]

creating a dataframe of each game ranked from highest to lowest on total sales

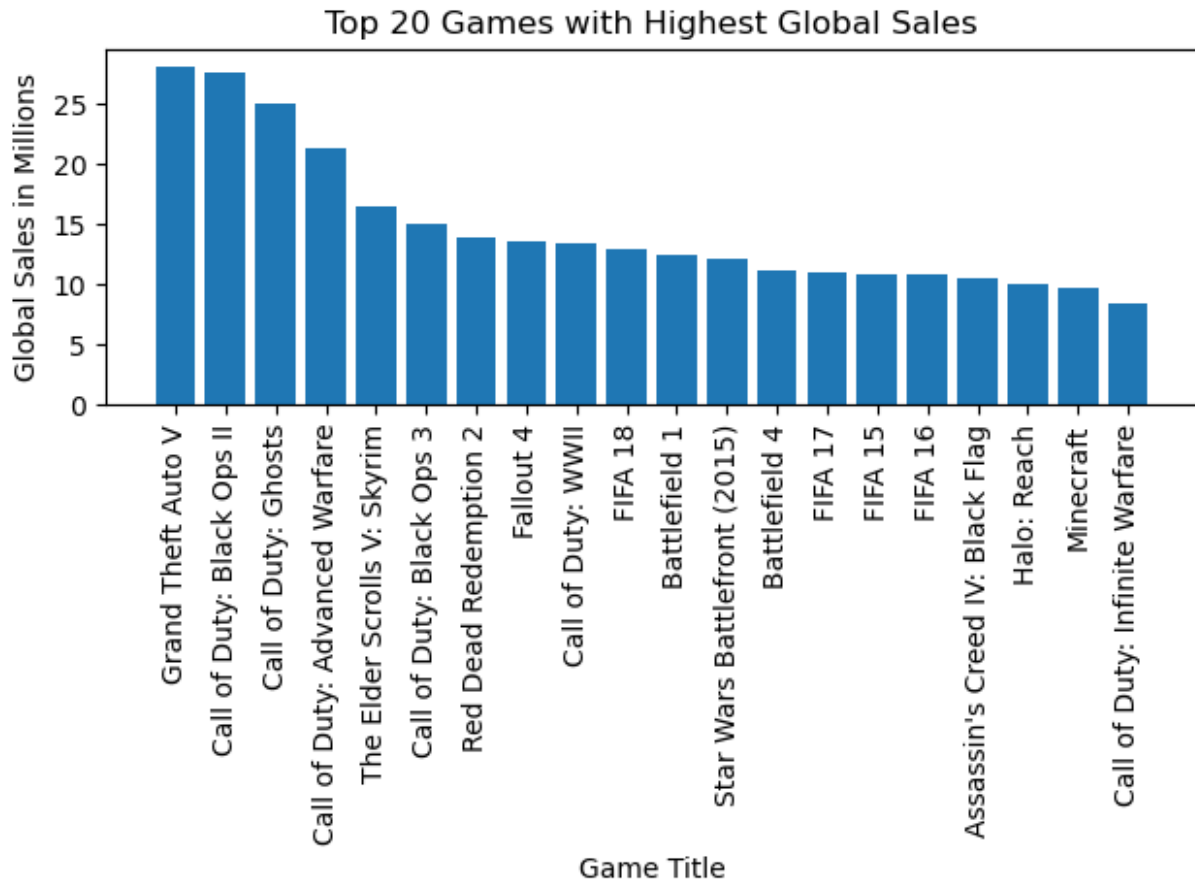
```
top_games = df_clean.groupby(['title'])
['total_sales'].sum().sort_values(ascending = False)
top_20_games = pd.DataFrame(top_games[:20]) # top 20 games
top_20_games
```

	total_sales
title	
Grand Theft Auto V	28.11
Call of Duty: Black Ops II	27.66
Call of Duty: Ghosts	25.06
Call of Duty: Advanced Warfare	21.36
The Elder Scrolls V: Skyrim	16.52
Call of Duty: Black Ops 3	15.09
Red Dead Redemption 2	13.94
Fallout 4	13.51
Call of Duty: WWII	13.40
FIFA 18	12.90
Battlefield 1	12.39
Star Wars Battlefront (2015)	12.18
Battlefield 4	11.12
FIFA 17	10.94
FIFA 15	10.88
FIFA 16	10.77
Assassin's Creed IV: Black Flag	10.47
Halo: Reach	9.97
Minecraft	9.69
Call of Duty: Infinite Warfare	8.48

The **top_20_games** DataFrame above contains the top 20 games with highest sales. A list of all games with their total sales is in the **top_games** DataFrame. The game with the highest sales is **Grand Theft Auto** with \$ 28.11 million total sales. **Call of Duty** and **FIFA** games are featured multiple times in the top games. The bar graph below shows how each of the top 20 games performed.

```
# a plot of the top games
fig, ax = plt.subplots()
ax.bar(top_20_games.index, top_20_games['total_sales'])
ax.set_ylabel('Global Sales in Millions')
```

```
ax.set_xlabel('Game Title')
ax.set_title('Top 20 Games with Highest Global Sales')
plt.xticks(rotation = 90)
plt.tight_layout();
```



After establishing the top-selling games it is prudent to investigate if any or all of the various game attributes have an impact on the sales. First we investigate the critic score.

```
# Get the critic scores of the top 20 games
highest_sales = df_clean[df_clean['title'].isin(top_20_games.index)]
highest_score = pd.DataFrame(highest_sales.groupby(['title'])
                             ['critic_score'].mean())
highest_score
```

title	critic_score
Assassin's Creed IV: Black Flag	8.500000
Battlefield 1	9.200000
Battlefield 4	8.566667
Call of Duty: Advanced Warfare	8.750000
Call of Duty: Black Ops 3	8.100000
Call of Duty: Black Ops II	8.200000

Call of Duty: Ghosts	7.675000
Call of Duty: Infinite Warfare	7.900000
Call of Duty: WWII	8.100000
FIFA 15	7.500000
FIFA 16	8.750000
FIFA 17	8.900000
FIFA 18	7.050000
Fallout 4	8.500000
Grand Theft Auto V	9.350000
Halo: Reach	9.300000
Minecraft	8.033333
Red Dead Redemption 2	9.800000
Star Wars Battlefront (2015)	7.000000
The Elder Scrolls V: Skyrim	8.966667

```
# get the critic scores of the games outside top 20 games
other_sales = df_clean[~df_clean['title'].isin(top_20_games.index)]
lowest_score = pd.DataFrame(other_sales.groupby(['title'])
['critic_score'].mean())
lowest_score.head()
```

	critic_score
title	
2014 FIFA World Cup Brazil	7.6
7th Dragon III Code: VFD	7.5
Animal Crossing: Amiibo Festival	4.8
Ape Escape	9.3
Ape Escape 2	8.0

```
# plot the difference in critic scores
scores = [highest_score.mean().iloc[0],lowest_score.mean().iloc[0]]
my_labels = ['top games critic score','other games critic score']

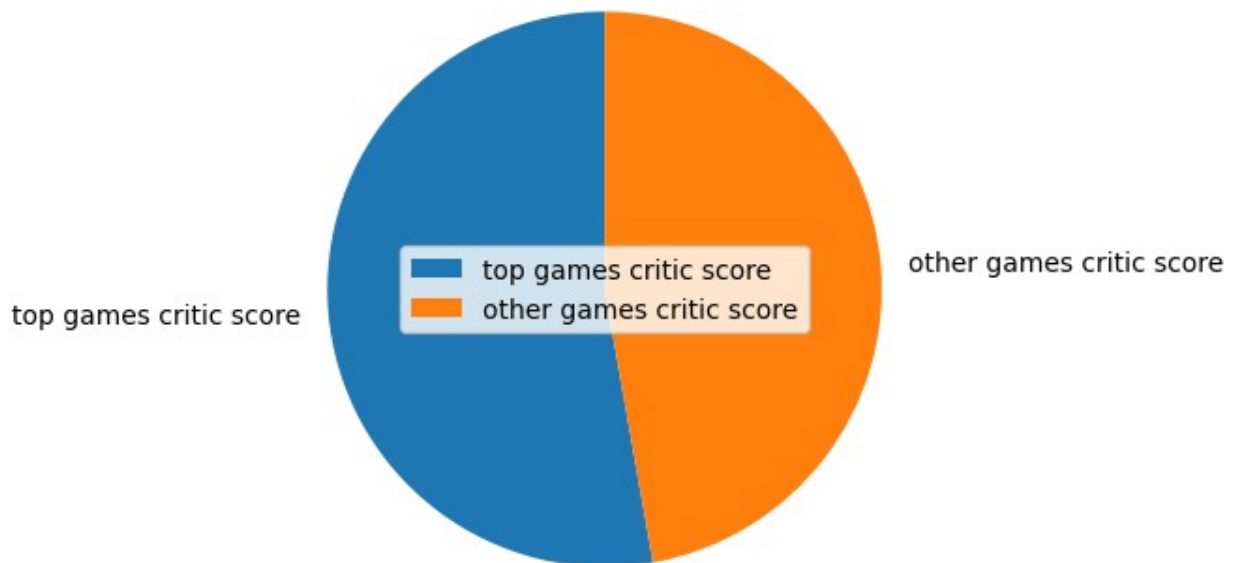
fig, ax = plt.subplots()
ax.pie(scores,labels = my_labels,startangle = 90)
ax.legend()
ax.set_title('Comparison of the critic scores for top games and other
games');
```

```
print(f'''From the results and as visualized in the pie chart below,
the
top 20 games have a higher average critic score of
{round(highest_score.mean().iloc[0],2)} compared to
the other games that have an average critic score of
{round(lowest_score.mean().iloc[0],2)}.
This confirms that the critic score affects the sales of a game. The
higher the critic score the more the sales''')
```

From the results and as visualized in the pie chart below, the top 20 games have a higher average critic score of 8.41 compared to

the other games that have an average critic score of 7.52. This confirms that the critic score affects the sales of a game. The higher the critic score the more the sales

Comparison of the critic scores for top games and other games



After establishing that the critic score affects the sales. We head on to investigate the effect of genre on the sales.

```
genre = highest_sales[['title', 'genre']].drop_duplicates()
genre
```

	title	genre
1	Grand Theft Auto V	Action
4	Call of Duty: Black Ops 3	Shooter
7	Red Dead Redemption 2	Action-Adventure
8	Call of Duty: Black Ops II	Shooter
11	Call of Duty: WWII	Shooter
15	FIFA 18	Sports
17	FIFA 17	Sports
20	Call of Duty: Ghosts	Shooter
22	Halo: Reach	Shooter
27	The Elder Scrolls V: Skyrim	Role-Playing
29	Call of Duty: Infinite Warfare	Shooter
30	Fallout 4	Role-Playing
31	FIFA 16	Sports
32	Star Wars Battlefront (2015)	Shooter
36	Call of Duty: Advanced Warfare	Shooter

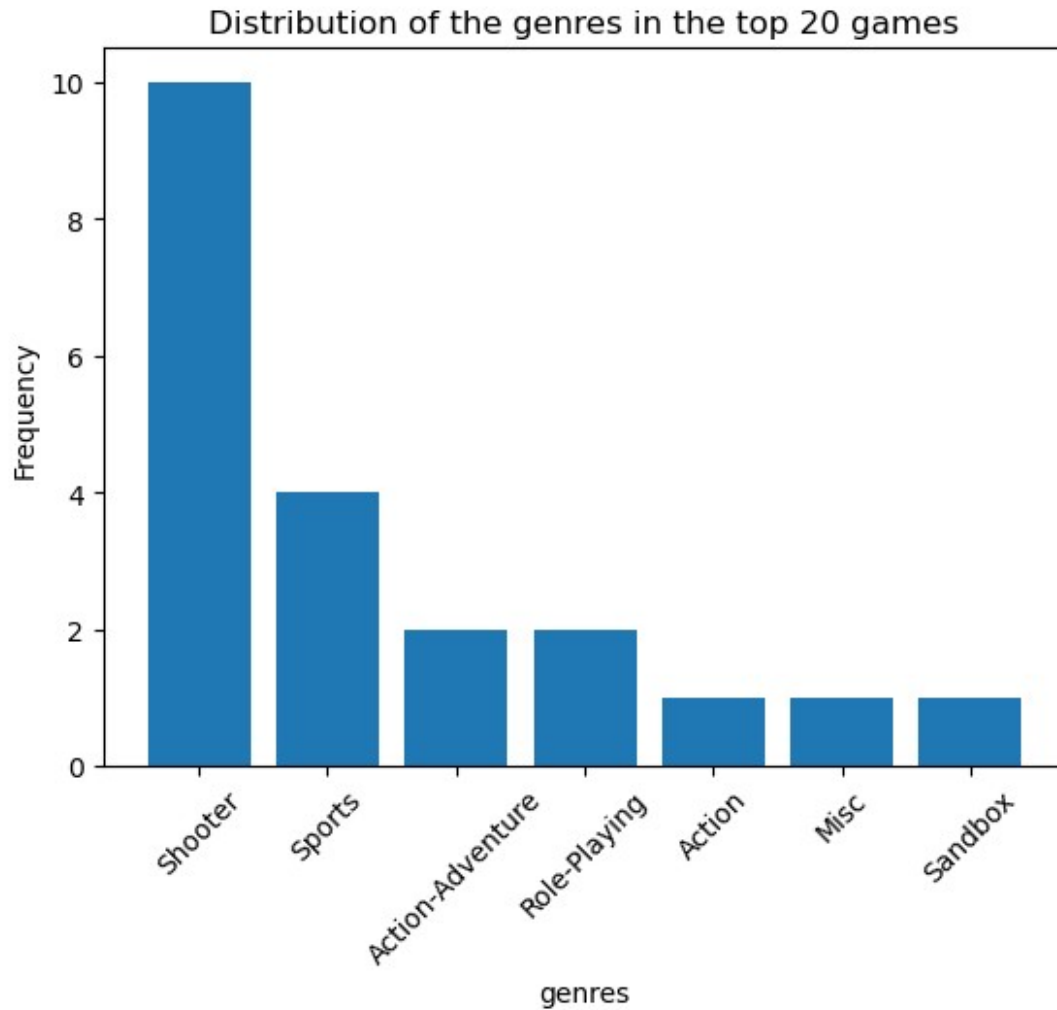
40	Battlefield 1	Shooter
54	Minecraft	Misc
55	FIFA 15	Sports
158	Battlefield 4	Shooter
159	Assassin's Creed IV: Black Flag	Action-Adventure
571	Minecraft	Sandbox

```
top_genres = genre['genre'].value_counts()
top_genres
```

```
genre
Shooter      10
Sports        4
Action-Adventure  2
Role-Playing  2
Action        1
Misc          1
Sandbox       1
Name: count, dtype: int64
```

The most selling genre is shooter, in the top 20 games, half of the games are of the shooter genre.

```
#plot of the top genres
fig,ax = plt.subplots()
ax.bar(top_genres.index,top_genres)
ax.set_xlabel('genres')
ax.set_ylabel('Frequency')
ax.set_title('Distribution of the genres in the top 20 games')
plt.xticks(rotation = 45);
```



```
console = highest_sales[['title', 'console']].drop_duplicates()
console
```

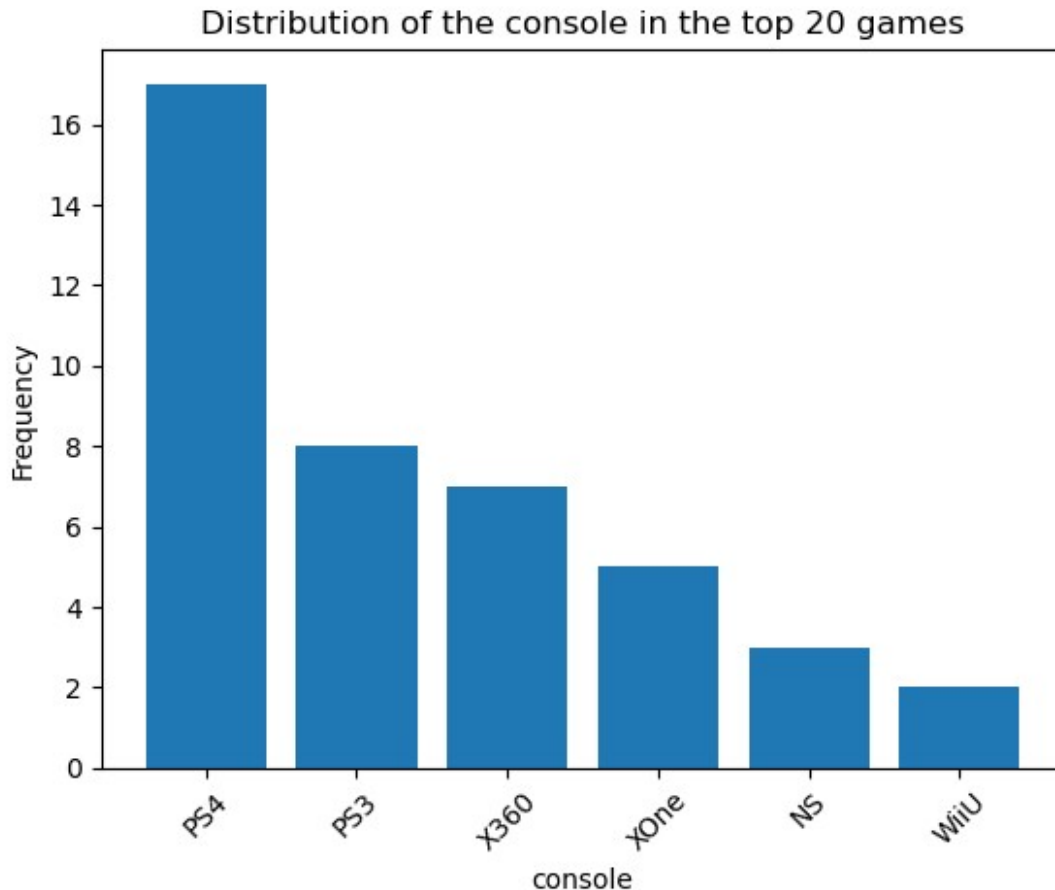
	title	console
1	Grand Theft Auto V	PS4
4	Call of Duty: Black Ops 3	PS4
7	Red Dead Redemption 2	PS4
8	Call of Duty: Black Ops II	X360
9	Call of Duty: Black Ops II	PS3
11	Call of Duty: WWII	PS4
15	FIFA 18	PS4
17	FIFA 17	PS4
20	Call of Duty: Ghosts	X360
21	Call of Duty: Ghosts	PS3
22	Halo: Reach	X360
27	The Elder Scrolls V: Skyrim	X360
28	Grand Theft Auto V	X0ne
29	Call of Duty: Infinite Warfare	PS4
30	Fallout 4	PS4

31	FIFA 16	PS4
32	Star Wars Battlefront (2015)	PS4
36	Call of Duty: Advanced Warfare	PS4
40	Battlefield 1	PS4
53	The Elder Scrolls V: Skyrim	PS3
54	Minecraft	PS4
55	FIFA 15	PS4
76	Call of Duty: Advanced Warfare	X0ne
78	Battlefield 1	X0ne
88	Fallout 4	X0ne
109	FIFA 15	PS3
123	Call of Duty: Advanced Warfare	X360
127	Call of Duty: Advanced Warfare	PS3
134	Call of Duty: Ghosts	PS4
138	Star Wars Battlefront (2015)	X0ne
158	Battlefield 4	PS4
159	Assassin's Creed IV: Black Flag	PS3
174	Battlefield 4	PS3
190	Battlefield 4	X360
219	Assassin's Creed IV: Black Flag	X360
234	Assassin's Creed IV: Black Flag	PS4
350	FIFA 16	PS3
571	Minecraft	NS
849	Minecraft	WiiU
1224	The Elder Scrolls V: Skyrim	NS
1299	FIFA 18	NS
4644	Call of Duty: Ghosts	WiiU

```
top_console = console['console'].value_counts()
top_console
```

```
console
PS4      17
PS3       8
X360      7
X0ne      5
NS         3
WiiU      2
Name: count, dtype: int64
```

```
#plot of the top consoles
fig,ax = plt.subplots()
ax.bar(top_console.index,top_console)
ax.set_xlabel('console')
ax.set_ylabel('Frequency')
ax.set_title('Distribution of the console in the top 20 games')
plt.xticks(rotation = 45);
```



The top console is PS4 indicating that the console affects the sales.

In conclusion the critic score, the genre and the console for which the game is made for affects the sales of a game. This is in line with the assumptions that the more a game is rated highly the more it will sell. A genre that is most liked will definitely sell more and finally the console that is most owned and most liked will have games made for them highly purchased.

2. Sales Trend Overtime

The second point of investigation is to analyze how the video games sales have improved over time.

```
# create a DataFrame with the total sales per year
sales_by_year = df[['total_sales', 'release_date']].dropna()
sales_by_year.loc[:, 'release_date'] =
sales_by_year.loc[:, 'release_date'].str[:4].astype('int')
sales_by_year.head()
```

	total_sales	release_date
0	20.32	2013
1	19.39	2014
2	16.15	2002

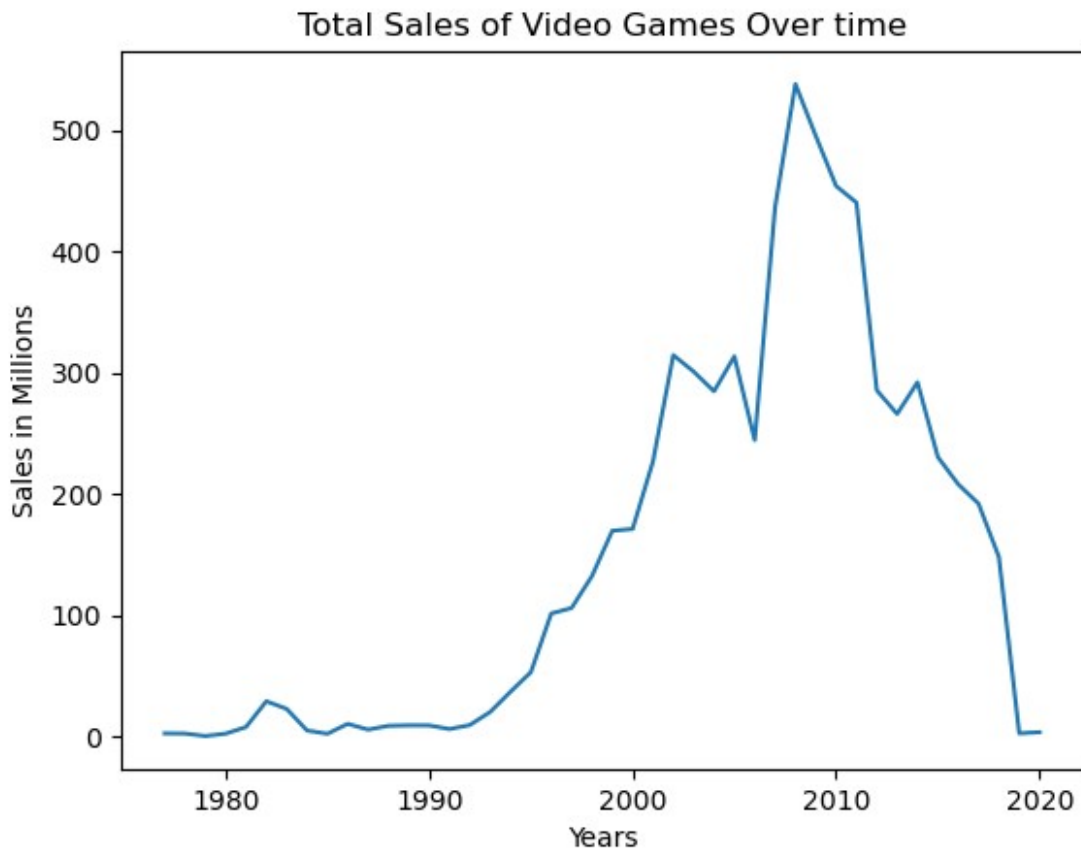
3	15.86	2013
4	15.09	2015

```

sales_trend = sales_by_year.groupby(['release_date'])
['total_sales'].sum()

fig, ax = plt.subplots()
ax.plot(sales_trend.index,sales_trend)
ax.set_title('Total Sales of Video Games Over time')
ax.set_xlabel('Years')
ax.set_ylabel('Sales in Millions');

```



The line graph above shows the trend of the sales of the games over time from the 70s to 2024. There was a steady rise in the sale of the games from the 1970s and it peaked at around 2010. But it has since experienced steady decline since then, hitting a low in 2020 during the pandemic.

3. Console Specialization in Genres

Here we investigate whether consoles specialize in a particular genre.

```
# A data frame of genres and consoles
genre_console = df[['genre', 'console']]
genre_console.head()
```

```
   genre console
0  Action    PS3
1  Action    PS4
2  Action    PS2
3  Action   X360
4 Shooter    PS4
```

```
# different genres classified to each console
```

```
g_c = pd.crosstab(index = genre_console['console'], columns =
genre_console['genre'])
g_c
```

```
genre      Action  Action-Adventure  Adventure  Board Game  Education
Fighting \
console
```

```
2600      302          0          3          0          0
3
```

```
3D0       12          0         43          0          0
13
```

```
3DS       198         34         63          0          1
18
```

```
5200      26          0          2          0          0
0
```

```
7800      21          0          0          0          0
3
```

```
...      ...          ...          ...          ...          ...
...
```

```
XOne      309        211        216          2          1
91
```

```
XS        63        121         94          1          0
18
```

```
ZXS        5          0          0          0          0
0
```

```
iOS       20         14         16          0          0
0
```

```
iQue       2          0          1          0          0
1
```

```
genre      MM0  Misc  Music  Party  Platform  Puzzle  Racing  Role-
Playing \
console
```

```
2600      0    21     0     0      17     19     19
4
```

```
3D0      0    88     0     0      7     13     10
```

```

10
3DS      1   382   11    8    76   78   22
125
5200     0    0    0    0    9    6    2
0
7800     0    0    0    0    6    3    3
0
...      ...   ...   ...   ...   ...   ...   .
..
XOne     10   67   27   10  120   67  129
197
XS        4    5    5    5   35   17   36
75
ZXS       0   13    0    0    1    0    0
0
iOS       0   21    1    0    9   13    4
24
iQue      0    0    0    0    2    2    4
2

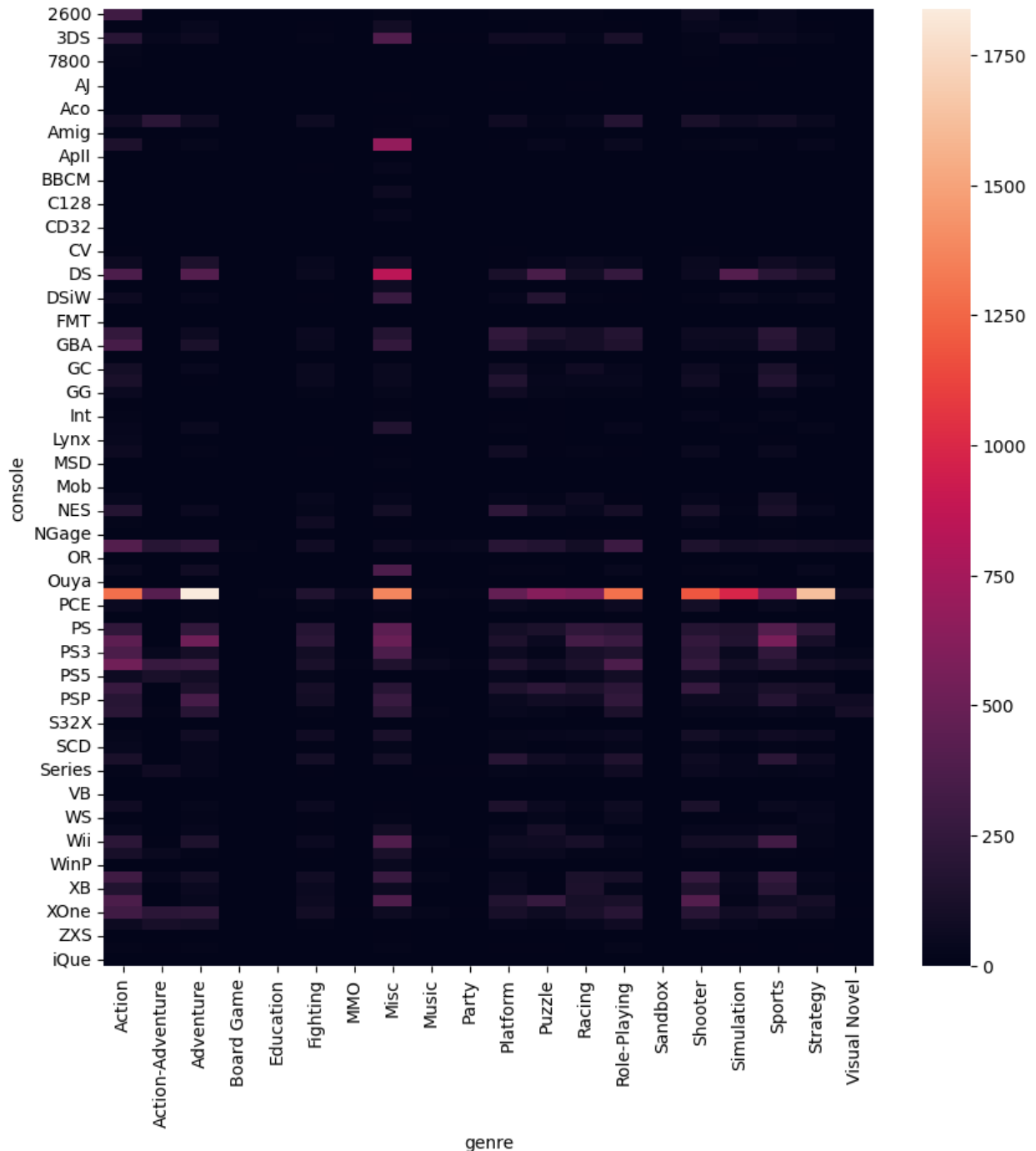
genre    Sandbox  Shooter  Simulation  Sports  Strategy  Visual Novel
console

2600      0     68      4    39     2      0
3D0       0     37     23    35    14      0
3DS       0     20     75    49    26      5
5200      0     20      1     7     0      0
7800      0      9      2    12     0      0
...       ...   ...   ...   ...   ...   ...
XOne       3    201     81   137    71     13
XS         0     65     37    16    20      3
ZXS        0      0      0     0     0      0
iOS        0      7     11     3    11      1
iQue       0      2      1     0     0      0

[81 rows x 20 columns]

plt.figure(figsize = (10,10))
sns.heatmap(g_c);

```



The heatmap above indicates that very few consoles have specialized in a particular genre. PC is one console that is shown to majorly specialize in Action, Strategy and Misc genres. Most consoles however take all genres shown.

4. Regional Popularity and Flops

Lastly we look into the games performance regionally to investigate any variance.


```

popularity = df.groupby(['title']
                        )
[['na_sales', 'jp_sales', 'pal_sales', 'other_sales', 'total_sales']].sum(
).sort_values(
                        by = 'total_sales', ascending = False)
popularity = popularity[popularity['total_sales']>0]
popularity.head()

```

	na_sales	jp_sales	pal_sales
other_sales \ title			
Grand Theft Auto V	26.19	1.66	28.14
8.32			
Call of Duty: Black Ops	17.65	0.59	9.45
3.31			
Call of Duty: Modern Warfare 3	15.57	0.62	11.26
3.26			
Call of Duty: Black Ops II	14.12	0.72	11.08
3.67			
Call of Duty: Ghosts	15.06	0.49	9.60
3.65			

	total_sales
title	
Grand Theft Auto V	64.29
Call of Duty: Black Ops	30.99
Call of Duty: Modern Warfare 3	30.71
Call of Duty: Black Ops II	29.59
Call of Duty: Ghosts	28.80

```

popularity['disparity'] =
popularity[['na_sales', 'jp_sales', 'pal_sales', 'other_sales']].std(axis
= 1)
popularity.head()

```

	na_sales	jp_sales	pal_sales
other_sales \ title			
Grand Theft Auto V	26.19	1.66	28.14
8.32			
Call of Duty: Black Ops	17.65	0.59	9.45
3.31			
Call of Duty: Modern Warfare 3	15.57	0.62	11.26
3.26			
Call of Duty: Black Ops II	14.12	0.72	11.08
3.67			
Call of Duty: Ghosts	15.06	0.49	9.60
3.65			

	total_sales	disparity
title		
Grand Theft Auto V	64.29	13.112458
Call of Duty: Black Ops	30.99	7.569218
Call of Duty: Modern Warfare 3	30.71	6.938983
Call of Duty: Black Ops II	29.59	6.251295
Call of Duty: Ghosts	28.80	6.459262

```
title_variation = popularity[popularity['disparity']>3].sort_values(by
= 'disparity', ascending = False)
title_variation
```

	na_sales	jp_sales	pal_sales
other_sales \			
title			
Grand Theft Auto V	26.19	1.66	28.14
8.32			
Call of Duty: Black Ops	17.65	0.59	9.45
3.31			
Call of Duty: Modern Warfare 3	15.57	0.62	11.26
3.26			
Call of Duty: Ghosts	15.06	0.49	9.60
3.65			
Call of Duty: Black Ops II	14.12	0.72	11.08
3.67			
Call of Duty: Modern Warfare 2	13.54	0.46	8.08
2.95			
Call of Duty: Black Ops 3	12.82	0.50	9.76
3.63			
FIFA 15	3.19	0.15	12.45
2.21			
FIFA 14	2.88	0.21	12.14
2.09			
FIFA 17	2.64	0.16	12.07
2.15			
FIFA 18	2.71	0.29	11.78
2.13			
Grand Theft Auto IV	11.60	0.58	7.64
2.72			
Guitar Hero III: Legends of Rock	11.14	0.04	2.57
2.64			
Call of Duty: Advanced Warfare	10.66	0.35	7.99
2.81			
FIFA 16	3.16	0.12	10.56
1.99			
The Elder Scrolls V: Skyrim	9.52	0.41	8.08
2.50			
FIFA 13	2.67	0.18	10.11

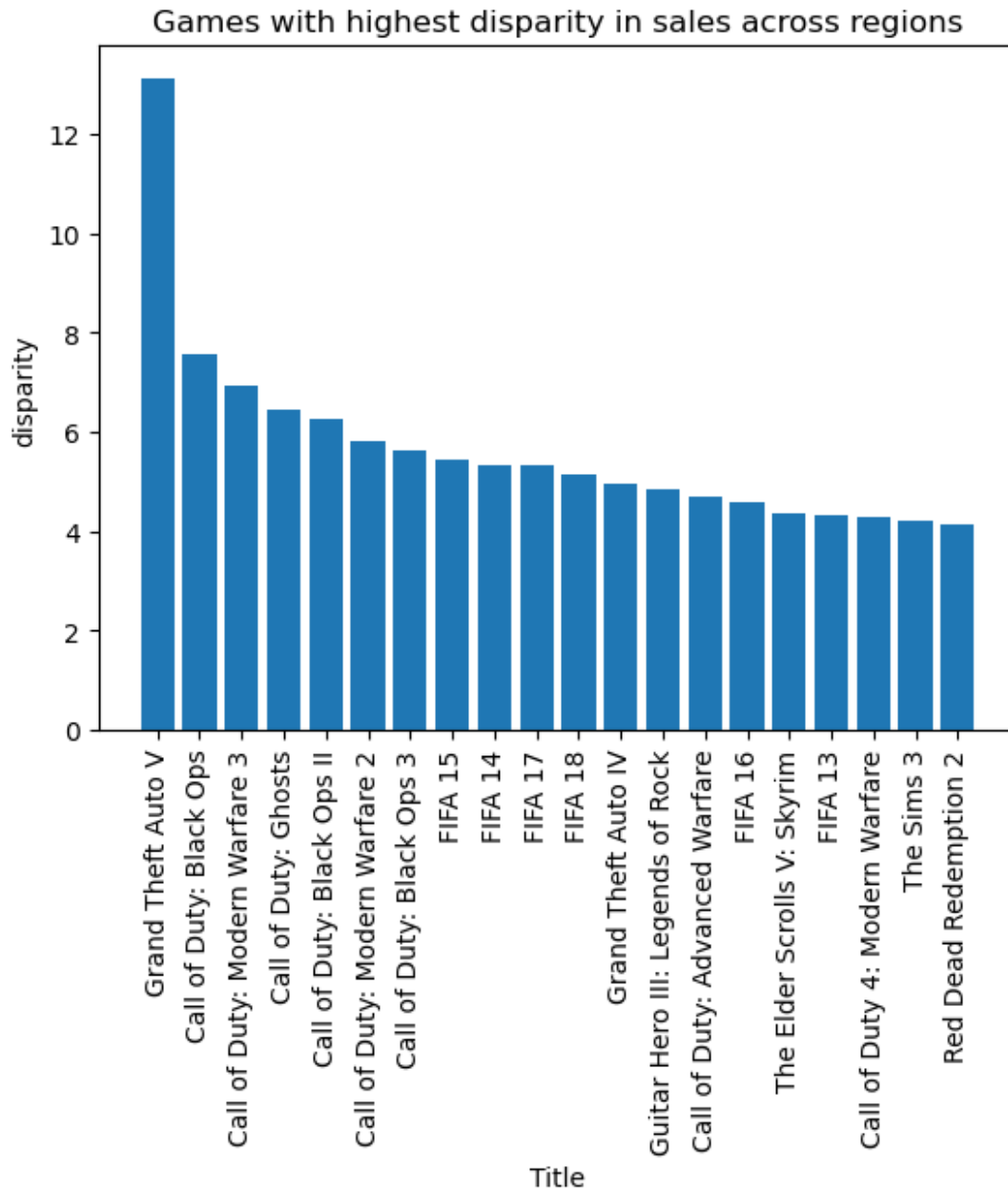
2.42			
Call of Duty 4: Modern Warfare	10.06	0.42	5.80
2.06			
The Sims 3	4.31	0.05	9.55
1.29			
Red Dead Redemption 2	9.02	0.21	7.68
2.80			
Call of Duty: World at War	9.38	0.00	4.72
1.83			
Call of Duty: WWII	8.49	0.40	8.23
2.70			
FIFA 19	1.58	0.16	8.93
1.56			
LEGO Star Wars: The Complete Saga	8.95	0.00	4.82
1.57			
Minecraft	9.07	2.50	9.48
2.96			
Battlefield 3	8.22	0.41	6.50
2.19			
Madden NFL 06	7.68	0.02	0.46
0.86			
FIFA Soccer 12	2.08	0.15	8.51
2.31			
Grand Theft Auto: Vice City	8.41	0.47	5.52
1.78			
Madden NFL 08	7.13	0.00	0.23
0.78			
FIFA Soccer 11	1.95	0.09	7.99
2.47			
Madden NFL 2005	6.85	0.01	0.53
0.21			
LEGO Batman: The Videogame	7.67	0.00	3.95
1.91			
Halo: Reach	7.08	0.08	2.01
0.80			
Madden NFL 10	6.55	0.00	0.34
0.68			
Grand Theft Auto III	6.99	0.30	4.52
1.30			
Call of Duty: Black Ops II	7.37	0.50	4.47
1.96			

	total_sales	disparity
title		
Grand Theft Auto V	64.29	13.112458
Call of Duty: Black Ops	30.99	7.569218
Call of Duty: Modern Warfare 3	30.71	6.938983
Call of Duty: Ghosts	28.80	6.459262
Call of Duty: Black Ops II	29.59	6.251295

Call of Duty: Modern Warfare 2	25.02	5.799623
Call of Duty: Black Ops 3	26.72	5.617855
FIFA 15	18.03	5.449318
FIFA 14	17.31	5.325743
FIFA 17	17.02	5.319226
FIFA 18	16.92	5.139600
Grand Theft Auto IV	22.53	4.954846
Guitar Hero III: Legends of Rock	16.38	4.848287
Call of Duty: Advanced Warfare	21.78	4.710813
FIFA 16	15.82	4.576260
The Elder Scrolls V: Skyrim	20.51	4.365389
FIFA 13	15.36	4.324107
Call of Duty 4: Modern Warfare	18.33	4.288539
The Sims 3	15.20	4.230256
Red Dead Redemption 2	19.71	4.127383
Call of Duty: World at War	15.94	4.089437
Call of Duty: WWII	19.82	4.043715
FIFA 19	12.22	3.971032
LEGO Star Wars: The Complete Saga	15.33	3.956889
Minecraft	24.01	3.787122
Battlefield 3	17.32	3.641749
Madden NFL 06	9.03	3.632901
FIFA Soccer 12	13.06	3.629944
Grand Theft Auto: Vice City	16.19	3.612003
Madden NFL 08	8.14	3.412394
FIFA Soccer 11	12.53	3.400446
Madden NFL 2005	7.60	3.306942
LEGO Batman: The Videogame	13.55	3.281985
Halo: Reach	9.97	3.160310
Madden NFL 10	7.57	3.117385
Grand Theft Auto III	13.11	3.060636
Call of Duty: Black Ops IIII	14.30	3.014791

The table above shows the games with the most variation with Grand Theft Auto V on top of the list with a standard deviation of \$ 13 million. The sales in Japan and other parts of the world apart from North America, Europe and Africa, were low. This can be attributed to population differences. This is the case for most of the other differences. The effect of availability factor or preferences is not significant on the sales. Most of the games had a standard deviation below 2 million showing that most did not experience disparities across regions.

```
fig,ax = plt.subplots()
ax.bar(title_variation.index[:20],title_variation['disparity'][:20])
ax.set_title('Games with highest disparity in sales across regions')
ax.set_xlabel('Title')
ax.set_ylabel('disparity')
plt.xticks(rotation = 90);
```



```
popularity1 = df.groupby(['console']
                          )
[['na_sales', 'jp_sales', 'pal_sales', 'other_sales', 'total_sales']].sum(
).sort_values(
    by = 'total_sales', ascending = False)
popularity1 = popularity1[popularity1['total_sales']>0]
popularity1.head()
```

	na_sales	jp_sales	pal_sales	other_sales	total_sales
console					
PS2	489.96	87.63	296.37	153.94	1027.76

X360	528.40	10.16	244.50	76.74	859.79
PS3	351.10	67.68	302.22	118.93	839.70
PS	263.26	88.32	163.73	31.10	546.25
PS4	192.18	30.69	233.40	83.55	539.92

```
popularity1['disparity'] =
popularity1[['na_sales', 'jp_sales', 'pal_sales', 'other_sales']].std(axis = 1)
popularity1.head()
```

	na_sales	jp_sales	pal_sales	other_sales	total_sales
disparity					
console					
PS2	489.96	87.63	296.37	153.94	1027.76
178.071069					
X360	528.40	10.16	244.50	76.74	859.79
231.059194					
PS3	351.10	67.68	302.22	118.93	839.70
137.795095					
PS	263.26	88.32	163.73	31.10	546.25
100.399205					
PS4	192.18	30.69	233.40	83.55	539.92
93.949956					

```
console_variation =
popularity1[popularity1['disparity']>3].sort_values(by = 'disparity',
ascending = False)
console_variation
```

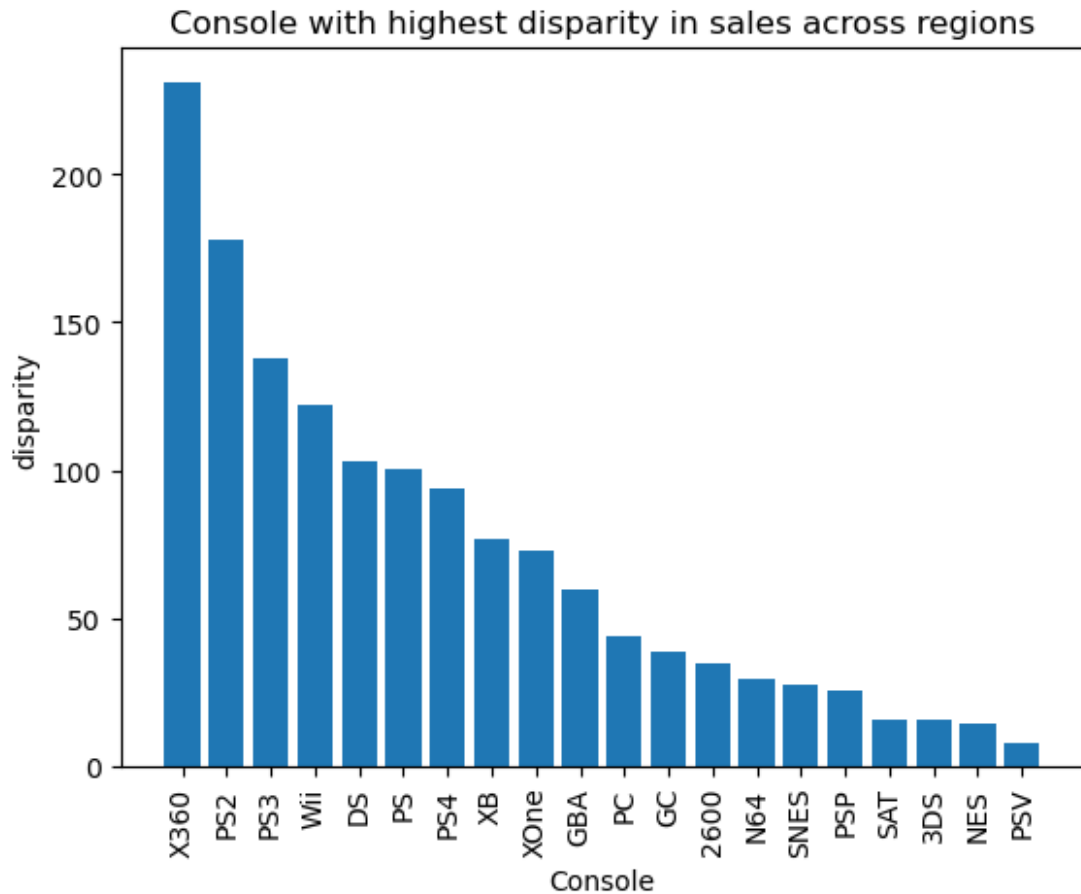
	na_sales	jp_sales	pal_sales	other_sales	total_sales
disparity					
console					
X360	528.40	10.16	244.50	76.74	859.79
231.059194					
PS2	489.96	87.63	296.37	153.94	1027.76
178.071069					
PS3	351.10	67.68	302.22	118.93	839.70
137.795095					
Wii	285.42	16.84	117.72	38.85	459.44
121.770605					
DS	264.84	64.20	94.33	33.88	458.17
103.341631					
PS	263.26	88.32	163.73	31.10	546.25
100.399205					
PS4	192.18	30.69	233.40	83.55	539.92
93.949956					
XB	166.88	0.83	55.62	8.17	232.05
76.620045					

XOne	164.53	0.38	79.44	24.53	268.96
72.820753					
GBA	140.41	23.94	54.86	4.47	224.48
60.026401					
PC	58.61	0.00	96.02	13.51	168.99
43.854790					
GC	86.31	6.09	23.77	2.90	119.53
38.796524					
2600	70.62	0.00	4.25	0.69	75.66
34.536911					
N64	66.36	8.44	16.67	2.10	93.79
29.259448					
SNES	6.84	57.14	1.46	0.26	65.71
27.293723					
PSP	96.56	52.29	59.79	36.59	245.29
25.412016					
SAT	0.72	32.34	0.54	0.07	33.67
15.950687					
3DS	30.37	41.74	22.24	4.88	99.27
15.506432					
NES	13.37	31.96	2.24	0.36	47.93
14.503155					
PSV	13.04	26.95	15.18	7.81	63.02
8.086048					
GB	1.84	16.87	1.00	0.14	19.84
7.968615					
NS	17.71	4.26	11.63	2.75	36.46
6.934947					
WiiU	16.44	3.84	12.39	2.73	35.42
6.650669					
GEN	12.55	2.05	3.64	0.61	18.83
5.369546					
DC	4.50	7.37	1.22	0.20	13.31
3.262886					

```

fig,ax = plt.subplots()
ax.bar(console_variation.index[:20],console_variation['disparity']
[:20])
ax.set_title('Console with highest disparity in sales across regions')
ax.set_xlabel('Console')
ax.set_ylabel('disparity')
plt.xticks(rotation = 90);

```



Similar to the games, the consoles sales disparity is attributed to the population density differences. X360 has the highest disparity.

```
popularity2 = df.groupby(['genre']
                        )
[['na_sales', 'jp_sales', 'pal_sales', 'other_sales', 'total_sales']].sum(
).sort_values(
    by = 'total_sales', ascending = False)
popularity2 = popularity2[popularity2['total_sales']>0]
popularity2.head()
```

genre	na_sales	jp_sales	pal_sales	other_sales	total_sales
Sports	607.47	109.27	341.92	128.19	1187.51
Action	589.69	80.30	342.52	112.41	1125.89
Shooter	528.27	33.99	324.74	108.43	995.50
Misc	295.66	56.24	146.71	58.52	557.79
Racing	269.72	20.23	179.14	56.44	525.75

```
popularity2['disparity'] =
popularity2[['na_sales', 'jp_sales', 'pal_sales', 'other_sales']].std(axes=1)
```



```
s = 1)
popularity2.head()
```

	na_sales	jp_sales	pal_sales	other_sales	total_sales
disparity					
genre					
Sports	607.47	109.27	341.92	128.19	1187.51
232.485459					
Action	589.69	80.30	342.52	112.41	1125.89
236.486147					
Shooter	528.27	33.99	324.74	108.43	995.50
223.396079					
Misc	295.66	56.24	146.71	58.52	557.79
112.439211					
Racing	269.72	20.23	179.14	56.44	525.75
114.585193					

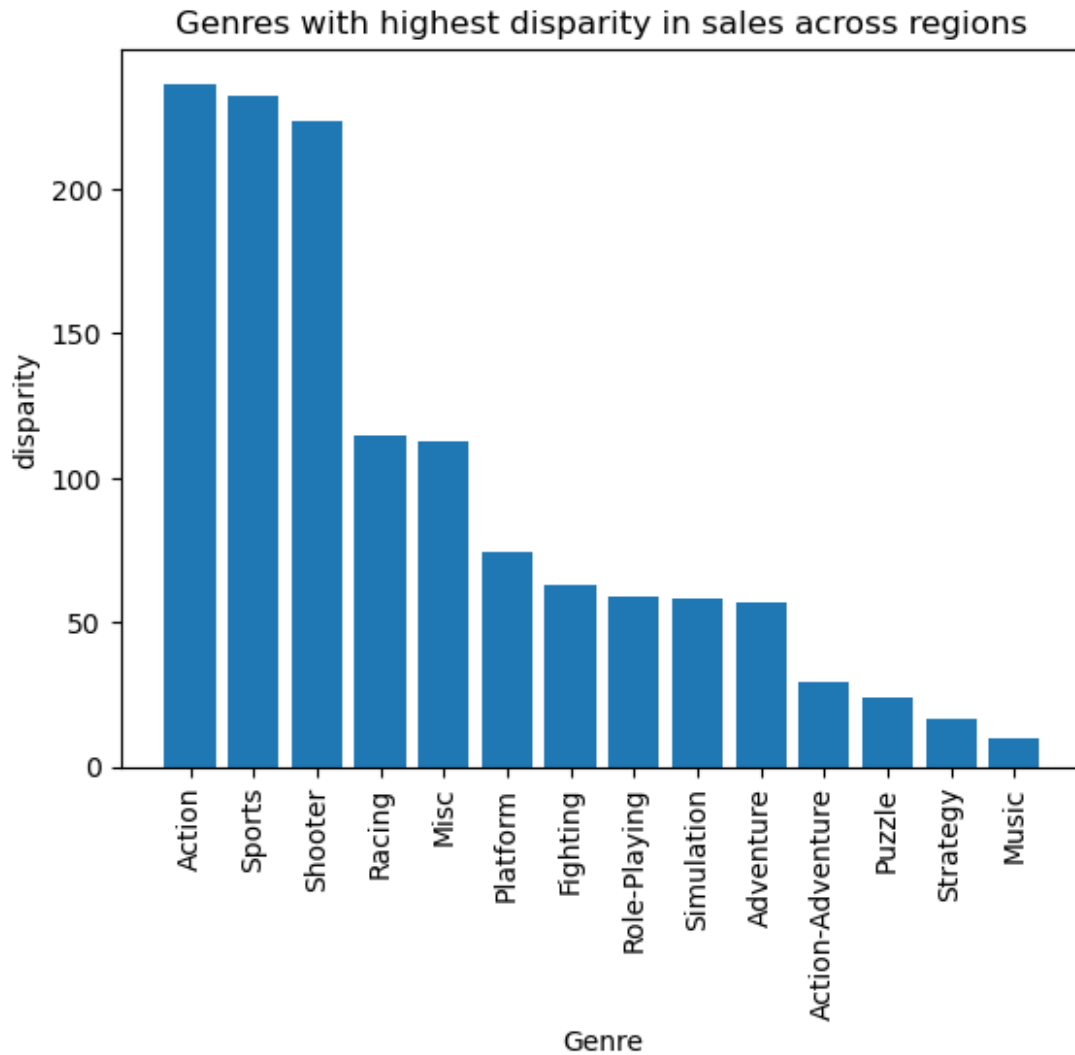
```
genre_variation =
popularity2[popularity2['disparity']>3].sort_values(by = 'disparity',
ascending = False)
genre_variation
```

	na_sales	jp_sales	pal_sales	other_sales
total_sales \				
genre				
Action	589.69	80.30	342.52	112.41
1125.89				
Sports	607.47	109.27	341.92	128.19
1187.51				
Shooter	528.27	33.99	324.74	108.43
995.50				
Racing	269.72	20.23	179.14	56.44
525.75				
Misc	295.66	56.24	146.71	58.52
557.79				
Platform	187.96	31.08	99.39	30.38
349.15				
Fighting	173.96	58.61	79.84	28.65
341.13				
Role-Playing	171.67	130.56	91.02	33.75
426.80				
Simulation	152.84	35.98	86.86	24.34
300.65				
Adventure	157.26	46.11	91.52	30.11
325.39				
Action-Adventure	65.09	5.46	59.52	18.40
148.52				
Puzzle	64.36	29.27	25.78	7.38

127.28				
Strategy	47.06	36.80	27.05	7.29
118.31				
Music	25.97	5.84	15.02	4.93
51.75				

	disparity
genre	
Action	236.486147
Sports	232.485459
Shooter	223.396079
Racing	114.585193
Misc	112.439211
Platform	74.563486
Fighting	62.748123
Role-Playing	58.759270
Simulation	58.562562
Adventure	56.959378
Action-Adventure	29.647231
Puzzle	23.798384
Strategy	16.940289
Music	9.809475

```
fig,ax = plt.subplots()
ax.bar(genre_variation.index[:20],genre_variation['disparity'][:20])
ax.set_title('Genres with highest disparity in sales across regions')
ax.set_xlabel('Genre')
ax.set_ylabel('disparity')
plt.xticks(rotation = 90);
```



Action genre has the highest disparity. Most sales are in North America as that is where most of the video games are produced. Otherwise, the population differences mainly leads to the disparity in sales.

Conclusion

The analysis clarifies most of the general assumptions and we discovered more findings in the course of the analysis.