

eda-on-video-games-sales

November 13, 2023

```
[6]: ## importing required libraries
import pandas as pd
from scipy.stats import ttest_ind, f_oneway
from sklearn.decomposition import PCA
from scipy.stats import chi2_contingency
import numpy as np
import warnings
warnings.filterwarnings('ignore')
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[7]: ## loading the dataset
df = pd.read_csv('vgsales.csv')
```

```
[203]: ## Viewing the data
print(df.head())
```

	Rank	Name	Platform	Year	Genre	Publisher	\
0	1	Wii Sports	Wii	2006.0	Sports	Nintendo	
1	2	Super Mario Bros.	NES	1985.0	Platform	Nintendo	
2	3	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	
3	4	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	
4	5	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo	

	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
0	41.49	29.02	3.77	8.46	82.74
1	29.08	3.58	6.81	0.77	40.24
2	15.85	12.88	3.79	3.31	35.82
3	15.75	11.01	3.28	2.96	33.00
4	11.27	8.89	10.22	1.00	31.37

0.1 Data Understanding

```
[202]: ## Shape of the dataset
print(df.shape)
```

(16598, 11)

```
[201]: print(df.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 16598 entries, 0 to 16597
Data columns (total 11 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Rank            16598 non-null  int64
1   Name            16598 non-null  object
2   Platform        16598 non-null  object
3   Year            16598 non-null  float64
4   Genre           16598 non-null  object
5   Publisher       16598 non-null  object
6   NA_Sales        16598 non-null  float64
7   EU_Sales        16598 non-null  float64
8   JP_Sales        16598 non-null  float64
9   Other_Sales     16598 non-null  float64
10  Global_Sales    16598 non-null  float64
dtypes: float64(6), int64(1), object(4)
memory usage: 1.4+ MB
None
```

```
[204]: print(df.dtypes)
```

```
Rank            int64
Name            object
Platform        object
Year            float64
Genre           object
Publisher       object
NA_Sales        float64
EU_Sales        float64
JP_Sales        float64
Other_Sales     float64
Global_Sales    float64
dtype: object
```

```
[205]: print(df.Year.value_counts())
```

```
Year
2007.0    1473
2009.0    1431
2008.0    1428
2010.0    1259
2011.0    1139
2006.0    1008
2005.0     941
2002.0     829
```

2003.0	775
2004.0	763
2012.0	657
2015.0	614
2014.0	582
2013.0	546
2001.0	482
1998.0	379
2000.0	349
2016.0	344
1999.0	338
1997.0	289
1996.0	263
1995.0	219
1994.0	121
1993.0	60
1981.0	46
1992.0	43
1991.0	41
1982.0	36
1986.0	21
1989.0	17
1983.0	17
1990.0	16
1987.0	16
1988.0	15
1985.0	14
1984.0	14
1980.0	9
2017.0	3
2020.0	1

Name: count, dtype: int64

```
[46]: print(df.Platform.value_counts())
```

```
[46]: Platform
DS      2163
PS2     2161
PS3     1329
Wii     1325
X360    1265
PSP     1213
PS      1196
PC       960
XB       824
GBA      822
GC       556
```

3DS	509
PSV	413
PS4	336
N64	319
SNES	239
XOne	213
SAT	173
WiiU	143
2600	133
NES	98
GB	98
DC	52
GEN	27
NG	12
SCD	6
WS	6
3DO	3
TG16	2
GG	1
PCFX	1

Name: count, dtype: int64

1 Information of Columns

Rank - Depicts the video game which is having the high overall sales

Name - Names of the Video Games

Platform - In which the Video Game is supported

Year - Year in which the video game is released

Genre - The most widely used game classifying system categorizes games into eight genres: action, adventure, fighting, puzzle, role-playing, simulation, sports, and strategy.

Publisher - The company which released the game

NA_Sales - Sales of Video Game ever recorded in North America

EU_Sales - Sales of Video Game ever recorded in Europe

JP_Sales - Sales of Video Game ever recorded in Japan

Other_Sales - Sales of Video Game ever recorded in Every other country except NA, EU and JP

Global_Sales - Sales of all Video Games ever recorded all Countries world wide

1.1 Some Important Information about Gaming Platfoms

WS – The WonderSwan is a handheld game console released in Japan by Bandai. It was developed by Gunpei Yokoi's company Koto Laboratory and Bandai.

3DO – The 3DO Interactive Multiplayer, also referred to as simply 3DO, is a home video game console developed by The 3DO Company.

DC – Sega's Dreamcast is the first console of the generation and had several features to show an advantage from the competition, including Internet gaming as an optional feature through its built-in modem, and a web browser.

SAT – The Sega Saturn is a home video game console developed by Sega and released on November 22, 1994.

SCD – The Sega CD plays CD-based games and adds hardware functionality such as a faster central processing unit and graphic enhancements like sprite scaling.

2600 – The Atari 2600 is a home video game console developed and produced by Atari, Inc. Released in September 1977.

PS – PlayStation Network's services are dedicated to an online marketplace (PlayStation Store), a premium subscription service for enhanced gaming and social features (PlayStation Plus), music streaming (PlayStation Music, based on Spotify), TV streaming (PlayStation Vue), and formerly a cloud gaming service

PSP – The PlayStation Portable (PSP) is a handheld game console developed and marketed by Sony Interactive Entertainment.

PCFX – The PC-FX is a 32-bit home video game console developed by NEC and Hudson Soft. It was released in 1994 and discontinued in February 1998

PS2 – The PlayStation 2 (PS2) is a home video game console developed and marketed by Sony Interactive Entertainment.

PSV – The PlayStation Vita (PS Vita, or Vita) is a handheld game console developed and marketed by Sony Interactive Entertainment.

PS3 – The PlayStation 3 (PS3) is a home video game console developed and marketed by Sony Interactive Entertainment. The successor to the PlayStation 2, it is part of the PlayStation brand of consoles.

TG16 – The TurboGrafx-16, known as the PC Engine outside North America, is a home video game console designed by Hudson Soft and sold by NEC Home Electronics.

PS4 – The PlayStation 4 (PS4) is a home video game console developed by Sony Interactive Entertainment. Announced as the successor to the PlayStation 3, it is part of the PlayStation brand of consoles.

PC – A personal computer game, also known as computer game or abbreviated PC game, is a electronic game .

N64 – The Nintendo 64 (N64) is a home video game console developed by Nintendo.

GB – The Game Boy is an 8-bit fourth generation handheld game console developed and manufactured by Nintendo.

Wii – The Wii is a home video game console developed and marketed by Nintendo.

WiiU – Wii is a series of simulation games published by Nintendo for the game console of the same name, as well as its successor, the Wii U.

DS – The Nintendo DS is a handheld game console manufactured by Nintendo and released worldwide from 2004 to 2005. DS is an acronym for “developer system” or “dual screen” and introduced new features characteristic of handheld games.

3DS – The Nintendo 3DS is a handheld game console produced by Nintendo. The console was announced in March 2010 and unveiled at E3 2010 as the successor to the Nintendo DS.

NES – The NES features a number of groundbreaking games, such as the 1985 platform game Super Mario Bros. and the 1986 action-adventure games The Legend of Zelda.

SNES – The Super Nintendo Entertainment System, commonly shortened to Super Nintendo, Super NES or SNES is a 16-bit home video game console developed by Nintendo.

GC – The Nintendo GameCube is a home video game console developed and released by Nintendo.

GBA – The Game Boy Advance (GBA) is a 32-bit handheld game console developed, manufactured and marketed by Nintendo as the successor to the Game Boy Color.

XB – The Xbox is a home video game console manufactured by Microsoft that is the first installment in the Xbox series of video game consoles.

XONE – The Xbox One is a home video game console developed by Microsoft. Announced in May 2013.

X360 – The Xbox 360 is a home video game console developed by Microsoft. As the successor to the original Xbox, it is the second console in the Xbox series.

GG – Guilty Gear, subtitled The Missing Link in Japan, is a 2D fighting video game developed by Team Neo Blood.

1.2 Data Cleaning

```
[12]: ## Finding the Null Values  
df.isnull().sum()
```

```
[12]: Rank          0  
      Name          0  
      Platform      0  
      Year          271  
      Genre          0  
      Publisher      58  
      NA_Sales       0  
      EU_Sales       0  
      JP_Sales       0  
      Other_Sales    0  
      Global_Sales   0  
      dtype: int64
```

```
[13]: percent_missing = df.isnull().sum() * 100 / len(df)  
      missing_table = pd.DataFrame({'column_name': df.columns,  
                                   'percent_missing': percent_missing})  
  
      print(missing_table)
```

	column_name	percent_missing
Rank	Rank	0.000000
Name	Name	0.000000
Platform	Platform	0.000000
Year	Year	1.632727
Genre	Genre	0.000000
Publisher	Publisher	0.349440
NA_Sales	NA_Sales	0.000000
EU_Sales	EU_Sales	0.000000
JP_Sales	JP_Sales	0.000000
Other_Sales	Other_Sales	0.000000
Global_Sales	Global_Sales	0.000000

```
[14]: ##Display of Null valued rows  
null_mask = df.isnull()  
  
null_rows = df[null_mask.any(axis=1)]  
  
ds = pd.DataFrame(null_rows)
```

```
ds.head(10)
```

```
[14]:
```

	Rank	Name	Platform	Year	\
179	180	Madden NFL 2004	PS2	NaN	
377	378	FIFA Soccer 2004	PS2	NaN	
431	432	LEGO Batman: The Videogame	Wii	NaN	
470	471	wwe Smackdown vs. Raw 2006	PS2	NaN	
607	608	Space Invaders	2600	NaN	
624	625	Rock Band	X360	NaN	
649	650	Frogger's Adventures: Temple of the Frog	GBA	NaN	
652	653	LEGO Indiana Jones: The Original Adventures	Wii	NaN	
711	713	Call of Duty 3	Wii	NaN	
782	784	Rock Band	Wii	NaN	

	Genre	Publisher	NA_Sales	EU_Sales	\
179	Sports	Electronic Arts	4.26	0.26	
377	Sports	Electronic Arts	0.59	2.36	
431	Action	Warner Bros. Interactive Entertainment	1.86	1.02	
470	Fighting	NaN	1.57	1.02	
607	Shooter	Atari	2.36	0.14	
624	Misc	Electronic Arts	1.93	0.34	
649	Adventure	Konami Digital Entertainment	2.15	0.18	
652	Action	LucasArts	1.54	0.63	
711	Shooter	Activision	1.19	0.84	
782	Misc	MTV Games	1.35	0.56	

	JP_Sales	Other_Sales	Global_Sales
179	0.01	0.71	5.23
377	0.04	0.51	3.49
431	0.00	0.29	3.17
470	0.00	0.41	3.00
607	0.00	0.03	2.53
624	0.00	0.21	2.48
649	0.00	0.07	2.39
652	0.00	0.22	2.39
711	0.00	0.23	2.26
782	0.00	0.20	2.11

```
[15]: ## Replacing the null values in Year using median
median_value = df['Year'].median()

df['Year'].fillna(median_value, inplace=True)

median_value
```

```
[15]: 2007.0
```



```
[16]: ## Replacing the null values in Publisher using mode
mode_value = df['Publisher'].mode()[0]

df['Publisher'].fillna(mode_value, inplace=True)

print(mode_value)
```

Electronic Arts

```
[17]: ## Finding the Null Values
df.isnull().sum()
```

```
[17]: Rank          0
      Name          0
      Platform      0
      Year          0
      Genre         0
      Publisher      0
      NA_Sales       0
      EU_Sales       0
      JP_Sales       0
      Other_Sales    0
      Global_Sales   0
      dtype: int64
```

```
[174]: # Check for duplicates in the "Name" column
name_duplicates = df['Rank'].duplicated()

# Print the result
print(f"Are there any duplicates in the 'Rank' column? {any(name_duplicates)}")
```

Are there any duplicates in the 'Rank' column? False

```
[42]: # Check for duplicates in the "Name" column
name_duplicates = df['Name'].duplicated()

# Print the result
print(f"Are there any duplicates in the 'Name' column? {any(name_duplicates)}")
```

Are there any duplicates in the 'Name' column? True

```
[172]: # Identify rows with duplicate names
duplicate_rows = df[name_duplicates]

# Print the duplicate rows
duplicate_rows.head()
```

```
[172]:
```

	Rank	Name	Platform	Year	Genre	\
23	24	Grand Theft Auto V	X360	2013.0	Action	
35	36	Call of Duty: Black Ops II	X360	2012.0	Shooter	
37	38	Call of Duty: Modern Warfare 3	PS3	2011.0	Shooter	
40	41	Call of Duty: Black Ops	PS3	2010.0	Shooter	
44	45	Grand Theft Auto V	PS4	2014.0	Action	

		Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	\
23	Take-Two	Interactive	9.63	5.31	0.06	1.38	
35		Activision	8.25	4.30	0.07	1.12	
37		Activision	5.54	5.82	0.49	1.62	
40		Activision	5.98	4.44	0.48	1.83	
44	Take-Two	Interactive	3.80	5.81	0.36	2.02	

	Global_Sales
23	16.38
35	13.73
37	13.46
40	12.73
44	11.98

```
[171]: # Filter the DataFrame to find rows containing "Grand Theft Auto V" in the
        ↪ "Name" column
gta_rows = df[df['Name'].str.contains('Grand Theft Auto V')]

# Print the filtered DataFrame
gta_rows.head()
```

```
[171]:
```

	Rank	Name	Platform	Year	\
16	17	Grand Theft Auto V	PS3	2013.0	
23	24	Grand Theft Auto V	X360	2013.0	
44	45	Grand Theft Auto V	PS4	2014.0	
197	198	Grand Theft Auto V	XOne	2014.0	
617	618	Rockstar Games Double Pack: Grand Theft Auto I...	XB	2003.0	

	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	\
16	Action	Take-Two Interactive	7.01	9.27	0.97	4.14	
23	Action	Take-Two Interactive	9.63	5.31	0.06	1.38	
44	Action	Take-Two Interactive	3.80	5.81	0.36	2.02	
197	Action	Take-Two Interactive	2.66	2.01	0.00	0.41	
617	Action	Take-Two Interactive	1.84	0.56	0.00	0.09	

	Global_Sales
16	21.40
23	16.38
44	11.98
197	5.08

1.3 Data Understanding

```
[114]: df.Genre.value_counts()
```

```
[114]: Genre
Action      3316
Sports      2346
Misc        1739
Role-Playing 1488
Shooter     1310
Adventure   1286
Racing      1249
Platform    886
Simulation   867
Fighting    848
Strategy    681
Puzzle      582
Name: count, dtype: int64
```

```
[115]: genre_counts = df['Genre'].value_counts()
genre_percentages = (genre_counts / len(df)) * 100
genre_percentages
```

```
[115]: Genre
Action      19.978311
Sports      14.134233
Misc        10.477166
Role-Playing  8.964936
Shooter      7.892517
Adventure    7.747921
Racing       7.525003
Platform     5.337993
Simulation   5.223521
Fighting     5.109049
Strategy     4.102904
Puzzle       3.506447
Name: count, dtype: float64
```

```
[173]: df.Rank.value_counts()
```

```
[173]: Rank
1      1
11089  1
11059  1
11060  1
```

```

11061    1
      ..
5539     1
5540     1
5541     1
5542     1
16600    1
Name: count, Length: 16598, dtype: int64

```

```
[175]: df.Year.value_counts()
```

```

[175]: Year
2007.0    1473
2009.0    1431
2008.0    1428
2010.0    1259
2011.0    1139
2006.0    1008
2005.0     941
2002.0     829
2003.0     775
2004.0     763
2012.0     657
2015.0     614
2014.0     582
2013.0     546
2001.0     482
1998.0     379
2000.0     349
2016.0     344
1999.0     338
1997.0     289
1996.0     263
1995.0     219
1994.0     121
1993.0      60
1981.0      46
1992.0      43
1991.0      41
1982.0      36
1986.0      21
1989.0      17
1983.0      17
1990.0      16
1987.0      16
1988.0      15
1985.0      14

```

```

1984.0      14
1980.0       9
2017.0       3
2020.0       1
Name: count, dtype: int64

```

```
[176]: df.Publisher.value_counts()
```

```

[176]: Publisher
Electronic Arts      1409
Activision           975
Namco Bandai Games   932
Ubisoft              921
Konami Digital Entertainment  832
...
Warp                  1
New                   1
Elite                 1
Evolution Games      1
UIG Entertainment    1
Name: count, Length: 578, dtype: int64

```

```
[177]: df.Platform.value_counts()
```

```

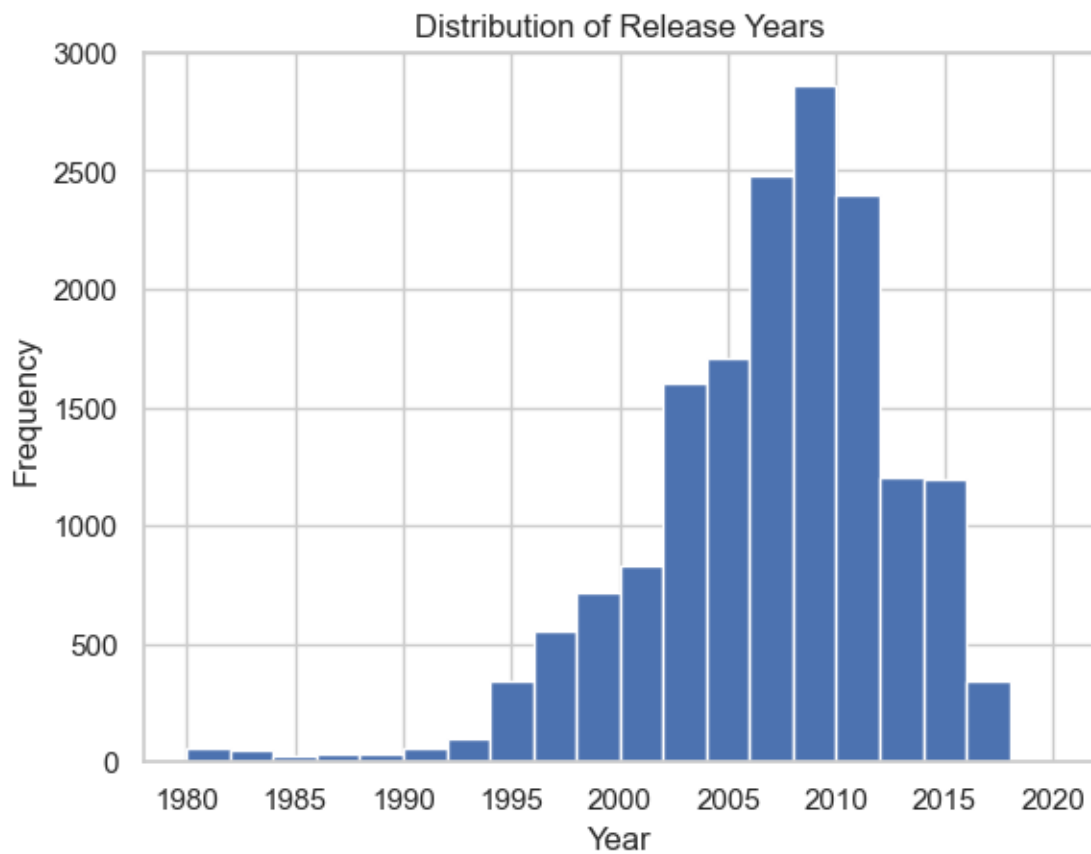
[177]: Platform
DS      2163
PS2     2161
PS3     1329
Wii     1325
X360    1265
PSP     1213
PS      1196
PC       960
XB       824
GBA      822
GC       556
3DS      509
PSV      413
PS4      336
N64      319
SNES     239
XOne     213
SAT      173
WiiU     143
2600     133
NES       98
GB        98

```

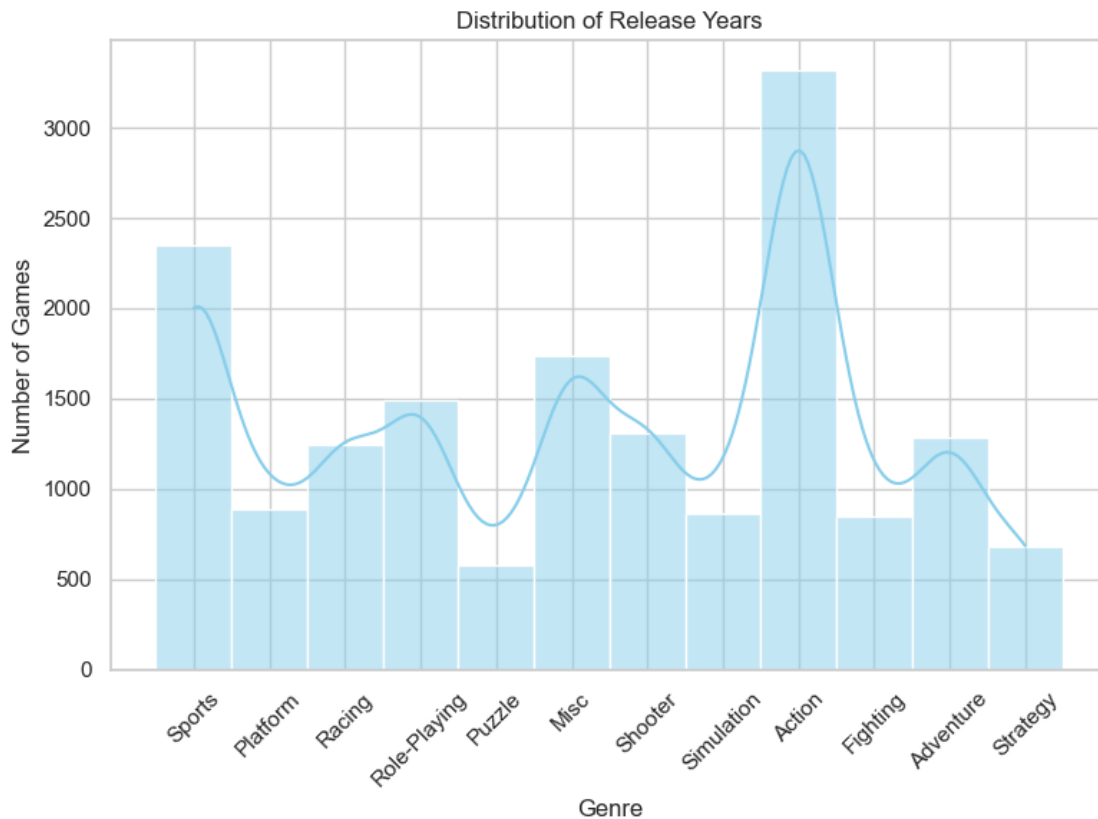
```
DC          52
GEN         27
NG          12
SCD         6
WS          6
3DO         3
TG16        2
GG          1
PCFX        1
Name: count, dtype: int64
```

1.4 UNIVARIAT ANALYSIS

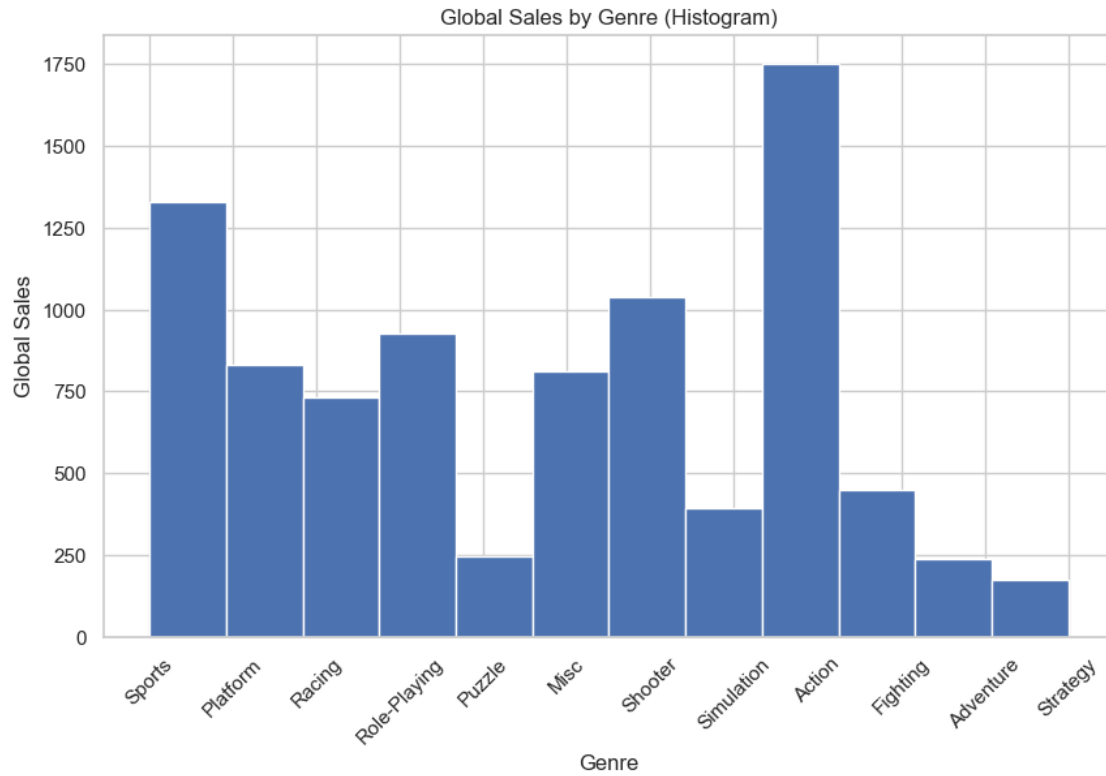
```
[111]: # Create histogram
plt.hist(df['Year'], bins=20)
plt.xlabel('Year')
plt.ylabel('Frequency')
plt.title('Distribution of Release Years')
plt.show()
```



```
[112]: # Create histogram with KDE
plt.figure(figsize=(8, 6))
sns.histplot(df['Genre'], bins=30, color='skyblue', kde=True)
plt.title('Distribution of Release Years')
plt.ylabel('Number of Games')
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

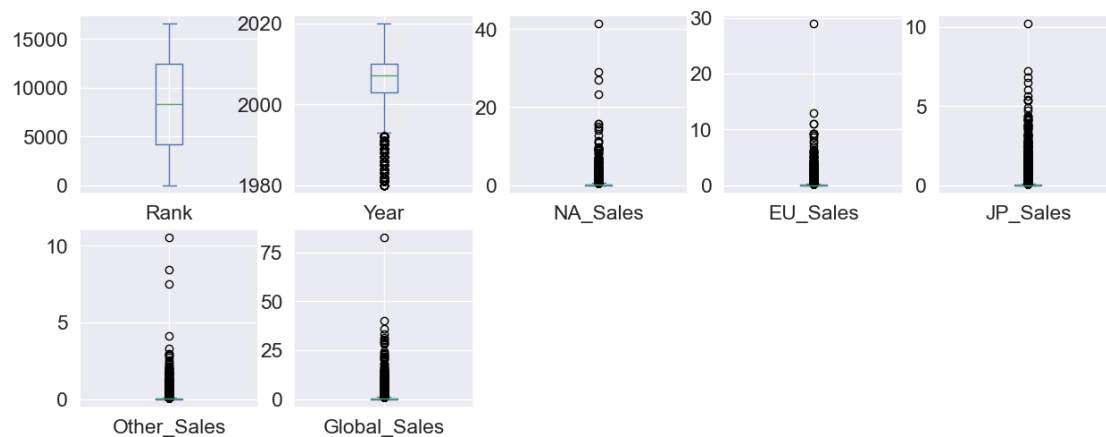


```
[116]: # Histogram for Global Sales by Genre
plt.figure(figsize=(10, 6))
plt.hist(df["Genre"], weights=df["Global_Sales"], bins=len(df["Genre"] .
    ↪unique()))
plt.xlabel("Genre")
plt.ylabel("Global Sales")
plt.title("Global Sales by Genre (Histogram)")
plt.xticks(rotation=45)
plt.show()
```



1.5 Bivariate Analysis

```
[140]: df.plot(kind='box', subplots=True, figsize=(15,15), layout=(5,5))
plt.show()
```



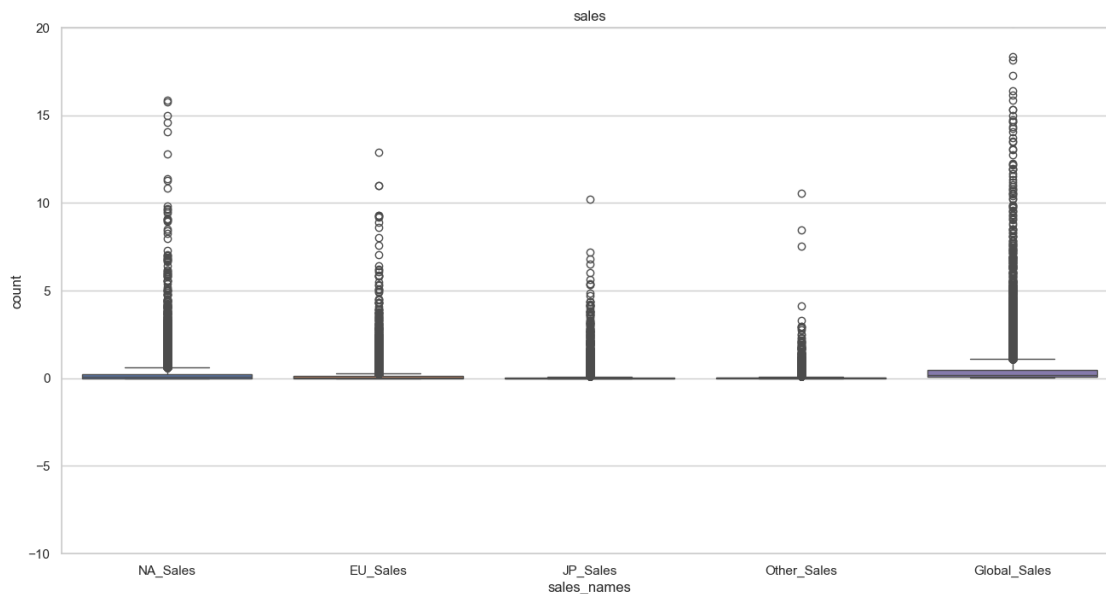

```
[141]: x=df[['NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales']]

# Set the style of the plot
sns.set(style="whitegrid")

# Create a box plot
plt.figure(figsize=(16, 8))
sns.boxplot(data=x)
plt.title('sales')
plt.xlabel('sales_names')
plt.ylabel('count')

# Set the y-axis limits to expand the range
plt.ylim(-10, 20) # Adjust the values as needed

plt.show()
```



```
[142]: plt.figure(figsize=(15, 10))

# Scatter plot 1
plt.subplot(2, 2, 1)
plt.scatter(df['Global_Sales'], df['Other_Sales'], color='blue', label='Other_
↵Sales')
plt.xlabel('Global Sales', fontsize=15)
plt.ylabel('Other Sales', fontsize=15)
plt.title('Scatter Plot of Global Sales vs Other Sales', fontsize=15)
plt.legend()
```

```

plt.grid(True)

# Scatter plot 2
plt.subplot(2, 2, 2)
plt.scatter(df['Global_Sales'], df['NA_Sales'], color='yellow', label='NA_Sales')
plt.xlabel('Global Sales', fontsize=15)
plt.ylabel('NA Sales', fontsize=15)
plt.title('Scatter Plot of Global Sales vs NA Sales', fontsize=15)
plt.legend()
plt.grid(True)

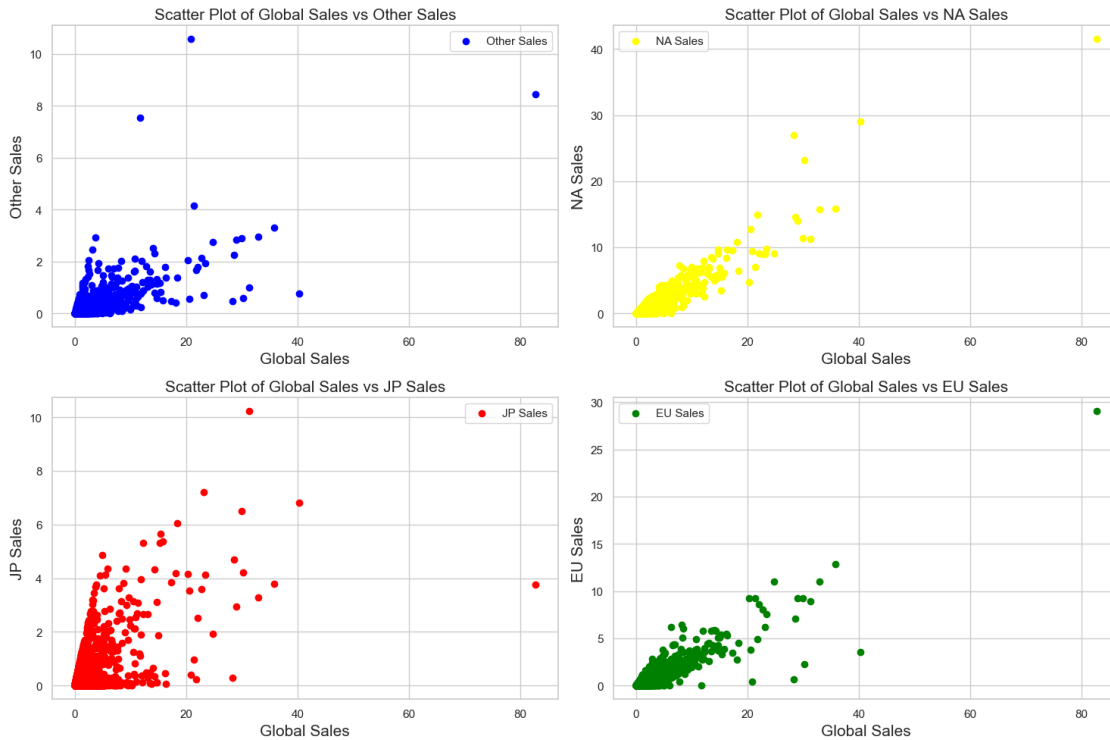
# Scatter plot 3
plt.subplot(2, 2, 3)
plt.scatter(df['Global_Sales'], df['JP_Sales'], color='red', label='JP Sales')
plt.xlabel('Global Sales', fontsize=15)
plt.ylabel('JP Sales', fontsize=15)
plt.title('Scatter Plot of Global Sales vs JP Sales', fontsize=15)
plt.legend()
plt.grid(True)

# Scatter plot 4
plt.subplot(2, 2, 4)
plt.scatter(df['Global_Sales'], df['EU_Sales'], color='green', label='EU Sales')
plt.xlabel('Global Sales', fontsize=15)
plt.ylabel('EU Sales', fontsize=15)
plt.title('Scatter Plot of Global Sales vs EU Sales', fontsize=15)
plt.legend()
plt.grid(True)

## Adjust the padding between and around subplots to avoid overlapping labels
plt.tight_layout()

## Display the graph
plt.show()

```



```
[143]: correlation = df['Global_Sales'].corr(df['NA_Sales'])
print('Correlation coefficient between Global and North America Sales:',
      ↪correlation)
```

Correlation coefficient between Global and North America Sales:
0.9410473571255523

```
[144]: correlation = df['Global_Sales'].corr(df['EU_Sales'])
print('Correlation coefficient between Global and Europe Sales:', correlation)
```

Correlation coefficient between Global and Europe Sales: 0.9028358134817434

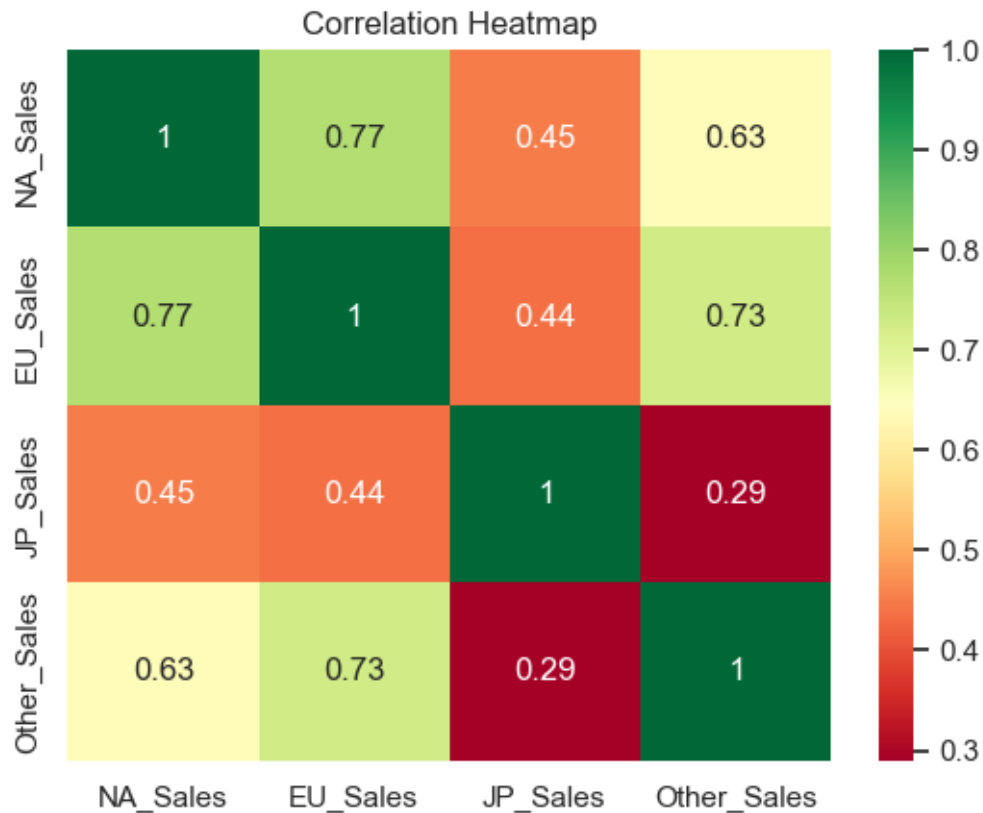
```
[145]: correlation = df['Global_Sales'].corr(df['JP_Sales'])
print('Correlation coefficient between Global and Japan Sales:', correlation)
```

Correlation coefficient between Global and Japan Sales: 0.6118155181564976

```
[146]: correlation = df['Global_Sales'].corr(df['Other_Sales'])
print('Correlation coefficient between Global and Other Countries Sales:',
      ↪correlation)
```

Correlation coefficient between Global and Other Countries Sales:
0.748330846407796

```
[147]: sns.heatmap(df[["NA_Sales", "EU_Sales", "JP_Sales", "Other_Sales"]].corr(), annot=True, cmap="RdYlGn")
plt.title("Correlation Heatmap")
plt.show()
```



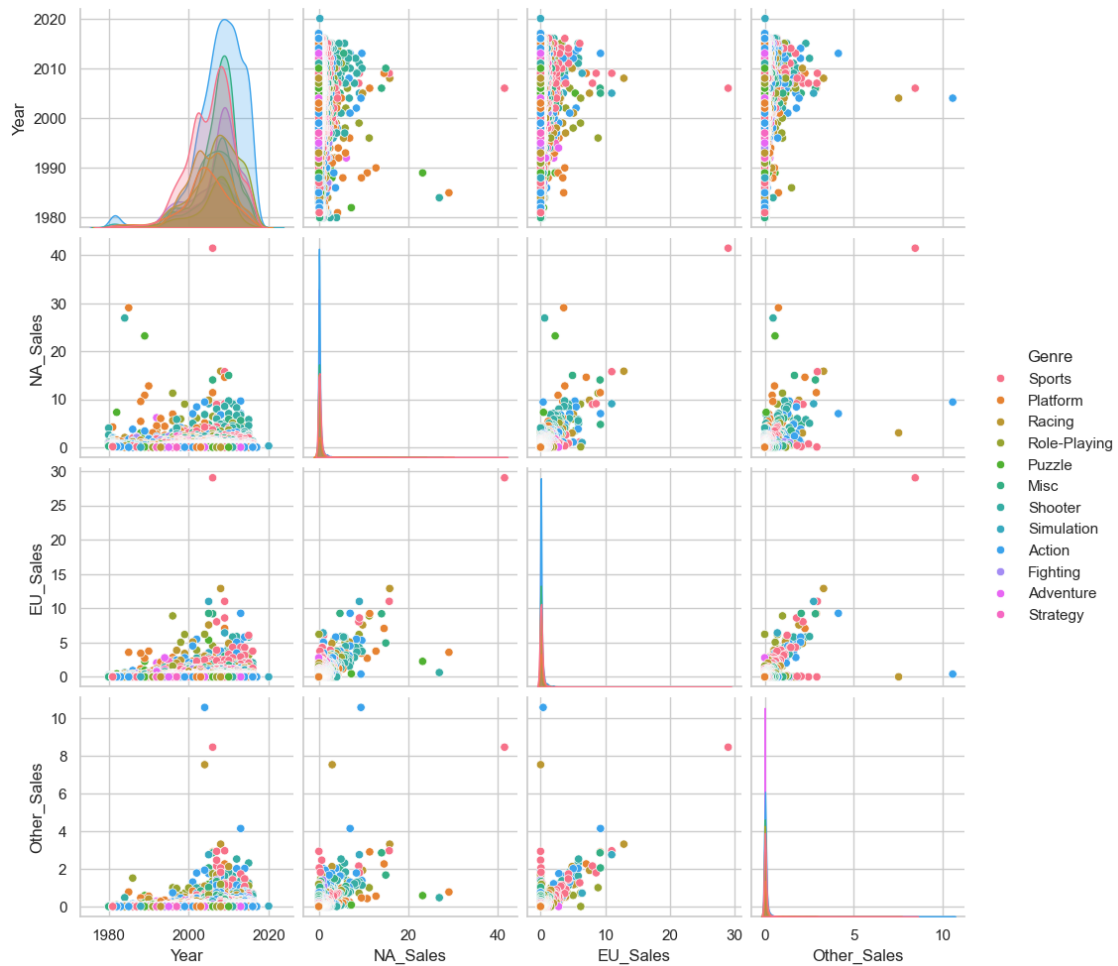
```
[148]: data_pair = df.loc[:, ["Year", "Platform", "Genre", "NA_Sales", "EU_Sales", "Other_Sales"]]
data_pair
```

```
[148]:
```

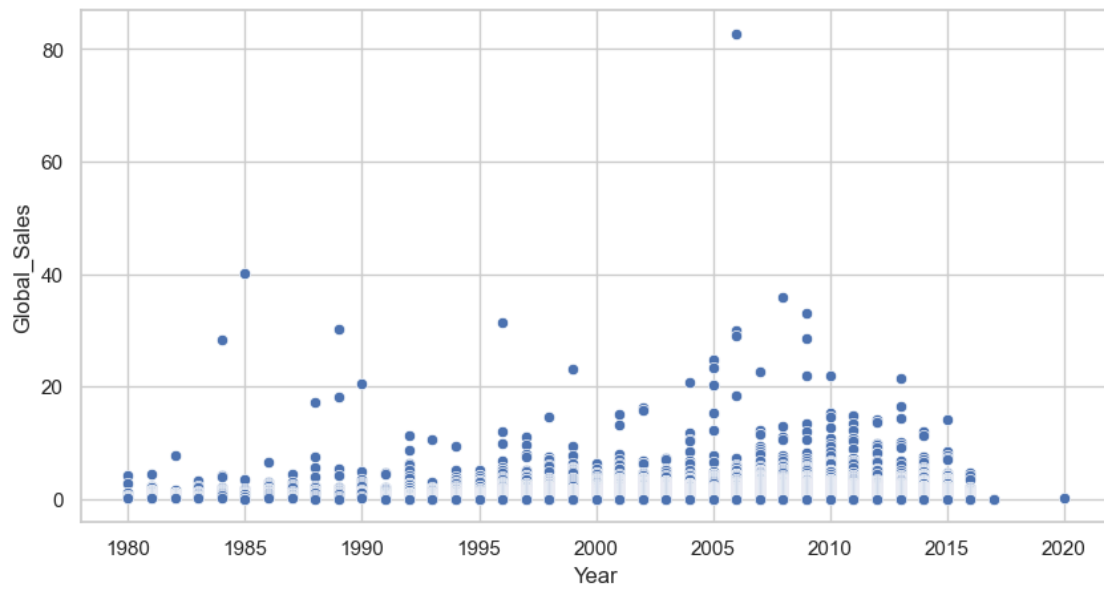
	Year	Platform	Genre	NA_Sales	EU_Sales	Other_Sales
0	2006.0	Wii	Sports	41.49	29.02	8.46
1	1985.0	NES	Platform	29.08	3.58	0.77
2	2008.0	Wii	Racing	15.85	12.88	3.31
3	2009.0	Wii	Sports	15.75	11.01	2.96
4	1996.0	GB	Role-Playing	11.27	8.89	1.00
...
16593	2002.0	GBA	Platform	0.01	0.00	0.00
16594	2003.0	GC	Shooter	0.01	0.00	0.00
16595	2008.0	PS2	Racing	0.00	0.00	0.00
16596	2010.0	DS	Puzzle	0.00	0.01	0.00
16597	2003.0	GBA	Platform	0.01	0.00	0.00

[16598 rows x 6 columns]

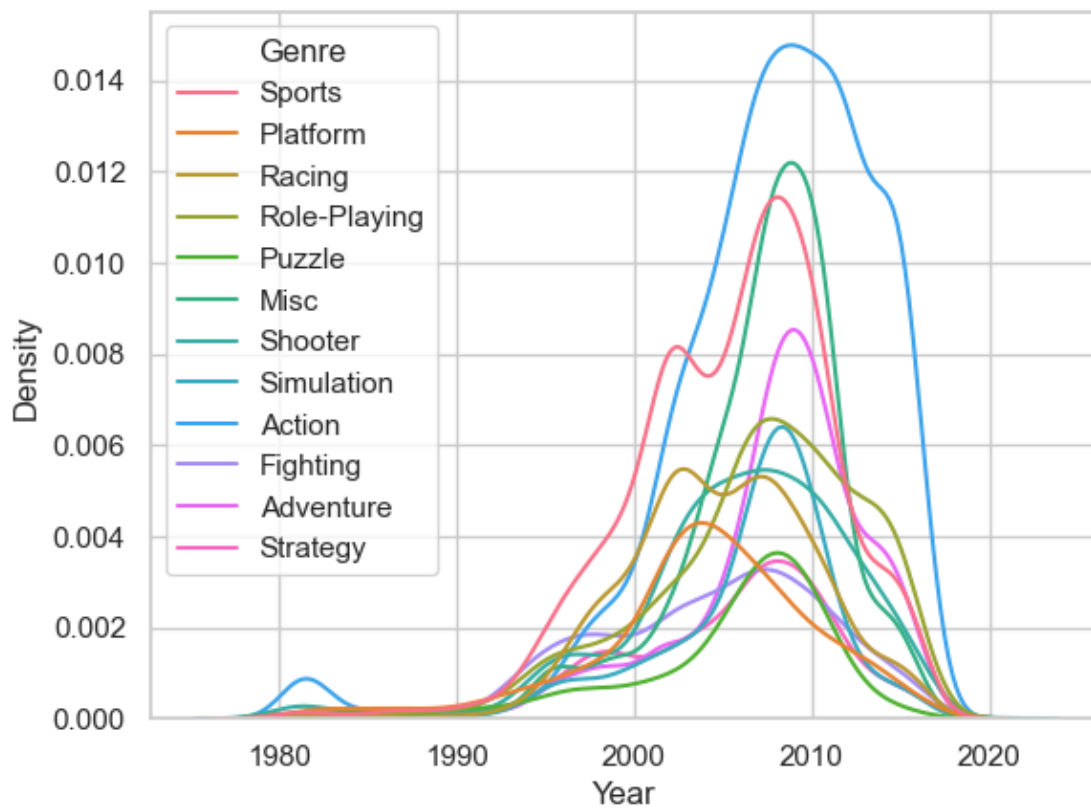
```
[149]: sns.pairplot(data_pair, hue='Genre')  
plt.show()
```



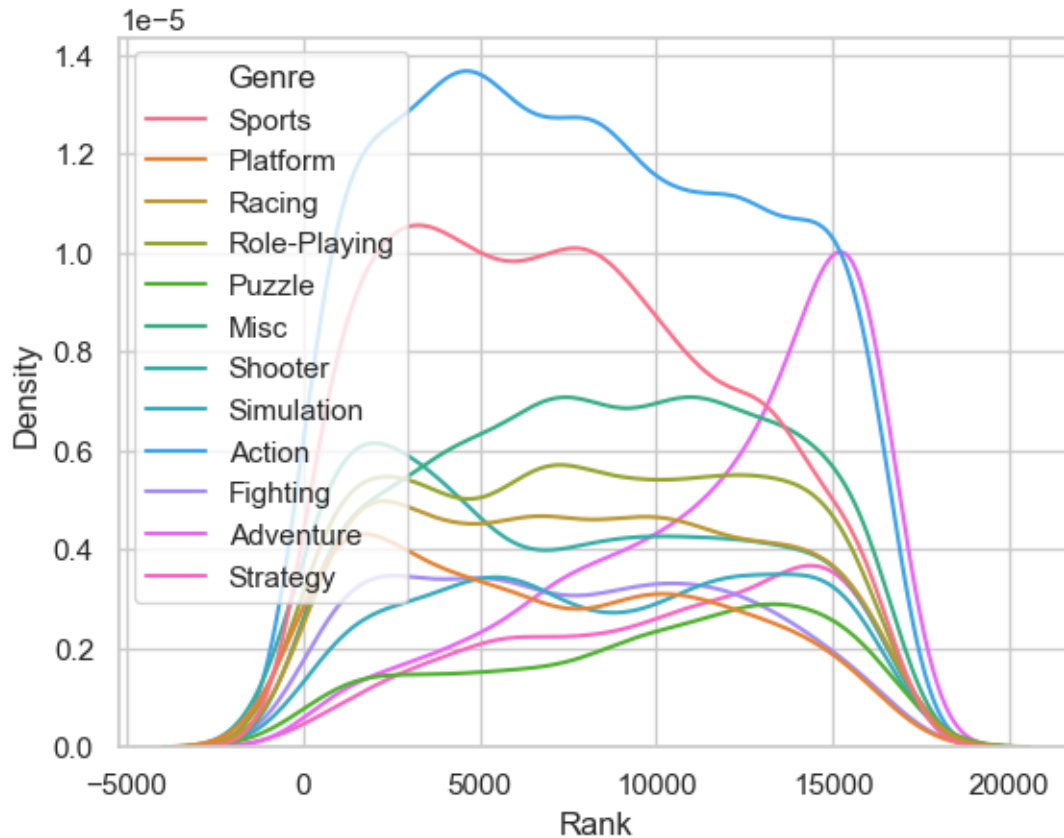
```
[150]: plt.figure(figsize=(10,5))  
sns.scatterplot(data=df, x="Year", y="Global_Sales")  
plt.show()
```



```
[151]: sns.kdeplot(x='Year',data = df,hue = 'Genre')
plt.show()
```



```
[152]: sns.kdeplot(x='Rank',data = df,hue = 'Genre')
plt.show()
```



```
[153]: # Set the style of the plot
sns.set(style="whitegrid")

plt.figure(figsize=(16, 12))

# Create a subplot for each distribution
plt.subplot(2, 2, 1)
plt.title("Distribution of NA Sales", fontsize=15)
sns.kdeplot(data=x['NA_Sales'], color='blue')

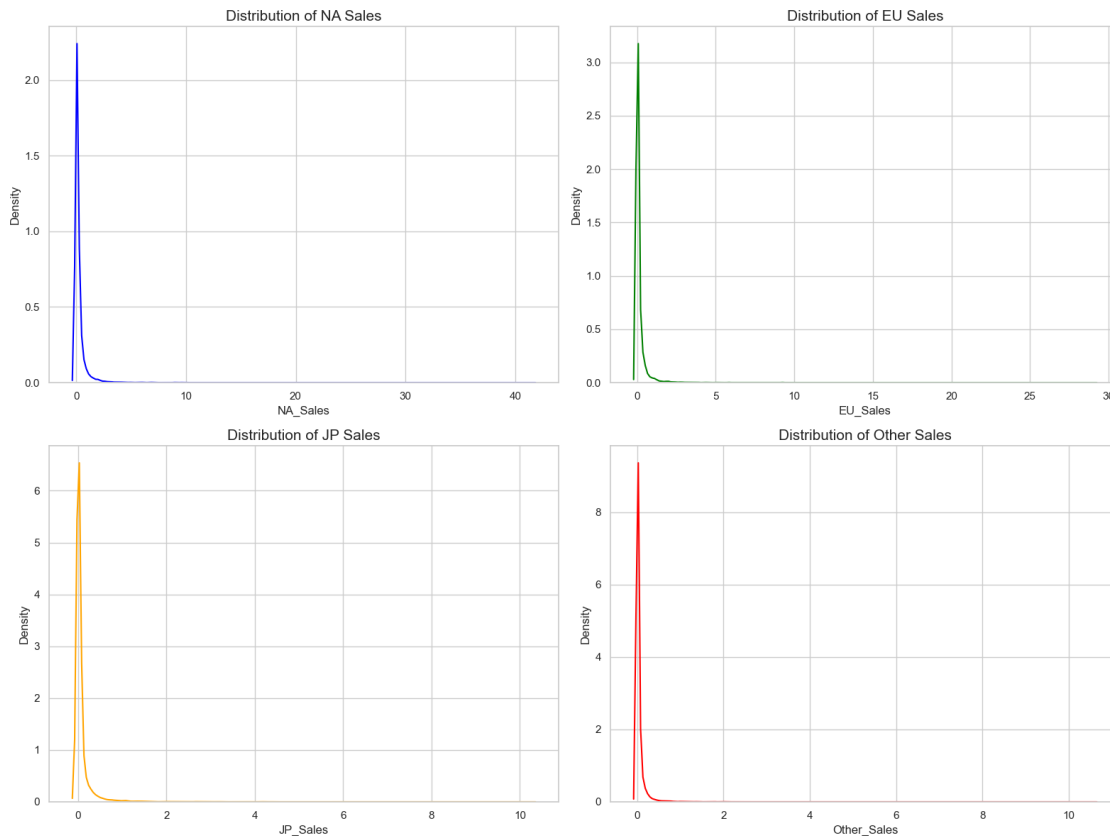
plt.subplot(2, 2, 2)
plt.title("Distribution of EU Sales", fontsize=15)
sns.kdeplot(data=x['EU_Sales'], color='green')

plt.subplot(2, 2, 3)
```

```
plt.title("Distribution of JP Sales", fontsize=15)
sns.kdeplot(data=x['JP_Sales'], color='orange')

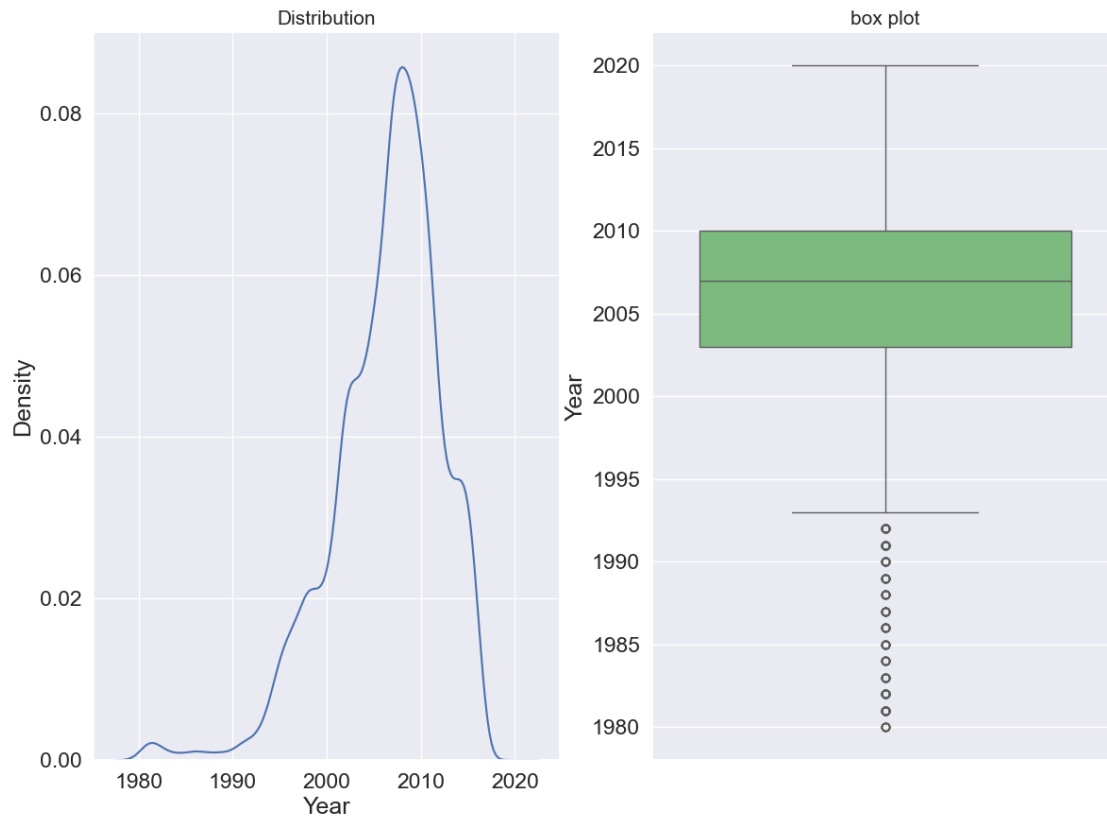
plt.subplot(2, 2, 4)
plt.title("Distribution of Other Sales", fontsize=15)
sns.kdeplot(data=x['Other_Sales'], color='red')

plt.tight_layout() # Ensures proper spacing between subplots
plt.show()
```



1.6 Multivariate Analysis

```
[93]: plt.figure(figsize=(14,10))
plt.subplot(1,2,1)
plt.title("Distribution", fontsize=15)
sns.kdeplot(data = df["Year"])
plt.subplot(1,2,2)
plt.title("box plot", fontsize=15)
sns.boxplot(data = df["Year"], palette="Greens")
plt.show()
```

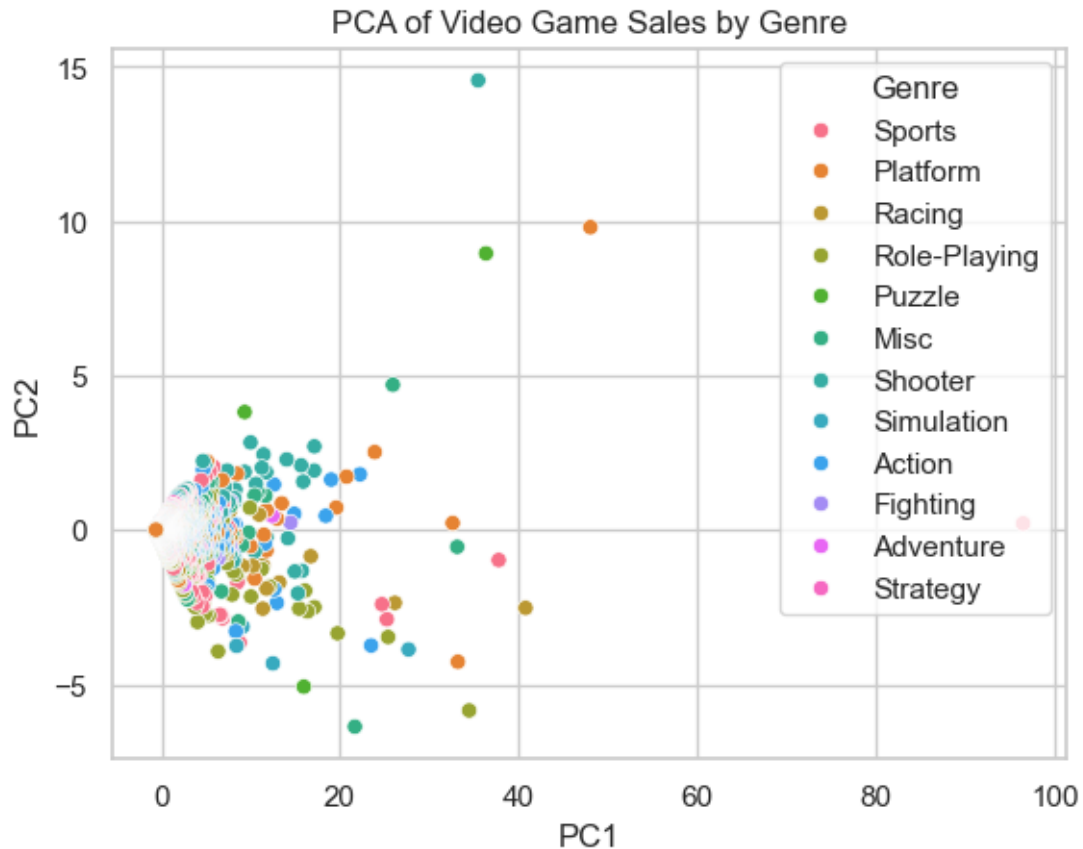
```
[160]: # Select features to use for PCA
features = ['NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales']
X = df[features]

# Perform PCA
pca = PCA(n_components=2)
principal_components = pca.fit_transform(X)

# Create new dataframe with principal components
df_pca = pd.DataFrame(data=principal_components, columns=['PC1', 'PC2'])

# Add target variable to new dataframe
df_pca['Genre'] = df['Genre']

# Plot results
sns.scatterplot(data=df_pca, x='PC1', y='PC2', hue='Genre')
plt.title('PCA of Video Game Sales by Genre')
plt.show()
```



1.7 Distributions

```
[166]: # Extract 'Global_Sales' data from DataFrame
global_sales = df['Global_Sales'].values

# Define the degrees of freedom (k)
k = 3 # Example: Degrees of freedom

# Generate random samples from a chi-square distribution
num_samples = 1000
samples = np.random.chisquare(df=k, size=num_samples)

# Create a histogram to visualize the distribution
plt.hist(samples, bins=30, density=True, alpha=0.6, color='b')

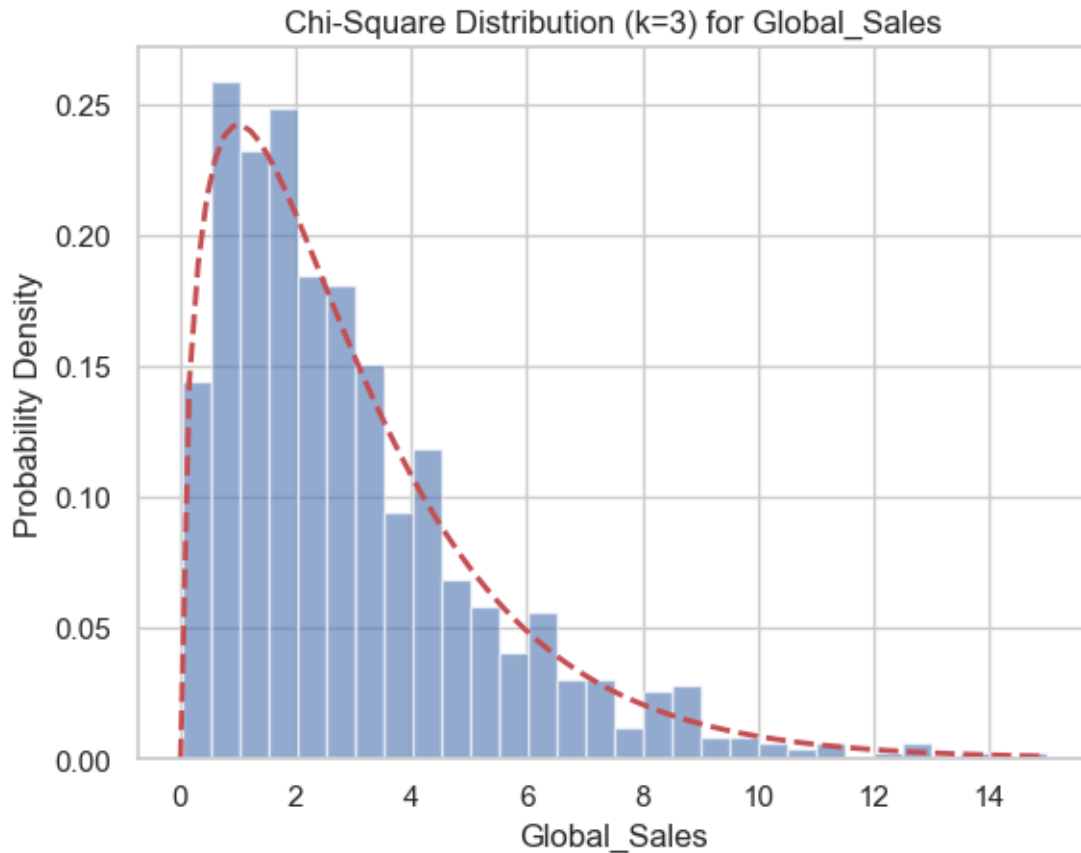
# Create the theoretical probability density function
x = np.linspace(0, max(samples), 100)
pdf = (1/(2**(k/2) * np.math.gamma(k/2))) * x**((k/2) - 1) * np.exp(-x/2)
plt.plot(x, pdf, 'r--', lw=2)
```

```

# Add labels and title to the plot
plt.xlabel("Global_Sales")
plt.ylabel("Probability Density")
plt.title(f"Chi-Square Distribution (k={k}) for Global_Sales")

# Show the plot
plt.show()

```



```

[78]: # Extract the global sales column
global_sales = df['Global_Sales']

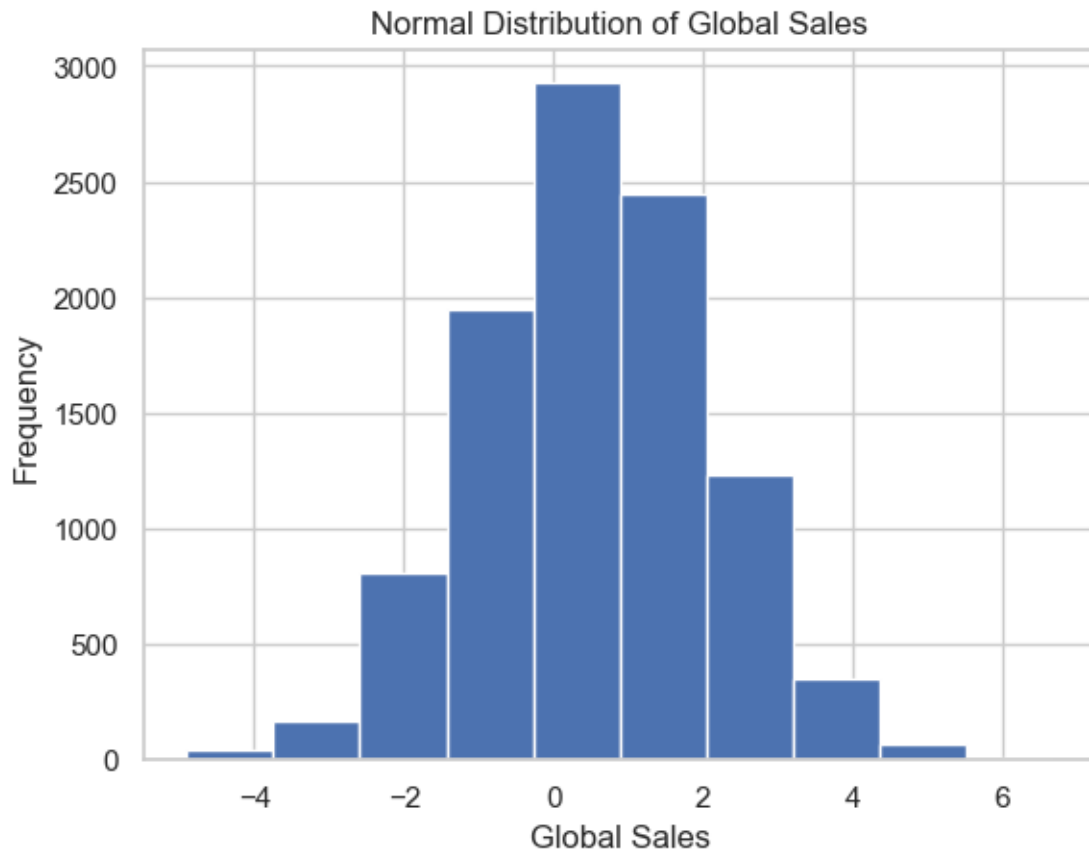
# Calculate the mean and standard deviation of global sales
mean = global_sales.mean()
std = global_sales.std()

# Create a normal distribution object
norm_dist = np.random.normal(mean, std, size=10000)

# Plot the normal distribution

```

```
plt.hist(norm_dist)
plt.xlabel('Global Sales')
plt.ylabel('Frequency')
plt.title('Normal Distribution of Global Sales')
plt.show()
```



```
[79]: # Extract 'Year' data from DataFrame
global_sales = df['Year'].values

# Define the degrees of freedom (k)
k = 3 # Example: Degrees of freedom

# Generate random samples from a chi-square distribution
num_samples = 1000
samples = np.random.chisquare(df=k, size=num_samples)

# Create a histogram to visualize the distribution
plt.hist(samples, bins=30, density=True, alpha=0.6, color='b')

# Create the theoretical probability density function
```

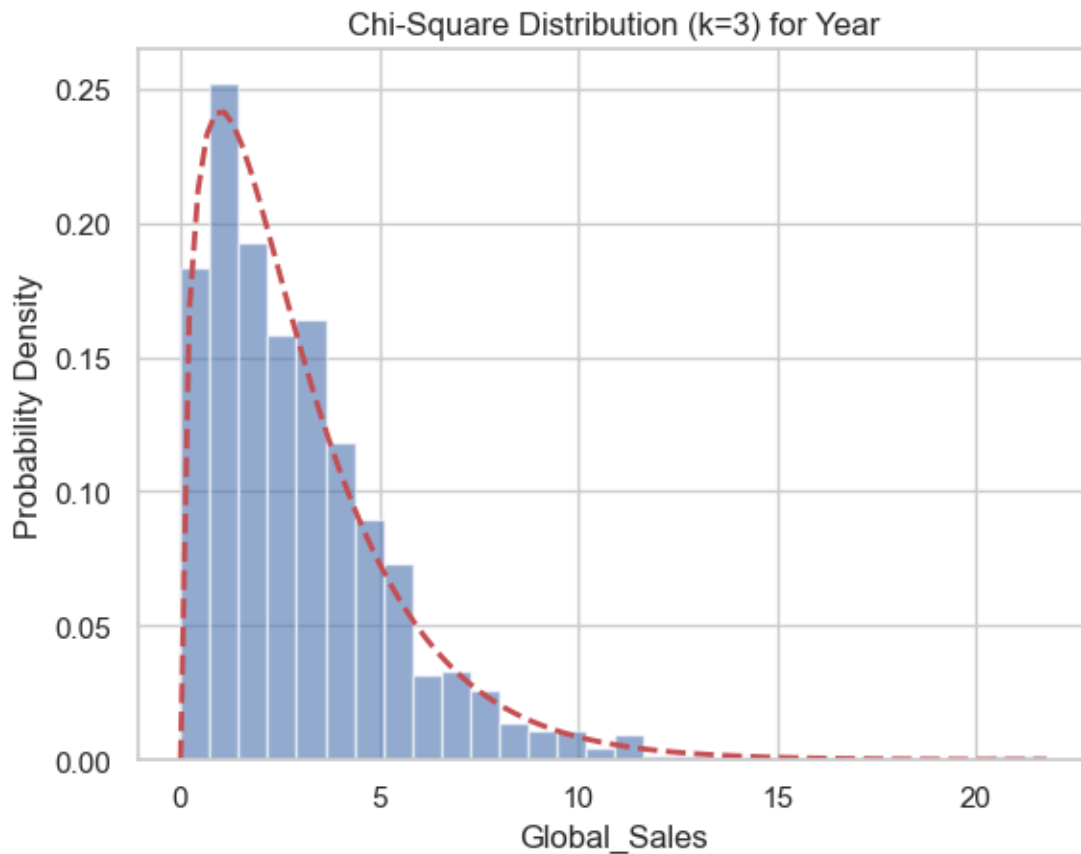
```

x = np.linspace(0, max(samples), 100)
pdf = (1/(2**(k/2) * np.math.gamma(k/2))) * x**((k/2) - 1) * np.exp(-x/2)
plt.plot(x, pdf, 'r--', lw=2)

# Add labels and title to the plot
plt.xlabel("Global_Sales")
plt.ylabel("Probability Density")
plt.title(f"Chi-Square Distribution (k={k}) for Year")

# Show the plot
plt.show()

```



```

[80]: # Extract the 'Year' column
global_sales = df['Year']

# Calculate the mean and standard deviation of Year
mean = global_sales.mean()
std = global_sales.std()

# Create a normal distribution object

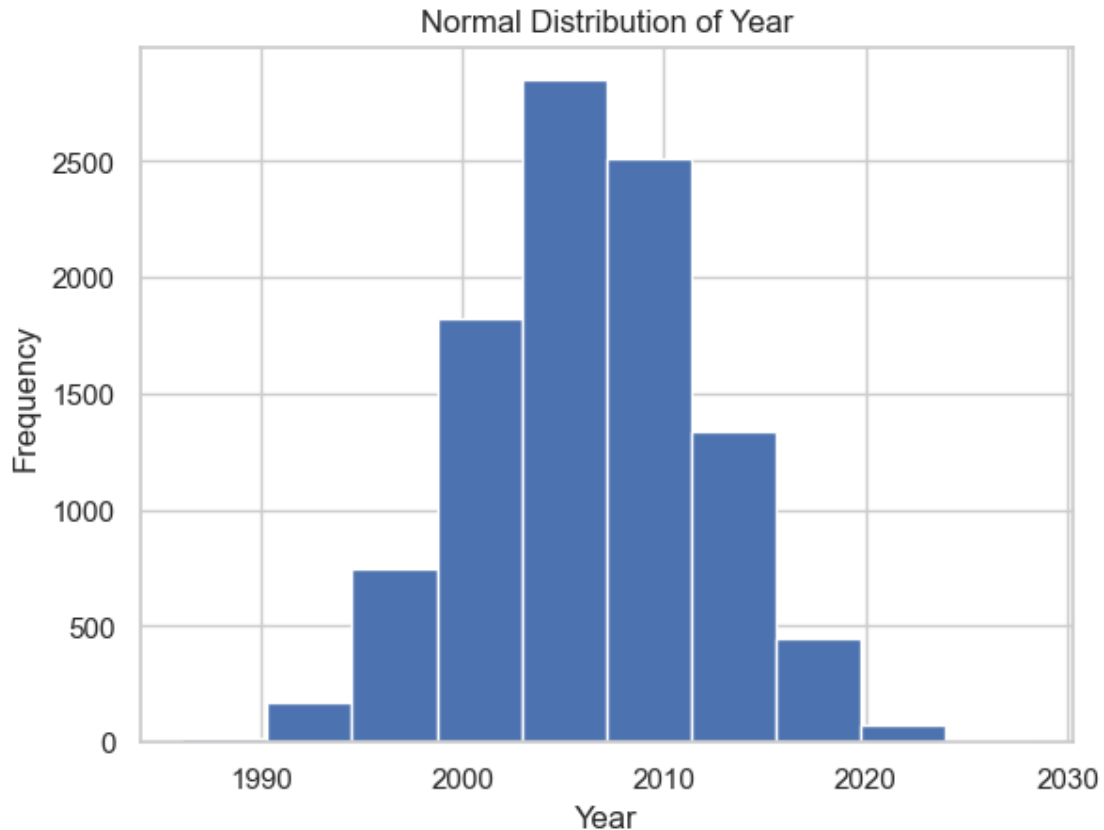
```

```

norm_dist = np.random.normal(mean, std, size=10000)

# Plot the normal distribution
plt.hist(norm_dist)
plt.xlabel('Year')
plt.ylabel('Frequency')
plt.title('Normal Distribution of Year')
plt.show()

```



```
[81]: data_hist_log = df.copy()
```

```
[82]: data_hist_log = data_hist_log[data_hist_log.NA_Sales != 0]
data_hist_log = data_hist_log[data_hist_log.EU_Sales != 0]
data_hist_log = data_hist_log[data_hist_log.Other_Sales != 0]
data_hist_log = data_hist_log[data_hist_log.JP_Sales != 0]
data_hist_log = data_hist_log[data_hist_log.Global_Sales != 0]
```

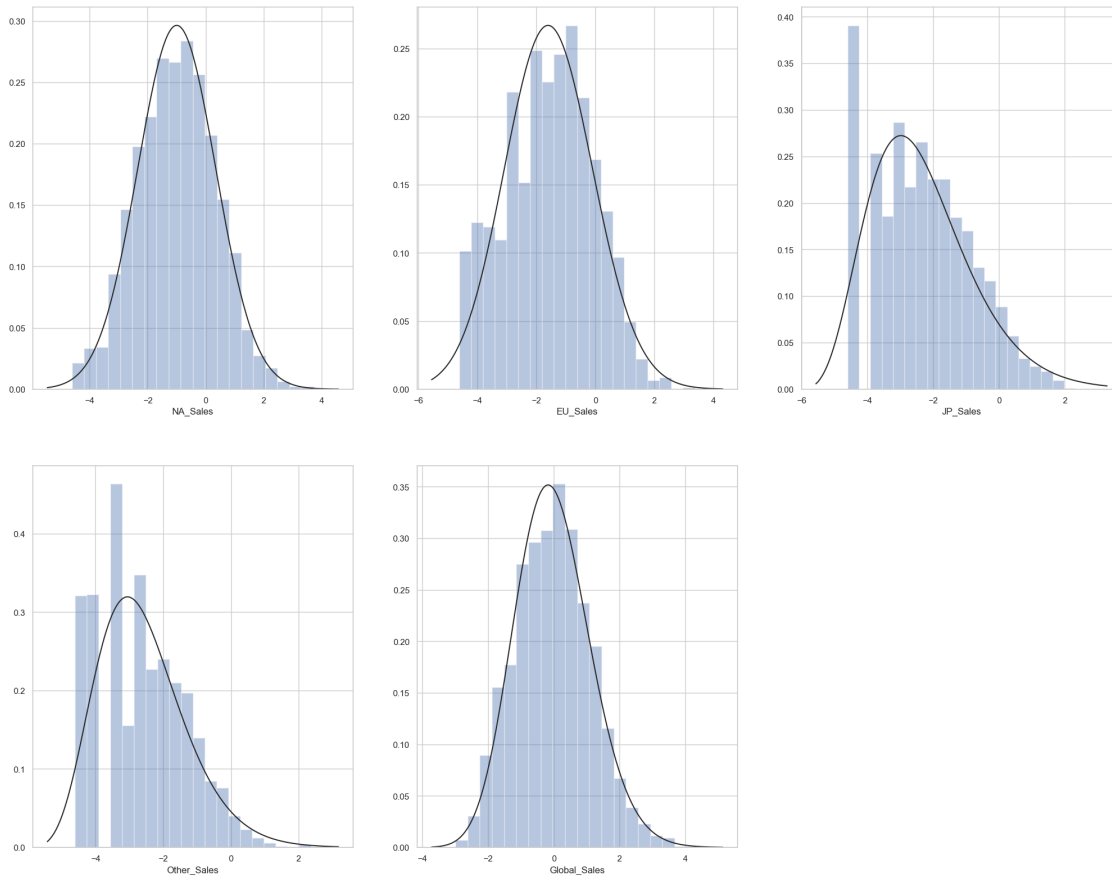
```
[83]: from scipy import stats

plt.figure(figsize=(25,30))
```

```

sales_columns = ['NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales']
for i, column in enumerate(sales_columns):
    plt.subplot(3,3,i+1)
    sns.distplot(np.log(data_hist_log[column]), bins=20, kde=False, fit=stats.gamma)
plt.show()

```



we can see their skewness and central tendency. we can say that our almost all data are right skewed and symmetric skewed after log transformation

1.8 Hypothesis Testing

Hypothesis 1: The average video game sales in North America is higher than the average video game sales in Europe. $H_0: \mu_{NA} \leq \mu_{EU}$

$H_a: \mu_{NA} > \mu_{EU}$

Hypothesis 2: The variance of video game sales in Japan is higher than the variance of video game sales in North America. $H_0: \text{var}_{JP} \leq \text{var}_{NA}$

Ha: $\text{var_JP} > \text{var_NA}$

1.8.1 a two-sample t-test to compare the means of the two groups.

$t = (\bar{x}_1 - \bar{x}_2) / \sqrt{s_1^2 / n_1 + s_2^2 / n_2}$ where:

\bar{x}_1 and \bar{x}_2 are the sample means s_1^2 and s_2^2 are the sample variances n_1 and n_2 are the sample sizes

1.8.2 F-test to compare the variances of the two groups.

$F = s_1^2 / s_2^2$

```
[84]: na_sales = df['NA_Sales']
      eu_sales = df['EU_Sales']

      # Conduct a two-sample t-test to compare the means of the two groups
      t_statistic, p_value = ttest_ind(na_sales, eu_sales)

      # Print the results of the t-test
      print('Two-sample t-test results:')
      print('t-statistic:', t_statistic)
      print('p-value:', p_value)

      # Extract the sales columns for Japan and North America
      jp_sales = df['JP_Sales']

      # Conduct an F-test to compare the variances of the two groups
      f_statistic, p_value = f_oneway(jp_sales, na_sales)

      # Print the results of the F-test
      print('F-test results:')
      print('F-statistic:', f_statistic)
      print('p-value:', p_value)
```

```
Two-sample t-test results:
t-statistic: 15.831371100049914
p-value: 3.0384925458015177e-56
F-test results:
F-statistic: 760.1394504077623
p-value: 1.8417918080111082e-165
```

Hypothesis 1: The t-statistic of 15.83 and p-value of 3.0384925458015177e-56 provide strong evidence that the mean video game sales in North America is significantly higher than the mean video game sales in Europe.

Hypothesis 2: The F-statistic of 760.14 and p-value of 1.8417918080111082e-165 provide even stronger evidence that the variance of video game sales in Japan is significantly higher than the variance of video game sales in North America.

2 SECTION __ II

2.1 Questions for Analysis

2.1.1 1.How is sales distribution across the global market?

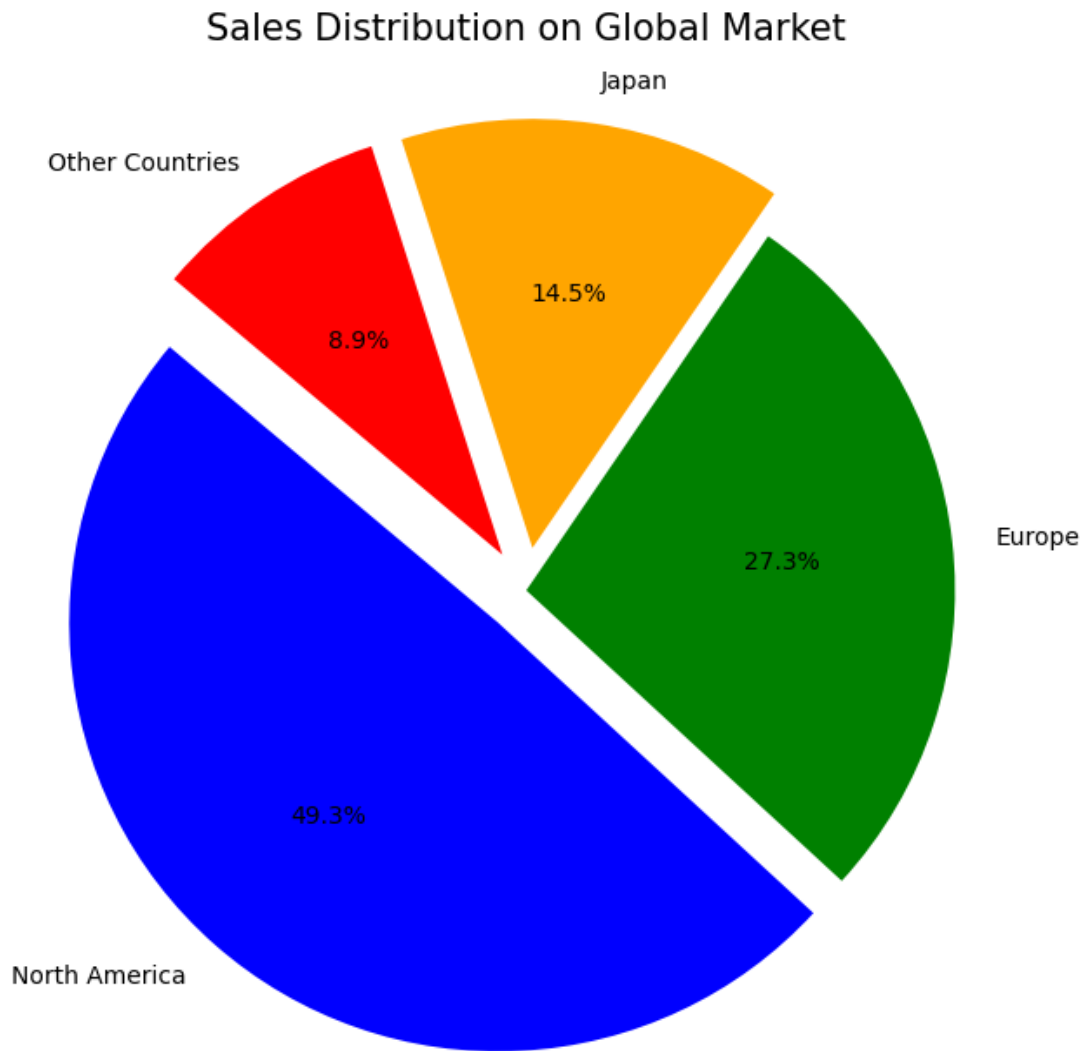
```
[47]: ## Sum of all Sales
total_na_sales = df['NA_Sales'].sum()
total_eu_sales = df['EU_Sales'].sum()
total_jp_sales = df['JP_Sales'].sum()
total_other_sales = df['Other_Sales'].sum()

# Creating a list of sales values for each region
sales_values = [total_na_sales, total_eu_sales, total_jp_sales,
                ↪total_other_sales]
regions = ['North America', 'Europe', 'Japan', 'Other Countries']

# Create a pie chart for the sales distribution
plt.figure(figsize=(8,8))
plt.pie(sales_values, labels=regions,explode = [0.1,0,0.1,0.1], autopct="%1.
    ↪1f%", startangle=140, colors=['blue', 'green', 'orange', 'red'])

## Title
plt.title("Sales Distribution on Global Market", fontsize=15)

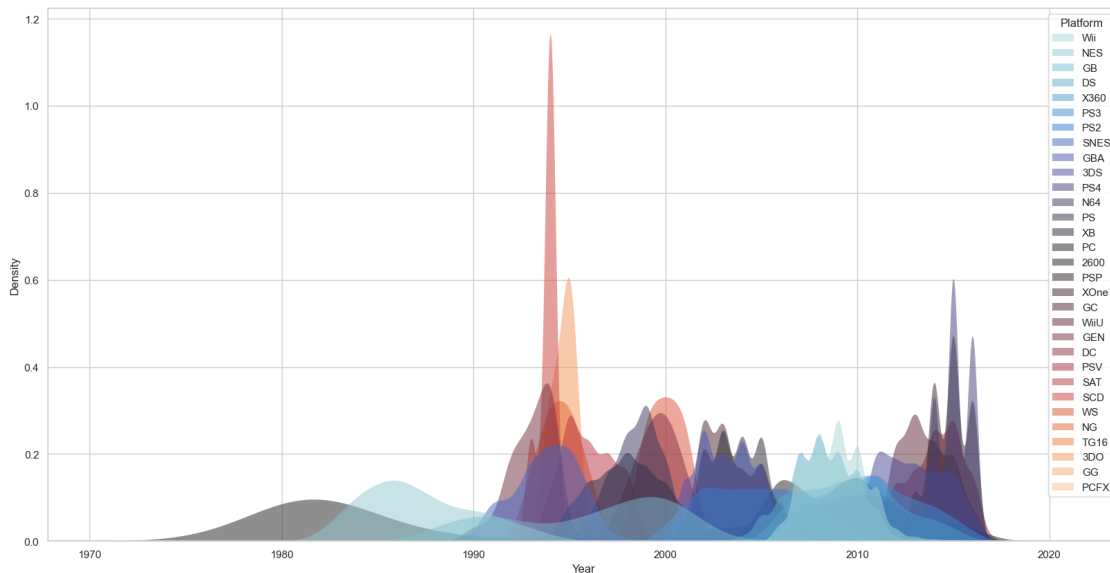
## Display of plot
plt.show()
```



North America has the highest shares

2.1.2 2 .list the platforms that support video games?

```
[158]: plt.figure(figsize=(20,10))
sns.kdeplot( data=df, x='Year',hue="Platform", fill=True, common_norm=False,
            palette="icefire", alpha=.5, linewidth=0,legend=True)
plt.show()
```



in the early 1980s, 2600 were the most popular platform.

by mid 1980s, NES became very popular.

at the start of the 1990s, GB was popular

in mid 1990s GB declined while SNES, DC, 3DO, WS, got more popular. SAT was highly grossing of all the years at that time.

in late 1990s, PS, XB, and NES resurfaced

early 2000s, SCD and SAT along with NES were highly grossing

mid 2000s, PSP, WiiU, PS2, SNES resurfaced, GBA,GC were popular

late 2000s, GB resurfaced, Wii, PS3, X360, DS were highly used gaming platforms.

early 2010s, GB still there, Wii, PS3, X360, DS, in addition to 3DS, WiiU, were highly grossing gaming platforms.

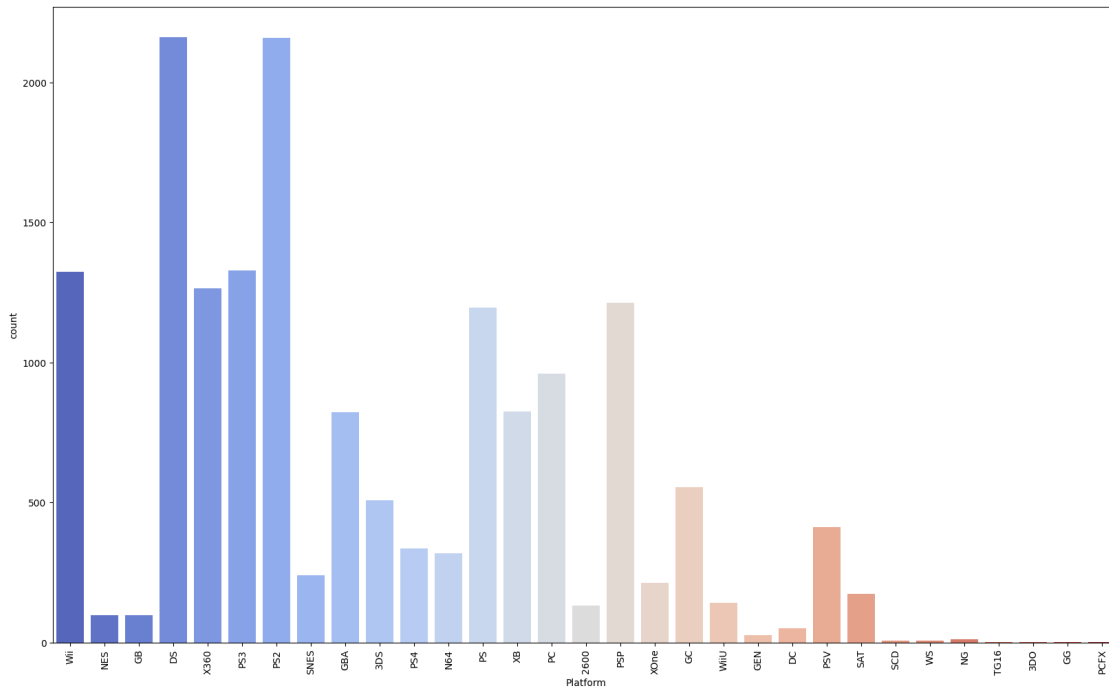
mid 2010s, PS4 was very popular, PSV, Wiii still grossing and so does 3DS, PCs were very popular.

SAT was the highest grossing gaming platform of all time

NES was the longest grossing gaming platform for up to 2 and a half decades!! 25 years of NES.

2.1.3 3. Number of games that support these Platforms?

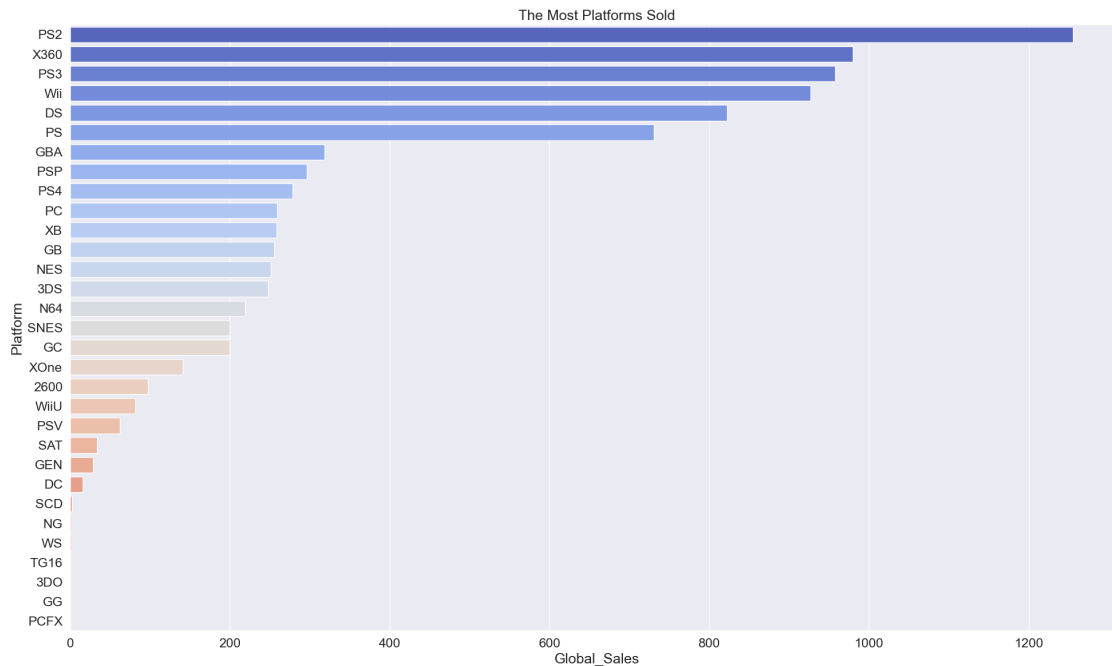
```
[48]: plt.figure(figsize=(20,12))
b=sns.countplot(x= 'Platform' ,data = df ,palette='coolwarm' )
plt.xticks(rotation=90)
plt.show()
```



DS, PS2, PS3, Wii, X360, PSP, PS, PC, XB, GBA, GC, 3DS, PSV, PS4, N64, SNES, XOne, SAT, WiiU, 2600, NES, GB, DC, GEN, NG, SCD, WS, 3DO, TG16, GG, PCFX

2.1.4 4. Global Sales of these Platforms?

```
[54]: plt.figure(figsize=(20, 12))
sns.set(font_scale = 1.5)
genre=df.groupby(["Platform"])["Global_Sales"].sum().reset_index()
genre=genre.sort_values(by="Global_Sales",ascending=False)
sns.barplot(y="Platform",x="Global_Sales",data=genre,palette='coolwarm' )
plt.title("The Most Platforms Sold")
plt.tight_layout()
plt.show()
```



PS2 has the highest sales globally

2.1.5 5. How many Games were released each year?

```
[58]: # Group the data by year and count the number of games released each year
yearly_releases = df.groupby("Year")["Name"].count()
yearly_releases
```

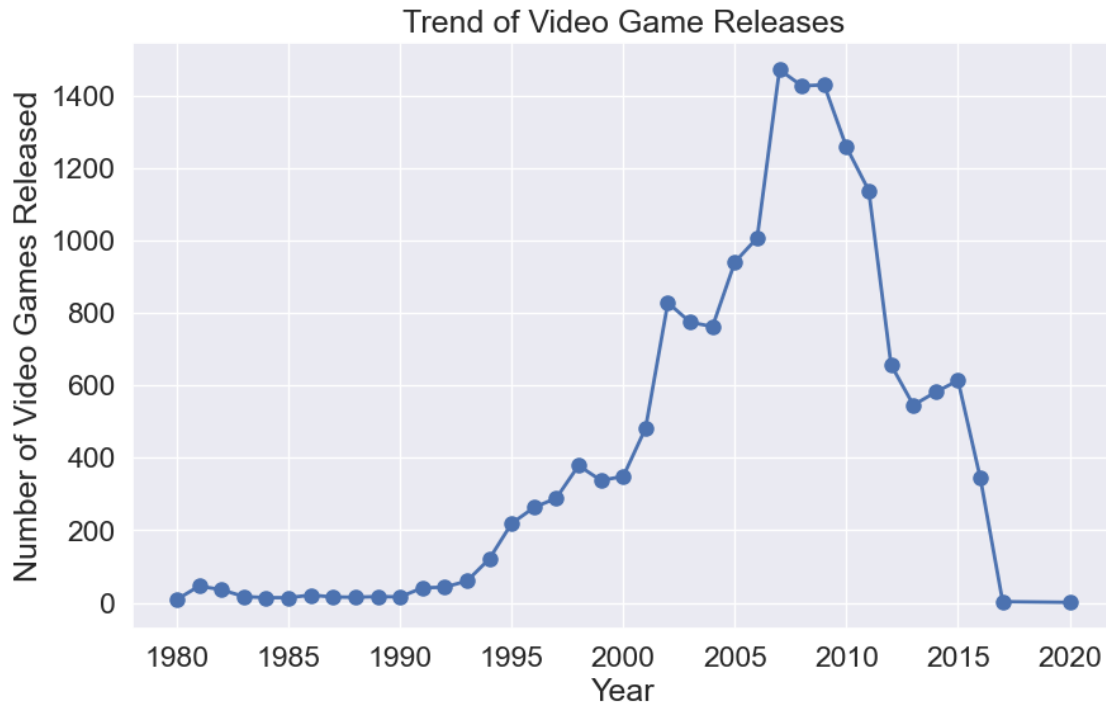
```
[58]: Year
1980.0    9
1981.0   46
1982.0   36
1983.0   17
1984.0   14
1985.0   14
1986.0   21
1987.0   16
1988.0   15
1989.0   17
1990.0   16
1991.0   41
1992.0   43
1993.0   60
1994.0  121
1995.0  219
1996.0  263
```

1997.0	289
1998.0	379
1999.0	338
2000.0	349
2001.0	482
2002.0	829
2003.0	775
2004.0	763
2005.0	941
2006.0	1008
2007.0	1473
2008.0	1428
2009.0	1431
2010.0	1259
2011.0	1139
2012.0	657
2013.0	546
2014.0	582
2015.0	614
2016.0	344
2017.0	3
2020.0	1

Name: Name, dtype: int64

2.1.6 6. Trends in Video Game releases each year.

```
[59]: # Create a line chart for the trend of video game releases each year
plt.figure(figsize=(10, 6))
plt.plot(yearly_releases.index, yearly_releases.values, marker='o', color='b',
         linestyle='-', linewidth=2, markersize=8)
plt.xlabel("Year")
plt.ylabel("Number of Video Games Released")
plt.title("Trend of Video Game Releases")
plt.grid(True)
plt.show()
```



Increasing Trend (2005-2009): The number of video game releases steadily increased from 941 in 2005 to 1473 in 2007, reaching a peak of 1473 in 2007. This period saw a significant rise in the gaming industry.

Decreasing Trend (2009-2016): After 2009, there was a gradual decline in the number of game releases. The numbers decreased from 1473 in 2007 to 344 in 2016. This decline could be attributed to various factors, such as market saturation or shifts in gaming platforms.

Stabilization (2016-2020): After 2016, the number of game releases stabilized around 300- 400 per year. There might be a new balance in the industry, with a consistent number of games being released annually.

Significant Drops (2017-2020): There was a sharp drop in 2017 with only 3 releases, and in 2020, there was only 1 release. These years might represent a shift in the industry, possibly due to emerging trends like mobile gaming or changes in consumer preferences.

2.1.7 7. What is the distribution of genre in Video Games?

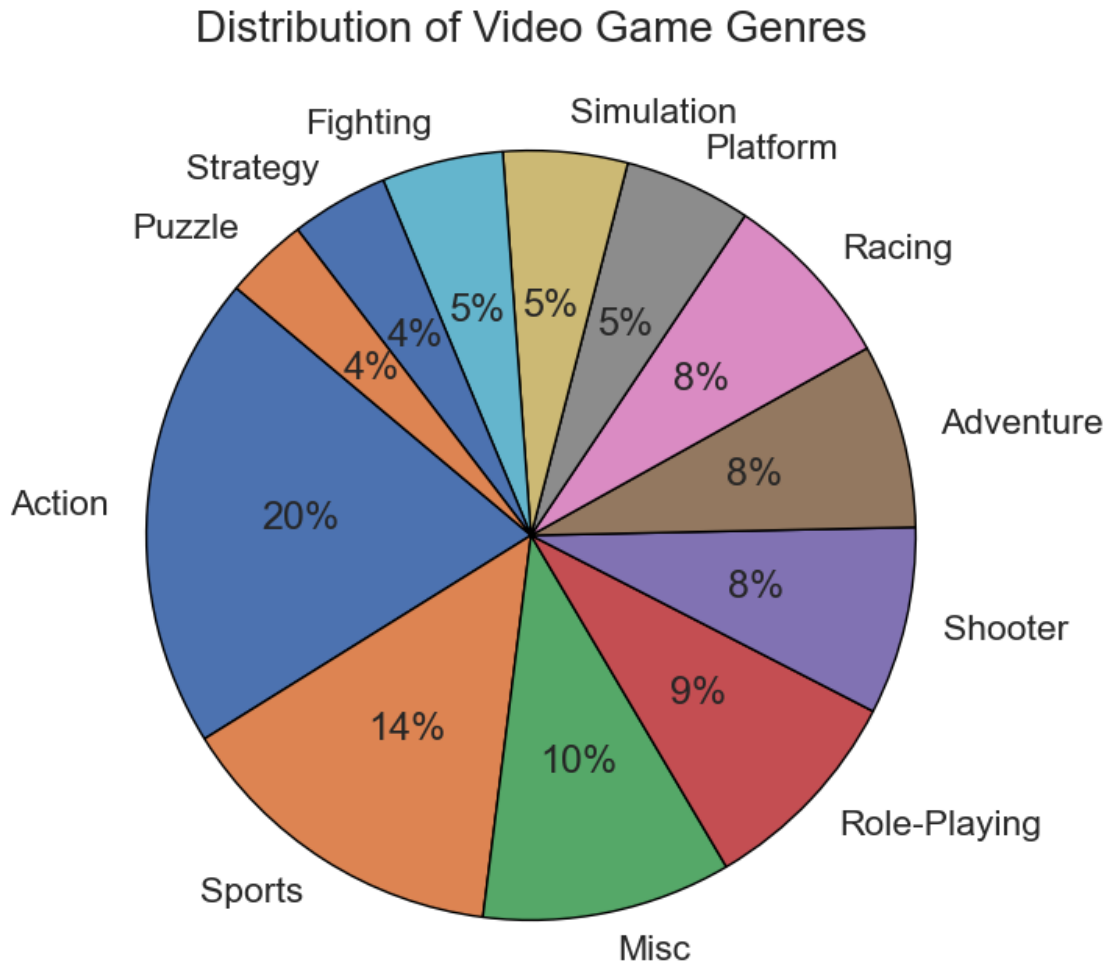
```
[63]: all_genres = df['Genre'].value_counts()

## Plotting the graph
plt.figure(figsize=(8,8))
plt.pie(all_genres, labels=all_genres.index, wedgeprops={"edgecolor":"black"},
        autopct='%1.0f%%', startangle=140)

## Title
```

```
plt.title('Distribution of Video Game Genres', fontsize=20)

## Display of plot
plt.show()
```



Action, sports, and misc games dominate the video game market. Role-playing, shooter, adventure, racing, platform, simulation, and fighting games have moderate popularity. Strategy and puzzle games have the lowest demand. Each genre appeals to different types of players and preferences.

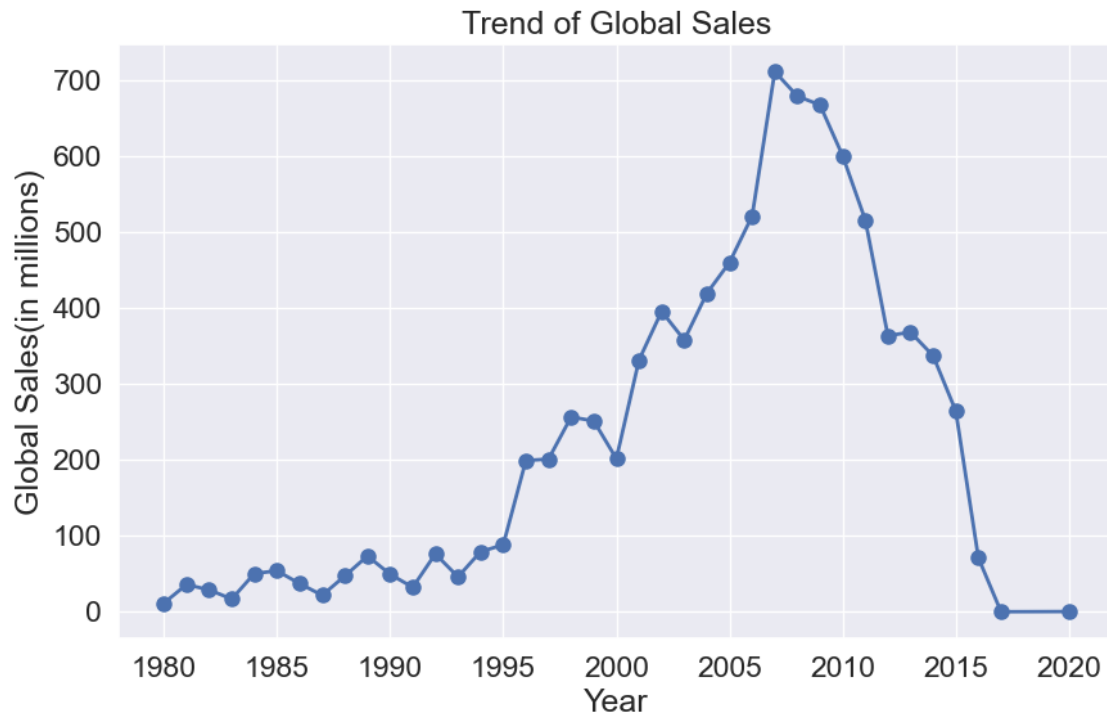
2.1.8 8. What is the trend of Global Sales each year?

```
[65]: # Group the data by year and sum the global sales for each year
yearly_sales = df.groupby("Year")["Global_Sales"].sum()

# Create a line chart for the trend of global sales each year
```



```
plt.figure(figsize=(10, 6))
plt.plot(yearly_sales.index, yearly_sales.values, marker='o', color='b',
        linestyle='-', linewidth=2, markersize=8)
plt.xlabel("Year")
plt.ylabel("Global Sales(in millions)")
plt.title("Trend of Global Sales")
plt.grid(True)
plt.show()
```



1980s to early 1990s: Global sales started relatively low in the early 1980s, gradually increased, and then spiked in 1989 and 1994.

The mid-1990s to early 2000s: After the spike in 1994, sales remained consistently high in the mid to late 1990s and early 2000s, peaking in 1998.

Early 2000s: There was a slight decline in the early 2000s, followed by a significant increase in 2001 and 2002.

The mid-2000s to 2010s: Sales remained high in the mid-2000s, reaching its peak in 2008, after which there was a notable decline. The sales gradually stabilized but remained relatively high until 2010.

2010s to 2020: From 2010 onwards, there was a general downward trend, with sales decreasing each year. There was a steep decline in 2013, followed by a slight increase in 2014, but the overall trend remained negative.

2016 to 2020: Sales continued to decrease significantly, reaching almost negligible levels in 2017 and remaining extremely low in 2020.

2.1.9 9. What is the top-sold Genre in the North American market?

```
[68]: # Group data by genre and calculate total sales for North America
genre_sales_na = df.groupby("Genre")["NA_Sales"].sum()

# Find the genre with the highest sales in North America
top_genre_na = genre_sales_na.idxmax()
top_sales_na = genre_sales_na.max()

print(f"The top-selling genre in North America is '{top_genre_na}' with total_
↳ sales of {top_sales_na} million copies.")
```

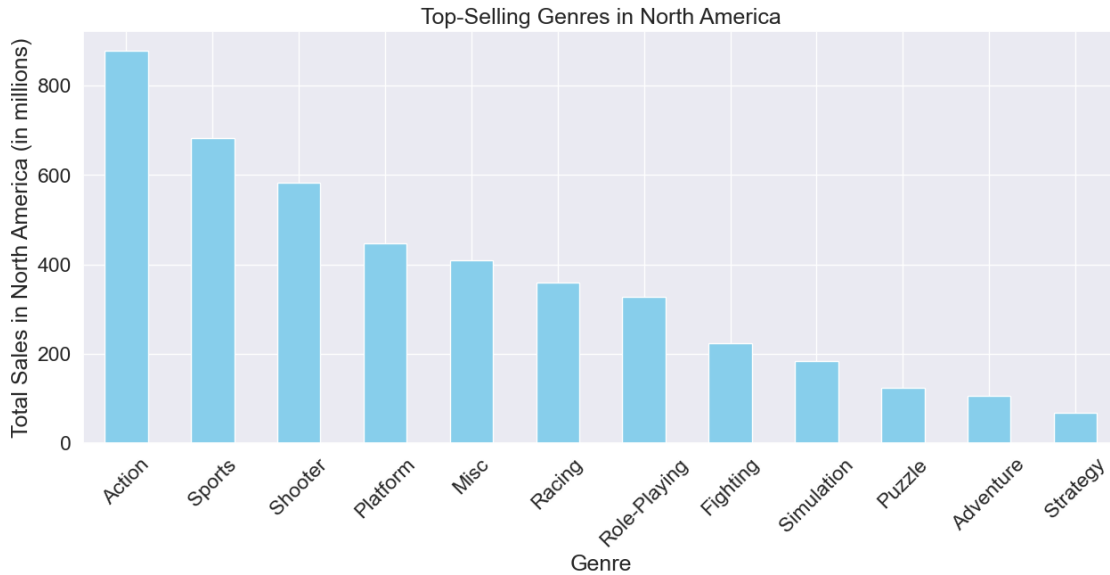
The top-selling genre in North America is 'Action' with total sales of 877.83 million copies.

```
[72]: genre_sales_na
```

```
[72]: Genre
Action      877.83
Sports      683.35
Shooter     582.60
Platform    447.05
Misc        410.24
Racing      359.42
Role-Playing 327.28
Fighting    223.59
Simulation   183.31
Puzzle       123.78
Adventure    105.80
Strategy      68.70
Name: NA_Sales, dtype: float64
```

```
[69]: genre_sales_na = genre_sales_na.sort_values(ascending=False)

# Bar plot for top-selling genres in North America
plt.figure(figsize=(15, 6))
genre_sales_na.plot(kind='bar', color='skyblue')
plt.xlabel("Genre")
plt.ylabel("Total Sales in North America (in millions)")
plt.title("Top-Selling Genres in North America")
plt.xticks(rotation=45)
plt.show()
```



2.1.10 10. What is the top-sold Genre in the European market?

```
[70]: # Group data by genre and calculate total sales for Europe
genre_sales_eu = df.groupby("Genre")["EU_Sales"].sum()

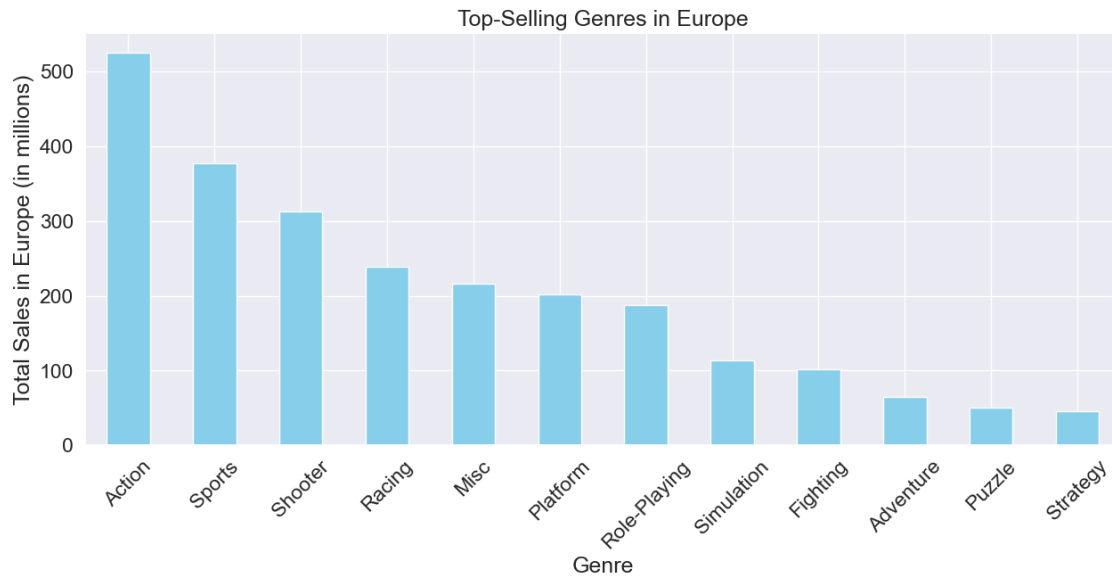
# Find the genre with the highest sales in Europe
top_genre_eu = genre_sales_eu.idxmax()
top_sales_eu = genre_sales_eu.max()

print(f"The top-selling genre in Europe is '{top_genre_eu}' with total sales of {top_sales_eu} million copies.")
```

The top-selling genre in Europe is 'Action' with total sales of 525.0 million copies.

```
[71]: genre_sales_eu = genre_sales_eu.sort_values(ascending=False)

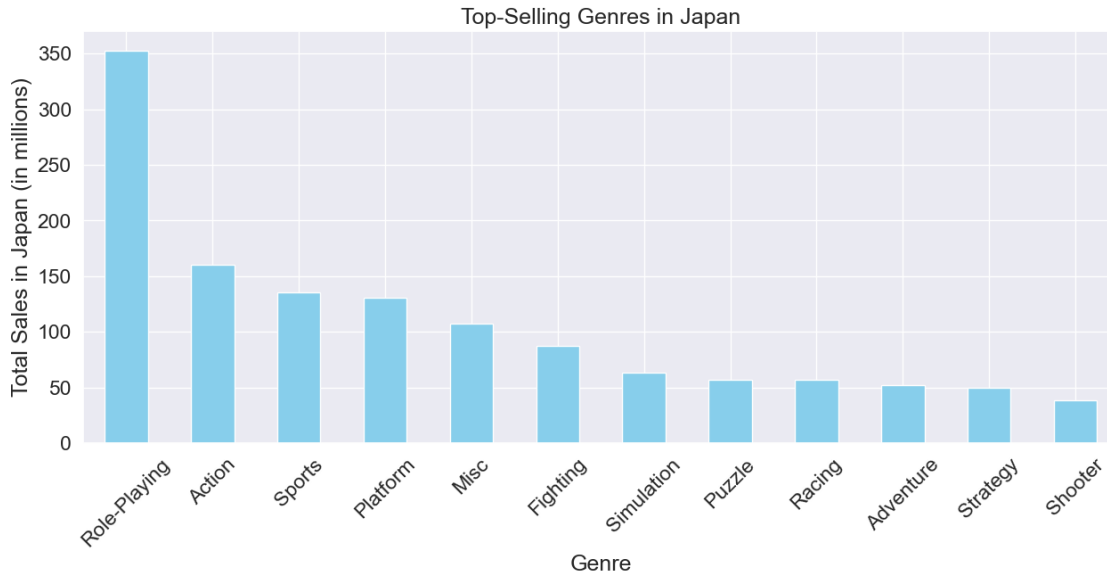
# Bar plot for top-selling genres in Europe
plt.figure(figsize=(15, 6))
genre_sales_eu.plot(kind='bar', color='skyblue')
plt.xlabel("Genre")
plt.ylabel("Total Sales in Europe (in millions)")
plt.title("Top-Selling Genres in Europe")
plt.xticks(rotation=45)
plt.show()
```



2.1.11 11. What is the top-sold Genre in the Japanese market?

```
[74]: genre_sales_jp = genre_sales_jp.sort_values(ascending=False)
```

```
# Bar plot for top-selling genres in Japan
plt.figure(figsize=(15, 6))
genre_sales_jp.plot(kind='bar', color='skyblue')
plt.xlabel("Genre")
plt.ylabel("Total Sales in Japan (in millions)")
plt.title("Top-Selling Genres in Japan")
plt.xticks(rotation=45)
plt.show()
```



2.1.12 12. What is the top-sold Genre in Other Country Markets?

```
[75]: # Group data by genre and calculate total sales for Other
genre_sales_Other = df.groupby("Genre")["Other_Sales"].sum()

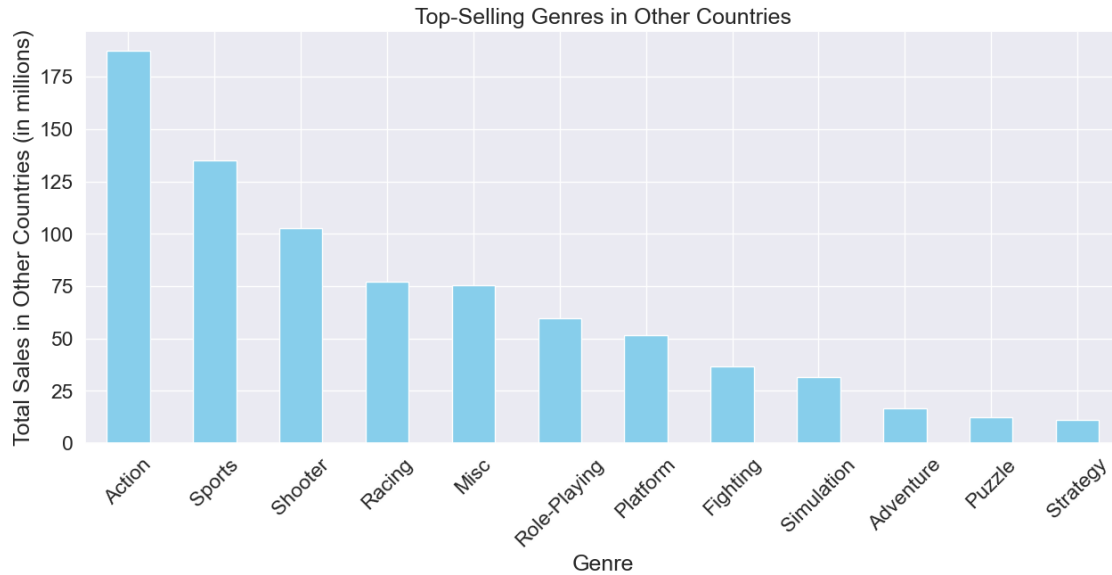
# Find the genre with the highest sales in Other Countries
top_genre_Other = genre_sales_Other.idxmax()
top_sales_Other = genre_sales_Other.max()

print(f"The top-selling genre in Other Countries is '{top_genre_Other}' with
↳total sales of {top_sales_Other} million copies.")
```

The top-selling genre in Other Countries is 'Action' with total sales of 187.38 million copies.

```
[76]: genre_sales_Other = genre_sales_Other.sort_values(ascending=False)

# Bar plot for top-selling genres in Other Countries
plt.figure(figsize=(15, 6))
genre_sales_Other.plot(kind='bar', color='skyblue')
plt.xlabel("Genre")
plt.ylabel("Total Sales in Other Countries (in millions)")
plt.title("Top-Selling Genres in Other Countries")
plt.xticks(rotation=45)
plt.show()
```



2.1.13 13. What is the top-sold Genre in the Global Market?

```
[77]: # Group data by genre and calculate total Global sales
genre_sales_global = df.groupby("Genre")["Global_Sales"].sum()

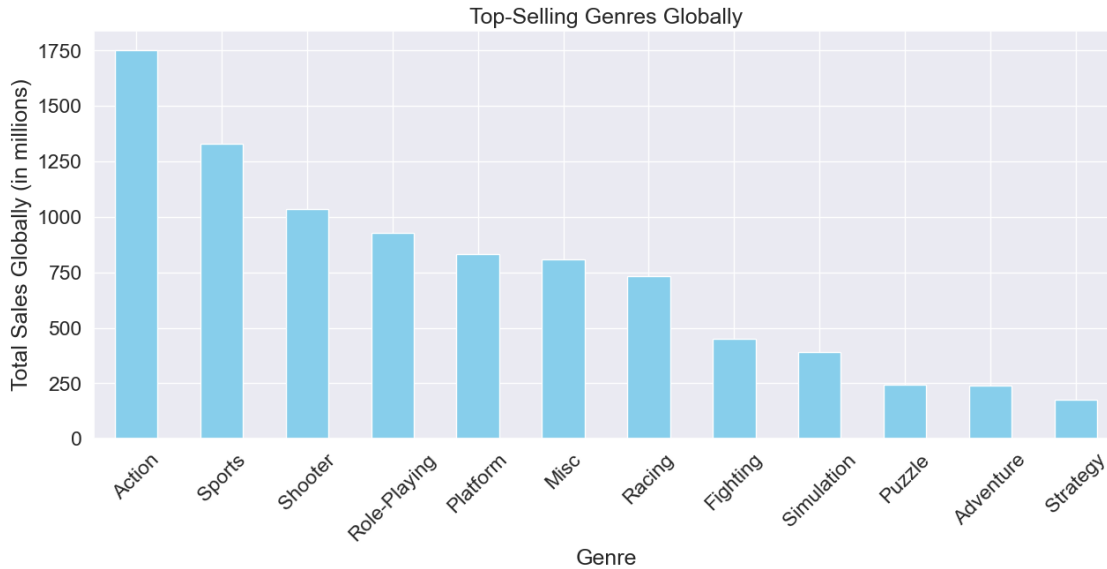
# Find the genre with the highest sales Globally
top_genre_global = genre_sales_global.idxmax()
top_sales_global = genre_sales_global.max()

print(f"The top-selling genre Globally is '{top_genre_global}' with total sales_
of {top_sales_global} million copies.")
```

The top-selling genre Globally is 'Action' with total sales of 1751.18 million copies.

```
[78]: genre_sales_global = genre_sales_global.sort_values(ascending=False)

# Bar plot for top-selling genres Globally
plt.figure(figsize=(15, 6))
genre_sales_global.plot(kind='bar', color='skyblue')
plt.xlabel("Genre")
plt.ylabel("Total Sales Globally (in millions)")
plt.title("Top-Selling Genres Globally")
plt.xticks(rotation=45)
plt.show()
```



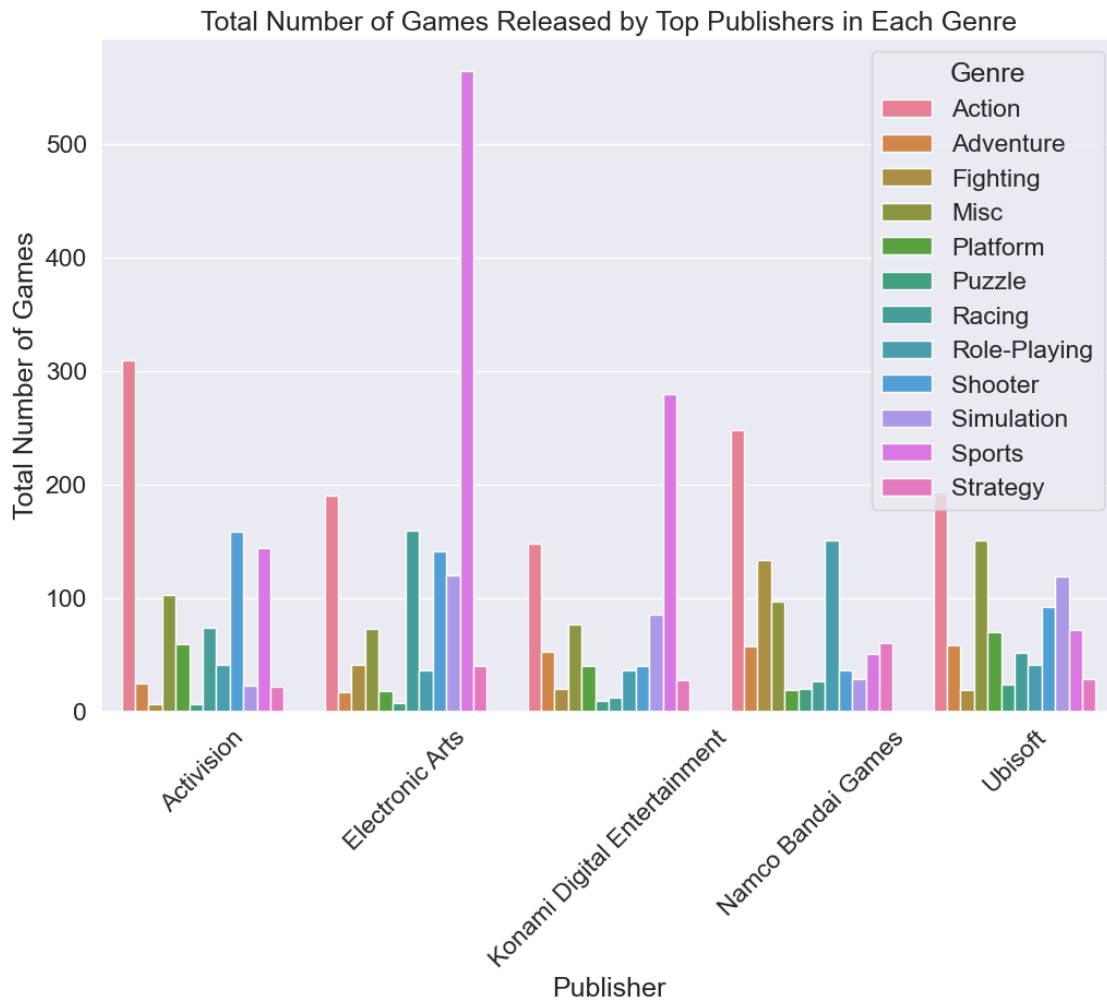
2.1.14 14. Top Publishers Who Have Released More Video Games?

```
[80]: df['Publisher'].value_counts()
```

```
[80]: Publisher
Electronic Arts          1409
Activision               975
Namco Bandai Games      932
Ubisoft                 921
Konami Digital Entertainment 832
...
Warp                    1
New                    1
Elite                  1
Evolution Games        1
UIG Entertainment      1
Name: count, Length: 578, dtype: int64
```

```
[151]: top_publishers = df.groupby('Publisher').size().nlargest(5).index
top_publisher_genre_counts = df[df['Publisher'].isin(top_publishers)].
    ↳groupby(['Publisher', 'Genre']).size().reset_index(name='Count')
plt.figure(figsize=(12, 8))
sns.barplot(data=top_publisher_genre_counts, x='Publisher', y='Count',
    ↳hue='Genre')
plt.xticks(rotation=45)
plt.xlabel('Publisher')
plt.ylabel('Total Number of Games')
```

```
plt.title('Total Number of Games Released by Top Publishers in Each Genre')
plt.show()
```



```
[91]: # Top 10 publishers by global sales
top_publishers = df.Publisher.value_counts().head(10)
top_publishers
```

```
[91]: Publisher
Electronic Arts      1409
Activision           975
Namco Bandai Games   932
Ubisoft              921
Konami Digital Entertainment  832
THQ                  715
Nintendo             703
Sony Computer Entertainment  683
```



```

Sega
Take-Two Interactive
Name: count, dtype: int64
639
413

```

```

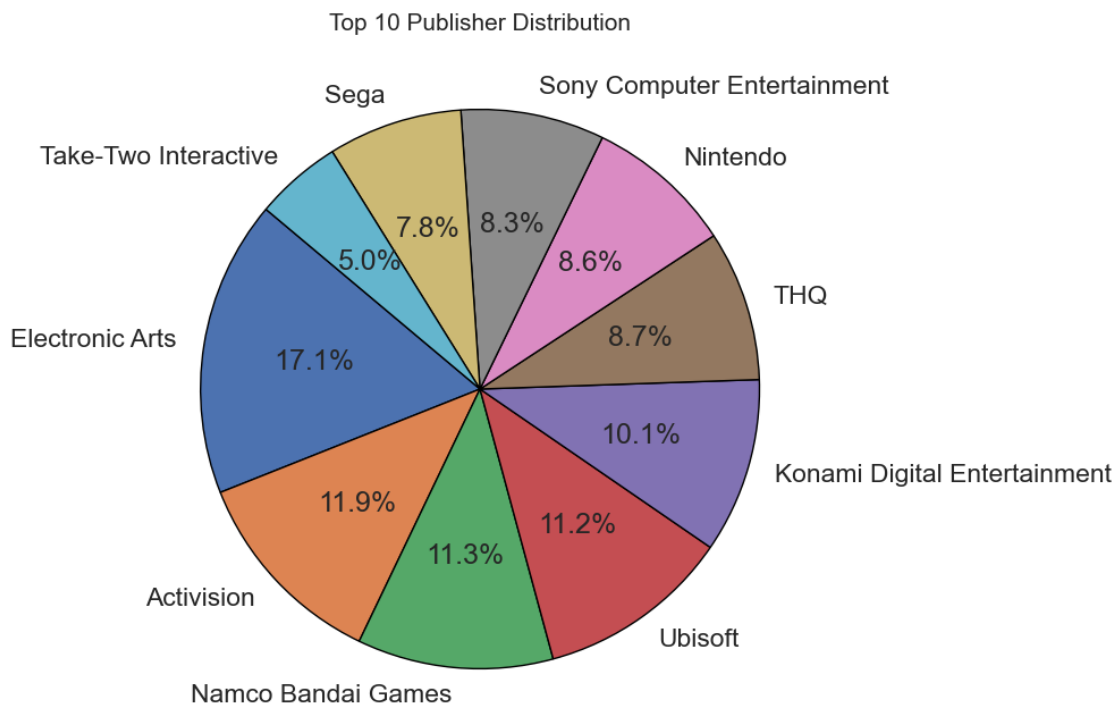
[92]: ## Plotting graph
plt.figure(figsize=(8,8))

plt.pie(top_publishers, labels=top_publishers.index, wedgeprops={"edgecolor": "black"}, autopct='%1.1f%%', startangle=140)

## Title
plt.title('Top 10 Publisher Distribution', fontsize = 15)

## Display of the plot
plt.show()

```



2.1.15 15. Top Publishers Who Have Made Most Sales on Video Games?

```

[99]: sales_publisher = df.groupby("Publisher").agg({"Global_Sales": pd.Series.sum})
sales_publisher.nlargest(10, "Global_Sales")

```

```

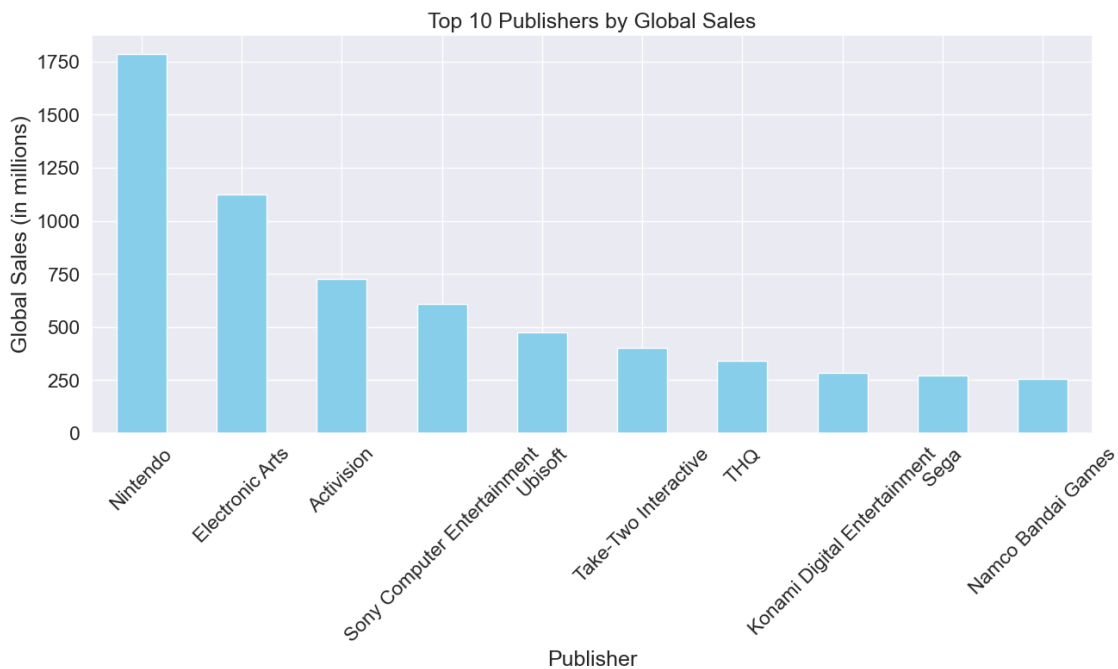
[99]:
Global_Sales
Publisher

```

Nintendo	1786.56
Electronic Arts	1125.20
Activision	727.46
Sony Computer Entertainment	607.50
Ubisoft	474.72
Take-Two Interactive	399.54
THQ	340.77
Konami Digital Entertainment	283.64
Sega	272.99
Namco Bandai Games	254.09

```
[102]: # Group the data by publisher and sum the global sales for each publisher
publisher_sales = df.groupby("Publisher")["Global_Sales"].sum().nlargest(10)

# Create a bar plot for the top 10 publishers by global sales
plt.figure(figsize=(15, 6))
publisher_sales.plot(kind='bar', color='skyblue')
plt.xlabel("Publisher")
plt.ylabel("Global Sales (in millions)")
plt.title("Top 10 Publishers by Global Sales")
plt.xticks(rotation=45)
plt.show()
```



```
[110]: comp_publisher = df[['Publisher', 'NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales', 'Global_Sales']]
```

```
comp_publisher.head()
```

```
[110]:
```

	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
0	Nintendo	41.49	29.02	3.77	8.46	82.74
1	Nintendo	29.08	3.58	6.81	0.77	40.24
2	Nintendo	15.85	12.88	3.79	3.31	35.82
3	Nintendo	15.75	11.01	3.28	2.96	33.00
4	Nintendo	11.27	8.89	10.22	1.00	31.37

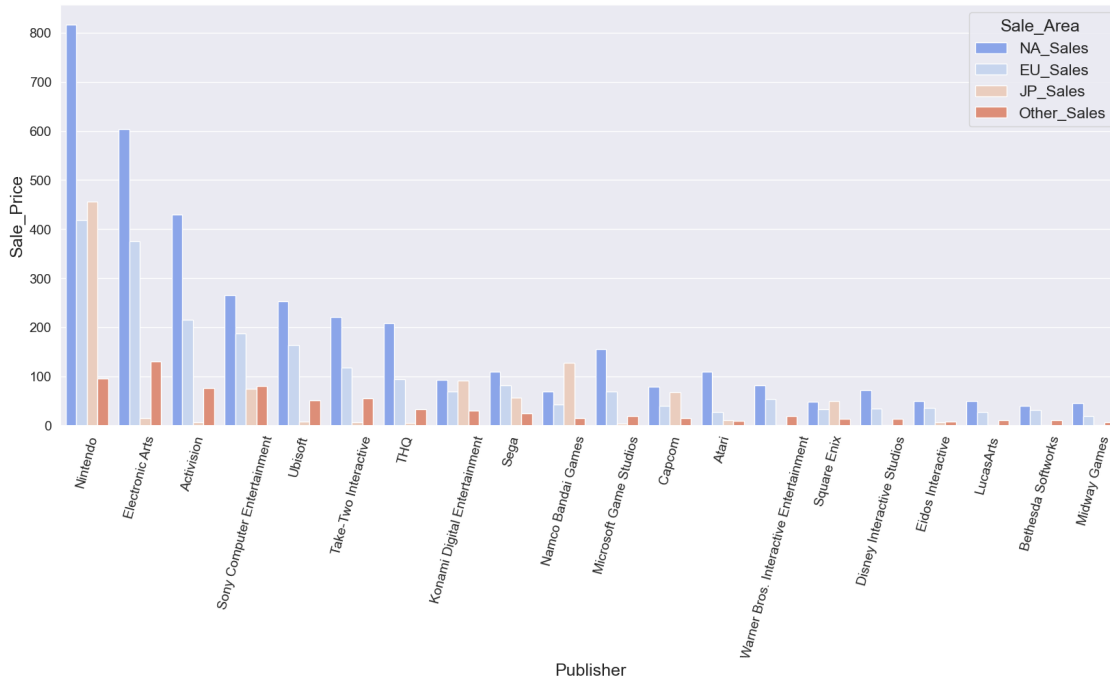
```
[111]: comp_publisher = comp_publisher.groupby(by=['Publisher']).sum().reset_index().
↳sort_values(by=['Global_Sales'], ascending=False)
comp_publisher = comp_publisher.head(20)
comp_publisher = pd.melt(comp_publisher,
↳id_vars=['Publisher'],value_vars=['NA_Sales', 'EU_Sales', 'JP_Sales',
↳'Other_Sales'],var_name='Sale_Area', value_name='Sale_Price')
comp_publisher
```

```
[111]:
```

	Publisher	Sale_Area	Sale_Price
0	Nintendo	NA_Sales	816.87
1	Electronic Arts	NA_Sales	603.61
2	Activision	NA_Sales	429.70
3	Sony Computer Entertainment	NA_Sales	265.22
4	Ubisoft	NA_Sales	253.43
..
75	Disney Interactive Studios	Other_Sales	13.18
76	Eidos Interactive	Other_Sales	8.02
77	LucasArts	Other_Sales	10.50
78	Bethesda Softworks	Other_Sales	10.16
79	Midway Games	Other_Sales	5.74

[80 rows x 3 columns]

```
[112]: plt.figure(figsize=(20,8))
sns.barplot(x='Publisher', y='Sale_Price',
↳hue='Sale_Area',data=comp_publisher,palette='coolwarm' )
plt.xticks(fontsize=14, rotation=75)
plt.yticks(fontsize=14)
plt.show()
```



2.1.16 16. What are the top-selling game genres across different market areas?

```
[178]: # Group data by genre and calculate total sales for each region
genre_sales_by_region = df.groupby("Genre")[["NA_Sales", "EU_Sales", "JP_Sales", "Other_Sales", "Global_Sales"]].sum()
genre_sales_by_region
```

```
[178]:
```

	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
Genre					
Action	877.83	525.00	159.95	187.38	1751.18
Adventure	105.80	64.13	52.07	16.81	239.04
Fighting	223.59	101.32	87.35	36.68	448.91
Misc	410.24	215.98	107.76	75.32	809.96
Platform	447.05	201.63	130.77	51.59	831.37
Puzzle	123.78	50.78	57.31	12.55	244.95
Racing	359.42	238.39	56.69	77.27	732.04
Role-Playing	327.28	188.06	352.31	59.61	927.37
Shooter	582.60	313.27	38.28	102.69	1037.37
Simulation	183.31	113.38	63.70	31.52	392.20
Sports	683.35	376.85	135.37	134.97	1330.93
Strategy	68.70	45.34	49.46	11.36	175.12

```
[113]: genre_sales = df.groupby('Genre')[['NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales']].sum()
```

```

top_na_genre = genre_sales['NA_Sales'].idxmax()
top_eu_genre = genre_sales['EU_Sales'].idxmax()
top_jp_genre = genre_sales['JP_Sales'].idxmax()
top_other_genre = genre_sales['Other_Sales'].idxmax()
print(f'Top-selling genre in North America: {top_na_genre}')
print(f'Top-selling genre in Europe: {top_eu_genre}')
print(f'Top-selling genre in Japan: {top_jp_genre}')
print(f'Top-selling genre in Other countries: {top_other_genre}')

```

Top-selling genre in North America: Action
 Top-selling genre in Europe: Action
 Top-selling genre in Japan: Role-Playing
 Top-selling genre in Other countries: Action

```

[117]: genre_sales = df.groupby('Genre')[['NA_Sales', 'EU_Sales', 'JP_Sales']].sum()

# Create a bar graph
genre_sales.plot(kind='bar',figsize=(10, 6))

## Title
plt.title('Major Markets')

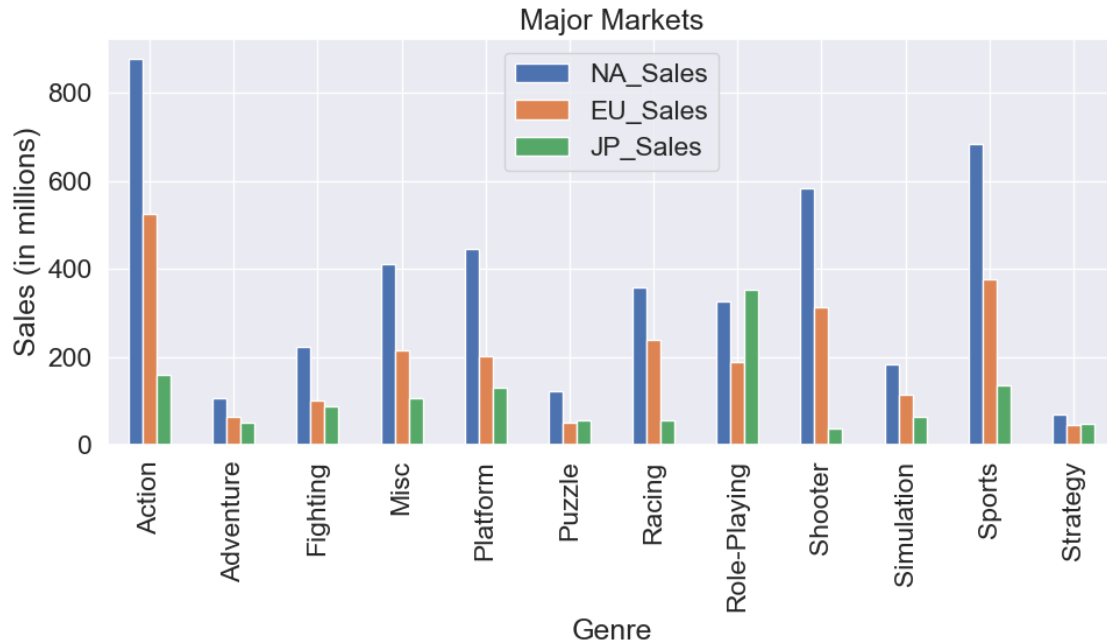
## Labels
plt.xlabel('Genre')
plt.ylabel('Sales (in millions)')

## Rotation
plt.xticks(rotation=90)

plt.legend()

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

```



2.1.17 17. Regional Analysis

```
[184]: highest_na_sales = df.loc[df.groupby('Genre')['NA_Sales'].idxmax()][['Name', 'Genre', 'Publisher', 'Year', 'NA_Sales']]
highest_na_sales = highest_na_sales.sort_values(by='NA_Sales', ascending=False)
highest_na_sales
```

```
[184]:
```

	Name	Genre	Publisher \
0	Wii Sports	Sports	Nintendo
1	Super Mario Bros.	Platform	Nintendo
9	Duck Hunt	Shooter	Nintendo
5	Tetris	Puzzle	Nintendo
2	Mario Kart Wii	Racing	Nintendo
15	Kinect Adventures!	Misc	Microsoft Game Studios
4	Pokemon Red/Pokemon Blue	Role-Playing	Nintendo
23	Grand Theft Auto V	Action	Take-Two Interactive
10	Nintendogs	Simulation	Nintendo
39	Super Smash Bros. Brawl	Fighting	Nintendo
50	Super Mario Land 2: 6 Golden Coins	Adventure	Nintendo
165	Pokemon Stadium	Strategy	Nintendo

	Year	NA_Sales
0	2006.0	41.49
1	1985.0	29.08
9	1984.0	26.93

5	1989.0	23.20
2	2008.0	15.85
15	2010.0	14.97
4	1996.0	11.27
23	2013.0	9.63
10	2005.0	9.07
39	2008.0	6.75
50	1992.0	6.16
165	1999.0	3.18

1. Nintendo dominates the list as the publisher of 10 out of 12 games being highest sold

2. Sports is the most popular genre in North America, with Wii Sports having the highest sales of all games.

3. Shooter and Action are the only genres that have non-Nintendo games as the top sellers, namely Duck Hunt and Grand Theft Auto V.

```
[185]: highest_eu_sales = df.loc[df.groupby('Genre')['EU_Sales'].idxmax()][['Name', 'Genre', 'Publisher', 'Year', 'EU_Sales']]
highest_eu_sales = highest_eu_sales.sort_values(by='EU_Sales', ascending=False)
highest_eu_sales
```

```
[185]:
```

	Name	Genre
0	Wii Sports	Sports
2	Mario Kart Wii	Racing
10	Nintendogs	Simulation
16	Grand Theft Auto V	Action
19	Brain Age: Train Your Brain in Minutes a Day	Misc
6	New Super Mario Bros.	Platform
4	Pokemon Red/Pokemon Blue	Role-Playing
34	Call of Duty: Black Ops II	Shooter
27	Brain Age 2: More Training in Minutes a Day	Puzzle
521	Myst	Adventure
39	Super Smash Bros. Brawl	Fighting
267	Warcraft II: Tides of Darkness	Strategy

	Publisher	Year	EU_Sales
0	Nintendo	2006.0	29.02
2	Nintendo	2008.0	12.88
10	Nintendo	2005.0	11.00
16	Take-Two Interactive	2013.0	9.27
19	Nintendo	2005.0	9.26
6	Nintendo	2006.0	9.23
4	Nintendo	1996.0	8.89
34	Activision	2012.0	5.88

27	Nintendo	2005.0	5.36
521	Red Orb	1994.0	2.79
39	Nintendo	2008.0	2.61
267	Activision	1995.0	2.27

1. Nintendo is also the dominant publisher in Europe, with 8 out of 12 games being highest sold.

2. Sports and Racing are the most popular genres in Europe, with Wii Sports and Mario Kart Wii having the highest sales of all games.

3. Simulation and Puzzle are only genres that have different top sellers in Europe than in North America, namely Nintendogs and Brain Age2.

```
[186]: highest_jp_sales = df.loc[df.groupby('Genre')['JP_Sales'].idxmax()][['Name', 'Genre', 'Publisher', 'Year', 'JP_Sales']]
highest_jp_sales = highest_jp_sales.sort_values(by='JP_Sales', ascending=False)
highest_jp_sales
```

```
[186]:
```

	Name	Genre \
4	Pokemon Red/Pokemon Blue	Role-Playing
1	Super Mario Bros.	Platform
41	Animal Crossing: Wild World	Simulation
27	Brain Age 2: More Training in Minutes a Day	Puzzle
19	Brain Age: Train Your Brain in Minutes a Day	Misc
11	Mario Kart DS	Racing
45	Pokemon HeartGold/Pokemon SoulSilver	Action
0	Wii Sports	Sports
136	Street Fighter II: The World Warrior	Fighting
50	Super Mario Land 2: 6 Golden Coins	Adventure
1143	Yu-Gi-Oh! Duel Monsters	Strategy
235	Splatoon	Shooter

	Publisher	Year	JP_Sales
4	Nintendo	1996.0	10.22
1	Nintendo	1985.0	6.81
41	Nintendo	2005.0	5.33
27	Nintendo	2005.0	5.32
19	Nintendo	2005.0	4.16
11	Nintendo	2005.0	4.13
45	Nintendo	2009.0	3.96
0	Nintendo	2006.0	3.77
136	Capcom	1992.0	2.87
50	Nintendo	1992.0	2.69
1143	Konami Digital Entertainment	1998.0	1.61
235	Nintendo	2015.0	1.44

1. Role-Playing is the most popular genre in Japan, with Pokemon Red/Pokemon Blue having the highest sales of all games.

2. Nintendo is also the dominant publisher in Japan, with 10 out of 12 games being highest sold.

3. Animal Crossing: Wild World and Splatoon are the only games that have higher sales in Japan than in other regions, indicating a strong local preference.

```
[187]: highest_other_sales = df.loc[df.groupby('Genre')['Other_Sales'].
      ↪idxmax()][['Name', 'Genre', 'Publisher', 'Year', 'Other_Sales']]
highest_other_sales = highest_other_sales.sort_values(by='Other_Sales',
      ↪ascending=False)
highest_other_sales
```

```
[187]:
```

	Name	Genre \
17	Grand Theft Auto: San Andreas	Action
0	Wii Sports	Sports
47	Gran Turismo 4	Racing
6	New Super Mario Bros.	Platform
7	Wii Play	Misc
10	Nintendogs	Simulation
34	Call of Duty: Black Ops II	Shooter
147	Final Fantasy XII	Role-Playing
676	WWE SmackDown vs Raw 2008	Fighting
27	Brain Age 2: More Training in Minutes a Day	Puzzle
218	Assassin's Creed	Adventure
217	StarCraft II: Wings of Liberty	Strategy

	Publisher	Year	Other_Sales
17	Take-Two Interactive	2004.0	10.57
0	Nintendo	2006.0	8.46
47	Sony Computer Entertainment	2004.0	7.53
6	Nintendo	2006.0	2.90
7	Nintendo	2006.0	2.85
10	Nintendo	2005.0	2.75
34	Activision	2012.0	2.52
147	Square Enix	2006.0	1.74
676	THQ	2007.0	1.41
27	Nintendo	2005.0	1.18
218	Ubisoft	2007.0	0.83
217	Activision	2010.0	0.59

1. Nintendo is also the dominant publisher in other countries, with 5 out of 12 games being highest sold.

2. Action and Sports are the most popular genres in this region, with Grand Theft Auto: San Andreas and Wii Sports having the highest sales of all games.

2.1.18 18. Analysis on Third Highest global Sales Publisher

```
[152]: ## Selecting all 'Activision' from 'Publisher'
Publisher_Activision = df[df['Publisher'] == 'Activision']

columns_to_include = ['Rank', 'Name', 'Platform', 'Year', 'Genre', 'Publisher',
                      'NA_Sales', 'EU_Sales', 'JP_Sales', 'Other_Sales',
                      'Global_Sales']

Activision_games = Publisher_Activision[columns_to_include]

Activision_games
```

```
[152]:
```

	Rank	Name	Platform	\
29	30	Call of Duty: Modern Warfare 3	X360	
31	32	Call of Duty: Black Ops	X360	
33	34	Call of Duty: Black Ops 3	PS4	
34	35	Call of Duty: Black Ops II	PS3	
35	36	Call of Duty: Black Ops II	X360	
...	
16411	16414	Teenage Mutant Ninja Turtles (3DS)	3DS	
16444	16447	Madagascar: Escape 2 Africa	PC	
16501	16504	Transformers: War for Cybertron (XBox 360, PS3...	PC	
16537	16540	Transformers: Fall of Cybertron	PC	
16595	16598	SCORE International Baja 1000: The Official Game	PS2	

	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	\
29	2011.0	Shooter	Activision	9.03	4.28	0.13	1.32	
31	2010.0	Shooter	Activision	9.67	3.73	0.11	1.13	
33	2015.0	Shooter	Activision	5.77	5.81	0.35	2.31	
34	2012.0	Shooter	Activision	4.99	5.88	0.65	2.52	
35	2012.0	Shooter	Activision	8.25	4.30	0.07	1.12	
...	
16411	2013.0	Action	Activision	0.01	0.00	0.00	0.00	
16444	2008.0	Action	Activision	0.01	0.00	0.00	0.00	
16501	2010.0	Shooter	Activision	0.01	0.00	0.00	0.00	
16537	2012.0	Action	Activision	0.01	0.00	0.00	0.00	
16595	2008.0	Racing	Activision	0.00	0.00	0.00	0.00	

	Global_Sales
29	14.76
31	14.64
33	14.24
34	14.03
35	13.73
...	...
16411	0.01

```

16444      0.01
16501      0.01
16537      0.01
16595      0.01

```

[975 rows x 11 columns]

IN Global Market

```

[154]: ## Extract the games with highest Global_Sales
highest_global_sales = Activision_games.groupby('Genre')['Global_Sales'].
    ↪idxmax()

highest_global_sales = Activision_games.loc[highest_global_sales][['Name',
    ↪'Genre', 'Global_Sales']]

highest_global_sales = highest_global_sales.sort_values(by='Global_Sales',
    ↪ascending=False)

highest_global_sales

```

```

[154]:

```

	Name	Genre	Global_Sales
29	Call of Duty: Modern Warfare 3	Shooter	14.76
137	World of Warcraft	Role-Playing	6.28
202	Tony Hawk's Pro Skater	Sports	5.02
207	Guitar Hero III: Legends of Rock	Misc	4.98
217	StarCraft II: Wings of Liberty	Strategy	4.83
239	Pitfall!	Platform	4.50
241	Spider-Man: The Movie	Action	4.48
1002	Star Wars: Starfighter	Simulation	1.76
1622	Vigilante 8	Racing	1.23
2669	Boxing	Fighting	0.77
2809	iCarly	Adventure	0.73
4337	Ghostbusters	Puzzle	0.45

```

[155]: plt.figure(figsize=(12, 6))

## Plot
plt.bar(highest_global_sales['Genre'], highest_global_sales['Global_Sales'],
    ↪color='skyblue')

## Labels
plt.xlabel('Genre')
plt.ylabel('Global Sales')

## Axis
plt.title('Highest Global Sales of Activision Publisher', size = 20)

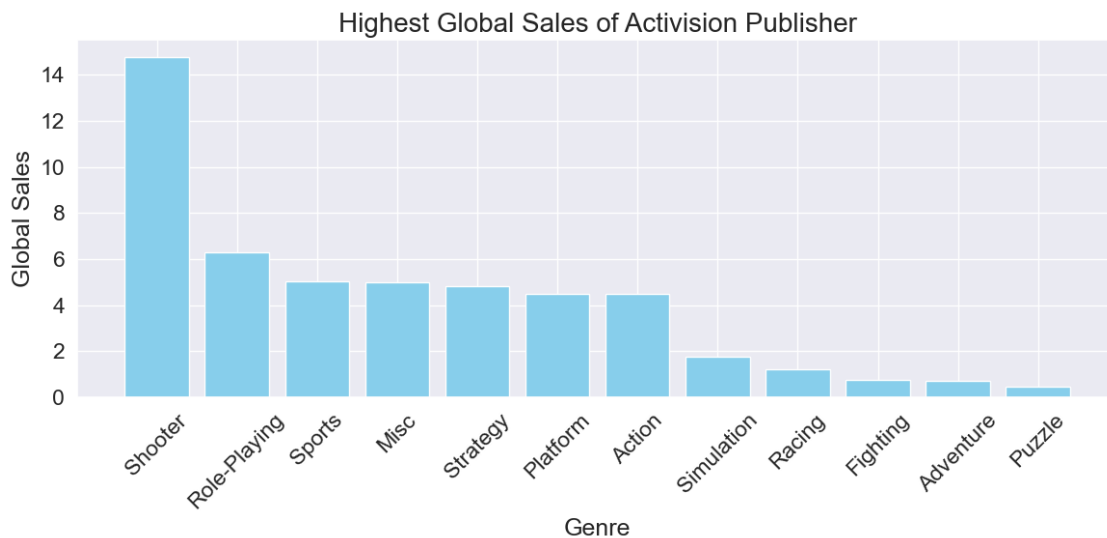
```

```

## Rotation of axis
plt.xticks(rotation=45)

plt.tight_layout()
##Display of graph
plt.show()

```



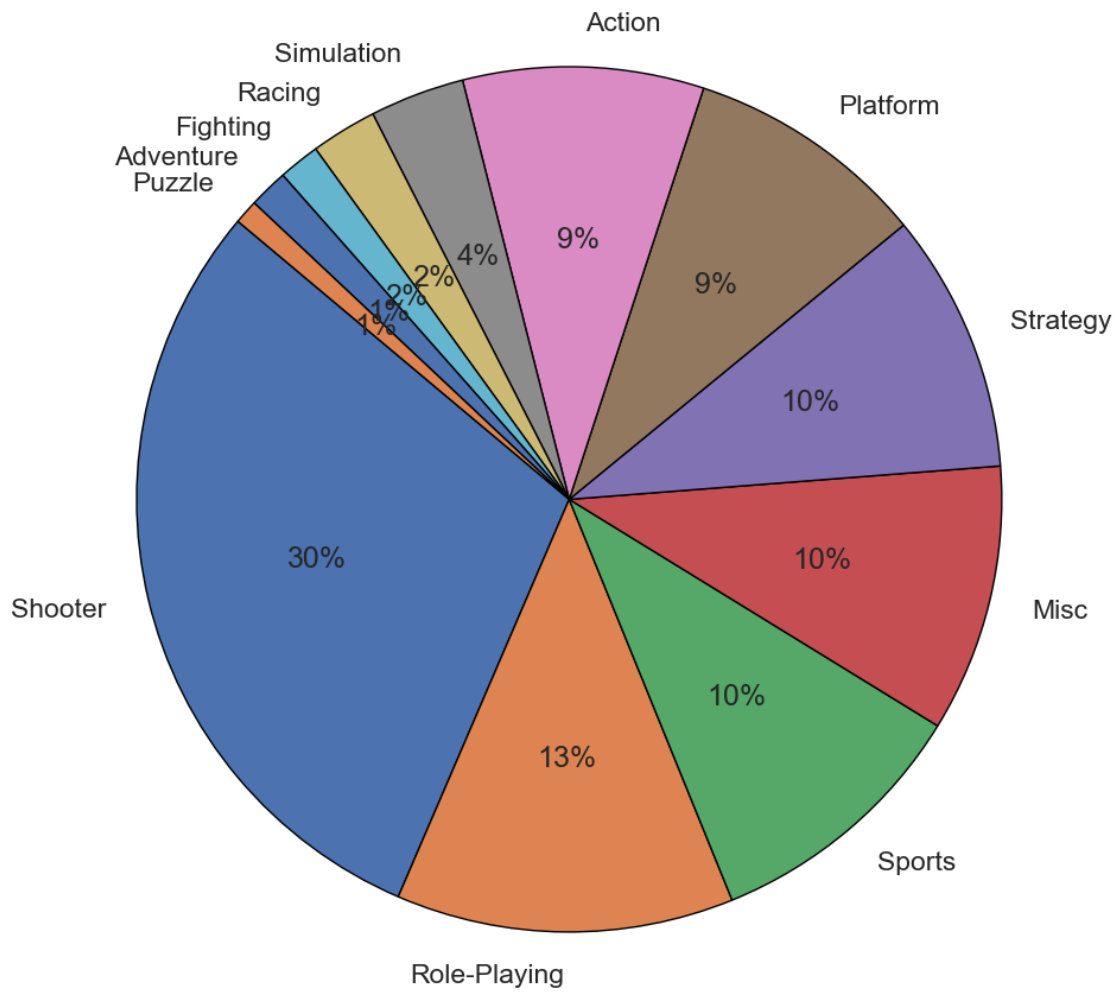
```

[156]: plt.figure(figsize=(10,10))
## plot
plt.pie(highest_global_sales['Global_Sales'],
        ↪labels=highest_global_sales['Genre'],wedgeprops={"edgecolor": "black"},
        ↪autopct='%1.0f%%', startangle=140)
## Title
plt.title('Distribution of Global Sales by Genre of Activision Publisher')

plt.tight_layout()
## Display of Plot
plt.show()

```

Distribution of Global Sales by Genre of Activision Publisher



In North America Market

```
[157]: ## Extract the games with highest NA_Sales
highest_na_sales = Activision_games.groupby('Genre')['NA_Sales'].idxmax()

highest_na_sales = Activision_games.loc[highest_na_sales][['Name', 'Genre', 'NA_Sales']]

highest_na_sales = highest_na_sales.sort_values(by='NA_Sales', ascending=False)

highest_na_sales
```

```
[157]:
```

	Name	Genre	NA_Sales
31	Call of Duty: Black Ops	Shooter	9.67
239	Pitfall!	Platform	4.21
207	Guitar Hero III: Legends of Rock	Misc	3.49
202	Tony Hawk's Pro Skater	Sports	3.42
241	Spider-Man: The Movie	Action	2.71
288	World of Warcraft: The Burning Crusade	Role-Playing	2.57
217	StarCraft II: Wings of Liberty	Strategy	2.56
2669	Boxing	Fighting	0.72
1622	Vigilante 8	Racing	0.68
2008	ZhuZhu Pets	Simulation	0.68
2809	iCarly	Adventure	0.67
4337	Ghostbusters	Puzzle	0.42

```
[158]: ## Plot
plt.figure(figsize=(12, 6))

plt.bar(highest_na_sales['Genre'], highest_na_sales['NA_Sales'],
        color='skyblue')

## Labels
plt.xlabel('Genre')

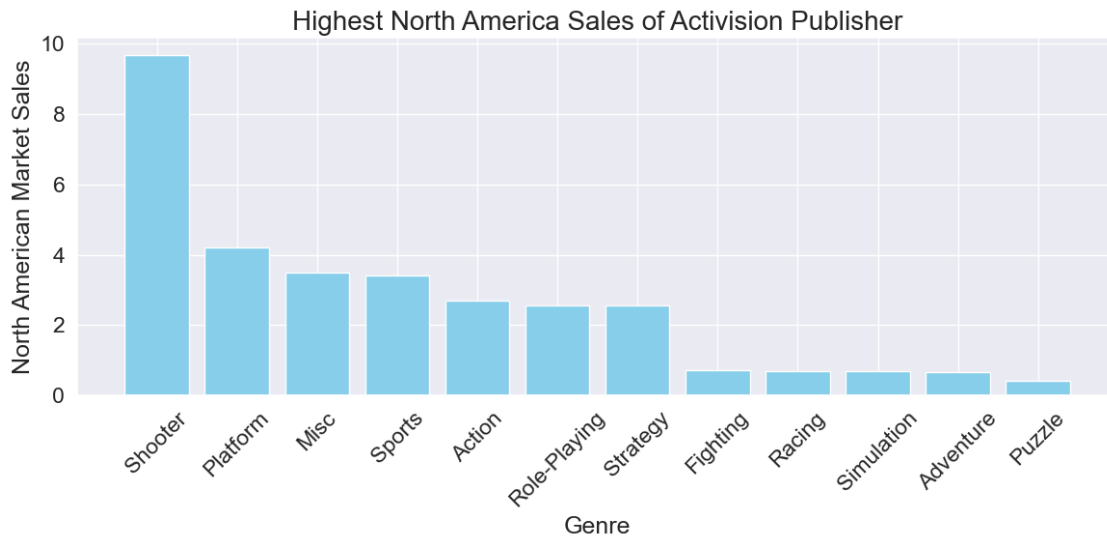
plt.ylabel('North American Market Sales')

## Title
plt.title('Highest North America Sales of Activision Publisher', size = 20)

## Rotating the axis
plt.xticks(rotation=45)

plt.tight_layout()

## Display of plot
plt.show()
```



```
[159]: ## Plot
plt.figure(figsize=(10,10))

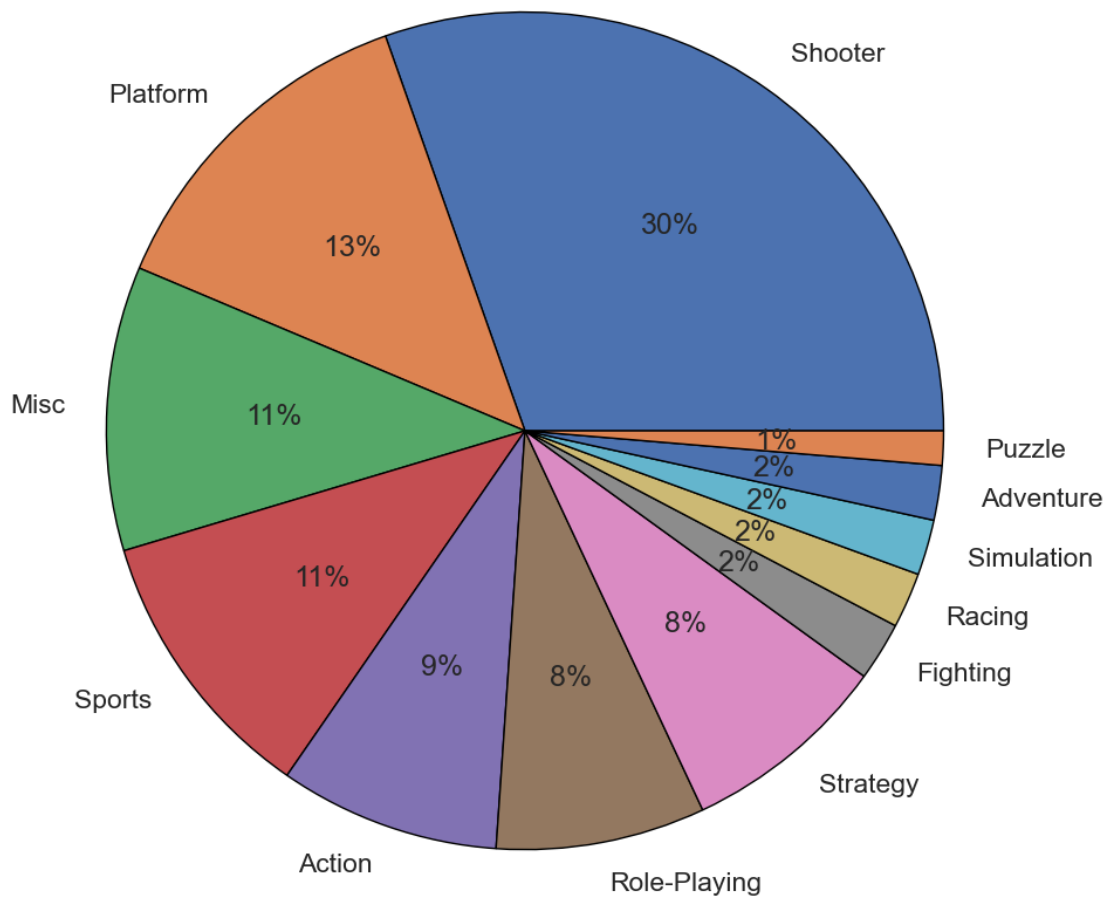
plt.pie(highest_na_sales['NA_Sales'],
        ↪labels=highest_na_sales['Genre'],wedgeprops={"edgecolor": "black"},
        ↪autopct='%1.0f%%')

## Title
plt.title('Distribution of North America Sales by Genre of Activision
        ↪Publisher')

plt.tight_layout()

## Display of Graph
plt.show()
```

Distribution of North America Sales by Genre of Activision Publisher



in European Market

```
[161]: ## Extract the games with highest EU_Sales
highest_eu_sales = Activision_games.groupby('Genre')['EU_Sales'].idxmax()

highest_eu_sales = Activision_games.loc[highest_eu_sales][['Name', 'Genre', '
↪ EU_Sales']]

highest_eu_sales = highest_eu_sales.sort_values(by='EU_Sales', ascending=False)

highest_eu_sales
```



```
[161]:
```

	Name	Genre	EU_Sales
137	World of Warcraft	Role-Playing	6.21
34	Call of Duty: Black Ops II	Shooter	5.88
267	Warcraft II: Tides of Darkness	Strategy	2.27
241	Spider-Man: The Movie	Action	1.51
225	Tony Hawk's Pro Skater 2	Sports	1.41
230	Guitar Hero III: Legends of Rock	Misc	1.12
1002	Star Wars: Starfighter	Simulation	0.87
675	Skylanders SWAP Force	Platform	0.83
1622	Vigilante 8	Racing	0.47
2840	X-Men: Mutant Academy 2	Fighting	0.27
3475	Shrek 2: Beg for Mercy	Adventure	0.15
6663	Bomberman Tournament	Puzzle	0.04

```
[162]: ## plot
plt.figure(figsize=(12, 6))

plt.bar(highest_eu_sales['Genre'], highest_eu_sales['EU_Sales'],
        color='skyblue')

## Labels
plt.xlabel('Genre')
plt.ylabel('European Market Sales')

##Title
plt.title('Highest Europe Sales of Activision Publisher', size = 20)
## Rotation of axis
plt.xticks(rotation=45)

plt.tight_layout()
## Display of Graph
plt.show()
```



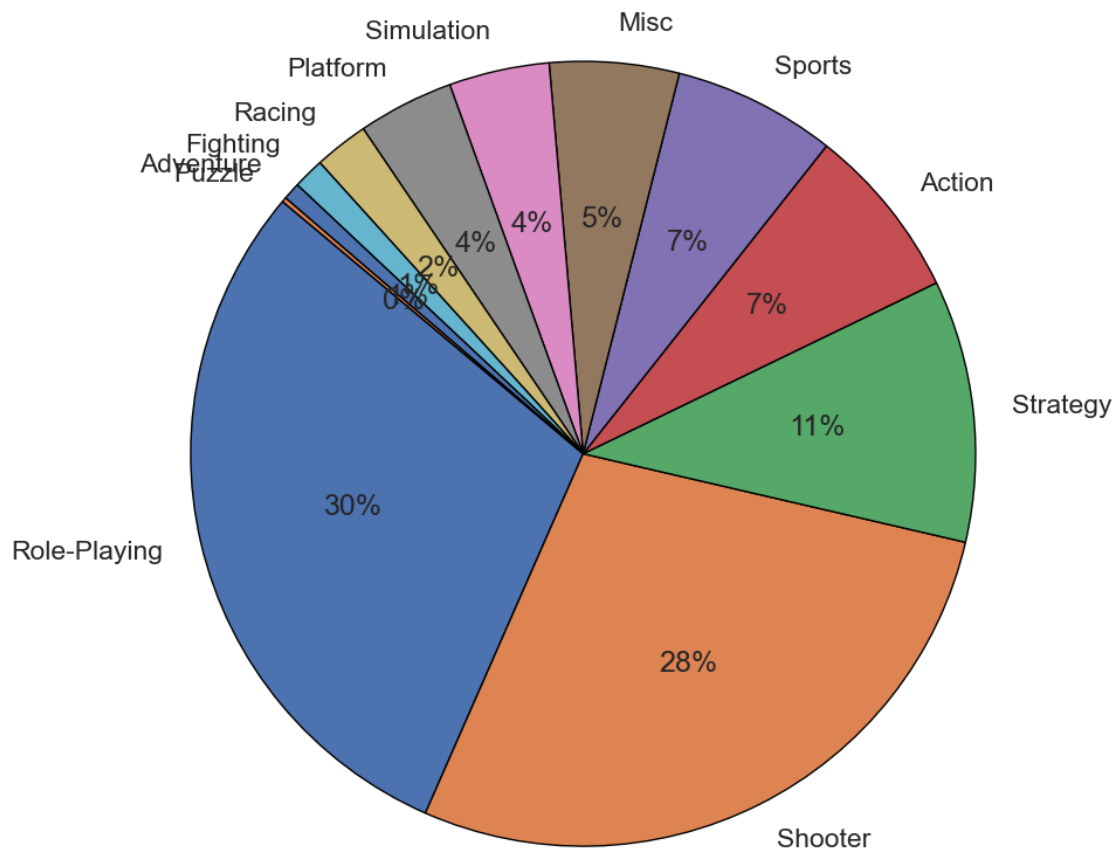
```
[163]: ## Plot
plt.figure(figsize=(10,10))

plt.pie(highest_eu_sales['EU_Sales'],
        ↪labels=highest_eu_sales['Genre'],wedgeprops={"edgecolor": "black"},
        ↪autopct='%1.0f%%', startangle=140)

## Title
plt.title('Distribution of Europe Sales by Genre of Activision Publisher')

plt.tight_layout()
## Display of plot
plt.show()
```

Distribution of Europe Sales by Genre of Activision Publisher



in Japan Market

```
[164]: ## Extract the games with highest JP_Sales
highest_jp_sales = Activision_games.groupby('Genre')['JP_Sales'].idxmax()

highest_jp_sales = Activision_games.loc[highest_jp_sales][['Name', 'Genre', 'JP_Sales']]

highest_jp_sales = highest_jp_sales.sort_values(by='JP_Sales', ascending=False)

highest_jp_sales
```

```
[164]:
```

	Name	Genre	JP_Sales
34	Call of Duty: Black Ops II	Shooter	0.65
851	Tenchu: Stealth Assassins	Action	0.27

826	Diablo III	Role-Playing	0.15
10538	Blue Stinger	Adventure	0.10
6663	Bomberman Tournament	Puzzle	0.10
925	Shrek 2	Platform	0.03
710	Guitar Hero III: Legends of Rock	Misc	0.02
2775	Blur	Racing	0.02
1002	Star Wars: Starfighter	Simulation	0.02
202	Tony Hawk's Pro Skater	Sports	0.02
2669	Boxing	Fighting	0.00
217	StarCraft II: Wings of Liberty	Strategy	0.00

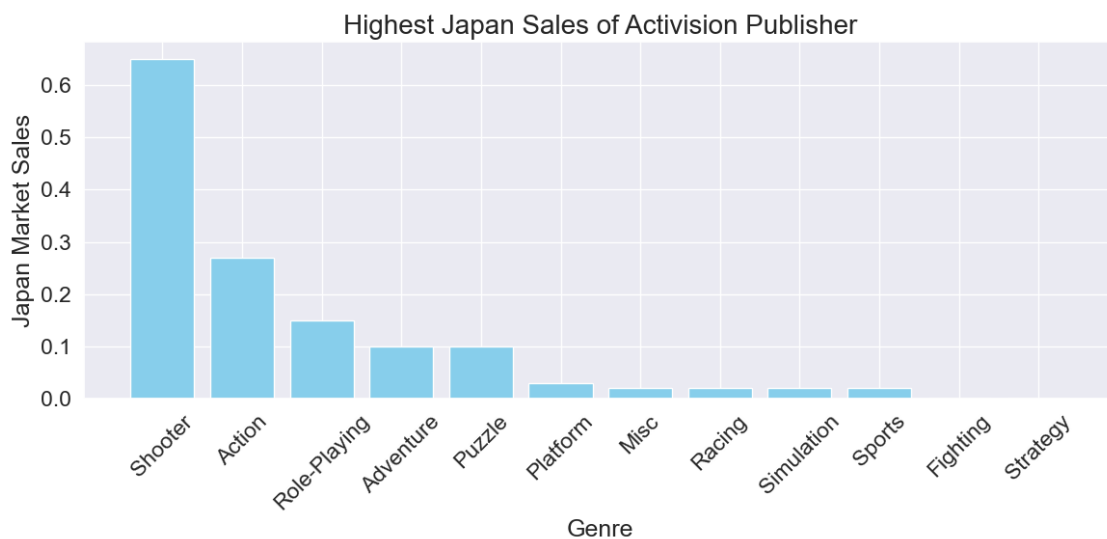
```
[165]: plt.figure(figsize=(12, 6))
## Plot
plt.bar(highest_jp_sales['Genre'], highest_jp_sales['JP_Sales'],
        color='skyblue')

## Labels
plt.xlabel('Genre')
plt.ylabel('Japan Market Sales')
## Title
plt.title('Highest Japan Sales of Activision Publisher', size = 20)

## Rotation of axis
plt.xticks(rotation=45)

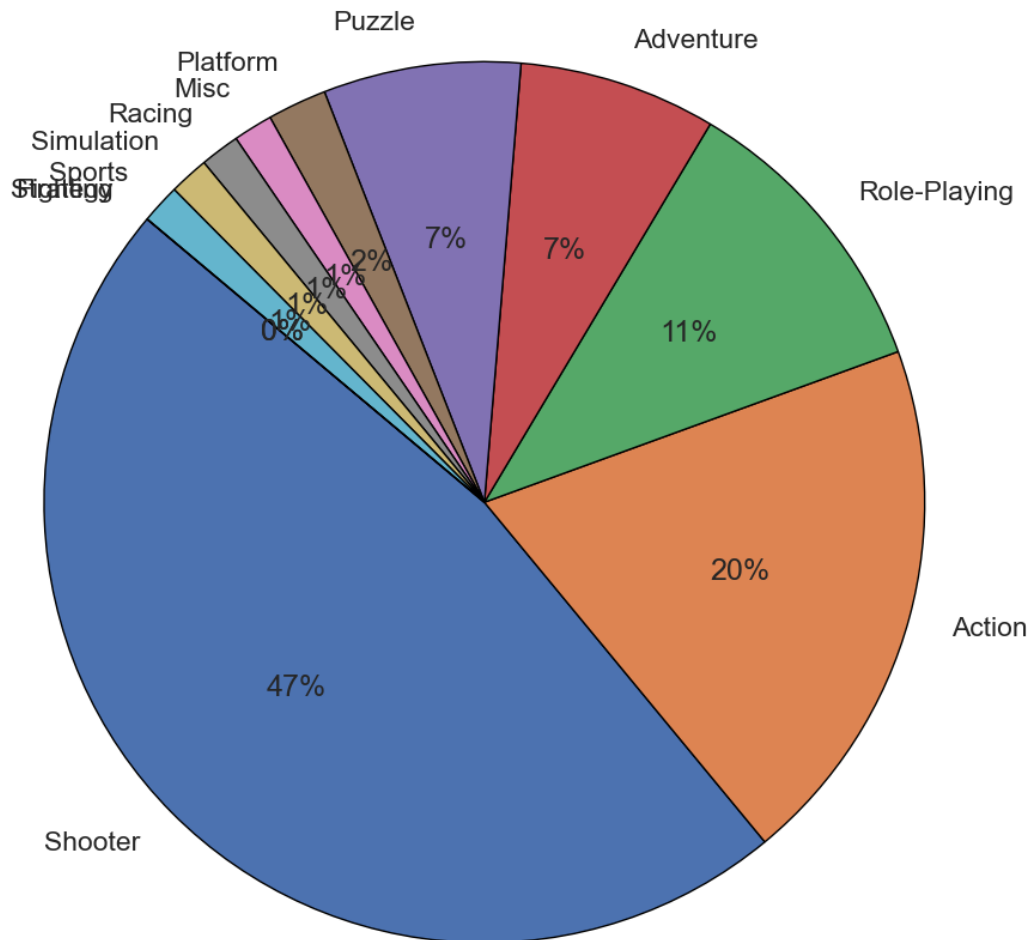
plt.tight_layout()

## Display of plot
plt.show()
```



```
[166]: plt.figure(figsize=(10,10))
## Plot
plt.pie(highest_jp_sales['JP_Sales'],
        labels=highest_jp_sales['Genre'], wedgeprops={"edgecolor": "black"},
        autopct='%1.0f%%', startangle=140)
##Title
plt.title('Distribution of Japan Sales by Genre of Activision Publisher')

plt.tight_layout()
##Display of plot
plt.show()
```



2.1.19 19. Top 10 Games by Global Sales

```
[188]: # Top 10 games by global sales
top_games = df.sort_values(by='Global_Sales', ascending=False).head(10)

# Print top 10 games
top_games[['Name', 'Platform', 'Year', 'Genre', 'Publisher', 'Global_Sales']]
```

```
[188]:
```

	Name	Platform	Year	Genre	Publisher	\
0	Wii Sports	Wii	2006.0	Sports	Nintendo	
1	Super Mario Bros.	NES	1985.0	Platform	Nintendo	
2	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	
3	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	
4	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo	
5	Tetris	GB	1989.0	Puzzle	Nintendo	
6	New Super Mario Bros.	DS	2006.0	Platform	Nintendo	
7	Wii Play	Wii	2006.0	Misc	Nintendo	
8	New Super Mario Bros. Wii	Wii	2009.0	Platform	Nintendo	
9	Duck Hunt	NES	1984.0	Shooter	Nintendo	

	Global_Sales
0	82.74
1	40.24
2	35.82
3	33.00
4	31.37
5	30.26
6	30.01
7	29.02
8	28.62
9	28.31

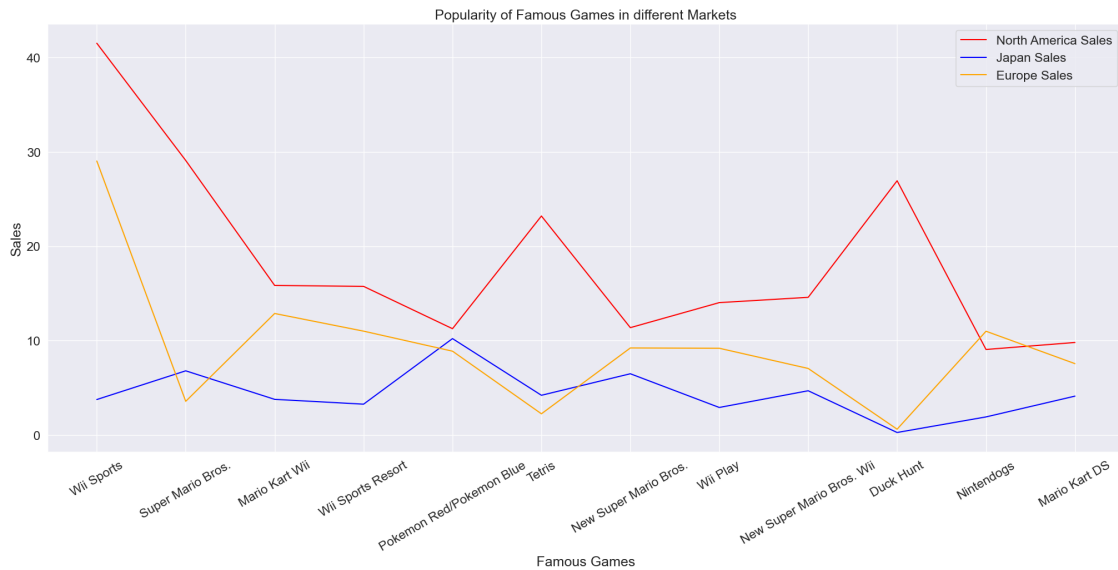
Nintendo is the most successful publisher in terms of global sales, as it has published all of the top 10 games in the table, with a total of 369.37 million units sold globally.

The most popular genres are Sports and Platform, with Wii Sports and Super Mario Bros. being the best-selling games in each genre. The most popular platforms are Wii and NES, with four games each in the top 10.

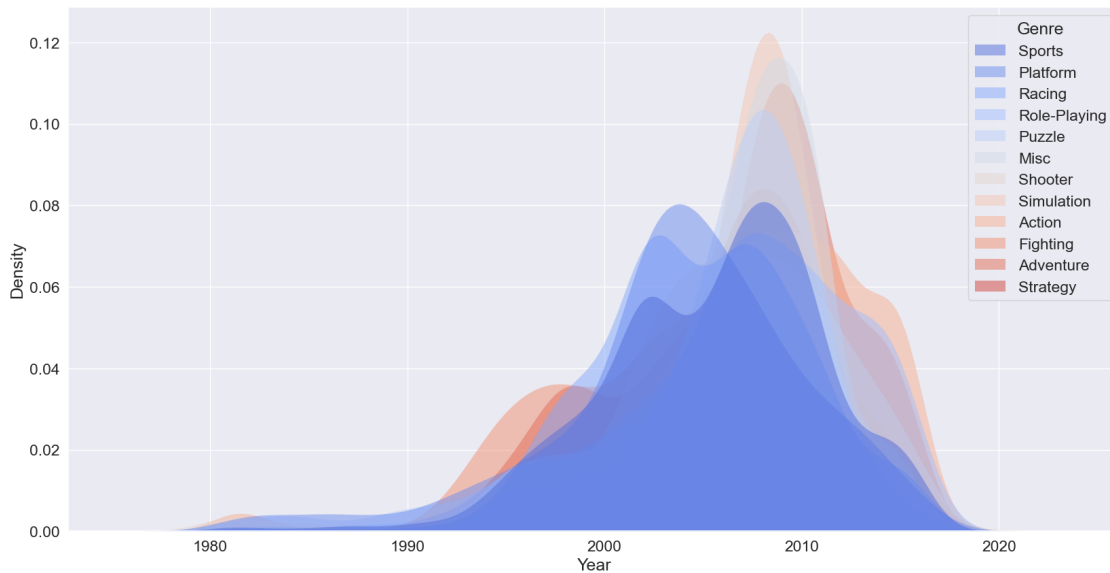
2.1.20 20 . Most Popular Games Trend

```
[198]: fig = plt.figure(figsize=(25,10))
plt.plot(df['Name'].head(12), df['NA_Sales'].head(12), color='red',
        label='North America Sales')
plt.plot(df['Name'].head(12), df['JP_Sales'].head(12), color='blue',
        label='Japan Sales')
```

```
plt.plot(df['Name'].head(12), df['EU_Sales'].head(12), color='orange',
        label='Europe Sales')
plt.legend()
plt.xlabel('Famous Games')
plt.xticks(rotation=30)
plt.ylabel('Sales')
plt.title('Popularity of Famous Games in different Markets')
plt.show()
```



```
[199]: plt.figure(figsize=(20,10))
sns.kdeplot( data=df, x='Year',hue="Genre", fill=True,
            common_norm=False,palette='coolwarm' , alpha=.5, linewidth=0,legend=True)
plt.show()
```



from the 1980s to 1990s, sports, racing, action, and shooter were very popular.

from 1990s to 2000s, all the genres kept increasing in addition to adventure and platform.

from 2000s to 2010s, all the genres kept increasing in addition to puzzle, action, simulation, fighting, misc were all at their peaks.

from 2010s to 2020, every genre is declining but action and role playing were the ones dominating while declining.

[]: