#### **Data Visualization**

```
In [1]: import numpy as np
    import pandas as pd
    import plotly.express as px
In [2]: import pandas as pd
    import matplotlib.pyplot as plt
    import seaborn as sns
```

#### **Load the Dataset**

```
In [3]: # Load the dataset
df = pd.read_csv(r"C:\Users\samee\Downloads\games_dataset.csv")
# Display the first few rows of the dataset
print(df.head())
```

```
Game Name
                                      Genre
                                                    Platform
                                                              Release Year \
0
      Sekiro: Shadows Die Twice
                                     Sports Nintendo Switch
                                                                      2014
1
                        Control
                                     Puzzle Nintendo Switch
                                                                      2000
2
                                   Strategy Nintendo Switch
                                                                      2007
                       Among Us
  Call of Duty: Modern Warfare
                                    Stealth
                                                 PlayStation
                                                                      2005
4
             League of Legends Simulation
                                                          PC
                                                                      2021
```

```
User Rating

0 9.654719

1 3.457386

2 5.367622

3 1.240325

4 3.535476
```

```
In [4]: # Get basic information about the dataset
print(df.info())

# Check for missing values
print(df.isnull().sum())

<class 'pandas.core.frame.DataFrame'>
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5000 entries, 0 to 4999
Data columns (total 5 columns):

	0010000						
#	Column	Non-Null Count	Dtype				
0	Game Name	5000 non-null	object				
1	Genre	5000 non-null	object				
2	Platform	5000 non-null	object				
3	Release Year	5000 non-null	int64				
4	User Rating	5000 non-null	float64				
<pre>dtypes: float64(1), int64(1), object(3)</pre>							
memory usage: 195.4+ KB							
None							
Game	Name 6	)					
Genre	<u> </u>	)					
Platf	orm (	)					
Relea	se Year (	)					
User	Rating 0	)					

dtype: int64

# **Descriptive Statistics**

```
In [6]: # Descriptive statistics for numerical columns
    print(df.describe())

# Descriptive statistics for categorical columns
    print(df['Genre'].value_counts())
    print(df['Platform'].value_counts())
```

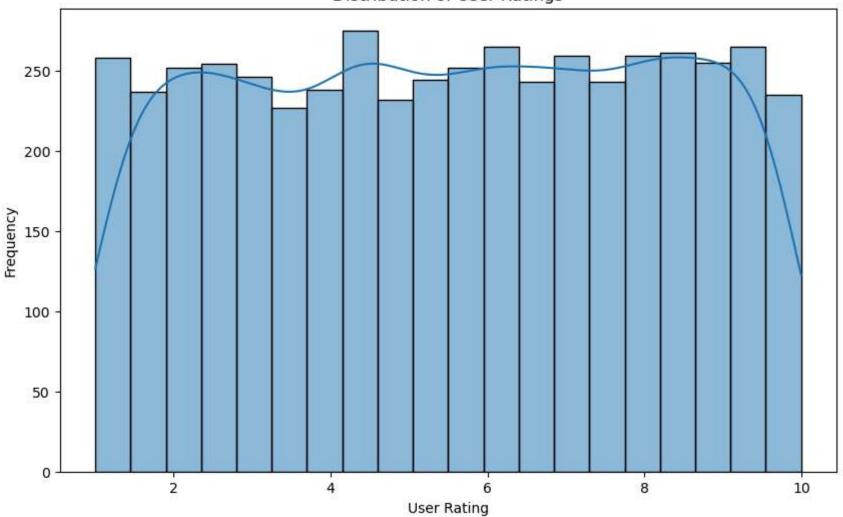
	Releas	se	Year	User	Rating		
count	5000	.00	90000	5000	.000000		
mean	2011	. 37	75400	5	.524913		
std	6	. 94	43984	2	.598429		
min	2000	.00	90000	1	.005878		
25%	2005	.00	90000	3	.260596		
50%	2011	.00	90000	5	.574233		
75%	2017	.00	90000	7	.802727		
max	2023	.00	90000	9	.996196		
Genre							
Sports			403				
Puzzle			384				
Racing			369				
Simulation 368							
Stealth 360							
Role-playing 360							
Horror 356							
Strate							
Survival 3							
Adventure 34							
Action			340				
Platfor	rmer		338				
Shooter	•		337				
Fightir	ng		337				
Name: d	count,	ď	type:	int64			
Platfor	rm						
PC			,	1056			
Xbox				1009			
Nintend	do Swit	989					
Mobile		986					
PlaySta	ation	960					
Name: d	count,	ď	type:	int64			

#### **Data Visualization**

```
In [9]: # Plot the distribution of user ratings
plt.figure(figsize=(10, 6))
sns.histplot(df['User Rating'], bins=20, kde=True)
plt.title('Distribution of User Ratings')
plt.xlabel('User Rating')
plt.ylabel('Frequency')
plt.show()
```

C:\Users\samee\anaconda3\Lib\site-packages\seaborn\\_oldcore.py:1119: FutureWarning: use\_inf\_as\_na option is de precated and will be removed in a future version. Convert inf values to NaN before operating instead. with pd.option\_context('mode.use\_inf\_as\_na', True):

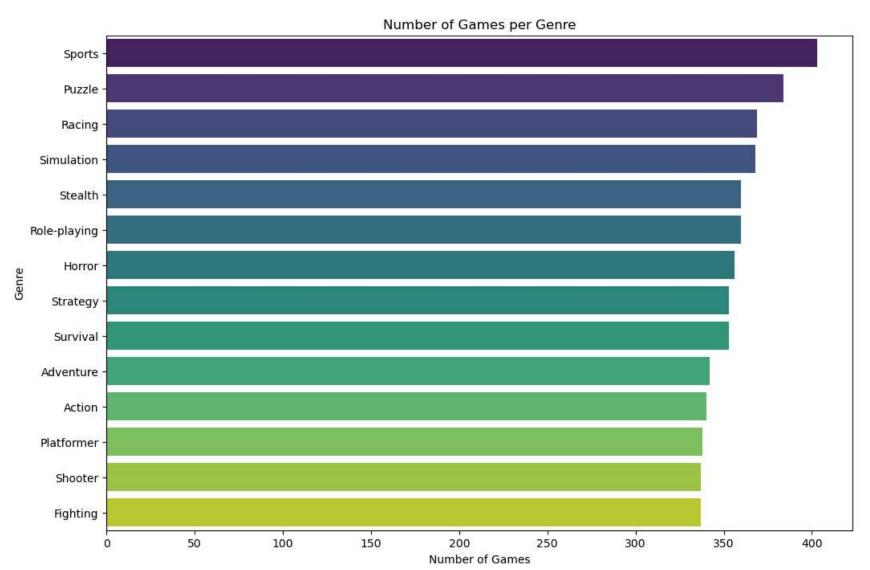
## Distribution of User Ratings



The graph above illustrates the distribution of User Rating.

# **Popular Genres**

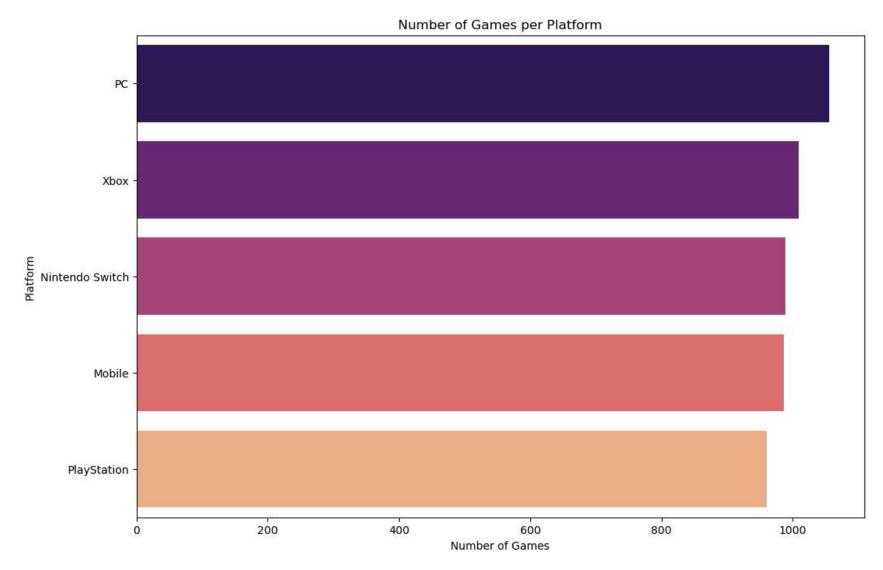
```
In [10]: # Plot the count of games in each genre
    plt.figure(figsize=(12, 8))
    sns.countplot(y='Genre', data=df, order=df['Genre'].value_counts().index, palette="viridis")
    plt.title('Number of Games per Genre')
    plt.xlabel('Number of Games')
    plt.ylabel('Genre')
    plt.show()
```



The graph above shows Sports and Puzzle games are the most common genres, while Fighting and Shooter games are less common.

# **Games by Platform**

```
In [11]: # Plot the count of games on each platform
plt.figure(figsize=(12, 8))
sns.countplot(y='Platform', data=df, order=df['Platform'].value_counts().index, palette='magma')
plt.title('Number of Games per Platform')
plt.xlabel('Number of Games')
plt.ylabel('Platform')
plt.show()
```



The graph above shows PC has the highest count of games, followed by Xbox and Nintendo Switch.

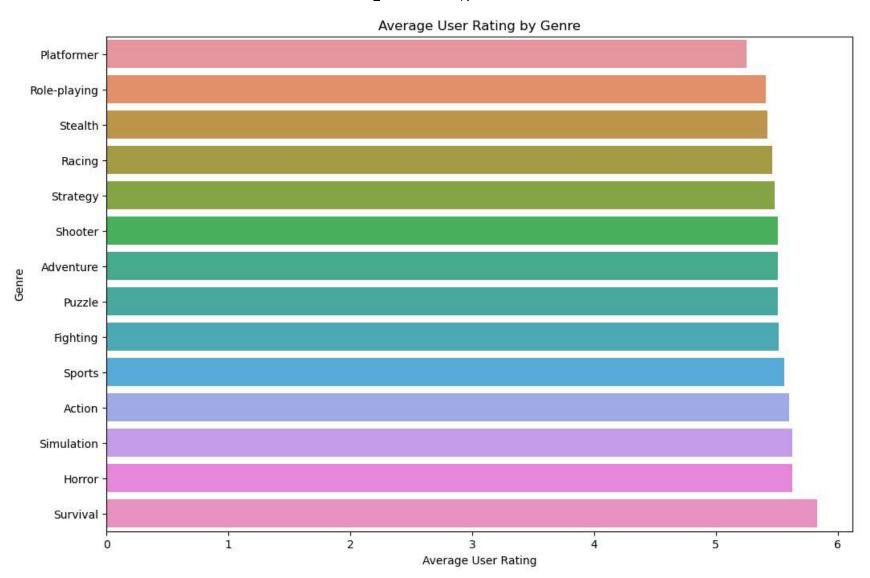
## **Average User Rating by Genre**

```
In [12]: # Plot average user rating by genre
    plt.figure(figsize=(12, 8))
    sns.barplot(x='User Rating', y='Genre', data=df, ci=None, estimator=pd.Series.mean, order=df.groupby('Genre')['Uplt.title('Average User Rating by Genre')
    plt.xlabel('Average User Rating')
    plt.ylabel('Genre')
    plt.show()

C:\Users\samee\AppData\Local\Temp\ipykernel_111296\1186505564.py:3: FutureWarning:

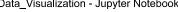
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

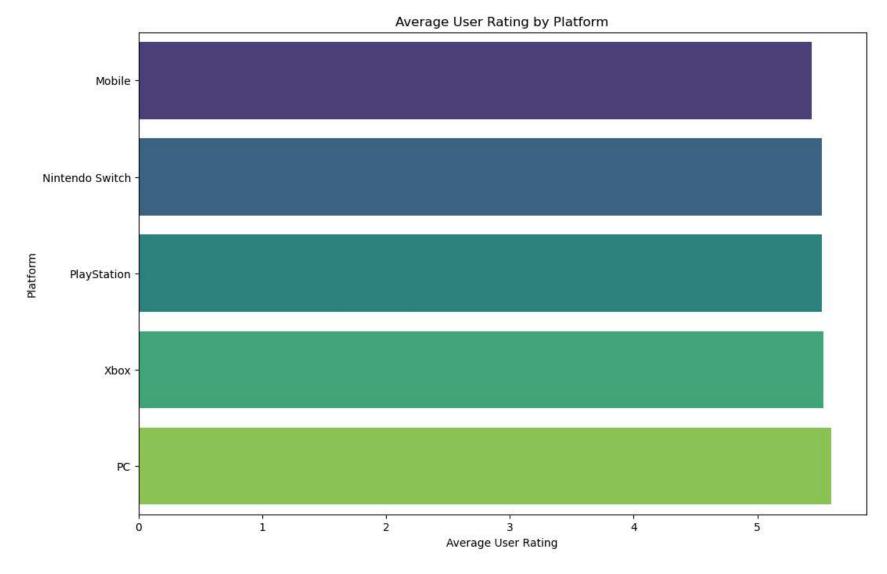
    sns.barplot(x='User Rating', y='Genre', data=df, ci=None, estimator=pd.Series.mean, order=df.groupby('Genre')['User Rating'].mean().sort_values().index)
```



The graph above illustrates the average user ratings across different gaming genres. Survival games received the highest average ratings, followed by Horror and Simulation.

## **Average User Rating by Platform**





The graph above illustrates the average user ratings across different gaming platforms. PC platforms received the highest average ratings, followed by Xbox, PlayStation, Nintendo Switch, and Mobile. This suggests that users may have a stronger preference for gaming experiences on PC platforms, while the other platforms have relatively similar ratings, indicating a more evenly distributed satisfaction among users across these platforms.

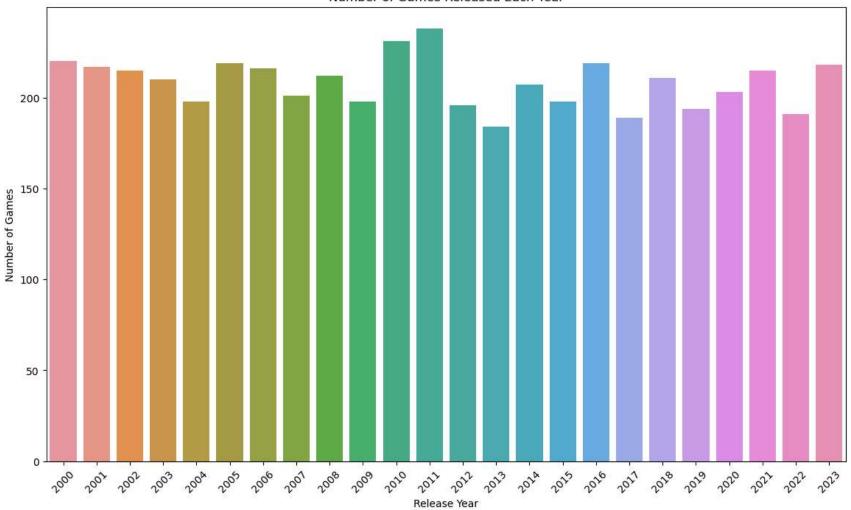
## **Top-Rated games**

```
In [6]: # Find the top-rated games
        top_rated_games = df.sort_values(by='User Rating', ascending=False).head(10)
        print(top_rated_games[['Game Name', 'User Rating']])
                             Game Name User Rating
                              NBA 2K21
                                           9.996196
        2627
        2666
                             Minecraft
                                           9.994527
                               Control
        3213
                                           9.993194
                              Valorant
                                           9.988694
        664
        3777
                            Elden Ring
                                           9.988145
        1469
                                           9.987679
                             Overwatch
        3647
                                           9.987600
                           PUBG Mobile
              The Witcher 3: Wild Hunt
                                           9.987243
        871
        1799
                     World of Warcraft
                                           9.987004
        1847
                   Super Mario Odyssey
                                           9.980222
```

# **Release Year Analysis**

```
In [72]: # Plot the number of games released each year
plt.figure(figsize=(14, 8))
sns.countplot(x='Release Year', data=df, order=df['Release Year'].value_counts().index.sort_values())
plt.title('Number of Games Released Each Year')
plt.xlabel('Release Year')
plt.ylabel('Number of Games')
plt.xticks(rotation=45)
plt.show()
```

#### Number of Games Released Each Year



The graph above shows a relatively steady number of game releases each year from 2000 to 2023, with a peak around 2011 and a slight decline in releases during the years following. This steady trend suggests consistent production within the gaming industry, with occasional increases or decreases.

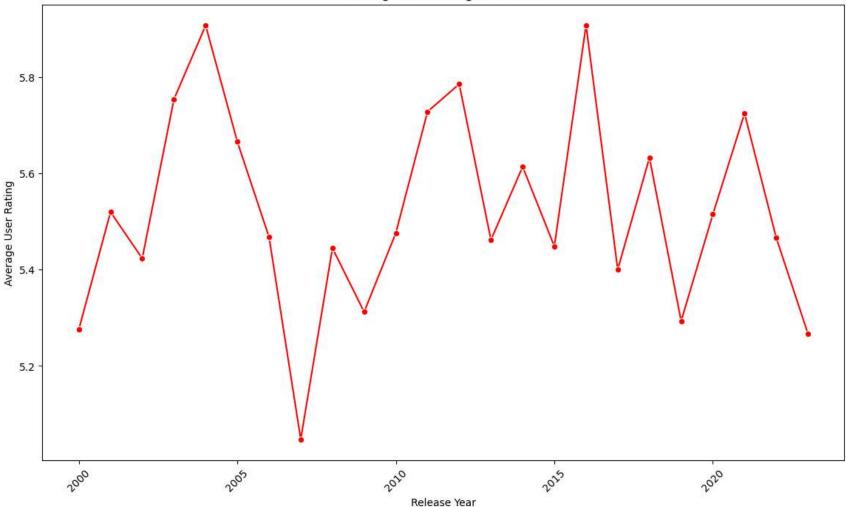
### Trend Analysis: Analyze trends over time, such as Average Ratings per Year.

```
In [59]: # Calculate average user rating per release year
         average rating per year = df.groupby('Release Year')['User Rating'].mean()
         # Plot the trend of average user rating over the years
         plt.figure(figsize=(14, 8))
         sns.lineplot(x=average rating per year.index, y=average rating per year.values, marker='o', color='r')
         plt.title('Average User Rating Over Time')
         plt.xlabel('Release Year')
         plt.ylabel('Average User Rating')
         plt.xticks(rotation=45)
         plt.show()
         C:\Users\samee\anaconda3\Lib\site-packages\seaborn\ oldcore.py:1119: FutureWarning: use inf as na option is de
         precated and will be removed in a future version. Convert inf values to NaN before operating instead.
```

with pd.option\_context('mode.use\_inf\_as\_na', True):

C:\Users\samee\anaconda3\Lib\site-packages\seaborn\ oldcore.py:1119: FutureWarning: use inf as na option is de precated and will be removed in a future version. Convert inf values to NaN before operating instead. with pd.option\_context('mode.use\_inf\_as\_na', True):

#### Average User Rating Over Time



The graph above shows that average user ratings have fluctuated significantly between 2000 and 2020, with peaks around 2004 and 2015, and dips around 2007 and 2018. This indicates varying user satisfaction over the years, influenced by factors like changes in game quality, technology, and player expectations.

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Go to Findings and Conclusion (./Findings\_and\_Conclusion.ipynb)

In [ ]: