C550-T301-Data-Mining_2241-1_week1_Samanta_Rajib

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0.1 Class : C550-T301 Data Mining (2241-1)

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0.2.1 Assignment : Week 1

Download the Video Game Sales with Ratings dataset from this link: Video Game Sales with Ratings. https://www.kaggle.com/code/jayagopal20/video-game-stats

1. Load the dataset as a Pandas data frame.
2. Display the first ten rows of data.
3. Find the dimensions (number of rows and columns) in the data frame. What do these two numbers represent in the context of the data?
4. Find the top five games by critic score.
5. Find the number of video games in the data frame in each genre.
6. Find the first five games in the data frame on the SNES platform.
7. Find the five publishers with the highest total global sales. Note: You will need to calculate the total global sales for each publisher to do this.
8. Create a new column in the data frame that calculates the percentage of global sales from North America. Display the first five rows of the new data frame.
9. Find the number NaN entries (missing data values) in each column.
10. Try to calculate the median user score of all the video games. You will likely run into an error because some of the user score entries are a non-numerical string that cannot be converted to a float. Find and replace this string with NaN and then calculate the median. Then, replace all NaN entries in the user score column with the median value.

[23]:
```
# Load the Libraries
import os
import pandas as pd
```

[24]:
```
# 1. Load the dataset as a Pandas data frame.
# 2. Display the first ten rows of data.
# Read in the Video Game Sales with Ratings data file␣
  ↪('Video_Games_Sales_as_at_22_Dec_2016.csv') from local:
directory = '/Users/rajibsamanta/Documents/Rajib/College/Sem6_fall_2023/week1'
# Set the working directory
os.chdir(directory)
print(os.getcwd())
dataset1_csv = pd.read_csv("Video_Games_Sales_as_at_22_Dec_2016.csv")
dataset1_csv.head(10)
```

```
# Display the DataFrame 10 rows
```

```
/Users/rajibsamanta/Documents/Rajib/College/Sem6_fall_2023/week1
```

[24]:

	Name	Platform	Year_of_Release	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	Critic_Score	Critic_Count	User_Score	User_Count	Developer	Rating
0	Wii Sports	Wii	2006.0	Sports	Nintendo	41.36	28.96	3.77	8.45	82.53	76.0	51.0	8	322.0	Nintendo	E
1	Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24	NaN	NaN	NaN	NaN	NaN	NaN
2	Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.68	12.76	3.79	3.29	35.52	82.0	73.0	8.3	709.0	Nintendo	E
3	Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.61	10.93	3.28	2.95	32.77	80.0	73.0	8	192.0	Nintendo	E
4	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo	11.27	8.89	10.22	1.00	31.37	NaN	NaN	NaN	NaN	NaN	NaN
5	Tetris	GB	1989.0	Puzzle	Nintendo	23.20	2.26	4.22	0.58	30.26	NaN	NaN	NaN	NaN	NaN	NaN
6	New Super Mario Bros.	DS	2006.0	Platform	Nintendo	11.28	9.14	6.50	2.88	29.80	89.0	65.0	8.5	431.0	Nintendo	E
7	Wii Play	Wii	2006.0	Misc	Nintendo	13.96	9.18	2.93	2.84	28.92	58.0	41.0	6.6	129.0	Nintendo	E
8	New Super Mario Bros. Wii	Wii	2009.0	Platform	Nintendo	14.44	6.94	4.70	2.24	28.32	87.0	80.0	8.4	594.0	Nintendo	E
9	Duck Hunt	NES	1984.0	Shooter	Nintendo	26.93	0.63	0.28	0.47	28.31	NaN	NaN	NaN	NaN	NaN	NaN

[25]:
```
# describe the data set
dataset1_csv.describe()
```

[25]:

	Year_of_Release	NA_Sales	EU_Sales	JP_Sales
count	16450.000000	16719.000000	16719.000000	16719.000000
mean	2006.487356	0.263330	0.145025	0.077602
std	5.878995	0.813514	0.503283	0.308818
min	1980.000000	0.000000	0.000000	0.000000

25%	2003.000000	0.000000	0.000000	0.000000
50%	2007.000000	0.080000	0.020000	0.000000
75%	2010.000000	0.240000	0.110000	0.040000
max	2020.000000	41.360000	28.960000	10.220000

	Other_Sales	Global_Sales	Critic_Score	Critic_Count	User_Count
count	16719.000000	16719.000000	8137.000000	8137.000000	7590.000000
mean	0.047332	0.533543	68.967679	26.360821	162.229908
std	0.186710	1.547935	13.938165	18.980495	561.282326
min	0.000000	0.010000	13.000000	3.000000	4.000000
25%	0.000000	0.060000	60.000000	12.000000	10.000000
50%	0.010000	0.170000	71.000000	21.000000	24.000000
75%	0.030000	0.470000	79.000000	36.000000	81.000000
max	10.570000	82.530000	98.000000	113.000000	10665.000000

[26]:
```
# 3. Find the dimensions (number of rows and columns) in the data frame.
#    What do these two numbers represent in the context of the data?

# DataFrame is named dataset1_csv
num_rows, num_columns = dataset1_csv.shape
print(' Number of rows '+ str(num_rows))
print(' Number of Columns '+ str(num_columns))

# Number of Rows (num_rows): This represents the total number of observations␣
 ↪or data points in your dataset. Each row typically corresponds to a single␣
 ↪data entry or record.
```

```
 Number of rows 16719
 Number of Columns 16
```

Number of Rows (num_rows): 16719

This represents the total number of observations or data points in your dataset. Each row typically corresponds to a single data entry or record.

Number of Columns (num_columns): 16

This represents the number of variables or features in your dataset. Each column typically corresponds to a different attribute or characteristic of the data. Columns can contain va types of information, such as numerical values, text, dates, or categorical data.

Understanding the dimensions of DataFrame is essential for data analysis because it helps to g size and structure of dataset, which is fundamental for making data-driven decisions and perfo various data manipulation and analysis tasks.

[27]:
```
# 4 - Find the top five games by critic score.
# Critic_score - Aggregate score compiled by Metacritic staff
# Sort the DataFrame by 'Critic_Score' column in descending order and get the␣
 ↪top 5 rows
```

```
top_five_games = dataset1_csv.nlargest(5, 'Critic_Score')

# Display the top five games by critic score in table format
#print(top_five_games.to_string)
top_five_games.head()
```

[27]:

	Name	Platform	Year_of_Release	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	Critic_Score	Critic_Count	User_Score	User_Count	Developer	Rating
51	Grand Theft Auto IV	X360	2008.0	Action	Take-Two Interactive	6.76	3.07	0.14	1.03	11.01	98.0	86.0	7.9	2951.0	Rockstar North	M
57	Grand Theft Auto IV	PS3	2008.0	Action	Take-Two Interactive	4.76	3.69	0.44	1.61	10.50	98.0	64.0	7.5	2833.0	Rockstar North	M
227	Tony Hawk's Pro Skater 2	PS	2000.0	Sports	Activision	3.05	1.41	0.02	0.20	4.68	98.0	19.0	7.7	299.0	Neversoft Entertainment	T
5350	SoulCalibur	DC	1999.0	Fighting	Namco Bandai Games	0.00	0.00	0.34	0.00	0.34	98.0	24.0	8.8	200.0	Namco	T
16	Grand Theft Auto V	PS3	2013.0	Action	Take-Two Interactive	7.02	9.09	0.98	3.96	21.04	97.0	50.0	8.2	3994.0	Rockstar North	M

[28]:
```
# 5 - Find the number of video games in the data frame in each genre.
## count based on 'Genre' column
genre_counts = dataset1_csv['Genre'].value_counts()

# Display the number of video games in each genre
print(genre_counts)
```

```
Action          3370
Sports          2348
Misc            1750
Role-Playing    1500
Shooter         1323
Adventure       1303
Racing          1249
```

```
Platform         888
Simulation       874
Fighting         849
Strategy         683
Puzzle           580
Name: Genre, dtype: int64
```

[29]:
```
# 6. Find the first five games in the data frame on the SNES platform.
snes_games = dataset1_csv[dataset1_csv['Platform'] == 'SNES'].head(5)

# Display the first five games on the SNES platform
snes_games.head()
```

[29]:

	Name	Platform	Year_of_Release	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	Critic_Score	Critic_Count	User_Score	User_Count	Developer	Rating
18	Super Mario World	SNES	1990.0	Platform	Nintendo	12.78	3.75	3.54	0.55	20.61	NaN	NaN	NaN	NaN	NaN	NaN
56	Super Mario All-Stars	SNES	1993.0	Platform	Nintendo	5.99	2.15	2.12	0.29	10.55	NaN	NaN	NaN	NaN	NaN	NaN
71	Donkey Kong Country	SNES	1994.0	Platform	Nintendo	4.36	1.71	3.00	0.23	9.30	NaN	NaN	NaN	NaN	NaN	NaN
76	Super Mario Kart	SNES	1992.0	Racing	Nintendo	3.54	1.24	3.81	0.18	8.76	NaN	NaN	NaN	NaN	NaN	NaN
137	Street Fighter II: The World Warrior	SNES	1992.0	Fighting	Capcom	2.47	0.83	2.87	0.12	6.30	NaN	NaN	NaN	NaN	NaN	NaN

[30]:
```
# 7. Find the five publishers with the highest total global sales.
#    Note: You will need to calculate the total global sales for each␣
 ↪publisher to do this.

# Group the DataFrame by 'Publisher' and calculate the sum of 'Global_Sales'␣
 ↪for each publisher
publisher_sales = dataset1_csv.groupby('Publisher')['Global_Sales'].sum()

# Sort the publishers by total global sales in descending order and get the top␣
 ↪5
top_publishers = publisher_sales.sort_values(ascending=False).head(5)

# Display the top 5 publishers with the highest total global sales
```

```
print(top_publishers)
```

```
Publisher
Nintendo                        1788.81
Electronic Arts                 1116.96
Activision                       731.16
Sony Computer Entertainment      606.48
Ubisoft                          471.61
Name: Global_Sales, dtype: float64
```

[31]:
```
# 8. Create a new column in the data frame that calculates the percentage of␣
 ↪global sales from North America.
#.  Display the first five rows of the new data frame.
# --> New column name is 'NA_Sales_Percentage'

# Calculate the percentage of global sales from North America and store it in a␣
 ↪new column 'NA_Sales_Percentage'
dataset1_csv['NA_Sales_Percentage'] = (dataset1_csv['NA_Sales'] /␣
 ↪dataset1_csv['Global_Sales']) * 100

# Display the first five rows of the DataFrame with the new column
dataset1_csv.head()
```

[31]:

	Name	Platform	Year_of_Release	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	

	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales	Critic_Score	\
0	41.36	28.96	3.77	8.45	82.53	76.0	
1	29.08	3.58	6.81	0.77	40.24	NaN	
2	15.68	12.76	3.79	3.29	35.52	82.0	
3	15.61	10.93	3.28	2.95	32.77	80.0	
4	11.27	8.89	10.22	1.00	31.37	NaN	

	Critic_Count	User_Score	User_Count	Developer	Rating	NA_Sales_Percentage
0	51.0	8	322.0	Nintendo	E	50.115110
1	NaN	NaN	NaN	NaN	NaN	72.266402
2	73.0	8.3	709.0	Nintendo	E	44.144144
3	73.0	8	192.0	Nintendo	E	47.635032
4	NaN	NaN	NaN	NaN	NaN	35.926044

[32]:
```
# 9. Find the number NaN entries (missing data values) in each column.
# Count the number of NaN entries in each column
nan_counts = dataset1_csv.isna().sum()
```

```
# Display the number of NaN entries in each column
print(nan_counts)
```

```
Name                    2
Platform                0
Year_of_Release       269
Genre                   2
Publisher              54
NA_Sales                0
EU_Sales                0
JP_Sales                0
Other_Sales             0
Global_Sales            0
Critic_Score         8582
Critic_Count         8582
User_Score           6704
User_Count           9129
Developer            6623
Rating               6769
NA_Sales_Percentage     0
dtype: int64
```

[33]:
```
# 10 . Try to calculate the median user score of all the video games.
#   You will likely run into an error because some of the user score entries␣
 ↪are a non-numerical string that cannot be converted to a float.
#.  Find and replace this string with NaN and then calculate the median.
#  Then, replace all NaN entries in the user score column with the median value.

# Replace non-numeric strings with NaN in the 'User_Score' column
dataset1_csv['User_Score'] = pd.to_numeric(dataset1_csv['User_Score'],␣
 ↪errors='coerce')

# Calculate the median of the 'User_Score' column, excluding NaN values
median_user_score = dataset1_csv['User_Score'].median()

# Replace NaN entries in the 'User_Score' column with the median value
dataset1_csv['User_Score'].fillna(median_user_score, inplace=True)

# Display the median user score and  few rows of the DataFrame
print(f"Median User Score: {median_user_score}")
dataset1_csv.head(10)
```

```
Median User Score: 7.5
```

[33]:

	Name	Platform	Year_of_Release	Genre	\
0	Wii Sports	Wii	2006.0	Sports	
1	Super Mario Bros.	NES	1985.0	Platform	

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

  Publisher  NA_Sales  EU_Sales  JP_Sales  Other_Sales  Global_Sales  \
0  Nintendo     41.36     28.96      3.77         8.45         82.53
1  Nintendo     29.08      3.58      6.81         0.77         40.24
2  Nintendo     15.68     12.76      3.79         3.29         35.52
3  Nintendo     15.61     10.93      3.28         2.95         32.77
4  Nintendo     11.27      8.89     10.22         1.00         31.37
5  Nintendo     23.20      2.26      4.22         0.58         30.26
6  Nintendo     11.28      9.14      6.50         2.88         29.80
7  Nintendo     13.96      9.18      2.93         2.84         28.92
8  Nintendo     14.44      6.94      4.70         2.24         28.32
9  Nintendo     26.93      0.63      0.28         0.47         28.31

   Critic_Score  Critic_Count  User_Score  User_Count Developer Rating  \
0          76.0          51.0         8.0       322.0  Nintendo      E
1           NaN           NaN         7.5         NaN       NaN    NaN
2          82.0          73.0         8.3       709.0  Nintendo      E
3          80.0          73.0         8.0       192.0  Nintendo      E
4           NaN           NaN         7.5         NaN       NaN    NaN
5           NaN           NaN         7.5         NaN       NaN    NaN
6          89.0          65.0         8.5       431.0  Nintendo      E
7          58.0          41.0         6.6       129.0  Nintendo      E
8          87.0          80.0         8.4       594.0  Nintendo      E
9           NaN           NaN         7.5         NaN       NaN    NaN

   NA_Sales_Percentage
0            50.115110
1            72.266402
2            44.144144
3            47.635032
4            35.926044
5            76.668870
6            37.852349
7            48.271093
8            50.988701
9            95.125397
```

[ ]: