

# VIDEO GAME SALES ANALYSIS USING PYTHON

Prepared by,

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## A. Data set:

https://www.kaggle.com/gregorut/videogamesales/version/2#vgsales.csv/

## **Data description:**

This dataset contains a list of video games with sales greater than 100,000 copies.

Rank - Ranking of overall sales

- · Name The games name
- Platform Platform of the games release (i.e. PC, PS4, etc.)
- · Year Year of the game's release
- · Genre Genre of the game
- **Publisher** Publisher of the game
- · NA\_Sales Sales in North America (in millions)
- EU Sales Sales in Europe (in millions)
- · JP Sales Sales in Japan (in millions)
- Other\_Sales Sales in the rest of the world (in millions)
- Global Sales Total worldwide sales.

Field Name	eld Name Type		Range of	Attributes/Form	Comments
		Value	values	at	
Year	Textual	1-01-1994	1-01-1985	dd-mm-yyyy	OUT OF
			to 1-01-		DATE
			2016		
rank	Quantitativ	3526	1 to 9303	Whole numbers	Some
	e				erroneous
					values
Name	Textual	FIFA 14	NA	Textual	
				FORMAT	
PLATFORM	Textual	X360	NA	Textual	MISSING
				FORMAT	VALUES
PUBLISHER	Textual	UBisoft	NA	Textual	
				FORMAT	
Genre	Textual	Action	NA	Textual	
				FORMAT	
NA_SALES	Quantitativ	3.19	0.1 - 41.49	Decimal	Shows no
	e	MILLIO	(MILLION)	numbers	of active
		N			users

EU_SALES	Quantitativ	0.92	0.1 - 29.02	Decimal	Some
	e	MILLIO		numbers	values
		N	(MILLION)		missing
JP_SALES	Quantitativ	0.01	0.1 - 10.22	Decimal	Some
	e	MILLIO	(MILLION)	numbers	values
		N			missing
OTHER_SALES	Quantitativ	0.42	0.1 - 10.57	Decimal	MISSING
	e	MILLIO	(MILLION)	numbers	VALUES
		N			
GLOBAL_SALE	Quantitativ	4.53	0.05 –	Decimal	MISSING
S	e	MILLIO	82.75	numbers	VALUES
		N	(MILLION)		

# B. Data cleaning

## 1. Removing rows with null values:

File contained Nan values as shown below in before screenshot, these Nan values was not in one single column but multiple columns through out the data set. To get rid of this null values we used **Drop function** and along with this function we have used **File function**.

The second screenshot displays the data set without Null values, all the Nan/null values in all of the column are been removed.

#### Before:

		taFrame.dropna of Rank			Name	Platform Year	Genre		Publisher	NA_Sales	EU_Sales JP_S
bal_Sa	les	Mild words							Towns I		
	1	Wii Sports	Wii	2006.0	Sports	Nintendo		29.02	NaN	8.46	82.74
	2	Super Mario Bros.	NES	1985.0	Platform	Nintendo		3.58	6.81	0.77	40.24
	3	Mario Kart Wii	Wii	2008.0	Racing	Nintendo		12.88	3.79	3.31	35.82
	4	Wii Sports Resort	Wii	2009.0	Sports	Nintendo		11.01	3.28	2 06	33.00
	5	Pokemon Red/Pokemon Blue	GB	1996.0	Role-Playing	Nintendo		8.89	10.22	NaN	31.37
	6	Tetris	GB	1989.0	Puzzle	Nintendo		2.26	4.22	0.58	30.26
	7	New Super Mario Bros.	DS	2006.0	Platform	Nintendo		9.23	6.50	2.90	30.01
	8	Wii Play	Wii	2006.0	Misc	Nintendo		9.20	2.93	2.85	29.02
	8	Wii Play	Wii	2006.0	Misc	Nintendo		9.20	2.93	2.85	29.02
		New Super Mario Bros. Wii	Wii	2009.0	Platform	Nintendo		7.06	4.70	2.26	28.62
ð	10	Duck Hunt	NES	1984.0	Shooter	Nintendo		0.63	0.28	0.47	28.31
	11	Nintendogs	DS	2005.0	Simulation	Nintendo		11.00	1.93	2.75	24.76
2	12	Mario Kart DS	DS	2005.0	Racing	Nintendo		7.57	4.13	1.92	23.42
3	13	Pokemon Gold/Pokemon Silver	GB	1999.0	Role-Playing	Nintendo		6.18	7.20	0.71	23.10
4	13	Pokemon Gold/Pokemon Silver	GB	1999.0	Role-Playing	Nintendo	9.00	6.18	7.20	0.71	23.10
5	14	Wii Fit	Wii	2007.0	Sports	Nintendo	8.94	8.03	3.60	2.15	22.72
5	15	Wii Fit Plus	Wii	2009.0	Sports	Nintendo	9.09	8.59	2.53	1.79	22.00
7	16	Kinect Adventures!	X360	2010.0	Misc	Microsoft Game Studios	14.97	4.94	0.24	1.67	21.82
8	17	Grand Theft Auto V	PS3	2013.0	Action	Take-Two Interactive	7.01	9.27	NaN	4.14	21.40
,	18	Grand Theft Auto: San Andreas	PS2	2004.0	Action	Take-Two Interactive	9.43	0.40	0.41	10.57	20.81
3	19	Super Mario World	SNES	1990.0	Platform	Nintendo	12 78	3.75	3.54	0.55	20.61
1	20	Brain Age: Train Your Brain in Minutes a Day	DS	2005.0	Misc	Nintendo	NaN	9.26	4.16	2.05	20.22
2	21	Pokemon Diamond/Pokemon Pearl	DS	2006.0	Role-Playing	Nintendo	6.42	4.52	6.04	1.37	18.36
3	22	Super Mario Land	GB	1989.0	Platform	Nintendo	10.83	2.71	4.18	0.42	18.14
	23	Super Mario Bros. 3	NES	1988.0	Platform	Nintendo	9.54	3.44	3.84	0.46	17.28
5	24	Grand Theft Auto V	X360	2013.0	Action	Take-Two Interactive	9.63	5.31	0.06	1.38	16.38
6	25	Grand Theft Auto: Vice City	PS2	2002.0	Action	Take-Two Interactive		5.49	NaN	1.78	16.15
7	26	Pokemon Ruby/Pokemon Sapphire	GBA	2002.0	Role-Playing	Nintendo		3.90	5.38	0.50	15.85
8	27	Pokemon Black/Pokemon White	DS	2010.0	Role-Playing	Nintendo		3.28	5.65	0.82	15.32
9	28	Brain Age 2: More Training in Minutes a Day	DS	2005.0	Puzzle	Nintendo		5.36	5.32	1.18	15.30

code:

```
6  @author: neelsavla
7   """
8
9   import pandas as pd
10   import matplotlib.pyplot as plt
11   games=pd.read_csv('vgsales.csv')
12   #print(games.head(100))
13   print(games.dropna())
14
```

## After:

### 2. Remove irrelevant columns:

In the data set their was an irrelevant column which was the copy of the first column 'Rank' named differently as 'Rank\_2000' which was not of any use, thus, the column was removed using **del function**, which is a inbuilt function used to delete the column.

The code we used is del games ['Rank\_2000'], here the game is the instance where we stored the csv and Rank\_2000 is the column to be deleted and finally used del function to delete the irrelevant column from our data set. We have used **File function** for this code

```
@author: neelsavla
import pandas as pd
import matplotlib.pyplot as plt
games=pd.read_csv('vgsales.csv')
del games['Rank_2000']
print(games.dropna())
```

#### Before:

```
| Neels-MedShork-Pro:spider neelsevies python project.py | Name Platform | Nam
```

#### After:

Nee	ls-MacBook-Pro	sspider neelsavla\$ python project.py									
	Rank		Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sales	Other_Sales	Global_Sales
1		Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24
2		Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82
3		Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.75	11.01	3.28	2.96	33.00
5		Tetris	GB	1989.0	Puzzle	Nintendo	23.20	2.26	4.22	0.58	30.26
6		New Super Mario Bros.	DS	2006.0	Platform	Nintendo	11.38	9.23	6.50	2.90	30.01
7	8	Wii Play	Wii	2006.0	Misc	Nintendo	14.03	9.20	2.93	2.85	29.02
8	8	Wii Play	Wii	2006.0	Misc	Nintendo	14.03	9.20	2.93	2.85	29.02
9		New Super Mario Bros. Wii	Wii	2009.0	Platform	Nintendo	14.59	7.06	4.70	2.26	28.62
10	10	Duck Hunt	NES	1984.0	Shooter	Nintendo	26.93	0.63	0.28	0.47	28.31
11	11	Nintendogs	DS	2005.0	Simulation	Nintendo	9.07	11.00	1.93	2.75	24.76
12	12	Mario Kart DS	DS	2005.0	Racing	Nintendo	9.81	7.57	4.13	1.92	23.42
13	13	Pokemon Gold/Pokemon Silver	GB	1999.0		Nintendo	9.00	6.18	7.20	0.71	23.10
14	13	Pokemon Gold/Pokemon Silver	GB	1999.0	Role-Playing	Nintendo	9.00	6.18	7.20	0.71	23.10
15	14	Wii Fit	Wii	2007.0	Sports	Nintendo	8.94	8.03	3.60	2.15	22.72
16	15	Wii Fit Plus	Wii	2009.0	Sports	Nintendo	9.09	8.59	2.53	1.79	22.00
17	16	Kinect Adventures!	X360	2010.0	Misc	Microsoft Game Studios	14.97	4.94	0.24	1.67	21.82
19	18	Grand Theft Auto: San Andreas	PS2	2004.0	Action	Take-Two Interactive	9.43	0.40	0.41	10.57	20.81
20	19	Super Mario World	SNES	1990.0	Platform	Nintendo	12.78	3.75	3.54	0.55	20.61
22	21	Pokemon Diamond/Pokemon Pearl	DS	2006.0		Nintendo	6.42	4.52	6.04	1.37	18.36
23	22	Super Mario Land	GB	1989.0	Platform	Nintendo	10.83	2.71	4.18	0.42	18.14
24	23	Super Mario Bros. 3	NES	1988.0	Platform	Nintendo	9.54	3.44	3.84	0.46	17.28
25	24	Grand Theft Auto V	X360	2013.0	Action	Take-Two Interactive	9.63	5.31	0.06	1.38	16.38
27	26	Pokemon Ruby/Pokemon Sapphire	GBA	2002.0		Nintendo	6.06	3.90	5.38	0.50	15.85
28	27	Pokemon Black/Pokemon White	DS	2010.0		Nintendo	5.57	3.28	5.65	0.82	15.32
29	28	Brain Age 2: More Training in Minutes a Day	DS	2005.0	Puzzle	Nintendo	3.44	5.36	5.32	1.18	15.30
30	29	Gran Turismo 3: A-Spec	PS2	2001.0	Racing	Sony Computer Entertainment	6.85	5.09	1.87	1.16	14.98
31	30	Call of Duty: Modern Warfare 3	X360	2011.0	Shooter	Activision	9.03	4.28	0.13	1.32	14.76
32	31	Pokémon Yellow: Special Pikachu Edition	GB	1998.0		Nintendo	5.89	5.04	3.12	0.59	14.64
34	33	Pokemon X/Pokemon Y	3DS	2013.0	Role-Playing	Nintendo	5.17	4.05	4.34	0.79	14.35
35	34	Call of Duty: Black Ops 3	PS4	2015.0	Shooter	Activision	5.77	5.81	0.35	2.31	14.24

## 3. Remove duplicate data:

The data set contained multiple duplicate values as show in screenshot for before in the second column called name contains duplicate data such as 'Wii Play' and 'Pokemon Gold/ Pokemon Silver' which are occurring twice in the data set. The column name 'Name' is a column containing unique values thus, it should not have duplicate values. Thus, we are removing the rows with duplicate value occurring only once.

Here we have used 'drop\_duplicates' functions to remove the row with duplicate value. As show in the second screenshot the row with duplicate values is removed. Now the value which was duplicate has only one occurence in the data set making the column unique which is exactly as per the requirement. We have used **Pandas Data Frame** for this code Before:

[Neels-MacBook-Pro:spider neelsavla\$ python project.py											
	Rank		Platform	Year	Genre	Publisher		EU_Sales		Other_Sales	Global_Sales
1		Super Mario Bros.	NES	1985.0	Platform	Nintendo	29.08	3.58	6.81	0.77	40.24
2		Mario Kart Wii	Wii	2008.0	Racing	Nintendo	15.85	12.88	3.79	3.31	35.82
3		Wii Sports Resort	Wii	2009.0	Sports	Nintendo	15.75	11.01	3.28	2.96	33.00
5		Tetris	GB	1989.0	Puzzle	Nintendo	23.20	2.26	4.22	0.58	30.26
٨	7	New Super Marin Bree	De	2006 0	Dlatform	Nintendo	11 28	0 22	4 F.A	2 00	20 01
7	8	Wii Play	Wii	2006.0	Misc	Nintendo	14.03	9.20	2.93	2.85	29.02
8	8	Wii Play	Wii	2006.0	Misc	Nintendo	14.03	9.20	2.93	2.85	29.02
y	y	New Super Mario bros. Wil		∠007.0	LTACIOIM	MILLEGIGO	14.57	/.00	4./0	۷،۷٥	20.02
10	10	Duck Hunt	NES	1984.0	Shooter	Nintendo	26.93	0.63	0.28	0.47	28.31
11	11	Nintendogs	DS	2005.0	Simulation	Nintendo	9.07	11.00	1.93	2.75	24.76
12	12	Mario Kart DS		2005.0	Racing	Nintendo	9.81	7.57	4.13	1.92	23.42
13	13	Pokemon Gold/Pokemon Silver			Role-Playing	Nintendo	9.00	6.18	7.20	0.71	23.10
14	13	Pokemon Gold/Pokemon Silver			Role-Playing	Nintendo	9.00	6.18	7.20	0.71	23.10
15	14	Wii Fit		2007.0	Sports	Nintendo	8.94	8.03	3.60	2.15	22.72
16	15	Wii Fit Plus	Wii	2009.0	Sports	Nintendo	9.09	8.59	2.53	1.79	22.00
17	16	Kinect Adventures!	X360	2010.0	Misc	Microsoft Game Studios	14.97	4.94	0.24	1.67	21.82
19	18	Grand Theft Auto: San Andreas	PS2	2004.0	Action	Take-Two Interactive	9.43	0.40	0.41	10.57	20.81
20	19	Super Mario World	SNES	1990.0	Platform	Nintendo	12.78	3.75	3.54	0.55	20.61
22	21	Pokemon Diamond/Pokemon Pearl	DS	2006.0	Role-Playing	Nintendo	6.42	4.52	6.04	1.37	18.36
23	22	Super Mario Land	GB	1989.0	Platform	Nintendo	10.83	2.71	4.18	0.42	18.14
24	23	Super Mario Bros. 3	NES	1988.0	Platform	Nintendo	9.54	3.44	3.84	0.46	17.28
25	24	Grand Theft Auto V	X360	2013.0	Action	Take-Two Interactive	9.63	5.31	0.06	1.38	16.38
27	26	Pokemon Ruby/Pokemon Sapphire	GBA	2002.0	Role-Playing	Nintendo	6.06	3.90	5.38	0.50	15.85
28	27	Pokemon Black/Pokemon White	DS		Role-Playing	Nintendo	5.57	3.28	5.65	0.82	15.32
29	28	Brain Age 2: More Training in Minutes a Day	DS	2005.0	Puzzle	Nintendo	3.44	5.36	5.32	1.18	15.30
30	29	Gran Turismo 3: A-Spec	PS2	2001.0	Racing	Sony Computer Entertainment	6.85	5.09	1.87	1.16	14.98
31	30	Call of Duty: Modern Warfare 3	X360	2011.0	Shooter	Activision	9.03	4.28	0.13	1.32	14.76
32	31	Pokémon Yellow: Special Pikachu Edition	GB	1998.0	Role-Playing	Nintendo	5.89	5.04	3.12	0.59	14.64
34	33	Pokemon X/Pokemon Y	3DS	2013.0	Role-Playing	Nintendo	5.17	4.05	4.34	0.79	14.35
35	34	Call of Duty: Black Ops 3	PS4	2015.0	Shooter	Activision	5.77	5.81	0.35	2.31	14.24

Code:

```
import pandas as pd
import matplotlib.pyplot as plt
games=pd.read_csv('vgsales.csv')
del games['Rank_2000']
games.drop_duplicates(subset="Name",keep= 'first' ,inplace= True)
print(games.dropna())
```

After:

```
| Name | Platform | Name |
```

#### C. Statistical Summary:

To find the statistical summary .describe() function is used and is applied to the required columns such as for sales in North America, Sales in Europe, Sales in Japan, Other sales and Global Sales. Using .describe() we found the count, mean, standard deviation, minimum, 25%, 50%, 75% and maximum value for each of the columns specified above.

For example, if we see the Sales for North America in the screenshot below we can see that the count of sale is 16599, mean of sales is 0.265753, standard deviation is 0.825670, minimum sales if 0.0000, 25% of the sales is 0.00000, 50% of the sales is 0.80000, 75% of the sales is 0.240000 and finally maximum value of sales for North America is 41.490000.

```
project_cleaning.py x project_statistics.py x

import pandas as pd
import matplotlib.pyplot as plt
games=pd.read_csv('vgsales1.csv')

print(" Summary Statistics Sales for North America :\n",games['NA_Sales'].describe())

print(" Summary Statistics Sales for Europe:\n",games['EU_Sales'].describe())

print(" Summary Statistics Sales for Japan:\n",games['JP_Sales'].describe())

print(" Summary Statistics Sales for Other Sales:\n",games['Other_Sales'].describe())

print(" Summary Statistics Global Sales:\n",games['Global_Sales'].describe())
```

```
[Neels-MacBook-Pro:spider neelsavla$ python project_statistics.py
 Summary Statistics Sales for North America:
          16599.000000
             0.265753
mean
std
             0.825670
min
             0.000000
25%
             0.000000
50%
             0.080000
75%
             0.240000
            41,490000
max
Name: NA_Sales, dtype: float64
 Summary Statistics Sales for Europe:
 count
          16600.000000
             0.147561
mean
std
             0.512327
min
             0.000000
25%
             0.000000
50%
             0.020000
75%
             0.110000
max
             29.020000
Name: EU Sales, dtype: float64
 Summary Statistics Sales for Japan:
         16597.000000
 count
mean
             0.078083
std
             0.313583
min
             0.000000
25%
             0.000000
50%
             0.000000
75%
             0.040000
max
            10.220000
Name: JP_Sales, dtype: float64
 Summary Statistics Sales for Other Sales:
         16598.000000
 count
mean
             0.048149
std
             0.189578
min
             0.000000
25%
             0.000000
50%
             0.010000
75%
             0.040000
max
            10.570000
Name: Other_Sales, dtype: float64
 Summary Statistics Global Sales:
 count
          16600.000000
mean
             0.540516
std
             1.580302
min
             0.010000
25%
             0.060000
50%
             0.170000
75%
             0.470000
            82.740000
max
Name: Global_Sales, dtype: float64
```

## D. Data cleaning Analysis & Visualization

## Question 1:

What is the Video game sales pattern in entire world?

1) Chart use- Pie Chart

```
🗀 viz.py 🛚
  1 import pandas as pd
  2 import matplotlib.pyplot as plt
  3 game=pd.read_csv('vgsales.csv')
  4 na=game['NA_Sales']
  5 na_game=sum(game['NA_Sales'])
  6 jp=game['JP_Sales']
  7 jp_game=sum(game['JP_Sales'])
  8 eu=game['EU_Sales']
  9 eu_game=sum(game['EU_Sales'])
  10 ot=game['Other_Sales']
  11 ot_game=sum(game['Other_Sales'])
  12 region=('North_America', 'Europe', 'Others', 'Japan')
  13 pie=(na_game,eu_game,ot_game,jp_game)
  14 colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728", "#8c564b"]
  15 plt.pie(pie, labels=region, colors=colors,
  16 autopct='%1.1f%%', shadow=True, startangle=140)
  17 plt.title("WORLD WIDE VIDEO GAME SALES")
  18 plt.show()
  19
  20
```

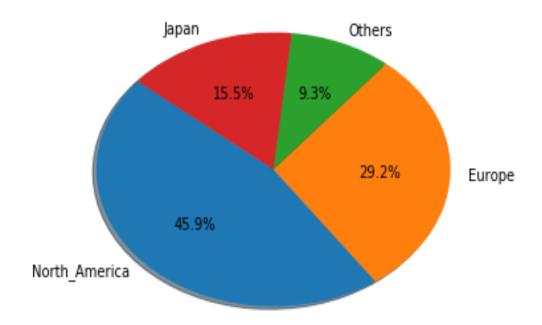
To get Insight of the data, we need to compare some important statistic of the data. This

Visualization helps us to understand the the global sales pattern. We have divided world into
four major region, those region are North America, Japan, Europe and Other part.

Not plotting the **Pie chart** with the attributes of region which is mentioned previously,
We see the best region to concentrate that's the region with major sales.

North America being the dominant sales region in the world containing 45.9% of the worlds game sales, followed by Europe region with 29.2%, Japan having 15.5% of the worlds sales share. We have used **Dictionary and List** for this code

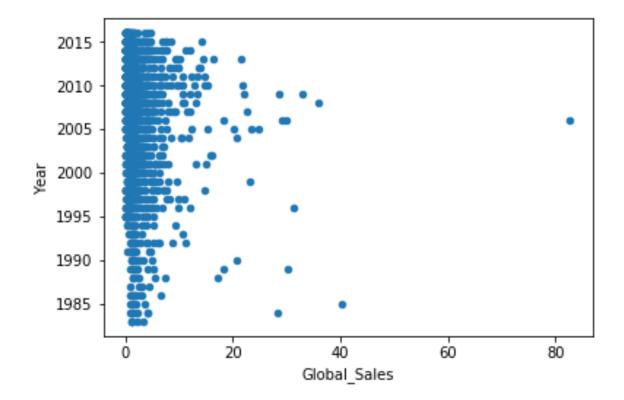


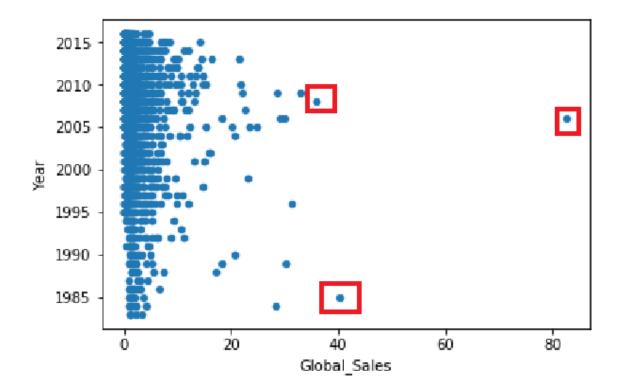


## Question 2:

What insights one might get by comparing top three performing games of all time?

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 game=pd.read_csv('vgsales.csv')
4 game.plot(kind='scatter', x= 'Global_Sales', y= 'Year')
5 plt.show()
6
```





Top Performing games Wii Sports - 80.74 (2006), Super Mario Bros -40.24 (1985), Wii Sports-82.74 (2008)
Interesting fact - Top three games have the same publisher Nintendo!

## Question 3:

Does different type of game has impact on it's sales?

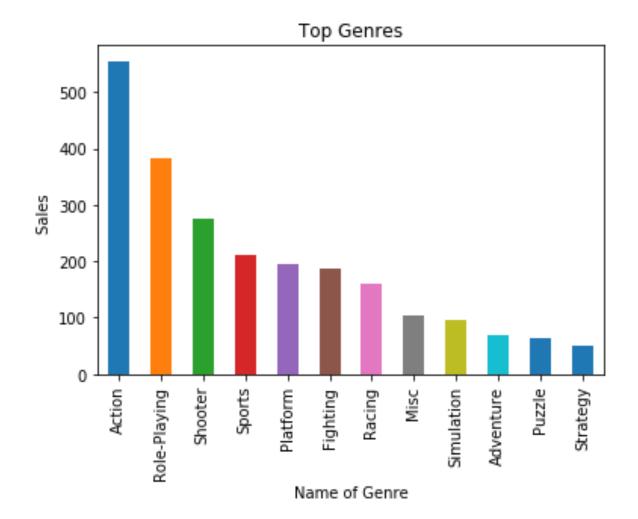
If person trying to make a debut in gaming industries, one of main things that person wants to choose is what type of video game will generate him more profits. Thus, now by looking at the bar chart visualization below one will have a clear view about the sales that video Genre in making, for instance if a person is planning to make a puzzle game which has only 20 millions compared to the sports games which has around 230 millions of sales worldwide.

One can also make a conclusion, that if action genre is the most sold video game in world then, perhaps the action video gaming has much more competition and when there is a less sales in strategic video game, then the competition is fever compared to action game.

We have used **String** for this code

Code:

```
import pandas as pd
import matplotlib.pyplot as plt
games=pd.read_csv('vgsales.csv')
count=pd.value_counts(games['Genre'].values)
print(count)
fig=count.plot(kind='bar')
fig.set_title('Top Genres')
fig.set_xlabel('Name of Genre')
fig.set_ylabel('Sales ')
plt.show()
```



## E. Consolidated Code:

```
Code for testing:
import pandas as pd
games=pd.read csv('vgsales.csv')
del games['Rank_2000']
games.drop duplicates(subset="Name",keep= 'first',inplace= True)
print(games.dropna())
Code for visuals:
Question 1:
import pandas as pd
import matplotlib.pyplot as plt
game=pd.read_csv('vgsales.csv')
na=game['NA_Sales']
na game=sum(game['NA Sales'])
jp=game['JP_Sales']
jp game=sum(game['JP Sales'])
eu=game['EU_Sales']
eu game=sum(game['EU Sales'])
ot=game['Other_Sales']
ot_game=sum(game['Other_Sales'])
region=('North_America','Europe','Others','Japan')
pie=(na game,eu game,ot game,jp game)
```

```
colors = ["#1f77b4", "#ff7f0e", "#2ca02c", "#d62728", "#8c564b"]
plt.pie(pie, labels=region, colors=colors,
autopct='%1.1f%%', shadow=True, startangle=140)
plt.title("WORLD WIDE VIDEO GAME SALES")
plt.show()
Question 2:
import pandas as pd
import matplotlib.pyplot as plt
game=pd.read csv('vgsales.csv')
game.plot(kind='line', x= 'Global_Sales', y= 'Year')
plt.show()
Question 3:
import pandas as pd
import matplotlib.pyplot as plt
games=pd.read csv('vgsales.csv')
count=pd.value_counts(games['Genre'].values)
print(count)
fig=count.plot(kind='bar')
fig.set title('Top Genres')
fig.set xlabel('Name of Genre')
fig.set_ylabel('Sales')
plt.show()
```

## Code for summary statistics:

```
import pandas as pd
import matplotlib.pyplot as plt
games=pd.read_csv('vgsales1.csv')
print(" Summary Statistics Sales for North America :\n",games['NA_Sales'].describe())
print(" Summary Statistics Sales for Europe:\n",games['EU_Sales'].describe())
print(" Summary Statistics Sales for Japan:\n",games['JP_Sales'].describe())
print(" Summary Statistics Sales for Other Sales:\n",games['Other_Sales'].describe())
print(" Summary Statistics Global Sales:\n",games['Global Sales'].describe())
```