Data Viz 2

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```
import pandas as pd
In [2]:
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
          data = pd.read_csv('final.csv')
In [3]:
          data.head()
Out[3]:
             Rank
                            Name Platform
                                                       Genre Publisher
                                               Year
                                                                        NA_Sales
                                                                                   EU_Sales
                                                                                             JP_Sa
          0
                             1942
              2061
                                        NES
                                             1985.0
                                                     Shooter
                                                               Capcom
                                                                         4.569217
                                                                                   3.033887
                                                                                             3.439
                    ¡Shin Chan Flipa
                                                                   505
              9137
                                         DS
                                             2007.0
                                                     Platform
                                                                         2.076955
                                                                                   1.493442 3.033
                        en colores!
                                                                 Games
                     .hack: Sekai no
                                                                Namco
          2 14279
                        Mukou ni +
                                        PS3
                                             2012.0
                                                       Action
                                                                 Bandai
                                                                         1.145709
                                                                                   1.762339 1.493
                            Versus
                                                                 Games
                                                                Namco
                         .hack//G.U.
                                                        Role-
                                        PS2 2006.0
          3
             8359
                                                                 Bandai
                                                                         2.031986
                                                                                   1.389856 3.2280
                      Vol.1//Rebirth
                                                      Playing
                                                                 Games
                                                                Namco
                         .hack//G.U.
                                                        Role-
                                        PS2 2006.0
                                                                 Bandai
                                                                         2.792725 2.592054 1.440
                    Vol.2//Reminisce
                                                      Playing
                                                                 Games
```

In [4]: data.describe()

Out[4]:

	Rank	Year	NA_Sales	EU_Sales	JP_Sales	Other_Sa
count	16652.000000	16381.000000	16652.000000	16652.000000	16652.000000	16652.0000
mean	8283.409620	2006.390513	2.752314	1.996875	2.499677	1.1518
std	4794.471477	5.863261	1.327002	1.322972	1.164023	1.0548
min	1.000000	1980.000000	0.140000	0.010000	0.000000	-0.4742
25%	4129.750000	2003.000000	1.781124	1.087977	1.781124	0.3948
50%	8273.500000	2007.000000	2.697415	1.714664	2.480356	0.4918
75 %	12436.250000	2010.000000	3.677290	2.795123	3.176299	1.7811
max	16600.000000	2020.000000	8.725452	8.367985	12.722984	7.3580

Continuous-Continuous

So far we have been analyzing only a single feature.

But what if we want to visualize two features at once?

What kind of questions can we ask regarding a continuous-continuous pair of features?

- Show relation between two features, like how does the sales vary over the years?
- Show how are the features associated, positively or negatively?

...and so on

Line Plot

How can we plot the sales trend over the years for the longest running game?

First, let's find the longest running game first.

```
In [5]: game_life = data.groupby('Name').agg(min_year = ('Year', 'min'), max_year =
    game_life['range'] = game_life['max_year'] - game_life['min_year']
    game_life.sort_values(['range'], ascending = False)[:5]
```

min_year max_year range

Out[5]:

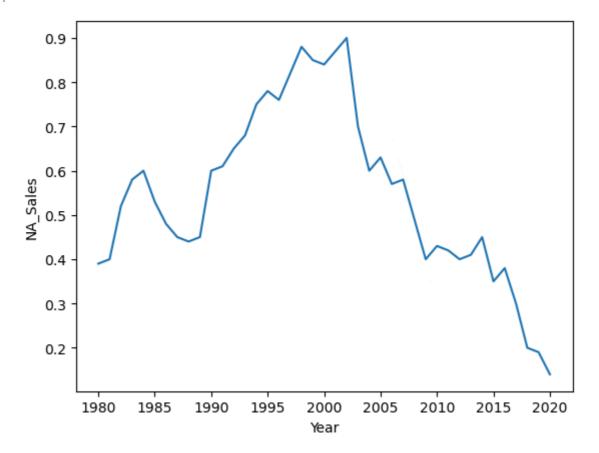
Name			
Ice Hockey	1980.0	2020.0	40.0
Baseball	1980.0	2019.0	39.0
Battlezone	1982.0	2006.0	24.0
Romance of the Three Kingdoms II	1991.0	2015.0	24.0
Bomberman	1985.0	2008.0	23.0

Great! So Ice Hockey is longer running than most of the games.

Let's try to find the sales trend in North America of the same across the years.

```
in [6]: ih = data.loc[data['Name']=='Ice Hockey']
sns.lineplot(x='Year', y='NA_Sales', data=ih)
```

Out[6]: <Axes: xlabel='Year', ylabel='NA_Sales'>



What can we infer from this graph?

- The sales across North America seem to have been boosted in the years of 1995-2005.
- Post 2010 though, the sales seem to have taken a dip.

Line plots are great for representing trends such as above, over time.

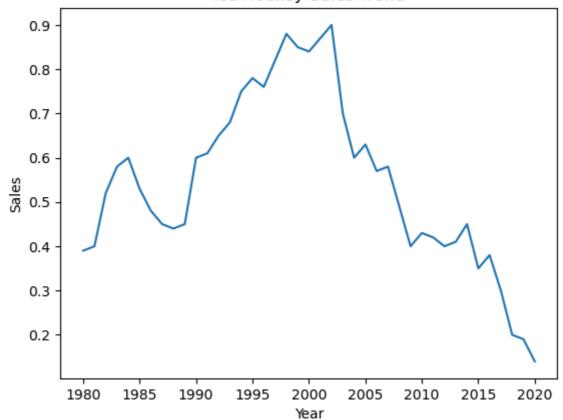
Style and Labelling

We already learnt how to add **titles**, **x-label** and **y-label** in barplot.

Let's do the same here.

```
In [7]: plt.title('Ice Hockey Sales Trend')
   plt.xlabel('Year')
   plt.ylabel('Sales')
   sns.lineplot(x='Year', y='NA_Sales', data=ih)
   plt.show()
```

Ice Hockey Sales Trend



What if we want to change the colour of the curve?

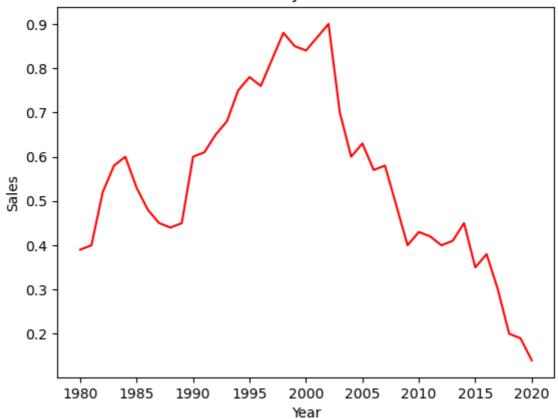
- sns.lineplot() contains an argument color
- It takes a matplotlib color or string for some defined colours.
 - black: k / black
 - red: r / red

But what all colours can we use?

- Matplotlib provides a variety of colors.
- Check the documentation for more https://matplotlib.org/2.0.2/api/colors_api.html

```
In [8]: plt.title('Ice Hockey Sales Trend')
   plt.xlabel('Year')
   plt.ylabel('Sales')
   sns.lineplot(x='Year', y='NA_Sales', data=ih, color='r')
   plt.show()
```

Ice Hockey Sales Trend



Now, let's say we only want to show the values from years 1990-2000.

How can we limit our plot to only the last decade of 20th century?

• This requires changing the range of x-axis.

But how can we change the range of an axis in matplotlib?

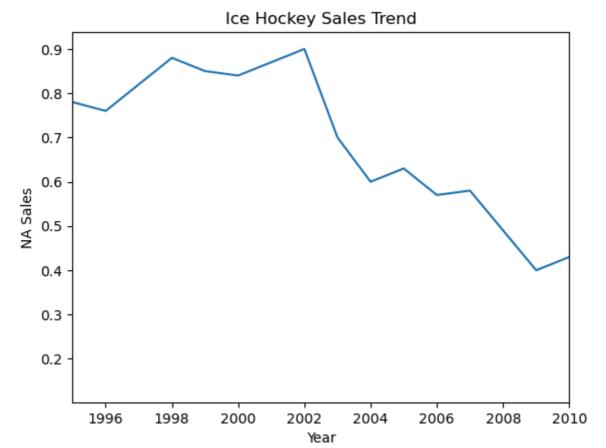
We can use:

plt.xlim(): x-axisplt.ylim(): y-axis

These functions take 2 arguments:

- 1. left: Starting point of range
- 2. right: End point of range

```
In [9]: plt.title('Ice Hockey Sales Trend')
  plt.xlabel('Year')
  plt.ylabel('NA Sales')
  plt.xlim(left=1995,right=2010)
  sns.lineplot(x='Year', y='NA_Sales', data=ih)
  plt.show()
```



So far we have visualised a single plot to understand it.

What if we want to compare it with some other plot?

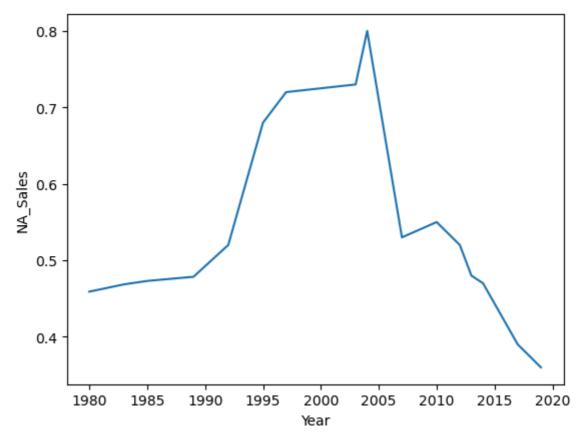
Say, we want to compare the same sales trend between two games.

- Ice Hockey
- Baseball

Let's first plot the trend for "Baseball".

```
In [10]: baseball = data.loc[data['Name']=='Baseball']
sns.lineplot(x='Year', y='NA_Sales', data=baseball)

Out[10]: <Axes: xlabel='Year', ylabel='NA_Sales'>
```



Now for the comparison, we will have to draw these plots in the same figure.

How can we plot multiple plots in the same figure?

```
In [11]: sns.lineplot(x='Year', y='NA_Sales', data=ih) sns.lineplot(x='Year', y='NA_Sales', data=baseball)

Out[11]: <Axes: xlabel='Year', ylabel='NA_Sales'>

0.9

0.8

0.7

0.4

0.3

0.2
```

1995

2000

Year

2005

2010

2015

2020

1980

1985

1990

We can use multiple sns.lineplot() functions.

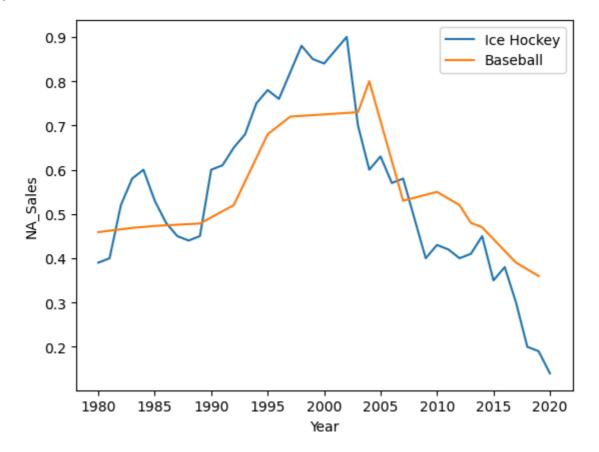
Observe that Seaborn automatically created 2 plots with different colors.

But how can we know which colour is of which plot?

- We can simply set the label of each plot.
- sns.lineplot() has another argument label to do so.

```
In [12]: sns.lineplot(x='Year', y='NA_Sales', data=ih, label='Ice Hockey')
sns.lineplot(x='Year', y='NA_Sales', data=baseball, label='Baseball')
```

Out[12]: <Axes: xlabel='Year', ylabel='NA_Sales'>

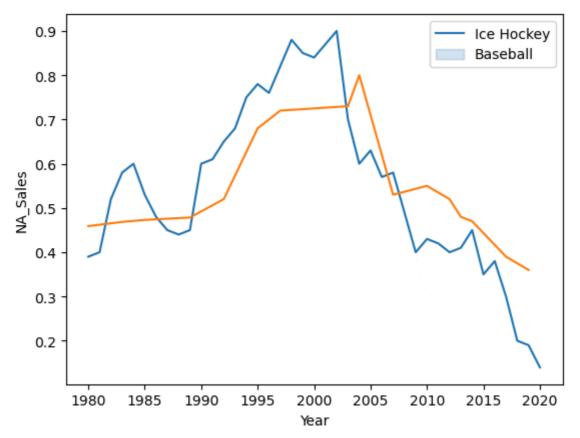


We can also pass these labels in plt.legend() as a list in the order the plots are created.

Please note that using labels in plt.legend() gives unexpected result in **Google**Colab. However, it works absolutely fine on other platforms.

Here, we will use plt.legend() for its other functionalities.

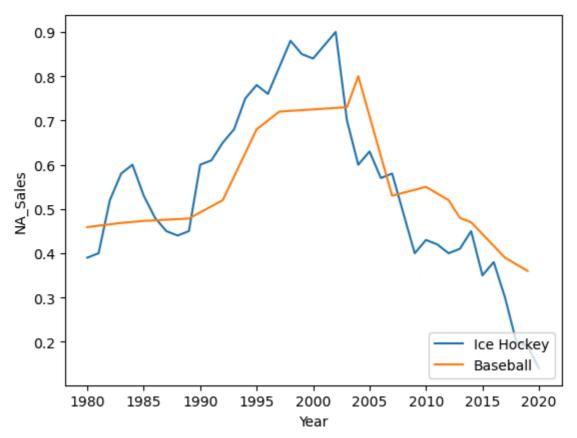
```
In [13]: sns.lineplot(x='Year', y='NA_Sales', data=ih)
    sns.lineplot(x='Year', y='NA_Sales', data=baseball)
    plt.legend(['Ice Hockey', 'Baseball'])
    plt.show()
```



Can we change the position of the legend, say to bottom-right corner?

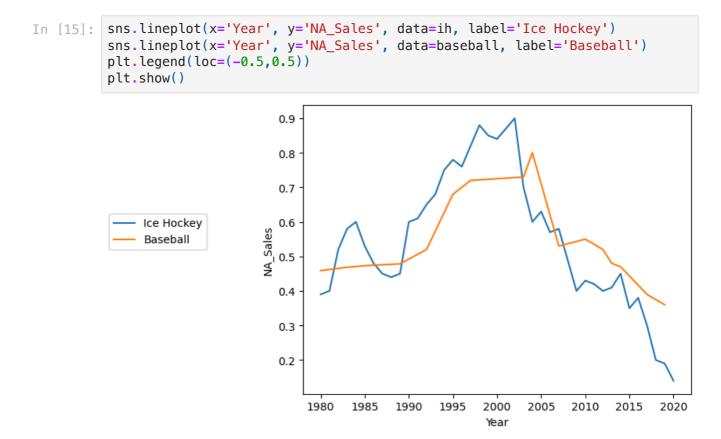
- Matplotlib automatically decides the best position for the legends.
- But we can also change it using the loc parameter.
- loc takes input as 1 of following strings:
 - upper center
 - upper left
 - upper right
 - lower right

```
In [14]: sns.lineplot(x='Year', y='NA_Sales', data=ih, label='Ice Hockey')
    sns.lineplot(x='Year', y='NA_Sales', data=baseball, label='Baseball')
    plt.legend(loc='lower right')
    plt.show()
```



What if we want the legend to be outside the plot?

- Maybe the plot is too congested to show the legend.
- We can use the same loc parameter for this too.



The pair of floats signify the (x,y) coordinates for the legend.

From this we can conclude that loc takes **two types of arguments**:

10/28

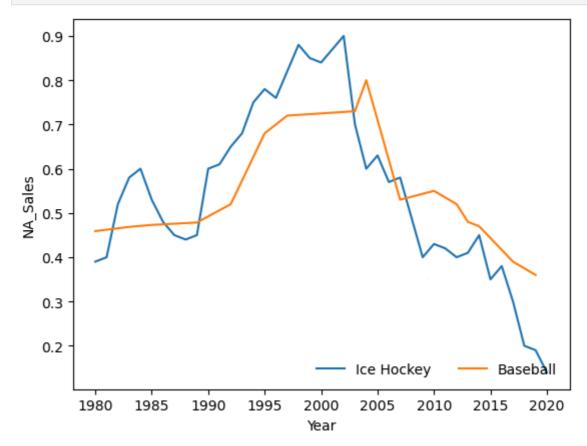
- The location in the form of string.
- The location in the form of coordinates.

\ What if we want to add other stylings to legends?

- Specify the number of rows/cols
 - Uses parameter ncols for this.
 - The number of rows are decided automatically.
- Decide if we want the box of legends to be displayed
 - Use the bool parameter frameon.

...and so on.

```
In [16]: sns.lineplot(x='Year', y='NA_Sales', data=ih, label='Ice Hockey')
sns.lineplot(x='Year', y='NA_Sales', data=baseball, label='Baseball')
plt.legend(loc='lower right', ncol=2, frameon=False)
plt.show()
```



Now say we want to highlight a point on our curve.

How can we highlight the maximum "Ice Hockey" sales across all years?

Let's first find this point.

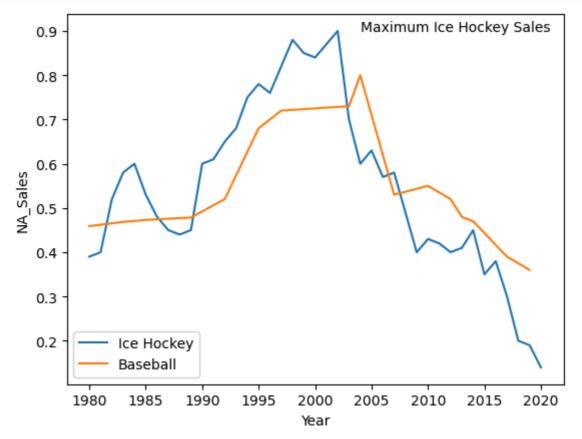
```
In [17]: print(max(ih['NA_Sales']))
0.9
```

If we observe, this point lies in the year 2004-2005.

Now we need to add text to this point (2004,0.9)

How can we add text to a point in a figure?

```
In [18]:
sns.lineplot(x='Year', y='NA_Sales', data=ih, label='Ice Hockey')
sns.lineplot(x='Year', y='NA_Sales', data=baseball, label='Baseball')
plt.legend(loc='lower left')
plt.text(2004,max(ih['NA_Sales']), 'Maximum Ice Hockey Sales')
plt.show()
```

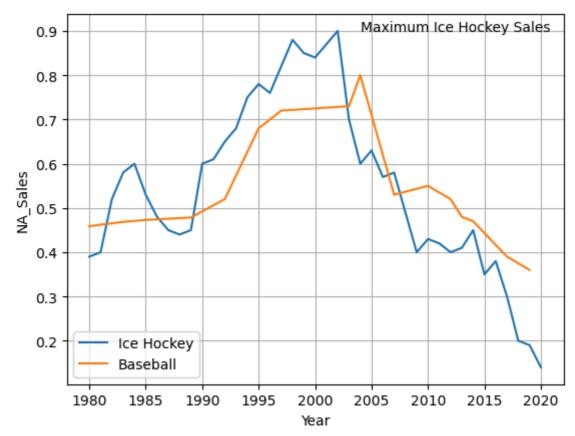


By using plt.text()

- Pass in the **x and y coordinates** where we want the text to appear.
- Pass in the text string.

We can also use plt.grid() to show the grid layout in the background.

```
In [19]: sns.lineplot(x='Year', y='NA_Sales', data=ih, label='Ice Hockey')
sns.lineplot(x='Year', y='NA_Sales', data=baseball, label='Baseball')
plt.legend(loc='lower left')
plt.text(2004, max(ih['NA_Sales']), 'Maximum Ice Hockey Sales')
plt.grid()
plt.show()
```



Note: We can pass in parameters inside <code>plt.grid()</code> to control its density, colour of grid lines, etc.

You can look it up later on how to customize the grid - https://matplotlib.org/stable/api/_as_gen/matplotlib.pyplot.grid.html

Scatter Plot

Suppose we want to find the relation between Rank and Sales of all games.

Are Rank and Sales positively or negatively correlated?

In this case, unlike line plot, there maybe multiple points in y-axis for each point in x-axis.

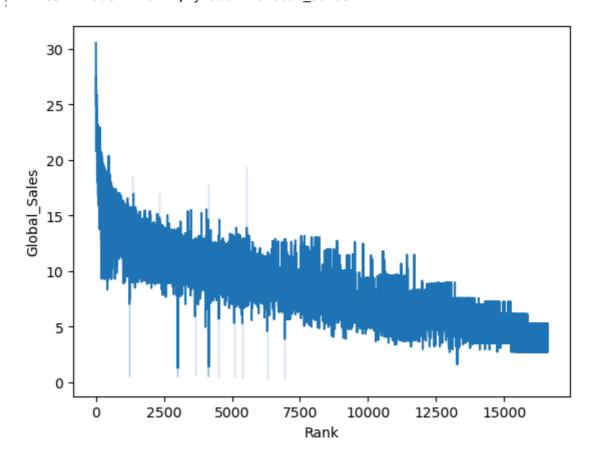
In [20]: data.head()

Out[20]:		Rank	Name	Platform	Year	Genre	Publisher	NA_Sales	EU_Sales	JP_Sa
	0	2061	1942	NES	1985.0	Shooter	Capcom	4.569217	3.033887	3.439
	1	9137	¡Shin Chan Flipa en colores!	DS	2007.0	Platform	505 Games	2.076955	1.493442	3.033
	2	14279	.hack: Sekai no Mukou ni + Versus	PS3	2012.0	Action	Namco Bandai Games	1.145709	1.762339	1.493
	3	8359	.hack//G.U. Vol.1//Rebirth	PS2	2006.0	Role- Playing	Namco Bandai Games	2.031986	1.389856	3.2280
	4	7109	.hack//G.U. Vol.2//Reminisce	PS2	2006.0	Role- Playing	Namco Bandai Games	2.792725	2.592054	1.440,

How can we plot the relation between Rank and Global Sales?

Can we use lineplot? Let's try it out.

```
In [21]: sns.lineplot(data=data, x='Rank', y='Global_Sales')
Out[21]: <Axes: xlabel='Rank', ylabel='Global_Sales'>
```



The plot itself looks very messy and it's hard to find any patterns from it.

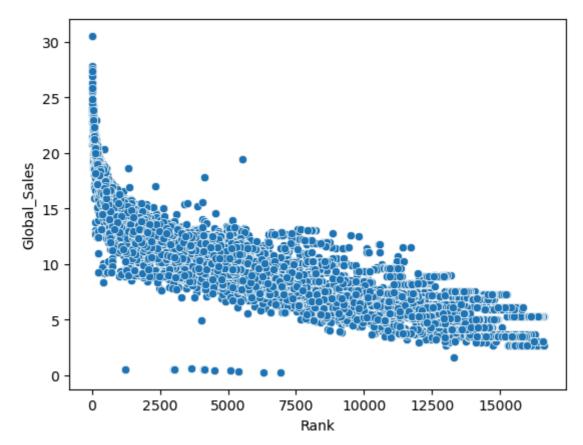
Is there any other way we can visualize this relation?

using scatterplot()

```
In [22]: sns.scatterplot(data=data, x='Rank', y='Global_Sales')
```

file:///Users/nikhil/Downloads/Viz2.html

Out[22]: <Axes: xlabel='Rank', ylabel='Global_Sales'>



Compared to lineplot, we are able to see the patterns and points more distinctly now! Notice,

- The two variables are negatively correlated with each other.
- With increase in ranks, the sales tend to go down, implying lower ranked games have higher sales overall!

Scatter plots help us visualize these relations and find any patterns in the data.

Sometimes, people also like to display the linear trend between two variables - **Regression Plot**.

If you notice, Genres, Publisher and Platform are categorical values.

Since we have a lot of categories of each of them, we will use top 3 of each to make our analysis easier.

```
In [23]: top3_pub = data['Publisher'].value_counts().index[:3]
    top3_gen = data['Genre'].value_counts().index[:3]
    top3_plat = data['Platform'].value_counts().index[:3]
    top3_data = data.loc[(data["Publisher"].isin(top3_pub)) & (data["Platform"]
    top3_data
```

0ut [23]:

Rank
Name
Platform
Year
Genre
Publisher
NA_Sales
EU_Sales
JP_Sales

.hack:
Sekai no
Mukou ni +
PS3 2012.0 Action
Bandai 1.145709 1.762339 1.4934

2	14279	.hack: Sekai no Mukou ni + Versus	PS3	2012.0	Action	Namco Bandai Games	1.145709	1.762339	1.4934
13	2742	[Prototype 2]	PS3	2012.0	Action	Activision	3.978349	3.727034	0.8488
16	1604	[Prototype]	PS3	2009.0	Action	Activision	4.569217	4.108402	1.1872
19	1741	007: Quantum of Solace	PS3	2008.0	Action	Activision	4.156030	4.346074	1.0879
21	4501	007: Quantum of Solace	PS2	2008.0	Action	Activision	3.228043	2.738800	2.5855
•••	•••			•••	•••				
16438	14938	Yes! Precure 5 Go Go Zenin Shu Go! Dream Festival	DS	2008.0	Action	Namco Bandai Games	1.087977	0.592445	1.0879
16479	10979	Young Justice: Legacy	PS3	2013.0	Action	Namco Bandai Games	2.186589	1.087977	3.4090
16601	11802	ZhuZhu Pets: Quest for Zhu	DS	2011.0	Misc	Activision	2.340740	1.525543	3.1038:
16636	9196	Zoobles! Spring to Life!	DS	2011.0	Misc	Activision	2.697415	1.087977	2.7607
16640	9816	Zubo	DS	2008.0	Misc	Electronic Arts	2.592054	1.493442	1.4934

617 rows × 11 columns

Categorical-Categorical

Earlier we saw how to work with continous-continuous pair of variables.

Now let's come to the second type of pair of i.e. Categorical-Categorical.

What questions comes to your mind when we say categorical-categorical pair?

Questions related to distribution of a category within another category.

- What is the distribution of genres for top-3 publishers?
- Which platforms do these top publishers use?

Which plot can we use to show distribution of one category with respect to another?

We can have can have multiple bars for each category

These multiple bars can be

- stacked together Stacked Countplot
- placed next to each other **Dodged Countplot**

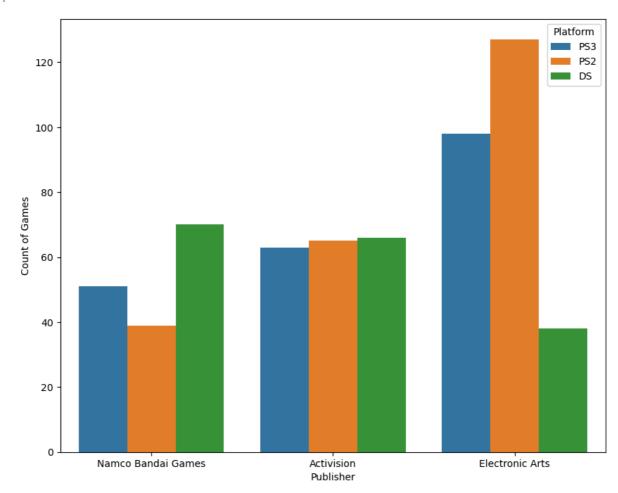
Dodged Countplot

How can we compare the top 3 platforms these publishers use?

• We can use a **dodged countplot** in this case.

```
In [24]: plt.figure(figsize=(10,8))
    sns.countplot(x='Publisher',hue='Platform',data=top3_data)
    plt.ylabel('Count of Games')
```

Out[24]: Text(0, 0.5, 'Count of Games')



What can we infer from the dodged countplot?

- EA releases PS2 games way more than any other publisher, or even platform!
- Activision has almost the same count of games for all 3 platforms.
- EA is leading in PS3 and PS2, but Namco leads when it comes to DS platform.

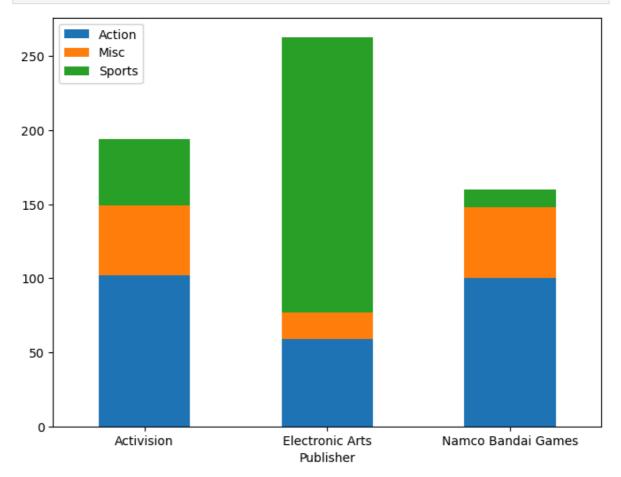
Stacked Countplot

How can we visualize the distribution of genres for top-3 publishers?

file:///Users/nikhil/Downloads/Viz2.html

• We can use a **stacked countplot** for this.

```
In [25]: df_stacked_plot = pd.crosstab(index=top3_data['Publisher'], columns=top3_data
    df_stacked_plot.plot(kind='bar', stacked=True, figsize=(8, 6))
    plt.xticks(rotation=0)
    plt.legend(loc='upper left')
    plt.show()
```



But stacked countplots can be misleading.

Some may find it difficult to understand if it starts from baseline or from on top of the bottom area.

How do we decide between a Stacked countplot and Dodged countlot?

- Stacked countplots are a good way to represent totals.
- Dodged countplots helps us to comapare values between various categories, and within the category itself too.

Continous-Categorical

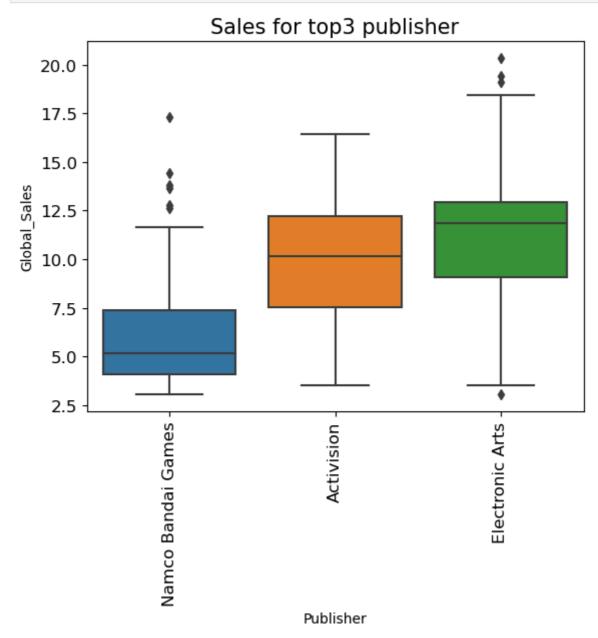
What kind of questions we may have regarding a continuous-categorical pair?

- We might to want calculate some numbers category-wise.
 - What is the average sales for every genre?
- We might be interested in checking the distribution of the data category-wise.
 - What is the distribution of sales for the top3 publishers?

What kind of plot can we make for every category?

• Either KDE plot or Boxplot per category.

```
In [27]: sns.boxplot(x='Publisher', y='Global_Sales', data=top3_data)
  plt.xticks(rotation=90, fontsize=12)
  plt.yticks(fontsize=12)
  plt.title('Sales for top3 publisher', fontsize=15)
  plt.show()
```



What can we infer from this plot?

- The overall sales of EA is higher, with a much larger spread than other publishers.
- Activision doesn't have many outliers.
 - If you notice, even though the spread is lesser than EA, the median is almost the same.

Bar Plot

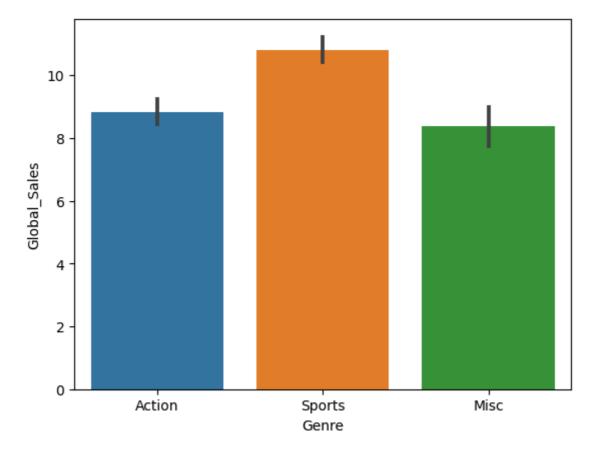
What if we want to compare the sales between the genres?

We have to use:

- Genre (categorical)
- Mean of global sales per genre (numerical)

How to visualize which genres bring higher average global sales?

In [28]: sns.barplot(data=top3_data, x="Genre", y="Global_Sales", estimator=np.mean)
Out[28]: <Axes: xlabel='Genre', ylabel='Global_Sales'>



What can we infer from this plot?

- If you remember, we had earlier seen EA had a larger market share of sales.
- Along with this fact, majority of games EA made was sports.
- This ultimately proves the fact that Sports has a high market share in the industry, as shown in the bar chart.

Subplots

So far we have **shown only 1 plot** using plt.show().

Let's say we want to plot the trend of NA and every other region separately in a single figure.

How can we plot multiple smaller plots at the same time?

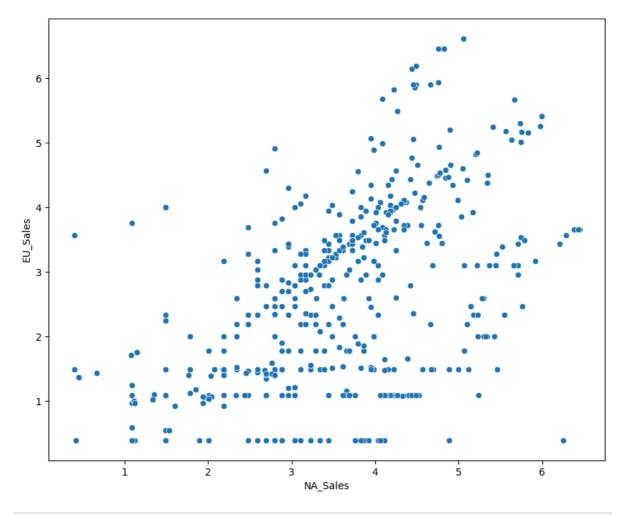
We will use **subplots**, i.e., **divide the figure into smaller plots**.

• We will be using plt.subplots()

- It mainly takes 2 arguments -
 - 1. No. of rows we want to divide our figure into.
 - 2. No. of columns we want to divide our figure into.
- It returns
 - Figure
 - Numpy matrix of subplots

```
In [29]: fig = plt.figure(figsize=(10,8))
    sns.scatterplot(x=top3_data['NA_Sales'], y=top3_data['EU_Sales'])
    fig.suptitle('Main title')
    plt.show()
```

Main title



```
In [30]: fig = plt.figure(figsize=(10,10))

plt.subplot(2, 3, 1)
    sns.scatterplot(x='NA_Sales', y='EU_Sales', data=top3_data)

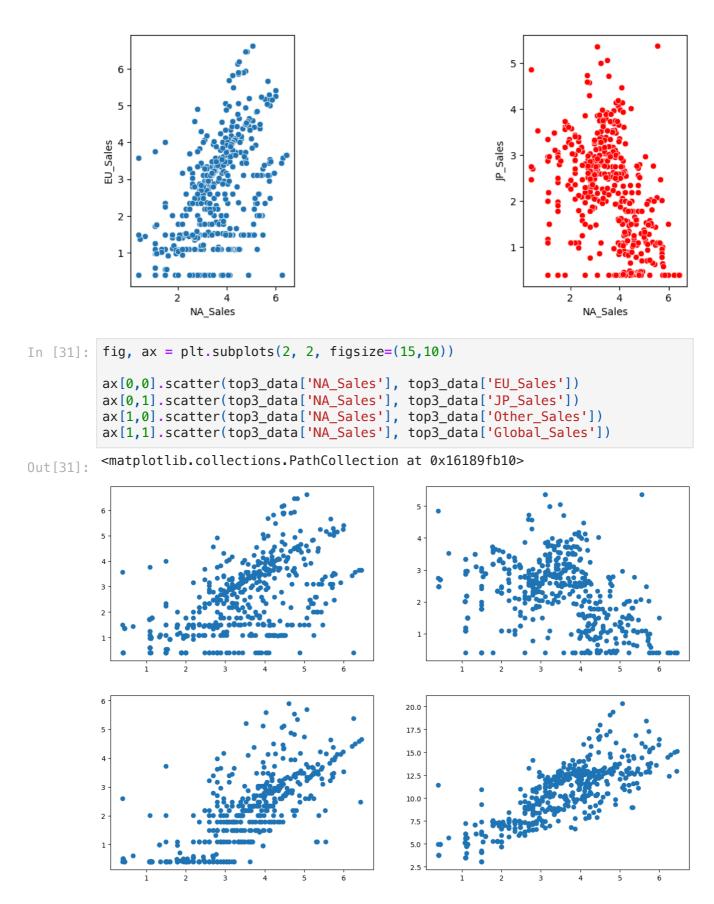
plt.subplot(2, 3, 3)
    sns.scatterplot(x='NA_Sales', y='JP_Sales', data=top3_data, color='red')

fig.suptitle('Main title')
```

Out[30]: Text(0.5, 0.98, 'Main title')

1/23/25, 3:51 PM

Main title



Notice that we are using 2 numbers during each plotting.

Think of subplots as a 2x2 grids, with the two numbers denoting x,y / row, column coordinate of each subplot.

What is this ax parameter exactly?

```
In [32]: print(ax)
    [[<Axes: > <Axes: >]
        [<Axes: > <Axes: >]]
```

We are plotting each plot on a single axes object.

It's a 2x2 matrix of multiple axes objects.

Hence, we are using a 2D notation to access each grid/axes object of the subplot.

Instead of accesing the individual axes using ax[0, 0], ax[1, 0], there is another method we can use too.

```
In [33]: plt.figure(figsize=(12,10)).suptitle("NA Sales vs regions",fontsize=20)
# using a 2x3 subplot
plt.subplot(2, 3, 1)
sns.scatterplot(x='NA_Sales', y='EU_Sales', data=top3_data)

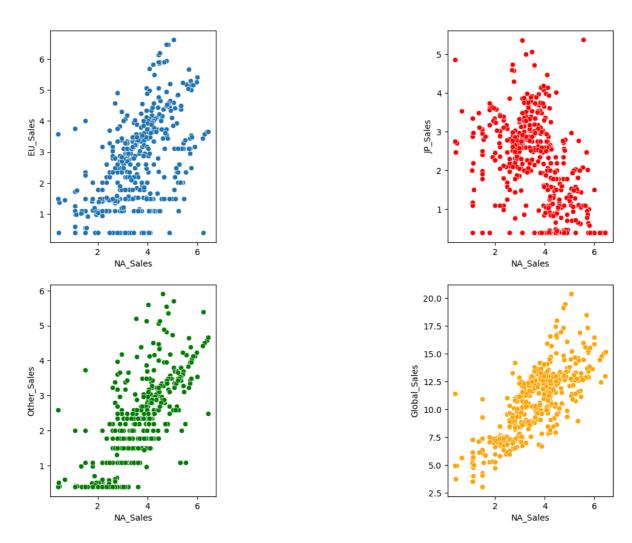
plt.subplot(2, 3, 3)
sns.scatterplot(x='NA_Sales', y='JP_Sales', data=top3_data, color='red')

plt.subplot(2, 3, 4)
sns.scatterplot(x='NA_Sales', y='Other_Sales', data=top3_data, color='green

plt.subplot(2, 3, 6)
sns.scatterplot(x='NA_Sales', y='Global_Sales', data=top3_data, color='orang
plt.show()
```

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NA Sales vs regions



Suptitle adds a title to the whole figure.

We need to observe a few things here -

- 1. The 3rd paramter defines the position of the plot.
- 2. The position/numbering starts from 1.
- 3. It goes on row-wise from start of row to its finish.
- 4. Empty subplots don't show any axes.

But how do we know which plot belongs to which category?

Basically the context of each plot.

We can use title, x/y label and every other functionality for the subplots too.

```
In [34]: plt.figure(figsize=(12,10)).suptitle("NA Sales vs regions",fontsize=20)

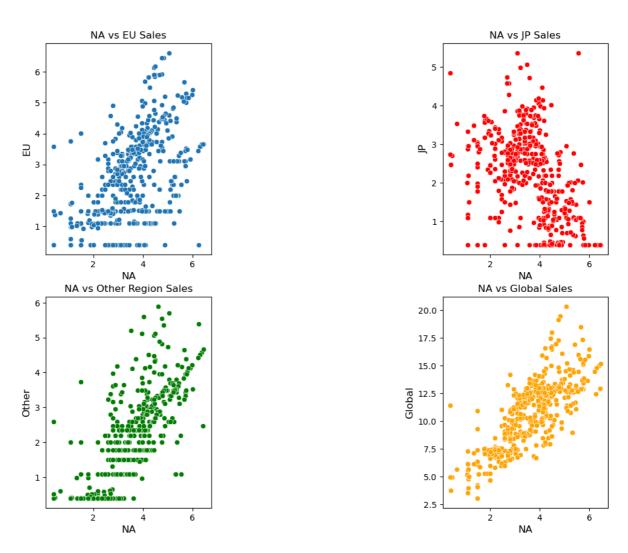
# using a 2x3 subplot
plt.subplot(2, 3, 1)
sns.scatterplot(x='NA_Sales', y='EU_Sales', data=top3_data)
plt.title('NA vs EU Sales', fontsize=12)
plt.xlabel('NA', fontsize=12)
plt.ylabel('EU', fontsize=12)
plt.subplot(2, 3, 3)
```

```
sns.scatterplot(x='NA_Sales', y='JP_Sales', data=top3_data, color='red')
plt.title('NA vs JP Sales', fontsize=12)
plt.xlabel('NA', fontsize=12)
plt.ylabel('JP', fontsize=12)

plt.subplot(2, 3, 4)
sns.scatterplot(x='NA_Sales', y='Other_Sales', data=top3_data, color='green
plt.title('NA vs Other Region Sales', fontsize=12)
plt.xlabel('NA', fontsize=12)
plt.ylabel('Other', fontsize=12)

plt.subplot(2, 3, 6)
sns.scatterplot(x='NA_Sales', y='Global_Sales', data=top3_data, color='orang
plt.title('NA vs Global Sales', fontsize=12)
plt.xlabel('NA', fontsize=12)
plt.ylabel('Global', fontsize=12)
```

NA Sales vs regions



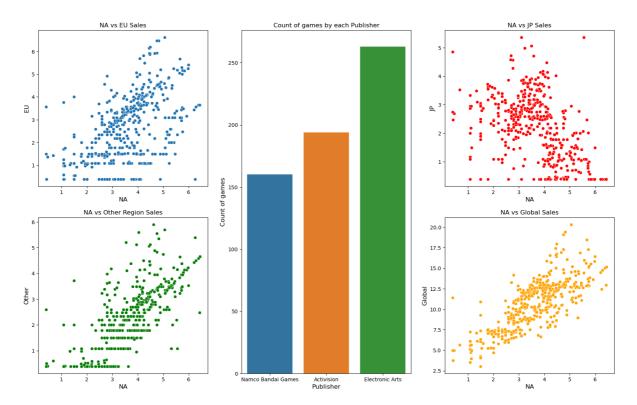
What if we want to span a plot across the full length of the plot?

- Think of this in terms of a grid.
- Currently we are dividing our plot into 2 rows and 3 columns.
- But we want our plot to be across the middle column, with **grids 2 and 5**.
- This can be said as a single column.

So this problem can be simplified to plotting the plot across **second column in a 1 row 3 column subplot**.

```
In [35]: plt.figure(figsize=(20,12)).suptitle("Video Games Sales Dashboard",fontsize=
         # using a 2x3 subplot
         plt.subplot(2, 3, 1)
         sns.scatterplot(x='NA_Sales', y='EU_Sales', data=top3_data)
         plt.title('NA vs EU Sales', fontsize=12)
         plt.xlabel('NA', fontsize=12)
         plt.ylabel('EU', fontsize=12)
         plt.subplot(2, 3, 3)
         sns.scatterplot(x='NA_Sales', y='JP_Sales', data=top3_data, color='red')
         plt.title('NA vs JP Sales', fontsize=12)
         plt.xlabel('NA', fontsize=12)
         plt.ylabel('JP', fontsize=12)
         # Countplot of publishers
         plt.subplot(1,3,2)
         sns.countplot(x='Publisher', data=top3_data)
         plt.title('Count of games by each Publisher', fontsize=12)
         plt.xlabel('Publisher', fontsize=12)
         plt.ylabel('Count of games', fontsize=12)
         plt.subplot(2, 3, 4)
         sns.scatterplot(x='NA\_Sales', y='0ther\_Sales', data=top3\_data, color='green')
         plt.title('NA vs Other Region Sales', fontsize=12)
         plt.xlabel('NA', fontsize=12)
         plt.ylabel('Other', fontsize=12)
         plt.subplot(2, 3, 6)
         sns.scatterplot(x='NA_Sales', y='Global_Sales', data=top3_data, color='orang
         plt.title('NA vs Global Sales', fontsize=12)
         plt.xlabel('NA', fontsize=12)
         plt.ylabel('Global', fontsize=12)
         plt.show()
```

Video Games Sales Dashboard



Question-1

Q. For the company "Toyota", we want to find which type of vehicle has made the maximum sales.

Which plot we will prefer to use here?

- a. Bar Plot
- b. Pie Chart
- c. Boxplot
- d. Line Plot

Answer: Bar Plot

Explanation:

A bar plot is ideal for comparing the quantities of different categories or groups, making it suitable for visualizing sales data across different types of vehicles. Each type of vehicle can be represented on the x-axis, while the sales quantity (or revenue) can be represented on the y-axis. This allows for a clear comparison of sales performance among different vehicle types, making it easy to identify which type has made the maximum sales.

Question-2

Q. Apple wanted to conduct an analysis and find the relationship between price and number of units sold for it's products.

Which of the following plots would you prefer?

- a. Scatter Plot
- b. Pie Chart
- c. Boxplot
- d. Line Plot

Answer: Scatter Plot

Explanation:

A scatter plot is specifically designed for visualizing the relationship between two continuous variables. In this case, you can plot the price of the products on the x-axis and the corresponding number of units sold on the y-axis. Each data point represents a product, with its price and the corresponding number of units sold. By examining the distribution of points on the plot, you can discern any patterns or trends in the relationship between price and sales volume for Apple products.

In []: