Matplotlib: 

Basic:

Title:

plt.title("Critic vs User Score", fontsize=14, fontweight='bold')

Labels:

plt.xlabel("Critic Score (0-10)", fontsize=12)

plt.ylabel("User Score (0-10)", fontsize=12)

Axis size:

plt.xlim(0, 10)

plt.ylim(0, 10)

Legend:

plt.legend(title="Game platform", loc="best")

loc: best-upper right-upper left-lower left-lower right-right-center left-center right-lower center-upper center.

title is hue default

Annotate:

plt.annotate("Top Rated", xy=(6.7,7.4), xytext=(4,9),

arrowprops=dict(facecolor='black', shrink=0))

xy: meaning the location on the plot where **x = 6.7** and **y = 7.4**.

xytext: meaning the location on the plot where the annotation text is placed.

Arrowprops-shrink: 0: long , 0.05: middle , 0.2: little.

Grid:

plt.grid(True, linestyle='-', alpha=1)

Background:

plt.gca().set_facecolor("lightblue")

Plot Line :

x_vals = np.linspace(0, 10)

plt.plot(x_vals, x_vals, color='red', linestyle='-', linewidth=2)

linspace(start,stop).

plt.plot(x,y).

Vlines:

plt.vlines(x=df['Critic_Score2'], ymin=0, ymax=df['User_Score'],

colors='gray', linestyles='-', linewidth=0.1)

x: plot line in Critic_score location from 0-10axis

ymin: start location .

ymax: finish location from 0-10 axis.

Hlines:

plt.hlines(y=df['User_Score'], xmin=0,
xmax=df['Critic_Score2'], colors='gray', linestyles='-', linewidth=0.1)

Xticks:

plt.xticks(rotation=45,fontsize=12)

Yticks:

plt.yticks(rotation=45,fontsize=12)

Figsize:

plt.figure(figsize=(1,100),dpi=100) : you have put it before plot

Libraries:

import numpy as np

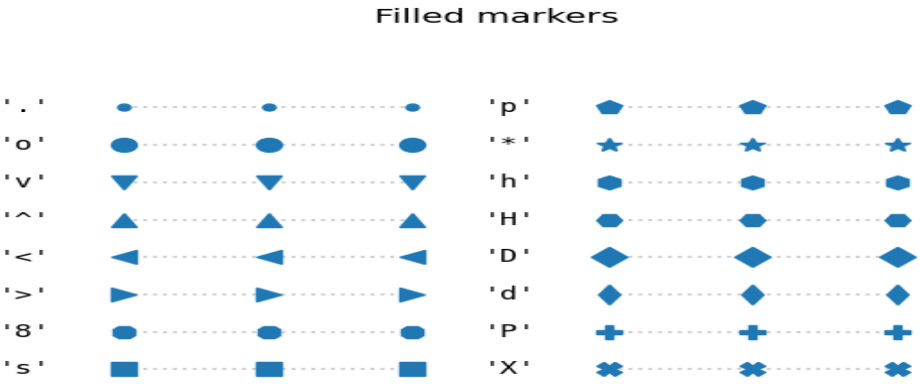
import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

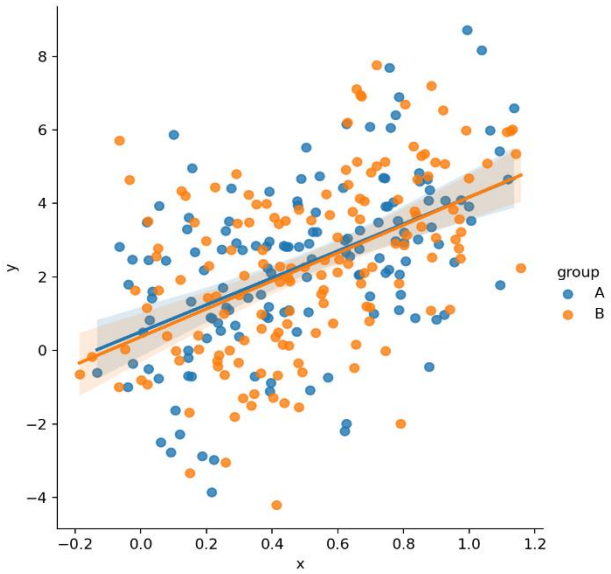
from mpl_toolkits.mplot3d import Axes3D

import mplfinance as mpf

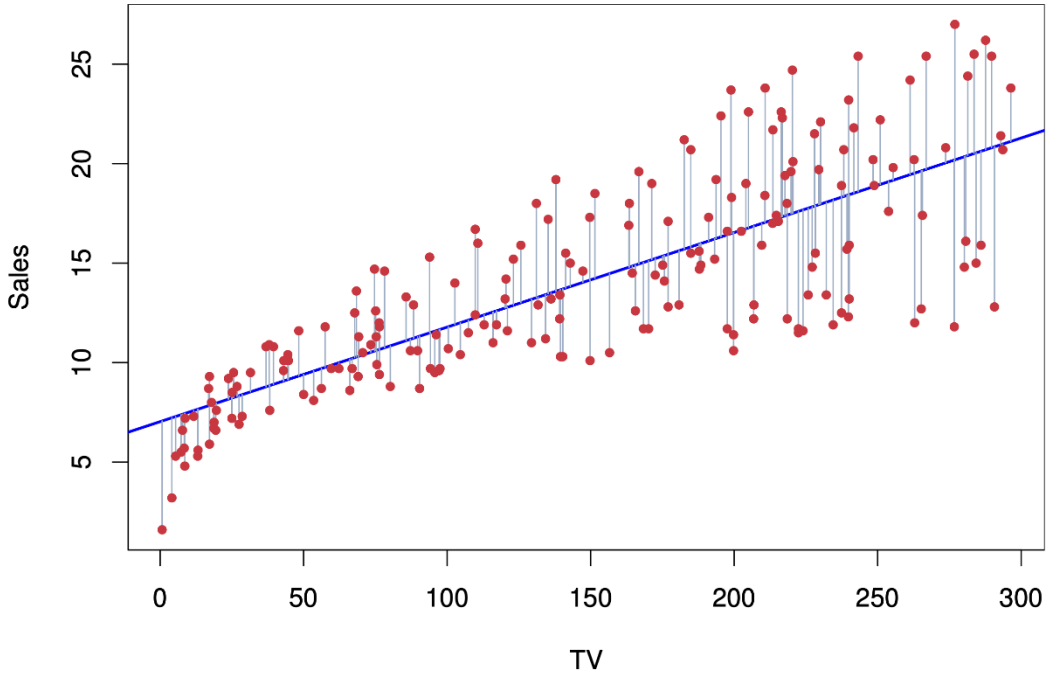


Scatter Plot:

```
sns.scatterplot(
    data=df,
    x='Critic_Score2',
    y='User_Score',
    marker='o',
    hue='Year_of_Release',
    alpha=1,
    legend='full')
```

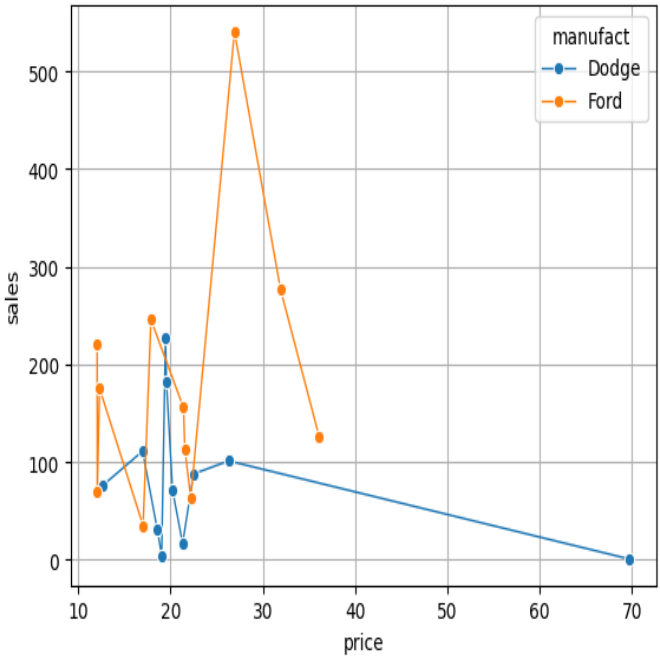


hue: hue means grouping by category.you can edit it with Legend.



Line Plot:

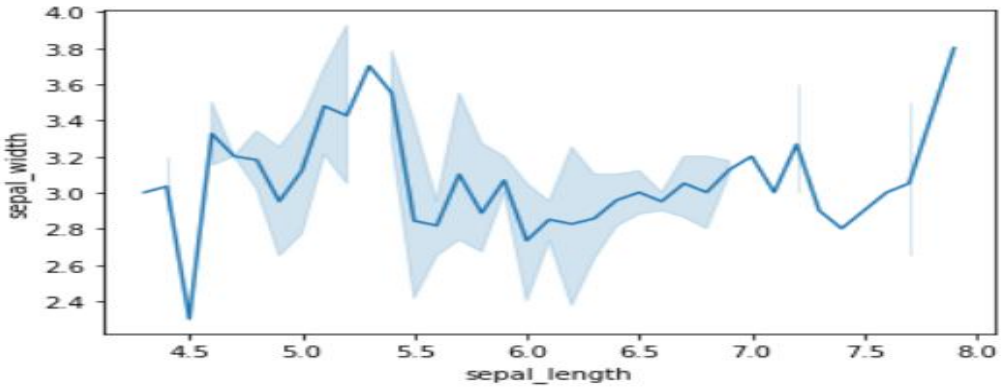
```
sns.lineplot(
    data=selcar,
    x='price',
    y='sales',
    hue='manufact',
    estimator=np.mean,
    ci=95,
    sort=True,
    linewidth=1,
    linestyle='-',
    marker='o'
)
```



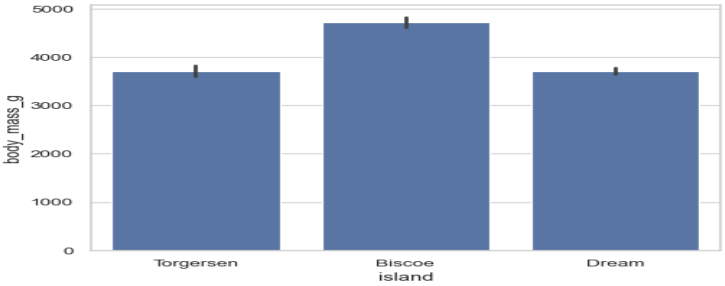
sort=False: Seaborn will plot the data in the order it appears in your DataFrame. If your x-values are not ordered, the line may zig-zag or look messy.

sort=True: Seaborn sorts the x-values before plotting, ensuring the line moves smoothly from left to right. This is especially useful when your data is not already sorted by the x-axis variable.

estimator=np.mean: then our data is some float it convert it to int to plot better.



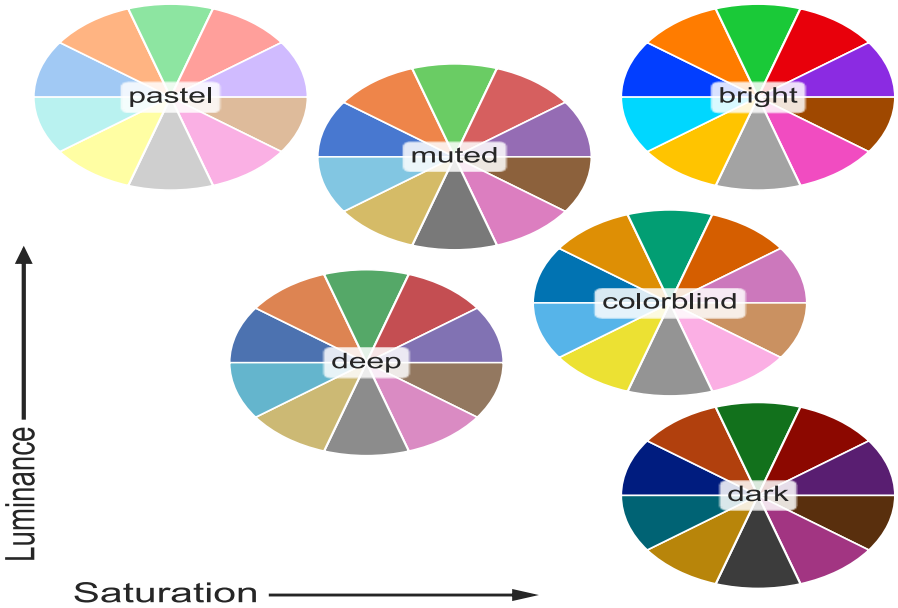
Bar Plot:




```
sns.barplot(
    data=selcar,
    x='model',
    y='sales',
    palette='bright',
    #order=['cat1','cat2','cat3']
    orient='v',
    width=0.8
)
```

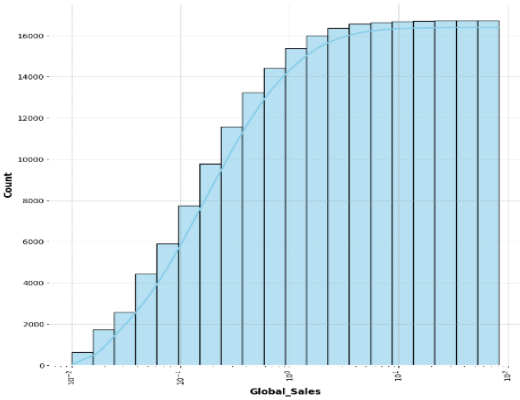
Orient: v = Vertical , h = horizontal.

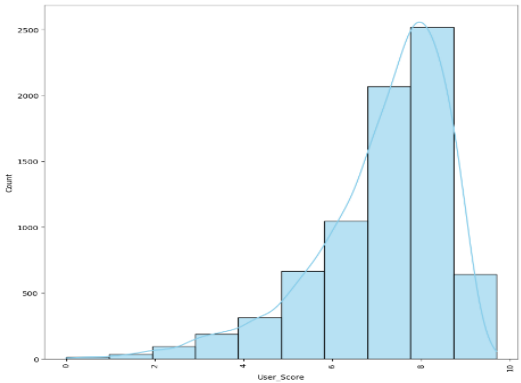
Order: order by cat from first to finally cat.



Histogram:

```
sns.histplot(
    dff['User_Score'],
    bins=100,
    kde=True,
    stat='count',
    color='skyblue',
    alpha=0.6,
    cumulative=False,
    log_scale=False
)
```





Cumulative: False :shows frequency in each bin normally.

Cumulative: True: shows cumulative frequency each bin adds up all previous bins.

Count stat: shows number of data points.

probability / percent stat: shows chance or percentage.

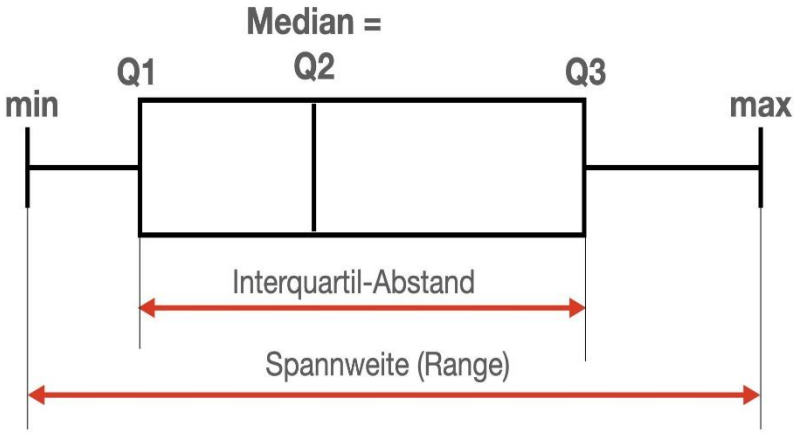
density stat: compares with continuous distributions or KDE.

KDE: Density diagram.

Log_scale: Logarithmic scale.

Box Plot:

```
sns.boxplot(
    data=car,
    x='manufact',
    y='sales',
    showfliers=True,
    notch=False,
    width=0.6,
    orient='v'
)
```



notch: (confidence interval for median).

Showfliers: Hide or show outliers.

Q1 → (Q1 / 25th percentile(25%))

Q3 → (Q3 / 75th percentile(75%))

Median → (Q2 / 50th percentile(50%))

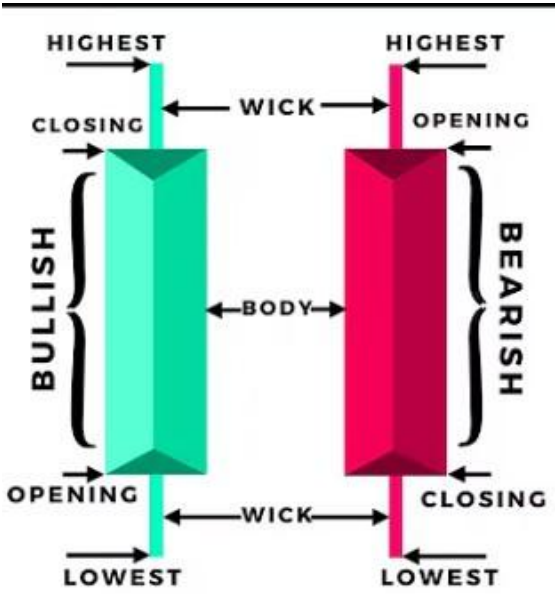
Max → Q3 + 1.5*IQR

Min →Q1 - 1.5*IQR

IQR → Q3 - Q1

You can clean data set outliers with datas that are out of max and min.

Candlestick:




```
mpf.plot(
   dfb,
   type='candle',
   style='charles',
   #volume=True,
   title='Bitcoin Candlestick',
   ylabel='Price ($)',
   ylabel_lower='Volume'
)
```

Index: you have to set inex to date column :

dfb['date'] = pd.to_datetime(dfb['date'], dayfirst=True)

dfb.set_index('date', inplace=True)

Dataset: must contain Open-Close-Low-High with just these names (if columns named like open_store or ... you have to rename them to right name) and,

Volume is Optional to use.(it`s name is important to use in dataset).

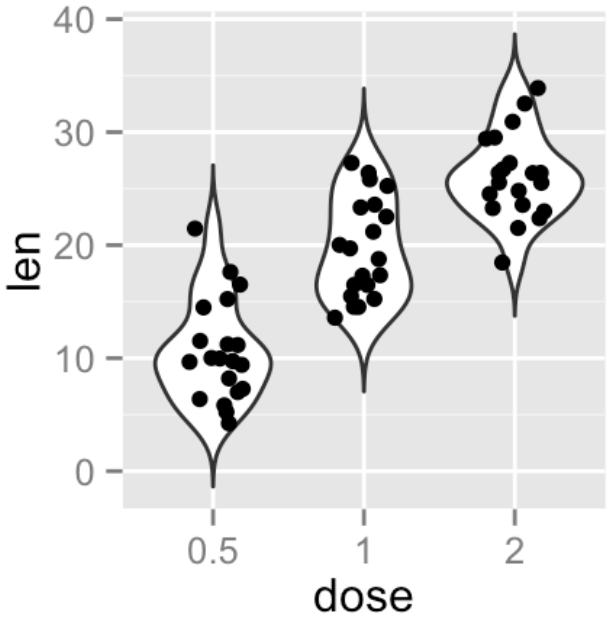
Volume: shows how many units of an asset were traded in a given time period and indicates the strength or validity of a price trend.

Rename:

dfb.rename(columns={'vol':'Volume'}, inplace=True)

Violin Plot:

```
sns.violinplot(
    data=car,
    x='manufact',
    y='sales',
    split=True,
    inner='quartile',
    scale='width',
    bw=0.2
    orient='v'
)
```



split: divide into two parts.

inner='box': Shows a small boxplot inside the violin.

inner='quartile': Shows only three horizontal lines.

inner='point': Shows only a single dot.

inner='stick': Shows many thin vertical lines.

inner=None: Shows nothing inside.

scale='area': All violins have the same total area.

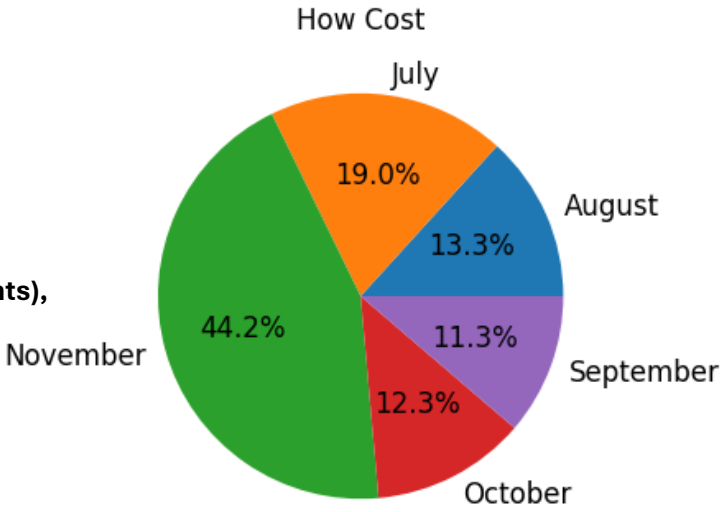
scale='count': Width depends on number of data points.

scale='width': All violins have the same maximum width.

bw: Bandwidth smoothing.

Pie:

```
counts = ex['Category'].value_counts()
plt.pie(
   counts,
   labels=counts.index,
   autopct='%1.1f%%',
   explode=[0.1]*len(counts),
   shadow=True,
   startangle=90
)
```



counts = ex['Category'].value_counts(): need to make them to category.

counts: categories names.

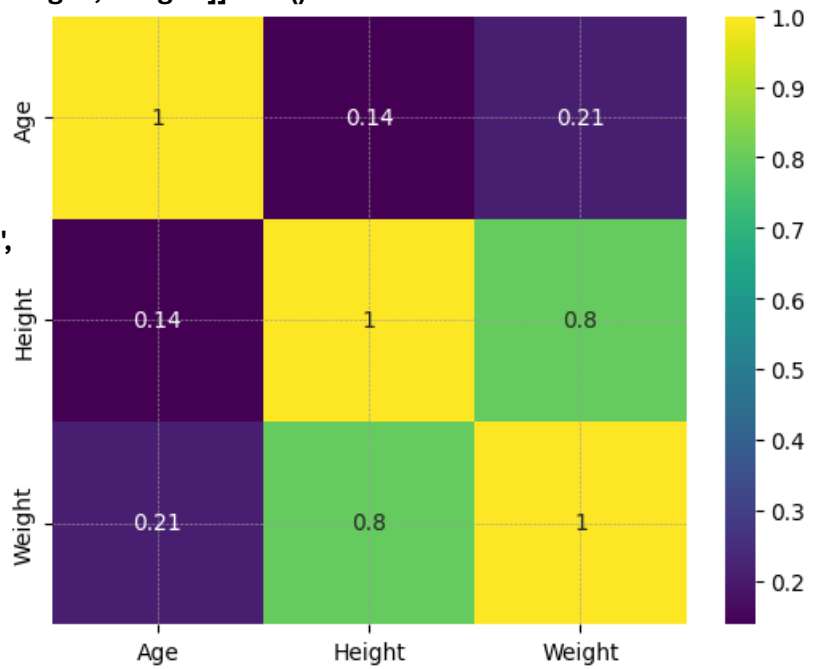
autopct='%1.1f%%': show percent .

explode=[0.1]*len(counts): distance amount.

startangle: angle amount.

Heatmap:

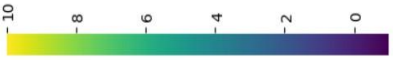
```
dfh = ath[['Age','Height','Weight']].corr()
sns.heatmap(
    dfh,
    annot=True,
    cmap='viridis',
    vmin=-1,
    vmax=1,
    linewidths=1,
    square=True,
    fmt='.2f'
)
```



annot: show percentages.

cmap: palletes color: viridis_coolwarm.

vmin_vmax: color max and min.



fmt: number format

Age and Age: **1.00** : Perfect correlation with itself

Age and Height: **0.14** : Weak positive correlation

Age and Weight: **0.2** : Weak positive correlation

Height and Height: **1.00** : Perfect correlation with itself

Height and Weight: **0.80** : Strong positive correlation

Weight and Weight: **1.00** : Perfect correlation with itself