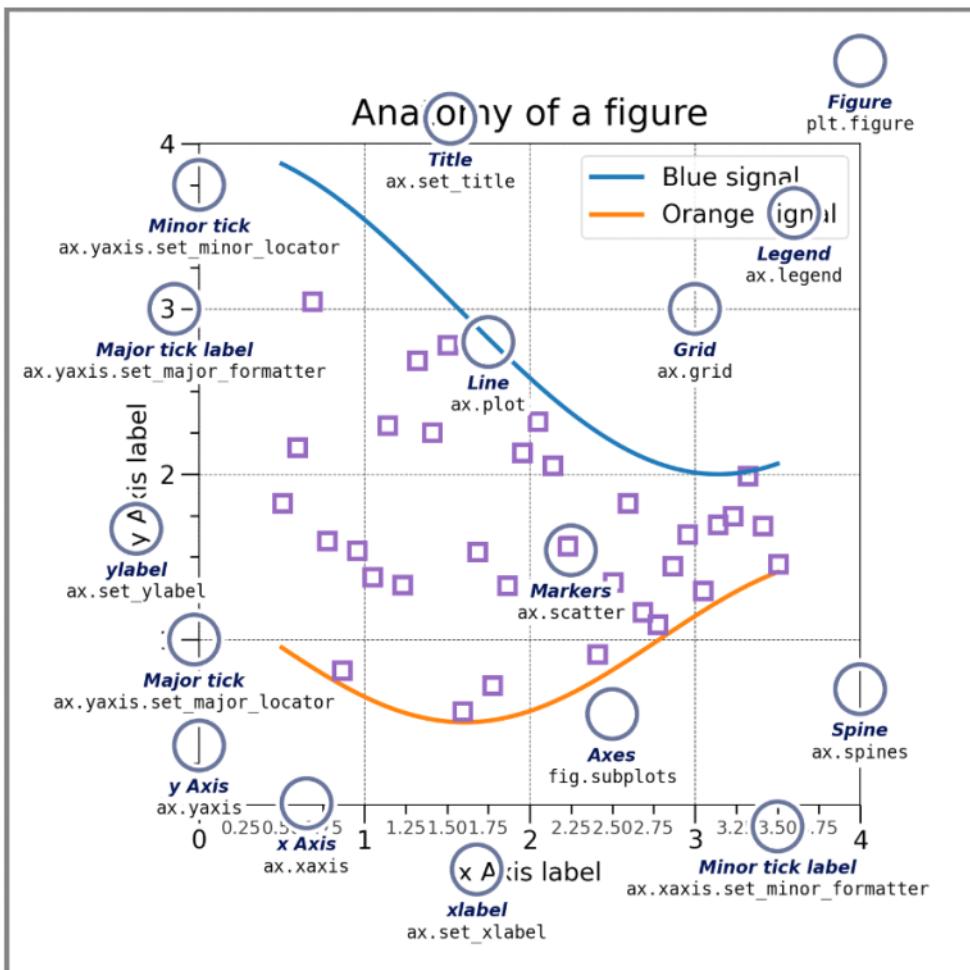


Matplotlib & Seaborn-2

Session Objectives:

- ✓ Understand what data visualization is and why it matters
- ✓ Use Matplotlib to plot different types of charts
- ✓ Customize plots with markers, colors, linewidth, and line styles
- ✓ Integrate Matplotlib with NumPy and Pandas
- ✓ Understand why Seaborn is important in visualization
- ✓ Recognize common Seaborn plot types



Bivariate Analysis

[Continuous VS Categorical]
 [Continuous VS Continuous]
 [Categorical Vs Categorical]

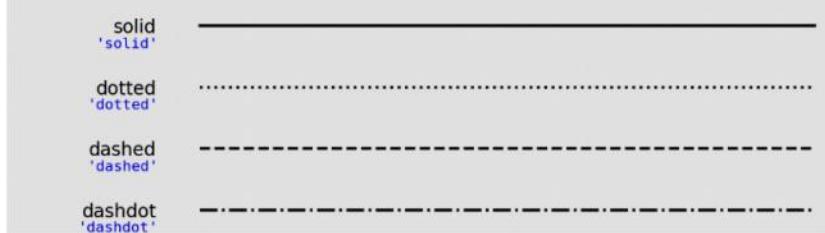
matplotlib.markers

Functions to handle markers; used by the marker functional

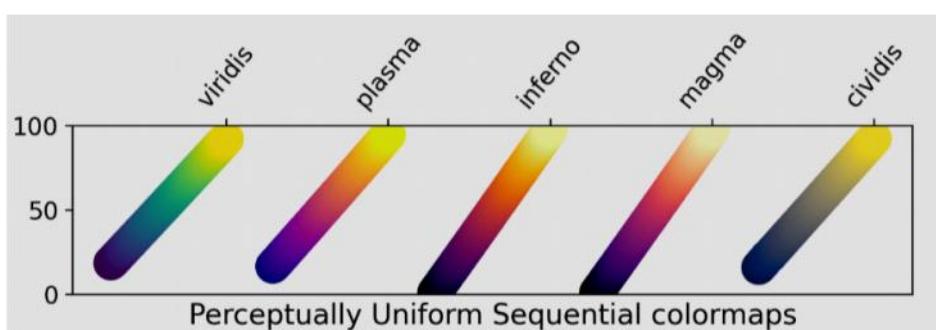
All possible markers are defined here:

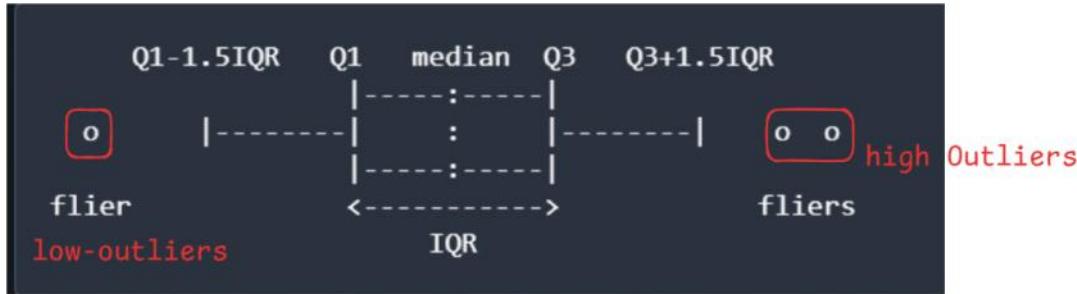
| marker | symbol | description |
|--------|--------|----------------|
| "." | ● | point |
| ■ | pixel | |
| "o" | ○ | circle |
| "v" | ▽ | triangle_down |
| "^" | △ | triangle_up |
| "<" | ◀ | triangle_left |
| ">" | ▶ | triangle_right |
| "1" | ▮ | tri_down |

Named line styles

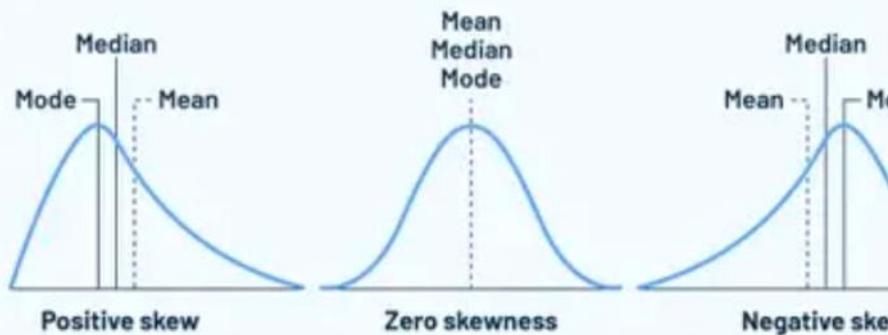


<https://matplotlib.org/stable/users/explain/colors/colormaps.html>





Types of Skewness

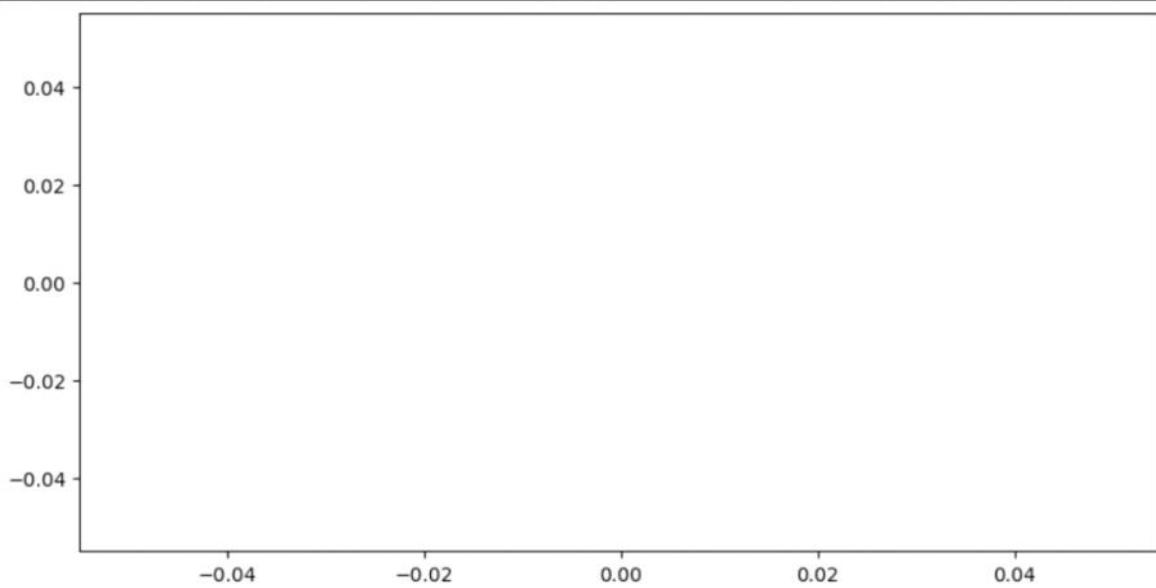


Note: Please complete all data cleaning steps before proceeding to the visualization stage.

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd

plt.figure(figsize = (10,5)) # (width, height)
plt.plot()

[]
```



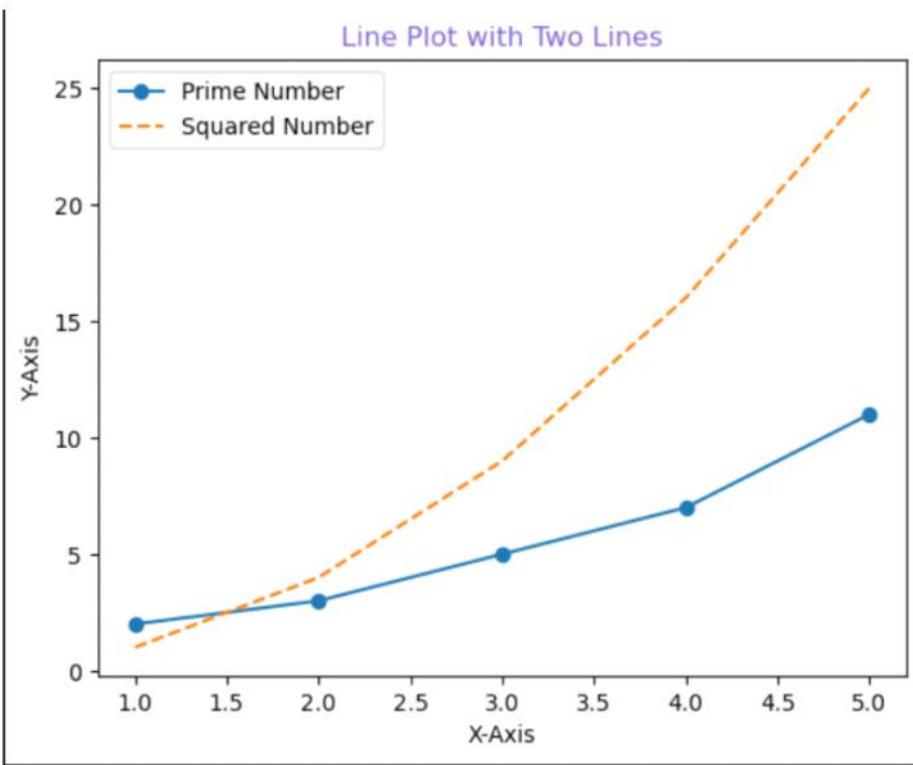
```

# Line Plot with Multiple Lines
x = [1,2,3,4,5]
y1 = [2,3,5,7,11] # Prime Numbers
y2 = [1,4,9,16,25] # Squared Number

plt.plot(x , y1, label = "Prime Number" , marker = 'o')
plt.plot(x , y2, label = "Squared Number" , linestyle = '--')

plt.title("Line Plot with Two Lines" , color = '#7E5ED1')
plt.xlabel('X-Axis')
plt.ylabel('Y-Axis')
plt.legend()
plt.show()

```

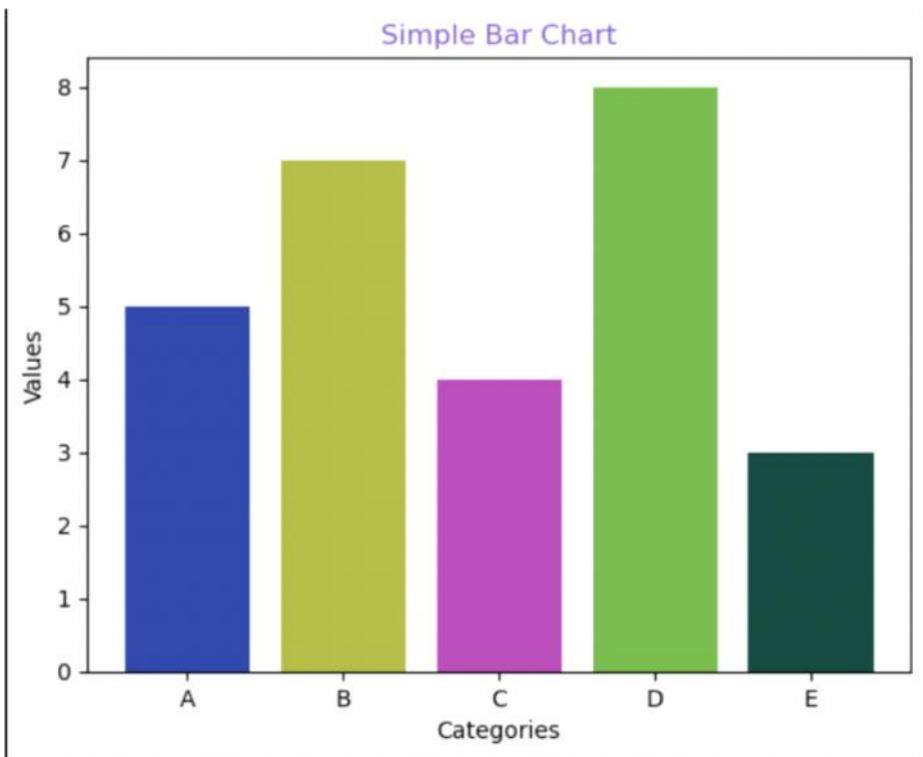


```

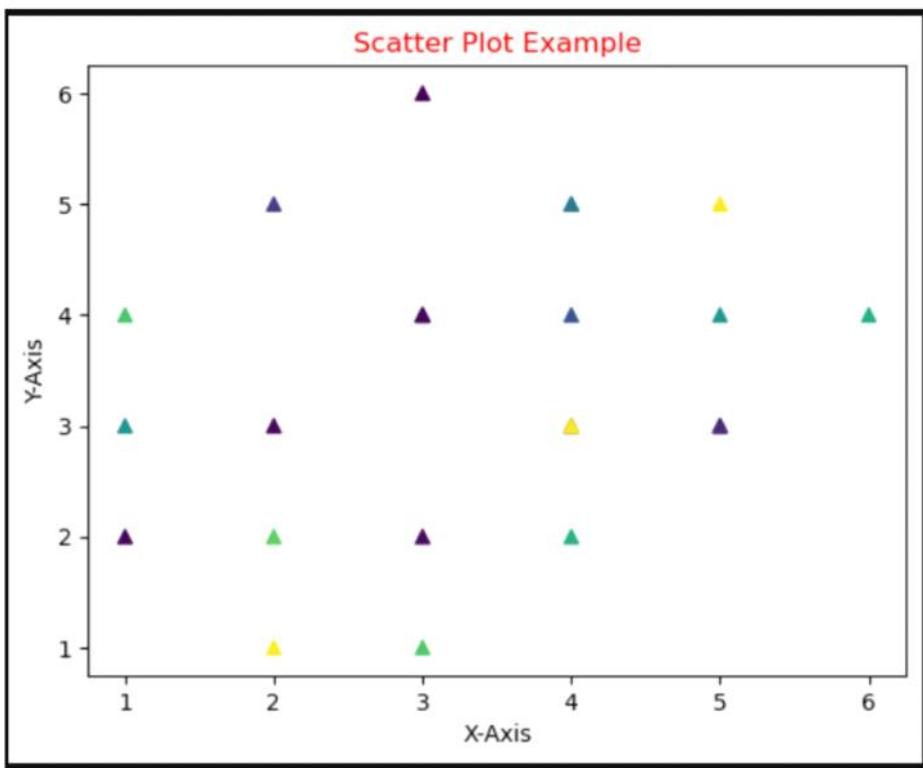
# Bar Chart [Categorical VS Continuous]
categories = ['A','B','C','D','E']
values = [5,7,4,8,3]
# colors = ['col1','col2',.....]
colors = ['#3147A8','#B0BA45','#B64DB8','#78B84D','#174A40']
plt.bar(categories, values, color = colors)

plt.title("Simple Bar Chart" , color = '#7E5ED1')
plt.xlabel('Categories')
plt.ylabel('Values')
plt.show()

```



```
# Scatter Plot [Continuous VS Continuous]
x = [5,4,3,3,4,5,3,2,1,2,3,4,5,6,4,4,3,2,1,2,3,4,5,1]
y = [4,5,6,4,4,3,2,1,2,3,4,5,5,4,3,3,4,5,3,2,1,2,3,4]
colors = np.random.rand(24)
plt.scatter(x,y,c=colors,cmap = 'viridis', marker = '^')
plt.title("Scatter Plot Example" , color = 'red')
plt.xlabel('X-Axis')
plt.ylabel('Y-Axis')
plt.show()
```



```
colors
```

```
array([0.50684192, 0.4916271 , 0.00892937, 0.4932071 , 0.25073179,
       0.04612673, 0.03462882, 0.97276215, 0.02366544, 0.0296304 ,
       0.10553535, 0.38785453, 0.95942208, 0.59700418, 0.05799306,
       0.95432777, 0.02186043, 0.1820911 , 0.50011489, 0.73553034,
       0.70239074, 0.61419957, 0.11905454, 0.69718322])
```

```
# Pie Chart [Composition Visuals]
```

```
sizes = [25,15,20,30,10]
```

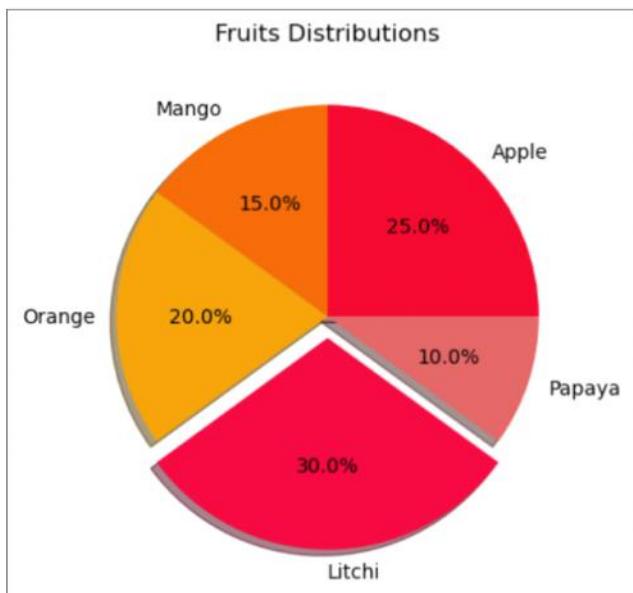
```
labels = ['Apple','Mango','Orange','Litchi','Papaya']
```

```
colors = ['#F00A30', '#F06A0A', '#F0A00A', '#F00A40', '#E06565']
```

```
plt.pie(sizes, labels = labels, colors = colors , explode = [0,0,0,0.1,0], autopct = '%1.1f%%', shadow=True)
```

```
plt.title("Fruits Distributions")
```

```
plt.show()
```



```
# Pie Chart [Composition Visuals]
```

```
sizes = [25,15,20,30,10]
```

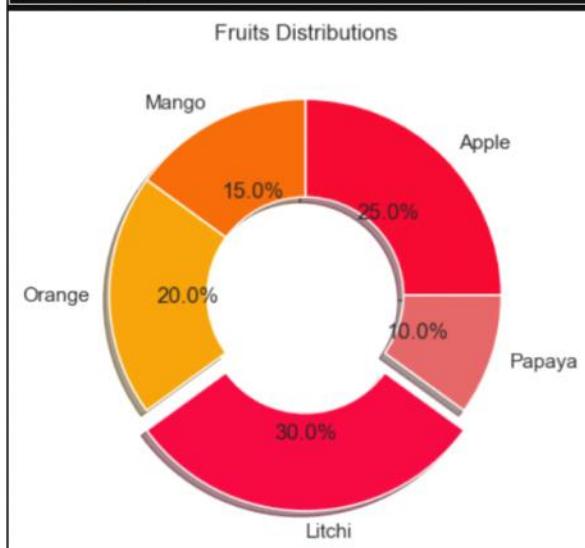
```
labels = ['Apple','Mango','Orange','Litchi','Papaya']
```

```
colors = ['#F00A30', '#F06A0A', '#F0A00A', '#F00A40', '#E06565']
```

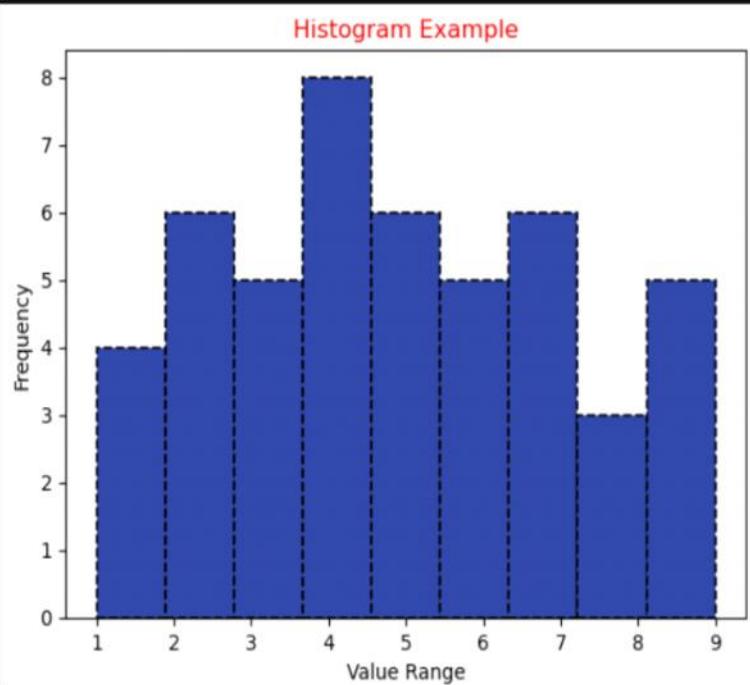
```
plt.pie(sizes, labels = labels, colors = colors , explode = [0,0,0,0.1,0], autopct = '%1.1f%%', shadow=True,
         wedgeprops=dict(width=0.5))
```

```
plt.title("Fruits Distributions")
```

```
plt.show()
```



```
# Histogram
data = [1,2,2,4,4,5,6,5,4,3,4,5,6,7,8,7,5,7,8,9,9,9,9,1,2,3,4,3,2,4,5,6,7,8,9,7,6,4,3,3,2,1,1,2,4,5,6,7]
plt.hist(data, bins=9, color = '#3147A8' , edgecolor = 'black', linewidth = 1.2 , linestyle = '--')
plt.title("Histogram Example" , color = 'red')
plt.xlabel('Value Range')
plt.ylabel('Frequency')
plt.show()
```

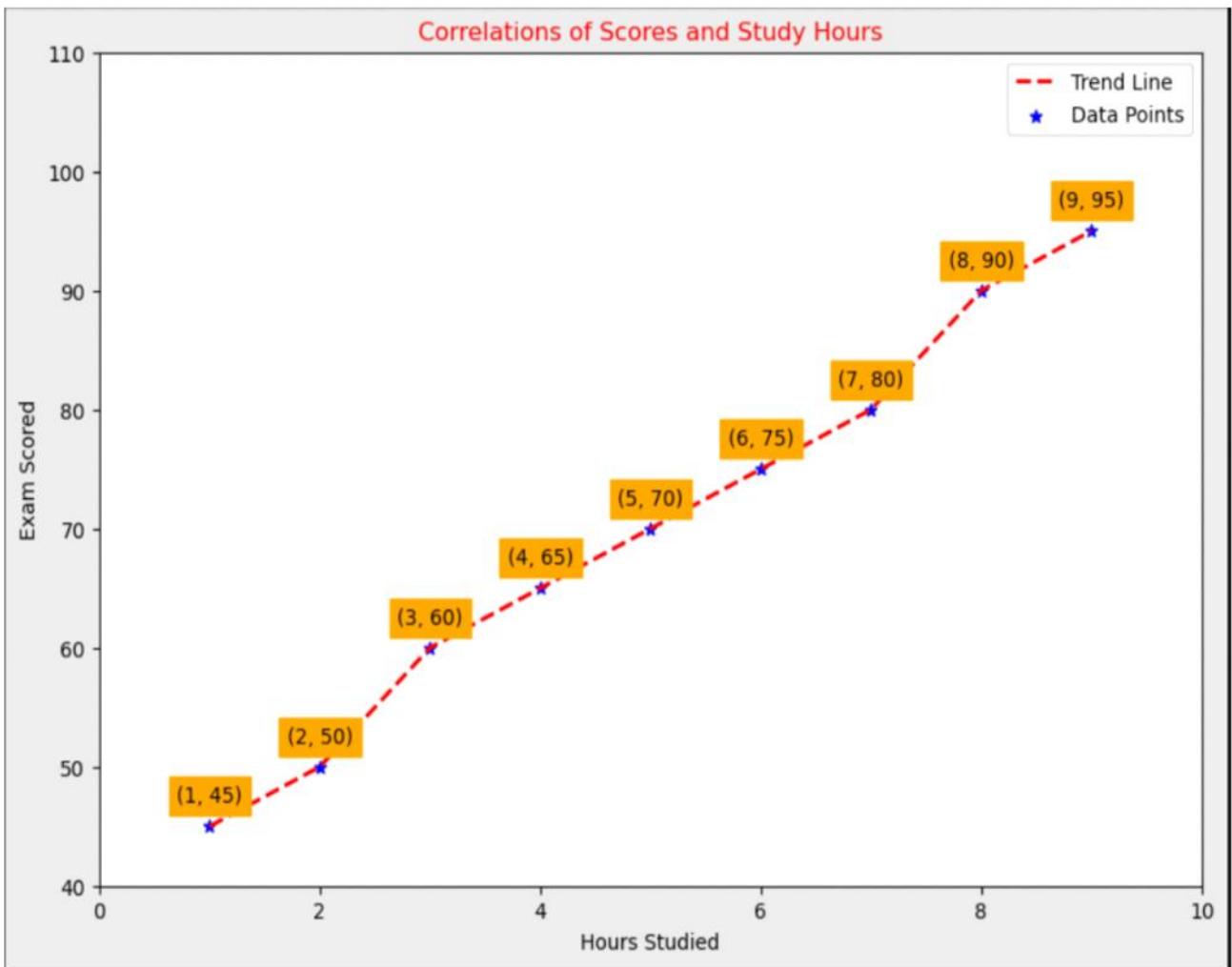


```
# Continuous Vs Continuous
hours = [1,2,3,4,5,6,7,8,9]
scores = [45,50,60,65,70,75,80,90,95]
plt.figure(figsize = (10,7) , facecolor = '#E8E8E8')

# Line + Scatter
plt.plot(hours , scores , color = 'red', linestyle = '--', linewidth = 2, label = 'Trend Line')
plt.scatter(hours, scores, color = 'blue' , marker = '*', label = 'Data Points')

# Add a Data Label
for x,y in zip(hours,scores):
    plt.text(x,y+2,f'{x,y}', ha = 'center', fontsize=10, color = 'black', backgroundcolor = 'orange')

plt.title('Correlations of Scores and Study Hours' , color = 'red')
plt.xlabel('Hours Studied')
plt.ylabel('Exam Scored')
plt.ylim(40,110)
plt.xlim(0,10)
plt.legend()
plt.show()
```

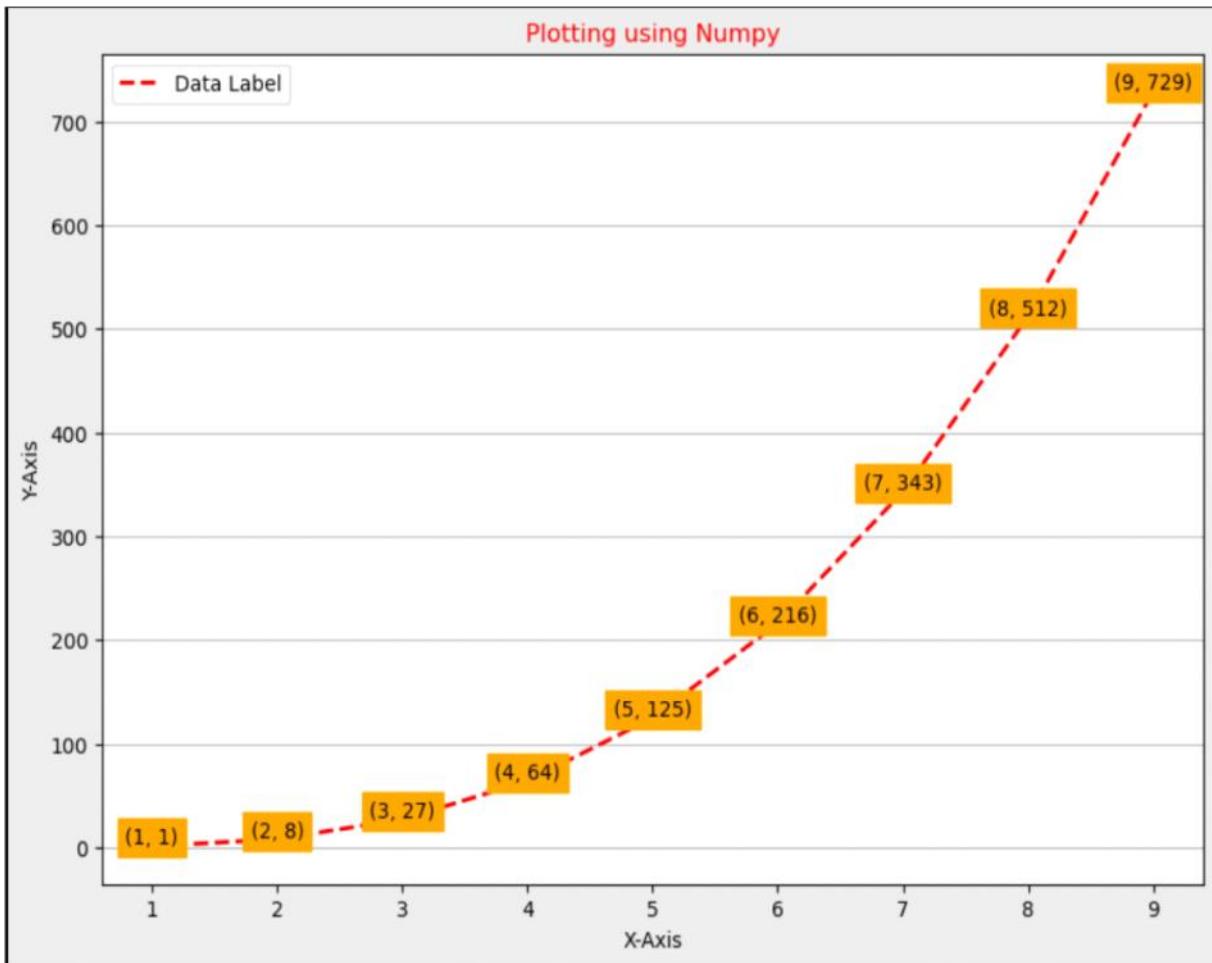


```
# Continuous Vs Continuous
import numpy as np
x = np.array([1,2,3,4,5,6,7,8,9])
y = x ** 3 # [1,8,27,64.....]
plt.figure(figsize = (10,7) , facecolor = '#E8E8E8')

# Line + Scatter
plt.plot(x , y , color = 'red', linestyle = '--', linewidth = 2, label = 'Data Label')

# Add a Data Label
for x,y in zip(x,y):
    plt.text(x,y+2,f'{x,y}', ha = 'center', fontsize=10, color = 'black', backgroundcolor = 'orange')

plt.title('Plotting using Numpy' , color = 'red')
plt.xlabel('X-Axis')
plt.ylabel('Y-Axis')
plt.grid(axis = 'y')
plt.legend()
plt.show()
```



```
# Customers Distribution ['genders']
gender_counts = customers['gender'].value_counts() #series
gender_counts

gender
Male      528
Female    472
Name: count, dtype: int64

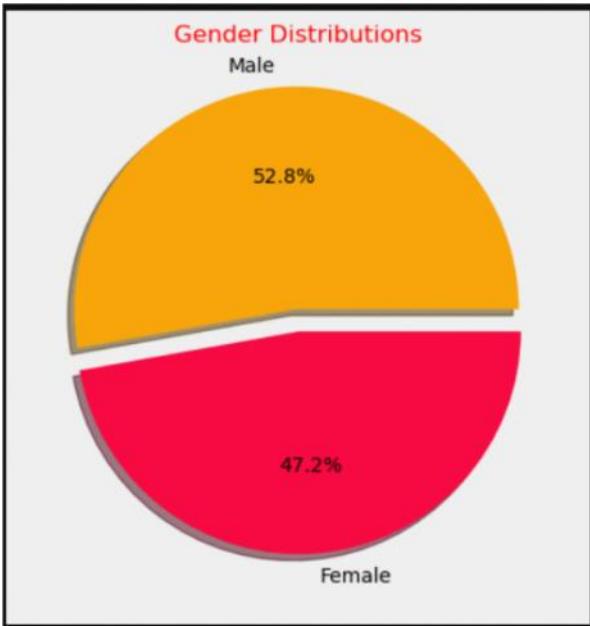
gender_counts.index

Index(['Male', 'Female'], dtype='object', name='gender')

gender_counts.values

array([528, 472], dtype=int64)

# Pie Chart [Composition Visuals]
plt.figure(figsize = (5,7) , facecolor = '#E8E8E8')
plt.pie(gender_counts, labels = gender_counts.index, colors = ['#F0A00A', '#F00A40'],
        explode = (0.1,0) , autopct = '%1.1f%%', shadow=True)
plt.title("Gender Distributions", color = 'red')
plt.show()
```



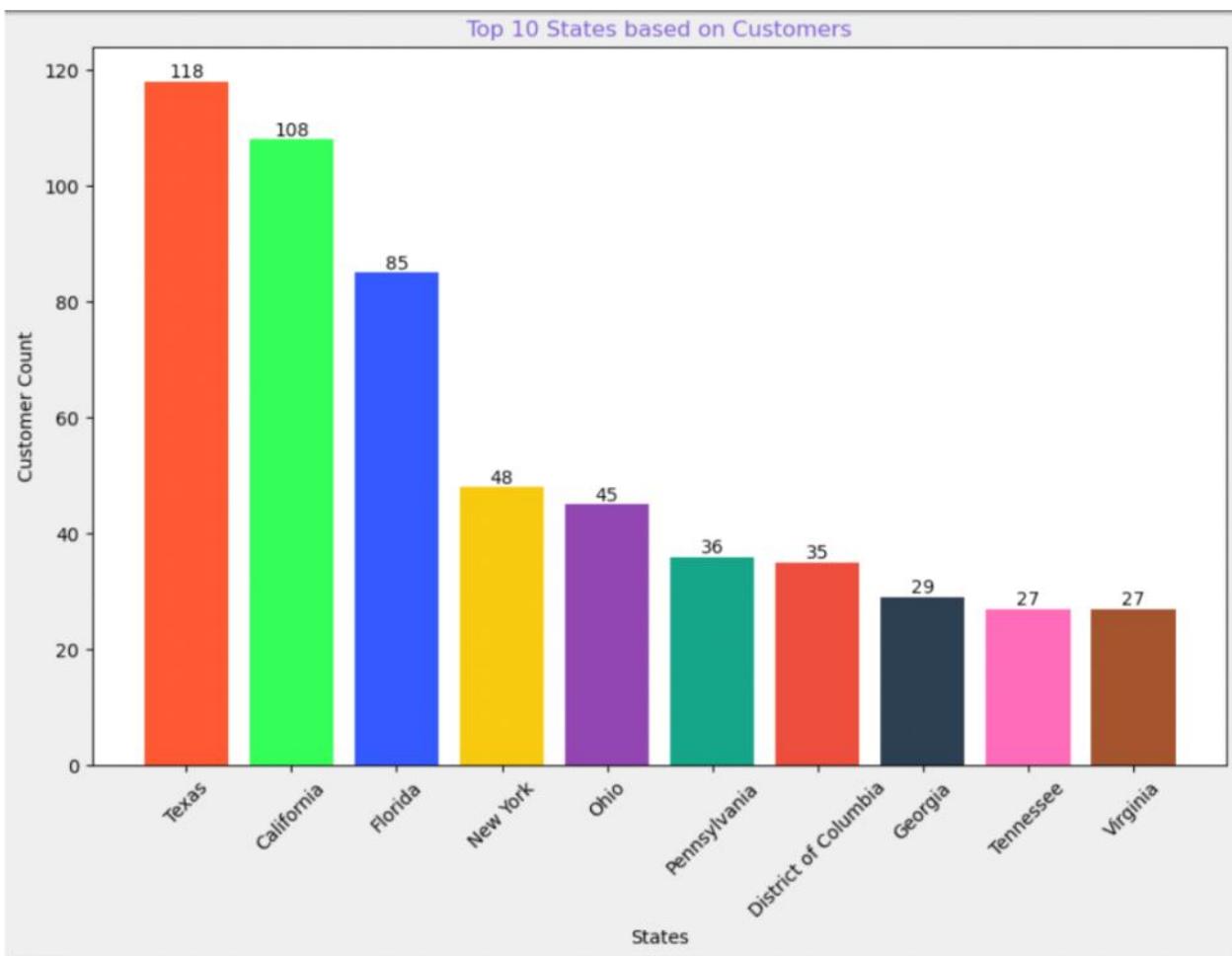
```
# Top 10 States based on count of customers.
top_states = customers['state'].value_counts().nlargest(10)
top_states # [Bar Chart]

state
Texas           118
California      108
Florida          85
New York         48
Ohio             45
Pennsylvania    36
District of Columbia 35
Georgia          29
Tennessee        27
Virginia         27
Name: count, dtype: int64

top_states.index
Index(['Texas', 'California', 'Florida', 'New York', 'Ohio', 'Pennsylvania',
       'District of Columbia', 'Georgia', 'Tennessee', 'Virginia'],
      dtype='object', name='state')

top_states.values
array([118, 108, 85, 48, 45, 36, 35, 29, 27, 27], dtype=int64)
```

```
# Plotting the Bar Plot
plt.figure(figsize = (11,7) , facecolor = '#E8E8E8')
colors = ["#FF5733", "#33FF57", "#3357FF", "#F1C40F", "#8E44AD",
          "#16A085", "#E74C3C", "#2C3E50", "#FF69B4", "#A0522D" ]
bar_state = plt.bar(top_states.index , top_states.values , color = colors)
plt.title("Top 10 States based on Customers" , color = '#7E5ED1')
plt.xlabel('States')
plt.ylabel('Customer Count')
plt.xticks(rotation=45)
plt.bar_label(bar_state)
plt.show()
```



```
# Calculated Columns [DAX] (row context , filtering)
# Add Column [Power Query Editor] [Date column -> Extract [Year,month,days]]
purchases['year'] = purchases['purch_date'].dt.year
purchases['month'] = purchases['purch_date'].dt.month
purchases['item_price'] = purchases['paid'] / purchases['amount']
purchases
```

| | id | purch_date | customer_num | product_num | amount | paid | year | month | item_price | |
|-------------|-----------|-------------------|---------------------|--------------------|---------------|-------------|-------------|--------------|-------------------|-------|
| 0 | 1 | 2019-01-03 | | 823 | 27 | 12 | 568.92 | 2019 | 1 | 47.41 |
| 1 | 2 | 2019-01-03 | | 606 | 28 | 14 | 395.36 | 2019 | 1 | 28.24 |
| 2 | 3 | 2019-01-03 | | 955 | 9 | 17 | 510.17 | 2019 | 1 | 30.01 |
| 3 | 4 | 2019-01-03 | | 577 | 19 | 3 | 68.49 | 2019 | 1 | 22.83 |
| 4 | 5 | 2019-01-03 | | 429 | 8 | 18 | 759.42 | 2019 | 1 | 42.19 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | |
| 5995 | 5996 | 2019-06-20 | | 893 | 33 | 5 | 411.10 | 2019 | 6 | 82.22 |
| 5996 | 5997 | 2019-06-20 | | 566 | 23 | 11 | 178.97 | 2019 | 6 | 16.27 |
| 5997 | 5998 | 2019-06-20 | | 114 | 19 | 9 | 205.47 | 2019 | 6 | 22.83 |
| 5998 | 5999 | 2019-06-20 | | 404 | 11 | 20 | 429.40 | 2019 | 6 | 21.47 |
| 5999 | 6000 | 2019-06-20 | | 88 | 57 | 4 | 274.52 | 2019 | 6 | 68.63 |

6000 rows × 9 columns

```
# Trend Axis [Time Intelligence] Over a Period of Time [Line Plot]
monthly_spendings = purchases.groupby('month')['amount'].sum().reset_index()
monthly_spendings
```

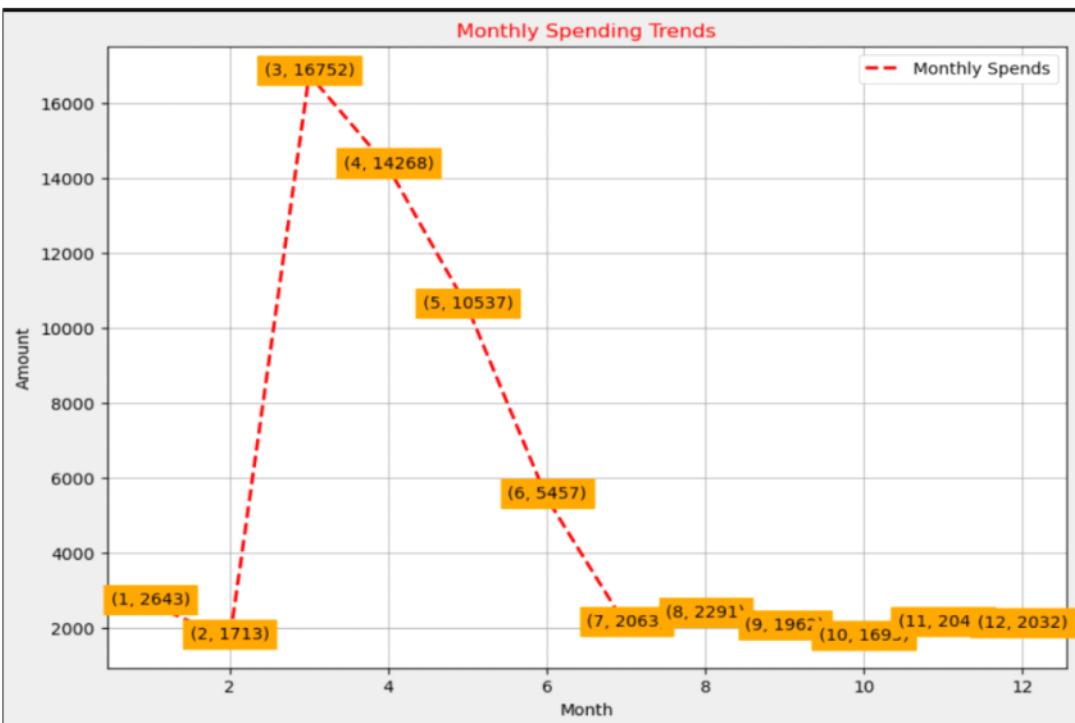
| month | amount |
|-------|---------|
| 0 | 1 2643 |
| 1 | 2 1713 |
| 2 | 3 16752 |
| 3 | 4 14268 |
| 4 | 5 10537 |
| 5 | 6 5457 |
| 6 | 7 2063 |
| 7 | 8 2291 |
| 8 | 9 1962 |
| 9 | 10 1695 |
| 10 | 11 2044 |
| 11 | 12 2032 |

```
plt.figure(figsize = (10,7) , facecolor = '#E8E8E8')

# Line
plt.plot(monthly_spendings['month'] , monthly_spendings['amount'] ,
         color = 'red', linestyle = '--', linewidth = 2, label = 'Monthly Spends')

# Add a Data Label
for x,y in zip(monthly_spendings['month'] , monthly_spendings['amount']):
    plt.text(x,y+2,f'{x,y}', ha = 'center', fontsize=10, color = 'black', backgroundcolor = 'orange')

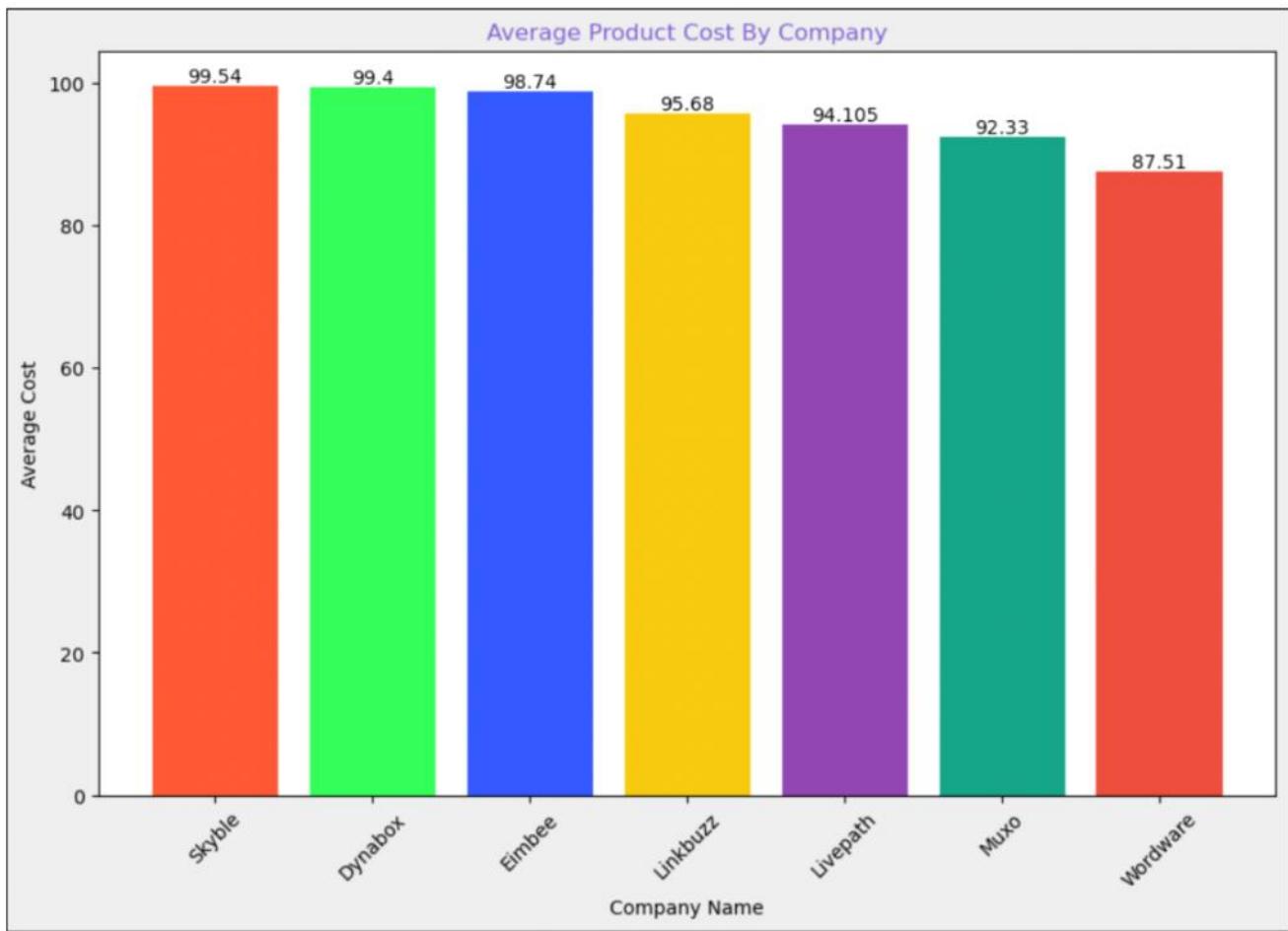
plt.title('Monthly Spending Trends' , color = 'red')
plt.xlabel('Month')
plt.ylabel('Amount')
plt.grid()
plt.legend()
plt.show()
```



```
# Average Product cost by company
avg_product_by_comp = products.groupby('company')['cost'].mean().nlargest(7)
avg_product_by_comp
```

```
company
Skyble    99.540
Dynabox   99.400
Eimbee    98.740
Linkbuzz  95.680
Livepath  94.105
Muxo      92.330
Wordware  87.510
Name: cost, dtype: float64
```

```
# Plotting the Bar Plot
plt.figure(figsize = (11,7) , facecolor = '#E8E8E8')
colors = ["#FF5733", "#33FF57", "#3357FF", "#F1C40F", "#8E44AD","#16A085", "#E74C3C" ]
company_bar = plt.bar(avg_product_by_comp.index , avg_product_by_comp.values , color = colors)
plt.title("Average Product Cost By Company" , color = '#7E5ED1')
plt.xlabel('Company Name')
plt.ylabel('Average Cost')
plt.xticks(rotation=45)
plt.bar_label(company_bar)
# Save the figure
plt.savefig('avg_cost_by_company.png')
plt.show()
```



```

# Univariate Analysis ['Single Column'] [Continuous Columns]
# Outlier Detection -> [Box-Wisker Plot]
purchases['paid'].describe() # +ve Skewness

count    6000.00000
mean     548.871353
std      444.237192
min      3.630000
25%     187.560000
50%     422.700000
75%     813.800000
max     1990.800000
Name: paid, dtype: float64

purchases['paid']

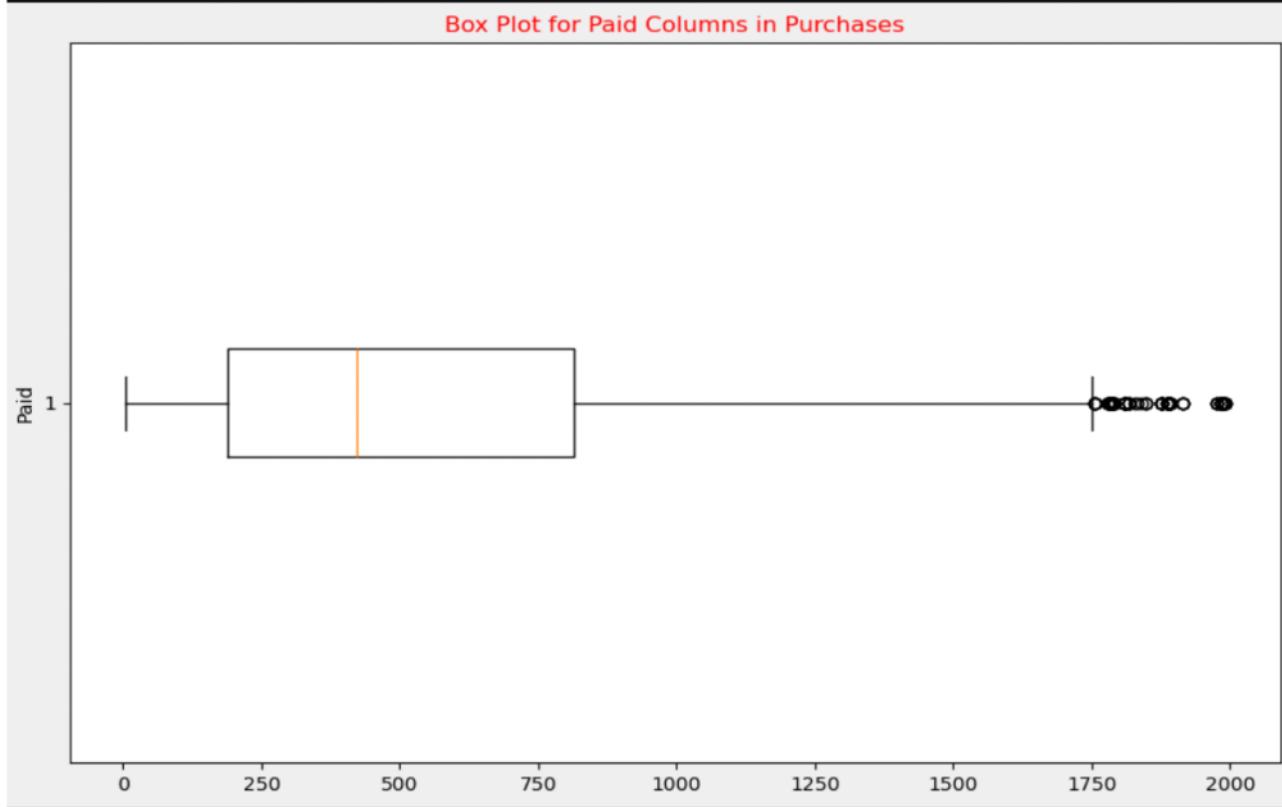
0      568.92
1      395.36
2      510.17
3      68.49
4      759.42
...
5995   411.10
5996   178.97
5997   205.47
5998   429.40
5999   274.52
Name: paid, Length: 6000, dtype: float64

```

```

# Plotting the Box Plot
plt.figure(figsize = (11,7) , facecolor = '#E8E8E8')
plt.boxplot(purchases['paid'] , vert = False)
plt.title('Box Plot for Paid Columns in Purchases' , color ='red')
plt.ylabel('Paid')
plt.show()

```

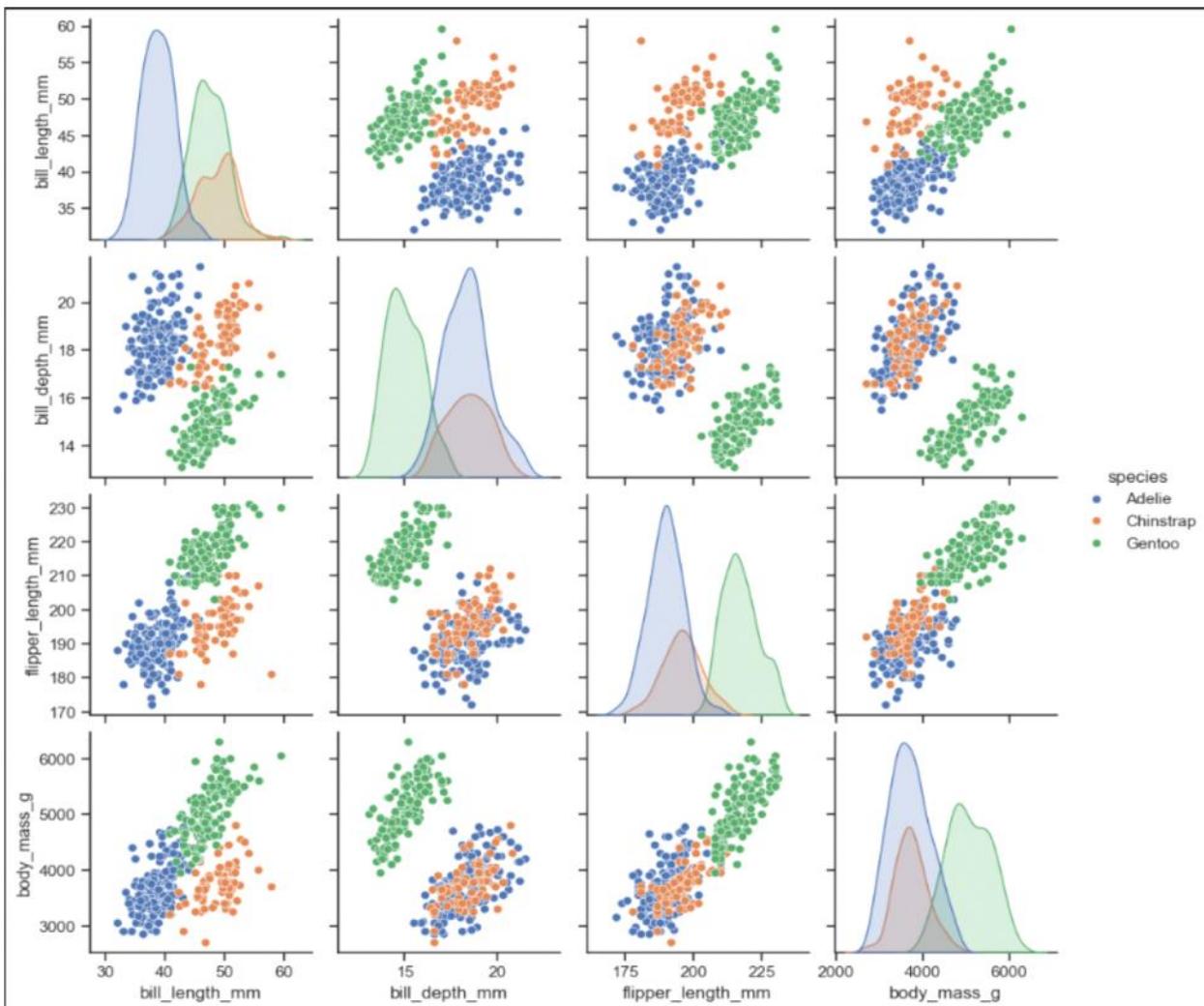


```

import seaborn as sns
sns.set_theme(style="ticks")

df = sns.load_dataset("penguins")
sns.pairplot(df, hue="species")

```



| df | | | | | | | |
|-----|---------|-----------|----------------|---------------|-------------------|-------------|--------|
| | species | island | bill_length_mm | bill_depth_mm | flipper_length_mm | body_mass_g | sex |
| 0 | Adelie | Torgersen | 39.1 | 18.7 | 181.0 | 3750.0 | Male |
| 1 | Adelie | Torgersen | 39.5 | 17.4 | 186.0 | 3800.0 | Female |
| 2 | Adelie | Torgersen | 40.3 | 18.0 | 195.0 | 3250.0 | Female |
| 3 | Adelie | Torgersen | NaN | NaN | NaN | NaN | NaN |
| 4 | Adelie | Torgersen | 36.7 | 19.3 | 193.0 | 3450.0 | Female |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 339 | Gentoo | Biscoe | NaN | NaN | NaN | NaN | NaN |
| 340 | Gentoo | Biscoe | 46.8 | 14.3 | 215.0 | 4850.0 | Female |
| 341 | Gentoo | Biscoe | 50.4 | 15.7 | 222.0 | 5750.0 | Male |
| 342 | Gentoo | Biscoe | 45.2 | 14.8 | 212.0 | 5200.0 | Female |
| 343 | Gentoo | Biscoe | 49.9 | 16.1 | 213.0 | 5400.0 | Male |

344 rows × 7 columns

```

df['species'].unique()

array(['Adelie', 'Chinstrap', 'Gentoo'], dtype=object)

df['island'].unique()

array(['Torgersen', 'Biscoe', 'Dream'], dtype=object)

df['sex'].unique()

array(['Male', 'Female', nan], dtype=object)

import seaborn as sns
sns.get_dataset_names()

['anagrams',
 'anscombe',
 'attention',
 'brain_networks',
 'car_crashes',
 'diamonds',
 'dots',
 'dowjones',
 'exercise',
 'flights',
 'fmri',
 'geyser',
 'glue',
 'healthexp',
 'iris',
 'mpg',
 'penguins',
 'planets',
 'seairce',
 'taxis',
 'tips',
 'titanic']

```

```

import seaborn as sns
sns.set_theme(style="ticks")

tips = sns.load_dataset("tips")
tips

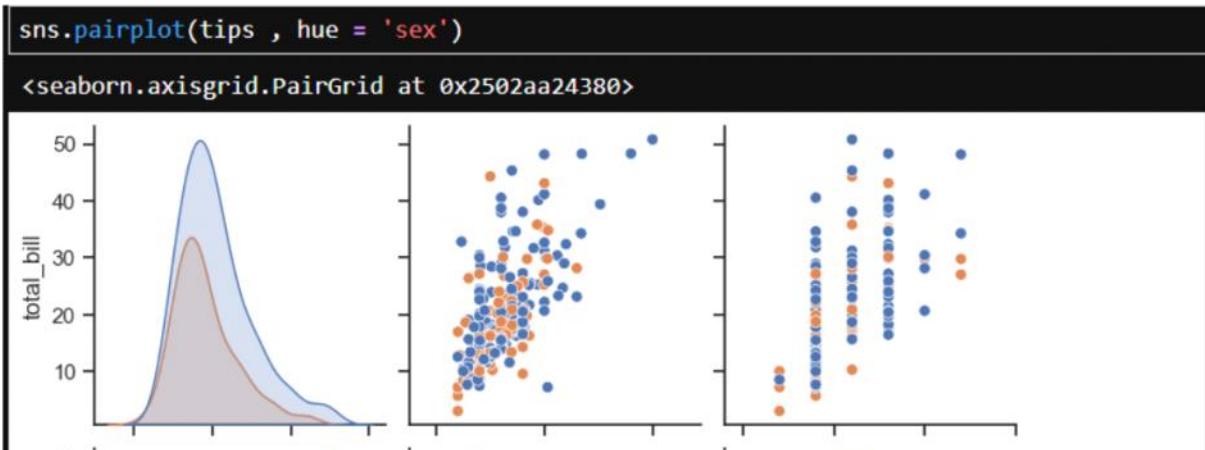
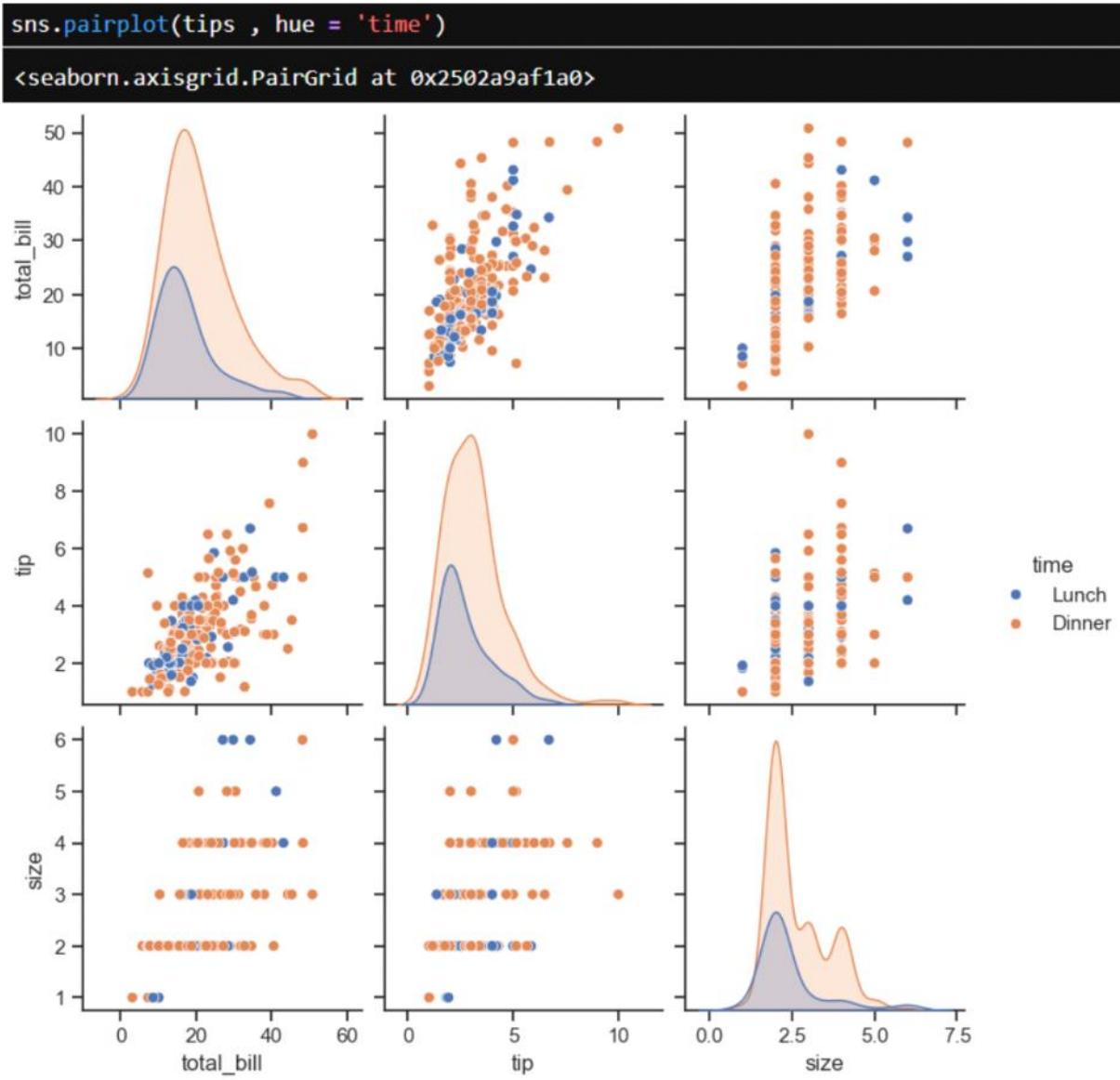
```

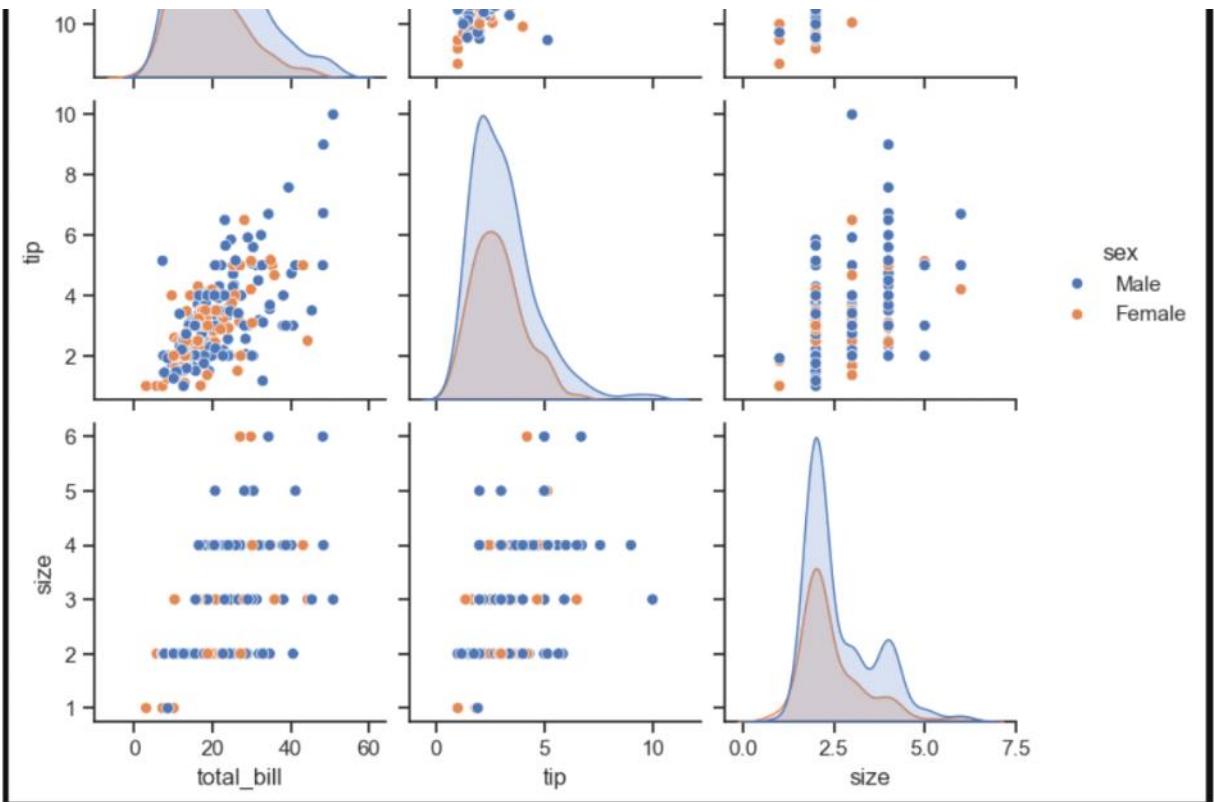
| | total_bill | tip | sex | smoker | day | time | size |
|------------|------------|------|--------|--------|------|--------|------|
| 0 | 16.99 | 1.01 | Female | No | Sun | Dinner | 2 |
| 1 | 10.34 | 1.66 | Male | No | Sun | Dinner | 3 |
| 2 | 21.01 | 3.50 | Male | No | Sun | Dinner | 3 |
| 3 | 23.68 | 3.31 | Male | No | Sun | Dinner | 2 |
| 4 | 24.59 | 3.61 | Female | No | Sun | Dinner | 4 |
| ... | ... | ... | ... | ... | ... | ... | ... |
| 239 | 29.03 | 5.92 | Male | No | Sat | Dinner | 3 |
| 240 | 27.18 | 2.00 | Female | Yes | Sat | Dinner | 2 |
| 241 | 22.67 | 2.00 | Male | Yes | Sat | Dinner | 2 |
| 242 | 17.82 | 1.75 | Male | No | Sat | Dinner | 2 |
| 243 | 18.78 | 3.00 | Female | No | Thur | Dinner | 2 |

244 rows × 7 columns

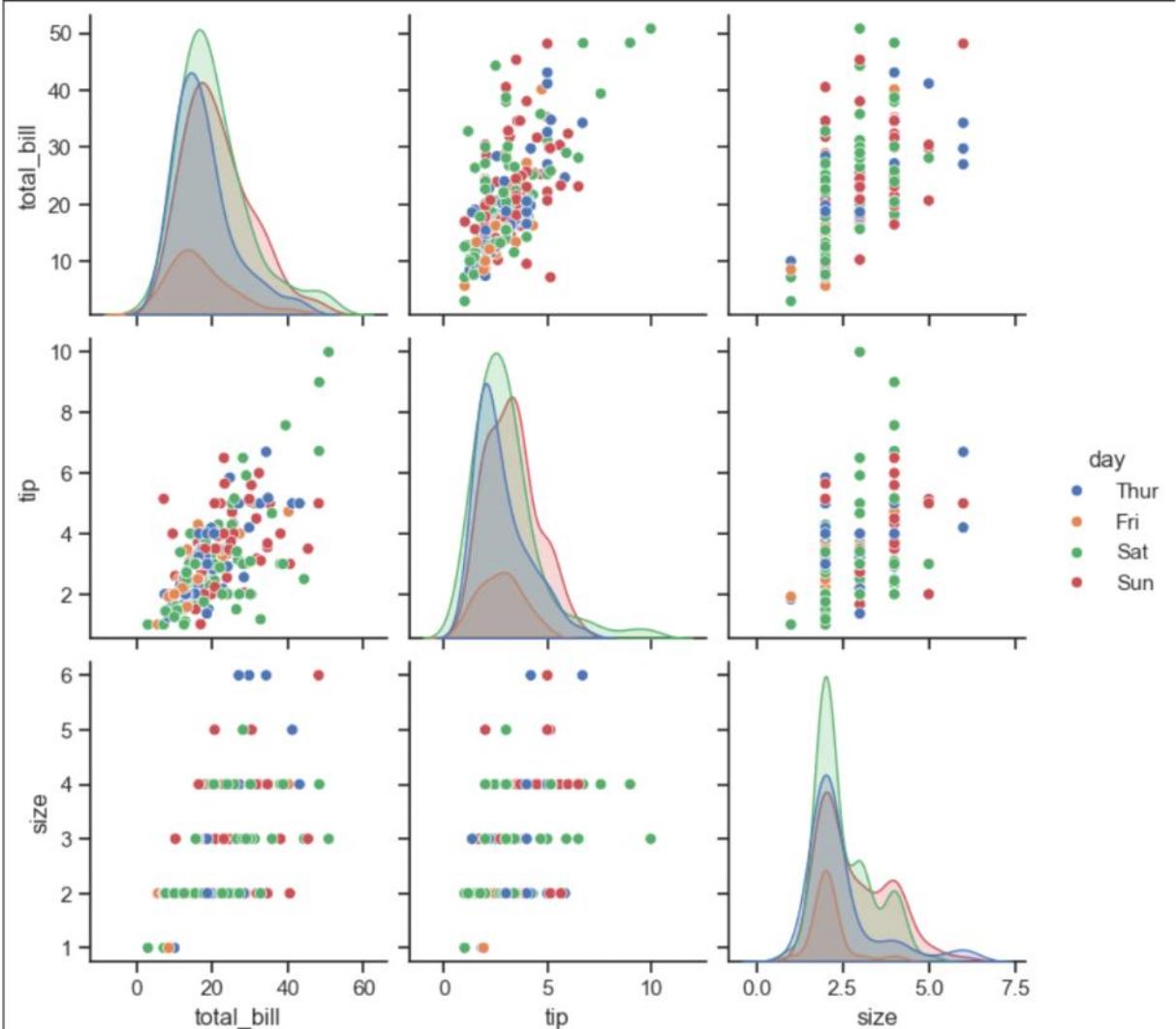
```
tips['time'].unique()
['Dinner', 'Lunch']
Categories (2, object): ['Lunch', 'Dinner']

tips['day'].unique()
['Sun', 'Sat', 'Thur', 'Fri']
Categories (4, object): ['Thur', 'Fri', 'Sat', 'Sun']
```



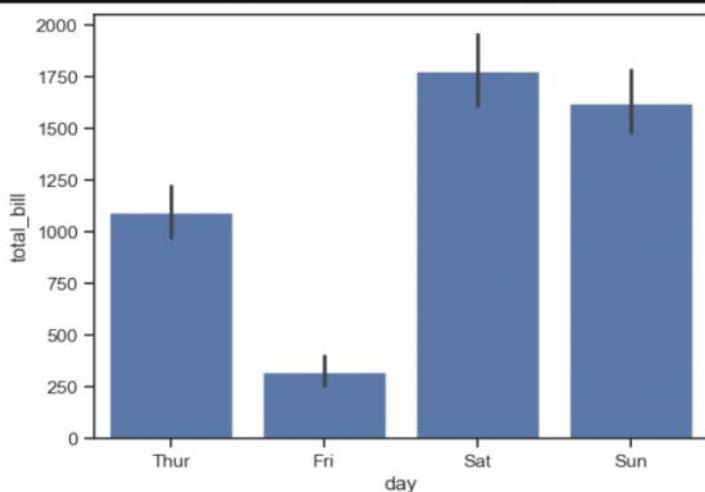


```
sns.pairplot(tips , hue = 'sex')
<seaborn.axisgrid.PairGrid at 0x2502aaca840>
```



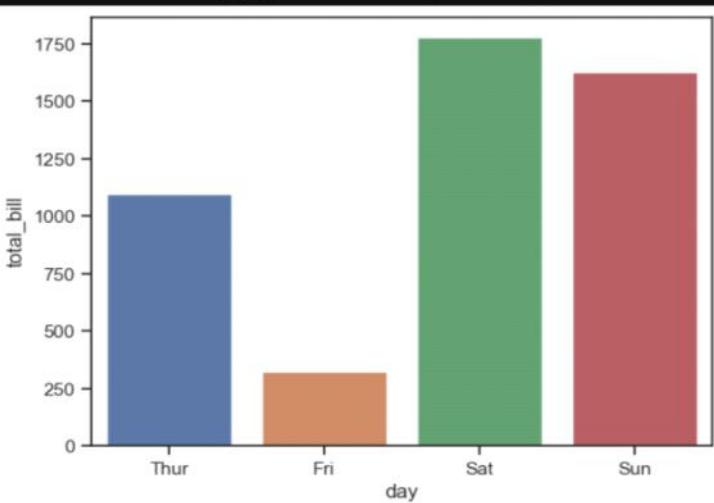
```
sns.barplot(tips, x="day", y="total_bill" , estimator='sum')
```

```
<Axes: xlabel='day', ylabel='total_bill'>
```

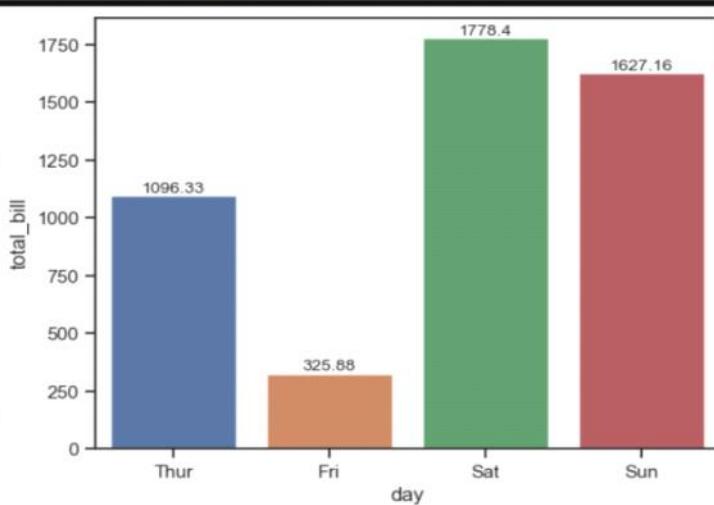


```
sns.barplot(tips, x="day", y="total_bill" , estimator='sum', errorbar=None, hue = 'day')
```

```
<Axes: xlabel='day', ylabel='total_bill'>
```



```
ax = sns.barplot(tips, x="day", y="total_bill" , estimator='sum', errorbar=None, hue = 'day')
ax.bar_label(ax.containers[0], fontsize=10)
ax.bar_label(ax.containers[1], fontsize=10)
ax.bar_label(ax.containers[2], fontsize=10)
ax.bar_label(ax.containers[3], fontsize=10);
```



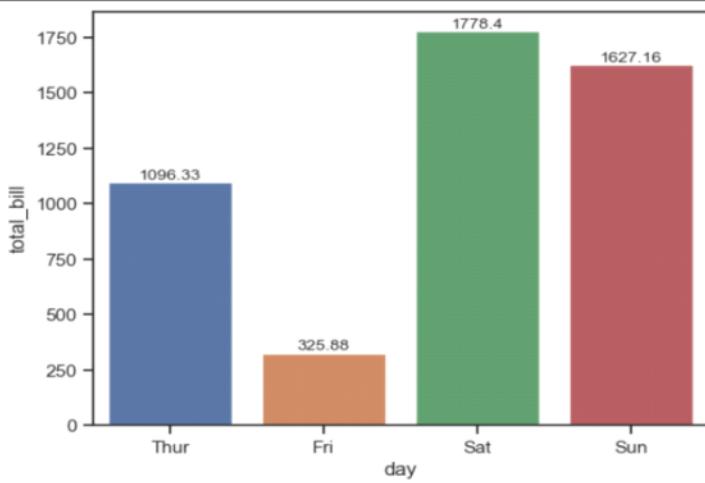
```

days_length = tips['day'].nunique()
days_length

4

ax = sns.barplot(tips, x="day", y="total_bill" , estimator='sum', errorbar=None, hue = 'day')
for i in range(0,days_length): # [0,1,2,3]
    ax.bar_label(ax.containers[i], fontsize=10)

```

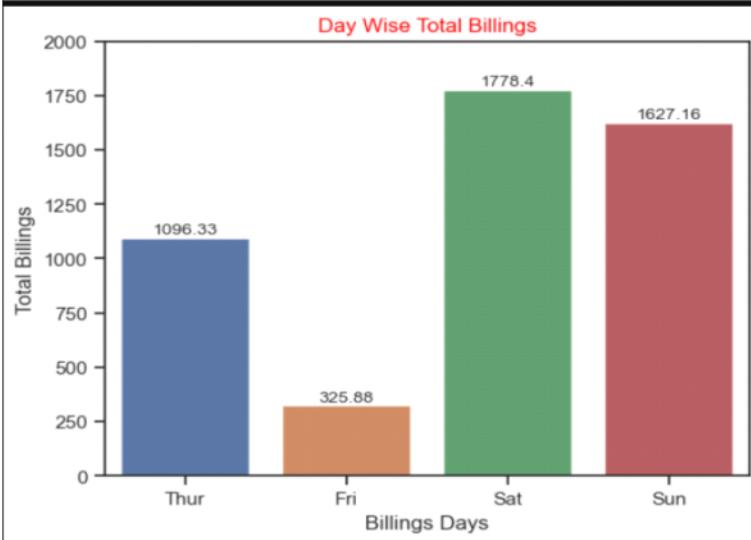


```

ax = sns.barplot(tips, x="day", y="total_bill" , estimator='sum', errorbar=None, hue = 'day')
for i in range(0,days_length): # [0,1,2,3]
    ax.bar_label(ax.containers[i], fontsize=10)

plt.title('Day Wise Total Billings' , color = 'red')
plt.ylim(0,2000)
plt.xlabel('Billings Days')
plt.ylabel('Total Billings')
plt.show()

```



```

containers = ax.containers # [<BarContainer object of 1 artists>, ...]
heights = []
if containers:
    for i in range(0,4):
        nth_container = containers[i] # First BarContainer
        for patch in nth_container.patches:
            heights.append(patch.get_height())
print(heights)

```

[1096.33, 325.88, 1778.399999999999, 1627.16]