

November 3, 2025

EXAMPLE

```
[ ]: print('hi')
```

hi

1 NUMPY

NumPy is a fundamental library for numerical computing in Python. It provides support for arrays and matrices, along with a large collection of mathematical functions to operate on these arrays.

```
[ ]: #IMPORTING NUMPY LIBRARY
      import numpy as np
```

```
[ ]: #single dimensional array
      n1=np.array([1,2,3,4,5])
      n1
```

```
[ ]: array([1, 2, 3, 4, 5])
```

```
[ ]: #multidimensional array
      n2=np.array([[1,2,3],[4,5,6]])
      n2
```

```
[ ]: array([[1, 2, 3],
           [4, 5, 6]])
```

```
[ ]: #index value
      n1[0]
```

```
[ ]: np.int64(1)
```

```
[ ]: n1[1]
```

```
[ ]: np.int64(2)
```

```
[ ]: n1=np.array([[1,2,3],[4,5,6],[7,8,9]])
      n1[:,1]
```

```
[ ]: array([2, 5, 8])
```

```
[ ]: n1[:,2]
```

```
[ ]: array([3, 6, 9])
```

```
[ ]: #Initializing Numpy array with zeros  
n1=np.zeros((1,2))  
n1
```

```
[ ]: array([[0., 0.]])
```

```
[ ]: #multidimensional array  
n2=np.zeros((4,4))  
n2
```

```
[ ]: array([[0., 0., 0., 0.],  
           [0., 0., 0., 0.],  
           [0., 0., 0., 0.],  
           [0., 0., 0., 0.]])
```

```
[ ]: #initializing numpy array with same numbers  
n1=np.full((2,2),10)  
n1
```

```
[ ]: array([[10, 10],  
           [10, 10]])
```

```
[ ]: #initializing numpy array  
n1=np.arange(10,20)  
n1
```

```
[ ]: array([10, 11, 12, 13, 14, 15, 16, 17, 18, 19])
```

```
[ ]: #with difference  
n1=np.arange(10,40,2)  
n1
```

```
[ ]: array([10, 12, 14, 16, 18, 20, 22, 24, 26, 28, 30, 32, 34, 36, 38])
```

```
[ ]: #random integers  
n1=np.random.randint(1,45,3)  
n1
```

```
[ ]: array([19, 23, 14])
```

```
[ ]: #shape
n1=np.array([[1,2,3],[4,5,6]])
n1.shape
```

```
[ ]: (2, 3)
```

```
[ ]: #vstack() stacks arrays vertically(row-wise)
n1=np.array([10,20,30])
n2=np.array([40,50,60])
np.vstack((n1,n2))
```

```
[ ]: array([10, 20, 30],
[40, 50, 60])
```

```
[ ]: #Hstack() stacks arrays horizontally ( column-wise)
n1= np.array([10,20,30])
n2=np.array([40,50,60])
np.hstack((n1,n2))
```

```
[ ]: array([10, 20, 30, 40, 50, 60])
```

```
[ ]: #COLUMN_STACK() stacks 1-D arrays as columns into a 2D-array
n1=np.array([10,20,30])
n2=np.array([40,50,60])
np.column_stack((n1,n2))
```

```
[ ]: array([[10, 40],
[20, 50],
[30, 60]])
```

```
[ ]: #intersection
np.intersect1d(n1,n2)
```

```
[ ]: array([], dtype=int64)
```

```
[ ]: #intersection
n3=np.array([20,40,60])
np.intersect1d(n1,n3)
```

```
[ ]: array([20])
```

```
[ ]: #difference
np.setdiff1d(n1,n2)
```

```
[ ]: array([10, 20, 30])
```

```
[ ]: np.setdiff1d(n2,n1)
```

```
[ ]: array([40, 50, 60])

[ ]: #total sum
    np.sum([n1,n2])

[ ]: np.int64(210)

[ ]: #sum of respective coordinate axis
    np.sum([n1,n2],axis=0)

[ ]: array([50, 70, 90])

[ ]: #addition
    n1=n1+1
    n1

[ ]: array([11, 21, 31])

[ ]: #multiplication
    n1=np.array([[1,2,3],[4,5,6],[7,8,9]])
    n2=np.array([[1,2,3],[3,4,5],[5,6,7]])
    n1.dot(n2)

[ ]: array([[ 22,   28,   34],
           [ 49,   64,   79],
           [ 76,  100,  124]])

[ ]: n2.dot(n1.T)

[ ]: array([[ 14,   32,   50],
           [ 26,   62,   98],
           [ 38,   92,  146]])

[ ]: n1=n1*2
    n1

[ ]: array([[ 2,   4,   6],
           [ 8,  10,  12],
           [14,  16,  18]])

[ ]: #division
    n1=n1/2
    n1

[ ]: array([[1., 2., 3.],
           [4., 5., 6.],
           [7., 8., 9.]])
```

```
[ ]: #mean  
np.mean(n1)
```

```
[ ]: np.float64(5.0)
```

```
[ ]: #median  
np.median(n1)
```

```
[ ]: np.float64(5.0)
```

```
[ ]: #standard deviation  
np.std(n1)
```

```
[ ]: np.float64(2.581988897471611)
```

```
[ ]: #transpose()  
n1.transpose()
```

```
[ ]: array([[1., 4., 7.],  
           [2., 5., 8.],  
           [3., 6., 9.]])
```

2 PANDA

```
[ ]: #importing pandas library  
import pandas as pd
```

```
[ ]: #series object is one-dimensional labeled array  
s1=pd.Series([10,20,30,40,50])  
s1
```

```
[ ]: 0    10  
1    20  
2    30  
3    40  
4    50  
dtype: int64
```

```
[ ]: #userdefined index  
s1=pd.Series([10,20,30],index=['x','y','z'])  
s1
```

```
[ ]: x    10  
y    20  
z    30  
dtype: int64
```

```
[ ]: #dictionary  
pd.Series({'a':10,'b':20,'c':30})
```

```
[ ]: a    10  
b    20  
c    30  
dtype: int64
```

```
[ ]: #missed data  
pd.Series({'a':10,'b':20,'c':30},index=['b','a','g'])
```

```
[ ]: b    20.0  
a    10.0  
g     NaN  
dtype: float64
```

```
[ ]: #extracting a single element  
s1=pd.Series([10,20,30,40,50])  
s1[3]
```

```
[ ]: np.int64(40)
```

```
[ ]: #extracting a sequence of elements  
s1=pd.Series([10,20,30,40,50])  
s1[:4]
```

```
[ ]: 0    10  
1    20  
2    30  
3    40  
dtype: int64
```

```
[ ]: #extracting elements from back  
s2=pd.Series([10,20,30,40,50])  
s2[-3:]
```

```
[ ]: 2    30  
3    40  
4    50  
dtype: int64
```

```
[ ]: s1+s2
```

```
[ ]: 0    20  
1    40  
2    60  
3    80
```

```
4      100  
dtype: int64
```

```
[ ]: s1+5
```

```
[ ]: 0    15  
1    25  
2    35  
3    45  
4    55  
dtype: int64
```

```
[ ]: #creating dataframes  
pd.DataFrame({'Name': ['A', 'B', 'C'], 'Marks': [90, 75, 82]})
```

```
[ ]:   Name  Marks  
0     A      90  
1     B      75  
2     C      82
```

```
[ ]: #defining through variable  
df=pd.DataFrame({'Name': ['A', 'B', 'C'], 'Marks': [90, 75, 82]})  
df
```

```
[ ]:   Name  Marks  
0     A      90  
1     B      75  
2     C      82
```

```
[ ]: type(df)
```

```
[ ]: pandas.core.frame.DataFrame
```

3 IRIS DATASET

```
[ ]: from google.colab import files  
uploaded=files.upload()
```

```
<IPython.core.display.HTML object>
```

```
Saving IRIS.csv to IRIS (1).csv
```

```
[ ]: df=pd.read_csv('IRIS.csv')  
print(df.head())
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	Iris-setosa

```
1          4.9      3.0      1.4      0.2 Iris-setosa
2          4.7      3.2      1.3      0.2 Iris-setosa
3          4.6      3.1      1.5      0.2 Iris-setosa
4          5.0      3.6      1.4      0.2 Iris-setosa
```

```
[ ]: #head()-displays first n rows by default 5
df.head(10)
```

```
[ ]:   sepal_length  sepal_width  petal_length  petal_width      species
0          5.1      3.5      1.4      0.2 Iris-setosa
1          4.9      3.0      1.4      0.2 Iris-setosa
2          4.7      3.2      1.3      0.2 Iris-setosa
3          4.6      3.1      1.5      0.2 Iris-setosa
4          5.0      3.6      1.4      0.2 Iris-setosa
5          5.4      3.9      1.7      0.4 Iris-setosa
6          4.6      3.4      1.4      0.3 Iris-setosa
7          5.0      3.4      1.5      0.2 Iris-setosa
8          4.4      2.9      1.4      0.2 Iris-setosa
9          4.9      3.1      1.5      0.1 Iris-setosa
```

```
[ ]: #displays last n rows by default 5
df.tail()
```

```
[ ]:   sepal_length  sepal_width  petal_length  petal_width      species
145         6.7      3.0      5.2      2.3 Iris-virginica
146         6.3      2.5      5.0      1.9 Iris-virginica
147         6.5      3.0      5.2      2.0 Iris-virginica
148         6.2      3.4      5.4      2.3 Iris-virginica
149         5.9      3.0      5.1      1.8 Iris-virginica
```

```
[ ]: df.tail(10)
```

```
[ ]:   sepal_length  sepal_width  petal_length  petal_width      species
140         6.7      3.1      5.6      2.4 Iris-virginica
141         6.9      3.1      5.1      2.3 Iris-virginica
142         5.8      2.7      5.1      1.9 Iris-virginica
143         6.8      3.2      5.9      2.3 Iris-virginica
144         6.7      3.3      5.7      2.5 Iris-virginica
145         6.7      3.0      5.2      2.3 Iris-virginica
146         6.3      2.5      5.0      1.9 Iris-virginica
147         6.5      3.0      5.2      2.0 Iris-virginica
148         6.2      3.4      5.4      2.3 Iris-virginica
149         5.9      3.0      5.1      1.8 Iris-virginica
```

```
[ ]: df.shape
```

```
[ ]: (150, 5)
```

```
[ ]: #generates descriptiv statistics of a dataframe  
df.describe()
```

```
[ ]:      sepal_length  sepal_width  petal_length  petal_width  
count    150.000000   150.000000   150.000000   150.000000  
mean     5.843333    3.054000    3.758667    1.198667  
std      0.828066    0.433594    1.764420    0.763161  
min      4.300000    2.000000    1.000000    0.100000  
25%     5.100000    2.800000    1.600000    0.300000  
50%     5.800000    3.000000    4.350000    1.300000  
75%     6.400000    3.300000    5.100000    1.800000  
max     7.900000    4.400000    6.900000    2.500000
```

```
[ ]: #integer-location based indexing/selection by position  
df.iloc[0:4,0:2]
```

```
[ ]:      sepal_length  sepal_width  
0          5.1         3.5  
1          4.9         3.0  
2          4.7         3.2  
3          4.6         3.1
```

```
[ ]: df.iloc[0:5,3:5]
```

```
[ ]:      petal_width      species  
0          0.2  Iris-setosa  
1          0.2  Iris-setosa  
2          0.2  Iris-setosa  
3          0.2  Iris-setosa  
4          0.2  Iris-setosa
```

```
[ ]: df.iloc[10:20,2:5]
```

```
[ ]:      petal_length  petal_width      species  
10          1.5         0.2  Iris-setosa  
11          1.6         0.2  Iris-setosa  
12          1.4         0.1  Iris-setosa  
13          1.1         0.1  Iris-setosa  
14          1.2         0.2  Iris-setosa  
15          1.5         0.4  Iris-setosa  
16          1.3         0.4  Iris-setosa  
17          1.4         0.3  Iris-setosa  
18          1.7         0.3  Iris-setosa  
19          1.5         0.3  Iris-setosa
```

```
[ ]: #removes specified rows or columns  
df.drop('sepal_length',axis=1)
```

```
[ ]:      sepal_width  petal_length  petal_width      species
 0          3.5          1.4          0.2  Iris-setosa
 1          3.0          1.4          0.2  Iris-setosa
 2          3.2          1.3          0.2  Iris-setosa
 3          3.1          1.5          0.2  Iris-setosa
 4          3.6          1.4          0.2  Iris-setosa
 ..
145         3.0          5.2          2.3  Iris-virginica
146         2.5          5.0          1.9  Iris-virginica
147         3.0          5.2          2.0  Iris-virginica
148         3.4          5.4          2.3  Iris-virginica
149         3.0          5.1          1.8  Iris-virginica
```

[150 rows x 4 columns]

```
[ ]: #label-based indexing/selection by label
df.loc[3:6,('sepal_length','petal_length')]
```

```
[ ]:      sepal_length  petal_length
 3          4.6          1.5
 4          5.0          1.4
 5          5.4          1.7
 6          4.6          1.4
```

```
[ ]: df.drop([1,2,3],axis=0)
```

```
[ ]:      sepal_length  sepal_width  petal_length  petal_width      species
 0          5.1          3.5          1.4          0.2  Iris-setosa
 4          5.0          3.6          1.4          0.2  Iris-setosa
 5          5.4          3.9          1.7          0.4  Iris-setosa
 6          4.6          3.4          1.4          0.3  Iris-setosa
 7          5.0          3.4          1.5          0.2  Iris-setosa
 ..
145         6.7          3.0          5.2          2.3  Iris-virginica
146         6.3          2.5          5.0          1.9  Iris-virginica
147         6.5          3.0          5.2          2.0  Iris-virginica
148         6.2          3.4          5.4          2.3  Iris-virginica
149         5.9          3.0          5.1          1.8  Iris-virginica
```

[147 rows x 5 columns]

```
[ ]: df.mean(numeric_only=True)
```

```
[ ]: sepal_length    5.843333
sepal_width     3.054000
petal_length    3.758667
petal_width     1.198667
```

```

dtype: float64

[ ]: df.median(numeric_only=True)

[ ]: sepal_length      5.80
     sepal_width       3.00
     petal_length      4.35
     petal_width       1.30
     dtype: float64

[ ]: df.min()

[ ]: sepal_length          4.3
     sepal_width           2.0
     petal_length          1.0
     petal_width           0.1
     species               Iris-setosa
     dtype: object

[ ]: df.max()

[ ]: sepal_length          7.9
     sepal_width           4.4
     petal_length          6.9
     petal_width           2.5
     species               Iris-virginica
     dtype: object

[ ]: print(df.loc[0:4,['sepal_length','petal_length']))
df.drop('petal_width',axis=1,inplace=True)
df.drop([0,1],axis=0,inplace=True)
num=df.select_dtypes(include=['number'])
print("Mean:\n",num.mean())
print("Median:\n",num.median())
print("Min:\n",num.min())
print("Max:\n",num.max())

sepal_length  petal_length
0            5.1          1.4
1            4.9          1.4
2            4.7          1.3
3            4.6          1.5
4            5.0          1.4
Mean:
sepal_length      5.854730
sepal_width       3.051351
petal_length      3.790541

```

```
dtype: float64
Median:
  sepal_length    5.8
  sepal_width     3.0
  petal_length    4.4
dtype: float64
Min:
  sepal_length    4.3
  sepal_width     2.0
  petal_length    1.0
dtype: float64
Max:
  sepal_length    7.9
  sepal_width     4.4
  petal_length    6.9
dtype: float64
```

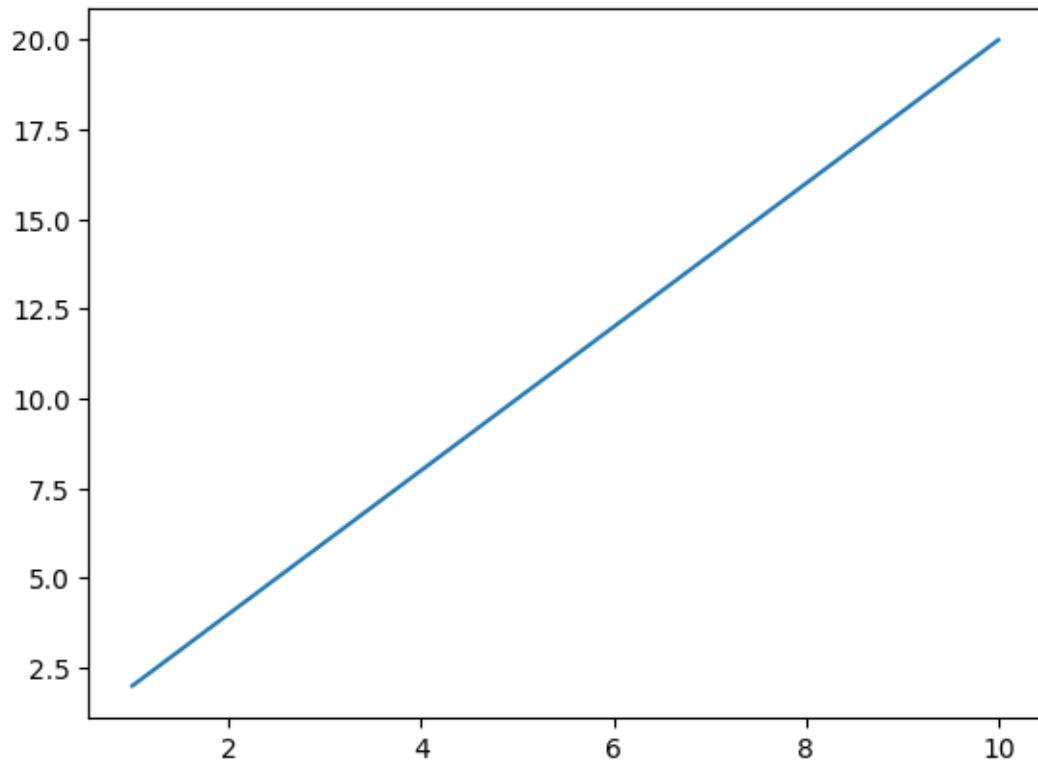
4 MATPLOTLIB

```
[ ]: from matplotlib import pyplot as plt
```

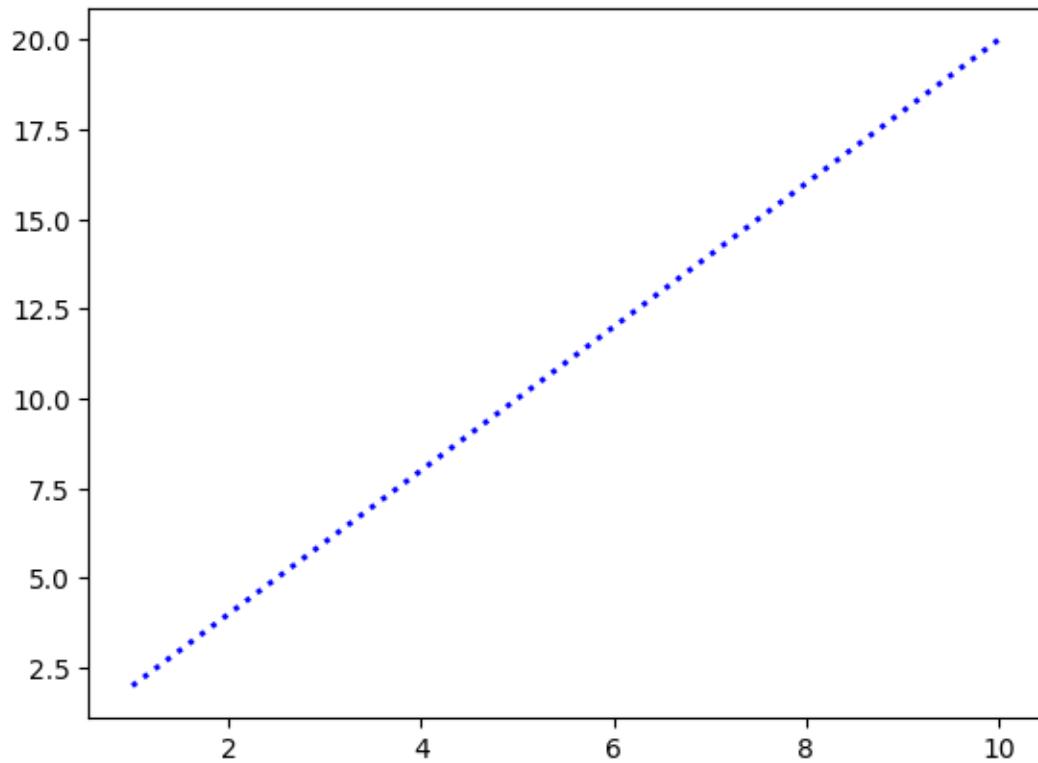
```
[ ]: x=np.arange(1,11)
y=2*x
print('x:',x)
print('y:',y)
```

```
x: [ 1  2  3  4  5  6  7  8  9 10]
y: [ 2  4  6  8 10 12 14 16 18 20]
```

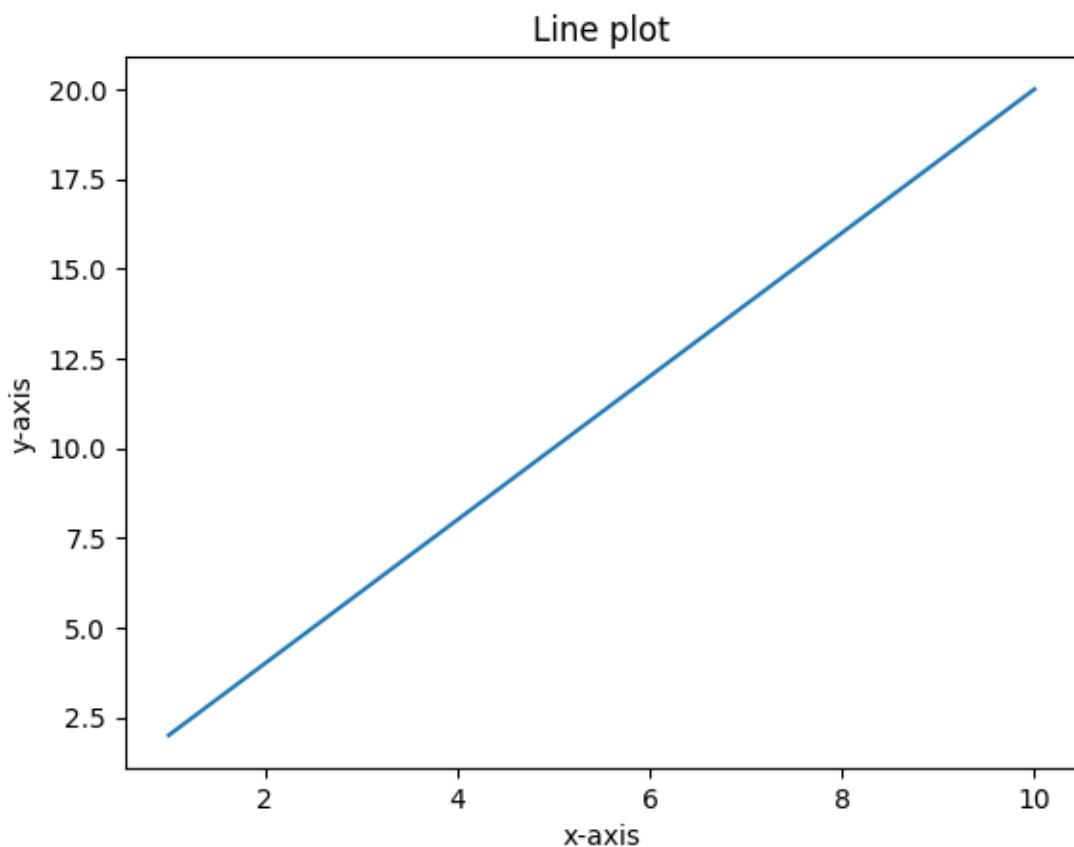
```
[ ]: plt.plot(x,y)
plt.show()
```



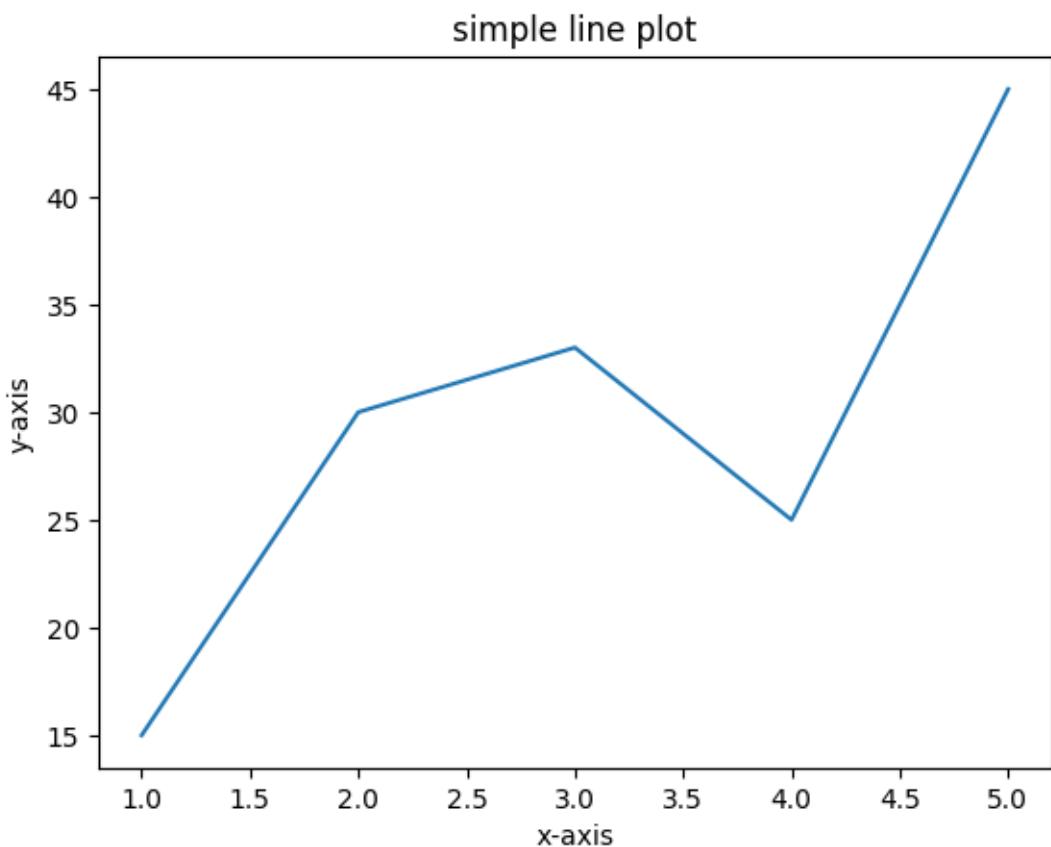
```
[ ]: plt.plot(x,y,color='blue',linestyle=':',linewidth=2)  
plt.show()
```



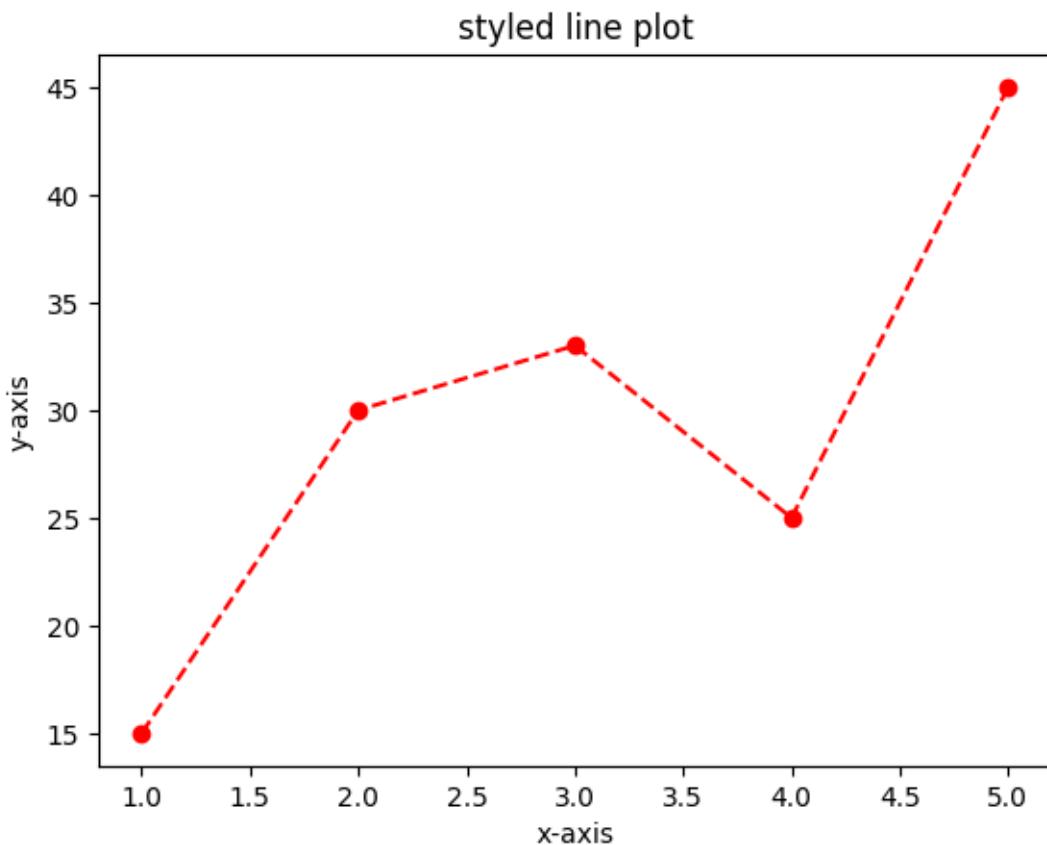
```
[ ]: plt.plot(x,y)
plt.title("Line plot")
plt.xlabel("x-axis")
plt.ylabel("y-axis")
plt.show()
```



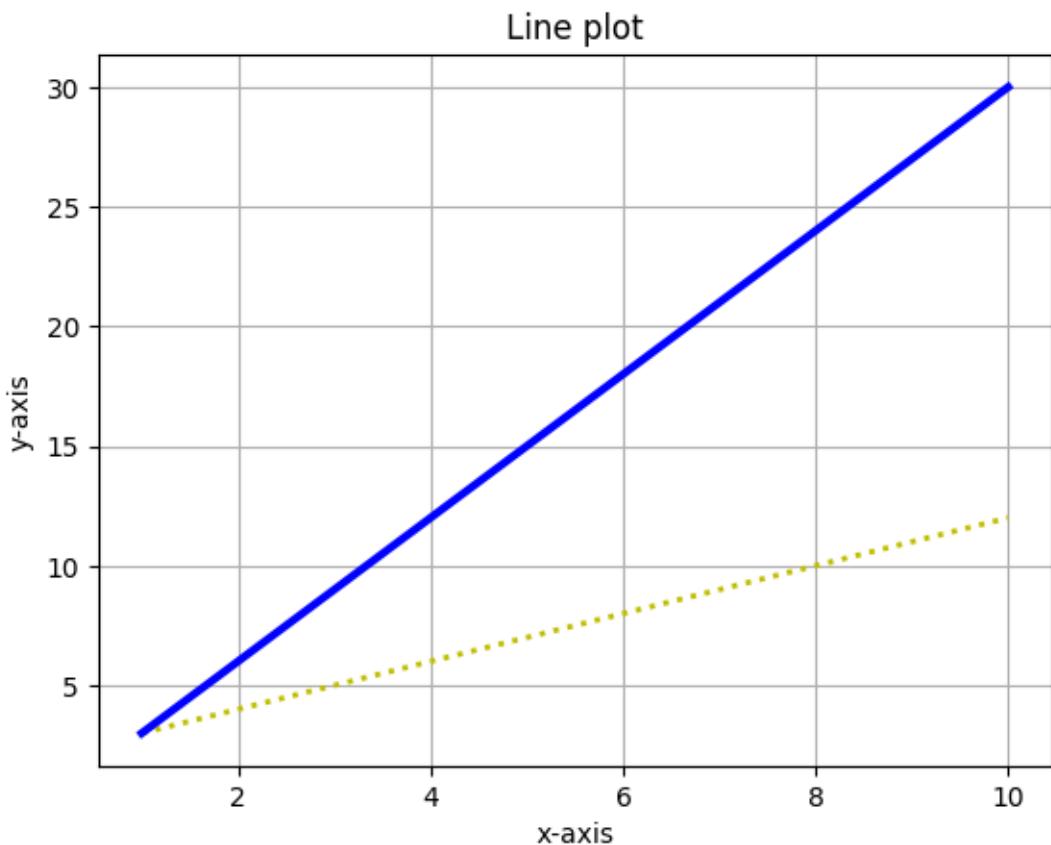
```
[ ]: x=[1,2,3,4,5]
y=[15,30,33,25,45]
plt.plot(x,y)
plt.title('simple line plot')
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
```



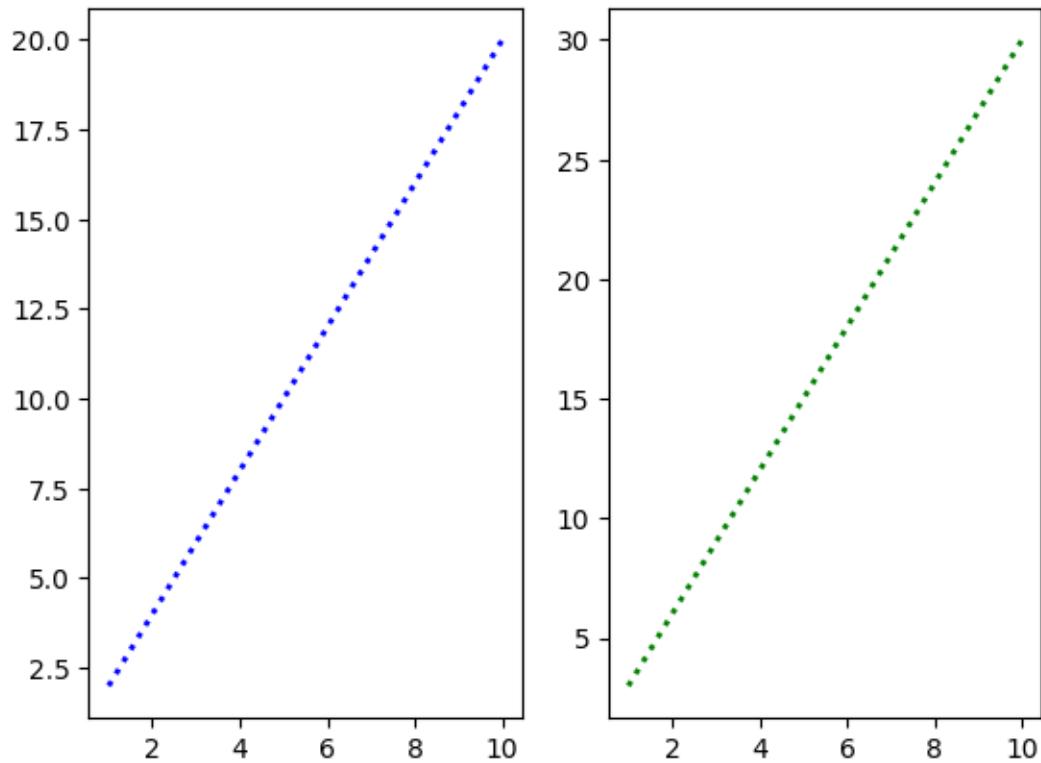
```
[ ]: plt.plot(x,y,color='red',linestyle='--',marker='o')
plt.title('styled line plot')
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.show()
```



```
[ ]: x=np.arange(1,11)
y1=2+x
y2=3*x
plt.plot(x,y1,color='y',linestyle=':',linewidth=2)
plt.plot(x,y2,color='b',linestyle='-',linewidth=3)
plt.title("Line plot")
plt.xlabel("x-axis")
plt.ylabel("y-axis")
plt.grid(True)
plt.show()
```

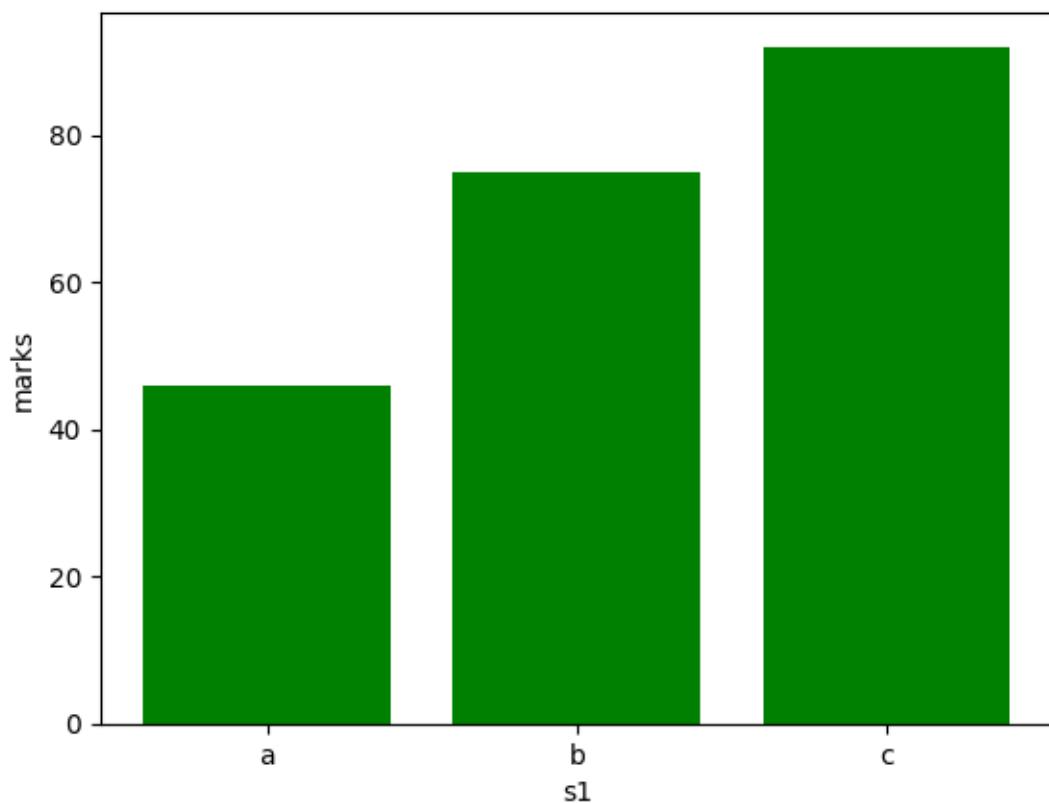


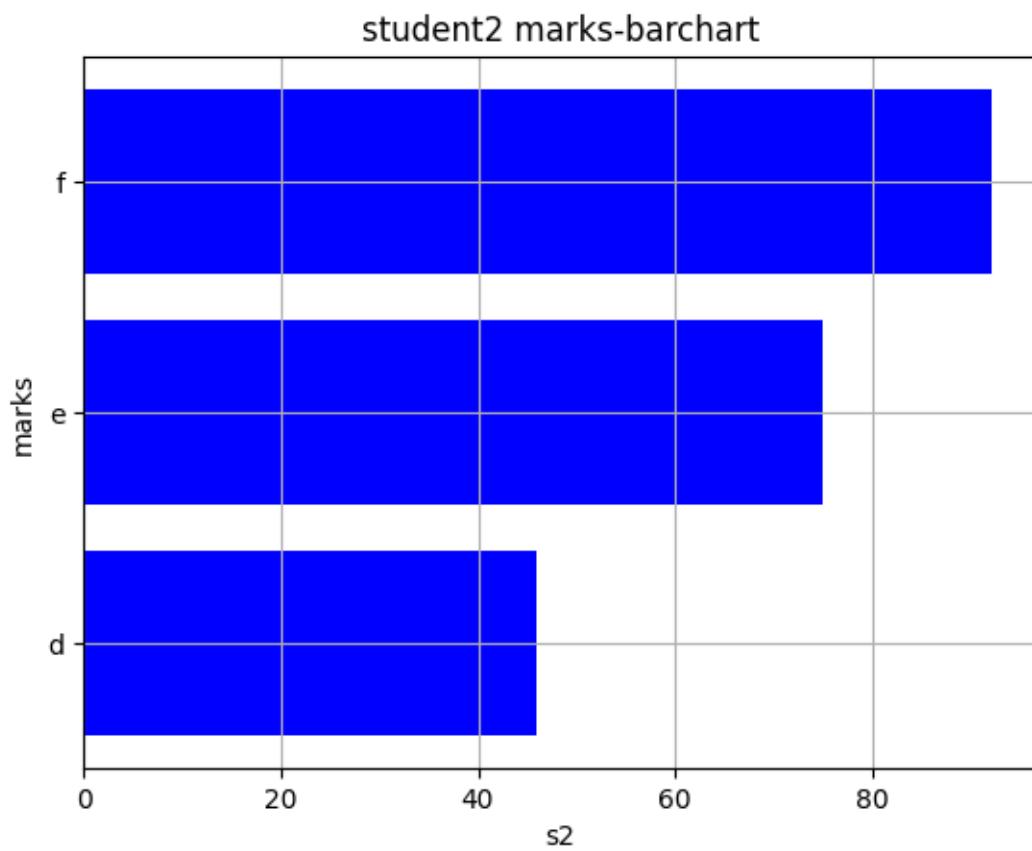
```
[ ]: #adds a subplot to a current figure
x=np.arange(1,11)
y1=2*x
y2=3*x
plt.subplot(1,2,1)
plt.plot(x,y1,color='b',linestyle=':',linewidth=2)
plt.subplot(1,2,2)
plt.plot(x,y2,color='g',linestyle=':',linewidth=2)
plt.show()
```



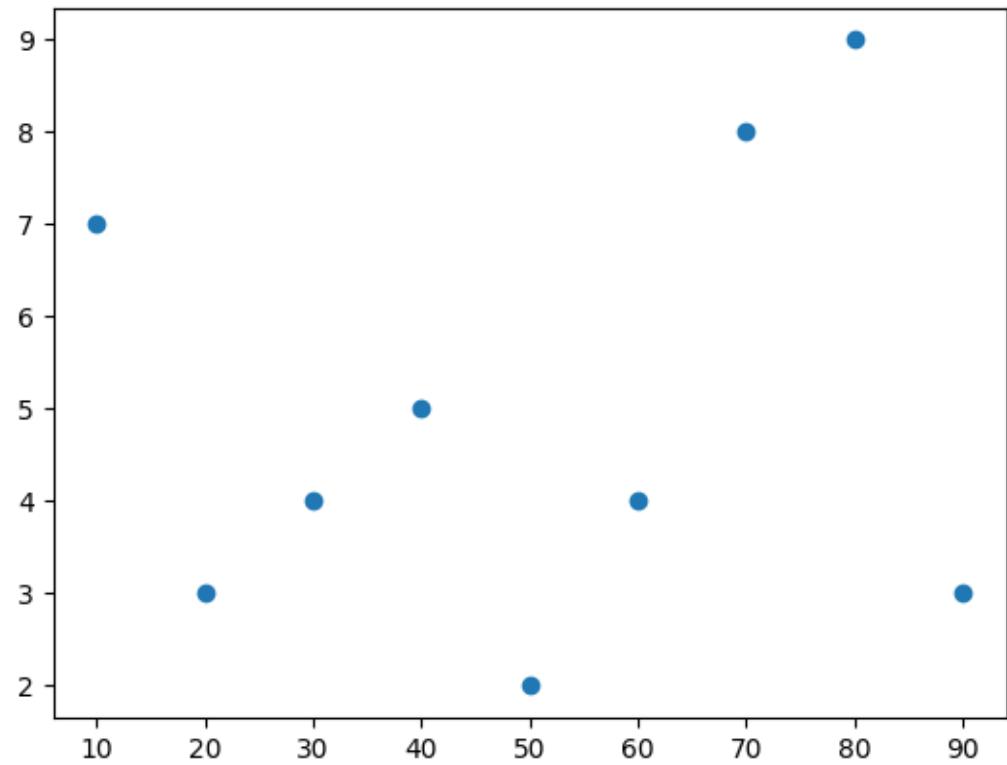
```
[ ]: s1=['a','b','c']
      s2=['d','e','f']
      marks=[46,75,92]
      plt.bar(s1,marks,color='g')
      plt.title('Student1 marks-barchart')
      plt.xlabel('s1')
      plt.ylabel('marks')
      plt.show()
      plt.bard(s2,marks,color='b')
      plt.title('student2 marks-barchart')
      plt.xlabel('s2')
      plt.ylabel('marks')
      plt.grid(True)
      plt.show()
```

Student1 marks-barchart

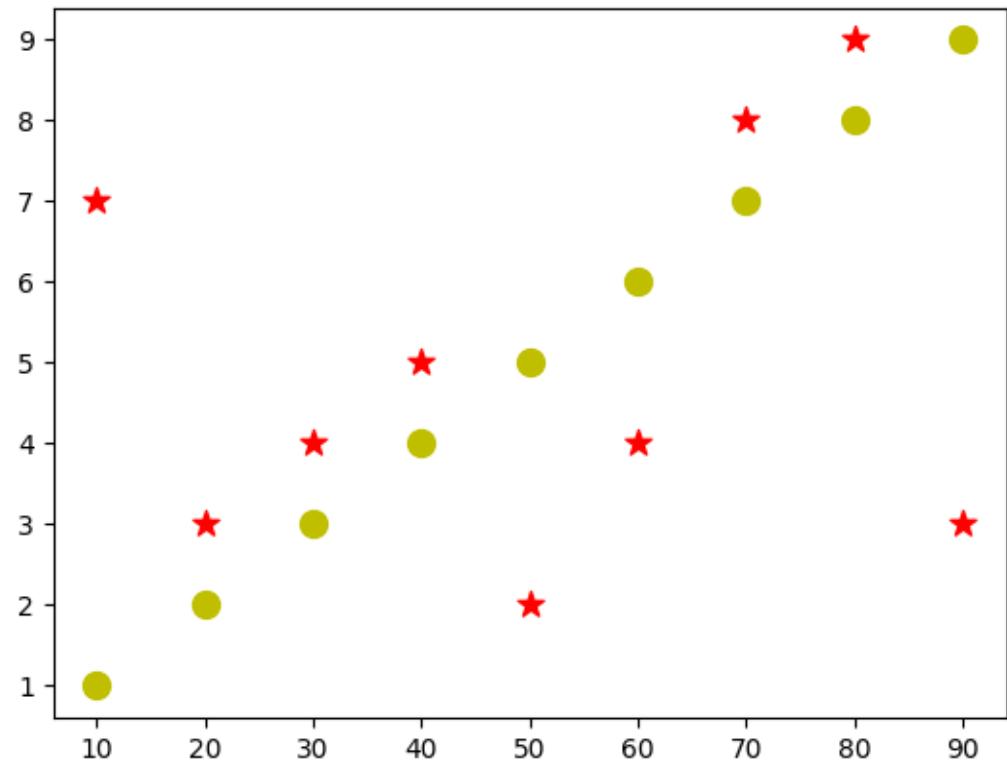




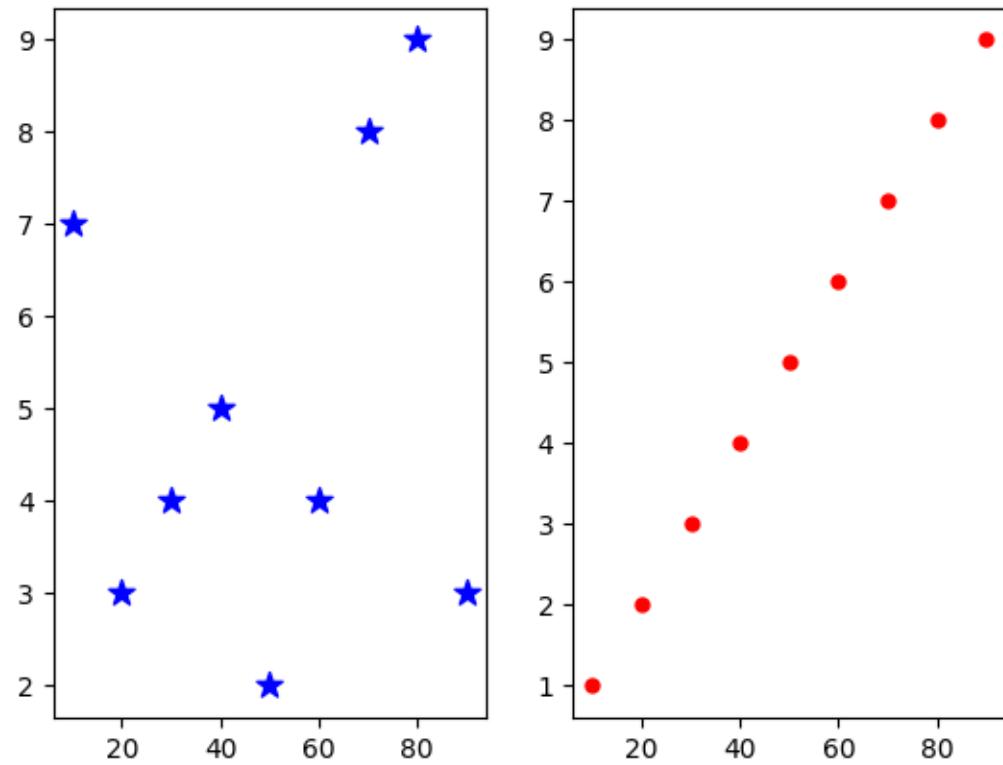
```
[ ]: x=[10,20,30,40,50,60,70,80,90]
a=[7,3,4,5,2,4,8,9,3]
plt.scatter(x,a)
plt.show()
```



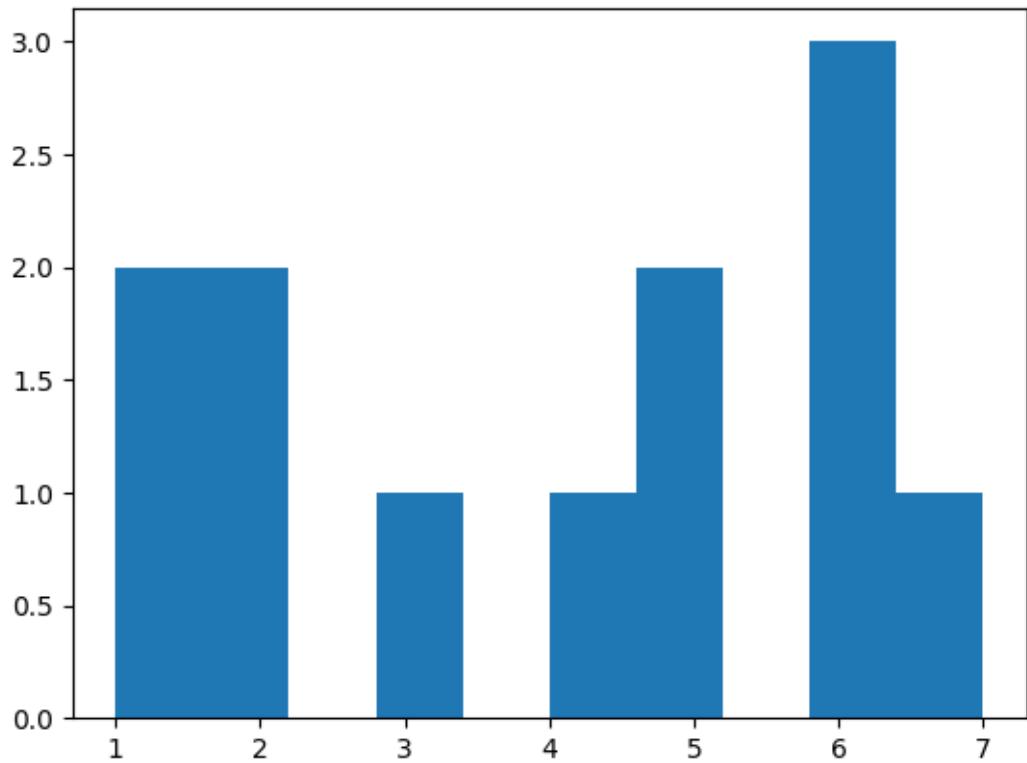
```
[ ]: b=[1,2,3,4,5,6,7,8,9]
plt.scatter(x,a,marker="*",c='r',s=100)
plt.scatter(x,b,marker='o',c='y',s=100)
plt.show()
```



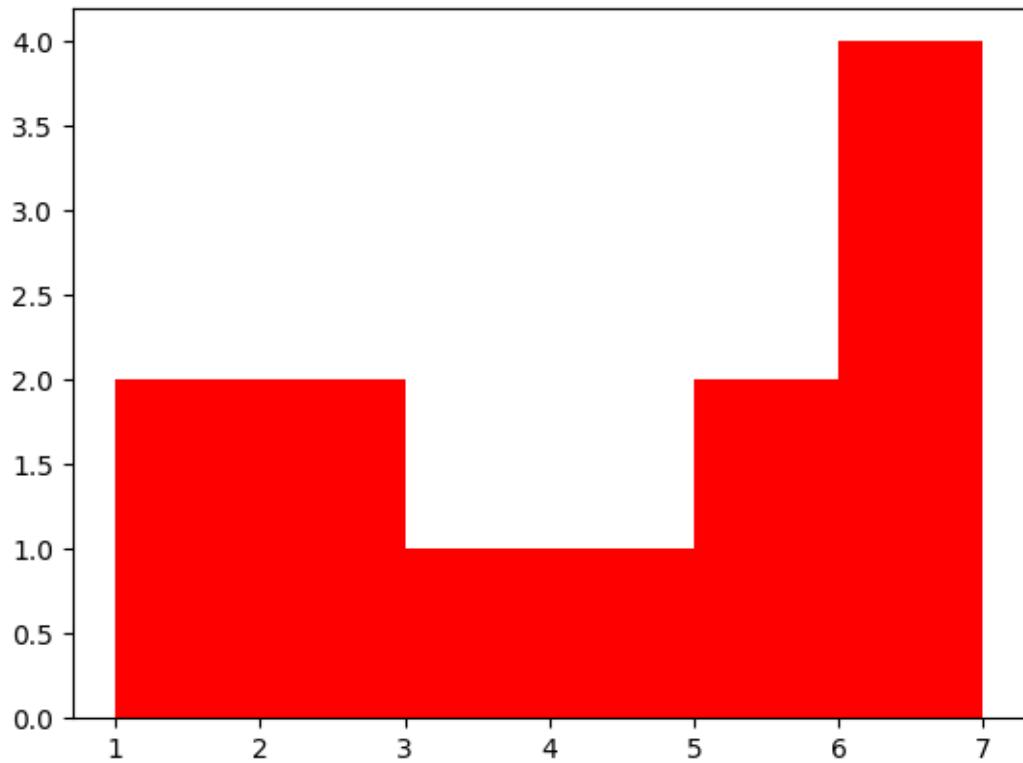
```
[ ]: plt.subplot(1,2,1)
plt.scatter(x,a,marker="*",c="b",s=100)
plt.subplot(1,2,2)
plt.scatter(x,b,marker=".",c="r",s=100)
plt.show()
```



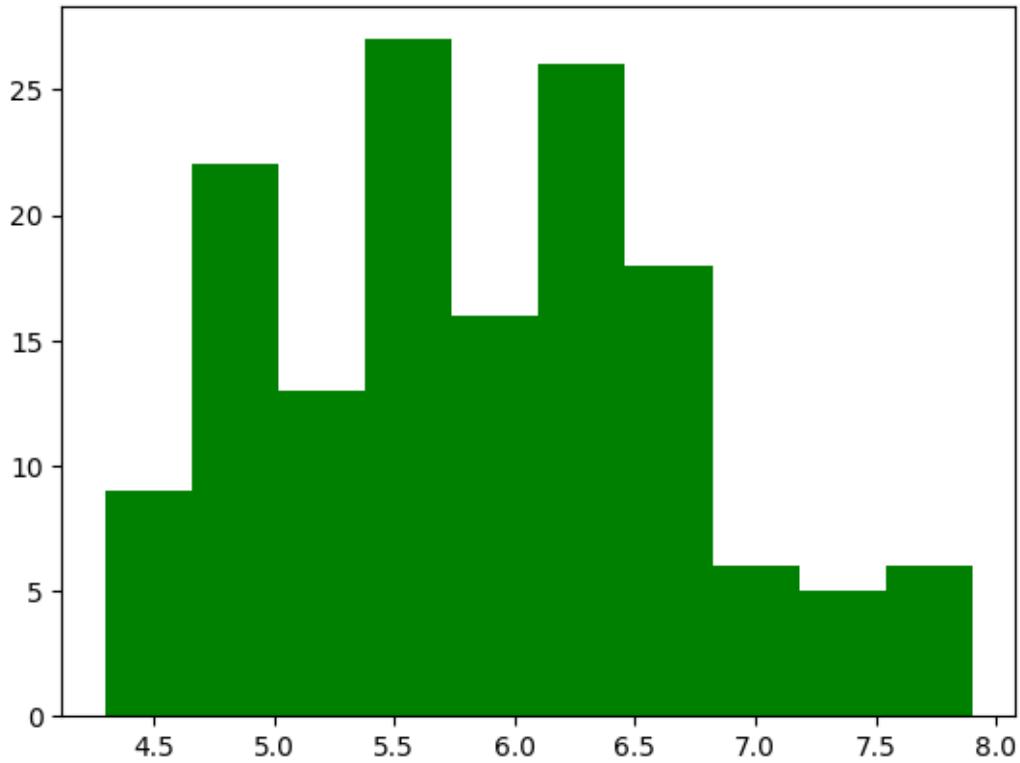
```
[ ]: data=[1,2,3,4,5,6,6,2,5,6,7,1]
plt.hist(data)
plt.show()
```



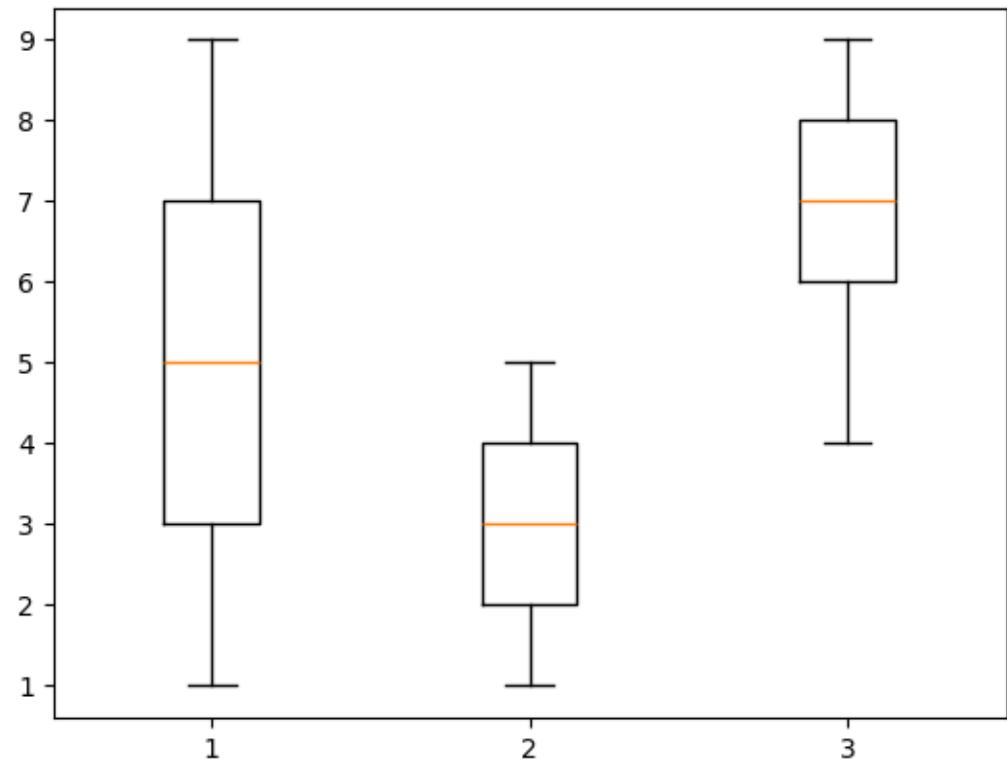
```
[ ]: plt.hist(data,color="r",bins=6)
plt.show()
```



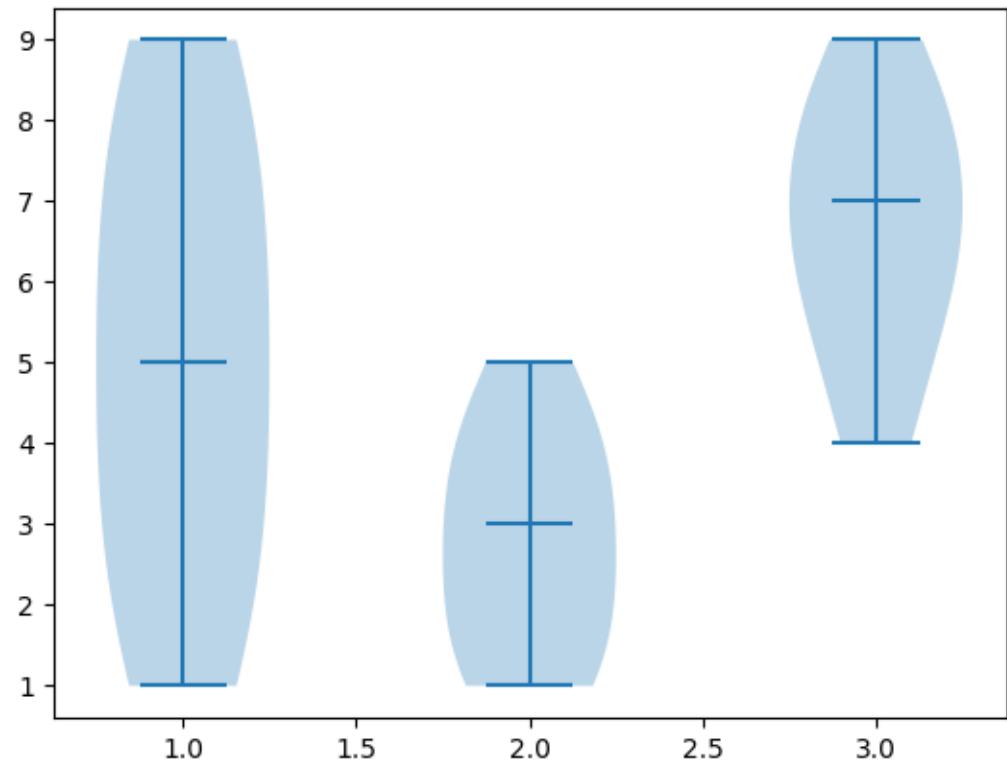
```
[ ]: #histogram of iris data - sepallength  
plt.hist(df['sepal_length'],color='g',bins=10)  
plt.show()
```



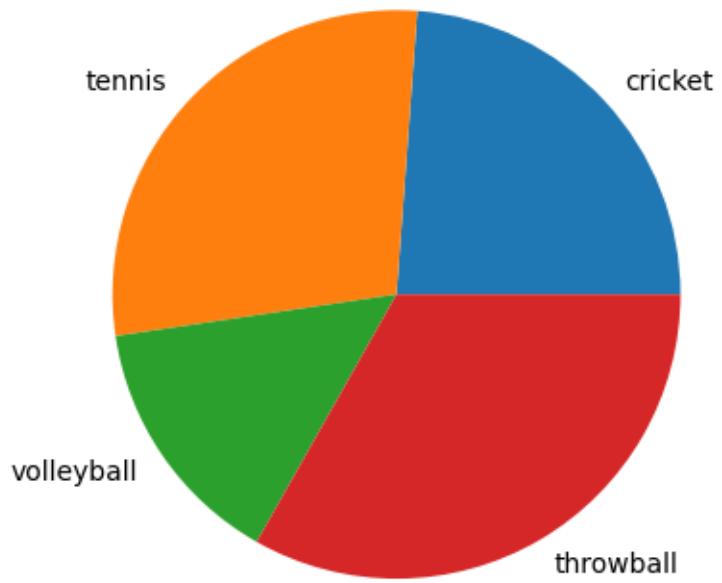
```
[ ]: one=[1,2,3,4,5,6,7,8,9]
two=[1,2,3,4,5,4,3,2,1]
three=[6,7,8,9,8,7,6,5,4]
data=list([one,two,three])
plt.boxplot(data)
plt.show()
```



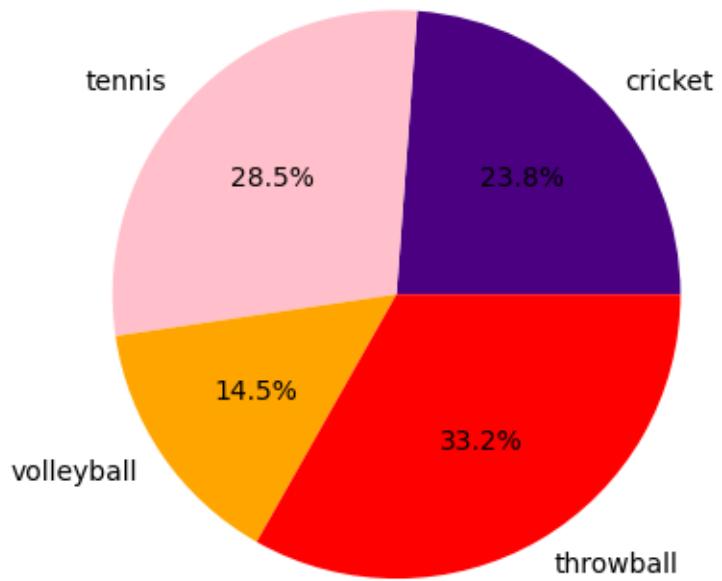
```
[ ]: plt.violinplot(data, showmedians=True)
plt.show()
```



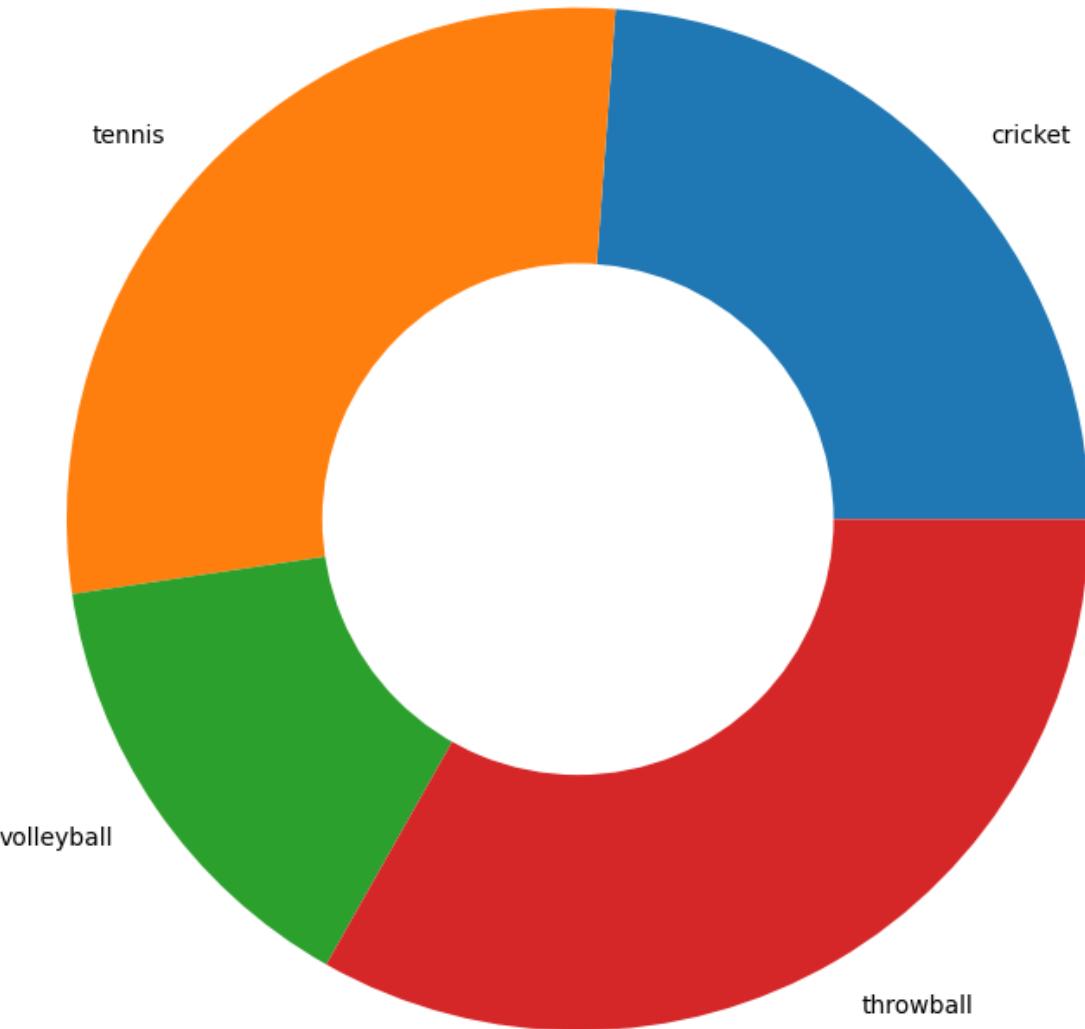
```
[ ]: games=['cricket','tennis','volleyball','throwball']
quantity=[56,67,34,78]
plt.pie(quantity,labels=games)
plt.show()
```



```
[ ]: #userdefined colours piechart
plt.pie(quantity,labels=games,autopct='%.0.
˓→1f%%',colors=['indigo','pink','orange','red'])
plt.show()
```



```
[ ]: #doughnut chart
plt.pie(quantity,labels=games,radius=2)
plt.pie([1],colors=['w'],radius=1)
plt.show()
```

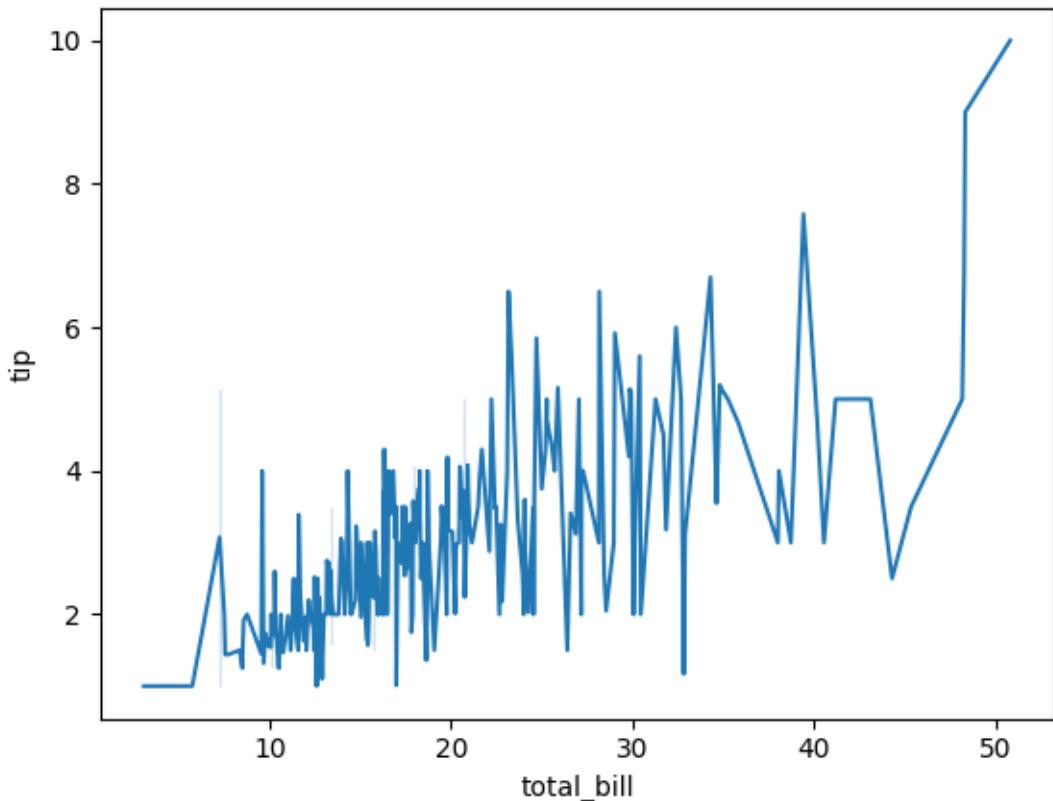


5 SEABORN

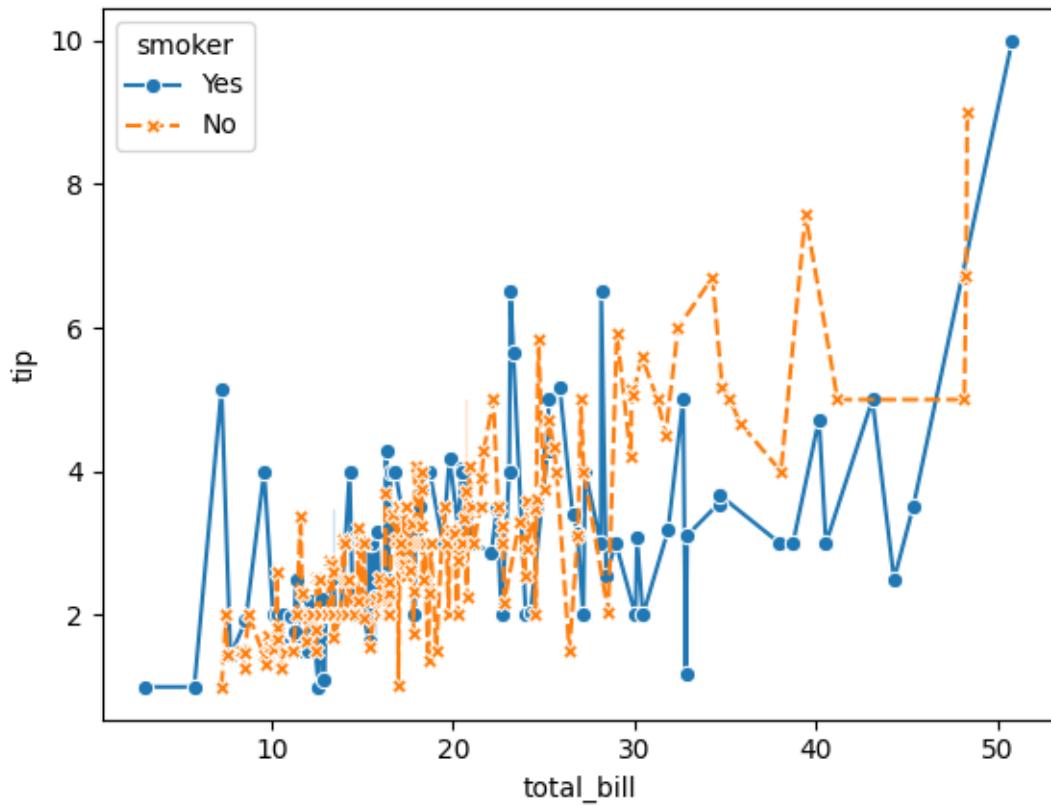
```
[ ]: import seaborn as sns
[ ]: print(sns.get_dataset_names())
['anagrams', 'anscombe', 'attention', 'brain_networks', 'car_crashes',
'diamonds', 'dots', 'dowjones', 'exercise', 'flights', 'fmri', 'geyser', 'glue',
'healthexp', 'iris', 'mpg', 'penguins', 'planets', 'seoice', 'taxis', 'tips',
'titanic']
[ ]: sb=sns.load_dataset('tips')
sb.head()
```

```
[ ]: 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
```

```
[ ]: sns.lineplot(x='total_bill',y='tip',data=sb)
plt.show()
```



```
[ ]: sns.
  ↪lineplot(x='total_bill',y='tip',hue='smoker',style='smoker',markers=True,data=sb)
plt.show()
```

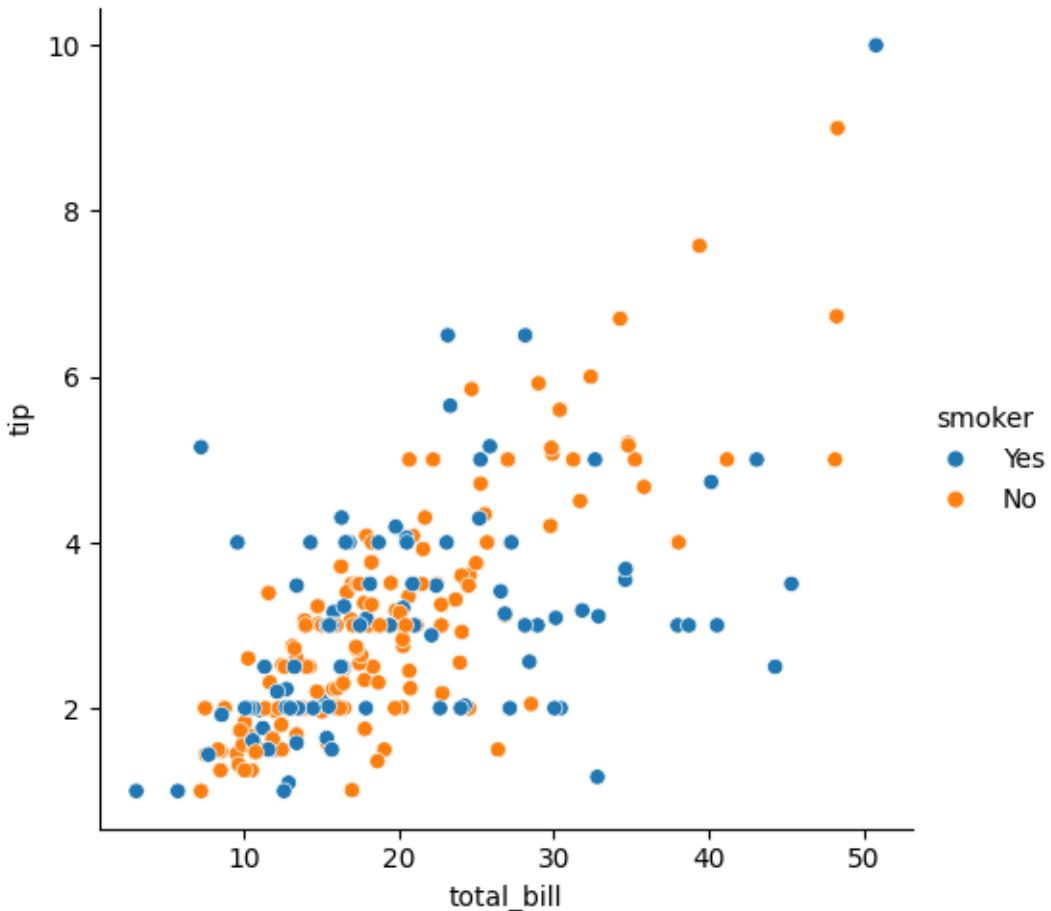


```
[ ]: sb.shape
```

```
[ ]: (244, 7)
```

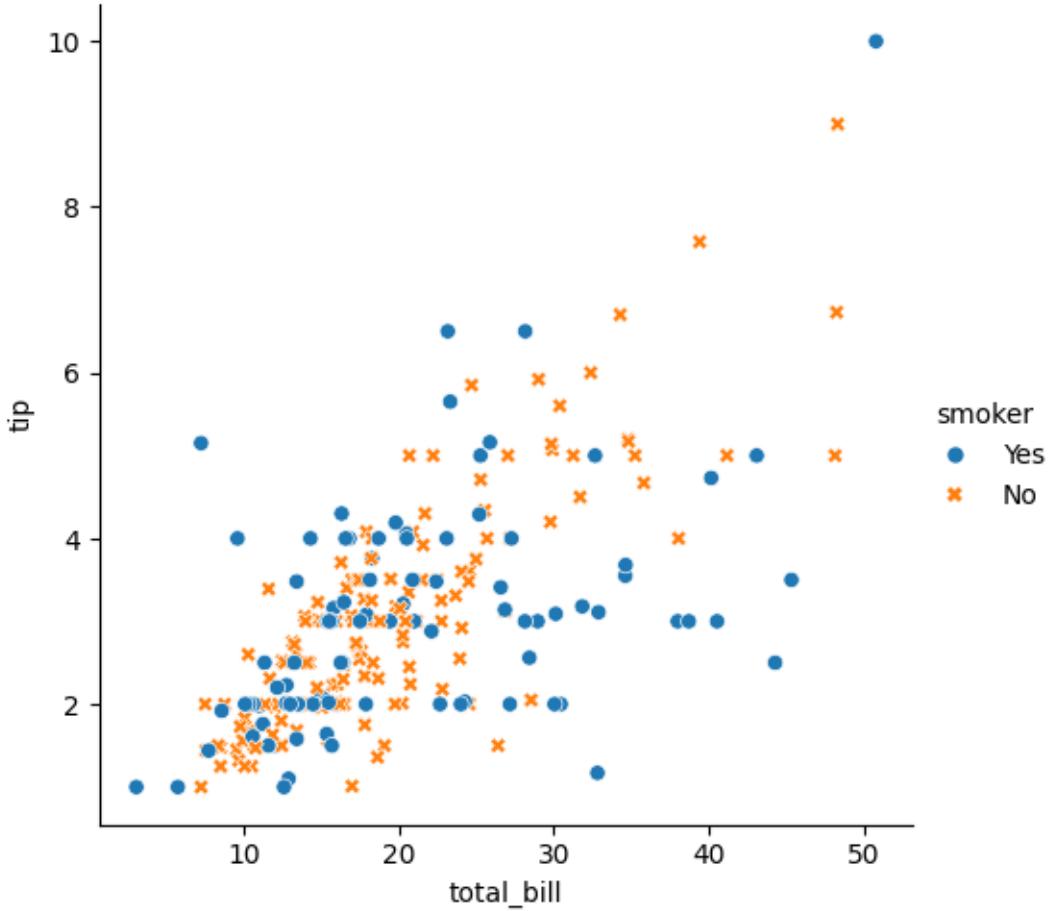
```
[ ]: #visualizing statistical relationships
      sns.relplot(data=sb,x='total_bill',y='tip',hue='smoker')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb57a0bda30>
```



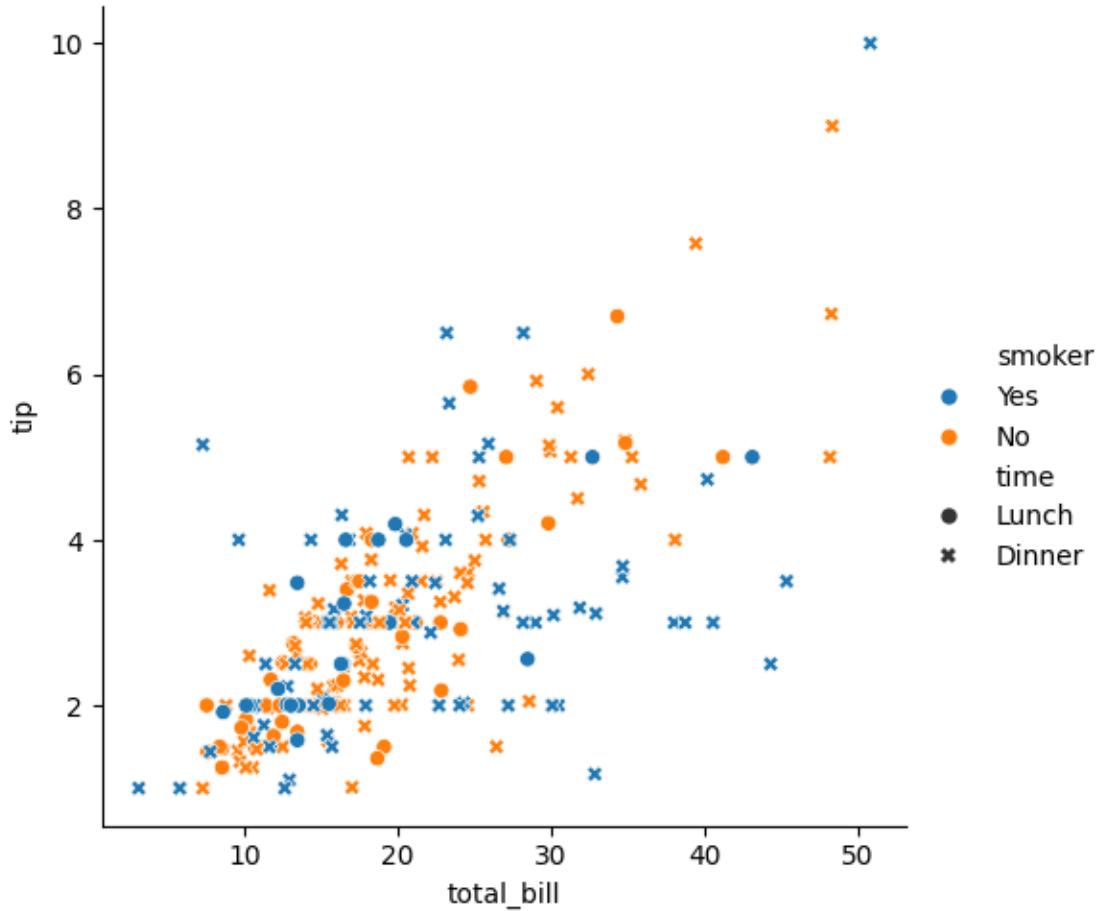
```
[ ]: #marker  
sns.relplot(data=sb,x='total_bill',y='tip',hue='smoker',style='smoker')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb57981dbb0>
```



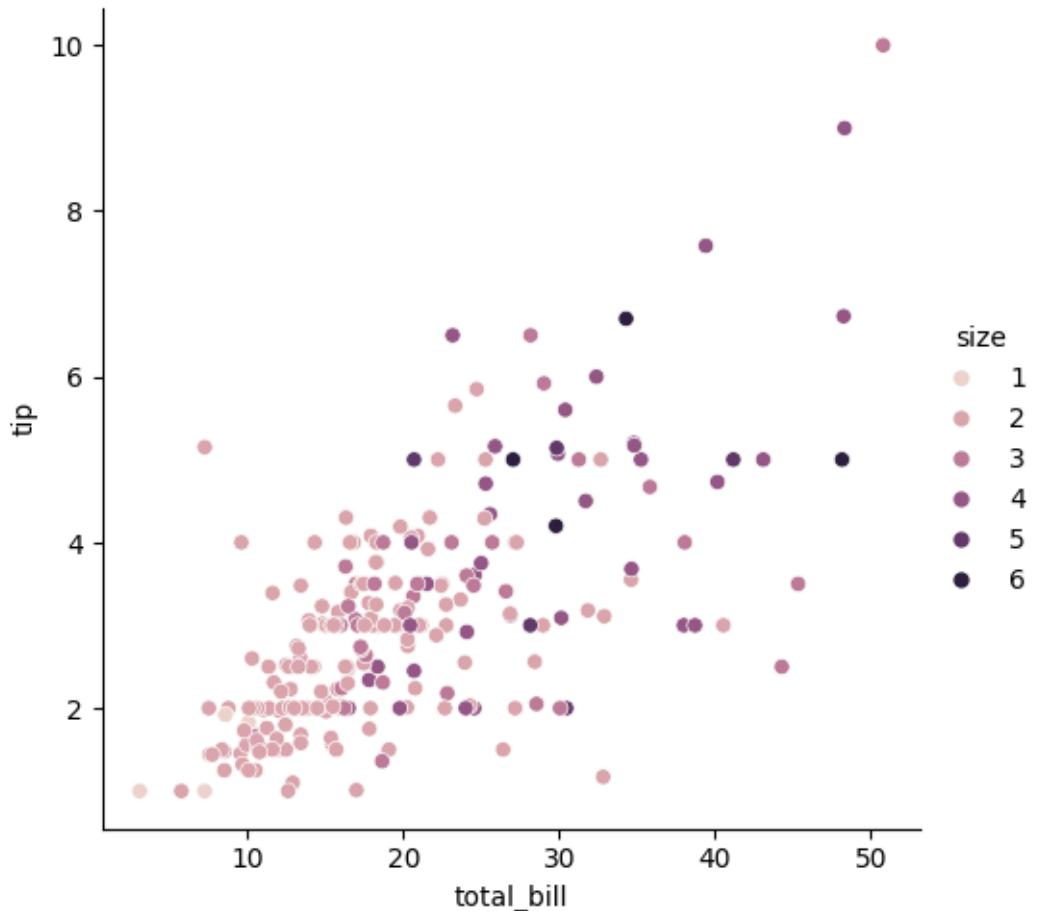
```
[ ]: sns.relplot(data=sb,x="total_bill",y="tip",hue="smoker",style="time")
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb578cc2870>
```



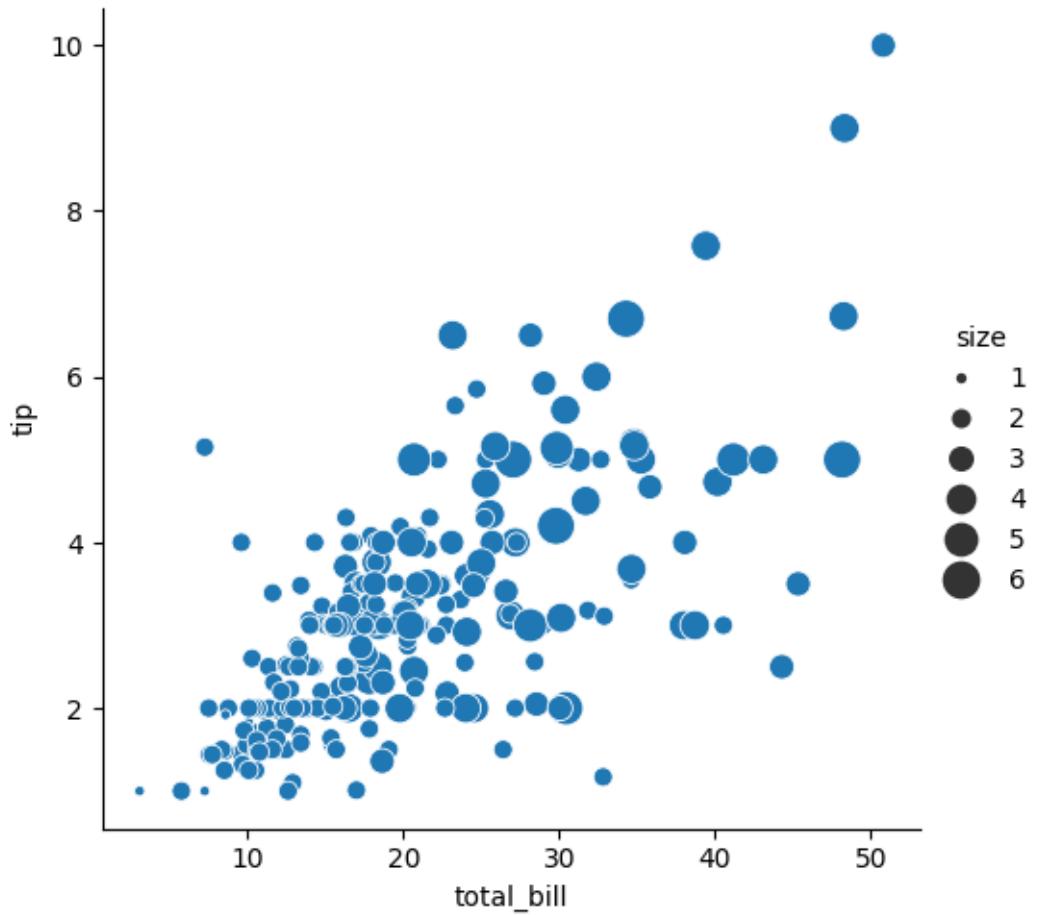
```
[ ]: sns.relplot(data=sb,x="total_bill",y="tip",hue="size",)
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb57981cd40>
```



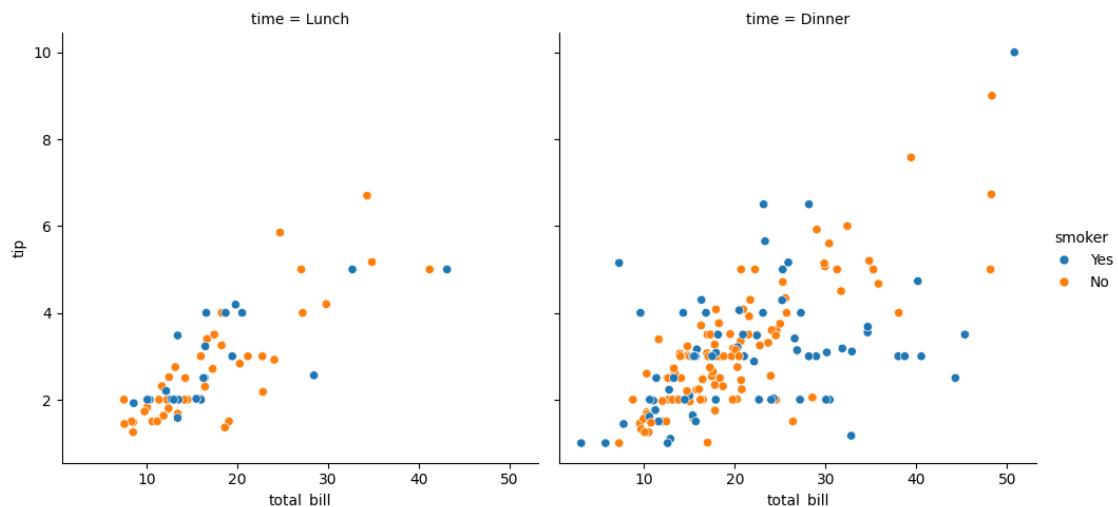
```
[ ]: sns.relplot(data=sb,x="total_bill",y="tip",size="size",sizes=(15,200))
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56f66dc70>
```



```
[ ]: sns.relplot(data=sb,x="total_bill",y="tip",hue="smoker",col="time",)
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb579d01a30>
```



```
[ ]: fmri=sns.load_dataset('fmri')
fmri.head()
```

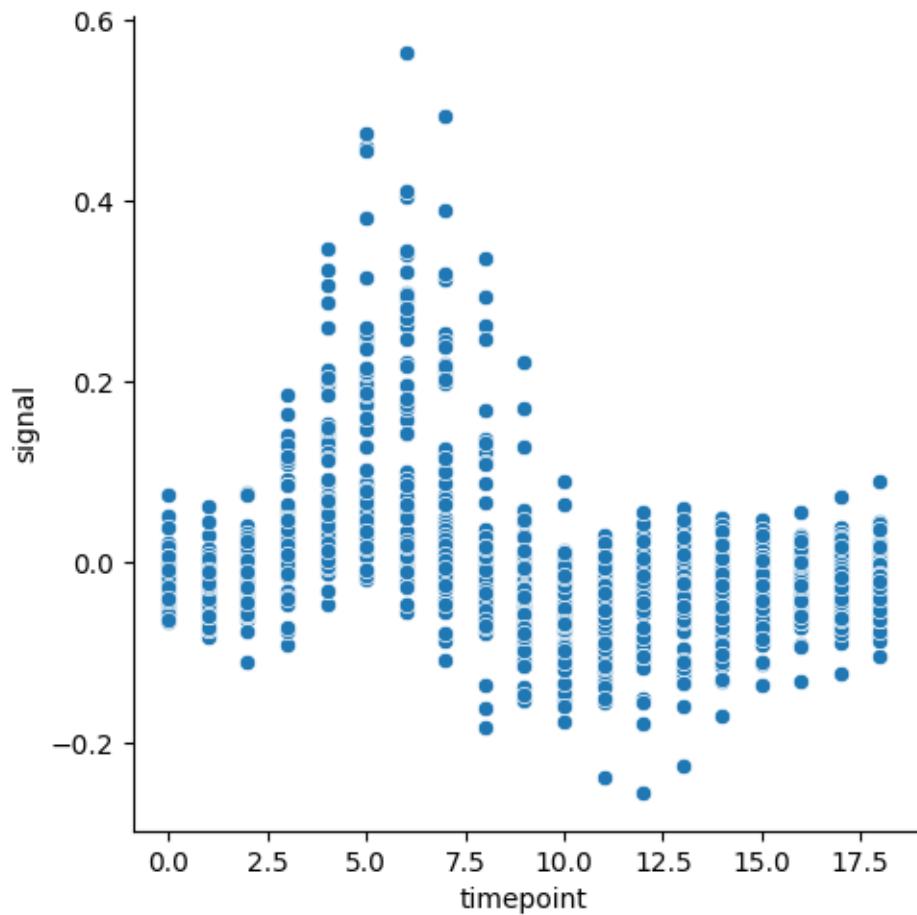
```
[ ]:   subject  timepoint  event    region    signal
 0      s13        18  stim  parietal -0.017552
 1      s5         14  stim  parietal -0.080883
 2      s12        18  stim  parietal -0.081033
 3      s11        18  stim  parietal -0.046134
 4      s10        18  stim  parietal -0.037970
```

```
[ ]: fmri.shape
```

```
[ ]: (1064, 5)
```

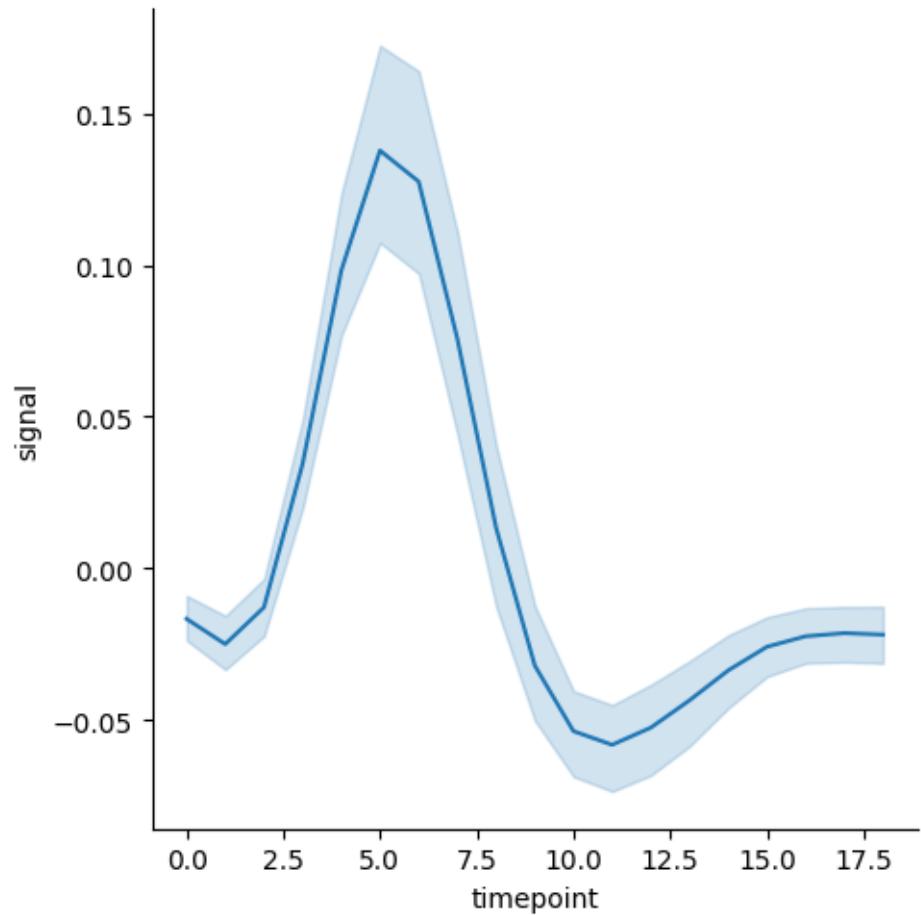
```
[ ]: sns.relplot(data=fmri,x='timepoint',y='signal')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56da34d40>
```



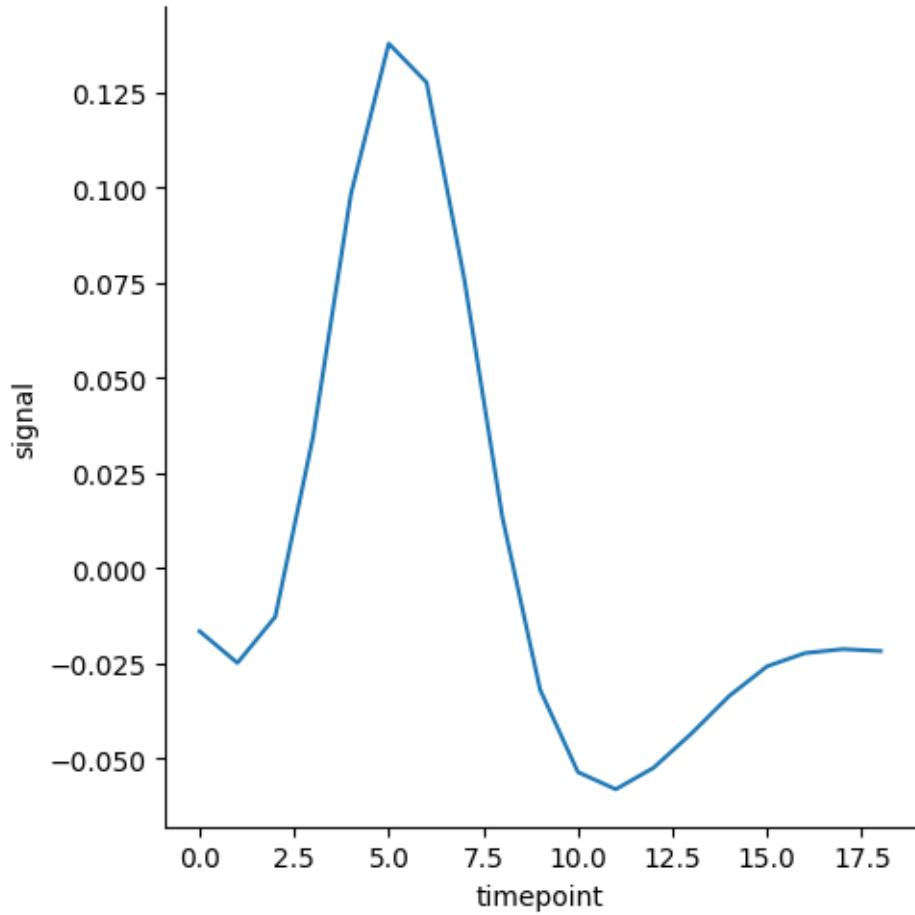
```
[ ]: sns.relplot(data=fMRI,x='timepoint',y='signal',kind='line')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56db13320>
```



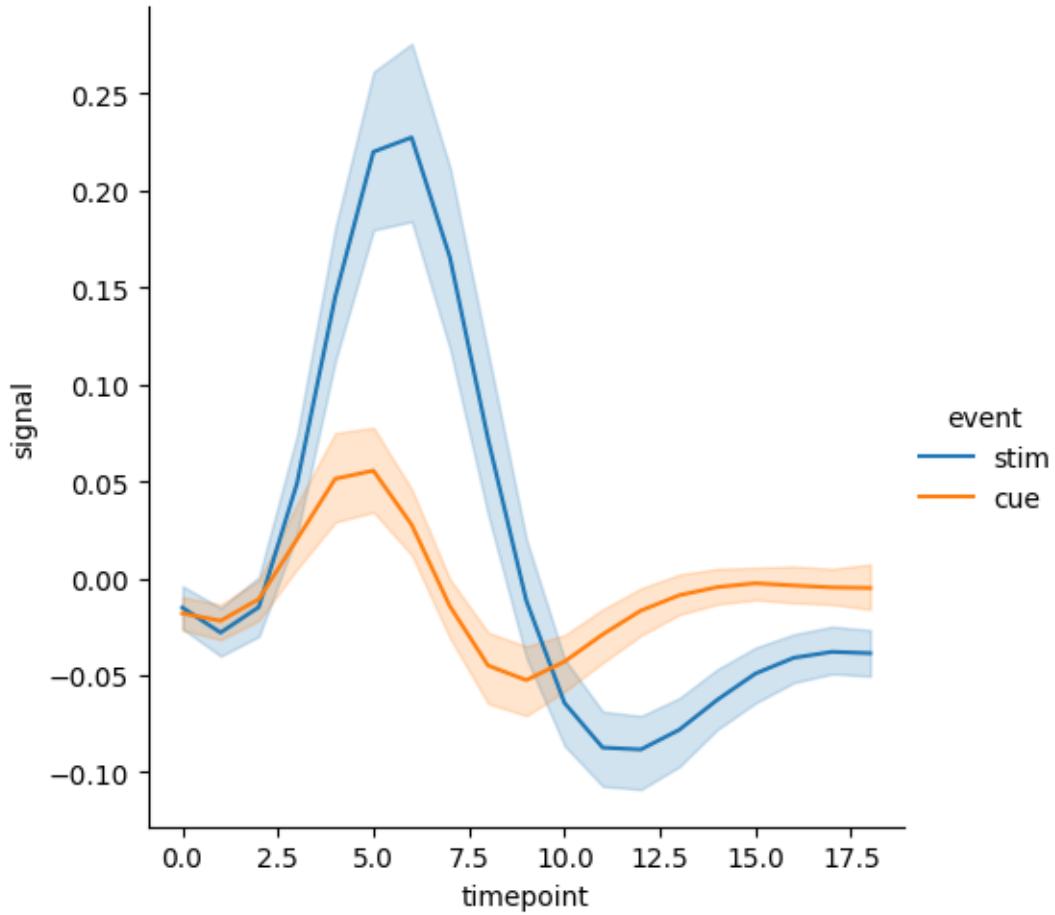
```
[ ]: #remove error band  
sns.relplot(data=fMRI,kind='line',x='timepoint',y='signal',errorbar=None)
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56f675c70>
```

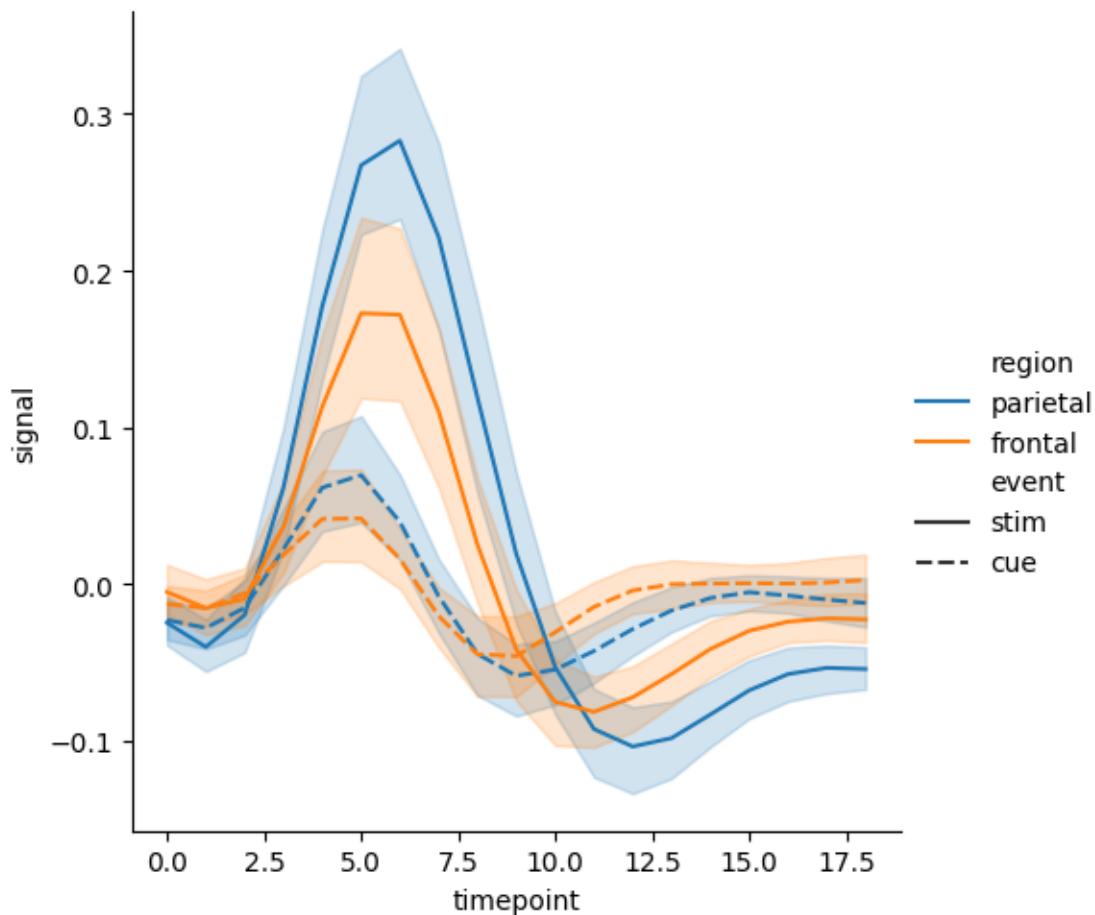


```
[ ]: #plotting into two lines and error bands
sns.relplot(data=fMRI,kind='line',x='timepoint',y='signal',hue='event')
```

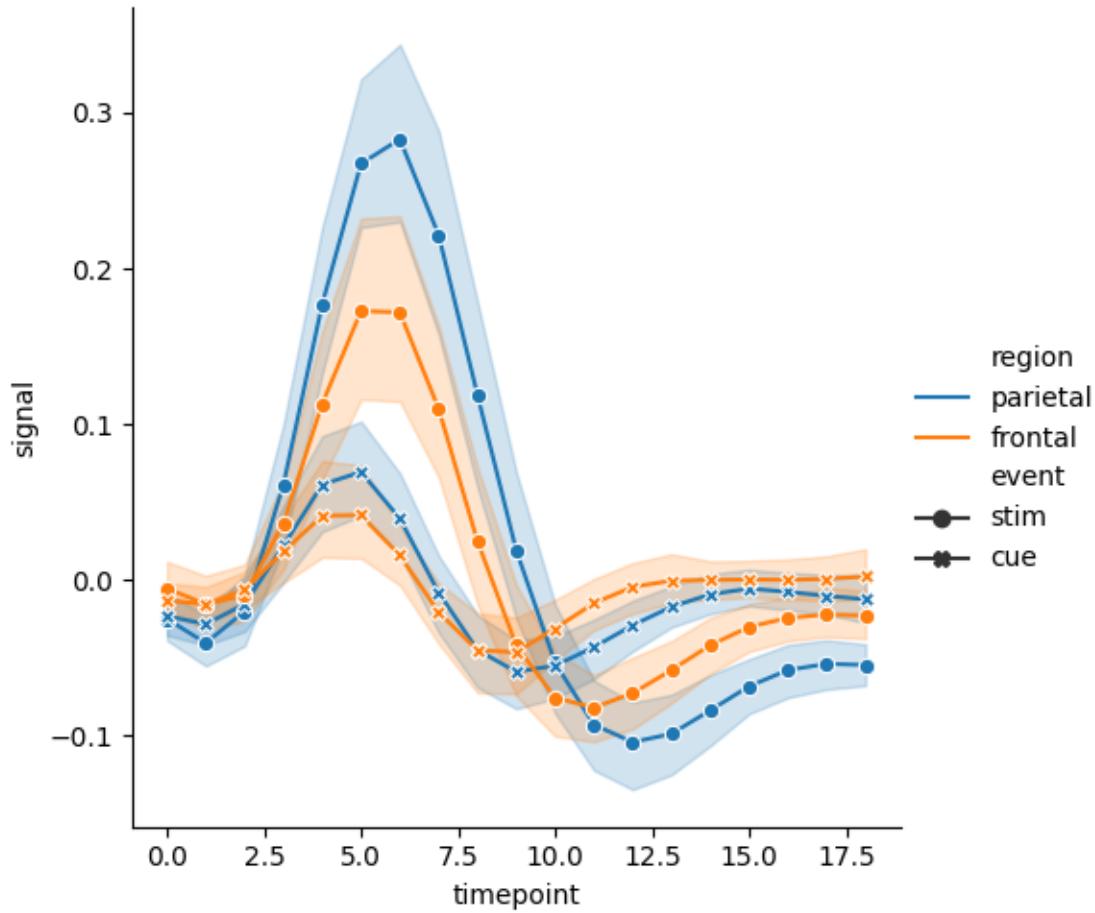
```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56f674d40>
```



```
[ ]: sns.  
     ↪relplot(data=fmri,kind='line',x='timepoint',y='signal',hue='region',style='event')  
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56da75250>
```



```
[ ]: sns.  
     ↵relplot(data=fMRI,kind='line',x='timepoint',y='signal',hue='region',style='event',dashes=Fa  
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56dabcad0>
```

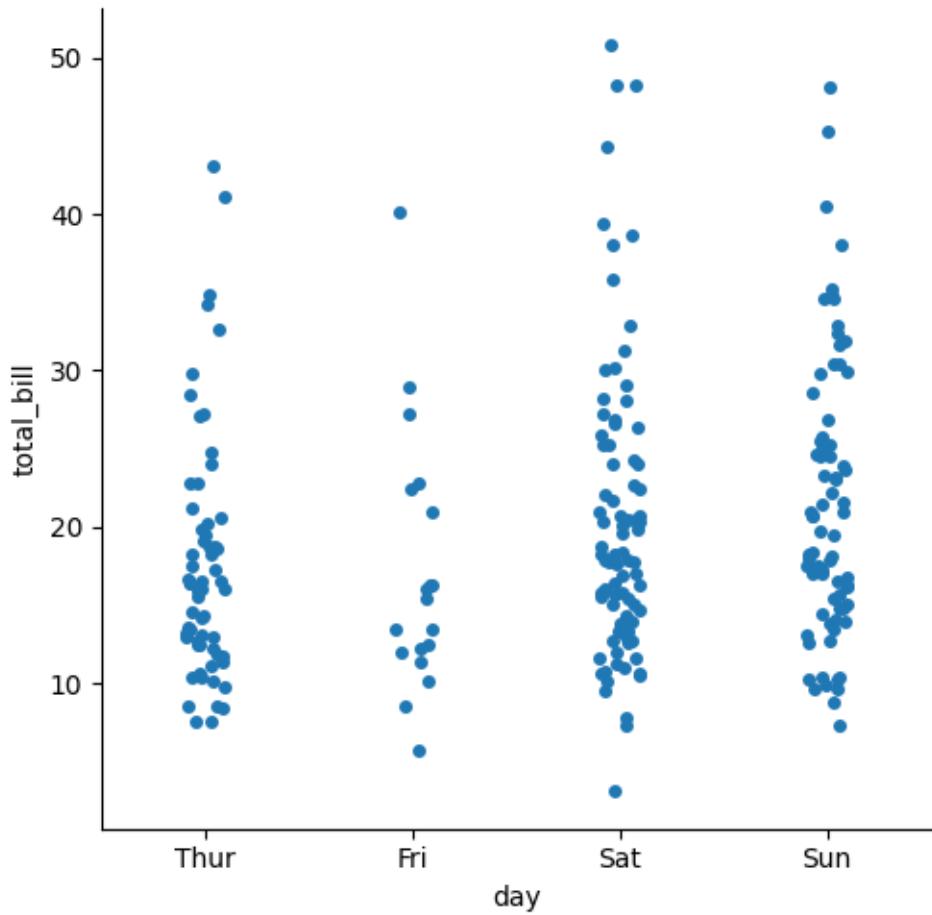


```
[ ]: tips=sns.load_dataset('tips')
      tips.head()
```

```
[ ]:   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
```

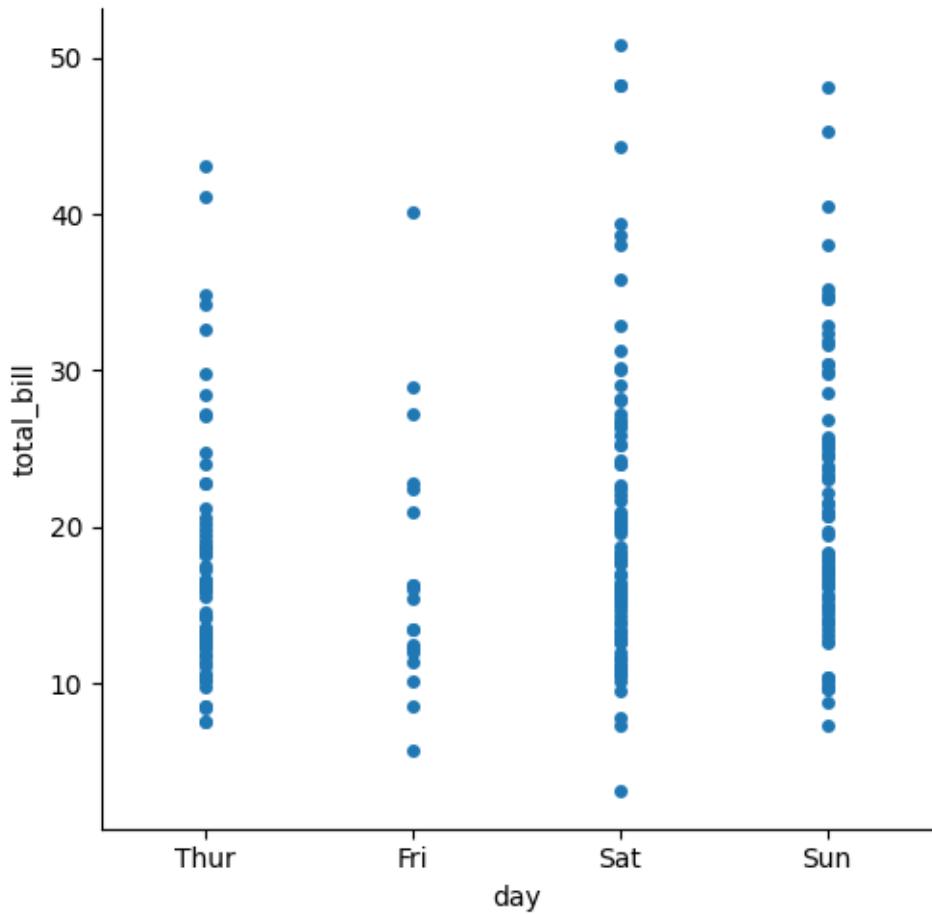
```
[ ]: #visualizing categorical data
      sns.catplot(data=tips,x='day',y='total_bill')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56f438e00>
```



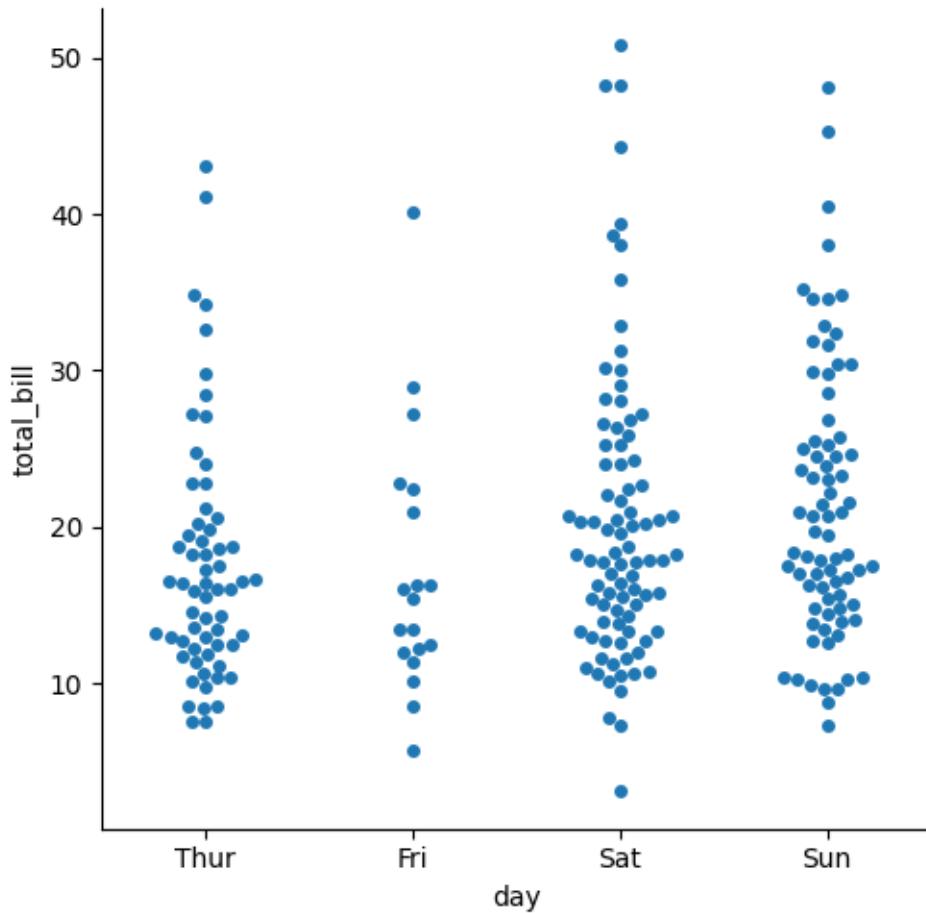
```
[ ]: #jitter  
sns.catplot(data=tips,x='day',y='total_bill',jitter=False)
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56d798470>
```



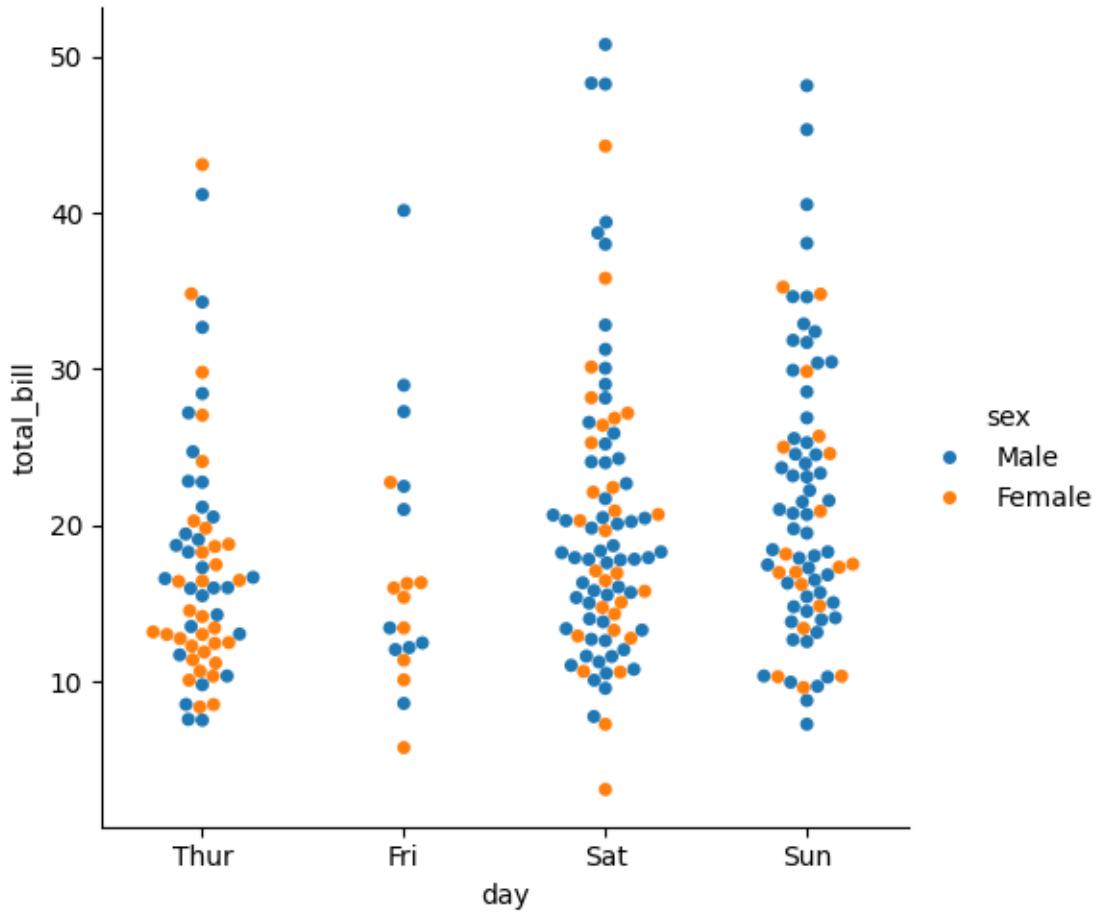
```
[ ]: #swarm plot  
sns.catplot(data=tips,x='day',y='total_bill',kind='swarm')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb57995bd70>
```



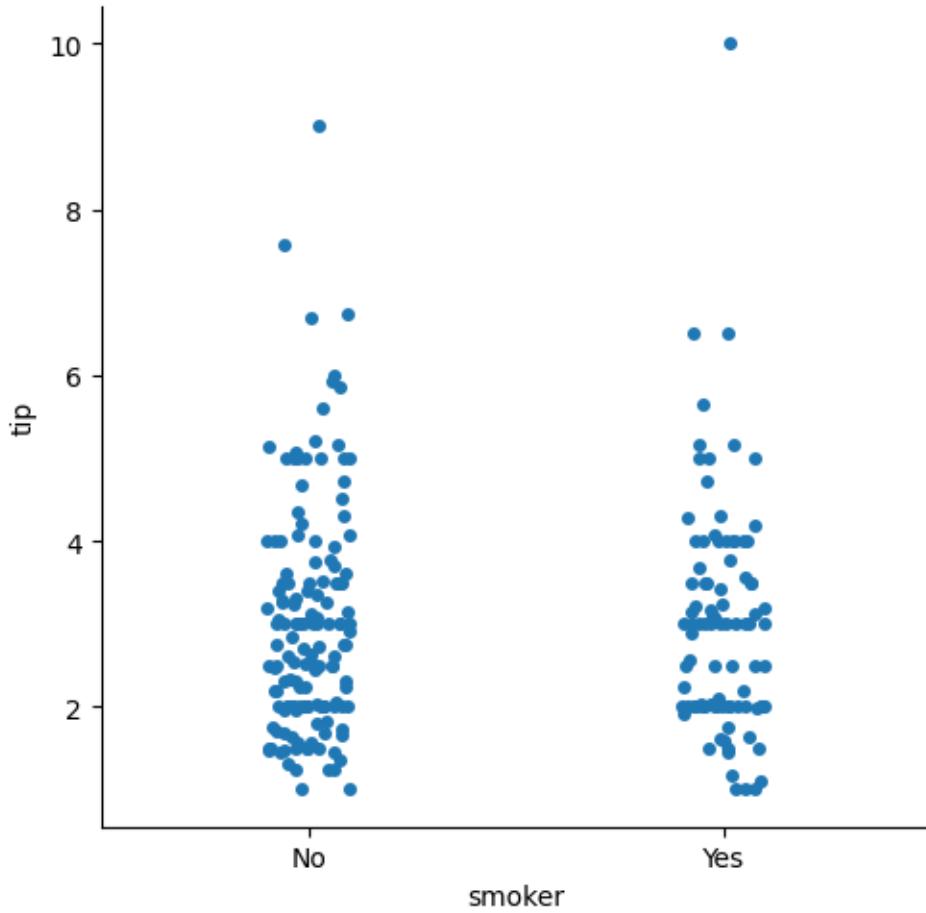
```
[ ]: sns.catplot(data=tips,x='day',y='total_bill',hue='sex',kind='swarm')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb5793da210>
```



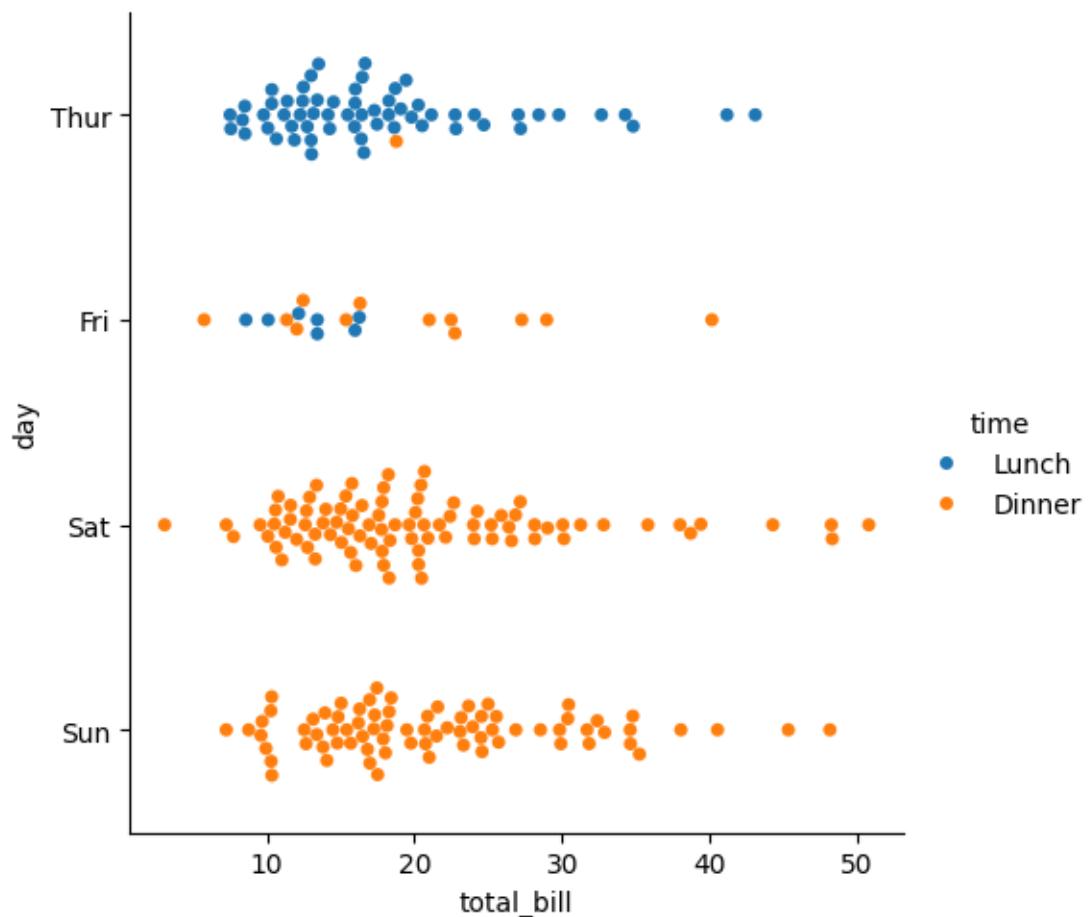
```
[ ]: sns.catplot(data=tips,x='smoker',y='tip',order=['No','Yes'])
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb584425e80>
```



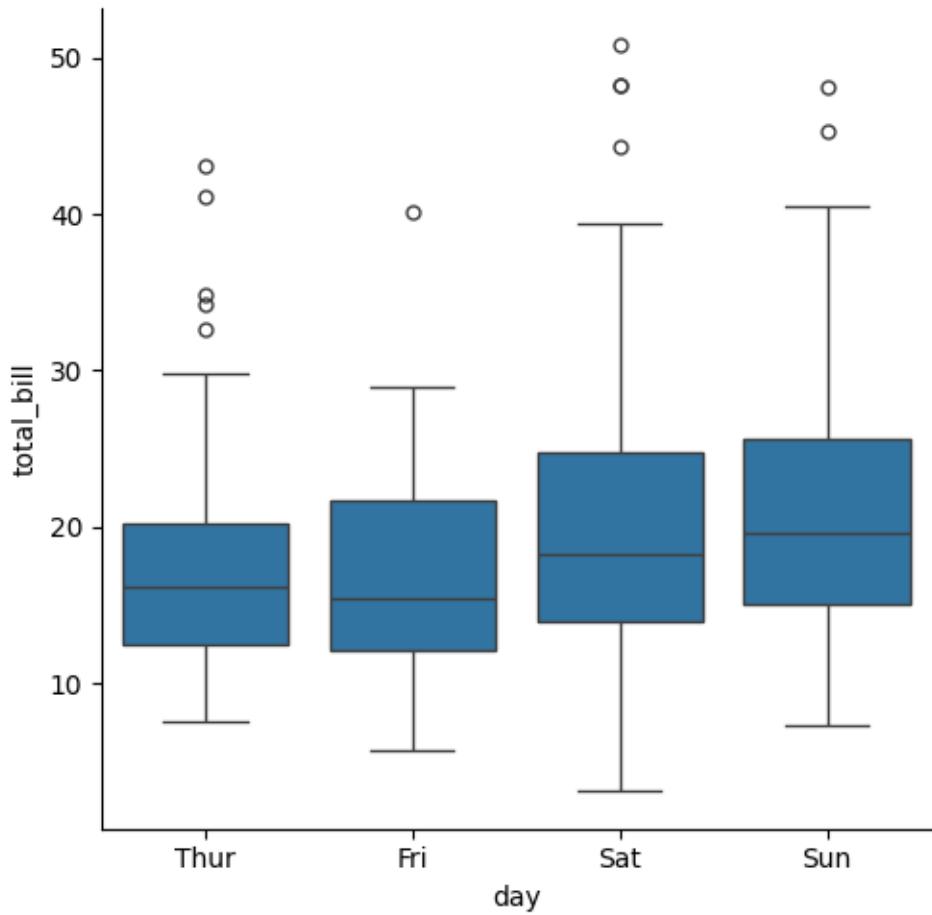
```
[ ]: sns.catplot(data=tips,x='total_bill',y='day',hue='time',kind='swarm')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56d74ecc0>
```



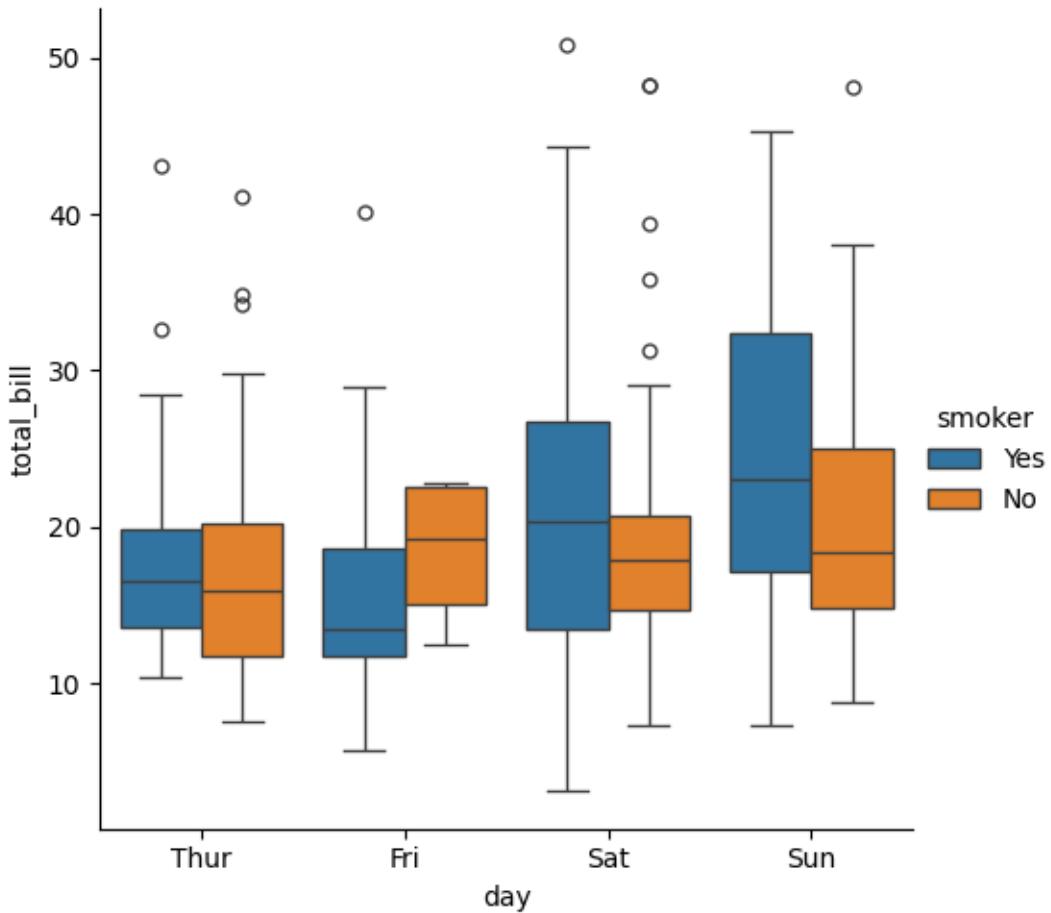
```
[ ]: sns.catplot(data=tips,x='day',y='total_bill',kind='box')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56f5070e0>
```



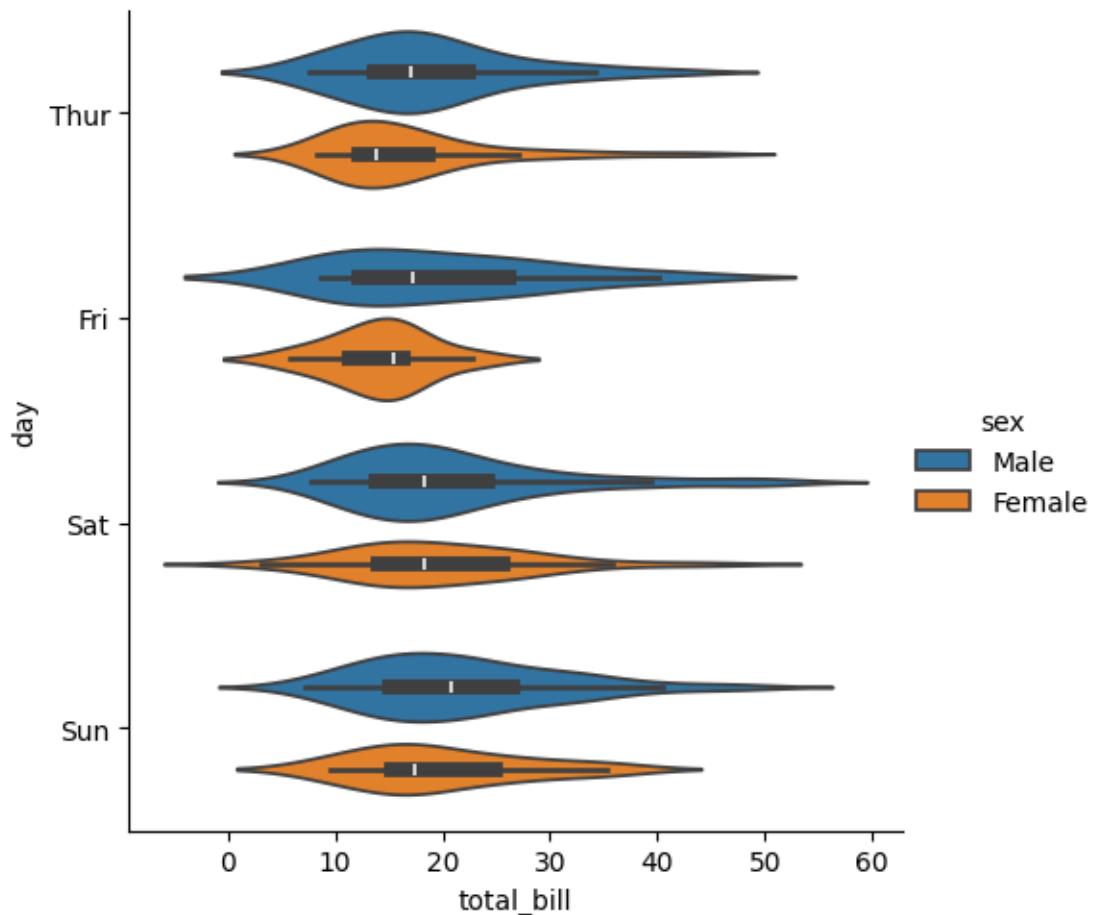
```
[ ]: sns.catplot(data=tips,x='day',y='total_bill',hue='smoker',kind='box')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56d626f00>
```



```
[ ]: sns.catplot(data=tips,x='total_bill',y='day',hue='sex',kind='violin')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56d74f650>
```



```
[ ]: sns.catplot(data=tips,x='day',y='total_bill',hue='sex',kind='violin',split=True)
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7fb56d4d0c20>
```

