

# iris-project

December 10, 2023

## IRIS DATASET

### Importing libraries and dataset

```
[ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df=pd.read_csv('/content/IRIS (2).csv')
df
```

```
[ ]:      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
..          ...           ...           ...           ...          ...
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
```

[150 rows x 5 columns]

### Data Preprocessing

```
[ ]: 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
```

```
[ ]: 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.dtypes
```

```
[ ]: sepal_length    float64
      sepal_width    float64
      petal_length    float64
      petal_width    float64
      species        object
      dtype: object
```

```
[ ]: df.columns
```

```
[ ]: Index(['sepal_length', 'sepal_width', 'petal_length', 'petal_width',
           'species'],
           dtype='object')
```

```
[ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   sepal_length    150 non-null    float64
1   sepal_width     150 non-null    float64
2   petal_length    150 non-null    float64
3   petal_width     150 non-null    float64
4   species         150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
[ ]: 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
-----	----------	----------	----------	----------

```
[ ]: df.isna().sum()
```

```
[ ]: sepal_length    0
      sepal_width    0
      petal_length   0
      petal_width    0
      species        0
      dtype: int64
```

```
[ ]: df.duplicated().sum()
```

```
[ ]: 3
```

```
[ ]: df.drop_duplicates(inplace=True)
```

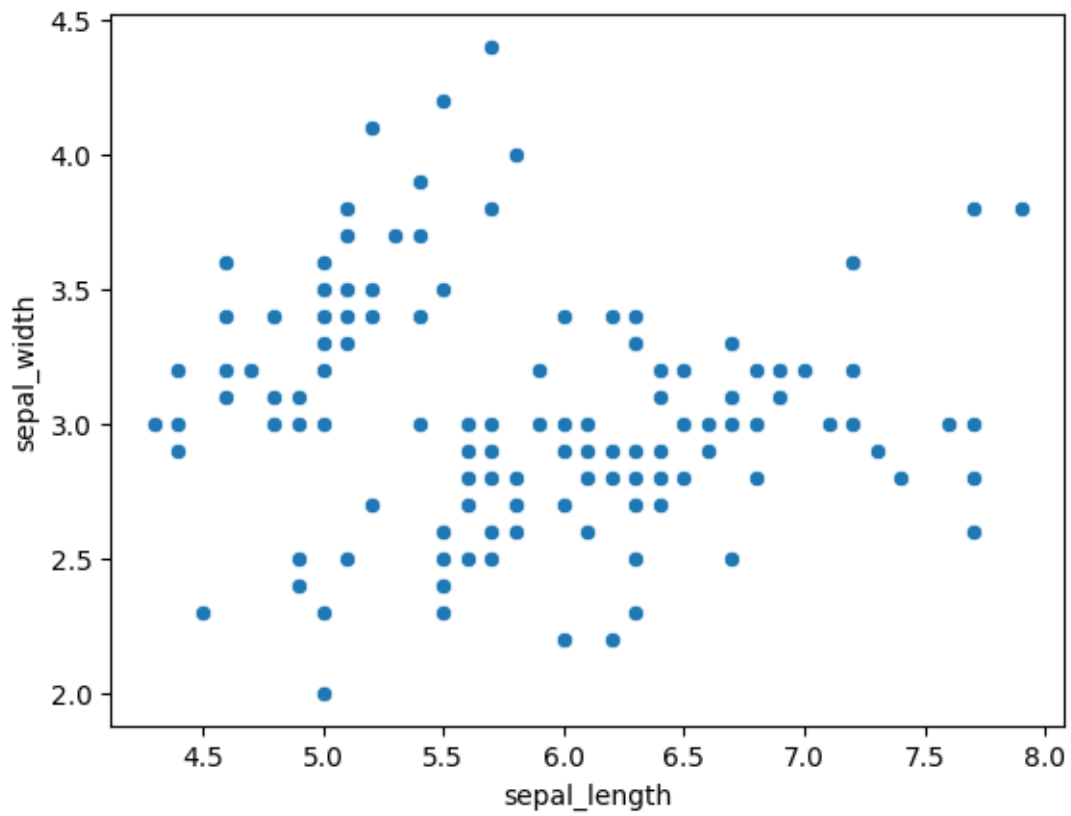
```
[ ]: df.shape
```

```
[ ]: (147, 5)
```

### Data Visualization

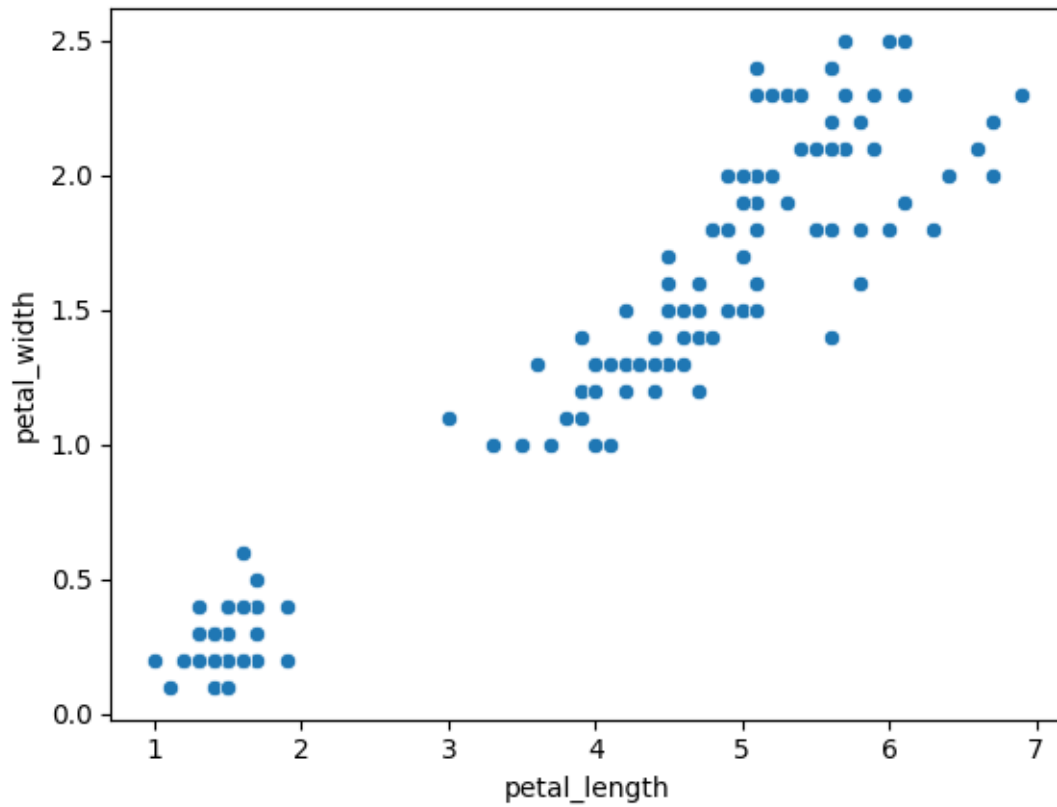
```
[ ]: sns.scatterplot(x='sepal_length',y='sepal_width',data=df)
```

```
[ ]: <Axes: xlabel='sepal_length', ylabel='sepal_width'>
```



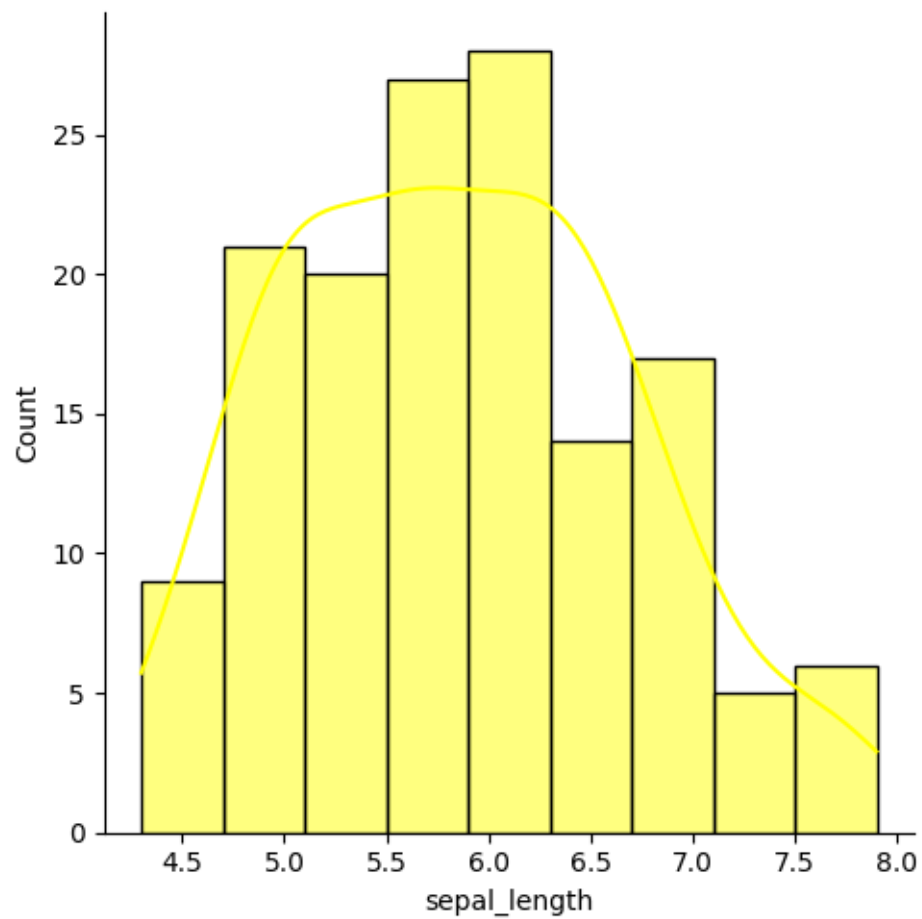
```
[ ]: sns.scatterplot(x='petal_length',y='petal_width',data=df)
```

```
[ ]: <Axes: xlabel='petal_length', ylabel='petal_width'>
```



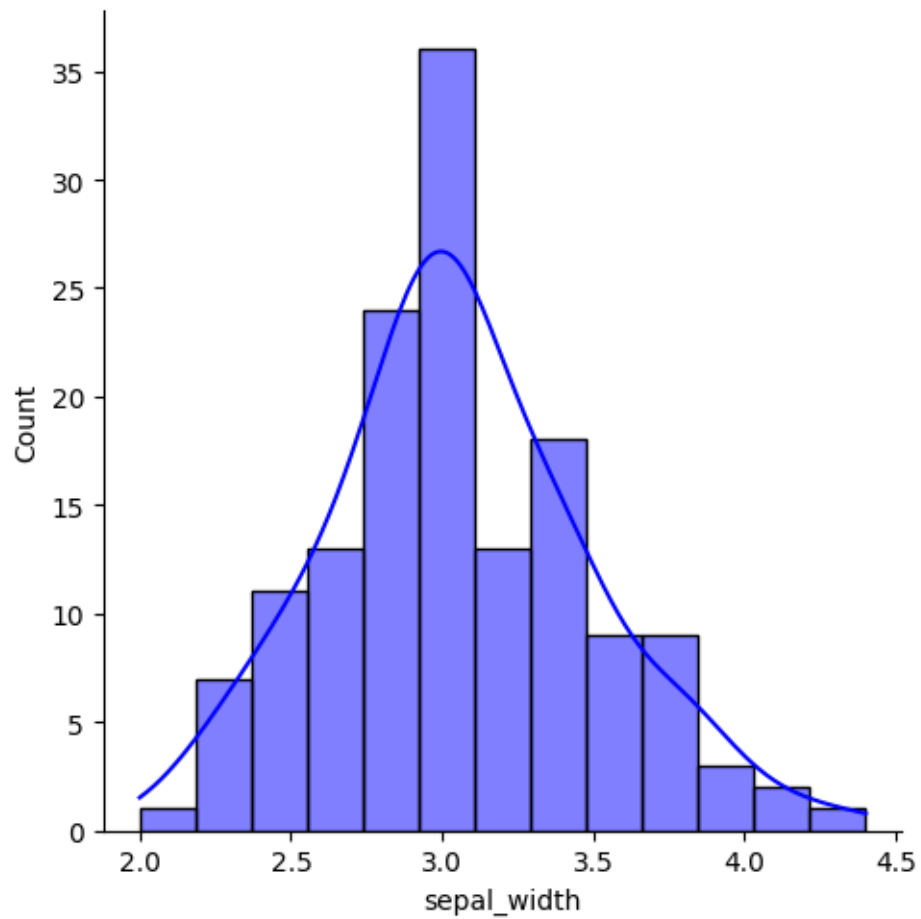
```
[ ]: sns.displot(df['sepal_length'],kde=True,color='yellow')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7d9b57371d80>
```



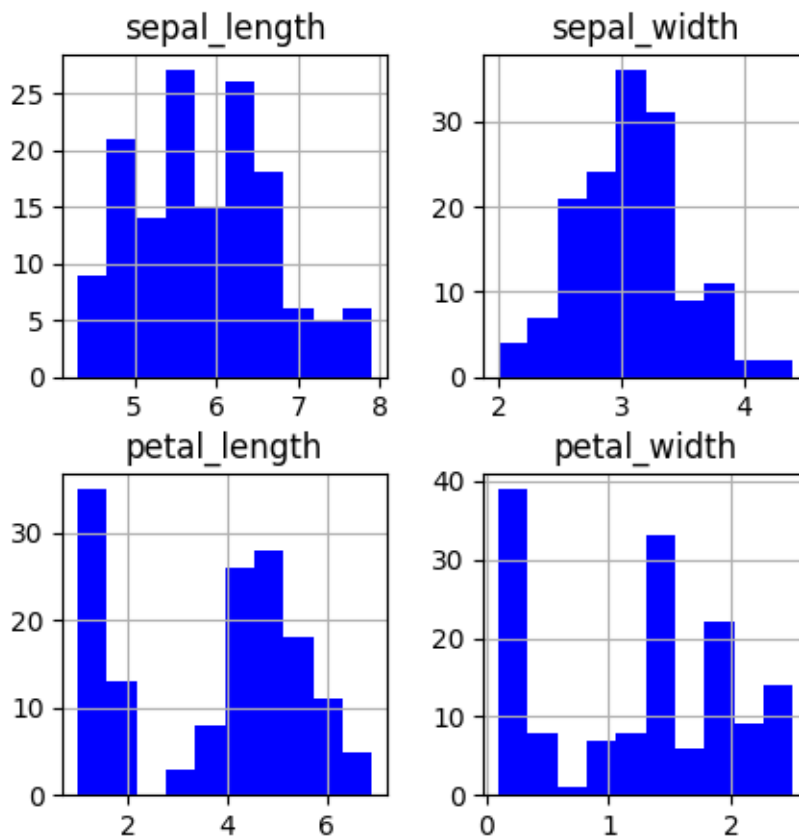
```
[ ]: sns.displot(df['sepal_width'],kde=True,color='blue')
```

```
[ ]: <seaborn.axisgrid.FacetGrid at 0x7d9b5503beb0>
```



```
[ ]: df.hist(figsize=(5,5),color='blue')
```

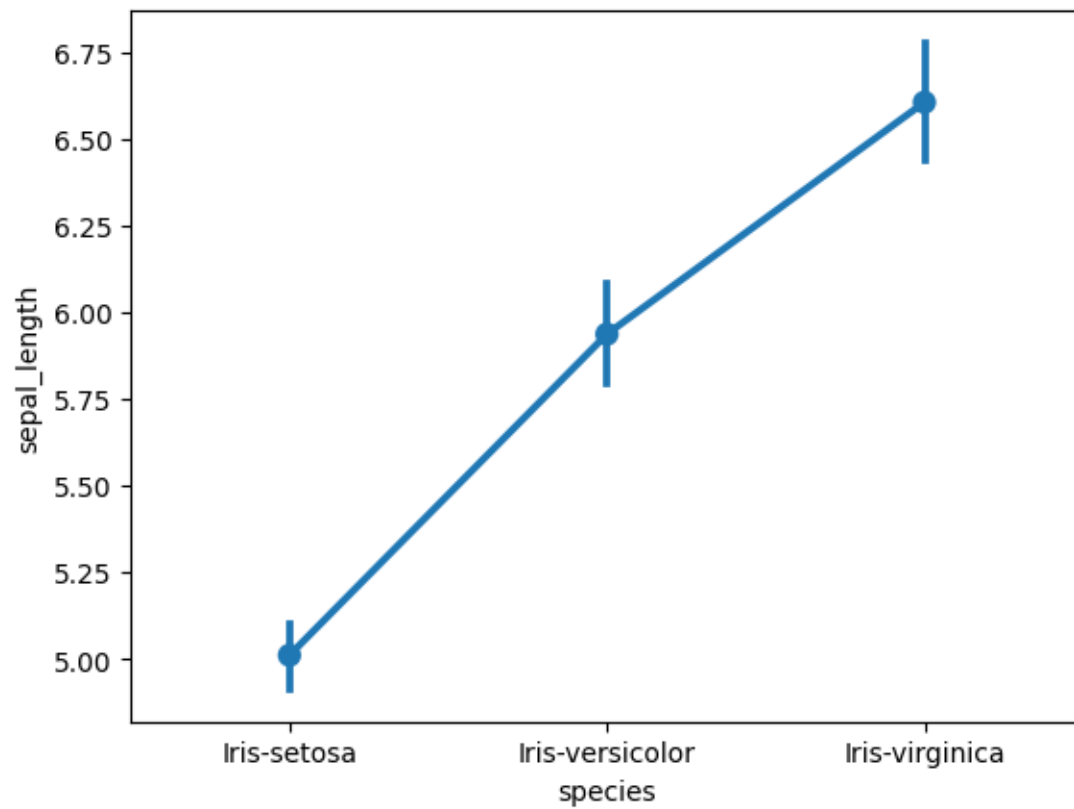
```
[ ]: array([[<Axes: title={'center': 'sepal_length'}>,  
          <Axes: title={'center': 'sepal_width'}>],  
          [<Axes: title={'center': 'petal_length'}>,  
          <Axes: title={'center': 'petal_width'}>]], dtype=object)
```



```
[ ]: sns.pointplot(x='species',y='sepal_length',data=df)
```

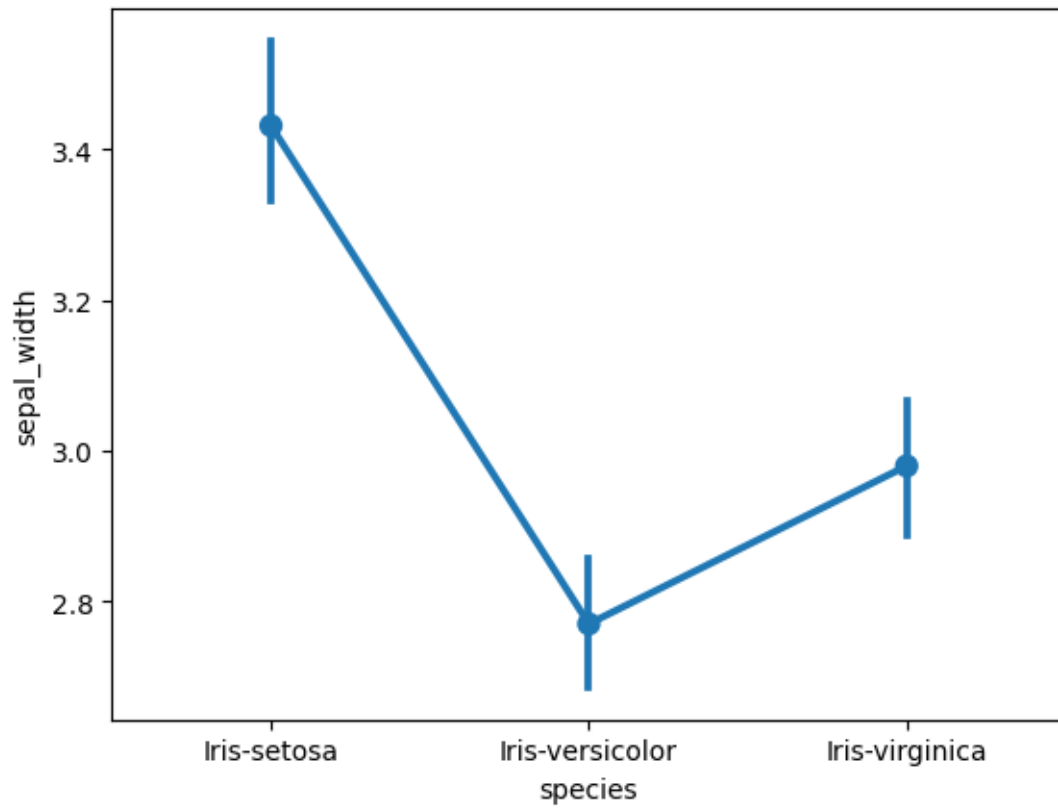
```
[ ]: <Axes: xlabel='species', ylabel='sepal_length'>
```





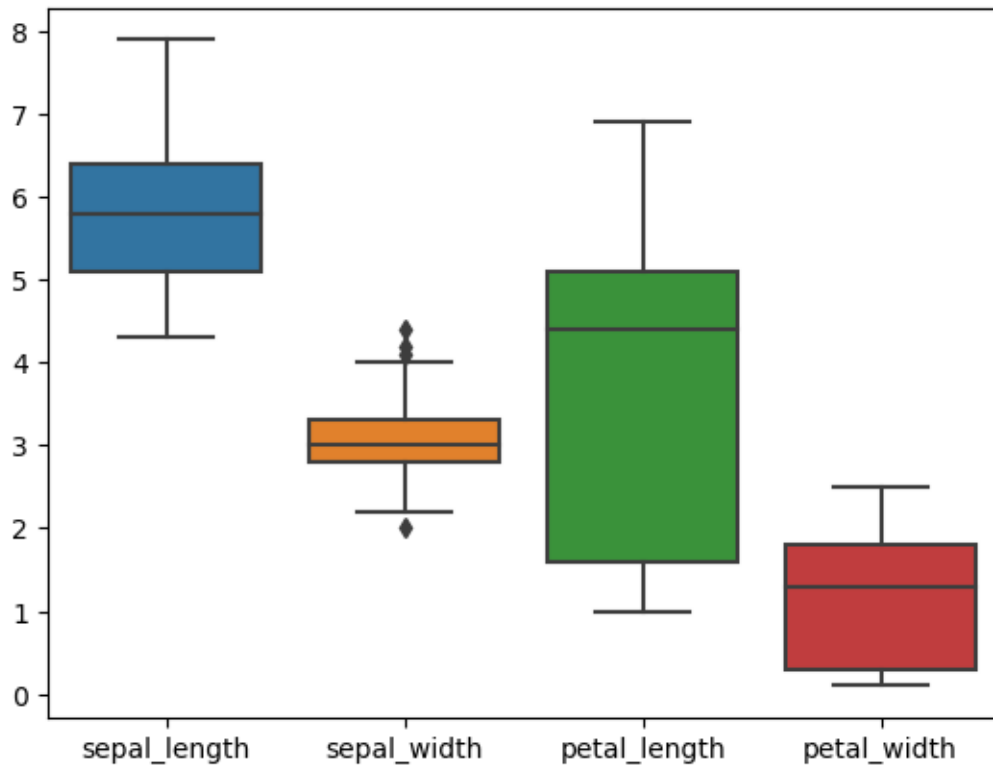
```
[ ]: sns.pointplot(x='species',y='sepal_width',data=df)
```

```
[ ]: <Axes: xlabel='species', ylabel='sepal_width'>
```



```
[ ]: sns.boxplot(df)
```

```
[ ]: <Axes: >
```

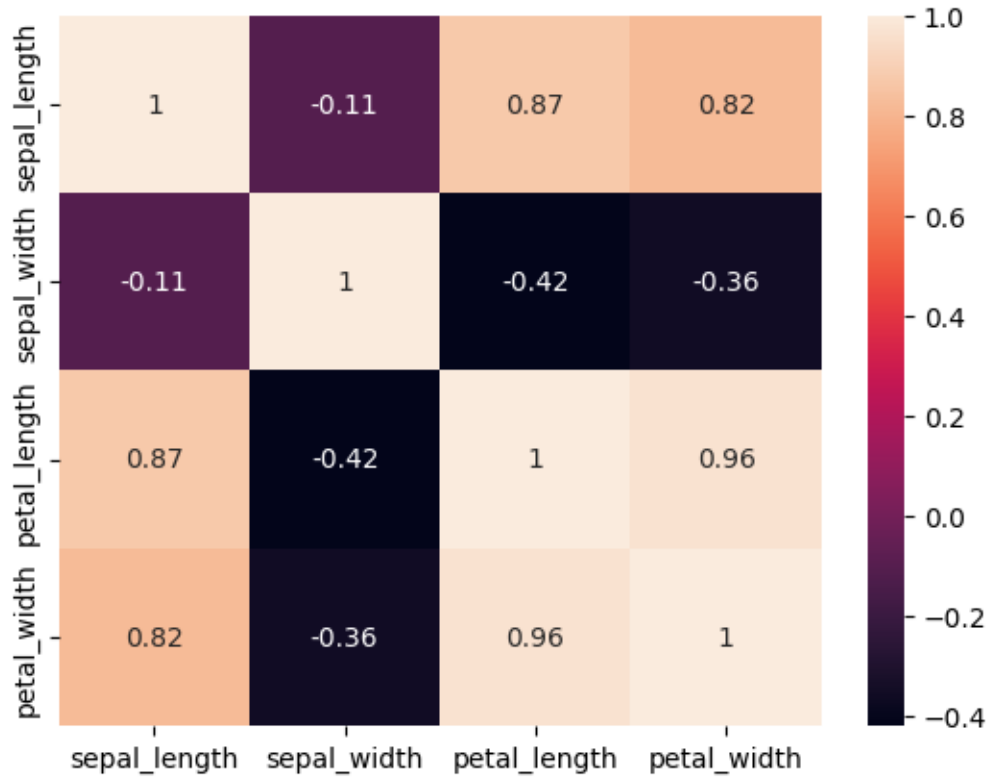


```
[ ]: sns.heatmap(df.corr(),annot=True)
```

<ipython-input-20-8df7bcac526d>:1: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
sns.heatmap(df.corr(),annot=True)
```

```
[ ]: <Axes: >
```



### Input and Output Separation

```
[ ]: x=df.iloc[:, :-1].values
      x
```

```
[ ]: array([[5.1, 3.5, 1.4, 0.2],
            [4.9, 3. , 1.4, 0.2],
            [4.7, 3.2, 1.3, 0.2],
            [4.6, 3.1, 1.5, 0.2],
            [5. , 3.6, 1.4, 0.2],
            [5.4, 3.9, 1.7, 0.4],
            [4.6, 3.4, 1.4, 0.3],
            [5. , 3.4, 1.5, 0.2],
            [4.4, 2.9, 1.4, 0.2],
            [4.9, 3.1, 1.5, 0.1],
            [5.4, 3.7, 1.5, 0.2],
            [4.8, 3.4, 1.6, 0.2],
            [4.8, 3. , 1.4, 0.1],
            [4.3, 3. , 1.1, 0.1],
            [5.8, 4. , 1.2, 0.2],
            [5.7, 4.4, 1.5, 0.4],
            [5.4, 3.9, 1.3, 0.4],
```

[5.1, 3.5, 1.4, 0.3],  
 [5.7, 3.8, 1.7, 0.3],  
 [5.1, 3.8, 1.5, 0.3],  
 [5.4, 3.4, 1.7, 0.2],  
 [5.1, 3.7, 1.5, 0.4],  
 [4.6, 3.6, 1. , 0.2],  
 [5.1, 3.3, 1.7, 0.5],  
 [4.8, 3.4, 1.9, 0.2],  
 [5. , 3. , 1.6, 0.2],  
 [5. , 3.4, 1.6, 0.4],  
 [5.2, 3.5, 1.5, 0.2],  
 [5.2, 3.4, 1.4, 0.2],  
 [4.7, 3.2, 1.6, 0.2],  
 [4.8, 3.1, 1.6, 0.2],  
 [5.4, 3.4, 1.5, 0.4],  
 [5.2, 4.1, 1.5, 0.1],  
 [5.5, 4.2, 1.4, 0.2],  
 [5. , 3.2, 1.2, 0.2],  
 [5.5, 3.5, 1.3, 0.2],  
 [4.4, 3. , 1.3, 0.2],  
 [5.1, 3.4, 1.5, 0.2],  
 [5. , 3.5, 1.3, 0.3],  
 [4.5, 2.3, 1.3, 0.3],  
 [4.4, 3.2, 1.3, 0.2],  
 [5. , 3.5, 1.6, 0.6],  
 [5.1, 3.8, 1.9, 0.4],  
 [4.8, 3. , 1.4, 0.3],  
 [5.1, 3.8, 1.6, 0.2],  
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 [5.3, 3.7, 1.5, 0.2],  
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 [7. , 3.2, 4.7, 1.4],  
 [6.4, 3.2, 4.5, 1.5],  
 [6.9, 3.1, 4.9, 1.5],  
 [5.5, 2.3, 4. , 1.3],  
 [6.5, 2.8, 4.6, 1.5],  
 [5.7, 2.8, 4.5, 1.3],  
 [6.3, 3.3, 4.7, 1.6],  
 [4.9, 2.4, 3.3, 1. ],  
 [6.6, 2.9, 4.6, 1.3],  
 [5.2, 2.7, 3.9, 1.4],  
 [5. , 2. , 3.5, 1. ],  
 [5.9, 3. , 4.2, 1.5],  
 [6. , 2.2, 4. , 1. ],  
 [6.1, 2.9, 4.7, 1.4],  
 [5.6, 2.9, 3.6, 1.3],  
 [6.7, 3.1, 4.4, 1.4],

[5.6, 3. , 4.5, 1.5],  
[5.8, 2.7, 4.1, 1. ],  
[6.2, 2.2, 4.5, 1.5],  
[5.6, 2.5, 3.9, 1.1],  
[5.9, 3.2, 4.8, 1.8],  
[6.1, 2.8, 4. , 1.3],  
[6.3, 2.5, 4.9, 1.5],  
[6.1, 2.8, 4.7, 1.2],  
[6.4, 2.9, 4.3, 1.3],  
[6.6, 3. , 4.4, 1.4],  
[6.8, 2.8, 4.8, 1.4],  
[6.7, 3. , 5. , 1.7],  
[6. , 2.9, 4.5, 1.5],  
[5.7, 2.6, 3.5, 1. ],  
[5.5, 2.4, 3.8, 1.1],  
[5.5, 2.4, 3.7, 1. ],  
[5.8, 2.7, 3.9, 1.2],  
[6. , 2.7, 5.1, 1.6],  
[5.4, 3. , 4.5, 1.5],  
[6. , 3.4, 4.5, 1.6],  
[6.7, 3.1, 4.7, 1.5],  
[6.3, 2.3, 4.4, 1.3],  
[5.6, 3. , 4.1, 1.3],  
[5.5, 2.5, 4. , 1.3],  
[5.5, 2.6, 4.4, 1.2],  
[6.1, 3. , 4.6, 1.4],  
[5.8, 2.6, 4. , 1.2],  
[5. , 2.3, 3.3, 1. ],  
[5.6, 2.7, 4.2, 1.3],  
[5.7, 3. , 4.2, 1.2],  
[5.7, 2.9, 4.2, 1.3],  
[6.2, 2.9, 4.3, 1.3],  
[5.1, 2.5, 3. , 1.1],  
[5.7, 2.8, 4.1, 1.3],  
[6.3, 3.3, 6. , 2.5],  
[5.8, 2.7, 5.1, 1.9],  
[7.1, 3. , 5.9, 2.1],  
[6.3, 2.9, 5.6, 1.8],  
[6.5, 3. , 5.8, 2.2],  
[7.6, 3. , 6.6, 2.1],  
[4.9, 2.5, 4.5, 1.7],  
[7.3, 2.9, 6.3, 1.8],  
[6.7, 2.5, 5.8, 1.8],  
[7.2, 3.6, 6.1, 2.5],  
[6.5, 3.2, 5.1, 2. ],  
[6.4, 2.7, 5.3, 1.9],  
[6.8, 3. , 5.5, 2.1],



[illegible]

## Training and testing data

### Splitting the data into training and testing data

```
[ ]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,random_state=42,test_size=0.
↪30)
```

```
[ ]: x_train
```



```

[ ]: array([[5.7, 3. , 4.2, 1.2],
           [6.8, 3.2, 5.9, 2.3],
           [6.5, 3.2, 5.1, 2. ],
           [5.1, 3.5, 1.4, 0.2],
           [6.6, 3. , 4.4, 1.4],
           [5.8, 2.7, 4.1, 1. ],
           [5.2, 3.4, 1.4, 0.2],
           [4.4, 3.2, 1.3, 0.2],
           [6. , 2.2, 4. , 1. ],
           [4.8, 3.4, 1.9, 0.2],
           [5. , 3. , 1.6, 0.2],
           [5.1, 3.3, 1.7, 0.5],
           [6. , 2.2, 5. , 1.5],
           [6.7, 3.1, 4.7, 1.5],
           [6.7, 3. , 5.2, 2.3],
           [5.1, 3.8, 1.6, 0.2],
           [5.7, 4.4, 1.5, 0.4],
           [6.3, 2.9, 5.6, 1.8],
           [4.5, 2.3, 1.3, 0.3],
           [5.9, 3.2, 4.8, 1.8],
           [6.5, 3. , 5.5, 1.8],
           [5. , 3.3, 1.4, 0.2],
           [5.7, 2.9, 4.2, 1.3],
           [6.8, 3. , 5.5, 2.1],
           [5.5, 4.2, 1.4, 0.2],
           [5.6, 3. , 4.1, 1.3],
           [6.3, 3.3, 6. , 2.5],
           [5.6, 2.9, 3.6, 1.3],
           [6.4, 2.8, 5.6, 2.1],
           [5.5, 2.4, 3.8, 1.1],
           [5.7, 2.8, 4.5, 1.3],
           [5.4, 3.9, 1.7, 0.4],
           [7.7, 2.8, 6.7, 2. ],
           [5.7, 2.8, 4.1, 1.3],
           [6.4, 3.2, 4.5, 1.5],
           [5.5, 3.5, 1.3, 0.2],
           [5.8, 2.7, 3.9, 1.2],
           [5.7, 2.6, 3.5, 1. ],
           [5. , 3.2, 1.2, 0.2],
           [5.7, 2.5, 5. , 2. ],
           [5. , 3.4, 1.5, 0.2],
           [4.8, 3. , 1.4, 0.3],
           [6.3, 2.5, 4.9, 1.5],
           [6.2, 2.9, 4.3, 1.3],
           [6. , 3.4, 4.5, 1.6],
           [6.3, 2.8, 5.1, 1.5],
           [6.1, 2.6, 5.6, 1.4],

```

[6.1, 3. , 4.6, 1.4],  
[4.4, 2.9, 1.4, 0.2],  
[4.3, 3. , 1.1, 0.1],  
[6. , 3. , 4.8, 1.8],  
[6.7, 3.3, 5.7, 2.1],  
[4.6, 3.1, 1.5, 0.2],  
[5.1, 3.5, 1.4, 0.3],  
[5. , 3.5, 1.3, 0.3],  
[6.4, 2.9, 4.3, 1.3],  
[7.4, 2.8, 6.1, 1.9],  
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[5.8, 2.8, 5.1, 2.4],  
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[6.4, 3.2, 5.3, 2.3],  
[6.1, 2.9, 4.7, 1.4],  
[6.5, 3. , 5.2, 2. ],  
[5.5, 2.4, 3.7, 1. ],  
[5.9, 3. , 4.2, 1.5],  
[5. , 2.3, 3.3, 1. ],  
[5. , 3.5, 1.6, 0.6],  
[5. , 2. , 3.5, 1. ],  
[5.8, 2.6, 4. , 1.2],  
[7. , 3.2, 4.7, 1.4],  
[5.5, 2.6, 4.4, 1.2],  
[7.2, 3.6, 6.1, 2.5],  
[5.1, 3.7, 1.5, 0.4],  
[5.2, 2.7, 3.9, 1.4],  
[6.7, 3. , 5. , 1.7],  
[5.2, 4.1, 1.5, 0.1],  
[6.7, 3.3, 5.7, 2.5],  
[5.1, 3.4, 1.5, 0.2],  
[6.9, 3.1, 5.4, 2.1],  
[4.9, 3. , 1.4, 0.2],  
[6.5, 2.8, 4.6, 1.5],  
[6.4, 2.8, 5.6, 2.2],  
[7.6, 3. , 6.6, 2.1],  
[5.8, 2.7, 5.1, 1.9],  
[7.7, 2.6, 6.9, 2.3],  
[5.5, 2.5, 4. , 1.3],  
[6.8, 2.8, 4.8, 1.4],

```

[6.3, 2.7, 4.9, 1.8],
[5.9, 3. , 5.1, 1.8],
[5.4, 3.4, 1.7, 0.2],
[6.1, 2.8, 4.7, 1.2],
[6.7, 2.5, 5.8, 1.8],
[5.8, 4. , 1.2, 0.2],
[5.6, 2.7, 4.2, 1.3],
[6.5, 3. , 5.8, 2.2]])

```

```
[ ]: x_test
```

```

[ ]: array([[6.1, 3. , 4.9, 1.8],
           [5.5, 2.3, 4. , 1.3],
           [6.7, 3.1, 5.6, 2.4],
           [5.1, 3.8, 1.5, 0.3],
           [4.9, 2.5, 4.5, 1.7],
           [4.8, 3. , 1.4, 0.1],
           [6. , 2.9, 4.5, 1.5],
           [5.4, 3.4, 1.5, 0.4],
           [6. , 2.7, 5.1, 1.6],
           [4.9, 3.1, 1.5, 0.1],
           [5. , 3.4, 1.6, 0.4],
           [5.1, 2.5, 3. , 1.1],
           [6.3, 2.5, 5. , 1.9],
           [5.6, 2.5, 3.9, 1.1],
           [6.3, 3.4, 5.6, 2.4],
           [6.2, 2.2, 4.5, 1.5],
           [5.7, 3.8, 1.7, 0.3],
           [6.1, 2.8, 4. , 1.3],
           [6.2, 2.8, 4.8, 1.8],
           [4.8, 3.1, 1.6, 0.2],
           [4.7, 3.2, 1.6, 0.2],
           [7.3, 2.9, 6.3, 1.8],
           [4.4, 3. , 1.3, 0.2],
           [6.9, 3.2, 5.7, 2.3],
           [4.9, 2.4, 3.3, 1. ],
           [4.6, 3.6, 1. , 0.2],
           [5.6, 3. , 4.5, 1.5],
           [7.7, 3. , 6.1, 2.3],
           [5.4, 3. , 4.5, 1.5],
           [4.8, 3.4, 1.6, 0.2],
           [6.4, 3.1, 5.5, 1.8],
           [4.6, 3.2, 1.4, 0.2],
           [5.6, 2.8, 4.9, 2. ],
           [5.2, 3.5, 1.5, 0.2],
           [5. , 3.6, 1.4, 0.2],
           [7.2, 3. , 5.8, 1.6],

```

```
[7.9, 3.8, 6.4, 2. ],
[6.2, 3.4, 5.4, 2.3],
[5.1, 3.8, 1.9, 0.4],
[5.4, 3.9, 1.3, 0.4],
[5.4, 3.7, 1.5, 0.2],
[7.7, 3.8, 6.7, 2.2],
[6.3, 2.3, 4.4, 1.3],
[6.9, 3.1, 5.1, 2.3],
[6.6, 2.9, 4.6, 1.3]])
```

```
[ ]: y_train
```

```
[ ]: array(['Iris-versicolor', 'Iris-virginica', 'Iris-virginica',
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'Iris-setosa', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
'Iris-setosa', 'Iris-virginica', 'Iris-versicolor',
'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica',
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'Iris-versicolor', 'Iris-virginica', 'Iris-versicolor',
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'Iris-versicolor', 'Iris-setosa', 'Iris-versicolor',
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'Iris-setosa', 'Iris-versicolor', 'Iris-versicolor',
'Iris-versicolor', 'Iris-virginica', 'Iris-virginica',
'Iris-versicolor', 'Iris-setosa', 'Iris-setosa', 'Iris-virginica',
'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',
'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
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'Iris-versicolor', 'Iris-versicolor', 'Iris-virginica',
'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
'Iris-virginica', 'Iris-versicolor', 'Iris-virginica',
'Iris-versicolor', 'Iris-virginica', 'Iris-virginica',
'Iris-setosa', 'Iris-versicolor', 'Iris-virginica', 'Iris-setosa',
'Iris-versicolor', 'Iris-virginica'], dtype=object)
```

```
[ ]: y_test
```

```
[ ]: array(['Iris-virginica', 'Iris-versicolor', 'Iris-virginica',
        'Iris-setosa', 'Iris-virginica', 'Iris-setosa', 'Iris-versicolor',
        'Iris-setosa', 'Iris-versicolor', 'Iris-setosa', 'Iris-setosa',
        'Iris-versicolor', 'Iris-virginica', 'Iris-versicolor',
        'Iris-virginica', 'Iris-versicolor', 'Iris-setosa',
        'Iris-versicolor', 'Iris-virginica', 'Iris-setosa', 'Iris-setosa',
        'Iris-virginica', 'Iris-setosa', 'Iris-virginica',
        'Iris-versicolor', 'Iris-setosa', 'Iris-versicolor',
        'Iris-virginica', 'Iris-versicolor', 'Iris-setosa',
        'Iris-virginica', 'Iris-setosa', 'Iris-virginica', 'Iris-setosa',
        'Iris-setosa', 'Iris-virginica', 'Iris-virginica',
        'Iris-virginica', 'Iris-setosa', 'Iris-setosa', 'Iris-setosa',
        'Iris-virginica', 'Iris-versicolor', 'Iris-virginica',
        'Iris-versicolor'], dtype=object)
```

### Normalization using Standard Scaler

```
[ ]: from sklearn.preprocessing import StandardScaler
scaler=StandardScaler()
scaler.fit(x_train)
x_train=scaler.transform(x_train)
x_test=scaler.transform(x_test)
```

```
[ ]: x_train
```

```
[ ]: array([[ -0.23690478, -0.08738704,  0.20877397, -0.03330265],
        [ 1.16208031,  0.38174338,  1.19788343,  1.46132027],
        [ 0.78053892,  0.38174338,  0.73242015,  1.05369584],
        [-0.99998756,  1.085439   , -1.42034749, -1.39205076],
        [ 0.90771938, -0.08738704,  0.32513979,  0.23844697],
        [-0.10972432, -0.79108266,  0.15059106, -0.30505227],
        [-0.8728071  ,  0.85087379, -1.42034749, -1.39205076],
        [-1.8902508  ,  0.38174338, -1.4785304  , -1.39205076],
        [ 0.14463661, -1.9639087  ,  0.09240815, -0.30505227],
        [-1.38152895,  0.85087379, -1.12943294, -1.39205076],
        [-1.12716803, -0.08738704, -1.30398167, -1.39205076],
        [-0.99998756,  0.61630858, -1.24579876, -0.98442633],
        [ 0.14463661, -1.9639087  ,  0.67423724,  0.37432178],
        [ 1.03489985,  0.14717817,  0.49968851,  0.37432178],
        [ 1.03489985, -0.08738704,  0.79060306,  1.46132027],
        [-0.99998756,  1.78913462, -1.30398167, -1.39205076],
        [-0.23690478,  3.19652586, -1.36216458, -1.12030114],
        [ 0.52617799, -0.32195225,  1.0233347  ,  0.78194622],
        [-1.76307034, -1.72934349, -1.4785304  , -1.25617595],
        [ 0.01745614,  0.38174338,  0.55787142,  0.78194622],
        [ 0.78053892, -0.08738704,  0.96515179,  0.78194622],
        [-1.12716803,  0.61630858, -1.42034749, -1.39205076],
```

[-0.23690478, -0.32195225, 0.20877397, 0.10257216],  
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 [-0.36408525, -0.32195225, -0.14032349, 0.10257216],  
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```
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[ 0.78053892, -0.55651745,  0.44150561,  0.37432178],
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[ 0.52617799, -0.79108266,  0.61605433,  0.78194622],
[ 0.01745614, -0.08738704,  0.73242015,  0.78194622],
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[ 0.27181707, -0.55651745,  0.49968851, -0.03330265],
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[-0.36408525, -0.79108266,  0.20877397,  0.10257216],
[ 0.78053892, -0.08738704,  1.13970052,  1.32544546]]])
```

```
[ ]: x_test
```

```
[ ]: array([[ 0.27181707, -0.08738704,  0.61605433,  0.78194622],
             [-0.49126571, -1.72934349,  0.09240815,  0.10257216],
             [ 1.03489985,  0.14717817,  1.0233347  ,  1.59719509],
             [-0.99998756,  1.78913462, -1.36216458, -1.25617595],
             [-1.25434849, -1.26021307,  0.3833227  ,  0.64607141],
             [-1.38152895, -0.08738704, -1.42034749, -1.52792557],
             [ 0.14463661, -0.32195225,  0.3833227  ,  0.37432178],
             [-0.61844617,  0.85087379, -1.36216458, -1.12030114],
             [ 0.14463661, -0.79108266,  0.73242015,  0.5101966 ],
             [-1.25434849,  0.14717817, -1.36216458, -1.52792557],
             [-1.12716803,  0.85087379, -1.30398167, -1.12030114],
```

```

[-0.99998756, -1.26021307, -0.48942094, -0.16917746],
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[ 0.39899753, -0.55651745,  0.55787142,  0.78194622],
[-1.38152895,  0.14717817, -1.30398167, -1.39205076],
[-1.50870941,  0.38174338, -1.30398167, -1.39205076],
[ 1.79798262, -0.32195225,  1.43061506,  0.78194622],
[-1.8902508 , -0.08738704, -1.4785304 , -1.39205076],
[ 1.28926077,  0.38174338,  1.08151761,  1.46132027],
[-1.25434849, -1.49477828, -0.31487221, -0.30505227],
[-1.63588988,  1.32000421, -1.65307913, -1.39205076],
[-0.36408525, -0.08738704,  0.3833227 ,  0.37432178],
[ 2.30670448, -0.08738704,  1.31424924,  1.46132027],
[-0.61844617, -0.08738704,  0.3833227 ,  0.37432178],
[-1.38152895,  0.85087379, -1.30398167, -1.39205076],
[ 0.65335846,  0.14717817,  0.96515179,  0.78194622],
[-1.63588988,  0.38174338, -1.42034749, -1.39205076],
[-0.36408525, -0.55651745,  0.61605433,  1.05369584],
[-0.8728071 ,  1.085439 , -1.36216458, -1.39205076],
[-1.12716803,  1.32000421, -1.42034749, -1.39205076],
[ 1.67080216, -0.08738704,  1.13970052,  0.5101966 ],
[ 2.5610654 ,  1.78913462,  1.48879797,  1.05369584],
[ 0.39899753,  0.85087379,  0.90696888,  1.46132027],
[-0.99998756,  1.78913462, -1.12943294, -1.12030114],
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[ 0.52617799, -1.72934349,  0.32513979,  0.10257216],
[ 1.28926077,  0.14717817,  0.73242015,  1.46132027],
[ 0.90771938, -0.32195225,  0.44150561,  0.10257216]])

```

## Model Creation

```

[ ]: from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
knn=KNeighborsClassifier(n_neighbors=7)
svmm=SVC()
dec=DecisionTreeClassifier(criterion='entropy')
model=RandomForestClassifier(n_estimators=10,criterion='entropy')
lst_model=[knn,svmm,dec,model]

```



```
[ ]: from sklearn.metrics import
      ↪confusion_matrix,accuracy_score,classification_report,ConfusionMatrixDisplay
for i in lst_model:
    print(i)
    i.fit(x_train,y_train)
    y_pred=i.predict(x_test)
    y_pred
    print(accuracy_score(y_test,y_pred))
    print("*****")
    print(classification_report(y_test,y_pred))
    print("*****")
    print(confusion_matrix(y_test,y_pred))
    print("*****")
    result=confusion_matrix(y_test,y_pred)
    labels=[0,1,2]
    cmd=ConfusionMatrixDisplay(result,display_labels=labels)
    cmd.plot()
```

KNeighborsClassifier(n\_neighbors=7)

0.9555555555555556

\*\*\*\*\*

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	17
Iris-versicolor	0.92	0.92	0.92	12
Iris-virginica	0.94	0.94	0.94	16
accuracy			0.96	45
macro avg	0.95	0.95	0.95	45
weighted avg	0.96	0.96	0.96	45

\*\*\*\*\*

```
[[17  0  0]
 [ 0 11  1]
 [ 0  1 15]]
```

\*\*\*\*\*

SVC()

0.9555555555555556

\*\*\*\*\*

	precision	recall	f1-score	support
Iris-setosa	1.00	1.00	1.00	17
Iris-versicolor	0.92	0.92	0.92	12
Iris-virginica	0.94	0.94	0.94	16
accuracy			0.96	45
macro avg	0.95	0.95	0.95	45

```

weighted avg      0.96      0.96      0.96      45

*****
[[17  0  0]
 [ 0 11  1]
 [ 0  1 15]]
*****
DecisionTreeClassifier(criterion='entropy')
0.9777777777777777
*****
              precision    recall  f1-score   support

 Iris-setosa          1.00      1.00      1.00        17
 Iris-versicolor      1.00      0.92      0.96        12
 Iris-virginica        0.94      1.00      0.97        16

 accuracy                   0.98        45
 macro avg              0.98      0.97      0.98        45
 weighted avg           0.98      0.98      0.98        45

*****
[[17  0  0]
 [ 0 11  1]
 [ 0  0 16]]
*****
RandomForestClassifier(criterion='entropy', n_estimators=10)
0.9555555555555556
*****
              precision    recall  f1-score   support

 Iris-setosa          1.00      1.00      1.00        17
 Iris-versicolor      0.92      0.92      0.92        12
 Iris-virginica        0.94      0.94      0.94        16

 accuracy                   0.96        45
 macro avg              0.95      0.95      0.95        45
 weighted avg           0.96      0.96      0.96        45

*****
[[17  0  0]
 [ 0 11  1]
 [ 0  1 15]]
*****

```

