```
import pandas as pd
```

from sklearn.model_selection import train_test_split
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score

data=pd.read_csv("/content/Iris.csv")

data

| ₹ | | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species | |
|---|-----|-----|---------------|--------------|---------------|--------------|----------------|-----|
| | 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa | ıl. |
| | 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa | +/ |
| | 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa | |
| | 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa | |
| | 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa | |
| | | | | | | | | |
| | 145 | 146 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica | |
| | 146 | 147 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica | |
| | 147 | 148 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica | |
| | 148 | 149 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica | |
| | 149 | 150 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica | |
| | | | | | | | | |

150 rows × 6 columns

Next steps: Generate code with data View recommended plots

data.head()

| | | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species | |
|-------------|---|----|---------------|--------------|---------------|--------------|-------------|-----|
| | 0 | 1 | 5.1 | 3.5 | 1.4 | 0.2 | Iris-setosa | ıl. |
| | 1 | 2 | 4.9 | 3.0 | 1.4 | 0.2 | Iris-setosa | |
| | 2 | 3 | 4.7 | 3.2 | 1.3 | 0.2 | Iris-setosa | |
| | 3 | 4 | 4.6 | 3.1 | 1.5 | 0.2 | Iris-setosa | |
| | 4 | 5 | 5.0 | 3.6 | 1.4 | 0.2 | Iris-setosa | |

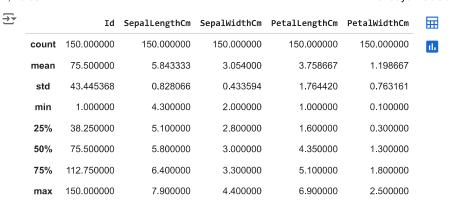
Next steps: Generate code with data View recommended plots

data.info()

<<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):

| | Ducu | COLUMNIS (COCCI | o coramiis). | | | | | |
|---------------------|-------|-----------------------|------------------|---------|--|--|--|--|
| | # | Column | Non-Null Count | Dtype | | | | |
| | | | | | | | | |
| | 0 | Id | 150 non-null | int64 | | | | |
| | 1 | SepalLengthCm | 150 non-null | float64 | | | | |
| | 2 | SepalWidthCm | 150 non-null | float64 | | | | |
| | 3 | PetalLengthCm | 150 non-null | float64 | | | | |
| | 4 | PetalWidthCm | 150 non-null | float64 | | | | |
| | 5 | Species | 150 non-null | object | | | | |
| dtypes: float64(4), | | | int64(1), object | t(1) | | | | |
| | memoi | memory usage: 7.2+ KB | | | | | | |
| | | | | | | | | |

data.describe()



data.shape

→ (150, 6)

data.size

→ 900

data.columns

data.tail()

| | | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species | |
|-------------|-----|-----|---------------|--------------|---------------|--------------|----------------|-----|
| | 145 | 146 | 6.7 | 3.0 | 5.2 | 2.3 | Iris-virginica | 11. |
| | 146 | 147 | 6.3 | 2.5 | 5.0 | 1.9 | Iris-virginica | |
| | 147 | 148 | 6.5 | 3.0 | 5.2 | 2.0 | Iris-virginica | |
| | 148 | 149 | 6.2 | 3.4 | 5.4 | 2.3 | Iris-virginica | |
| | 149 | 150 | 5.9 | 3.0 | 5.1 | 1.8 | Iris-virginica | |

data.isnull()

| → | | Id | SepalLengthCm | SepalWidthCm | PetalLengthCm | PetalWidthCm | Species | |
|----------|--------|---------|---------------|--------------|---------------|--------------|---------|-----|
| | 0 | False | False | False | False | False | False | 11. |
| | 1 | False | False | False | False | False | False | |
| | 2 | False | False | False | False | False | False | |
| | 3 | False | False | False | False | False | False | |
| | 4 | False | False | False | False | False | False | |
| | | | | | | | | |
| | 145 | False | False | False | False | False | False | |
| | 146 | False | False | False | False | False | False | |
| | 147 | False | False | False | False | False | False | |
| | 148 | False | False | False | False | False | False | |
| | 149 | False | False | False | False | False | False | |
| | 150 rd | ows × 6 | columns | | | | | |

X = data.drop(columns=['Id','Species'])

y = data['Species']

Χ

```
\overline{\Rightarrow}
           SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm
                                                                       \overline{\Pi}
       0
                     5.1
                                   3.5
                                                   1.4
                                                                 0.2
                                                                       ıl.
       1
                     4.9
                                   3.0
                                                   1.4
                                                                 0.2
       2
                     4.7
                                    3.2
                                                   1.3
                                                                 0.2
       3
                     4.6
                                   3.1
                                                   1.5
                                                                 0.2
       4
                     5.0
                                    3.6
                                                   1.4
                                                                 0.2
                     6.7
                                   3.0
                                                   5.2
                                                                 2.3
      145
                     6.3
                                    2.5
                                                   5.0
                                                                 1.9
      146
                     6.5
                                    3.0
                                                   5.2
                                                                 2.0
      147
                                                                 2.3
      148
                     6.2
                                    3 4
                                                   5.4
      149
                     5.9
                                    3.0
                                                   5.1
                                                                 1.8
     150 rows × 4 columns
 Next steps:
              Generate code with X
                                     View recommended plots
у
\overline{\mathbf{T}}
    0
               Iris-setosa
               Iris-setosa
     1
               Iris-setosa
     3
               Iris-setosa
               Iris-setosa
     4
     145
            Iris-virginica
            Iris-virginica
     146
     147
            Iris-virginica
     148
            Iris-virginica
            Iris-virginica
     149
     Name: Species, Length: 150, dtype: object
X = data.drop(columns=['Id', 'Species'])
data.columns
'Species'],
           dtype='object')
 X\_train, \ X\_test, \ y\_train, \ y\_test = train\_test\_split(X,y,test\_size=0.2,random\_state=42) 
knn = KNeighborsClassifier(n_neighbors=3)
knn.fit(X_train,y_train)
₹
              KNeighborsClassifier
     KNeighborsClassifier(n_neighbors=3)
y pred = knn.predict(X test)
accuracy = accuracy_score(y_test,y_pred)
print("Accuracy",accuracy)
→ Accuracy 1.0
import pickle
with open('iris_model.pkl', 'wb') as file:
    pickle.dump(data, file)
```