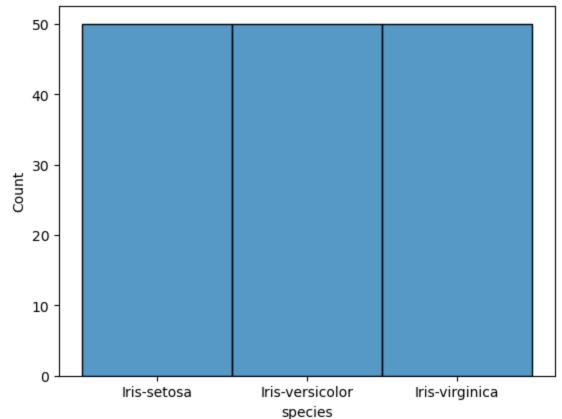
```
import numpy as np
In [1]:
          import pandas as pd
          import seaborn as sns
          from sklearn.model_selection import train_test_split
          from sklearn.preprocessing import LabelEncoder
          # loading the dataset to a pandas DataFrame
In [2]:
          df = pd.read_csv("C:\\Users\\Rames\\Project\\IRIS.csv")
          df
              sepal_length sepal_width petal_length petal_width
Out[2]:
                                                                   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
                                                            0.2
           4
                       5.0
                                   3.6
                                                1.4
                                                                 Iris-setosa
                                   3.0
                                                5.2
          145
                       6.7
                                                            2.3
                                                                Iris-virginica
          146
                       6.3
                                   2.5
                                                5.0
                                                                Iris-virginica
          147
                                   3.0
                       6.5
                                                5.2
                                                            2.0 Iris-virginica
                                                                Iris-virginica
          148
                       6.2
                                   3.4
                                                5.4
          149
                       5.9
                                   3.0
                                                5.1
                                                            1.8 Iris-virginica
         150 rows × 5 columns
         # 5 Rows of the Dataset
In [3]:
          df.tail()
               sepal_length sepal_width petal_length petal_width
Out[3]:
                                                                   species
          145
                       6.7
                                   3.0
                                                5.2
                                                            2.3 Iris-virginica
          146
                       6.3
                                   2.5
                                                5.0
                                                               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
                                                5.1
                       5.9
                                   3.0
                                                            1.8 Iris-virginica
         # dataset information
In [4]:
          df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 150 entries, 0 to 149
         Data columns (total 5 columns):
           #
               Column
                                Non-Null Count
                                                   Dtype
          - - -
                                                   float64
               sepal_length 150 non-null
           0
                                                   float64
           1
               sepal_width
                                150 non-null
           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
```

Loading [MathJax]/extensions/Safe.js

```
In [5]: # checking the number of missing values in each columns
        df.isnull().sum()
        sepal_length
                         0
Out[5]:
        sepal_width
                         0
        petal_length
                         0
        petal_width
                         0
        species
                         0
        dtype: int64
In [6]: # distribution of legit transactions and fraudulent transactions
        visualization = df['species'].value_counts()
        visualization
        Iris-setosa
                            50
Out[6]:
        Iris-versicolor
                            50
        Iris-virginica
                            50
        Name: species, dtype: int64
In [7]:
        #visualizing the data using histograms
        sns.histplot(x = "species", data = df)
        <Axes: xlabel='species', ylabel='Count'>
Out[7]:
```



```
In [8]: #convert species column binary format
en = LabelEncoder()
df["species"] =en.fit_transform(df.species)
df
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0
145	6.7	3.0	5.2	2.3	2
146	6.3	2.5	5.0	1.9	2
147	6.5	3.0	5.2	2.0	2
148	6.2	3.4	5.4	2.3	2
149	5.9	3.0	5.1	1.8	2

150 rows × 5 columns

```
In [9]: df.describe()
```

Out[9]:

Out[8]:

	sepal_length	sepal_width	petal_length	petal_width	species
count	150.000000	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667	1.000000
std	0.828066	0.433594	1.764420	0.763161	0.819232
min	4.300000	2.000000	1.000000	0.100000	0.000000
25%	5.100000	2.800000	1.600000	0.300000	0.000000
50%	5.800000	3.000000	4.350000	1.300000	1.000000
75%	6.400000	3.300000	5.100000	1.800000	2.000000
max	7.900000	4.400000	6.900000	2.500000	2.000000

```
In [10]: # select column
x = df.iloc[:,[0,1,2,3]]
y = df.iloc[:,[4]]
```

In [11]: # Split the dataset into training and testing sets
 x\_train,x\_test,y\_train,y\_test = train\_test\_split(x,y,test\_size = 0.3,random\_state = 42)

In [16]: # algorithm like K-Nearest Neighbours
from sklearn.neighbors import KNeighborsClassifier
model = KNeighborsClassifier(n\_neighbors=3)

In [17]: # training the Logistic Tegression model with Training data
model.fit(x\_train,y\_train)

I:\Install Sofware\Anaconda\Lib\site-packages\sklearn\neighbors\\_classification.py:228:
DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n\_samples,), for example using ravel().
 return self.\_fit(X, y)

Out[17]: ▼ KNeighborsClassifier

KNeighborsClassifier(n neighbors=3)

Loading [MathJax]/extensions/Safe.js

```
In [18]: y_pred = model.predict(x_test)
         y_pred
         array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
Out[18]:
                0, 2, 2, 2, 2, 2, 0, 0, 0, 0, 1, 0, 0, 2, 1, 0, 0, 0, 2, 1, 1, 0,
                0])
In [33]: # accyracy and matrix test data
         from sklearn.metrics import confusion_matrix, accuracy_score, classification_report
         result = confusion_matrix(y_test,y_pred)
         print("confusion Matrix : ")
         print(result)
         print("Accuracy Score : ",accuracy_score(y_test,y_pred))
         confusion Matrix :
         [[19 0 0]
          [ 0 13 0]
          [ 0 0 13]]
         Accuracy Score: 1.0
In [23]: # new values for sepal length, sepal width, petal length, and petal width to see the pre
         k = np.array([[5, 35, 14, 2]])
         k.shape
         (1, 4)
Out[23]:
In [29]:
         pred = model.predict(k)
         pred
         I:\Install Sofware\Anaconda\Lib\site-packages\sklearn\base.py:464: UserWarning: X does n
         ot have valid feature names, but KNeighborsClassifier was fitted with feature names
           warnings.warn(
         array([2])
Out[29]:
In [32]:
         df = pd.read_csv("C:\\Users\\Rames\\Project\\IRIS.csv")
         print("Predicted Species:", df['species'][pred[0]])
         Predicted Species: Iris-setosa
 In [ ]:
 In [ ]:
```