6/3/24, 10:03 AM Iris data set - Colab

import pandas as pd
import numpy as np

from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, classification_report,confusion_matrix
import joblib

joblib is used for saving the data train_test_split is used to train the data and split it into train and test data. randomforest classifier - model training standard scaler-feature scaling sklearn-simplicty,easier to train the model,and used to find accuracy easier.

df=pd.read_csv("Iris.csv")

₹		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species	
	0	1	5.1	3.5	1.4	0.2	Iris-setosa	11.
	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 rc	ws ×	6 columns					

df.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	

df.describe()

₹

,		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	
	count	150.000000	150.000000	150.000000	150.000000	150.000000	ılı
	mean	75.500000	5.843333	3.054000	3.758667	1.198667	
	std	43.445368	0.828066	0.433594	1.764420	0.763161	
	min	1.000000	4.300000	2.000000	1.000000	0.100000	
	25%	38.250000	5.100000	2.800000	1.600000	0.300000	
	50%	75.500000	5.800000	3.000000	4.350000	1.300000	
	75%	112.750000	6.400000	3.300000	5.100000	1.800000	
	max	150.000000	7.900000	4.400000	6.900000	2.500000	

```
df.info()
```

```
<pr
   RangeIndex: 150 entries, 0 to 149
   Data columns (total 6 columns):
            Non-Null Count Dtype
   # Column
   0 Id
               150 non-null
```

int64 1 SepalLengthCm 150 non-null
2 SepalWidthCm 150 non-null
3 PetalLengthCm 150 non-null float64 float64 3 PetalLengthCm 150 non-null float 4 PetalWidthCm 150 non-null float 5 Species 150 non-null object(1) dtypes: float64(4), int64(1), object(1) float64 float64 object

memory usage: 7.2+ KB

df.shape

→ (150, 6)

df.size

→ 900

df.tail()

→		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species	
	145	146	6.7	3.0	5.2	2.3	Iris-virginica	ıl.
	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	

print(df.isnull().sum())

<u>→</u> Id SepalLengthCm 0 SepalWidthCm 0 PetalLengthCm 0 PetalWidthCm 0 Species 0 dtype: int64

print(df.isnull())

$\overline{}$		T 4	6 11 11 6	C 1112 del-C-	D . t . 11 t l . C	D - 1 - 1112 del-C-	
₹		Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
	0	False	False	False	False	False	False
	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 rows x 6 columns]

df.isnull()

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_	Тd	Senall ength(m	SenalWidthCm	PetalLengthCm	DetalWidthCm	Snacias
	False	False	False	False	False	False
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 r	ows × 6	columns				
df.columns	5					
→ Index	'Spe	, 'SepalLength(cies'], ='object')	Cm', 'SepalWid	thCm', 'PetalLer	ngthCm', 'Peta	lWidthCm'
print(df)						
₹ .				PetalLengthCm F		\
0 1	1 2	5.1 4.9	3.5 3.0	1.4 1.4	0.2 0.2	
2	3 4	4.7 4.6	3.2 3.1	1.3 1.5	0.2 0.2	
4	5	5.0	3.6	1.4	0.2	
 145	146	6.7	3.0	5.2	2.3	
146 147	147 148	6.3 6.5	2.5 3.0	5.0 5.2	1.9 2.0	
148	149	6.2	3.4	5.4	2.3	
149	150	5.9	3.0	5.1	1.8	
0 1 2	Iri	Species s-setosa s-setosa s-setosa				
3 4 		s-setosa s-setosa 				
145 146 147 148 149	Iris-v Iris-v Iris-v	irginica irginica irginica irginica irginica				
		6 columns]				
[250	. 0.113 X	0 00145]				
df.drop(': print(df)	Id', ax	is=1)				
→	Id S	epalLengthCm 5	SepalWidthCm I 3.5	PetalLengthCm F 1.4	PetalWidthCm 0.2	\
1	2	4.9	3.0	1.4	0.2	
2	3 4	4.7 4.6	3.2 3.1	1.3 1.5	0.2 0.2	
4	5	5.0	3.6	1.4	0.2	
 145	 146	6.7	3.0	5.2	2.3	
146	147	6.3	2.5	5.0	1.9	
147 148	148 149	6.5 6.2	3.0 3.4	5.2 5.4	2.0 2.3	
149	150	5.9	3.0	5.1	1.8	
0 1 2 3 4	Iri Iri Iri	Species s-setosa s-setosa s-setosa s-setosa				
145 146 147	Iris-v	irginica irginica irginica irginica				

```
148 Iris-virginica
     149 Iris-virginica
     [150 rows x 6 columns]
df.columns

    Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',

           'Species'],
dtype='object')
X=df.drop(columns=['Id','Species'])
y=df['Species']
df.columns
→ Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm',
             'Species'],
           dtype='object')
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.2,random_state=42)
from \ sklearn.neighbors \ import \ KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=3)
knn.fit(X_train,y_train)
\overline{2}
              KNeighborsClassifier
     KNeighborsClassifier(n_neighbors=3)
y_pred = knn.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print("Accuracy:", accuracy)
→ Accuracy: 1.0
df.describe().T
⋺
                                                                                       Ī
```

*		count	mean	std	min	25%	50%	75%	max	
	ld	150.0	75.500000	43.445368	1.0	38.25	75.50	112.75	150.0	ılı
	SepalLengthCm	150.0	5.843333	0.828066	4.3	5.10	5.80	6.40	7.9	
	SepalWidthCm	150.0	3.054000	0.433594	2.0	2.80	3.00	3.30	4.4	
	PetalLengthCm	150.0	3.758667	1.764420	1.0	1.60	4.35	5.10	6.9	
	PetalWidthCm	150.0	1.198667	0.763161	0.1	0.30	1.30	1.80	2.5	

```
#Save model
import pickle
```

with open('iris_model.pkl', 'wb') as file:
 pickle.dump(df, file)