



বরেন্দ্র বিশ্ববিদ্যালয়
VARENDRA UNIVERSITY



Department of Computer Science and Engineering

29th Batch

Lab Report 7

Course title : Artificial Intelligence Lab

Course Code : CSE - 414

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Signature

➤ Question: Making Predictions Using Iris Dataset

❖ **Solution(Code & Output):**

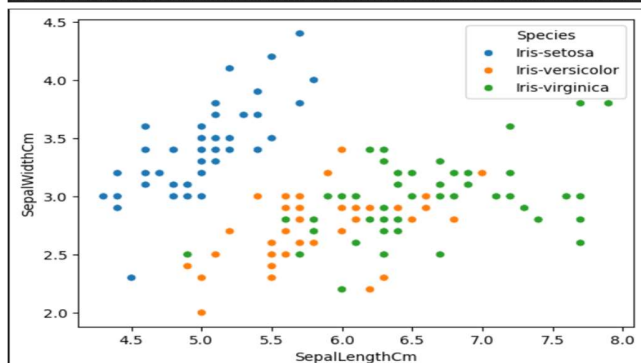
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

df = pd.read_csv('iris.csv')
df.head()

df.describe()
df.info()
df.columns

sns.scatterplot(data=df, x =
'SepalLengthCm', y =
'SepalWidthCm', hue='Species')
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
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
5	6	5.4	3.9	1.7	0.4	Iris-setosa
6	7	4.6	3.4	1.4	0.3	Iris-setosa
7	8	5.0	3.4	1.5	0.2	Iris-setosa
8	9	4.4	2.9	1.4	0.2	Iris-setosa
9	10	4.9	3.1	1.5	0.1	Iris-setosa



```
from sklearn.preprocessing import  
LabelEncoder,StandardScaler  
  
x = df.drop("Species", axis = 1)  
y = df["Species"]  
  
x  
y  
  
le = LabelEncoder()  
y_encoded = le.fit_transform(y)  
  
y_encoded
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidth
	0	1	5.1	3.5	1.4
	1	2	4.9	3.0	1.4
	2	3	4.7	3.2	1.3
	3	4	4.6	3.1	1.5
	4	5	5.0	3.6	1.4

145	146	6.7	3.0	5.2	
146	147	6.3	2.5	5.0	
147	148	6.5	3.0	5.2	
148	149	6.2	3.4	5.4	
149	150	5.9	3.0	5.1	

150 rows × 5 columns

```
0      Iris-setosa
1      Iris-setosa
2      Iris-setosa
3      Iris-setosa
4      Iris-setosa
...
145    Iris-virginica
146    Iris-virginica
147    Iris-virginica
148    Iris-virginica
149    Iris-virginica
Name: Species, Length: 150, dtype: object
```

```
array([[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
        0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
        1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2,
        2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
        2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

ID- 221311131

```
from sklearn.model_selection import
train_test_split
from sklearn.metrics import
classification_report, confusion_matrix
```

```
X_train, X_test, y_train, y_test =
train_test_split(x, y_encoded,
test_size= 0.2, random_state=42)
```

```
X_train
X_train.shape
X_test.shape
X_test
```

```
scaler = StandardScaler()
X_train_scaled =
scaler.fit_transform(X_train)
X_train_scaled
```

```
X_test_scaled =
scaler.transform(X_test)
X_test_scaled
```

	Id	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
22	23	4.6	3.6	1.0	0.2
15	16	5.7	4.4	1.5	0.4
65	66	6.7	3.1	4.4	1.4
11	12	4.8	3.4	1.6	0.2
42	43	4.4	3.2	1.3	0.2
...
71	72	6.1	2.8	4.0	1.3
106	107	4.9	2.5	4.5	1.7
14	15	5.8	4.0	1.2	0.2
92	93	5.8	2.6	4.0	1.2
102	103	7.1	3.0	5.9	2.1

120 rows x 5 columns

(120, 5)

(30, 5)

```
array([[ -1.21030717, -1.47393679,  1.22037928, -1.5639872 , -1.30948358],
       [ -1.37240188, -0.13307079,  3.02001693, -1.27728011, -1.04292204],
       [ -0.21458252,  1.08589829,  0.09560575,  0.38562104,  0.28988568],
       [ -1.46502743, -1.23014297,  0.77046987, -1.21993869, -1.30948358],
       [ -0.74717943, -1.7177306 ,  0.32056046, -1.39196294, -1.30948358],
       [  1.66108484,  0.59831066, -1.25412249,  0.72966956,  0.95628954],
       [ -0.53877194,  0.72020757,  0.32056046,  0.44296246,  0.42316645],
       [ -1.09452523, -0.74255534,  0.99542457, -1.27728011, -1.30948358],
       [ -1.62712214, -0.98634915,  1.22037928, -1.33462153, -1.30948358],
       [ -0.9787433 , -0.74255534,  2.34515281, -1.27728011, -1.44276436],
       [  1.56845929, -0.01117388, -0.80421307,  0.78701097,  0.95628954],
       [  0.24854522,  0.23261993,  0.77046987,  0.44296246,  0.55644722],
       [  0.27170161,  1.08589829,  0.09560575,  0.5576453 ,  0.42316645],
       [ -1.34924549, -0.49876152,  1.8952434 , -1.39196294, -1.04292204],
       [ -1.48818382, -0.49876152,  1.44533399, -1.27728011, -1.30948358],
       [  0.15591967, -0.37686461, -1.47907719, -0.01576889, -0.24323741],
       [  1.36005181,  0.59831066, -0.57925837,  0.78701097,  0.42316645],
```

```
from sklearn.neighbors import
KNeighborsClassifier
```

```
model=KNeighborsClassifier(n_neighbors=
3)
model.fit(X_train_scaled, y_train)
```

```
pred = model.predict(X_test_scaled)
pred
```

```
y_test
cmatri = confusion_matrix(p, y_test)
cmatri
```

KNeighborsClassifier ⓘ ?

KNeighborsClassifier(n_neighbors=3)

```
array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
       0, 2, 2, 2, 2, 2, 0, 0])
```

```
array([1, 0, 2, 1, 1, 0, 1, 2, 1, 1, 2, 0, 0, 0, 1, 2, 1, 1, 2, 0, 2,
       0, 2, 2, 2, 2, 2, 0, 0])
```

```
array([[10,  0,  0],
       [ 0,  9,  0],
       [ 0,  0, 11]])
```

❖ Conclusion:

In this lab, I learned how to use Python to train a model and make predictions. I used pandas for data, seaborn and matplotlib for graphs, and scikit-learn for machine learning. It was helped me understand how prediction works with real data. I feel more confident now about building simple ML projects.