Ex. No. : 2	IMPLEMENTATION OF SIMPLE LAYER PERCEPTRON
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Aim:

To build, train, and evaluate a single-layer perceptron model to predict the species of iris flowers based on their features and measure the model's performance.

Algorithm:

// Import Libraries //

- 1. *numpy* and *matplotlib* for numerical operations and plotting.
- 2. *sklearn* libraries for loading and splitting data, standardizing features, creating the perceptron model, and measuring performance.

// Load the Dataset //

3. Load the Iris dataset using *load_iris()* from sklearn.datasets.

// Separate Features and Labels //

4. Extract features (X) and labels (y) from the dataset.

// Standardize the Features //

5. Use *StandardScaler()* to normalize the feature values.

// Split the Data //

6. Split the standardized data by *train_test_split()* as 70% training and 30% for testing.

// Initialize the Perceptron Model //

7. Create a single-layer perceptron model using *Perceptron()* with parameters.

// Train the Model //

8. Train the model by calling *fit()* on the model with *X_train* and *y_train*.

// Make Predictions //

9. Predict the labels for the test data using predict() on the model with X_test .

// Evaluate the Model //

- 10. Calculate the accuracy of the model using *accuracy_score()* by comparing *y_test* and *y_pred*.
- 11. Generate a confusion matrix with *confusion_matrix()* and a detailed classification report using *classification_report()* for metrics like precision, recall, and F1-score.

// Display Results //

- 12. Print the accuracy, confusion matrix, and classification report for a detailed evaluation of the model.
- 13. Use *ConfusionMatrixDisplay* to plot the confusion matrix, labeling each class appropriately.

Page No.: 2

Program:

IMPLEMENTATION OF SIMPLE LAYER PERCEPTRON

Importing Libraries

```
import numpy as np
from sklearn.datasets import load_iris
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import train_test_split
from sklearn.linear_model import Perceptron
from sklearn.metrics import accuracy_score, confusion_matrix,
classification_report, ConfusionMatrixDisplay
import matplotlib.pyplot as plt
```

Loading Data Set

```
iris = load_iris()
X = iris.data[:, [2, 3]]
y = iris.target
```

Standardize the data

```
scaler = StandardScaler()
X1 = scaler.fit transform(X)
```

Data Split

```
x_train, x_test, y_train, y_test = train_test_split(X1, y,
test_size = 0.1, random_state = 42, shuffle = True)
```

x_train.shape, x_test.shape, y_train.shape, y_test.shape

Create the Perceptron Model

```
myperceptron = Perceptron(max_iter = 2000, tol = 1e-3,
random_state = 42)
```

Train the Model to Fit

```
myperceptron.fit(x train, y train)
```

Prediction

```
y pred = myperceptron.predict(x test)
```

Calculate performance metrics

```
accuracy = accuracy_score(y_test, y_pred)
conf_matrix = confusion_matrix(y_test, y_pred)
class_report = classification_report(y_test, y_pred,
target names = iris.target names)
```

Print the Performance Metrics

```
print(f'Accuracy: {accuracy:.3f}')
print("Confusion Matrix:\n", conf_matrix)
print("\nClassification Report:\n", class report)
```

Plot the Confusion Matrix

```
fig, ax = plt.subplots(figsize=(6, 6))
ConfusionMatrixDisplay.from_estimator(myperceptron, x_test,
y_test, display_labels=iris.target_names, ax=ax)
plt.title(f'Single Layer Perceptron - Accuracy:
{accuracy:.3f}')
plt.show()
```

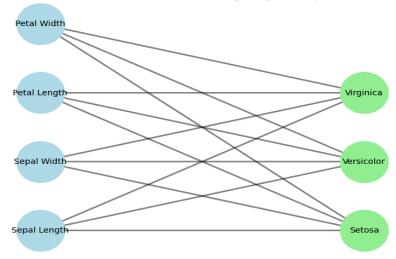
Outputs:

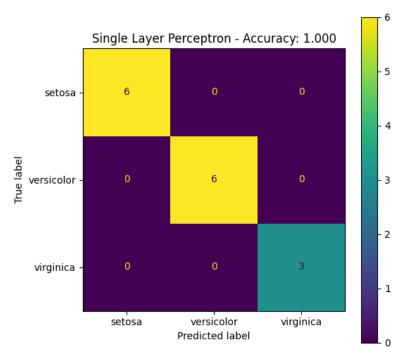
Accuracy: 0.98
Confusion Matrix:
[[6 0 0]
[0 6 0]
[0 0 3]]

Classification Report:

	precision	recall	f1-score	support
setosa	1.00	1.00	1.00	6
versicolor	1.00	1.00	1.00	6
virginica	1.00	1.00	1.00	3
accuracy			1.00	15
macro avg	1.00	1.00	1.00	15
weighted avg	1.00	1.00	1.00	15







Confusion Matrix

Result

Building, training, and evaluating a single-layer perceptron for predicting iris species has been successfully completed.