

Exp No: 4

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A Python Program to implement single Layer Perceptron

Aim:-

To implement python program for the single layer Perceptron

Algorithm :-

Step 1: Import Necessary Libraries

* Define the number of input features (input_dim).

* Initialize weights (w) and bias (b) to 0 or small random values.

Step 2: Initialize the perceptron

* Define the number of input features (input_dim)

* Initialize weights (w) and bias (b) to 280 or small random values.

Step 3: Define Activation function!

* Choose an activation function (eg. step function, sigmoid, or ReLU)

* Use defined function - sigmoid_func()

→ Compute $1/(1 + \exp(-x))$ and return the value

* Use defined function - del_Ge

→ Compute the product of value of sigmoid_func(x) and (1 - Sigmoid_func(x)) and return the value

Step 4: Define Training Data:

* Define input features (x_i) and corresponding target labels (y_i)

Step 5: Define Learning Rate and Number of Epochs:

* Choose a learning rate (α) and the number of training epochs

Step 6: Training the perceptron:

* For each epoch:

→ For each input sample in the training data:

→ Compute the weighted sum of inputs (z) as the dot product of input features and weights plus bias ($z = \mathbf{w} \cdot \mathbf{x}[i] + b$)

→ Apply the activation function to get the predicted output (y_{pred})

→ Compute the error ($error = y[i] - y_{pred}$)

→ Update the weights and bias using the learning rate and error ($w += \alpha * error * x[i]$; $b += \alpha * error$)

Step 7: Prediction

* Use the trained perceptron to predict the output for new input data

Step 8: Evaluate the model

* Measure the performance of the model using metrics such as accuracy, precision, recall etc.

Program:

```
import numpy as np
import random as rd
# input and output data
input_value = np.array([[-0.7, -0.1], [0.1, 0.5], [0.5, 0.9]])
shape (4, 2)
output = np.array([0, 0, 1]).reshape(4, 1) # shape (4, 1)
# initialize weights and bias
weights = np.array([[0.5, 0.1], [0.1, 0.3]]) # shape (2, 2)
bias = 0.2
# sigmoid activation function
def sigmoid_func(x):
    return 1/(1+np.exp(-x))
# Derivation of sigmoid
def dfunc():
    return sigmoid_func(x) * (1 - sigmoid_func(x))
# Training loop
for epoch in range(15000):
    input_val = input_value
    # Forward pass.
    weighted_sum = np.dot(input_val, weights) + bias
    first_output = sigmoid_func(weighted_sum)
    # Error
    error = first_output - output
    total_error = np.sqrt(error).mean()
```

```

# Backpropagation
first_da = error
second_da = da(weighted_sum) # Derivative
derivative = first_da * second_da
# Update weights
t_input = input_bias.T
final_derivative = np.dot(t_input, derivative)
weights = weights - (0.05 * final_derivative)
# Update bias
for i in derivative:
    bias = bias - (0.05 * i)
# final weights and bias
print("Weights :\n", weights)
print("Bias :\n", bias)

# Test prediction
def predict(x):
    result = np.dot(x, weights) + bias
    return sigmoid_func(result)

# Predict for each input
test_inputs = [np.array([1, 0]), np.array([1, 1]),
               np.array([0, 0]), np.array([0, 1])]

for pred in test_inputs:
    res = predict(pred)
    print("Input: ", pred, "Output: ", res)

```

Program Flow:

Input → Activation Function → Output

Ans:

The program will take input from user and calculate output by applying the activation function.

Algorithm:

- Step 1: Import the random module
- * Import module os
- Step 2: Create a variable x and assign it to 0
- Step 3: Read input from user
- Step 4: Put float(x) in x variable
- Step 5: Create a variable y and assign it to 0
- Step 6: Assign the x value to y and print the value of y
- Step 7: Display the result
- Step 8: Handle the exception
- Step 9: Display the message "Program ended"
- Step 10: Use the stop() method to stop the execution of the code (when the user presses Ctrl+C)
- Step 11: Import the time module
- Step 12: Import the math module

Result:

Thus the python program to implement single Layer Perceptron has been executed successfully