

A python program to implement single layer perceptron

Aim:

To implement python program for the Single layer perceptron

Algorithm:

Step 1: Import necessary libraries.

- * Import numpy for numerical operation

Step 2: Initialize the perception.

- * Define the number of input feature (input - aim).

* Initialize weight (w) and bias (b) to zero or small random values.

Step 3: Define Activation function.

- * choose an activation function (e.g. step function, Sigmoid, ReLU).

* user Defined function - sigmoid_func(x):

- * Compute $\frac{1}{1 + \exp(-x)}$ and return the value.

Step 4: Define Training Data:

- * Define input features (x) and corresponding target values (y).

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 ($z = \mathbf{w} \cdot \mathbf{x} + b$).

 >> Compute the error ($\text{error} = y_{\text{true}} - y_{\text{pred}}$).

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 pandas as pd
```

```
input_values = np.array([[-0.5, 0.1, 1.2, 0.9]])
```

```
input_values.shape(4, 2)
```

```
output = np.array([0, 0, 1, 0])
```

```
output = output.reshape(4, 1)
```

```
output.shape
```

```
weights = np.array([[0.1], [0.3]])
```

```
weights
```

```
bias = 0.2
```

```
def sigmoid_func(x):
```

```
return 1 / (1 + np.exp(-x))
```

def der(f):

return sigmoid-func(x)* (1 - sigmoid-func(x))

for epochs in range(15000):

$$\text{input_corr} = \text{input_value}$$

$$\text{weighted_sum} = \text{np}. \text{dot}(\text{input_corr}, \text{weights})$$

$$\text{bias_first_output} = \text{sigmoid_func}(\text{weighted_sum})$$

$$\text{error} = \text{first_output} - \text{output}$$

$$\text{tot_error} = \text{np}. \text{square}(\text{np}. \text{subtract}(\text{first_output}, \text{output}). \text{mean})$$

$$\text{first_der} = \text{error}$$

$$\text{second_der} = \text{der}(\text{first_output})$$

$$\text{derivative} = \text{first_der} * \text{second_der}$$

$$t_input = \text{input_value} : T$$

$$\text{final_derivative} = \text{np}. \text{dot}(t_input, \text{derivative})$$

$$\text{weights} = \text{weights} - (\text{0.05} * \text{final_derivative})$$

for e in derivative

$$\text{bias} = \text{bias} - (\text{0.05} * e)$$

print(weights)

print(bias)

$\text{pred} = \text{np.array}([1, 0])$

$\text{result} = \text{np.dot}(\text{pred}, \text{weights}) + \text{bias}$

$\text{res} = \text{sigmoid_func(result)}$

print(res)

$\text{pred} = \text{np.array}([1, 1])$

$\text{result} = \text{np.dot}(\text{pred}, \text{weights}) + \text{bias}$

$\text{res} = \text{sigmoid_func(result)}$

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$\text{pred} = \text{np.array}([0, 0])$

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$\text{result} = \text{np.dot}(\text{pred}, \text{weights}) + \text{bias}$

$\text{res} = \text{sigmoid_func(result)}$

print(res)

Results:

~~Ans 30~~ Thus, the python program for implement single layer perceptron was executed successfully.