

Experiment – 7

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Aim: Perceptron training algorithm for L and M classification

Theory:

Perceptrons are a type of artificial neuron that predates the sigmoid neuron. It appears that they were invented in 1957 by Frank Rosenblatt at the Cornell Aeronautical Laboratory.

A perceptron can have any number of inputs, and produces a binary output, which is called its activation.

First, we assign each input a weight, loosely meaning the amount of influence the input has over the output.

To determine the perceptron's activation, we take the weighted sum of each of the inputs and then determine if it is above or below a certain threshold, or *bias, *represented by b.

The formula for perceptron neurons can be expressed like this:

$$\text{output} = \begin{cases} 0 & \text{if } \sum_j w_j x_j \leq \text{threshold} \\ 1 & \text{if } \sum_j w_j x_j > \text{threshold} \end{cases}$$

Algorithm:

```
def perceptron(inputs, bias)

    weighted_sum = sum {
    for each input in inputs
    input.value * input.weight
    }

    if weighted_sum <= bias return 0
    if weighted_sum > bias
    return 1
end
```

Code:

```
def sgn(net_input):
    if net_input <= 0 :
        return -1
    return 1

def pattern_classifier(n_iterations, input, weight, desired_output, learning_rate):
    for iteration in range(n_iterations):
        print(f'Iteration {iteration+1}')
        output = []
        for i,X in enumerate(input):
            net_input = 0
            for j in range(len(X)):
                net_input+=weight[j]*X[j]
            generated_output = sgn(net_input)
            output.append(generated_output)
            if generated_output != desired_output[i]:
                difference = desired_output[i] - generated_output
                for position in range(len(weight)):
                    weight[position] = float("{:.2f}".format(weight[position] +
learning_rate*difference*X[position]))
        print(f'Generated Output vector for Iteration {iteration+1} : {output}')
        print(f'Weight vector after Iteration {iteration+1} : {weight}')
        print("-----"*25)
        if output == desired_output:
            break
    return output,weight

def main():
    input = [
        [1,0,0,0,0,1,0,0,0,0,1,0,0,0,0,1,1,1,1,1], #L starts here
        [1,0,0,0,0,1,0,0,0,0,1,0,0,0,0,0,1,1,1,1],
        [1,1,0,0,0,1,0,0,0,0,1,0,0,0,0,1,1,1,1,1],
        [0,1,0,0,0,0,1,0,0,0,0,1,0,0,0,1,1,1,1,1],
        [1,0,0,0,0,1,0,0,0,0,1,0,0,0,1,1,1,1,1,1],
        [0,1,0,0,0,0,1,0,0,0,1,0,0,0,0,1,1,1,1,1],
        [0,1,0,0,0,1,0,0,0,0,1,0,0,0,0,1,1,1,1,1],
        [1,0,0,0,0,1,0,0,0,0,1,0,1,0,0,1,1,0,1,1],
        [0,1,0,0,0,0,1,0,0,0,0,1,1,0,0,1,1,0,1,1],
        [1,0,0,0,0,1,0,0,0,0,1,0,0,0,0,1,1,0,1,1],
```

```

[0,1,0,1,0,1,1,0,1,1,0,1,0,1,1,0,0,0,1], #M starts here
[1,0,0,0,1,1,1,0,1,1,1,0,1,0,1,1,0,0,0,1],
[1,0,0,0,1,1,1,0,1,1,1,0,1,0,1,1,0,1,0,1],
[1,1,0,1,1,1,0,1,0,1,1,0,1,0,1,1,0,0,0,1],
[1,1,0,1,1,1,0,1,0,1,1,0,0,0,1,1,0,0,0,1],
[1,0,0,0,1,1,1,0,1,1,1,0,0,0,1,1,0,0,0,1],
[1,0,0,0,1,1,1,0,1,1,1,0,1,1,1,0,1,0,1,1],
[1,1,0,1,1,1,0,1,0,1,1,0,1,0,1,1,0,1,0,1],
[1,0,0,0,1,1,1,0,1,1,1,0,1,0,1,0,0,0,0,0],
[1,0,0,0,1,1,1,1,1,1,1,0,1,0,1,1,0,0,0,1],
]
desired_output = [1,1,1,1,1,1,1,1,1,-1,-1,-1,-1,-1,-1,-1,-1,-1,-1]
initial_weight = [1,1,0,1,1,0,1,1,0,1,1,0,1,1,0,1,1,0,1,1]
learning_rate = 0.05 n_iterations = 3

classification_output, weight_vector = pattern_classifier(n_iterations, input,
initial_weight, desired_output, learning_rate)

count = 0
for i, output in enumerate(classification_output):
    if output == desired_output[i]: count+=1

accuracy = (count / len(input))*100

print(f'Accuracy of Classifier : {accuracy} %')

print('Classifying an Unknown Sample of L (Output = 1)')
unknown_sample = [1,1,0,0,0,1,0,0,0,0,1,0,0,0,0,1,1,1,1,0]
print('Unknown Sample : ',unknown_sample)
net_input=0
for i in range(len(unknown_sample)):
    net_input+=weight_vector[i]*unknown_sample[i]
predicted_output = sgn(net_input)
print('Predicted Output : ', predicted_output)
print("\n")
main()

```

Output:

```

Iteration 1
Generated Output vector for Iteration 1 : [1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, -1, 1]

```

Weight vector after Iteration 1 : [0.2, 0.6, 0.0, 0.6, 0.2, -0.9, 0.4, 0.6, -0.6, 0.1, 0.1, -0.1, 0.4, 0.9, -0.9, 0.1, 1.0, -0.3, 1.0, 0.1]

Iteration 2

Generated Output vector for Iteration 2 : [1, 1, 1, 1, 1, 1, 1, 1, 1, 1, -1, -1, -1, 1, -1, -1, -1, -1, -1, -1]

Weight vector after Iteration 2 : [0.1, 0.5, 0.0, 0.5, 0.1, -1.0, 0.4, 0.5, -0.6, 0.0, 0.0, -0.1, 0.3, 0.9, -1.0, 0.0, 1.0, -0.3, 1.0, 0.0]

Iteration 3

Generated Output vector for Iteration 3 : [1, 1, 1, 1, -1, 1, 1, 1, 1, 1, -1, -1, -1, 1, -1, -1, -1, -1, -1, -1]

Weight vector after Iteration 3 : [0.1, 0.4, 0.0, 0.4, 0.0, -1.0, 0.4, 0.4, -0.6, -0.1, 0.0, -0.1, 0.2, 0.9, -1.0, 0.0, 1.1, -0.2, 1.1, 0.0]

Accuracy of Classifier : 90.0 %

Classifying an Unknown Sample of L (Output = 1)

Unknown Sample : [1, 1, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 1, 1, 1, 0]

Predicted Output : 1

Conclusion: Hence we have performed **Perceptron training algorithm for L and M classification**