

Date of Performance :

Date of Submission :

EXPERIMENT NUMBER: 1

Aim:Implement McCulloch Pitts model for AND and XOR logic gates.

Objective:

To implement basic neural network models for simulating logic gate

Software Used :Python

Theory:

McCulloch Pitt Model : The McCulloch-Pitts (M-P) model, introduced in 1943 by Warren McCulloch and Walter Pitts, is a foundational model of artificial neural networks. The model is based on the concept of binary neurons, which are either on (1) or off (0), and the model defines a set of logical rules that determine the state of a neuron based on the states of its inputs.

In the M-P model, a neuron is represented as a threshold gate, which receives inputs from other neurons and outputs a binary value depending on whether the sum of the inputs exceeds a threshold value. The threshold gate acts as a simple decision maker, allowing a representation of logical statements such as "if A and B are true, then output 1."

One important aspect of the M-P model is that it allows for the creation of a network of neurons that can work together to solve a problem. For example, multiple neurons can be connected to recognize patterns in input data, making it possible to use the M-P model for applications such as image recognition or decision making.

Despite its simplicity, the M-P model was a major contribution to the development of artificial neural networks and provided a basis for further research in the field. The M-P model inspired the development of more sophisticated neural network models with continuous activation functions and multi-layer networks, which expanded the capability of artificial neural networks from simple decision making to more complex pattern recognition and data analysis.

In conclusion, the McCulloch-Pitts model is a simple yet powerful model of artificial neurons that laid the foundation for the development of artificial intelligence and artificial neural networks. It introduced the idea of binary neurons and showed how a network of neurons can work together to solve problems, paving the way for the development of more advanced neural network models.

Algorithm:

Here is a high-level pseudocode for a McCulloch-Pitts (M-P) model:

Input: binary inputs x_1, x_2, \dots, x_n

Output: binary output y

Step 1: Initialize weights w_1, w_2, \dots, w_n and threshold value θ
Step 2: Calculate the weighted sum of inputs:

weighted_sum = $w_1 * x_1 + w_2 * x_2 + \dots + w_n * x_n$

Step 3: Compare weighted_sum to threshold θ :

if weighted_sum > θ :

y = 1

else:

y = 0

Step 4: Return output y

Note that this is just a high-level pseudocode and not a complete implementation of the M-P model. A complete implementation would typically involve training the network to optimize the weights and threshold based on a set of input-output examples, and would involve additional steps and algorithms for weight update and error calculation.

Program:

```
*****Mc-Culloch Pitts model for AND gate*****
```

```
import numpy as np
```

```
def ANDThreshold(v):
```

```
    if v > 1:
```

```
        return 1
```

```
    else:
```

```
        return 0
```

```
def MCPModel(x, w):
```

```
    v = np.dot(w, x)
```

```
    y = ANDThreshold(v)
```

```
    return y
```

```
def ANDGate(x):
```

```
    w = np.array([1, 1])
```

```
    return MCPModel(x, w)
```

```
test1 = np.array([0, 0])
```

```
test2 = np.array([0, 1])
```

```
test3 = np.array([1, 0])
```

```
test4 = np.array([1, 1])
```

```
print(f"AND({0}, {0}) = { ANDGate(test1)}")
```

```
print(f"AND({0}, {1}) = { ANDGate(test2)}")
```

```
print(f"AND({1}, {0}) = { ANDGate(test3)}")
```

```
print(f"AND({1}, {1}) = { ANDGate(test4)}")
```

```
*****Mc-Culloch Pitts model for XOR gate*****
```

```
def XORThreshold(v):
```

```
    if v == 1:
```

```
        return 1
```

```
    else:
```

```

    return 0

def MCPModel(x, w):
    v = np.dot(w, x)
    y = XORThreshold(v)
    return y

def XORGate(x):
    w = np.array([1, 1])
    return MCPModel(x, w)

test1 = np.array([0, 0])
test2 = np.array([0, 1])
test3 = np.array([1, 0])
test4 = np.array([1, 1])

print(f"XOR({0}, {0}) = { XORGate(test1)}")
print(f"XOR({0}, {1}) = { XORGate(test2)}")
print(f"XOR({1}, {0}) = { XORGate(test3)}")
print(f"XOR({1}, {1}) = { XORGate(test4)}")

```

Output:

Mc-Culloch Pitts model for AND gate

```

    return MCPModel(x, w)
test1 = np.array([0, 0])
test2 = np.array([0, 1])
test3 = np.array([1, 0])
test4 = np.array([1, 1])
print(f"XOR({0}, {0}) = { XORGate(test1)}")
print(f"XOR({0}, {1}) = { XORGate(test2)}")
print(f"XOR({1}, {0}) = { XORGate(test3)}")
print(f"XOR({1}, {1}) = { XORGate(test4)}")

C XOR(0, 0) = 0
XOR(0, 1) = 1
XOR(1, 0) = 1
XOR(1, 1) = 0

```

The screenshot shows a Jupyter Notebook cell with the following code and output. The code defines a function XORGate that returns the result of a dot product between input x and weight vector w. It then tests four cases of XOR with inputs (0,0), (0,1), (1,0), and (1,1). The output shows the expected results for each case.

Mc-Culloch Pitts model for OR gate

```

    return MCPModel(x, w)
test1 = np.array([0, 0])
test2 = np.array([0, 1])
test3 = np.array([1, 0])
test4 = np.array([1, 1])
print(f"XOR({0}, {0}) = { XORGate(test1)}")
print(f"XOR({0}, {1}) = { XORGate(test2)}")
print(f"XOR({1}, {0}) = { XORGate(test3)}")
print(f"XOR({1}, {1}) = { XORGate(test4)}")

C XOR(0, 0) = 0
XOR(0, 1) = 1
XOR(1, 0) = 1
XOR(1, 1) = 0

```

The screenshot shows a Jupyter Notebook cell with the same code as the previous one, but for an OR gate. The output shows the expected results for each case, where the result is 1 for all inputs except (0,0).

Conclusion/Outcome:

Thus we have implemented Mc-Culloch Pitts model for AND and XOR logic gates
We also understood that implement basic neural network models for simulating logic gate

Marks & Signature:

R1 (4 Marks)	R2 (4 Marks)	R3 (4 Marks)	R4 (3 Mark)	Total (15 Marks)	Signature