

EXPERIMENTNO.9

Aim: Implement McCulloch Pitts model.

Theory:

The McCulloch-Pitts (MCP) neuron model, proposed by Warren McCulloch and Walter Pitts in 1943, laid the foundation for modern artificial neural networks. It is a simplified mathematical model of a biological neuron and provides insights into how neurons might process information in the brain. The MCP neuron model consists of three main components: input signals, weights, and an activation function.

Basic Concepts

- **Neuron:** The fundamental building block of neural networks, a neuron receives input signals, processes them, and produces an output signal. In the MCP neuron model, the neuron performs a simple computation based on the inputs and weights.
- **Input Signals:** Neurons receive input signals from other neurons or external sources. Each input signal is associated with a weight, which represents the strength of the connection between the input and the neuron.
- **Weights:** Weights determine the influence of input signals on the neuron's output. Larger weights indicate stronger connections, while smaller weights indicate weaker connections. The weights can be adjusted during the learning process to improve the performance of the neuron.

- Activation Function: The activation function determines whether the neuron will produce an output signal based on the weighted sum of its inputs. In the MCP neuron model, the activation function is a threshold function that compares the weighted sum to a threshold. If the weighted sum exceeds the threshold, the neuron fires and produces an output signal; otherwise, it remains inactive.

McCulloch-Pitts Neuron Model

The McCulloch-Pitts neuron model describes how a neuron processes input signals to produce an output signal. Mathematically, the output of an MCP neuron y is computed as follows:

$$y = \begin{cases} 1 & \text{if } \sum_{i=1}^n w_i \cdot x_i \geq \text{threshold} \\ 0 & \text{otherwise} \end{cases}$$

Where:

- y is the output of the neuron.
- x_i are the input signals.
- w_i are the weights associated with the input signals.
- threshold is a threshold value.
- n is the number of input signals.

Properties

- Binary Output: The output of an MCP neuron is binary, typically represented as 0 (inactive) or 1 (active). This binary output simplifies computations and enables the modeling of logical functions.

- **Thresholding Operation:** The MCP neuron performs a thresholding operation, where the weighted sum of input signals is compared to a threshold. If the sum exceeds the threshold, the neuron fires; otherwise, it remains inactive.
- **Simplified Model:** The MCP neuron model is a simplified representation of the biological neuron. It neglects details such as temporal dynamics, synapse strengths, and non-linearities observed in real neurons.

Applications

The McCulloch-Pitts neuron model has influenced various fields, including:

- **Artificial Neural Networks:** The MCP neuron model forms the basis of artificial neural networks, which are used for pattern recognition, classification, regression, and other machine learning tasks.
- **Computational Neuroscience:** The MCP neuron model provides insights into the computational properties of biological neurons and helps researchers understand how neurons process information in the brain.
- **Cognitive Science:** The MCP neuron model informs theories of cognition and provides a framework for understanding how the brain performs cognitive functions such as perception, learning, and memory.

The McCulloch-Pitts neuron model is a foundational concept in neuroscience and artificial intelligence. It describes a simplified mathematical model of a biological neuron and serves as the building block for artificial neural networks. By understanding the principles of the MCP neuron model, researchers can develop more complex neural network architectures and learning algorithms for solving a wide range of tasks in machine learning and cognitive science.

Program Code:

```
import numpy as np
```

```
classMCPNeuron:
```

```
    def __init__(self,weights,threshold):
```

```
        self.weights = weights
```

```
        self.threshold = threshold
```

```
    defactivate(self,inputs):
```

```
        #Checkifthesumofweightedinputsexceedsthethreshold net_input =
```

```
        np.dot(self.weights, inputs)
```

```
        ifnet_input>=self.threshold:
```

```
            return1#Neuronfires(outputis1) else:
```

```
            return0#Neurondoesnotfire(outputis0)
```

```
#Exampleusage
```

```
if __name__=="__main__":
```

```
    # Define weights and threshold for the MCP neuron
```

```
    weights=np.array([0.5,-0.3,0.2])#Exampleweights
```

```
    threshold = 0# Example threshold
```

```
#CreateanMCPneuronwiththedefinedweightsandthreshold neuron =
```

```
MCPNeuron(weights, threshold)
```

```
#Defineinputpattern
```

```
inputs=np.array([1,0,1])#Exampleinputpattern
```

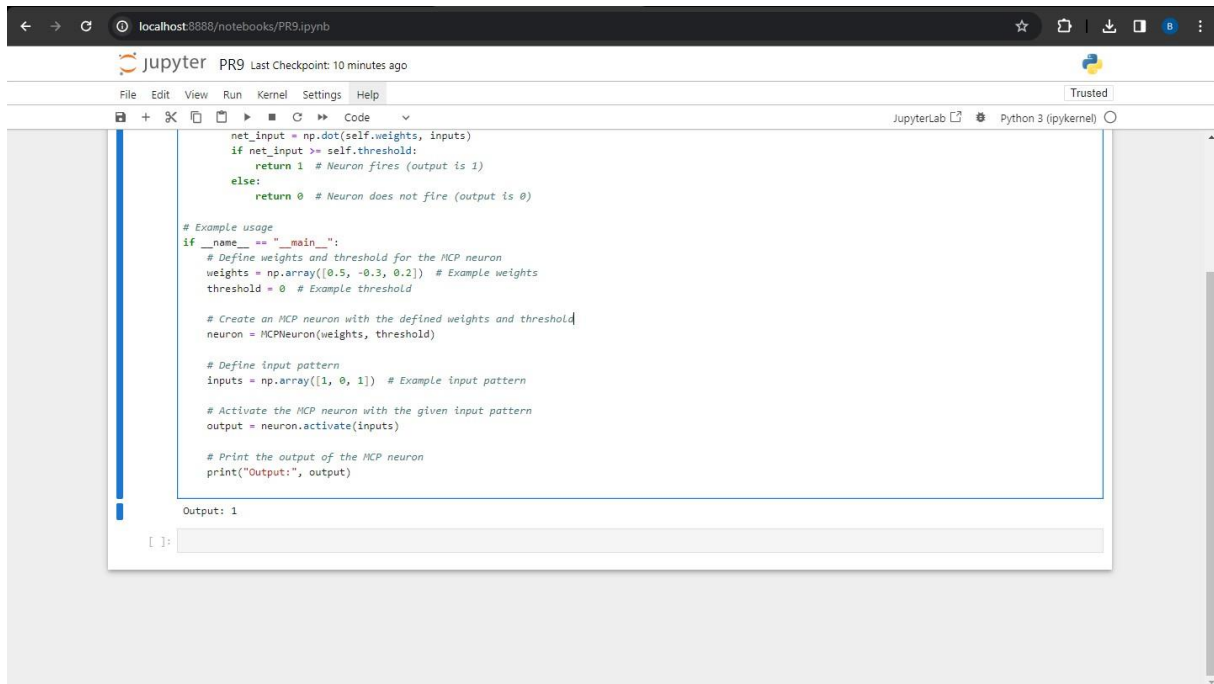
```
#ActivatetheMCPneuronwiththegiveninputpattern output
```

```
= neuron.activate(inputs)
```

#PrinttheoutputoftheMCPneuron

print("Output:", output)

Output:



```
net_input = np.dot(self.weights, inputs)
if net_input >= self.threshold:
    return 1 # Neuron fires (output is 1)
else:
    return 0 # Neuron does not fire (output is 0)

# Example usage
if __name__ == "__main__":
    # Define weights and threshold for the MCP neuron
    weights = np.array([0.5, -0.3, 0.2]) # Example weights
    threshold = 0 # Example threshold

    # Create an MCP neuron with the defined weights and threshold
    neuron = MCPNeuron(weights, threshold)

    # Define input pattern
    inputs = np.array([1, 0, 1]) # Example input pattern

    # Activate the MCP neuron with the given input pattern
    output = neuron.activate(inputs)

    # Print the output of the MCP neuron
    print("Output:", output)
```

Output: 1

Conclusion:

By performing this practical we learnt to implement McCulloch Pitts model in python.