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Batch: D

Branch: IT

```
In [1]:
```

```
import numpy as np
```

In [2]:

```
def unipolar_activation(input):
    if input > 0:
        return 1
    else:
        return 0
```

In [3]:

```
alpha=0.05
```

In [4]:

```
def perceptron activation(input, output y, input size):
    weights = np.zeros(input size + 1)
    for iteration in range(0,50):
        for num, i in enumerate(input):
            unipolar activation output = unipolar activation(np.sum(i * w
eights[1:]) + weights[0])
            difference = output y[num] - unipolar activation output
            weights[1:] += difference*alpha*i
            weights[0] += difference*alpha
    output = np.array([])
    print('The weights are : \n \n', weights)
    for i in input:
        unipolar activation output = unipolar activation(np.sum(i * weigh
ts[1:]) + weights[0])
        output = np.append(output, [unipolar_activation_output])
    difference = output y - unipolar activation output
    return output
```

```
In [5]:
input = [[0,0], [0,1], [1,0], [1,1]]
In [6]:
output = [0, 0, 0, 1]
In [7]:
print("----")
print(perceptron_activation(np.array(input), np.array(output), 2))
----AND----
The weights are:
[-0.1 0.1]
              0.05]
[0. 0. 0. 1.]
In [8]:
input = [[0,0], [0,1], [1,0], [1,1]]
In [9]:
output = [0, 1, 1, 1]
In [10]:
print("----")
print(perceptron_activation(np.array(input), np.array(output), 2))
----OR----
The weights are:
[0. 0.05 0.05]
[0. 1. 1. 1.]
In [11]:
input = [[0], [1]]
In [12]:
output = [1, 0]
```

```
In [13]:
```

```
print("----NOT-----")
print(perceptron_activation(np.array(input), np.array(output), 1))
----NOT----
The weights are :
   [ 0.05 -0.05]
[1. 0.]

In [14]:

def bipolar_activation(input):
   if input > 0:
        return 1
   else:
        return 1
```

In [15]:

```
alpha=0.05
```

In [16]:

```
def perceptron_activation_bipolar(input, output_y, input_size):
    weights = np.zeros(input_size + 1)
    for iteration in range(0,50):
        for num, i in enumerate(input):
            op = bipolar_activation(np.sum(i * weights[1:]) + weights[0])
            difference = output_y[num] - op
            weights[1:] += difference*alpha*i
                weights[0] += difference*alpha
    output = np.array([])
    print('The weights are : \n \n', weights)
    for i in input:
            op = bipolar_activation(np.sum(i * weights[1:]) + weights[0])
            output = np.append(output, [op])
    difference = output_y - op
    return output
```

In [17]:

```
input = [[-1, 1], [-1,1], [1,-1], [1,1]]
```

In [18]:

```
output = [-1, -1, -1, 1]
```

```
In [19]:
print("----")
print(perceptron_activation_bipolar(np.array(input), np.array(output), 2
))
----AND----
The weights are:
[-15. 5. -5.]
[1. 1. 1. 1.]
In [20]:
input = [[-1,-1], [-1,1], [1,-1], [1,1]]
In [21]:
output = [-1, 1, 1, 1]
In [22]:
print("----")
print(perceptron_activation_bipolar(np.array(input), np.array(output), 2
))
----OR----
The weights are:
[-5. 5. 5.]
[1. 1. 1. 1.]
In [23]:
input = [[-1], [1]]
In [24]:
output = [1, -1]
```

```
In [25]:
```

```
print("----NOT-----")
print(perceptron_activation_bipolar(np.array(input), np.array(output), 1
))
----NOT----
The weights are :
  [-5. -5.]
[1. 1.]
```

Conclusion:

The difference between the two activation functions

-> The threshold value that we use

Unipolar -- its 1 and 0

Bipolar its 1 and -1

- -> A single neuron can be used to train and implement the AND, OR and NOT logic gates successfully.
- ->More complex logical gates will require more neurons or more layers of single neurons.