

▼ Roll No.: 19BCE041

Course : Machine Learning

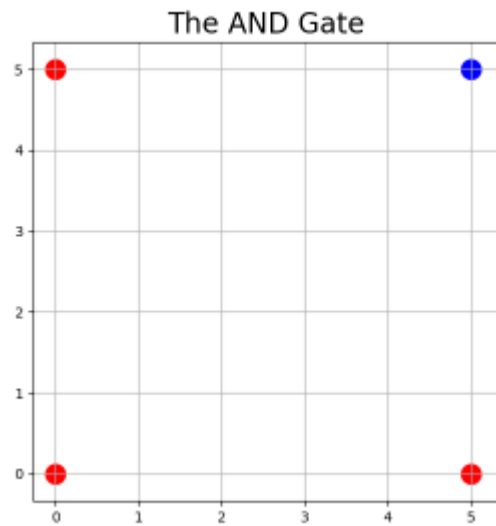
Course Code : 2CS501

Practical : 8

AND gate using Perceptron Learning

In the field of Machine Learning, the Perceptron is a Supervised Learning Algorithm for binary classifiers

A single perceptron can learn any function, as long as the instances in the dataset are linearly separable, like AND, OR, NAND, and NOR!



```
import numpy as np
```

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

```
n_datapoints = Data.shape[0]
```

```
n_dimensions = Data.shape[1]-1
```

```
w = 2*np.random.random_sample((n_dimensions)) - 1  
b=np.random.random()
```

```
lr = 0.1
```

epoches = 50

```
for ep in range(epoches):
    for i in range(n_datapoints):
        net_input = np.dot(w, Data[i, 0:n_dimensions]) + b

        A = net_input >= 0
        E = Data[i, n_dimensions] - A
        w = w + lr * E * (Data[i, 0:n_dimensions].T)
        b = b + lr * E
    print("Epoc:", ep, "weights: ", w, "bias: ", b)
```

```
Epoc: 0 weights: [ 0.47859906  0.14774628  0.07197222] bias: 0.013530876591565849
Epoc: 5 weights: [-0.47859906  0.04774628 -0.02802778] bias: -0.08646912340843416
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Epoc: 8 weights: [-0.07859906  0.14774628  0.07197222] bias: -0.18646912340843416
Epoc: 9 weights: [-0.07859906  0.14774628  0.07197222] bias: -0.18646912340843416
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Epoc: 9 weights: [0.02140094 0.14774628 0.07197222] bias: -0.18646912340843416
Epoc: 10 weights: [0.02140094 0.14774628 0.07197222] bias: -0.18646912340843416
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Epoc: 10 weights: [ 0.02140094  0.04774628 -0.02802778] bias: -0.28646912340843417
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Epoc: 10 weights: [ 0.02140094  0.04774628 -0.02802778] bias: -0.28646912340843417
Epoc: 10 weights: [ 0.02140094  0.04774628 -0.02802778] bias: -0.28646912340843417
Epoc: 10 weights: [0.12140094 0.14774628 0.07197222] bias: -0.18646912340843416
Epoc: 11 weights: [0.12140094 0.14774628 0.07197222] bias: -0.18646912340843416
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Epoc: 11 weights: [ 0.12140094  0.04774628 -0.02802778] bias: -0.28646912340843417
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Epoc: 11 weights: [ 0.12140094  0.04774628 -0.02802778] bias: -0.28646912340843417
Epoc: 11 weights: [0.22140094 0.14774628 0.07197222] bias: -0.18646912340843416
Epoc: 12 weights: [0.22140094 0.14774628 0.07197222] bias: -0.18646912340843416
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Epoc: 12 weights: [ 0.22140094  0.04774628 -0.02802778] bias: -0.28646912340843417
```

```
print("Final weights",w)
print("Final bias ",b)
```

```
Final weights [0.22140094 0.14774628 0.07197222]
Final bias -0.3864691234084342
```

```
prediciton = (np.dot(Data[:,0:n_dimensions],w)+b)>=0
prediciton
```

```
array([False, False, False, False, False, False, False,  True])
```

```
Final=[]
for i in prediciton:
    if i==True:
        Final.append(1)
    else:
        Final.append(0)
```

Final

```
[0, 0, 0, 0, 0, 0, 0, 1]
```

