

# Tarea 12 Perceptron

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```
In [1]: import numpy as np
        LEARNING_RATE = .1
```

## Funcion de activacion

```
In [2]: def activation(a):
        threshold = .5
        if(a >= threshold):
            return 1
        return 0
```

## Funcion de perceptron

```
In [3]: def perceptron(data,w,evaluate=False):
        global LEARNING_RATE
        if(evaluate):
            return np.array(list(map(activation,np.dot(data,w))))
        else:
            nW = w
            epoch = 0
            inp = data[:, :-1]
            results = data[:, -1]
            while(epoch < 100):
                epoch += 1
                purePrediction = np.dot(inp,nW)
                pred = np.array(list(map(activation,purePrediction)))
                error = (results - pred)
                nW = nW + LEARNING_RATE * np.dot(error,inp)
            return nW
```

## Training the Not Perceptron

```
In [4]: notTable = [
        [1,1,0],
        [0,1,1]
        ]
        notTable = np.array(notTable)
        notWeights = np.ones(2)
```

```
In [5]: notWeights = perceptron(notTable,notWeights)
        print(notWeights)
```

```
[-0.1  0.5]
```

```
In [6]: perceptron(notTable[:, :-1],notWeights,True) == notTable[:, -1]
```

```
Out[6]: array([ True,  True])
```

## Training for AND Perceptron

```
In [7]: andTable = [
        [0,0,1,0],
        [0,1,1,0],
        [1,0,1,0],
        [1,1,1,1]
        ]
        andTable = np.array(andTable)
        andWeights = np.ones(3)
```

```
In [8]: andWeights = perceptron(andTable,andWeights)
        print(andWeights)
```

```
[ 0.5  0.5 -0.2]
```

```
In [9]: perceptron(andTable[:, :-1], andWeights, True) == andTable[:, -1]
```

```
Out[9]: array([ True,  True,  True,  True])
```

## Training the OR Perceptron

```
In [10]: orTable = [
          [0,0,1,0],
          [0,1,1,1],
          [1,0,1,1],
          [1,1,1,1]
        ]
orTable = np.array(orTable)
orWeights = np.ones(3)
```

```
In [11]: orWeights = perceptron(orTable, orWeights)
print(orWeights)
```

```
[1.  1.  0.4]
```

```
In [12]: perceptron(orTable[:, :-1], orWeights, True) == orTable[:, -1]
```

```
Out[12]: array([ True,  True,  True,  True])
```

## Training the NOR Perceptron

```
In [13]: norTable = [
          [0,0,1,1],
          [0,1,1,0],
          [1,0,1,0],
          [1,1,1,0]
        ]
norTable = np.array(norTable)
norWeights = np.ones(3)
```

```
In [14]: norWeights = perceptron(norTable, norWeights)
print(norWeights)
```

```
[-0.1 -0.1  0.6]
```

```
In [15]: perceptron(norTable[:, :-1], norWeights, True) == norTable[:, -1]
```

```
Out[15]: array([ True,  True,  True,  True])
```

## Training for the XNOR Perceptron

```
In [16]: xnorTable = [
          [0,0,1,1],
          [0,1,1,0],
          [1,0,1,0],
          [1,1,1,1]
        ]
xnorTable = np.array(xnorTable)
xnorWeights = np.ones(3)
```

```
In [17]: xnorWeights = perceptron(xnorTable, xnorWeights)
print(xnorWeights)
```

```
[0.1 0.1 0.5]
```

```
In [18]: perceptron(xnorTable[:, :-1], xnorWeights, True) == xnorTable[:, -1]
```

```
Out[18]: array([ True, False, False,  True])
```

Debido a que no es posible resolver el problema del XNOR o del XOR con 1 solo perceptron se tiene que recurrir a conectar varios perceptrones en una red para resolver problema

```
In [19]: def andPerceptron(inputs):
          return perceptron(inputs, andWeights, True)
          def orPerceptron(inputs):
          return perceptron(inputs, orWeights, True)
          def notPerceptron(inputs):
          return perceptron(inputs, notWeights, True)
```

la compuerta XNOR se puede descomponer en  $\sim (\sim AB + \sim BA)$

definiremos  $C = \sim AB$

definiremos  $D = \sim BA$

por lo tanto  $XNOR = \sim(C+D)$

Y  $XNOR = \sim XOR$

por lo que empezaremos por la compuerta XOR

## Compuerta XOR

```
In [20]: xorTable = [  
    [0,0,1,0],  
    [0,1,1,1],  
    [1,0,1,1],  
    [1,1,1,0]  
]  
xorTable = np.array(xorTable)
```

```
In [21]: def xorNN(inputs):  
    A = inputs[:,0] ## Sacar columna A  
    B = inputs[:,1] ## Sacar columna B  
    ones = np.ones(inputs.shape[0]) ## Columna de 1  
    ## Agregamos Bias a las columnas  
    A = np.c_[A,ones]  
    B = np.c_[B,ones]  
    ## Usando el perceptron de NOT negamos los valores  
    notA = notPerceptron(A)  
    notB = notPerceptron(B)  
    ## Sacamos los valores de C y D  
    temp = np.c_[notA,B]  
    C = andPerceptron(temp)  
    temp = np.c_[notB,A]  
    D = andPerceptron(temp)  
    ## Agregamos una columna de bias para el perceptron que sigue  
    temp = np.c_[C,D,ones]  
    return orPerceptron(temp) ## Regresamos el resultado de el perceptron OR
```

```
In [22]: xorNN(xorTable[:, :-1]) == xorTable[:, -1]
```

```
Out[22]: array([ True,  True,  True,  True])
```

## NN XNOR

```
In [23]: def xnorNN(inputs):  
    return notPerceptron(np.c_[xorNN(inputs), np.ones(inputs.shape[0])])
```

```
In [24]: xnorNN(xnorTable[:, :-1]) == xnorTable[:, -1]
```

```
Out[24]: array([ True,  True,  True,  True])
```