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# Cálculo do gradiente

```
In [1]:
import numpy as np
In [2]:
def sigmoid(x):
    return 1/(1+np.exp(-x))
In [3]:
def sigmoid prime(x):
        return sigmoid(x) * (1 - sigmoid(x))
In [4]:
learnrate = 0.5
x = np.array([1, 3, 4, 7])
y = np.array(0.5)
b = 0.5
In [5]:
# Pesos iniciais
w = np.array([0.5, -0.5, 0.3, 0.2])
In [6]:
h = np.dot(x, w)+b
In [7]:
nn_output = sigmoid(h)
print(nn_output)
0.8909031788043871
In [8]:
# TODO: erro Calcular de rede neural
error = y - nn_output
print(error)
-0.3909031788043871
In [9]:
# TODO: Calcule o termo de erro
# Lembre-se, isso requer o gradiente de saída, para o qual não adicionamos
# especificamente uma variável.
error_term = error * sigmoid_prime(h)
```

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```
In [10]:
```

```
# TODO: Calcule a mudança nos pesos
del_w = learnrate * error_term * x
```

# In [11]:

```
print(del_w)
```

[-0.02256441 -0.04512882 -0.06769323 -0.09025764]

## In [12]:

```
w = w + del_w
```

#### In [13]:

```
print(w)
```

[ 0.47743559 -0.54512882 0.23230677 0.10974236]

# In [14]:

```
h = np.dot(x, w)+b
```

## In [15]:

```
nn_output = sigmoid(h)
print(nn_output)
```

0.7355697130589234