

```
self.biases = [np.random.randn(y, 1) for y in sizes[1:]]
self.weights = [np.random.randn(y, x)
                 for x, y in zip(sizes[:-1], sizes[1:])]

```

Since pass over for process

sizes = [764, 15, 10]

↑ ↑ ↑
0 1 2

self.weights = $\begin{pmatrix} w_{11} & \dots & \dots \\ w_{21} & \dots & \dots \\ \vdots & \ddots & \vdots \end{pmatrix}$ } 15

764

$[w]_{10,15}$

} 10

```
def feedforward(self, a):
    """Return the output of the network if 'a' is input."""
    for b, w in zip(self.biases, self.weights):
        a = sigmoid(np.dot(w, a) + b)
    return a

```

Sigmoid

iteraciones de ciclo

stochastic gradient descent

```
def SGD(self, training_data, epochs, mini_batch_size, eta,
        test_data=None):
    """Train the neural network using mini-batch stochastic gradient descent."""

```

$$\text{training_data} = \left[(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots, \right]$$

One hot-encoding:

```
mini_batches = [
    training_data[k:k+mini_batch_size]
    for k in range(0, n, mini_batch_size)]
for mini_batch in mini_batches:
```

Índice de la lista en cada mini_batch

$$\text{training_data} = \left[\underbrace{(x_1, y_1), (x_2, y_2), (x_3, y_3), \dots,}_{1 \text{ mini_batch}} \right] \underbrace{\hspace{10em}}_{2^{\circ} \text{ mini_batch}}$$

```
mini_batches = [ → lista de mini_batches
    training_data[k:k+mini_batch_size]
    for k in range(0, n, mini_batch_size)]
for mini_batch in mini_batches:
    self.update_mini_batch(mini_batch, eta)
if test_data:
```

Nos pasa el código y nos reclutamos

Separamos los datos en mini batches

Estocástico →