# **MicroPython - Neural Network**

Implement Neural Network Deep Feed Forward on micro-controller using MicroPython

Olivier Lenoir

olivier.len02@gmail.com

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#### Abstract

Implement Neural Network Deep Feed Forward on micro-controller using MicroPython. This project is designed in pure MicroPython.



# 1 Requirements

Download *matrix.py*<sup>1</sup> and *neuralnetwork.py*<sup>2</sup> and copy them on your MicroPython board. The same code can be used on your computer with Python.

# 2 Get Started

I'm going to describe how work *MicroPython - Neural Network* with a very small an simple example. In this classifier we are using a *Sigmoid* activation function as  $\sigma(x)$  and his derivative as  $\sigma'(x)$ . Here is what we want to predict, with  $i_n$  as inputs and  $s_n$  as the expected classification.

$i_1$	$i_2$	$i_3$	$s_1$	$s_2$
0	0	0	1	0
0	0	1	0	1
0	1	1	0	1
0	1	0	1	0
1	1	0	0	1
1	1	1	1	0
1	0	1	0	1
1	0	0	1	0



We use a neural network (3, 4, 2). With 3 values in the input layer, 4 values in the hidden layer and 2 values in the output layer.

<sup>&</sup>lt;sup>1</sup>mafrix.py: https://gitlab.com/olivier.len02/MicroPython-Matrix/-/blob/master/micropython/matrix.py <sup>2</sup>neuralnetwork.py: https://gitlab.com/olivier.len02/MicroPython-NeuralNetwork/-/blob/master/micropython/ neuralnetwork.py



Figure 1: Neural network (3, 4, 2) detail

Input layer is represented by the matrix I and the output by the matrix O. Hidden layer is matrix H. Weights between matrix I and H is matrix  $W_{ih}$ . Weights between matrix H and O is matrix  $W_{ho}$ .

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$
  
$$\sigma'(x) = \sigma(x) \cdot (1 - \sigma(x))$$
(1)

$$\sigma'(x_{\sigma}) = x_{\sigma} \cdot (1 - x_{\sigma})$$

$$I = \begin{pmatrix} i_1 & i_2 & i_3 \end{pmatrix} \tag{2}$$

$$W_{ih} = \begin{pmatrix} w_{1,1}^{ih} & w_{1,2}^{ih} & w_{1,3}^{ih} & w_{1,4}^{ih} \\ w_{2,1}^{ih} & w_{2,2}^{ih} & w_{2,4}^{ih} \\ w_{3,1}^{ih} & w_{3,2}^{ih} & w_{3,3}^{ih} & w_{3,4}^{ih} \end{pmatrix}$$
(3)

$$H = \begin{pmatrix} h_1 & h_2 & h_3 & h_4 \end{pmatrix} \tag{4}$$

$$W_{ho} = \begin{pmatrix} w_{1,1}^{ho} & w_{1,2}^{ho} \\ w_{2,1}^{ho} & w_{2,2}^{ho} \\ w_{3,1}^{ho} & w_{3,2}^{ho} \\ w_{4,1}^{ho} & w_{4,2}^{ho} \end{pmatrix}$$
(5)

$$O = \begin{pmatrix} o_1 & o_2 \end{pmatrix} \tag{6}$$

$$S = \begin{pmatrix} s_1 & s_2 \end{pmatrix} \tag{7}$$

From those matrices you can propagate the input I to the output O using the following calculations:

$$H = \sigma(I \cdot W_{ih})$$

$$O = \sigma(H \cdot W_{ho})$$
(8)

If the network is properly trained, the output O should be very close from the expected S matrix. If not, this mind we need to train the artificial neural network with the training data-set and back-propagate the error to adjust weights.

We are now going to use propagated results to back-propagate the error of each layers.  $E_o$  and  $E_h$  are error of layers O and H. Matrix  $gW_{ho}$  and  $gW_{ih}$  are the gradient to adjust weights.

$$E_o = (S - O) \times \sigma'(O)$$

$$gW_{ho} = H^T \cdot E_o$$
(9)

$$E_h = (E_o \cdot W_{ho}) \times \sigma'(H)$$

$$gW_{ih} = I^T \cdot E_h$$
(10)

$$W_{ho} = W_{ho} + gW_{ho}$$

$$W_{ih} = W_{ih} + gW_{ih}$$
(11)

Weights are updated. Training continue until we are satisfied with the result.

#### 2.1 Train Neural Network

I recommend training *Neural Network* on a computer. Otherwise you may quickly run into memory error on your MicroPython board, even if you use garbage collect.

```
from matrix import Matrix
import neuralnetwork as nn
# Create neural network with an input layer of 3, an hidden layer of 4
# and an output layer of 2 by default the activation function is sigmoid()
# but a ReLU also exist as relu()
ann = nn.DFF((3, 4, 2))
```

Now let's create a training set with input matrix and output matrix.

```
training_set = [
    [Matrix([[0, 0, 0]]), Matrix([[1, 0]])],
    [Matrix([[0, 0, 1]]), Matrix([[0, 1]])],
    [Matrix([[0, 1, 1]]), Matrix([[0, 1]])],
    [Matrix([[0, 1, 0]]), Matrix([[1, 0]])],
    [Matrix([[1, 1, 0]]), Matrix([[0, 1]])],
    [Matrix([[1, 1, 1]]), Matrix([[1, 0]])],
    [Matrix([[1, 0, 1]]), Matrix([[0, 1]])],
    [Matrix([[1, 0, 0]]), Matrix([[1, 0]])],
    [Matrix([[1, 0]])]],
```

We train the network a thousand times with the training set. The learning rate (*Irate*) is set to one by default

```
print('Learning progress')
for i in range(1000):
    for a, s in training_set:
        ann.train(a, s, lrate=1)
    if i % 10 == 0:
        print('.', end='')
```

Check if you are satisfied with the training.

```
def short(a):
    return round(a, 3)

print('=' * 20)
score = True
for a, s in training_set:
    p = ann.predict(a)
    scr = str(p.map(round)) == str(s.map(short))
    print(a.map(short), p.map(short), p.map(round), s.map(short), scr)
    score &= scr
print('Good_learning?', score)
```

Print out weights.

```
print('=' * 20)
print('Rounded_weights')
for i, w in enumerate(ann.weights):
    print('W{}'.format(i), w.map(round))
```

### 2.2 Predict

With the trained weights, we can now use our network.

# References

- (1) Jean-Claude Heudin, Comprendre le deep learning, une introduction aux réseaux de neurones, Science-eBook, Octobre 2016, ISBN 979-10-91245-44-9.
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- (3) Damien George, *MicroPython*, George Robotics Limited, https://micropython.org/.
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- (5) Olivier Lenoir, *MicroPython Matrix*, GitLab, January 2021, https://gitlab.com/olivier.len02/ MicroPython-Matrix/.
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