### Assignment 1

太空三 109607504 賴以芸

### 1. 編譯結果

eeeeyun0301@LAPTOP-V9DEG1U3:/mnt/d/assignment\_1-yeeeeyun0301/src\$ gcc -o main main.c layer.c neuron.c -1m,

## 2. 執行結果

```
Enter input to test:

0

0

Output: 0

Enter input to test:
0
1

Output: 1

Enter input to test:
1

Output: 1

Enter input to test:
1

Enter input to test:
1

Output: 1

Enter input to test:
1

Output: 0
```

```
yeeceyun0301@LAPTOP-V9DBG1U3:/mnt/d/assignment_1-yeeceyun0301/src$ ./main Enter the number of Layers in Neuval Network:

3 Enter number of neurons in layer[1]:
2 Enter number of neurons in layer[2]:
4 Enter number of neurons in layer[3]:
1 Created Layer: 1
Number of Neurons in Layer 1: 2
Number of Neurons in Layer 1: 2
Number of Neurons in Layer 1 created
Neuron 1 in Layer 1 created
Created Layer: 2
Number of Neurons in Layer 2: 4
Neuron 1 in Layer 2 created
Neuron 3 in Layer 2 created
Neuron 4 in Layer 2 created
Neuron 5 in Layer 2 created
Neuron 6 Neurons in Layer 3: 1
Number of Neurons in Layer 3: 1
Neuron 1 in Layer 3 created

Initializing weights...
0:w[0][0]: 0.840934
1:w[0][0]: 0.962804
2:w[0][0]: 0.325800
3:w[0][1]: 0.272300
1:w[0][1]: 0.272300
1:w[0][1]: 0.539701
3:w[0][1]: 0.539701
3:w[0][1]: 0.572473
0:w[1][0]: 0.133101
0:w[1][1]: 0.272046
0:w[1][3]: 0.298000
Neural Network Created Successfully...
```

```
Enter the learning rate (Usually 0.15):
0.15

Enter the number of training examples:
4

Enter the Inputs for training example[0]:
0

Enter the Inputs for training example[1]:
0

Enter the Inputs for training example[2]:
1
0

Enter the Inputs for training example[3]:
1

Enter the Desired Outputs (Labels) for training example[0]:
0

Enter the Desired Outputs (Labels) for training example[1]:
1

Enter the Desired Outputs (Labels) for training example[2]:
1

Enter the Desired Outputs (Labels) for training example[2]:
1

Enter the Desired Outputs (Labels) for training example[3]:
0
```

# 3. 分析

隱藏層層數	隱藏層 neuron 數		執行成果	結果
1	2		Enter input to test:  0 0 Output: 1 Enter input to test: 0 1 Output: 1 Enter input to test: 1 1 Output: 0 Enter input to test: 1 0 Output: 1	(0,0)的輸出結果錯誤
1	3		Enter input to test:  0 0 0 Output: 0 Enter input to test: 0 1 Output: 1 Enter input to test: 1 0 Output: 1 Enter input to test: 1 1 Coutput: 0	皆正確
2	2	2	Enter input to test:  0 0 0 Output: 0 Enter input to test: 0 1 Output: 1 Enter input to test: 1 0 Output: 1 Enter input to test: 1 0 Output: 0	皆正確

可以看到利用類神經學習在計算 XOR 運算中,只要中間隱藏層有多一層,或是在隱藏層多一個神經元,都可以達到我們預想的計算成果。

## 4. 心得

在學期初的時候我一直無法理解類神經網路學習的架構,為什麼要訓練模型時要 backpropagation,權重更新,但從老師提供的網站連結以及助教課,再加上我剛好這個學期系上的必修課內容為機器學習,讓我搞清楚類神經網路學習的步驟,以及模型訓練的相關知識。雖然今天要運算的是 XOR 而已,但此架構可大幅運用在較複雜的計算中,也難怪類神經網路成為現今熱門的話題,雖然一開始要花時間理解,但會使用之後很多複雜的問題都能迎刃而解,不需要人類反覆計算,丟給電腦運算就好,人類就能去做更多有意義的思考或精進。