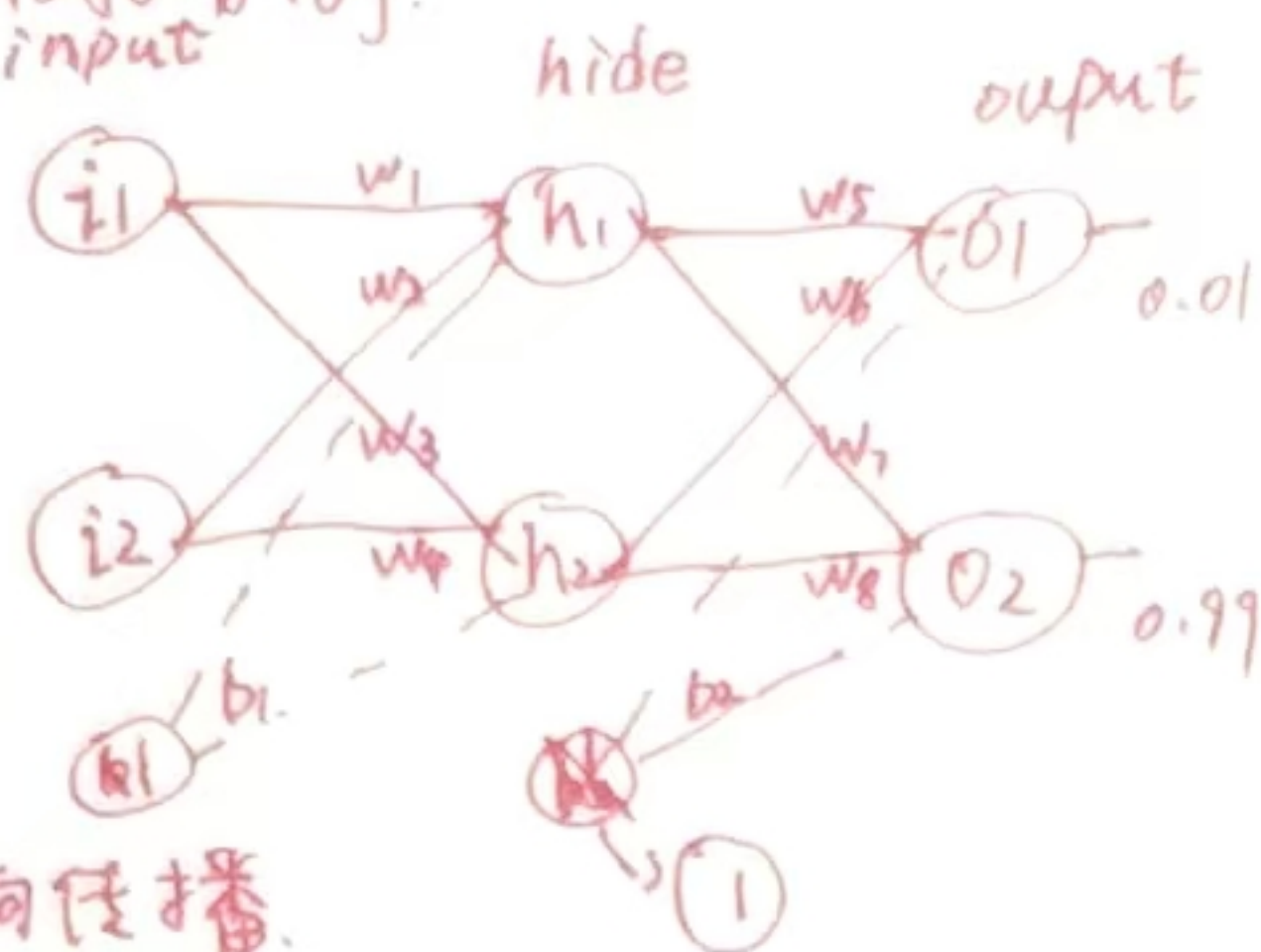


# 手推神经网络训练过程

以课上的例子为例子。



$i_1 = 0.05$	$w_1 = 0.15$	$w_7 = 0.50$
$i_2 = 0.1$	$w_2 = 0.20$	$w_8 = 0.55$
	$w_3 = 0.25$	
	$w_4 = 0.30$	$b_1 = 0.35$
	$w_5 = 0.40$	$b_2 = 0.60$
	$w_6 = 0.45$	

step 1. 前向传播.

input  $\rightarrow$  hide:

( $z$ : 神经元的加权输入和  
中间经过激活函数 sigmoid.  
 $a$ : 神经元的加权输出和)

$$z_{h1} = i_1 \times w_1 + i_2 \times w_2 + b_1 \times 1$$

$$= 0.05 \times w_1 + 0.1 \times 0.2 + 0.35$$

$$= 0.3775$$

$\downarrow$  sigmoid

$$a_{h1} = \frac{1}{1 + e^{-z_{h1}}} = \frac{1}{1 + e^{-0.3775}}$$

$$= 0.0593269992$$

$$z_{h2} = i_1 \times w_3 + i_2 \times w_4 + b_1 \times 1$$

$$= 0.3925$$

$$a_{h2} = \frac{1}{1 + e^{-z_{h2}}} = 0.596884378$$

hide  $\rightarrow$  output:

$$z_{o1} = a_{h1} \times w_5 + a_{h2} \times w_6 + b_2 \times 1$$

$$= 1.105905967$$

$$a_{o1} = \frac{1}{1 + e^{-z_{o1}}} = 0.751365069$$

$$z_{o2} = a_{h1} \times w_7 + a_{h2} \times w_8 + b_2 \times 1$$

$$= 1.2249214$$

$$a_{o2} = \frac{1}{1 + e^{-z_{o2}}} = 0.77292846$$

step 1 正向传播结束.

输出结果.	0.75136..., 0.77292...	} 差很多.
正确答案	0.01, 0.99.	



反向传播, 更新权重, 重新输出



## step 2. 反向传播.

### 1. 使用MSE损失函数.

$$E_{total} = \sum \frac{1}{n} (\text{target} - \text{output})^2$$

↓  
输出值个数.

分别计算两个值的E, 再相加.

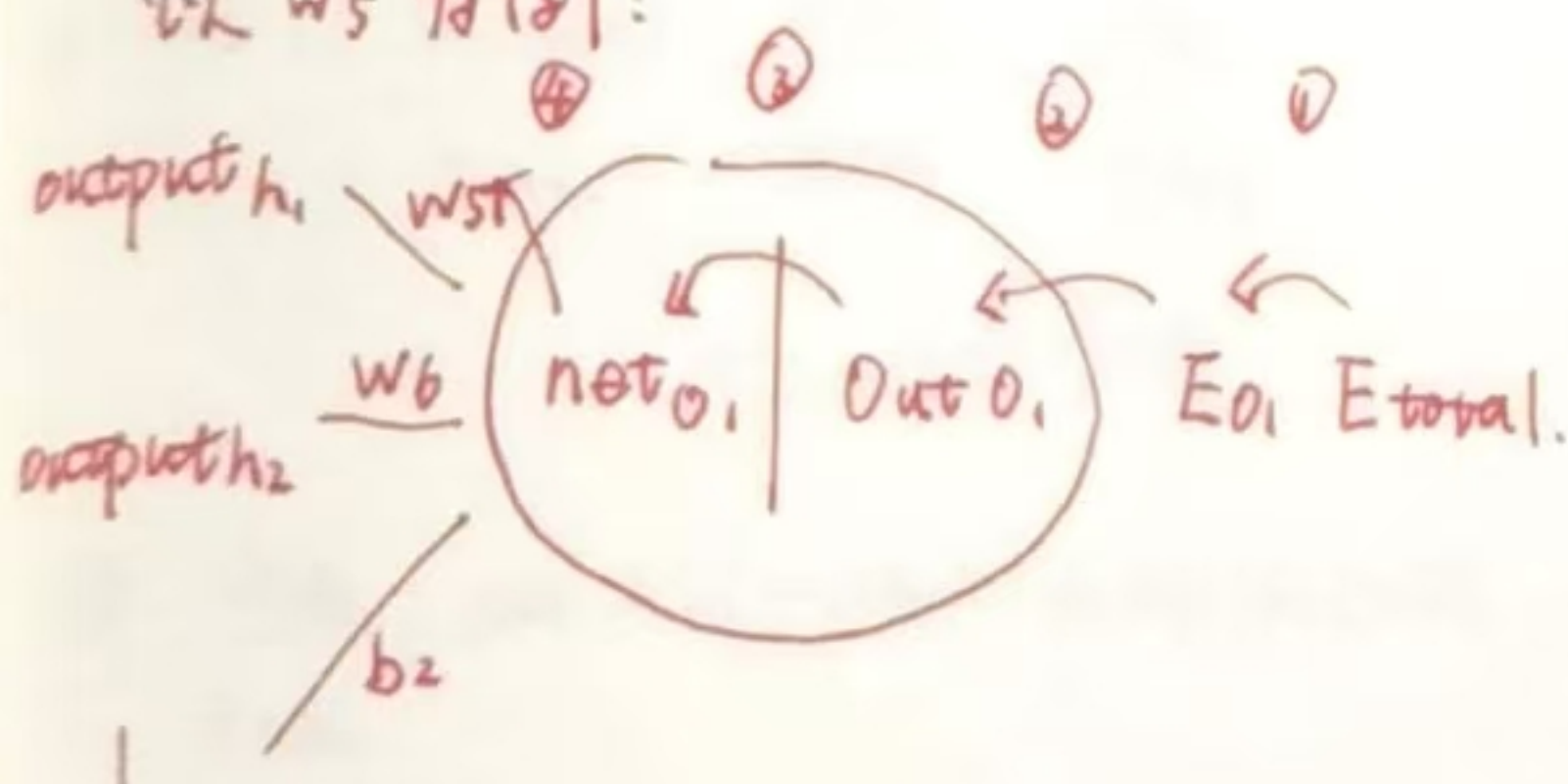
$$E_{01} = \frac{1}{2} (0.01 - 0.751365069)^2 = 0.274811083$$

$$E_{02} = \frac{1}{2} (0.99 - 0.72928465)^2 = 0.023560026$$

$$E_{total} = E_{01} + E_{02} = 0.298371109$$

### 2. 隐藏层 → 输出层 权值更新

以  $w_5$  为例:



↓ 求偏导

$$\frac{\partial E_{total}}{\partial w_5} = \frac{\partial E_{total}}{\partial \text{out}_{01}} \times \frac{\partial \text{out}_{01}}{\partial \text{net}_{01}} \times \frac{\partial \text{net}_{01}}{\partial w_5}$$

↓ 拆开算

$$\textcircled{1} E_{total} = \frac{1}{2} (\text{target}_{01} - a_{01})^2 + \frac{1}{2} (\text{target}_{02} - a_{02})^2$$

$$\frac{\partial E_{total}}{\partial a_{01}} = 2 \times \frac{1}{2} (\text{target}_{01} - a_{01}) \times (-1) = 0.741365069$$

$$\textcircled{2} a_{01} = \frac{1}{e^{-z_{01}} + 1}$$

$$\frac{\partial a_{01}}{\partial z_{01}} = a_{01} \times (1 - a_{01}) = 0.18685062$$

$$\textcircled{4} z_{01} = w_5 \times a_{h1} + w_6 \times a_{h2} + b_2 \times 1 \quad \frac{\partial z_{01}}{\partial w_5} = a_{h1} = 0.593269992$$

$$\therefore \frac{\partial E_{total}}{\partial w_5} = \textcircled{2} \times \textcircled{3} \times \textcircled{4} = 0.082167041$$

↓ 更新  $w_5$ .

$$w_5^+ = w_5 - \eta \times \frac{\partial E_{total}}{\partial w_5}$$

↓  
学习率.

$$= 0.4 - \underbrace{0.5}_{\text{经验值}} \times 0.082167041 = 0.35891648$$

同理:

$$w_6^+ = 0.408666186$$

$$w_7^+ = 0.511301270$$

$$w_8^+ = 0.561370121$$

### 3. 隐藏层 → 输入层

以  $w_1$  为例.

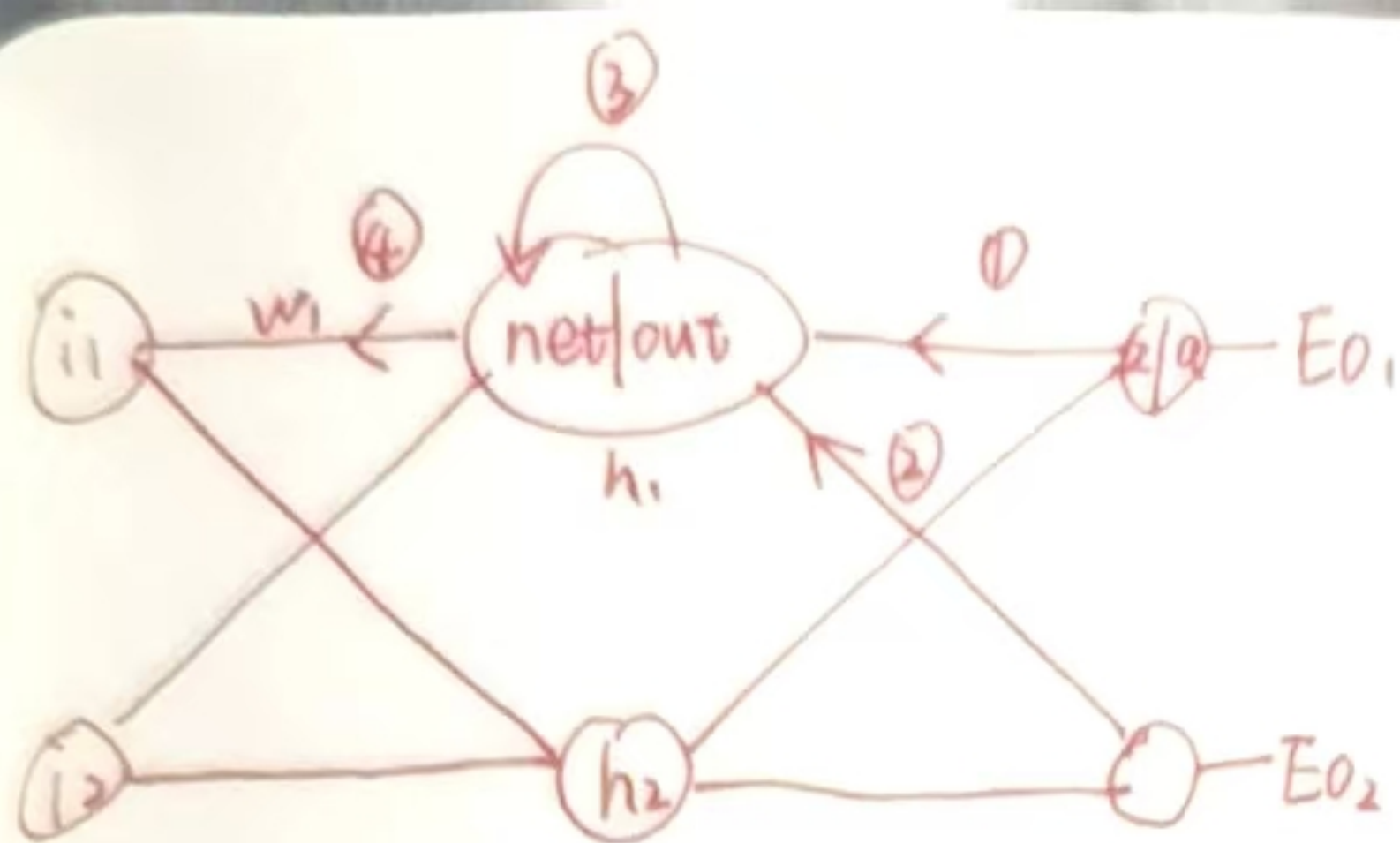
先看左  
再看右

$$\frac{\partial E_{total}}{\partial w_1} = \frac{\partial E_{total}}{\partial \text{out}_{h1}} \times \frac{\partial \text{out}_{h1}}{\partial \text{net}_{h1}} \times \frac{\partial \text{net}_{h1}}{\partial w_1}$$

$$\frac{\partial E_{total}}{\partial \text{out}_{h1}} = \frac{\partial E_{01}}{\partial \text{out}_{h1}} + \frac{\partial E_{02}}{\partial \text{out}_{h1}}$$

$\because \text{out}_{h1}$  同时影响  $E_{01}$  和  $E_{02}$





$$E_{total} = E_{01} + E_{02}$$

$$\textcircled{1}: \frac{\partial E_{01}}{\partial a_{h1}} = \frac{\partial E_{01}}{\partial a_{01}} \times \frac{\partial a_{01}}{\partial z_{01}} \times \frac{\partial z_{01}}{\partial a_{h1}}$$

$$= 0.0055399425$$

$$\text{同理 } \textcircled{2}: \frac{\partial E_{02}}{\partial a_{h1}} = -0.19049119$$

$$\textcircled{1} + \textcircled{2}: \frac{\partial E_{total}}{\partial a_{h1}} = \frac{\partial E_{01}}{\partial a_{h1}} + \frac{\partial E_{02}}{\partial a_{h1}}$$

$$= 0.036350306$$

$$\textcircled{3} \frac{\partial a_{h1}}{\partial z_{h1}} = a_{h1} \times (1 - a_{h1}) = 0.2413007086$$

$$\textcircled{4} \frac{\partial z_{h1}}{\partial w_1} = i_1 = 0.05$$

↓

$$\frac{\partial E_{total}}{\partial w_1} = (\textcircled{1} + \textcircled{2}) \times \textcircled{3} \times \textcircled{4} = 0.000438568$$

↓

更新  $w_1$  权值.

$$w_1^+ = w_1 - \eta \times \frac{\partial E_{total}}{\partial w_1} = 0.15 - 0.5 \times 0.000438568 = 0.149780716$$

同理.  $w_2^+ = 0.19956143$

$$w_3^+ = 0.24975114$$

$$w_4^+ = 0.29950229$$

一次正向、反向完成.  $w_1 \sim w_4$  都更新了.

↓

再迭代.

↓

直到误差满足要求, 训练完成