

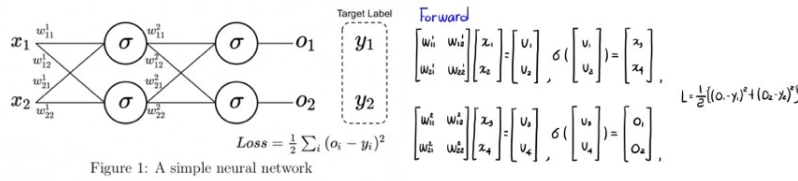
E412 Foundation of Big Data Analytics, Fall 2021

HW4

20170616 정희진

1. Neural Nets and Deep Learning

(a) Compute the gradients using the chain rule



Backward

$$\frac{\partial L}{\partial o_i} = o_i - y_i$$

$$\frac{\partial L}{\partial o_2} = o_2 - y_2$$

$$\frac{\partial L}{\partial u_3} = \frac{\partial L}{\partial o_1} \cdot \frac{\partial o_1}{\partial u_3} = o_1(1-o_1) \frac{\partial L}{\partial o_1} = o_1(1-o_1)(o_1 - y_1)$$

$$\frac{\partial L}{\partial u_4} = \frac{\partial L}{\partial o_2} \cdot \frac{\partial o_2}{\partial u_4} = o_2(1-o_2) \frac{\partial L}{\partial o_2} = o_2(1-o_2)(o_2 - y_2)$$

$$\frac{\partial L}{\partial w_{11}} = \frac{\partial L}{\partial u_3} \cdot \frac{\partial u_3}{\partial w_{11}} = z_3 \frac{\partial L}{\partial u_3} = z_3 o_1(1-o_1)(o_1 - y_1)$$

$$\frac{\partial L}{\partial w_{12}} = \frac{\partial L}{\partial u_3} \cdot \frac{\partial u_3}{\partial w_{12}} = z_3 \frac{\partial L}{\partial u_3} = z_3 o_1(1-o_1)(o_1 - y_1)$$

$$\frac{\partial L}{\partial w_{21}} = \frac{\partial L}{\partial u_4} \cdot \frac{\partial u_4}{\partial w_{21}} = z_4 \frac{\partial L}{\partial u_4} = z_4 o_2(1-o_2)(o_2 - y_2)$$

$$\frac{\partial L}{\partial w_{22}} = \frac{\partial L}{\partial u_4} \cdot \frac{\partial u_4}{\partial w_{22}} = z_4 \frac{\partial L}{\partial u_4} = z_4 o_2(1-o_2)(o_2 - y_2)$$

$$\frac{\partial L}{\partial x_1} = \frac{\partial L}{\partial u_3} \cdot \frac{\partial u_3}{\partial x_1} + \frac{\partial L}{\partial u_4} \cdot \frac{\partial u_4}{\partial x_1} = w_{11}^* \frac{\partial L}{\partial u_3} + w_{21}^* \frac{\partial L}{\partial u_4} = w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2)$$

$$\frac{\partial L}{\partial x_2} = \frac{\partial L}{\partial u_3} \cdot \frac{\partial u_3}{\partial x_2} + \frac{\partial L}{\partial u_4} \cdot \frac{\partial u_4}{\partial x_2} = w_{12}^* \frac{\partial L}{\partial u_3} + w_{22}^* \frac{\partial L}{\partial u_4} = w_{12}^* o_1(1-o_1)(o_1 - y_1) + w_{22}^* o_2(1-o_2)(o_2 - y_2)$$

$$\frac{\partial L}{\partial u_1} = \frac{\partial L}{\partial x_1} \cdot \frac{\partial x_1}{\partial u_1} = z_3 \frac{\partial L}{\partial x_1} = z_3 \left(w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2) \right)$$

$$\frac{\partial L}{\partial u_2} = \frac{\partial L}{\partial x_2} \cdot \frac{\partial x_2}{\partial u_2} = z_4 \frac{\partial L}{\partial x_2} = z_4 \left(w_{12}^* o_1(1-o_1)(o_1 - y_1) + w_{22}^* o_2(1-o_2)(o_2 - y_2) \right)$$

$$\frac{\partial L}{\partial w_{11}} = \frac{\partial L}{\partial u_1} \cdot \frac{\partial u_1}{\partial w_{11}} = z_1 \frac{\partial L}{\partial u_1} = z_1 z_3 \left(w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2) \right)$$

$$\frac{\partial L}{\partial w_{12}} = \frac{\partial L}{\partial u_1} \cdot \frac{\partial u_1}{\partial w_{12}} = z_1 \frac{\partial L}{\partial u_1} = z_1 z_3 \left(w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2) \right)$$

$$\frac{\partial L}{\partial w_{21}} = \frac{\partial L}{\partial u_2} \cdot \frac{\partial u_2}{\partial w_{21}} = z_2 \frac{\partial L}{\partial u_2} = z_2 z_4 \left(w_{12}^* o_1(1-o_1)(o_1 - y_1) + w_{22}^* o_2(1-o_2)(o_2 - y_2) \right)$$

$$\frac{\partial L}{\partial w_{22}} = \frac{\partial L}{\partial u_2} \cdot \frac{\partial u_2}{\partial w_{22}} = z_2 \frac{\partial L}{\partial u_2} = z_2 z_4 \left(w_{12}^* o_1(1-o_1)(o_1 - y_1) + w_{22}^* o_2(1-o_2)(o_2 - y_2) \right)$$

Answer

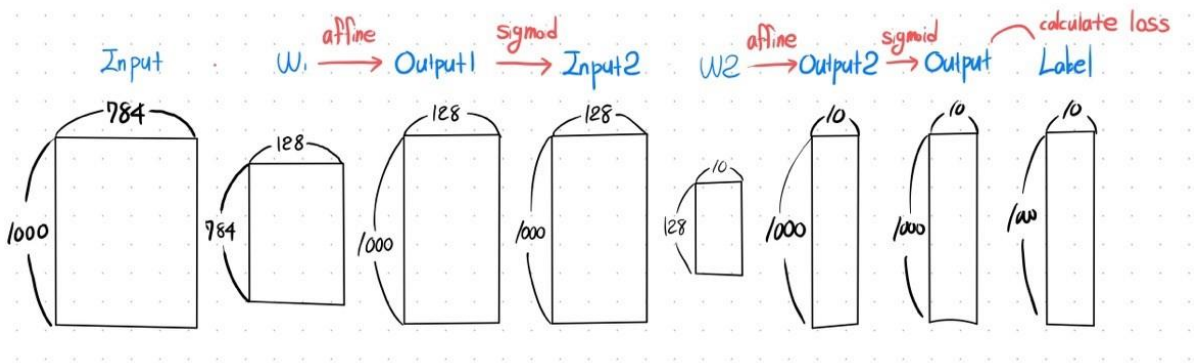
$$\begin{aligned} \frac{\partial L}{\partial w_{11}} &= z_1 z_3 (1-z_3) \left(w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2) \right), & \frac{\partial L}{\partial w_{12}} &= z_1 z_3 (1-o_1)(o_1 - y_1) \\ \frac{\partial L}{\partial w_{21}} &= z_2 z_4 (1-z_4) \left(w_{12}^* o_1(1-o_1)(o_1 - y_1) + w_{22}^* o_2(1-o_2)(o_2 - y_2) \right), & \frac{\partial L}{\partial w_{22}} &= z_2 z_4 (1-o_2)(o_2 - y_2) \\ \frac{\partial L}{\partial u_1} &= z_1 z_3 (1-z_3) \left(w_{11}^* o_1(1-o_1)(o_1 - y_1) + w_{21}^* o_2(1-o_2)(o_2 - y_2) \right), & \frac{\partial L}{\partial u_2} &= z_2 z_4 (1-o_2)(o_2 - y_2) \\ \frac{\partial L}{\partial u_3} &= z_3 o_1(1-o_1)(o_1 - y_1), & \frac{\partial L}{\partial u_4} &= z_4 o_2(1-o_2)(o_2 - y_2) \end{aligned}$$

or

$$\begin{cases} z_3 = \sigma(w_{11}^* x_1 + w_{12}^* x_2) \\ z_4 = \sigma(w_{21}^* z_3 + w_{22}^* z_4) \end{cases}$$

(b) Implement a fully-connected network to distinguish digits using Python.

The structure of FCN (without bias) is below figure. However, in hw4_1_p2.py, bias is added to that figure.



Result is below figure.

```
0.965
0.782
10000
1.0
```

2. Mining Data Streams

(a) Solve the following problems, which are based on the exercises in the Mining of Massive Datasets 3rd edition (MMDS) textbook.

Exercise 4.4.1 and 4.4.2

(a)	x	h(x)	h(x)	tail length
	3	7	00111	0
	1	3	00011	0
	4	9	01001	0
	1	3	00011	0
	5	11	01011	0
	9	19	10011	0
	2	5	00101	0
	6	13	01101	0
	5	11	01011	0

Estimate: $2^8 \cdot 1$

(b)	x	h(x)	h(x)	tail length
	3	16	10000	4
	1	10	01010	1
	4	19	10011	0
	1	10	01010	1
	5	22	10110	1
	9	2	00010	1
	2	13	01101	0
	6	25	11001	0
	5	22	10110	1

Estimate: $2^8 \cdot 16$

(c)	x	h(x)	h(x)	tail length
	3	12	01100	2
	1	4	00100	2
	4	16	10000	4
	1	4	00100	2
	5	20	10100	2
	9	2	00010	1
	2	8	01000	3
	6	24	11000	3
	5	20	10100	2

Estimate: $2^8 \cdot 16$

(a)에서처럼 2는 무조건 짝수를 만들어내고 거기다 차를 하면 홀수를 만들어 버므로 마지막 bit가 무조건 1이되어 tail length가 무조건 0이 된다. 이러한 상황은 미리 방지해야 한다.

Exercise 4.5.3

i	1	2	3	4	5	6	7	8	9
X_i	3	1	4	1	3	4	2	1	2
X_i value	2	3	2	2	1	1	2	1	1
Estimate	27	45	27	27	9	9	27	9	9

Average of Estimate = $\frac{27 \times 4 + 45 + 9 \times 4}{9} = 21$

(b) Implement the DGIM algorithm.

When $k_1=1$, $k_2=10$, $k_3=100$, The result is

```
0.5  
2.0  
23.0  
200.5
```