
Machine Learning

Answer Sheet for Homework 8

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Problem 1

1. Forward:

$$(A + 1) \times B + (B + 1) \times 1 = (A + 2) B + 1 \quad (1)$$

2. Backward:

$\delta_1^{(L)} = -2 \left(y_n - s_1^{(L)} \right) x_i^{(L-1)}$ counts and

$$\frac{\partial e_n}{\partial w_{ij}^{(\ell)}} = \delta_j^{(\ell)} x_i^{(\ell-1)} \text{ for } 0 \leq i \leq d^{(\ell-1)} \text{ and } 1 \leq j \leq d^{(\ell)} \quad (2)$$

with

$$\delta_j^{(\ell)} = \sum_k \left(\delta_k^{(\ell+1)} \right) \left(w_{jk}^{(\ell+1)} \right) \left(\tanh' \left(s_j^\ell \right) \right) \quad (3)$$

So one backward counts

$$\underbrace{(B + 1) \times 1}_{\text{output layer}} + \underbrace{B \times (A + 1)}_{\text{hidden layer}} + \underbrace{B}_{\text{hidden layer } \delta_j^{(\ell)}} = (A + 3) B + 1 \quad (4)$$

Hence, total number of operations required in a single iteration of backpropagation is

$$((A + 2) B + 1) + ((A + 3) B + 1) = (2A + 5) B + 2 \quad (5)$$

□

Problem 2

Suppose we have k hidden layers, which means $L = k + 1$, with $d^{(1)}, d^{(2)}, \dots, d^{(k)}$ units ($x_0^{(\ell)}$ is not counted here) in each layer. The number of total weights is

$$\sum_{i=0}^{k-1} (d^{(i)} + 1) d^{(i+1)} + (d^{(k)} + 1) \times 1 = \sum_{i=0}^{k-1} d^{(i)} d^{(i+1)} + \sum_{j=1}^k d^{(j)} + (d^{(k)} + 1) := N_w \quad (6)$$

with

$$\sum_{j=1}^k (d^{(j)} + 1) = \left(\sum_{j=1}^k d^{(j)} \right) + k = 36 \text{ and } d^{(0)} = 9 \quad (7)$$

So we have

$$N_w = (37 - k) + 9d^{(1)} + \left(\sum_{i=1}^{k-1} d^{(i)} d^{(i+1)} \right) + d^{(k)} \quad (8)$$

Since $d^{(\ell)} \geq 1$ for $0 \leq \ell \leq k + 1$, so we have $1 \leq k \leq 18$.

Claim: $k = 18$ minimizes N_w .

Proof of Claim:

If $k = 18$, we have 2 units in each hidden layer (one is $x_0^{(\ell)}$, not counted in $d^{(\ell)}$), so

$$N_w|_{k=18} = (37 - 18) + 9 \times 1 + \left(\sum_{i=1}^{17} 1 \times 1 \right) + 1 = 46 \quad (9)$$

If $k = 18 - m$, $m \in \mathbb{N}$ and $1 \leq m \leq 17$, we have

$$N_w|_{k=18-m} = (19 + m) + 9d'^{(1)} + \left(\sum_{i=1}^{17-m} d'^{(i)} d'^{(i+1)} \right) + d'^{(18-m)} \quad (10)$$

$$\geq (19 + m) + 9 + (17 - m) + 1 \quad (11)$$

$$= 19 + 9 + 17 + 1 = N_w|_{k=18} \quad (12)$$

where $d'^{(\ell)}$ is the new number of each hidden layer if $k = 18 - m$ and (11) holds due to $d'^{(i)} d'^{(i+1)} \geq 1$ and $d'^{(i)} \geq 1, \forall i$ (by definition).

Hence, we have $N_w \geq 46$.

□

Problem 3

Following the setting of Problem 2.

Claim: $k = 2$ with 21 units (not included $x_0^{(1)}$) in $d^{(1)}$ and 13 units (not included $x_0^{(2)}$) in $d^{(2)}$ maximizes N_w .

Proof of Claim:

If $k = 2$ with 21 units (not included $x_0^{(1)}$) in $d^{(1)}$ and 13 units (not included $x_0^{(2)}$) in $d^{(2)}$, we have

$$N_w|_{k=2} = (37 - 2) + 9 \times 21 + (21 \times 13) + 13 = 510 \quad (13)$$

Consider following cases,

1. If $k = 2$ with $34 - m$ units (not included $x_0^{(1)}$) in $d^{(1)}$ and m units (not included $x_0^{(2)}$) in $d^{(2)}$, where $m \in \mathbb{N}$ and $1 \leq m \leq 33$, we have

$$N_w|_{k=2} = (37 - 2) + 9 \times (34 - m) + ((34 - m) \times m) + m = -(m - 13)^2 + 510 \quad (14)$$

Hence, $m = 13$ maximize $N_w|_{k=2}$.

2. If $k = 1$.

$$N_w|_{k=1} = (37 - 1) + 9 \times 35 + 35 = 386 < 510 \quad (15)$$

3. If $k = 3$ with $33 - n_1 - n_2$ units (not included $x_0^{(1)}$) in $d^{(1)}$, n_1 units (not included $x_0^{(2)}$) in $d^{(2)}$ and n_2 units (not included $x_0^{(3)}$) in $d^{(3)}$, where $n_1, n_2 \in \mathbb{N}$ and $1 \leq n_1, n_2 \leq 32$, we have

$$N_w|_{k=3} = (37 - 3) + 9 \times (33 - n_1 - n_2) + ((33 - n_1 - n_2) \times n_1 + n_1 \times n_2) + n_2 \quad (16)$$

$$= -(n_1 - 12)^2 - 8n_2 + 475 \leq -8n_2 + 475 \leq 467 < 510 \quad (17)$$

We can see that if we have no $d^{(3)}$ layer (which means $n_2 = 0$), then $N_w|_{k=3}$ can be larger.

4. If $18 \geq k = s \geq 4$ with with $(36 - s) - \sum_{i=1}^{s-1} n_i$ units (not included $x_0^{(1)}$) in $d^{(1)}$, n_1 units (not included $x_0^{(2)}$) in $d^{(2)}$, n_2 units (not included $x_0^{(3)}$) in $d^{(3)}$, ..., n_{s-1} units (not included $x_0^{(s)}$) in $d^{(s)}$ where $n_i \in \mathbb{N}$ and $1 \leq n_i \leq (35 - s)$, $\forall i$, we have

$$\begin{aligned} N_w|_{k=s} &= (37 - s) + 9 \times \left((36 - s) - \sum_{i=1}^{s-1} n_i \right) \\ &\quad + \left(\left((36 - s) - \sum_{i=1}^{s-1} n_i \right) \times n_1 + \cdots + n_{s-2} \times n_{s-1} \right) + n_{s-1} \end{aligned} \quad (18)$$

We can find that there is no $n_1 n_2$ term, only $n_2 n_3$ term exists and no other terms contains n_2 . We have

$$\frac{\partial N_w|_{k=s}}{\partial n_2} = n_3 = 0 \text{ as } N_w|_{k=s} \text{ reaches maximum} \quad (19)$$

This implies $N_w|_{k=s}$ can be larger without $d^{(4)}$ layer. So $k = s \geq 4$ cannot maximize N_w .

Hence, we have $N_w \geq 510$.

□

Problem 4

$$\nabla_{\mathbf{w}} \text{err}_n(\mathbf{w}) = \frac{\partial}{\partial \mathbf{w}} \|\mathbf{x}_n - \mathbf{w} \mathbf{w}^T \mathbf{x}_n\|^2 = \frac{\partial}{\partial \mathbf{w}} (\mathbf{x}_n - \mathbf{w} \mathbf{w}^T \mathbf{x}_n)^T (\mathbf{x}_n - \mathbf{w} \mathbf{w}^T \mathbf{x}_n) \quad (20)$$

$$= \frac{\partial}{\partial \mathbf{w}} (\mathbf{x}_n^T - \mathbf{x}_n^T \mathbf{w}^T \mathbf{w}) (\mathbf{x}_n - \mathbf{w} \mathbf{w}^T \mathbf{x}_n) \quad (21)$$

$$= (-2 \mathbf{x}_n^T \mathbf{w}) (\mathbf{x}_n - \mathbf{w} \mathbf{w}^T \mathbf{x}_n) + (\mathbf{x}_n^T - \mathbf{x}_n^T \mathbf{w}^T \mathbf{w}) (-2 \mathbf{w} \mathbf{x}_n) \quad (22)$$

$$= -2 (\mathbf{x}_n^T \mathbf{w}) \mathbf{x}_n + 2 (\mathbf{x}_n^T \mathbf{w}) \mathbf{w} (\mathbf{x}_n^T \mathbf{w})^T - 2 (\mathbf{x}_n^T \mathbf{w}) \mathbf{x}_n + 2 \mathbf{x}_n^T (\mathbf{w}^T \mathbf{w}) \mathbf{w} \mathbf{x}_n \quad (23)$$

$$= -4 (\mathbf{x}_n^T \mathbf{w}) \mathbf{x}_n + 2 (\mathbf{x}_n^T \mathbf{w})^2 \mathbf{w} + 2 \mathbf{x}_n^T (\mathbf{w}^T \mathbf{w}) \mathbf{w} \mathbf{x}_n \quad (24)$$

where we have used

$$(\mathbf{x}_n^T \mathbf{w}) \mathbf{w} (\mathbf{x}_n^T \mathbf{w})^T = \begin{pmatrix} x_1 & x_2 & \cdots & x_n \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix}^T (\mathbf{x}_n^T \mathbf{w})^T \quad (25)$$

$$= c \begin{pmatrix} w_1 \\ w_2 \\ \vdots \\ w_n \end{pmatrix} c = \begin{pmatrix} c^2 w_1 \\ c^2 w_2 \\ \vdots \\ c^2 w_n \end{pmatrix} = (\mathbf{x}_n^T \mathbf{w})^2 \mathbf{w} \quad (26)$$

with $c := \sum_{i=1}^n x_i w_i$.

□

Problem 5

$$E_{\text{in}}(\mathbf{w}) = \frac{1}{N} \sum_{n=1}^N (\mathbf{x}_n - \mathbf{w}\mathbf{w}^T(\mathbf{x}_n + \boldsymbol{\epsilon}_n))^T (\mathbf{x}_n - \mathbf{w}\mathbf{w}^T(\mathbf{x}_n + \boldsymbol{\epsilon}_n)) \quad (27)$$

$$= \frac{1}{N} \sum_{n=1}^N \left(\mathbf{x}_n^T - (\mathbf{x}_n + \boldsymbol{\epsilon}_n)^T \mathbf{w}^T \mathbf{w} \right) (\mathbf{x}_n - \mathbf{w}\mathbf{w}^T(\mathbf{x}_n + \boldsymbol{\epsilon}_n)) \quad (28)$$

$$= \frac{1}{N} \sum_{n=1}^N \left\| \mathbf{x}_n - \mathbf{w}\mathbf{w}^T \mathbf{x}_n \right\|^2 - \boldsymbol{\epsilon}_n^T \mathbf{w}^T \mathbf{w} (\mathbf{x}_n - \mathbf{w}\mathbf{w}^T \mathbf{x}_n) - \mathbf{w}\mathbf{w}^T \boldsymbol{\epsilon}_n (\mathbf{x}_n^T - \mathbf{x}_n^T \mathbf{w}^T \mathbf{w}) + (\boldsymbol{\epsilon}_n)^2 (\mathbf{w}^T \mathbf{w})^2 \quad (29)$$

Since $\boldsymbol{\epsilon}_n$ is generated from a zero-mean, unit variance Gaussian distribution, so $\mathcal{E}(\boldsymbol{\epsilon}) = 0$ and $\mathcal{E}(\|\boldsymbol{\epsilon}\|^2) = 1$. Hence

$$\mathcal{E}(E_{\text{in}}(\mathbf{w})) = \frac{1}{N} \sum_{n=1}^N \left\| \mathbf{x}_n - \mathbf{w}\mathbf{w}^T \mathbf{x}_n \right\|^2 + (\mathbf{w}^T \mathbf{w})^2 \quad (30)$$

So $\Omega(\mathbf{w}) = (\mathbf{w}^T \mathbf{w})^2$.

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Problem 6

Claim: $\mathbf{w} = 2(\mathbf{x}_+ - \mathbf{x}_-)$, $b = -\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2$.

Proof of Claim:

Consider the following cases.

1. If $\mathbf{x} = \mathbf{x}_+$, $\mathbf{w}^T \mathbf{x} + b > 0$.

$$\mathbf{w}^T \mathbf{x} + b = 2(\mathbf{x}_+^T - \mathbf{x}_-^T) \mathbf{x}_+ + (-\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2) \quad (31)$$

$$= \|\mathbf{x}_+\|^2 - 2\mathbf{x}_-^T \mathbf{x}_+ + \|\mathbf{x}_-\|^2 = \|\mathbf{x}_+\|^2 - \mathbf{x}_-^T \mathbf{x}_+ - \mathbf{x}_+^T \mathbf{x}_- + \|\mathbf{x}_-\|^2 \quad (32)$$

$$= \|\mathbf{x}_+ - \mathbf{x}_-\|^2 > 0 \quad (33)$$

2. If $\mathbf{x} = \mathbf{x}_-$, $\mathbf{w}^T \mathbf{x} + b < 0$.

$$\mathbf{w}^T \mathbf{x} + b = 2(\mathbf{x}_+^T - \mathbf{x}_-^T) \mathbf{x}_- + (-\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2) \quad (34)$$

$$= -\|\mathbf{x}_+\|^2 + 2\mathbf{x}_-^T \mathbf{x}_+ - \|\mathbf{x}_-\|^2 = -\|\mathbf{x}_+\|^2 + \mathbf{x}_-^T \mathbf{x}_+ + \mathbf{x}_+^T \mathbf{x}_- - \|\mathbf{x}_-\|^2 \quad (35)$$

$$= -\|\mathbf{x}_+ - \mathbf{x}_-\|^2 < 0 \quad (36)$$

3. If $\mathbf{x} = (\mathbf{x}_+ + \mathbf{x}_-)/2$, $\mathbf{w}^T \mathbf{x} + b = 0$.

$$\mathbf{w}^T \mathbf{x} + b = 2 (\mathbf{x}_+^T - \mathbf{x}_-^T) (\mathbf{x}_+ + \mathbf{x}_-)/2 + (-\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2) \quad (37)$$

$$= (\|\mathbf{x}_+\|^2 - \|\mathbf{x}_-\|^2) + (-\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2) = 0 \quad (38)$$

4. If $\mathbf{x} = (\mathbf{x}_+ + \mathbf{x}_-)/2 + \mathbf{x}'$, $\mathbf{w}^T \mathbf{x} + b > 0$ if $\|\mathbf{x}_+ - \mathbf{x}'\|^2 < \|\mathbf{x}_- - \mathbf{x}'\|^2$; $\mathbf{w}^T \mathbf{x} + b < 0$ if $\|\mathbf{x}_+ - \mathbf{x}'\|^2 > \|\mathbf{x}_- - \mathbf{x}'\|^2$.

$$\mathbf{w}^T \mathbf{x} + b = 2 (\mathbf{x}_+^T - \mathbf{x}_-^T) ((\mathbf{x}_+ + \mathbf{x}_-)/2 + \mathbf{x}') + (-\|\mathbf{x}_+\|^2 + \|\mathbf{x}_-\|^2) \quad (39)$$

$$= 2 (\mathbf{x}_+^T - \mathbf{x}_-^T) \mathbf{x}' \quad (40)$$

If $\|\mathbf{x}_+ - \mathbf{x}'\|^2 < \|\mathbf{x}_- - \mathbf{x}'\|^2$, then $\mathbf{x}_+^T \mathbf{x}' > 0$ and $\mathbf{x}_-^T \mathbf{x}' < 0 \Rightarrow 2 (\mathbf{x}_+^T - \mathbf{x}_-^T) \mathbf{x}' > 0$; if $\|\mathbf{x}_+ - \mathbf{x}'\|^2 > \|\mathbf{x}_- - \mathbf{x}'\|^2$, then $\mathbf{x}_+^T \mathbf{x}' < 0$ and $\mathbf{x}_-^T \mathbf{x}' > 0 \Rightarrow 2 (\mathbf{x}_+^T - \mathbf{x}_-^T) \mathbf{x}' < 0$.

Hence, we have proved the claim.

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Problem 7

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Problem 8

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Problem 9

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Problem 10

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Problem 11

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Problem 12

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Problem 13

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Problem 14

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Problem 15

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Problem 16

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Problem 17

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Problem 18

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Problem 19

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Problem 20



Reference

- [1] Lecture Notes by Hsuan-Tien LIN, Department of Computer Science and Information Engineering, National Taiwan University, Taipei 106, Taiwan.