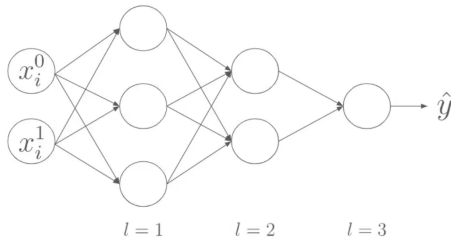


Homework 1

Problem 1

Question 1



$$\text{setup: } x = \begin{pmatrix} 2 \\ 1 \end{pmatrix} \quad y = 10 \quad w^{(1)} = \begin{pmatrix} 1 & 0 \\ 3 & 4 \\ 1 & 2 \end{pmatrix} \quad b^{(1)} = \begin{pmatrix} 0 \\ -2 \\ 5 \end{pmatrix} \quad w^{(2)} = \begin{pmatrix} 2 & 6 & 1 \\ 0 & 3 & 1 \end{pmatrix} \quad b^{(2)} = \begin{pmatrix} -23 \\ -11 \end{pmatrix}$$

$$w^{(3)} = [-2 \ 3] \quad b^{(3)} = 18$$

$$\text{Analysis: } ① z^{(0)} = x = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$② a^{(1)} = w^{(1)} z^{(0)} + b^{(1)} = \begin{pmatrix} 1 & 0 \\ 3 & 4 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} + \begin{pmatrix} 0 \\ -2 \\ 5 \end{pmatrix} = \begin{pmatrix} 2 \\ 10 \\ 4 \end{pmatrix} + \begin{pmatrix} 0 \\ -2 \\ 5 \end{pmatrix} = \begin{pmatrix} 2 \\ 8 \\ 9 \end{pmatrix}$$

$$③ z^{(1)} = \text{ReLU}(a^{(1)}) = \begin{pmatrix} 2 \\ 8 \\ 9 \end{pmatrix}$$

$$④ a^{(2)} = w^{(2)} z^{(1)} + b^{(2)} = \begin{pmatrix} 2 & 6 & 1 \\ 0 & 3 & 1 \end{pmatrix} \begin{pmatrix} 2 \\ 8 \\ 9 \end{pmatrix} + \begin{pmatrix} -23 \\ -11 \end{pmatrix} = \begin{pmatrix} 61 \\ 33 \end{pmatrix} + \begin{pmatrix} -23 \\ -11 \end{pmatrix} = \begin{pmatrix} 38 \\ 22 \end{pmatrix}$$

$$⑤ z^{(2)} = \text{ReLU}(a^{(2)}) = \begin{pmatrix} 38 \\ 22 \end{pmatrix}$$

$$⑥ a^{(3)} = w^{(3)} z^{(2)} + b^{(3)} = [-2 \ 3] \begin{pmatrix} 38 \\ 22 \end{pmatrix} + 18 = 8$$

$$⑦ z^{(3)} = \text{ReLU}(a^{(3)}) = 8$$

$$⑧ \hat{y} = z^{(3)} = 8$$

Question 2 $L(\hat{y}, y) = (y - \hat{y})^2$ $\text{ReLU}(x) = \max(0, x) \Rightarrow \text{ReLU}'(x) = \begin{cases} 0, & x \leq 0 \\ 1, & x > 0 \end{cases}$

$$① \frac{\partial L}{\partial \hat{y}} = -2(y - \hat{y}) = 2(\hat{y} - y)$$

$$② \frac{\partial L}{\partial z^{(3)}} = \frac{\partial L}{\partial \hat{y}} \cdot \frac{\partial \hat{y}}{\partial z^{(3)}} = 2(\hat{y} - y)$$

$$③ \frac{\partial L}{\partial a^{(3)}} = \frac{\partial L}{\partial z^{(3)}} \cdot \frac{\partial z^{(3)}}{\partial a^{(3)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)})$$

$$④ \frac{\partial L}{\partial w^{(3)}} = \frac{\partial L}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial w^{(3)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot z^{(2)}$$

$$⑤ \frac{\partial L}{\partial b^{(3)}} = \frac{\partial L}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial b^{(3)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)})$$

$$⑥ \frac{\partial L}{\partial z^{(2)}} = \frac{\partial L}{\partial a^{(3)}} \cdot \frac{\partial a^{(3)}}{\partial z^{(2)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)}$$

$$⑦ \frac{\partial L}{\partial a^{(2)}} = \frac{\partial L}{\partial z^{(2)}} \cdot \frac{\partial z^{(2)}}{\partial a^{(2)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)})$$

$$⑧ \frac{\partial L}{\partial w^{(2)}} = \frac{\partial L}{\partial a^{(2)}} \cdot \frac{\partial a^{(2)}}{\partial w^{(2)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)}) \cdot z^{(1)}$$

$$⑨ \frac{\partial L}{\partial b^{(2)}} = \frac{\partial L}{\partial a^{(2)}} \cdot \frac{\partial a^{(2)}}{\partial b^{(2)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)})$$

$$⑩ \frac{\partial L}{\partial z^{(1)}} = \frac{\partial L}{\partial a^{(2)}} \cdot \frac{\partial a^{(2)}}{\partial z^{(1)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)}) \cdot w^{(2)}$$

$$⑪ \frac{\partial L}{\partial a^{(1)}} = \frac{\partial L}{\partial z^{(1)}} \cdot \frac{\partial z^{(1)}}{\partial a^{(1)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)}) \cdot w^{(2)} \cdot \text{ReLU}'(a^{(1)})$$

$$⑫ \frac{\partial L}{\partial w^{(1)}} = \frac{\partial L}{\partial a^{(1)}} \cdot \frac{\partial a^{(1)}}{\partial w^{(1)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)}) \cdot w^{(2)} \cdot \text{ReLU}'(a^{(1)}) \cdot z^{(0)}$$

$$⑬ \frac{\partial L}{\partial b^{(1)}} = \frac{\partial L}{\partial a^{(1)}} \cdot \frac{\partial a^{(1)}}{\partial b^{(1)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot w^{(3)} \cdot \text{ReLU}'(a^{(2)}) \cdot w^{(2)} \cdot \text{ReLU}'(a^{(1)})$$

$$\textcircled{14} \frac{\partial L}{\partial z^{(0)}} = \frac{\partial L}{\partial a^{(1)}} \frac{\partial a^{(1)}}{\partial z^{(0)}} = 2(\hat{y} - y) \text{ReLU}'(a^{(1)}) w^{(1)} \text{ReLU}'(a^{(0)}) w^{(0)} \text{ReLU}'(a^{(1)}) w^{(1)}$$

$$\textcircled{15} \frac{\partial L}{\partial x} = \frac{\partial L}{\partial z^{(0)}} \frac{\partial z^{(0)}}{\partial x} = 2(\hat{y} - y) \text{ReLU}'(a^{(1)}) w^{(1)} \text{ReLU}'(a^{(2)}) w^{(2)} \text{ReLU}'(a^{(0)}) w^{(0)}$$

Question 3

$$\textcircled{1} \ell=3: \frac{\partial \ell}{\partial w^{(3)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) \cdot z^{(3)} = 2 \times (8-10) \times 1 \times \begin{pmatrix} 38 \\ 22 \end{pmatrix} = -4 \times \begin{pmatrix} 38 \\ 22 \end{pmatrix} = \begin{pmatrix} -152 \\ -88 \end{pmatrix}$$

$$\frac{\partial \ell}{\partial b^{(3)}} = 2(\hat{y} - y) \cdot \text{ReLU}'(a^{(3)}) = 2 \times (8-10) \times 1 = -4$$

$$\begin{aligned} \textcircled{2} \ell=2: \frac{\partial \ell}{\partial w^{(2)}} &= 2(\hat{y} - y) \text{ReLU}'(a^{(3)}) w^{(3)} \text{ReLU}'(a^{(2)}) \cdot z^{(2)} = 2 \times (8-10) \times 1 \times (-2 \ 3) \times 1 \times \begin{pmatrix} 2 \\ 8 \\ 9 \end{pmatrix}^T \\ &= -4 \times (-2 \ 3)^T (2 \ 8 \ 9) = \begin{pmatrix} 8 \\ -12 \end{pmatrix} (2 \ 8 \ 9) = \begin{pmatrix} 16 & 64 & 72 \\ -24 & -46 & -108 \end{pmatrix} \end{aligned}$$

$$\frac{\partial \ell}{\partial b^{(2)}} = 2(\hat{y} - y) \text{ReLU}'(a^{(3)}) w^{(3)} \text{ReLU}'(a^{(2)}) = 2 \times (8-10) \times 1 \times (-2 \ 3)^T \times 1 = \begin{pmatrix} 8 \\ -12 \end{pmatrix}$$

$$\textcircled{3} \ell=1: \frac{\partial \ell}{\partial w^{(1)}} = 2(\hat{y} - y) \text{ReLU}'(a^{(3)}) w^{(3)} \text{ReLU}'(a^{(2)}) w^{(2)} \text{ReLU}'(a^{(1)}) \cdot z^{(1)}$$

$$= (2 \times (8-10) \times 1 \times (-2 \ 3) \times 1 \times \begin{pmatrix} 2 & 6 & 1 \\ 0 & 3 & 1 \end{pmatrix})^T \times 1 \times \begin{pmatrix} 2 \\ 1 \end{pmatrix}^T$$

$$= (-4 \times (-2 \ 3) \times \begin{pmatrix} 2 & 6 & 1 \\ 0 & 3 & 1 \end{pmatrix})^T \times \begin{pmatrix} 2 \\ 1 \end{pmatrix}^T$$

$$= \begin{pmatrix} 8 & -12 \end{pmatrix} \times \begin{pmatrix} 2 & 6 & 1 \\ 0 & 3 & 1 \end{pmatrix}^T \times \begin{pmatrix} 2 \\ 1 \end{pmatrix}^T = \begin{pmatrix} 16 \\ 12 \\ -4 \end{pmatrix} (2 \ 1) = \begin{pmatrix} 32 & 16 \\ 24 & 12 \\ -8 & -4 \end{pmatrix}$$

$$\begin{aligned} \frac{\partial \ell}{\partial b^{(1)}} &= 2(\hat{y} - y) \text{ReLU}'(a^{(3)}) w^{(3)} \text{ReLU}'(a^{(2)}) w^{(2)} \text{ReLU}'(a^{(1)}) \\ &= \begin{pmatrix} 16 \\ 12 \\ -4 \end{pmatrix} \end{aligned}$$

Problem 2

Question 1: number of d : 10, number of λ_2 : 6, number of p : 5

Hence, the total number of combination: $10 \times 6 \times 5 = 300$

10 epochs for one combination, then the total epochs are $300 \times 10 = 3000$

And the total time = $3000 \text{ epoch} \times 2 \text{ min} = 6000 \text{ min}$

conclusion: 3000 epochs, 6000 min

Question 2: $P(\text{the combination results in good performance}) = 0.05$

$P(\text{the combination results in bad performance}) = 1 - 0.05 = 0.95$

$P(\text{at least one combination has good results in } N \text{ trials}) = 1 - 0.95^N$

$$\text{let } 1 - 0.95^N \geq q \Rightarrow 0.95^N \leq 1 - q \Rightarrow N \cdot \ln 0.95 \leq \ln(1 - q) \Rightarrow N \geq \frac{\ln(1 - q)}{\ln(0.95)}$$

$$\textcircled{1} \text{ let } q = 0.95: N \geq \frac{\ln 0.05}{\ln 0.95} \approx 58.4 \Rightarrow 59 \text{ trials}$$

$$\textcircled{2} \text{ let } q = 0.995: N \geq \frac{\ln 0.005}{\ln 0.95} \approx 103.29 \Rightarrow 104 \text{ trials}$$

Problem 3

Question 1: SGD $W_{t+1} = W_t - \alpha \nabla \ell(W_t, x_t)$

- ① reduce the computational cost
- ② faster than GD
- ③ can help the optimizer escape shallow local minima or saddle point

Question 2: Minibatch GD $W_{t+1} = W_t - \alpha \frac{1}{b} \sum_{i=1}^b \nabla \ell(W_t, x_i)$

- ① compared with SGD, it has lower variance
- ② reduce the computational cost than GD

Question 3: SGD with Momentum $\begin{cases} m_{t+1} = m_t - \alpha \nabla \ell(W_t, x_t) \\ W_{t+1} = W_t + m_{t+1} \end{cases}$

- ① reduces oscillations and accelerate convergence, especially in problems where gradients vary significantly across dimensions
- ② enable faster convergence and more stable

Problem 4

Question 1: (i) setup: (i) input: 36×36 (ii) filter size: 3×3 (iii) padding: 0 (iv) stride: 1
(ii) analysis: if the filter starts at column c , then it covers: $c, c+1, c+2$
the last covered column should be: $c+2 \leq 36 \Rightarrow c \leq 34$
Hence, the starting columns could be 1 through 34, and same logic for row
Then, the output size should be 34×34

Question 2: (i) setup: (i) input: $36 \times 36 \times 3$ (ii) filter: $3 \times 3 \times 3$ (iii) padding: 3 (iv) stride: 2 (v) filter size: 2
(ii) analysis: for each channel, it will be calculated by one filter.
(i) If padding size = 3, then the input size will be $42 \times 42 \times 3$ ($36+3+3$)
same logic as Question 1, suppose c is the starting column, then the filter covers: $c, c+1, c+2$.
Then, $c+2 \leq 42 \Rightarrow c \leq 40$
★ (ii) If stride = 2, then the starting position should be 1, 3, 5, ..., 39
From 1 to 39, there will be $1+3+\dots+39 = \frac{39+1}{2} \times 39 = 800$ steps
So there are 20 horizontal positions
(iii) There are 2 filters, then each produces one feature map
so output has 2 channels
Conclusion: $20 \times 20 \times 2$ (height \times width \times channels)

Question 3: ① setup: i) input: $n \times n \times c$ ii) filter size: $k \times k \times c$

iii) number of filters: ℓ iv) stride: S v) padding: P

② analysis:

i) from padding $= P$, we could know the input should be $(n+2P) \times (n+2P) \times c$

ii) if we want to get the valid starting position C , then $(C, C+1, \dots, C+k-1)$

$$C+k-1 \leq n+2P \Rightarrow C \leq n+2P-k+1 \text{ (starting row/column)}$$

iii) from stride $= S$, we could analyze the starting position for row/column
it should be: $1, 1+S, 1+2S, \dots, 1+mS \leq n+2P-k+1$

then $m \leq \frac{1}{S}(n+2P-k)$, so the max step should be: $\lfloor \frac{1}{S}(n+2P-k) \rfloor$

the number of starting points = number of steps + 1 = $\lfloor \frac{1}{S}(n+2P-k) \rfloor + 1$

iv) there are ℓ filters, then there are ℓ feature maps, and the output should have ℓ channels

conclusion: $n_{out} \times n_{out} \times \ell$ and $n_{out} = \lfloor \frac{1}{S}(n+2P-k) \rfloor + 1$

Question 4 @ setup: (i) input: $I = \begin{pmatrix} 6 & 0 & 1 \\ 3 & 4 & 1 \\ 1 & 7 & 0 \end{pmatrix}$

(ii) convolutional kernel: $K = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$

(iii) stride: 1

(iv) padding: 0

@ analysis: output =
$$\begin{pmatrix} 6 \times 1 + 0 \times 0 + 3 \times 0 + 4 \times (-1) & 0 \times 1 + 1 \times 0 + 4 \times 0 + 1 \times (-1) \\ 3 \times 1 + 4 \times 0 + 1 \times 0 + 7 \times (-1) & 4 \times 1 + 1 \times 0 + 7 \times 0 + 0 \times (-1) \end{pmatrix}$$
$$= \begin{pmatrix} 2 & -1 \\ -4 & 4 \end{pmatrix}$$

Problem 5

Question 1

$$\frac{\partial L}{\partial z_i} = \frac{\partial L}{\partial z_m} \frac{\partial z_m}{\partial z_{m-1}} \dots \frac{\partial z_{i+1}}{\partial z_i}$$

if $\frac{\partial z_{i+1}}{\partial z_i} \ll 1$, it will make the dot of gradients close to 0, then $\frac{\partial L}{\partial z_i} \rightarrow 0$, which will make model cannot learn weights effectively

Question 2

$$\text{setup: } \frac{\partial L}{\partial z_2} = 0.01, \frac{\partial z_2}{\partial z_1} = 0.005, \frac{\partial z_1}{\partial x} = 0.1$$

$$\text{analysis: } \frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_2} \times \frac{\partial z_2}{\partial z_1} \times \frac{\partial z_1}{\partial x} = 0.01 \times 0.005 \times 0.1 = 0.000005$$

Question 3

$$\text{setup: } \frac{\partial L}{\partial y} = 0.01 \quad y = z_2 + x \Rightarrow \frac{\partial y}{\partial x} = \frac{\partial z_2}{\partial x} + 1$$

$$\frac{\partial L}{\partial x} = \frac{\partial L}{\partial y} \frac{\partial y}{\partial x} = \frac{\partial L}{\partial y} \left(\frac{\partial z_2}{\partial x} + 1 \right) = \frac{\partial L}{\partial y} \left(\frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} + 1 \right)$$

$$= 0.01 \times (0.005 \times 0.1 + 1) = 0.010005$$

Question 4

It is obvious that the gradient with residual connection is larger than that without it ($0.010005 > 0.000005$)

Question 5

1 layer: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5 \times 0.5 = 0.25$

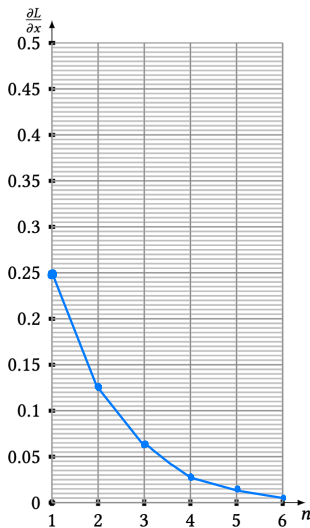
2 layers: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_2} \frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5^2 \times 0.5 = 0.125$

3 layers: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_3} \frac{\partial z_3}{\partial z_2} \frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5^4 = 0.0625$

4 layers: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_4} \frac{\partial z_4}{\partial z_3} \frac{\partial z_3}{\partial z_2} \frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5^5 = 0.03125$

5 layers: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_5} \frac{\partial z_5}{\partial z_4} \frac{\partial z_4}{\partial z_3} \frac{\partial z_3}{\partial z_2} \frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5^6 = 0.015625$

6 layers: $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial z_6} \frac{\partial z_6}{\partial z_5} \frac{\partial z_5}{\partial z_4} \frac{\partial z_4}{\partial z_3} \frac{\partial z_3}{\partial z_2} \frac{\partial z_2}{\partial z_1} \frac{\partial z_1}{\partial x} = 0.5^7 = 0.0078125$



Question 6

① gradient of a single residual block:

analysis: n conv layers and each residual block has 2 conv, then the number of residual blocks is $n/2$

output of j residual block: y_{j+1}

outputs of two conv: z_{2j-1} z_{2j}

output of block: $y_j = z_{2j} + y_{j-1}$

$$\frac{\partial y_j}{\partial y_{j-1}} = \frac{\partial z_{2j}}{\partial y_{j-1}} + 1 \Rightarrow \frac{\partial z_{2j}}{\partial y_{j-1}} = \frac{\partial z_{2j}}{\partial z_{2j-1}} \frac{\partial z_{2j-1}}{\partial y_{j-1}} = 0.5^2 = 0.25$$

$$\Rightarrow \frac{\partial y_j}{\partial y_{j-1}} = 0.25 + 1 = 1.25$$

② general: let $g_j = \frac{\partial L}{\partial y_j}$ and $g_B = \frac{\partial L}{\partial y_B} = 0.5$

$$g_{j-1} = \frac{\partial L}{\partial y_{j-1}} = \frac{\partial L}{\partial y_j} \frac{\partial y_j}{\partial y_{j-1}} = g_j \cdot 1.25 \Rightarrow g_0 = (1.25)^B g_B = 0.5 (1.25)^B$$

$$n=2 \Rightarrow B=1 \Rightarrow \frac{\partial L}{\partial x} = 0.5 \times 1.25 = 0.625$$

$$n=4 \Rightarrow B=2 \Rightarrow \frac{\partial L}{\partial x} = 0.5 \times 1.25^2 = 0.78125$$

$$n=6 \Rightarrow B=3 \Rightarrow \frac{\partial L}{\partial x} = 0.5 \times 1.25^3 = 0.9765625$$

