

1.1.1

$$\nabla_{w_t} = 2(x_i \hat{w} - t) x_i$$

$$w_{t+1} \leftarrow w_t - \eta \nabla_{w_t} L_i(x_i, w_t)$$

$$w_{t+1} \leftarrow w_t - 2\eta (x_i \hat{w} - t) x_i$$

$$\Leftrightarrow w_{t+1} \leftarrow w_t + a x_i \quad a \in \mathbb{R}$$

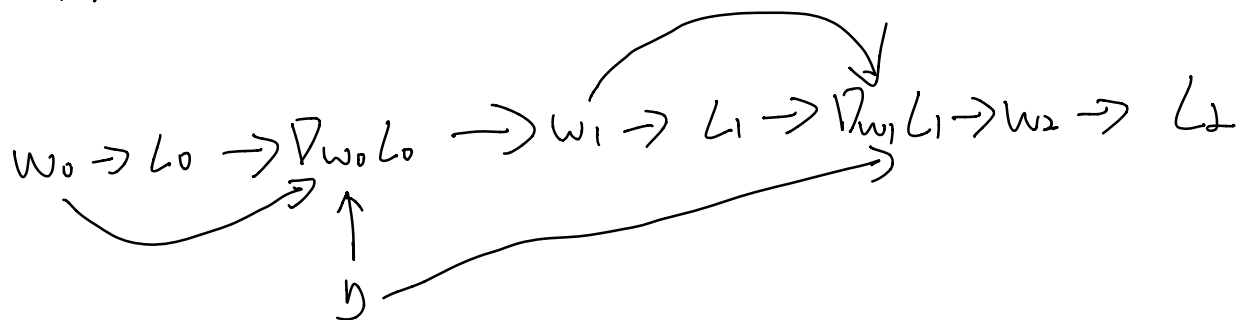
if $w_0 = 0$, then SGD $\hat{w} \in \text{span}\{x_1, x_2, \dots, x_n\}$

From HW1 Q3.4, we know GD $w^* \in \text{span}\{x_1, x_2, \dots, x_n\}$

If both w^* , \hat{w} converge and share same objective function, then $\hat{w} = w^*$.

2.1 Computation Graph of Learning Rates.

2.1.1



2.1.2

forward: $O(1)$

backward: $O(t)$

2.1.3

The memory cost linearly increases when taking many iterations for using back propagation.

2.2 Learning Learning Rates

2.2.1

$$\frac{\partial L}{\partial w_0} = \frac{2}{n} X^T (X w_0 - t)$$

$$\begin{aligned} w_1 &= w_0 - \frac{2\eta}{n} X^T (X w_0 - t) \\ &= w_0 - \frac{2\eta}{n} X^T a \end{aligned}$$

$$L_1 = \frac{1}{n} \|X w_1 - t\|_2^2$$

$$= \frac{1}{n} \|X (w_0 - \frac{2\eta}{n} X^T a) - t\|_2^2$$

2.2.2

$$\frac{dL_1}{dw_1} = \frac{2}{n} X^T (X w_1 - t) \quad \frac{dw_1}{d\eta} = -\frac{2}{n} X^T a$$

$$\frac{dL_1}{d\eta} = \frac{dL_1}{dw_1} \cdot \frac{dw_1}{d\eta}$$

$$= \frac{2}{n} \left(-\frac{2}{n} X^T a \right)^T X^T (X w_1 - t)$$

$$= -\frac{4}{n} (x^T a)^T x^T (x w_1 - t)$$

$$\frac{d^2 \mathcal{L}_1}{dy^2} = \frac{d}{dy} \left(\frac{d\mathcal{L}_1}{dy} \right)$$

$$= -\frac{4}{n} (x^T a)^T x^T x \left(-\frac{2}{n} x^T a \right)$$

$$= \frac{8}{n^2} (x^T a)^T x^T x x^T a$$

$$= \frac{8}{n^2} a^T x x^T x x^T a$$

$$= \frac{8}{n^2} \|x x^T a\|_2^2 > 0$$

$\Rightarrow \mathcal{L}_1$ is convex

2.2.3

$$\frac{d\mathcal{L}_1}{dy} = -\frac{4}{n} (x^T a)^T x^T (x w_1 - t) = 0$$

$$-\frac{4}{n} (x^T a)^T x^T x w_1 = -\frac{4}{n} (x^T a)^T x^T t$$

$$a^T x x^T x (w_0 - \frac{2y}{n} x^T a) = a^T x x^T t$$

$$a^T x x^T x w_0 - \frac{2y}{n} a^T x x^T x x^T a = a^T x x^T t$$

$$\Rightarrow y = \frac{n}{2} \frac{a^T x x^T x w_0 - a^T x x^T t}{\|x x^T a\|_2^2}$$

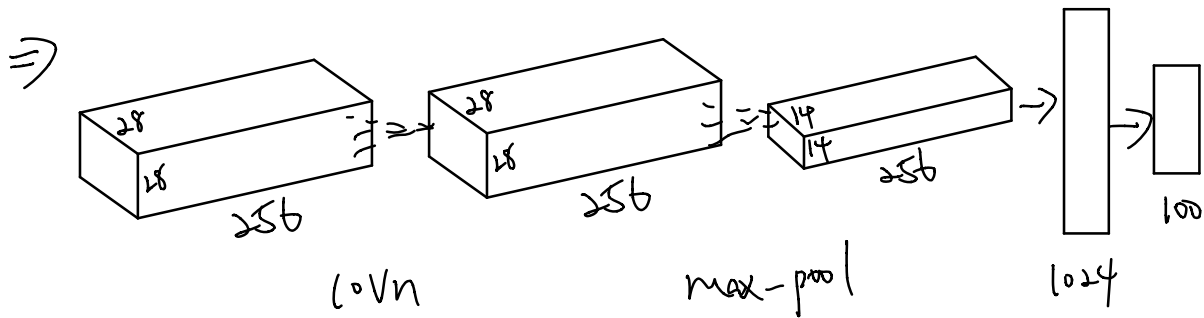
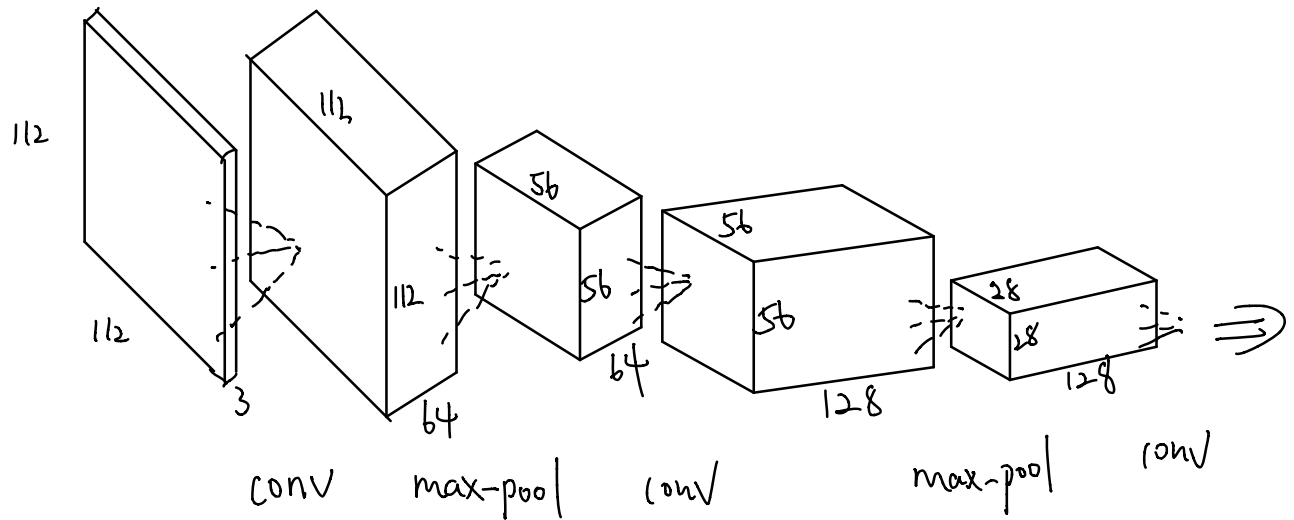
3. CNN

$$3.1 \quad I = \begin{bmatrix} 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 2 \end{bmatrix} \quad J = \begin{pmatrix} 0 & -1 & 0 \\ -1 & 4 & -1 \\ 0 & -1 & 0 \end{pmatrix}$$

$$I * J = \begin{bmatrix} -1 & 2 & 2 & -2 & 0 \\ -2 & 1 & 0 & 2 & -1 \\ 3 & 0 & 0 & 1 & -1 \\ -2 & 2 & 0 & 2 & -1 \\ 0 & -2 & 3 & -2 & 0 \end{bmatrix}$$

feature: edge detect

3-2 $k=3$



	# of parameter
C1	$3 \times 3 \times 3 \times 64 + 64 = 1728 + 64 = 1792$
C2	$3 \times 3 \times 64 \times 128 + 128 = 73728 + 128 = 73856$
C3	$3 \times 3 \times 128 \times 256 + 256 = 294912 + 256 = 295168$
C4	$3 \times 3 \times 256 \times 256 + 256 = 589824 + 256 = 590080$
F1	$14 \times 14 \times 256 \times 1024 + 1024 = 51381248$
F2	$1024 \times 100 + 100 = 102500$

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$$\text{total} = 52444644$$