



Convolution layer C1

$$k_{1,p}^1 = (5, 5) \quad (p = 1, 2, \dots, 6)$$

$$b_p^1 = (1, 1) \quad (p = 1, 2, \dots, 6)$$

$$C_p^1 = \sigma(I * k_{1,p}^1 + b_p^1) \quad (p = 1, 2, \dots, 6)$$

$$C_p^1(i, j) = \sigma \left(\sum_{u=-2}^2 \sum_{v=-2}^2 I(i-u, j-v) \cdot k_{1,p}^1(u, v) + b_p^1 \right) = (24, 24)$$

$$\sigma(x) = \frac{1}{1 + e^{-x}}$$

Pooling Layer S1

$$S_p^1(i, j) = \frac{1}{4} \sum_{u=0}^1 \sum_{v=0}^1 C_p^1(2i-u, 2j-v) = (12, 12) \quad (i, j = 1, 2, \dots, 12)$$

The indexing $2i-u, 2j-v$ accesses the 2×2 grids that are pooled.

Convolution layer C2

$$k_{p,q}^2 = (5, 5) \quad (q = 1, 2, \dots, 12)$$

$$b_q^2 = (1, 1) \quad (q = 1, 2, \dots, 12)$$

$$C_q^2 = \sigma \left(\sum_{p=1}^6 S_p^1 * k_{p,q}^2 + b_q^2 \right) \quad (q = 1, 2, \dots, 12)$$

$$C_q^2(i, j) = \sigma \left(\sum_{p=1}^6 \sum_{u=-2}^2 \sum_{v=-2}^2 S_p^1(i-u, j-v) \cdot k_{p,q}^2(u, v) + b_q^2 \right) = (8, 8)$$

Pooling Layer S2

$$S_q^2(i, j) = \frac{1}{4} \sum_{u=0}^1 \sum_{v=0}^1 C_q^2(2i-u, 2j-v) = (4, 4) \quad (i, j = 1, 2, \dots, 4)$$

Vectorisation and concatenation

Each S_q^2 is a 4×4 matrix, and there are 12 such matrices on the S2 layer.

$$f = \text{concatenate} \left(\text{asColumnVector} \left(\{S_q^2\}_{q=1,2,\dots,12} \right) \right) = F \left(\{S_q^2\}_{q=1,2,\dots,12} \right)$$

$$\{S_q^2\}_{q=1,2,\dots,12} = F^{-1}(f)$$

FC layer

$$W = (10, 192)$$

$$b = (10, 1)$$

$$\hat{y} = \sigma(W \times f + b) = a$$

$$\hat{y}_i = \sigma \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) = \sigma(Z_i)$$

$$L = \frac{1}{2} \sum_{i=1}^{10} (\hat{y}_i - y_i)^2$$

Backpropagation

We update the parameters in reverse order. $W, b, k_{p,q}^2, b_q^2, k_{1,p}^1, b_p^1$.

Backpropagation, part I

Backpropagation 3a, $\Delta W = (10, 192)$

$$\begin{aligned} \Delta W_{i,j} &= \frac{\partial L}{\partial W_{i,j}} = \frac{\partial L}{\partial \hat{y}_i} \frac{\partial \hat{y}_i}{\partial W_{i,j}} = (\hat{y}_i - y_i) \frac{\partial}{\partial W_{i,j}} \sigma \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) \\ &= (\hat{y}_i - y_i) \sigma'(Z_i) \frac{\partial}{\partial W_{i,j}} \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) = (\hat{y}_i - y_i) \sigma(Z_i) (1 - \sigma(Z_i)) f_j \\ &= (\hat{y}_i - y_i) \hat{y}_i (1 - \hat{y}_i) f_j = (10, 1) \end{aligned}$$

Denote $\Delta \hat{y}_i = (\hat{y}_i - y_i) \hat{y}_i (1 - \hat{y}_i)$

$$\begin{aligned} \Delta W_{i,j} &= \Delta \hat{y}_i f_j \\ \Delta W &= \Delta \hat{y} \times f^T \quad (10, 192) = (10, 1) \times (1, 192) \end{aligned}$$

Backpropagation 3b, $\Delta b = (10, 1)$

$$\begin{aligned} \Delta b_i &= \frac{\partial L}{\partial b_i} = \frac{\partial L}{\partial \hat{y}_i} \frac{\partial \hat{y}_i}{\partial b_i} = (\hat{y}_i - y_i) \frac{\partial}{\partial b_i} \sigma \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) \\ &= (\hat{y}_i - y_i) \sigma'(Z_i) \frac{\partial}{\partial b_i} \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) = (\hat{y}_i - y_i) \sigma(Z_i) (1 - \sigma(Z_i)) \\ &= (\hat{y}_i - y_i) \hat{y}_i (1 - \hat{y}_i) = \Delta \hat{y}_i \\ \Delta b &= \Delta \hat{y} \quad (10, 1) = (10, 1) \end{aligned}$$

Backpropagation, part II

Backpropagation 2a, $\Delta S_q^2 = (4, 4)$

$$\begin{aligned} \Delta f_j &= \frac{\partial L}{\partial f_j} = \sum_{i=1}^{10} \frac{\partial L}{\partial \hat{y}_i} \frac{\partial \hat{y}_i}{\partial f_j} = \sum_{i=1}^{10} \left\{ (\hat{y}_i - y_i) \frac{\partial}{\partial f_j} \sigma \left(\sum_{j=1}^{192} W_{i,j} f_j + b_i \right) \right\} \\ &= \sum_{i=1}^{10} (\hat{y}_i - y_i) \hat{y}_i (1 - \hat{y}_i) W_{i,j} = \sum_{i=1}^{10} \Delta \hat{y}_i W_{i,j} \\ \Delta f &= W^T \times \Delta \hat{y} \quad (192, 1) = (192, 10) \times (10, 1) \end{aligned}$$

Backpropagation 2b, $\Delta C_q^2 = (8, 8)$

We reshape Δf by

$$\{\Delta S_q^2\}_{q=1,2,\dots,12} = F^{-1}(\Delta f)$$

which gives the error on the S2 layer made of twelve 4×4 maps. There are no parameters on the S2 layer so we do not need to take any derivatives. Perform upsampling to obtain the error on the C2 layer.

$$\Delta C_q^2(i, j) = \frac{1}{4} \Delta S_q^2 \left(\left\lceil \frac{i}{2} \right\rceil, \left\lceil \frac{j}{2} \right\rceil \right) \quad (i, j = 1, 2, \dots, 8)$$

Backpropagation 2c, $\Delta k_{p,q}^2 = (5, 5)$

$$\Delta k_{p,q}^2(u, v) = \frac{\partial L}{\partial k_{p,q}^2(u, v)} = \sum_{i=1}^8 \sum_{j=1}^8 \frac{\partial L}{\partial C_q^2(i, j)} \frac{\partial C_q^2(i, j)}{\partial k_{p,q}^2(u, v)}$$

$$\begin{aligned}
&= \sum_{i=1}^8 \sum_{j=1}^8 \left\{ \Delta C_q^2(i, j) \frac{\partial}{\partial k_{p,q}^2(u, v)} \sigma \left(\sum_{p=1}^6 \sum_{u=-2}^2 \sum_{v=-2}^2 S_p^1(i-u, j-v) \cdot k_{p,q}^2(u, v) + b_p^2 \right) \right\} \\
&= \sum_{i=1}^8 \sum_{j=1}^8 \Delta C_q^2(i, j) C_q^2(i, j) (1 - C_q^2(i, j)) S_p^1(i-u, j-v)
\end{aligned}$$

Denote $\Delta C_{q,\sigma}^2(i, j) = \Delta C_q^2(i, j) C_q^2(i, j) (1 - C_q^2(i, j))$, which is actually the error before the sigmoid activation on the C2 layer.

$$\begin{aligned}
C_{q,\sigma}^2(i, j) &= \sum_{p=1}^6 \sum_{u=-2}^2 \sum_{v=-2}^2 S_p^1(i-u, j-v) \cdot k_{p,q}^2(u, v) + b_p^2 \\
\Delta C_{q,\sigma}^2(i, j) &= \Delta C_q^2(i, j) C_q^2(i, j) (1 - C_q^2(i, j)) \\
\Delta k_{p,q}^2(u, v) &= \sum_{i=1}^8 \sum_{j=1}^8 \Delta C_{q,\sigma}^2(i, j) S_p^1(i-u, j-v)
\end{aligned}$$

Rotating S_p^1 by 180 degrees, we get $S_{p,180}^1(u-i, v-j) = S_p^1(i-u, j-v)$

$$\begin{aligned}
\Delta k_{p,q}^2(u, v) &= \sum_{i=1}^8 \sum_{j=1}^8 S_{p,180}^1(u-i, v-j) \Delta C_{q,\sigma}^2(i, j) \\
\Delta k_{p,q}^2 &= S_{p,180}^1 * \Delta C_{q,\sigma}^2
\end{aligned}$$

Backpropagation 2d, $\Delta b_q^2 = (1, 1)$

$$\begin{aligned}
\Delta b_q^2 &= \frac{\partial L}{\partial b_q^2} = \sum_{i=1}^8 \sum_{j=1}^8 \frac{\partial L}{\partial C_q^2(i, j)} \frac{\partial C_q^2(i, j)}{\partial b_q^2} \\
&= \sum_{i=1}^8 \sum_{j=1}^8 \left\{ \Delta C_q^2(i, j) \frac{\partial}{\partial b_q^2} \sigma \left(\sum_{p=1}^6 \sum_{u=-2}^2 \sum_{v=-2}^2 S_p^1(i-u, j-v) \cdot k_{p,q}^2(u, v) + b_p^2 \right) \right\} \\
&= \sum_{i=1}^8 \sum_{j=1}^8 \Delta C_q^2(i, j) C_q^2(i, j) (1 - C_q^2(i, j)) = \sum_{i=1}^8 \sum_{j=1}^8 \Delta C_{q,\sigma}^2(i, j)
\end{aligned}$$

Backpropagation, part III

Backpropagation 3a, $\Delta S_p^1 = (12, 12)$

$$\begin{aligned}
\Delta S_p^1(i, j) &= \frac{\partial L}{\partial S_p^1(i, j)} = \sum_{q=1}^{12} \sum_{u=-2}^2 \sum_{v=-2}^2 \frac{\partial L}{\partial C_{q,\sigma}^2(i+u, j+v)} \frac{\partial C_{q,\sigma}^2(i+u, j+v)}{\partial S_p^1(i, j)} \\
&= \sum_{q=1}^{12} \sum_{u=-2}^2 \sum_{v=-2}^2 \left\{ \Delta C_{q,\sigma}^2(i+u, j+v) \frac{\partial}{\partial S_p^1(i, j)} \left(\sum_{p=1}^6 \sum_{u=-2}^2 \sum_{v=-2}^2 S_p^1(i, j) \cdot k_{p,q}^2(u, v) + b_q^2 \right) \right\} \\
&= \sum_{q=1}^{12} \sum_{u=-2}^2 \sum_{v=-2}^2 \Delta C_{q,\sigma}^2(i+u, j+v) k_{p,q}^2(u, v)
\end{aligned}$$

Backpropagation 3b, $\Delta C_p^1 = (24, 24)$

Rotating $k_{p,q}^2$ by 180 degrees, we get $k_{p,q,180}^2(-u, -v) = k_{p,q}^2(u, v)$

$$\begin{aligned}
\Delta S_p^1(i, j) &= \sum_{q=1}^{12} \sum_{u=-2}^2 \sum_{v=-2}^2 \Delta C_{q,\sigma}^2(i - (-u), j - (-v)) k_{p,q,180}^2(-u, -v) \\
\Delta S_p^1(i, j) &= \sum_{q=1}^{12} \Delta C_{q,\sigma}^2 * k_{p,q,180}^2
\end{aligned}$$

Perform upsampling to obtain the error on the C1 layer

$$\Delta C_p^1(i, j) = \frac{1}{4} \Delta S_p^1 \left(\left\lceil \frac{i}{2} \right\rceil, \left\lceil \frac{j}{2} \right\rceil \right) \quad (i, j = 1, 2, \dots, 24)$$

Backpropagation 3c, $\Delta k_{1,p}^1 = (5, 5)$

$$\begin{aligned} \Delta k_{1,p}^1(u, v) &= \frac{\partial L}{\partial k_{1,p}^1(u, v)} = \sum_{i=1}^{24} \sum_{j=1}^{24} \frac{\partial L}{\partial C_p^1(i, j)} \frac{\partial C_p^1(i, j)}{\partial k_{1,p}^1(u, v)} \\ &= \sum_{i=1}^{24} \sum_{j=1}^{24} \left\{ \Delta C_p^1(i, j) \frac{\partial}{\partial k_{1,p}^1(u, v)} \sigma \left(\sum_{u=-2}^2 \sum_{v=-2}^2 I(i-u, j-v) \cdot k_{1,p}^1(u, v) + b_p^1 \right) \right\} \\ &= \sum_{i=1}^{24} \sum_{j=1}^{24} \Delta C_p^1(i, j) C_p^1(i, j) (1 - C_p^1(i, j)) I(i-u, j-v) \end{aligned}$$

Denote $\Delta C_{p,\sigma}^1(i, j) = \Delta C_p^1(i, j) C_p^1(i, j) (1 - C_p^1(i, j))$ and rotating I by 180 degrees

$$\begin{aligned} \Delta k_{1,p}^1(u, v) &= \sum_{i=1}^{24} \sum_{j=1}^{24} I_{180}(u-i, v-j) \Delta C_{p,\sigma}^1(i, j) \\ \Delta k_{1,p}^1 &= I_{180} * \Delta C_{p,\sigma}^1 \end{aligned}$$

Backpropagation 3d, $\Delta b_p^1 = (1, 1)$

$$\begin{aligned} \Delta b_p^1 &= \frac{\partial L}{\partial b_p^1} = \sum_{i=1}^{24} \sum_{j=1}^{24} \frac{\partial L}{\partial C_p^1(i, j)} \frac{\partial C_p^1(i, j)}{\partial b_p^1} \\ &= \sum_{i=1}^{24} \sum_{j=1}^{24} \left\{ \Delta C_{p,\sigma}^1(i, j) \frac{\partial}{\partial b_p^1} \sigma \left(\sum_{u=-2}^2 \sum_{v=-2}^2 I(i-u, j-v) \cdot k_{1,p}^1(u, v) + b_p^1 \right) \right\} \\ &= \sum_{i=1}^{24} \sum_{j=1}^{24} \Delta C_{p,\sigma}^1(i, j) C_p^1(i, j) (1 - C_p^1(i, j)) = \sum_{i=1}^{24} \sum_{j=1}^{24} \Delta C_{p,\sigma}^1(i, j) \end{aligned}$$