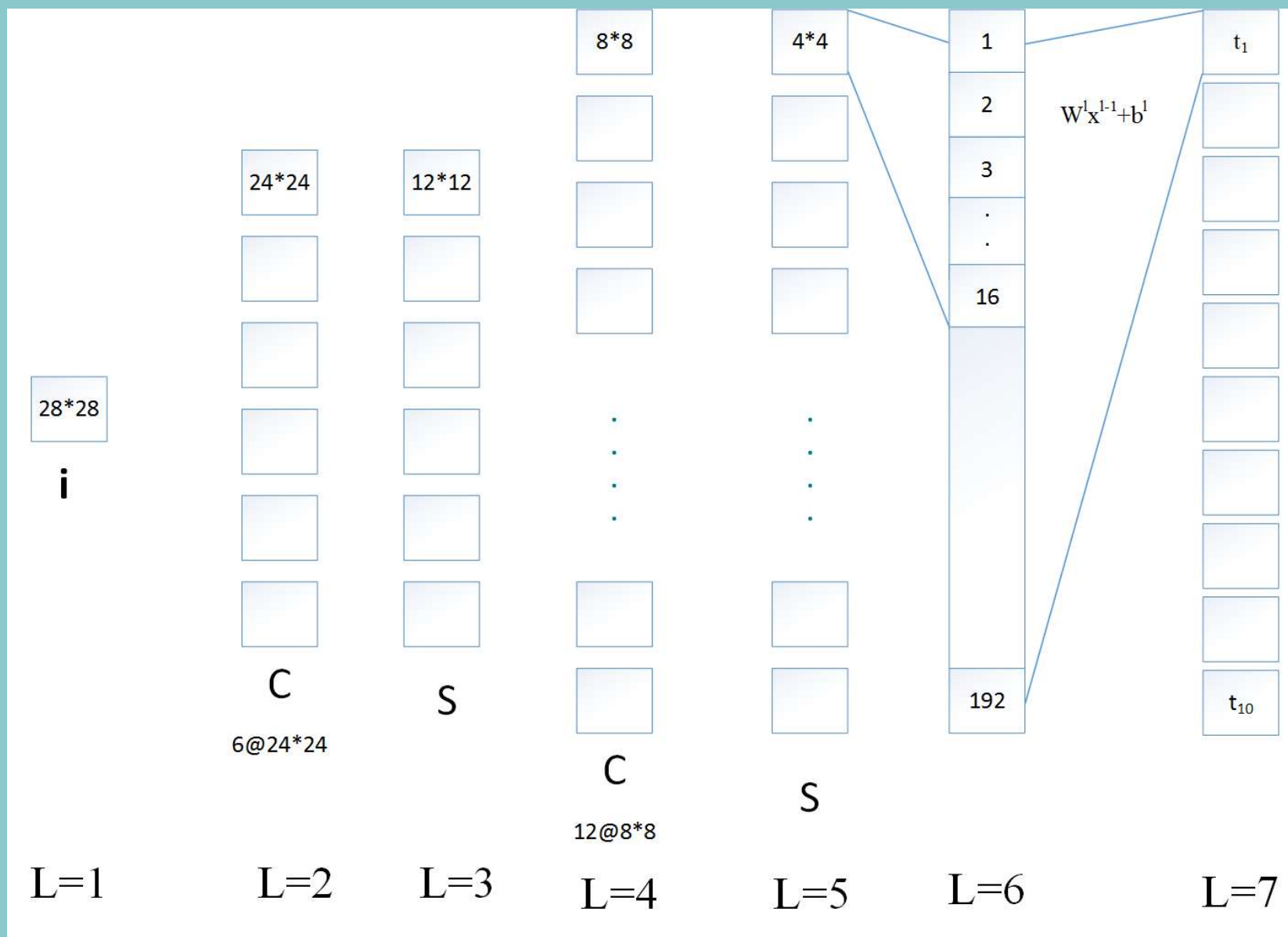


Lenet :



2

Right:

$t_k = [0.01 \ 0.02 \ 0.01 \ 0.03 \ 0.01 \ 0.03 \ 0.05 \ 0.02 \ 0.04 \ 0.98]$ $k=1:10$

$y_k = [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1]$

Wrong:

$t = [0.01 \ 0.02 \ 0.78 \ 0.03 \ 0.01 \ 0.03 \ 0.05 \ 0.02 \ 0.04 \ 0.02]$

$y = [0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 1]$

$y_k = [1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$ (汽车)

$y_k = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$ (飞机)

$$E^N = \frac{1}{2} \sum_{n=1}^N \sum_{k=1}^c (t_k^n - y_k^n)^2$$

C:类别个数 (10)

N:训练样本个数

用l表示第几层,不考虑样本数 (每份样本都是一样的), cnn中每一层:

$$E = \frac{1}{2} \sum_{k=1}^c (t_k^l - y_k^l)^2$$

理想: 误差E=0

$$t^l = f(u^l) = \text{sigmod}(u^l)$$

$$u^l = W^l x^{l-1} + b^l$$

(我们要求的是变量W和b)

反向传播（**Backpropagation Pass**, BP）

$$W = W - \alpha \frac{\partial}{\partial W} J(W, b)$$

$$b = b - \alpha \frac{\partial}{\partial b} J(W, b)$$

$$J(W, b) = E = \frac{1}{2} \sum_{k=1}^c (t_k^l - y_k^l)^2$$

$$\frac{\partial E}{\partial b} = \frac{\partial E}{\partial u} \frac{\partial u}{\partial b} = \frac{\partial E}{\partial u} = \delta \quad \left(\frac{\partial u}{\partial b} = 1 \right)$$

$$f(u_1^{l-1})$$

δ 称为这一层偏置 b 的灵敏度或残差

求最后一层的偏置:

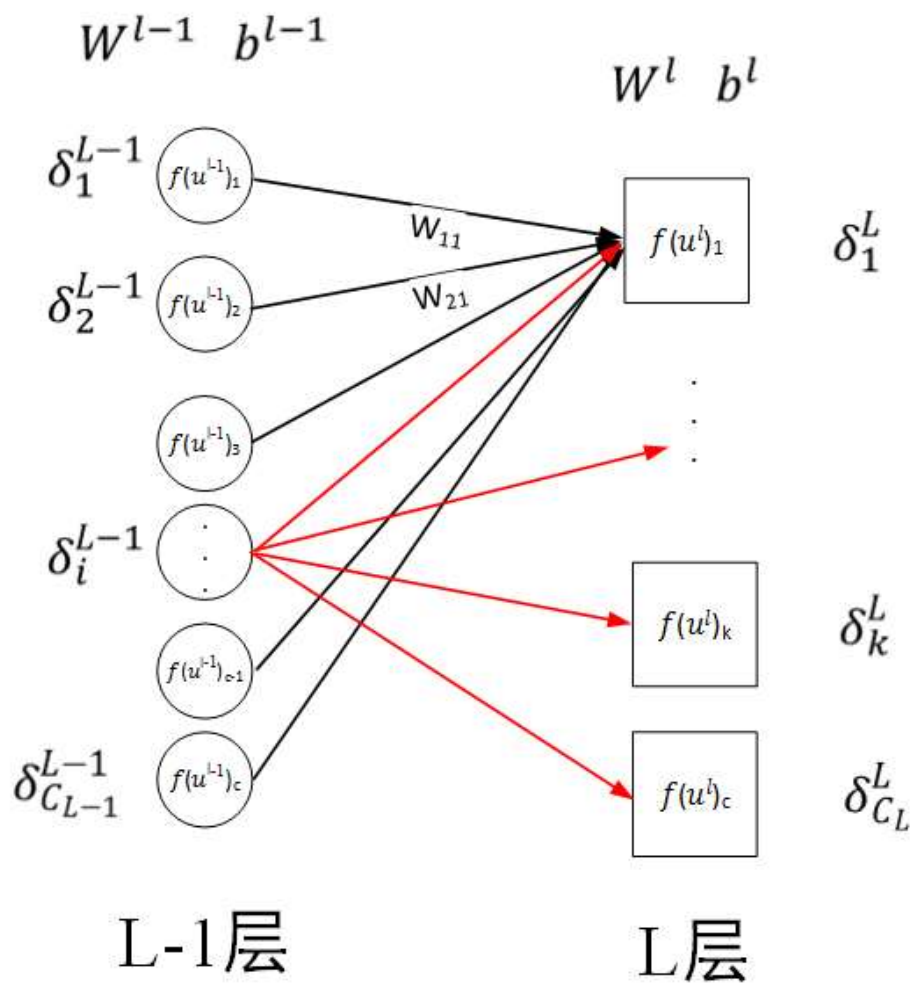
$$t^l = f(u^l) = \text{sigmoid}(u^l)$$

$$\delta_k^l = \frac{\partial E}{\partial u_k^l} = \frac{\partial}{\partial u_k^l} \left\{ \frac{1}{2} [f(u_1^l) - y_1^l]^2 + \frac{1}{2} [f(u_2^l) - y_2^l]^2 + \cdots + \frac{1}{2} [f(u_k^l) - y_k^l]^2 + \cdots + \frac{1}{2} [f(u_n^l) - y_n^l]^2 \right\}$$

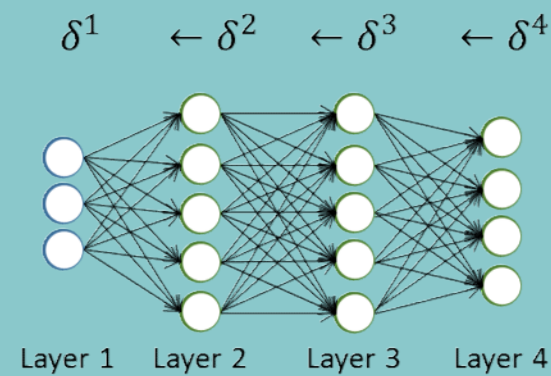
$$= (f(u_k^l) - y_k^l)(f(u_k^l) - y_k^l)'$$

$$= (f(u_k^l) - y_k^l)f'(u_k^l)$$

$$f'(u^l) = \left(\frac{1}{1 + e^{-u^l}} \right)' = \frac{1}{1 + e^{-u^l}} \left(1 - \frac{1}{1 + e^{-u^l}} \right) = f(u^l)(1 - f(u^l))$$



$$u^l = W^l f(u^{l-1}) + b^l$$



L-1层残差:

$$\begin{aligned}\delta_i^{l-1} &= \frac{\partial E}{\partial u_i^{l-1}} = \frac{\partial}{\partial u_i^{l-1}} \frac{1}{2} \sum_{k=1}^{c_l} (f(u_k^l) - y_k^l)^2 \\&= \sum_{k=1}^{c_l} (f(u_k^l) - y_k^l) \frac{\partial}{\partial u_i^{l-1}} (f(u_k^l) - y_k^l) \\&= \sum_{k=1}^{c_l} (f(u_k^l) - y_k^l) \frac{\partial}{\partial u_i^{l-1}} f(u_k^l) \frac{\partial u_k^l}{\partial u_i^l} \\&= \sum_{k=1}^{c_l} (f(u_k^l) - y_k^l) f'(u_k^l) \frac{\partial u_k^l}{\partial u_i^{l-1}} \\&= \sum_{k=1}^{c_l} \delta_k^l \frac{\partial u_k^l}{\partial u_i^{l-1}}\end{aligned}$$

$$u^l = W^l f(u^{l-1}) + b^l$$

$$= \sum_{k=1}^{c_l} \delta_k^l \frac{\partial}{\partial u_i^{l-1}} \left[\sum_{j=1}^{c_{l-1}} W_{jk}^l f(u_j^{l-1}) + b_k^l \right]$$

取j=i时才有值

$$= \sum_{k=1}^{c_l} \delta_k^l W_{ik}^l f'(u_i^{l-1})$$

$$\delta^{l-1} = (W^l)^T \delta^{l.*} f'(u^{l-1})$$

L层:

$$b = b - \alpha \frac{\partial}{\partial b} J(W, b)$$

$$= b - \alpha * \delta^l$$

$$W = W - \alpha \frac{\partial}{\partial W} J(W, b)$$

$$\frac{\partial E}{\partial W} = \frac{\partial E}{\partial u} \frac{\partial u}{\partial W} = \delta^l * f(u^{l-1})$$

$$\frac{\partial E}{\partial W_{ik}} = \frac{\partial E}{\partial u_k^l} \frac{\partial u_k^l}{\partial W_{ik}} = \delta_k^l * f(u_i^{l-1})$$

$$W = W - \alpha \frac{\partial}{\partial W} J(W, b) = W - \alpha * \delta_k^l * f(u_i^{l-1})$$

L=7

$$\delta_k^l = (f(u_k^l) - y_k^l) f'(u_k^l)$$

L=6

$$\delta^{l-1} = (W^l)^T \delta^{l.*} f'(u^{l-1})$$

L=5 S

Reshape 12@4*4

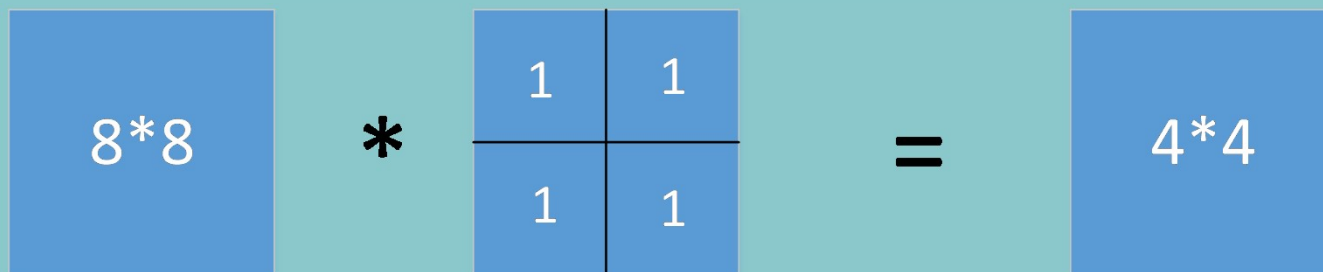
L=4 C

$\delta =$

0.3517	0.9172	0.3804	0.5308
0.8308	0.2858	0.5678	0.7792
0.5853	0.7572	0.0759	0.9340
0.5497	0.7537	0.0540	0.1299

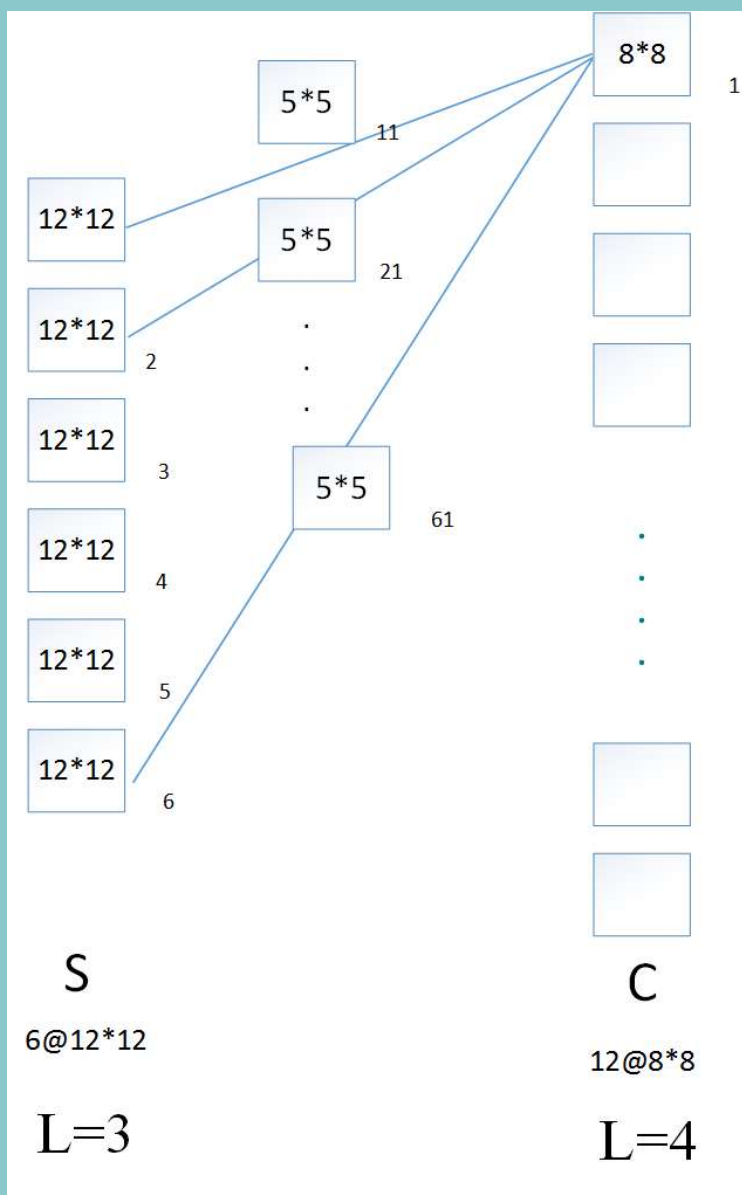
ans =

0.3517	0.3517	0.9172	0.9172	0.3804	0.3804	0.5308	0.5308
0.3517	0.3517	0.9172	0.9172	0.3804	0.3804	0.5308	0.5308
0.8308	0.8308	0.2858	0.2858	0.5678	0.5678	0.7792	0.7792
0.8308	0.8308	0.2858	0.2858	0.5678	0.5678	0.7792	0.7792
0.5853	0.5853	0.7572	0.7572	0.0759	0.0759	0.9340	0.9340
0.5853	0.5853	0.7572	0.7572	0.0759	0.0759	0.9340	0.9340
0.5497	0.5497	0.7537	0.7537	0.0540	0.0540	0.1299	0.1299
0.5497	0.5497	0.7537	0.7537	0.0540	0.0540	0.1299	0.1299



$$\delta^{l-1} = (W^l)^T \delta^{l,*} f'(u^{l-1})$$

L=3 S



一共6*12个卷积核

$$\begin{matrix} 12*12 \\ 1 \end{matrix} * \begin{matrix} 5*5 \\ 11 \end{matrix} = \begin{matrix} 8*8 \\ 11 \end{matrix}$$

$$\begin{matrix} 12*12 \\ 2 \end{matrix} * \begin{matrix} 5*5 \\ 21 \end{matrix} = \begin{matrix} 8*8 \\ 21 \end{matrix}$$

\vdots
 \vdots
 \vdots

$$\begin{matrix} 12*12 \\ 6 \end{matrix} * \begin{matrix} 5*5 \\ 61 \end{matrix} = \begin{matrix} 8*8 \\ 61 \end{matrix}$$

Sum



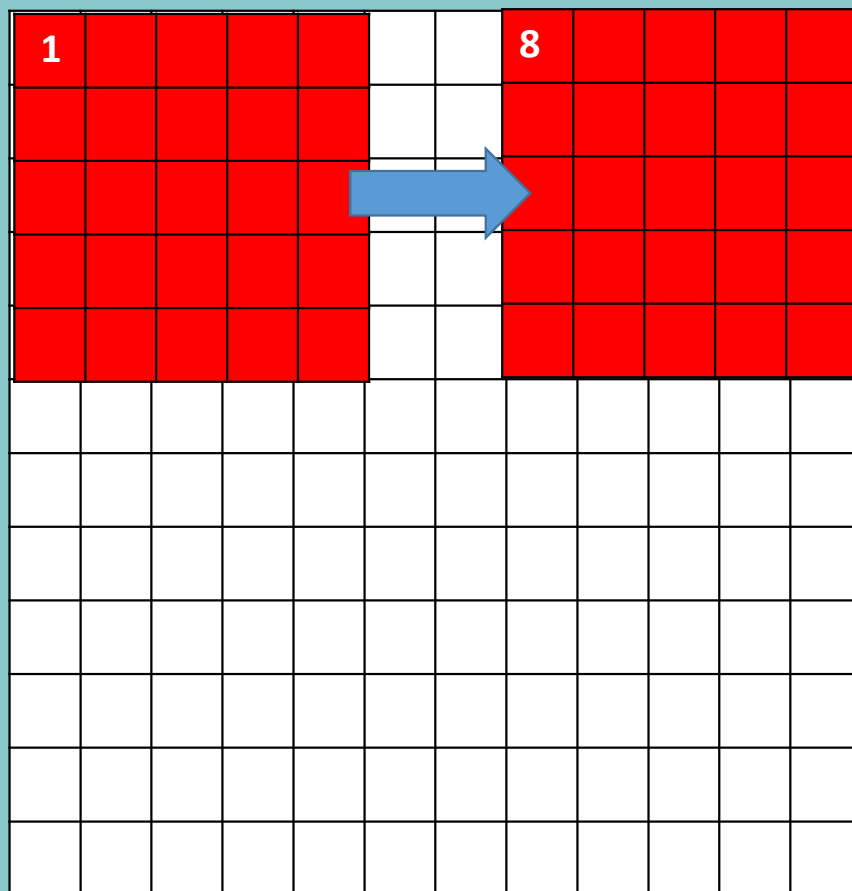
$$8*8$$

+ b

Sigmoid



$$\begin{matrix} 8*8 \\ 1 \end{matrix}$$

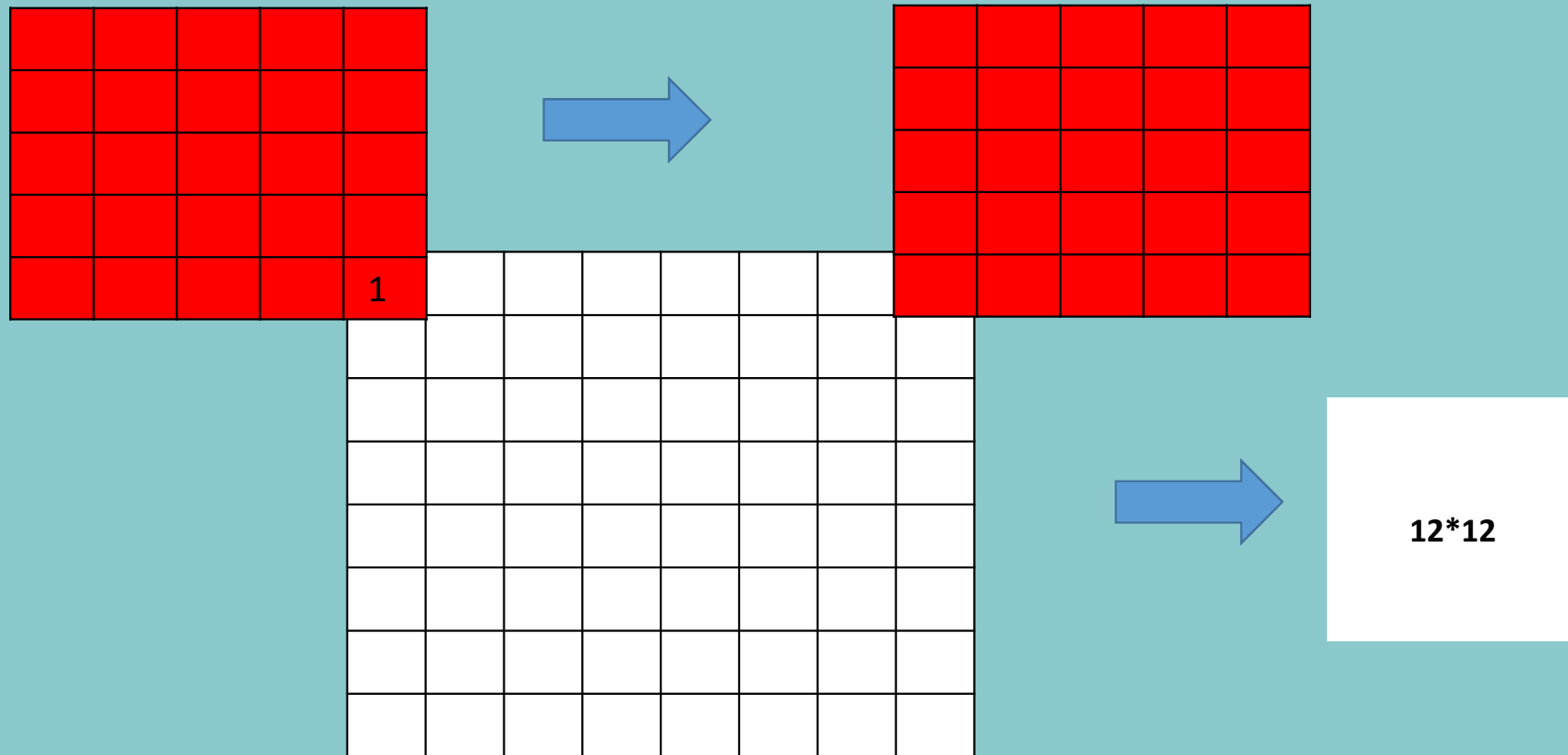


Image_size: $L \times L = [12, 12]$

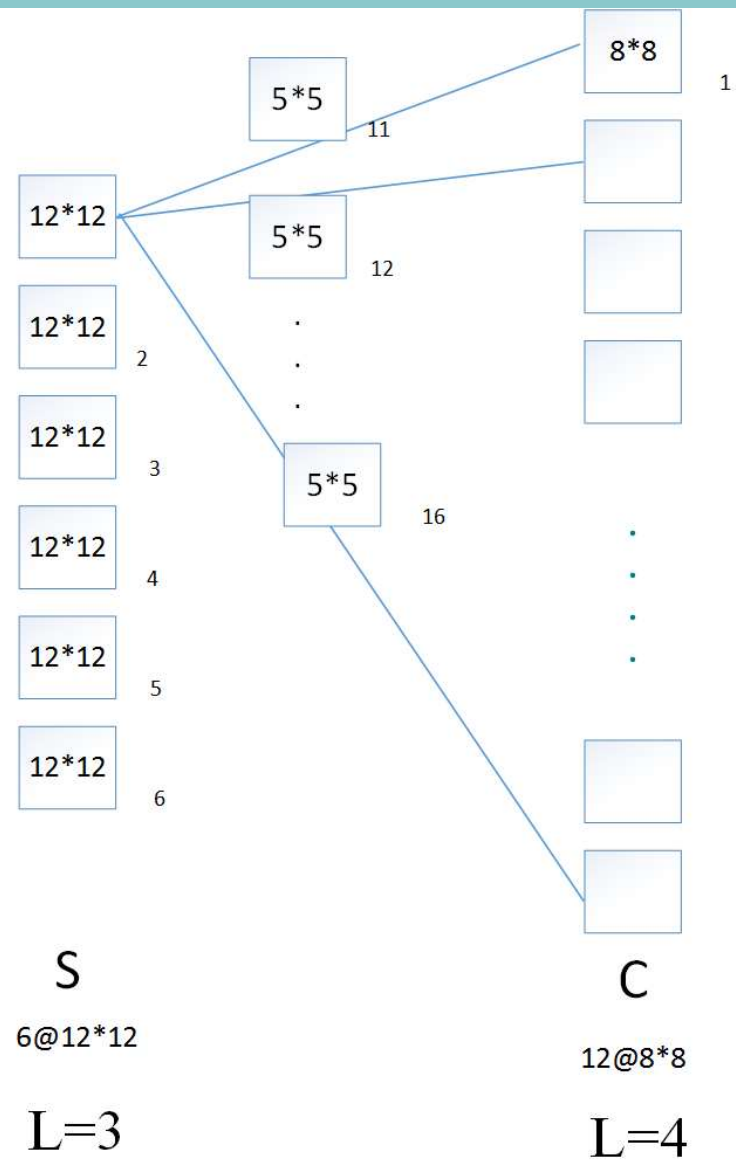
Weight_size: $k \times k = [5 \times 5]$

移动步长 : stride:1

Output_image_size:
 $(L-k)/\text{stride}+1 = (12-5)/1+1 = 8 \times 8$



$12 \times 12 = \text{convn}(8 \times 8, 5 \times 5, 'full')$



12个12*12的残差相加得到L3第一个12*12残差

L=2 C

类似L=4 扩展，公式。

L=1 I

无参数及残差

更新参数W、b

$$W = W - \alpha \frac{\partial}{\partial W} J(W, b)$$

$$b = b - \alpha \frac{\partial}{\partial b} J(W, b)$$

再输入一批训练图片得到输出t，继续训练