# Convolution and Deconvolution

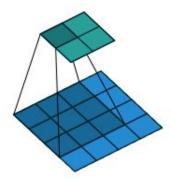
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### 1 2D Convolution



The blue grid is input image  $(4 \times 4)$ , and the green grid is output image  $(2 \times 2)$ . The kernel size is  $3 \times 3$ . Stride is 1 and padding is 0. The weight of the kernel at the localization (i, j) is donated as  $w_{i,j}$ . The value of the image at the location (i, j) is donated as  $x_{i,j}$ . The kernel is:

$$w = \begin{bmatrix} w_{0,0} & w_{0,1} & w_{0,0} \\ w_{1,0} & w_{1,1} & w_{1,2} \\ w_{2,0} & w_{2,1} & w_{2,2} \end{bmatrix}$$

Rewrite image X as a vector:  $\begin{bmatrix} x_{0,0} & x_{0,1} & \dots & x_{3,2} & x_{3,3} \end{bmatrix}^T$ . Therefore, convolution can be written as matrix multiplication:

$$Y = w * X = WX =$$

$$\begin{bmatrix} w_{0,0} \ w_{0,1} \ w_{0,2} \ 0 \ w_{1,0} \ w_{1,1} \ w_{1,2} \ 0 \ w_{2,0} \ w_{2,1} \ w_{2,2} \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \\ 0 \ w_{0,0} \ w_{0,1} \ w_{0,2} \ 0 \ w_{1,0} \ w_{1,1} \ w_{1,2} \ 0 \ w_{2,0} \ w_{2,1} \ w_{2,2} \ 0 \ 0 \ 0 \ 0 \ 0 \\ 0 \ 0 \ 0 \ 0 \ w_{0,0} \ w_{0,1} \ w_{0,2} \ 0 \ w_{1,0} \ w_{1,1} \ w_{1,2} \ 0 \ w_{2,0} \ w_{2,1} \ w_{2,2} \ 0 \\ 0 \ 0 \ 0 \ 0 \ 0 \ w_{0,0} \ w_{0,1} \ w_{0,2} \ 0 \ w_{1,0} \ w_{1,1} \ w_{1,2} \ 0 \ w_{2,0} \ w_{2,1} \ w_{2,2} \ 0 \\ x_{2,1} \ x_{2,2} \ x_{2,3} \ x_{3,0} \ x_{3,1} \ x_{3,2} \ x_{3,3} \end{bmatrix}$$

 $x_{0,0}$   $x_{0,1}$   $x_{0,2}$   $x_{0,3}$ 

W is a doubly block circulant matrix (a special case of Toeplitz matrix). Here we define:

$$W_0 = \begin{bmatrix} w_{0,0} & w_{0,1} & w_{0,2} & 0 \\ 0 & w_{0,0} & w_{0,1} & w_{0,2} \end{bmatrix} \ W_1 = \begin{bmatrix} w_{1,0} & w_{1,1} & w_{1,2} & 0 \\ 0 & w_{1,0} & w_{1,1} & w_{1,2} \end{bmatrix}$$

$$W_3 = \begin{bmatrix} w_{2,0} & w_{2,1} & w_{2,2} & 0\\ 0 & w_{2,0} & w_{2,1} & w_{2,2} \end{bmatrix} \ \mathbf{0} = \begin{bmatrix} 0 & 0 & 0 & 0\\ 0 & 0 & 0 & 0 \end{bmatrix}$$

 $W_0, W_1, W_2$ , and **0** are all Toeplitz matrix. W can be written as:

$$W = \begin{bmatrix} W_0 & W_1 & W_2 & \mathbf{0} \\ \mathbf{0} & W_0 & W_1 & W_2 \end{bmatrix}$$

Therefore, W is a doubly block circulant matrix.

#### 2 2D Deconvolution

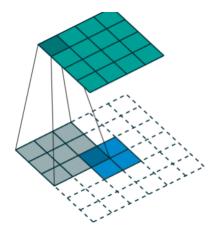
In the above example, the size of X is  $16 \times 1$  and the size of Y is  $4 \times 1$ . The dimension of W is  $4 \times 16$ . In order to recover the size of Y to the size of X, we just need  $W^TY$  (We cannot recover the value of X, but just size):

$$Y' = W^T Y = W^T (WX)$$

If we do nate Y as:  $\begin{bmatrix} y_{0,0} & y_{0,1} & y_{1,0} & y_{1,1} \end{bmatrix}^T$ , the deconvolution can be written as:

$$Y' = W^T Y = \begin{bmatrix} w_{0,0} & 0 & 0 & 0 \\ w_{0,1} & w_{0,0} & 0 & 0 \\ w_{0,2} & w_{0,1} & 0 & 0 \\ 0 & w_{0,2} & 0 & 0 \\ w_{1,0} & 0 & w_{0,0} & 0 \\ w_{1,1} & w_{1,0} & w_{0,1} & w_{0,0} \\ w_{1,2} & w_{1,1} & w_{0,2} & w_{0,1} \\ 0 & w_{1,2} & 0 & w_{0,2} \\ w_{2,0} & 0 & w_{1,0} & 0 \\ w_{2,1} & w_{2,0} & w_{1,1} & w_{1,0} \\ w_{2,2} & w_{2,1} & w_{1,2} & w_{1,1} \\ 0 & w_{2,2} & 0 & w_{1,2} \\ 0 & 0 & w_{2,0} & 0 \\ 0 & 0 & w_{2,1} & w_{2,0} \\ 0 & 0 & w_{2,2} & w_{2,1} \\ 0 & 0 & 0 & w_{2,2} \end{bmatrix}$$

If we rewrite it in convolution format Y' = w' \* Y, where w' is the top-bottom AND left-right flip of the original filter w.



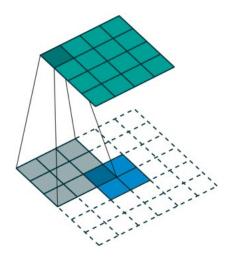
In the above image, the blue grid is Y, and the green grid is Y'. Stride is 1 and padding is 2.

We can write this process as:

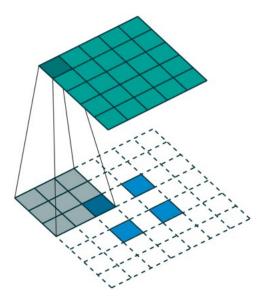
$$X - > Y = conv(X) - > deconv(Y) - > Y'$$

To keep Y' has the same size as X, we need to add paddings to conv(X). Usually we add 0 paddings to conv(X). There are two common padding strategries:

#### 1. Add paddings around an image



#### 2. Add paddings inside an image



These two strategies yield different results. How to choose this two strategies depends on which one can keep the image size.