

Convolutional Neural Networks

The step size taken by a kernel while sliding over an input image is called as a Stride

We have our usual 6×6 matrix as before

$$A = \begin{pmatrix} 3 & 2 & 7 & 1 & 6 & 3 \\ 4 & 1 & 1 & 0 & 9 & 8 \\ 8 & 2 & 0 & 1 & 2 & 4 \\ 5 & 6 & 0 & 7 & 1 & 4 \\ 0 & 5 & 4 & 9 & 6 & 0 \\ 4 & 6 & 2 & 4 & 0 & 5 \end{pmatrix}$$

We have a convolutional kernel K as below.

$$K = \begin{pmatrix} 1 & 2 & 2 \\ 2 & 1 & 2 \\ 2 & 2 & 1 \end{pmatrix}$$

$$m = 6$$

$$f = 3 \quad n = 3$$

Let us consider a stride $(s) = 3$

$$\begin{aligned} \text{The size of output matrix} &= \left\lfloor \frac{m-f}{s} + 1 \right\rfloor \times \left\lfloor \frac{m-f}{s} + 1 \right\rfloor \\ &= \left\lfloor \frac{6-3}{3} + 1 \right\rfloor \times \left\lfloor \frac{6-3}{3} + 1 \right\rfloor \\ &= 2 \times 2. \end{aligned}$$

$$\text{We will have output matrix } H = \begin{pmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \end{pmatrix}$$

$$\begin{aligned} \text{Here, } h_{11} &= \underline{3 \times 1} + \underline{2 \times 2} + \underline{7 \times 2} + \\ &\quad \underline{4 \times 2} + \underline{1 \times 1} + \underline{1 \times 2} + \\ &\quad \underline{8 \times 2} + \underline{2 \times 2} + \underline{0 \times 1} \\ &= \underline{52} \end{aligned}$$

$$A = \begin{pmatrix} 3 & 2 & 7 & 1 & 6 & 3 \\ 4 & 1 & 1 & 0 & 9 & 8 \\ 8 & 2 & 0 & 1 & 2 & 4 \\ 5 & 6 & 0 & 7 & 1 & 4 \\ 0 & 5 & 4 & 9 & 6 & 0 \\ 4 & 6 & 2 & 4 & 0 & 5 \end{pmatrix}$$

$$\begin{aligned}
 \underline{\underline{h_{13}}} &= \underline{1} \times 1 + \underline{6} \times 2 + \underline{3} \times 2 + \\
 &= \underline{0} \times 2 + \underline{9} \times 1 + \underline{8} \times 2 + \\
 &= \underline{1} \times 2 + \underline{2} \times 2 + \underline{4} \times 1 \\
 &= \underline{\underline{54}}
 \end{aligned}$$

$$A = \begin{pmatrix} 3 & 2 & 7 & 1 & 6 & 3 \\ 4 & 1 & 1 & 0 & 9 & 8 \\ 8 & 2 & 0 & 1 & 2 & 4 \\ 5 & 6 & 0 & 7 & 1 & 4 \\ 0 & 5 & 4 & 9 & 6 & 0 \\ 4 & 6 & 2 & 4 & 0 & 5 \end{pmatrix}$$

$$\begin{aligned}
 h_{21} &= \underline{5} \times 1 + \underline{6} \times 2 + \underline{0} \times 2 + \\
 &= \underline{0} \times 2 + \underline{5} \times 1 + \underline{4} \times 2 + \\
 &= 4 \times 2 + 6 \times 2 + 2 \times 1 \\
 &= \underline{\underline{52}}
 \end{aligned}$$

$$A = \begin{pmatrix} 3 & 2 & 7 & 1 & 6 & 3 \\ 4 & 1 & 1 & 0 & 9 & 8 \\ 8 & 2 & 0 & 1 & 2 & 4 \\ 5 & 6 & 0 & 7 & 1 & 4 \\ 0 & 5 & 4 & 9 & 6 & 0 \\ 4 & 6 & 2 & 4 & 0 & 5 \end{pmatrix}$$

$$\begin{aligned}
 h_{22} &= 7 \times 1 + 1 \times 2 + 4 \times 2 + \\
 &= 9 \times 2 + 6 \times 1 + 0 \times 2 + \\
 &= 4 \times 2 + 0 \times 2 + 5 \times 1 \\
 &= \underline{\underline{54}}
 \end{aligned}$$

The output matrix is

$$H = \begin{pmatrix} \underline{\underline{52}} & \underline{\underline{54}} \\ \underline{\underline{52}} & \underline{\underline{54}} \end{pmatrix}$$

$$6 \times 6, 3 \times 3, 3$$

$$H \rightarrow 2 \times 2$$

Padding:- Adding extra layers of zeros along the edges of input image before performing convolution

In case, input size = output size we apply padding.

$$\text{Input size} = \underline{m \times m}$$

$$\text{filter size} = \underline{f \times f}$$

$$\text{Output size} = \underline{m \times m} \text{ if } \boxed{\beta = \frac{f-1}{2}}$$

where β is padding size.

$$f \times f = 3 \times 3$$

$$\beta = \frac{3-1}{2} = 1$$

In our example, padded input A_β will be.

$$A_\beta = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 2 & 7 & 1 & 6 & 3 & 0 \\ 0 & 4 & 1 & 1 & 0 & 9 & 8 & 0 \\ 0 & 8 & 2 & 0 & 1 & 2 & 4 & 0 \\ 0 & 5 & 6 & 0 & 7 & 1 & 4 & 0 \\ 0 & 0 & 5 & 4 & 9 & 6 & 0 & 0 \\ 0 & 4 & 6 & 2 & 4 & 0 & 5 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

The padded input A_β will now be of size 8×8 .

[Reason:- As $f=3$; $\beta = \frac{3-1}{2} = 1$ [we pad by 1]]

$$n = 8$$

$$f = 3$$

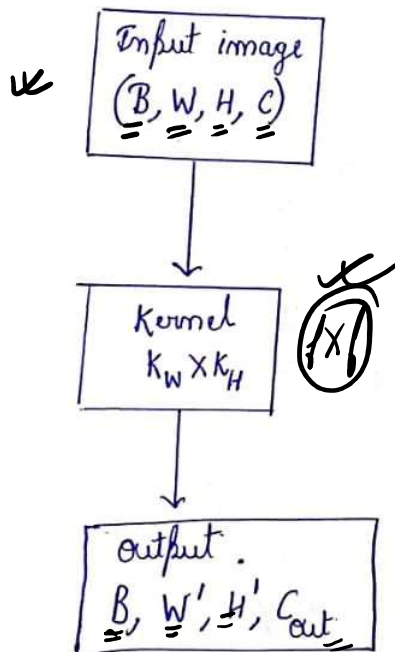
$$\begin{aligned} \text{output size} &= \left[\frac{8-3}{1} + 1 \right] \times \left[\frac{8-3}{1} + 1 \right] \\ &= 6 \times 6 = \text{same as input} \end{aligned}$$

$$\begin{aligned} D &= 1 \\ \left[\frac{n-f}{2} + 1 \right] &= 6 \end{aligned}$$

$$n = 6$$

$$n + 2\beta = 6 + 2(1) = 8$$

For an input image of $m \times m$ size, a kernel of $f \times f$ will give a padding of $\beta = \frac{f-1}{2}$. The size of output image after applying kernel on padded input is $(m + 2\beta - f + 1) \times (m + 2\beta - f + 1)$



we have a batch of input images each of width W , height H & channel C .

we consider C_{out} kernels of size $K_W \times K_H$
 ↳ Number of kernels

Output is of size

$$\left[\begin{array}{l} W' = \left\lfloor \frac{W - K_W}{2} + 1 \right\rfloor \\ H' = \left\lfloor \frac{H - K_H}{2} + 1 \right\rfloor \\ C_{out} = \text{Number of kernels} \end{array} \right]$$

Dilated Convolution: The convolution performed by inflating the kernel by inserting spaces between the kernel elements

we have input matrix of size $m \times m$ ✓
 kernel matrix of size $f \times f$ ✓
 padding done of size β ✓
 stride of s ✓
 dilation rate = d

$$\left\lceil \frac{m + 2\beta - f}{s} + 1 \right\rceil$$

$-(f-1)(d-1)$

The size of output = $\left\lceil \frac{m + 2\beta - f - \underbrace{(f-1)(d-1)}}{s} + 1 \right\rceil \times$

✓ $\left\lceil \frac{m + 2\beta - f - (f-1)(d-1)}{s} + 1 \right\rceil$

$$\beta = \frac{f-1}{2} = 1$$

we will consider our usual matrix A with padding $\beta=1$
 $m = \underline{6}$ [$\because A$ was of size 6×6], $f = \underline{3}$ [$\because k$ was 3×3 matrix] ✓
 we consider a stride $s = \underline{1}$ & dilation rate of 3 (i.e. $d=3$)

we have size of output = $\left\lceil \frac{6 + 2(1) - \underline{3} - \underline{(3-1)(3-1)}}{1} + 1 \right\rceil \times \left\lceil \frac{6 + 2(1) - 3 - (3-1)(\underline{3})}{1} + 1 \right\rceil$
 $= 2 \times 2$ ✓

we will have a 2×2 matrix as output

We will have padded matrix as below -

$$A_p = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 2 & 7 & 1 & 6 & 3 & 0 \\ 0 & 4 & 1 & 1 & 0 & 9 & 8 & 0 \\ 0 & 8 & 2 & 0 & 1 & 2 & 4 & 0 \\ 0 & 5 & 6 & 0 & 7 & 1 & 4 & 0 \\ 0 & 0 & 5 & 4 & 9 & 6 & 0 & 0 \\ 0 & 4 & 6 & 2 & 4 & 0 & 5 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix} = (a'_{ij})_{8 \times 8}$$

$$d = 3$$

To calculate h_{11} , we have the below submatrix

$$A_p^{(11)} = \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 4 \\ 0 & 2 & 5 \end{pmatrix} = \begin{pmatrix} a'_{11} & a'_{14} & a'_{17} \\ a'_{41} & a'_{44} & a'_{47} \\ a'_{71} & a'_{74} & a'_{77} \end{pmatrix}$$

$$h_{11} = 0 \times 1 + 0 \times 2 + 0 \times 2 + \\ 0 \times 2 + 0 \times 1 + 4 \times 2 + \\ 0 \times 2 + 2 \times 2 + 5 \times 1 \\ = 17$$

$$A_p = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 3 & 2 & 7 & 1 & 6 & 3 & 0 \\ 0 & 4 & 1 & 1 & 0 & 9 & 8 & 0 \\ 0 & 8 & 2 & 0 & 1 & 2 & 4 & 0 \\ 0 & 5 & 6 & 0 & 7 & 1 & 4 & 0 \\ 0 & 0 & 5 & 4 & 9 & 6 & 0 & 0 \\ 0 & 4 & 6 & 2 & 4 & 0 & 5 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

To calculate h_{12} , we have below submatrix

$$A_p^{(12)} = \begin{pmatrix} 0 & 0 & 0 \\ 8 & 1 & 0 \\ 4 & 4 & 0 \end{pmatrix} = \begin{pmatrix} a'_{12} & a'_{15} & a'_{18} \\ a'_{42} & a'_{45} & a'_{48} \\ a'_{72} & a'_{75} & a'_{78} \end{pmatrix}$$

$$\begin{aligned}
 h_{12} &= 0 \times 1 + 0 \times 2 + 0 \times 2 + \\
 &\quad 8 \times 2 + 1 \times 1 + 0 \times 2 + \\
 &\quad 4 \times 2 + 4 \times 2 + 0 \times 1 \\
 &= 33
 \end{aligned}$$

To calculate h_{21} , we have the below submatrix

$$A_p^{(21)} = \begin{pmatrix} 0 & 7 & 3 \\ 0 & 0 & 4 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} a'_{21} & a'_{24} & a'_{27} \\ a'_{51} & a'_{54} & a'_{57} \\ a'_{81} & a'_{84} & a'_{87} \end{pmatrix}$$

$$\begin{aligned}
 h_{21} &= 0 \times 1 + 7 \times 2 + 3 \times 2 + \\
 &\quad 0 \times 2 + 0 \times 1 + 4 \times 2 + \\
 &\quad 0 \times 2 + 0 \times 2 + 0 \times 1 \\
 &= 14 + 6 + 8 = 28
 \end{aligned}$$

To calculate h_{22} , we have the below submatrix

$$A_p^{(22)} = \begin{pmatrix} a'_{22} & a'_{25} & a'_{28} \\ a'_{52} & a'_{55} & a'_{58} \\ a'_{82} & a'_{85} & a'_{88} \end{pmatrix} = \begin{pmatrix} 3 & 1 & 0 \\ 5 & 7 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\begin{aligned}
 h_{22} &= 3 \times 1 + 1 \times 2 + 0 \times 2 + \\
 &\quad 5 \times 2 + 7 \times 1 + 0 \times 2 + \\
 &\quad 0 \times 2 + 0 \times 2 + 0 \times 1 = 22
 \end{aligned}$$

$$A_p = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 8 & 0 \\ 0 & 4 & 1 & 1 & 0 & 9 & 8 & 0 \\ 0 & 8 & 2 & 0 & 1 & 2 & 4 & 0 \\ 0 & 5 & 6 & 0 & 7 & 1 & 4 & 0 \\ 0 & 0 & 5 & 4 & 9 & 6 & 0 & 0 \\ 0 & 4 & 6 & 2 & 4 & 0 & 5 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$H = \begin{pmatrix} 17 & 33 \\ 28 & 22 \end{pmatrix}$ is output matrix

$$H = \begin{pmatrix} 17 & 33 \\ 28 & 22 \end{pmatrix}$$

Dilated convolution numerical example

Consider a 7x7 input and a 3x3 convolution kernel, as shown below. Point (i, j) corresponds to the i-th row (ranging 1, 2, ...) from top to bottom and j-th column (ranging 1, 2, ...) from left to right.

0.5	1	1	1	1	1	1
1	0.5	0	0	0	0	0
0	0	0.5	1	1	1	1
1	1	1	0.5	0	0	0
0	0	0	0	0.5	1	1
1	1	1	1	1	0.5	0
0	0	0	0	0	0	0.5

→ A

1	0	-1
0	4	0
-1	0	-1

→ 15

$p = 0$

$d = 3$

Assuming zero padding and 3-dilated convolution, what will be value at (1,1) in the output plane?

We have an input matrix A as below -

$$A = \begin{pmatrix} 0.5 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0.5 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0.5 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0.5 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0.5 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0.5 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0.5 \end{pmatrix}$$

The kernel is given as

$$K = \begin{pmatrix} 1 & 0 & -1 \\ 0 & 4 & 0 \\ -1 & 0 & -1 \end{pmatrix}$$

We have $m=7$, $f=3$, $\beta=0$, $d=3$, $z=1$ (assumed)

$$\text{output size} = \left\lfloor \frac{m + 2\beta - f - (f-1)(d-1)}{z} + 1 \right\rfloor \times \left\lfloor \frac{m + 2\beta - f - (f-1)(d-1)}{z} + 1 \right\rfloor$$

$$= \left\lfloor \frac{7 + 2(0) - 3 - (3-1)(3-1)}{1} + 1 \right\rfloor \times \left\lfloor \frac{7 + 2(0) - 3 - (3-1)(3-1)}{1} + 1 \right\rfloor$$

$$= \underline{\underline{1 \times 1}}$$

We will just get a matrix $H = (\underline{h_{11}})$

The submatrix to calculate dilated convolution is

$$A'_{11} = \begin{pmatrix} 0.5 & 1 & 1 \\ 1 & 0.5 & 0 \\ 0 & 0 & 0.5 \end{pmatrix} = \begin{pmatrix} a_{11} & a_{14} & a_{17} \\ a_{41} & a_{44} & a_{47} \\ a_{71} & a_{74} & a_{77} \end{pmatrix}$$

$$h_{11} = \begin{matrix} 0.5 \times 1 & + 1 \times 0 & + 1 \times (-1) \\ 1 \times 0 & + 0.5 \times 4 & + 0 \times 0 \\ 0 \times (-1) & + 0 \times 0 & + 0.5 \times (-1) \end{matrix}$$

$$= \underline{\underline{1}}$$

Transposed Convolution

We have an input matrix $A = \begin{pmatrix} 1 & 3 & 2 \\ 2 & 5 & 6 \\ 7 & 1 & 3 \end{pmatrix}$ ✓

$$\begin{aligned} n &= 3 \\ f &= 2 \\ d &= 1 \\ \hline 1 & 10 \end{aligned}$$

We have an input matrix $N = \begin{pmatrix} 1 & 5 & 7 \\ 2 & 5 & 6 \\ 7 & 1 & 3 \end{pmatrix}$ ✓

The kernel $K = \begin{pmatrix} 0 & 1 \\ 3 & 2 \end{pmatrix}$. We will calculate transposed convolution with stride 1 as below.

$$D = 1$$

$$\boxed{m + (p-1) \times D}$$

$$n + (q-1) \times D$$

$$3 + (2-1)$$

$$= 4 \times 4$$

$$A_1 = \begin{pmatrix} 1 \times 0 & 1 \times 1 & 0 & 0 \\ 3 \times 1 & 2 \times 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 3 & 2 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_2 = \begin{pmatrix} 0 & 3 \times 0 & 3 \times 1 & 0 \\ 0 & 3 \times 3 & 3 \times 2 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 3 & 0 \\ 0 & 9 & 6 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_3 = \begin{pmatrix} 0 & 0 & 2 \times 0 & 2 \times 1 \\ 0 & 0 & 2 \times 3 & 2 \times 2 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 2 \\ 0 & 0 & 6 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_4 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 2 \times 0 & 2 \times 1 & 0 & 0 \\ 2 \times 3 & 2 \times 2 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 \\ 6 & 4 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_5 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 5 \times 0 & 5 \times 1 & 0 \\ 0 & 5 \times 3 & 5 \times 2 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 5 & 0 \\ 0 & 15 & 10 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_6 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 6 \times 0 & 6 \times 1 \\ 0 & 0 & 6 \times 3 & 6 \times 2 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 6 \\ 0 & 0 & 18 & 12 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$A_7 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 \times 7 & 0 \times 7 & 0 & 0 \\ 3 \times 7 & 2 \times 7 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 7 & 0 & 0 \\ 21 & 14 & 0 & 0 \end{pmatrix}$$

$$A_8 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 3 & 2 & 0 \end{pmatrix}$$

$$A_9 = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 6 & 6 \end{pmatrix}$$

$$\text{Transposed convolution output} = A_1 + A_2 + \dots + A_9$$

$$= \begin{pmatrix} 0 & 1 & 3 & 2 \\ 3 & 13 & 17 & 10 \\ 6 & 26 & 29 & 15 \\ 21 & 17 & 11 & 6 \end{pmatrix} \quad \checkmark \quad \underline{\underline{4 \times 4}}$$

If we consider a stride of 2, our output matrix will be of shape 6×6 . We get following as an output matrix

0	1	0	3	0	2
3	2	9	6	6	4
0	2	0	5	0	6
6	4	15	10	18	12
0	7	0	1	0	3
21	14	3	2	9	6

6×6

$$8 \times (n-1) + 1$$

$$2(3-1) + 2$$

$$= 6$$

$$d = 2$$

$$A = \begin{pmatrix} 1 & 3 & 2 \\ 2 & 5 & 6 \\ 7 & 1 & 3 \end{pmatrix}$$

$$K = \begin{pmatrix} 0 & 1 \\ 3 & 2 \end{pmatrix}$$