

Convolutional Neural Networks

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A dark blue diagonal gradient bar that starts from the bottom left corner and extends towards the top right corner, covering the lower half of the slide.

Convolutional Neural Networks (CNNs)

Components:

1. Kernels and filters
2. Convolution
3. Strides
4. Pooling
5. Padding
6. Dropout

Kernels & Filters

1. Filters are square matrices with $n_k \times n_k$ dimensions
2. n_k is usually 3 or 5

The following kernel will allow the detection of horizontal edges

$$\mathfrak{J}_H = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ -1 & -1 & -1 \end{pmatrix}$$

Kernels & Filters

The following kernel will allow the detection of vertical edges

$$\mathfrak{I}_V = \begin{pmatrix} 1 & 0 & -1 \\ 1 & 0 & -1 \\ 1 & 0 & -1 \end{pmatrix}$$

The following kernel will allow the detection of edges when luminosity changes drastically

$$\mathfrak{I}_L = \begin{pmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{pmatrix}$$

Kernels & Filters

The following kernel will blur edges in an image

$$\mathfrak{J}_B = -\frac{1}{9} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

Convolution

1. Convolution is multiplying corresponding positional values in two tensors and summing them up.
2. Convolution is not tensor or matrix multiplication

$$\begin{pmatrix} a_1 & a_2 & a_3 \\ a_4 & a_5 & a_6 \\ a_7 & a_8 & a_9 \end{pmatrix} * \begin{pmatrix} k_1 & k_2 & k_3 \\ k_4 & k_5 & k_6 \\ k_7 & k_8 & k_9 \end{pmatrix} = \sum_{i=1}^9 a_i k_i$$

Convolution: Illustration

- Assume matrix A is 4x4 & Kernel K is 3x3 matrix
- Convolution of A*K results in matrix B (2x2 matrix)

$$A = \begin{pmatrix} a_1 & a_2 & a_3 & a_4 \\ a_5 & a_6 & a_7 & a_8 \\ a_9 & a_{10} & a_{11} & a_{12} \\ a_{13} & a_{14} & a_{15} & a_{16} \end{pmatrix} \quad K = \begin{pmatrix} k_1 & k_2 & k_3 \\ k_4 & k_5 & k_6 \\ k_7 & k_8 & k_9 \end{pmatrix} \quad B = \begin{pmatrix} B_1 & B_2 \\ B_3 & B_4 \end{pmatrix}$$

Convolution: Illustration

- First Convolution of $A * K$

$$A = \begin{pmatrix} a_1 & a_2 & a_3 & a_4 \\ a_5 & a_6 & a_7 & a_8 \\ a_9 & a_{10} & a_{11} & a_{12} \\ a_{13} & a_{14} & a_{15} & a_{16} \end{pmatrix} \quad A_1 = \begin{pmatrix} a_1 & a_2 & a_3 \\ a_5 & a_6 & a_7 \\ a_9 & a_{10} & a_{11} \end{pmatrix} \quad K = \begin{pmatrix} k_1 & k_2 & k_3 \\ k_4 & k_5 & k_6 \\ k_7 & k_8 & k_9 \end{pmatrix} \quad B = \begin{pmatrix} B_1 & B_2 \\ B_3 & B_4 \end{pmatrix}$$

$$B_1 = A_1 * K = a_1 k_1 + a_2 k_2 + a_3 k_3 + k_4 a_5 + k_5 a_5 + k_6 a_7 + k_7 a_9 + k_8 a_{10} + k_9 a_{11}$$

Convolution: Illustration

- Second Convolution of A*K

$$A = \begin{pmatrix} a_1 & a_2 & a_3 & a_4 \\ a_5 & a_6 & a_7 & a_8 \\ a_9 & a_{10} & a_{11} & a_{12} \\ a_{13} & a_{14} & a_{15} & a_{16} \end{pmatrix} \quad K = \begin{pmatrix} k_1 & k_2 & k_3 \\ k_4 & k_5 & k_6 \\ k_7 & k_8 & k_9 \end{pmatrix} \quad B = \begin{pmatrix} B_1 & B_2 \\ B_3 & B_4 \end{pmatrix}$$

$$B_2 = A_2 * K = a_2 k_1 + a_3 k_2 + a_4 k_3 + a_6 k_4 + a_7 k_5 + a_8 k_6 + a_{10} k_7 + a_{11} k_8 + a_{12} k_9$$

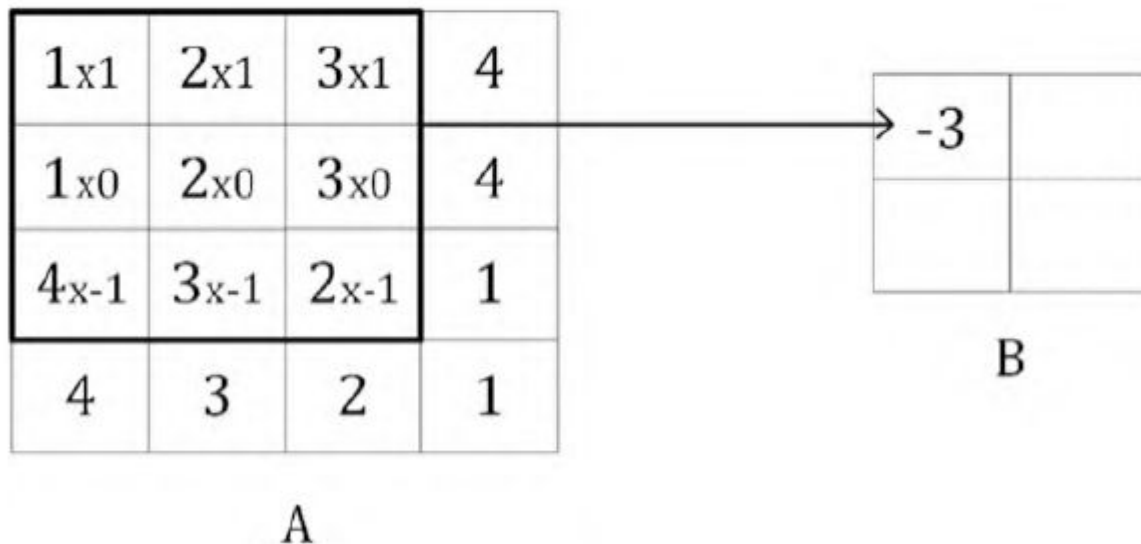
Stride

- **Stride:** The number of rows to move down and columns to move to right when performing convolution between a tensor and the kernel
- In the previous example between $A * K$, the 3×3 kernel region was moved one column to the right and one row down.
 - In this example , the stride is 1 ($s=1$).
- Depending on S , the new matrix created after convolution will have the following shape:

$$n_B = \frac{n_A - n_K}{s} + 1$$

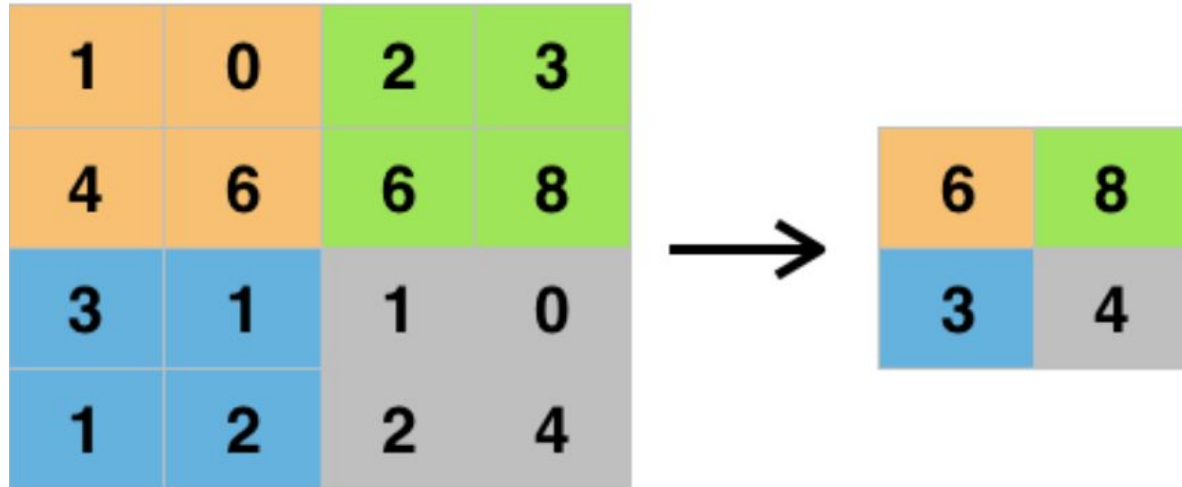
Convolution: Visual Illustration

$$B_{11} = 1 \times 1 + 2 \times 1 + 3 \times 1 + 1 \times 0 + 2 \times 0 + 3 \times 0 + 4 \times (-1) + 3 \times (-1) + 2 \times (-1) = -3$$



Pooling

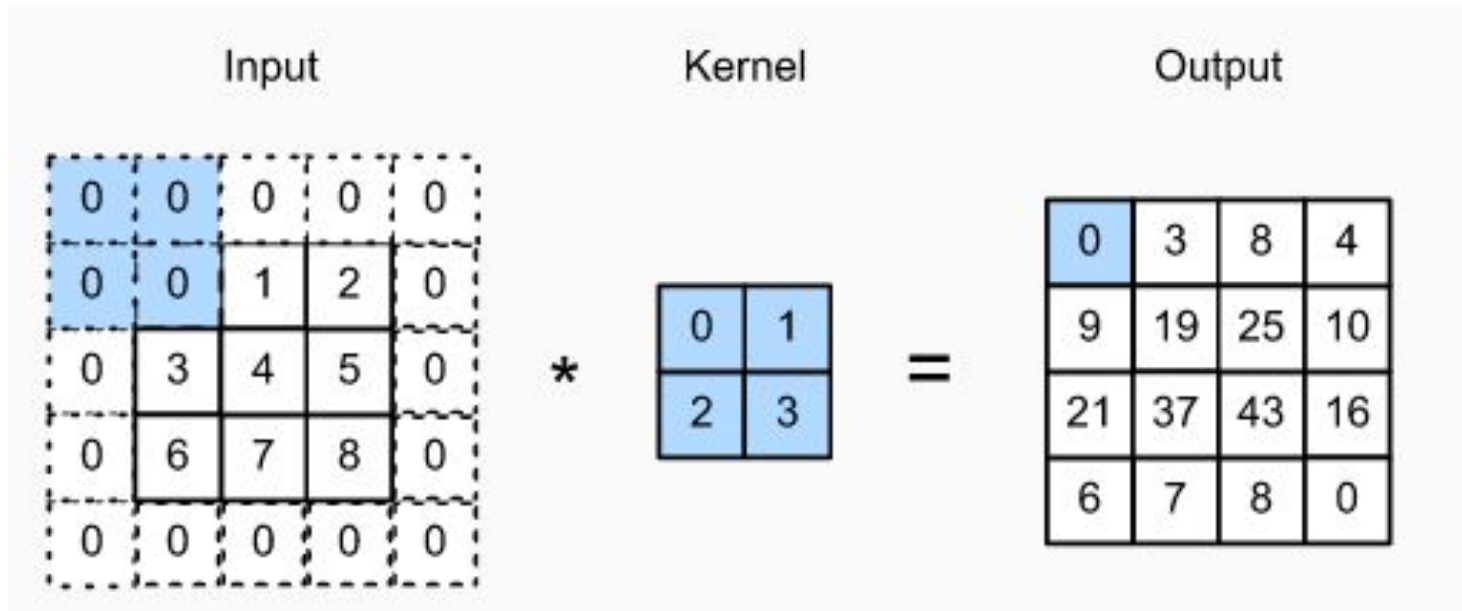
- **Pooling**: From a given matrix, stride and kernel, identify the maximum or average in that block
- Most often used pooling technique is **max pooling**



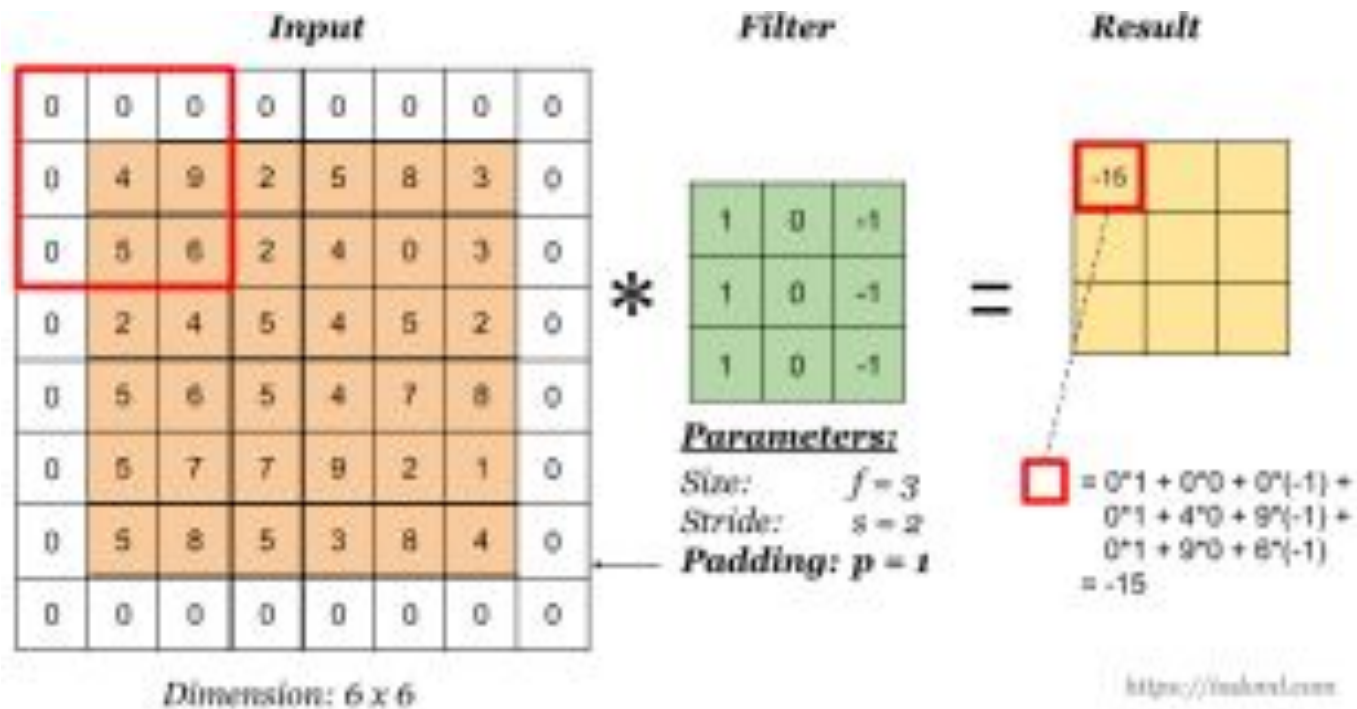
Padding

- **Padding**
- Add rows of pixels to the top and bottom and to the columns of pixels on the right and on the left of the final images so that resulting matrix after convolution with kernel is same as original image matrix
- Some strategies for padding are to fill the added pixels with zeros, with the values of the closest pixels, and so on.

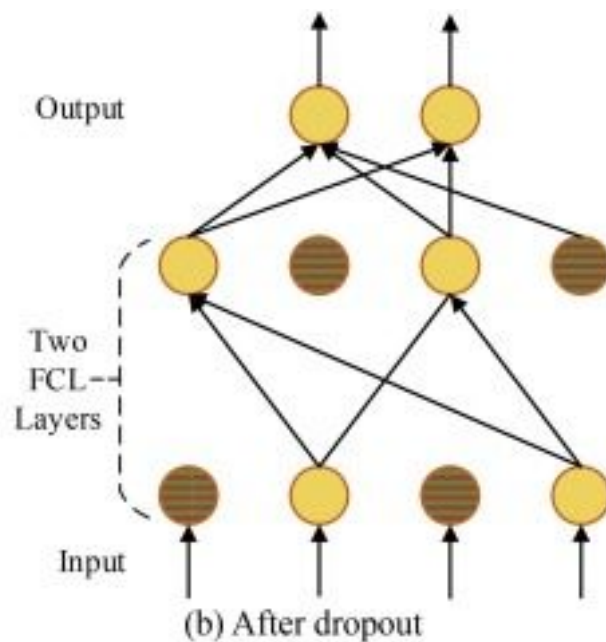
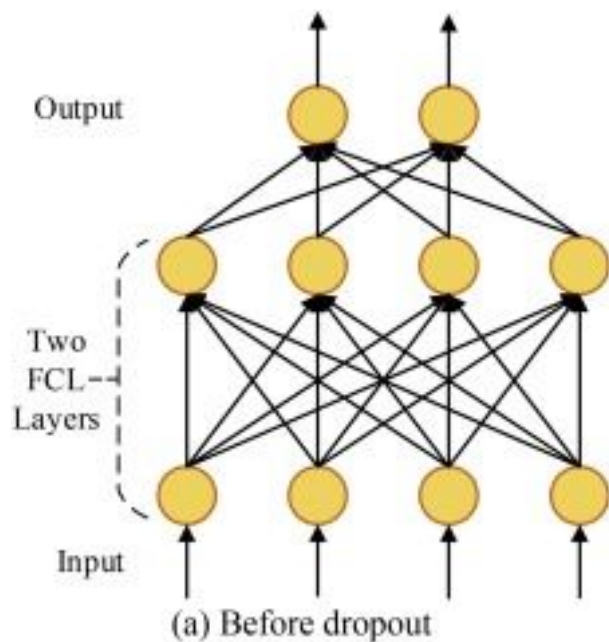
Padding: Illustration



Padding: Illustration



Dropout: Illustration



Convolutional Neural Network: Illustration

