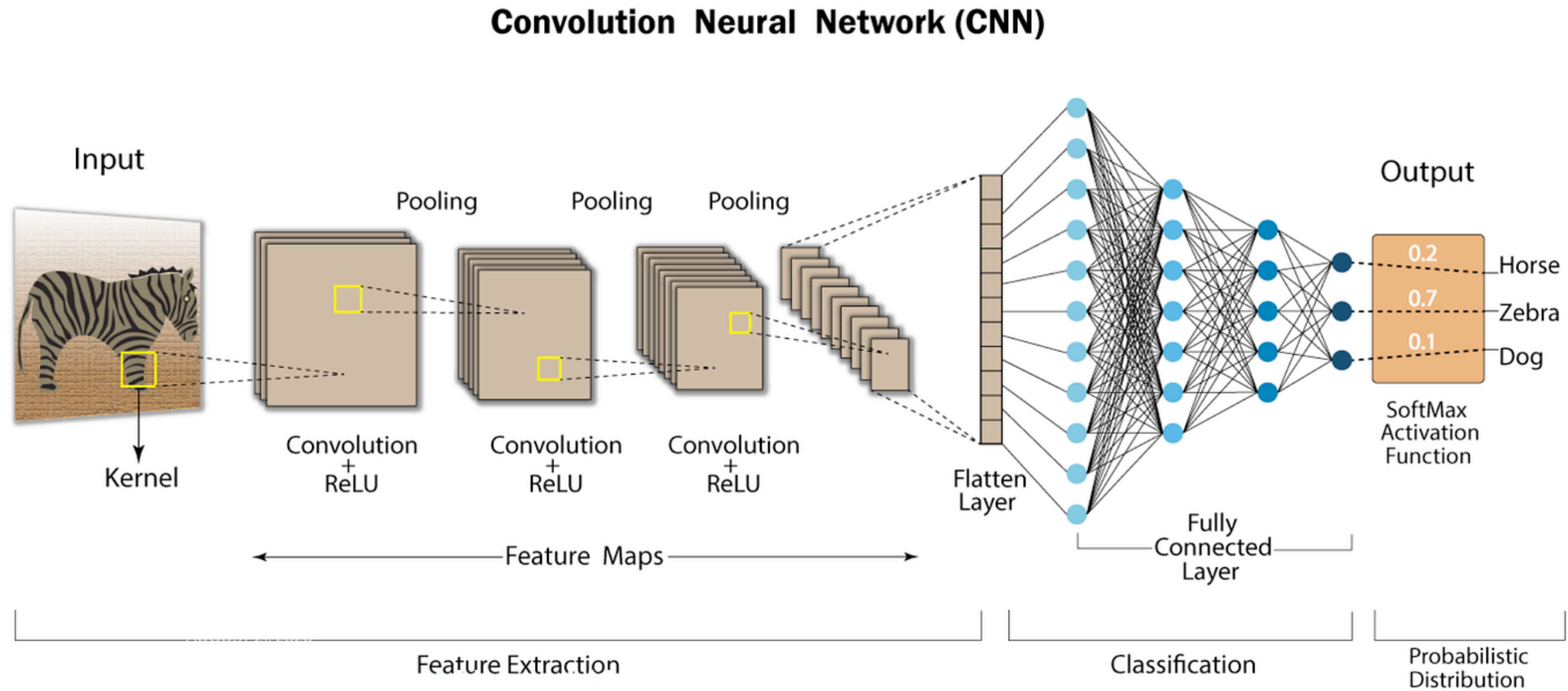


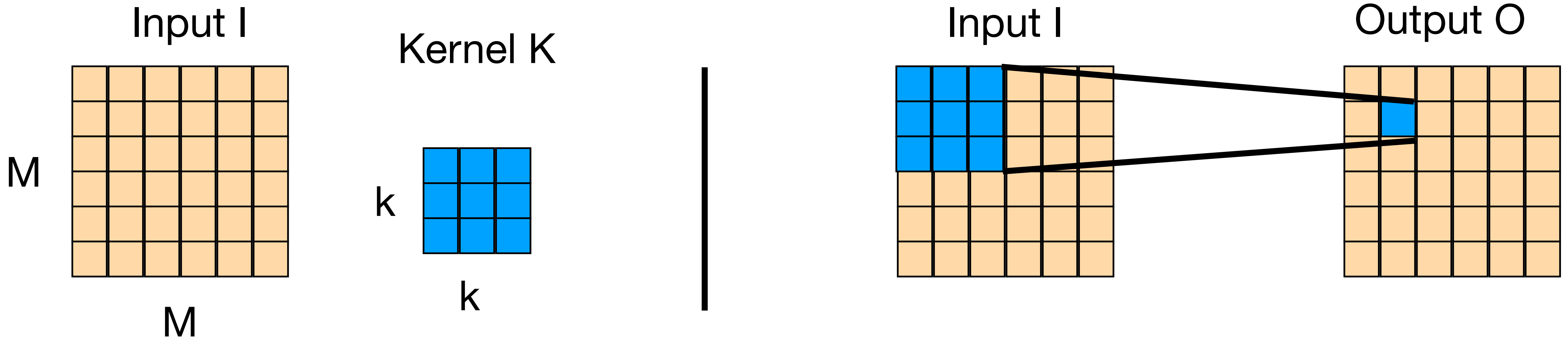
# CNN recap

## CNNs for classification - general structure



# CNN recap

1 input channel 1 output channel



$$O_{i,j} = \sum_{n=0, m=0}^{k-1, k-1} K_{n,m} I_{i-k//2+n, j-k//2+m}$$

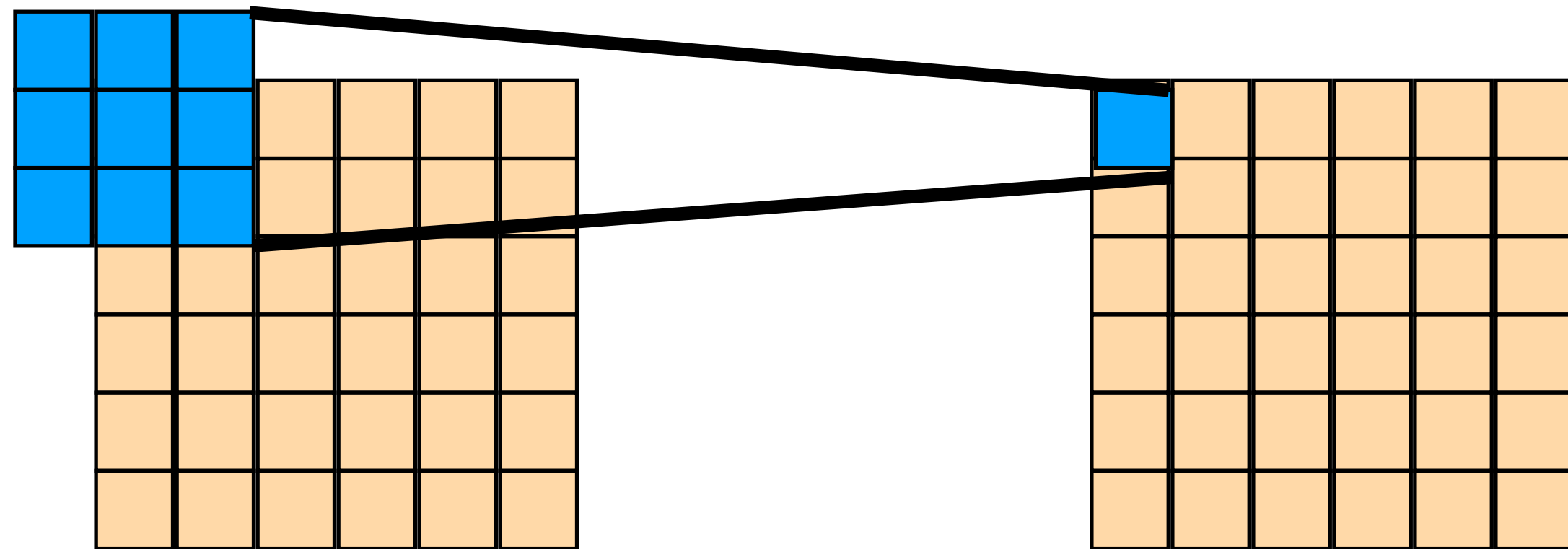
# CNN recap

## Different paddings

Zero padding

Input I

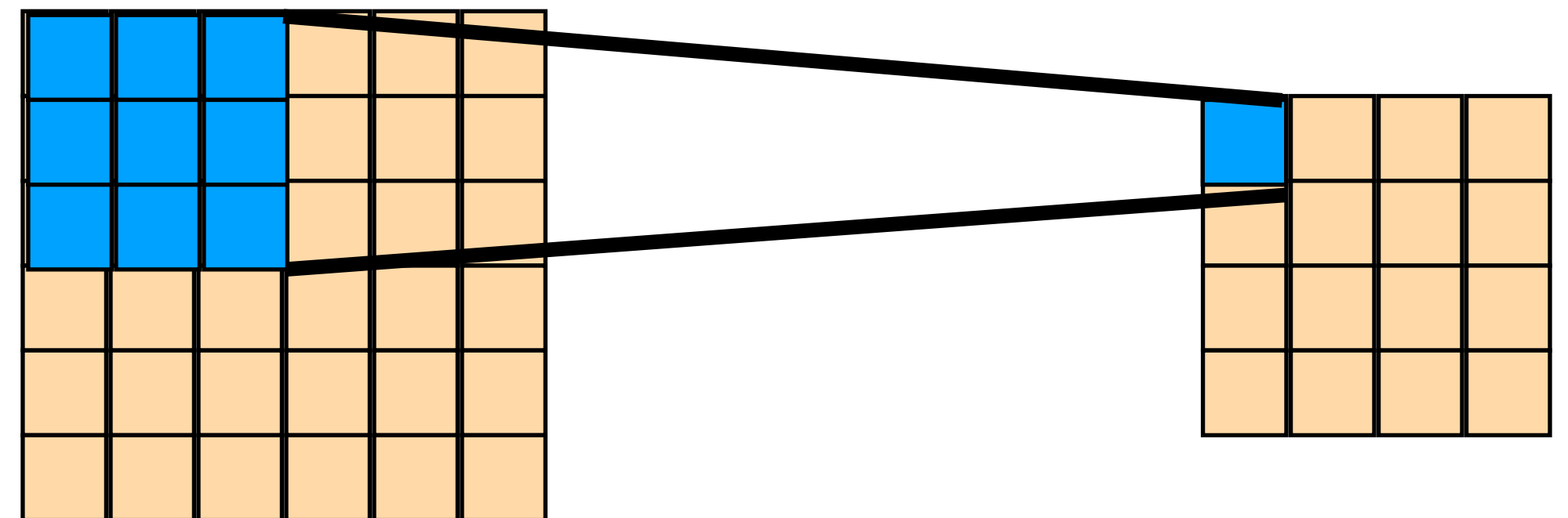
Output O



No padding

Input I

Output O



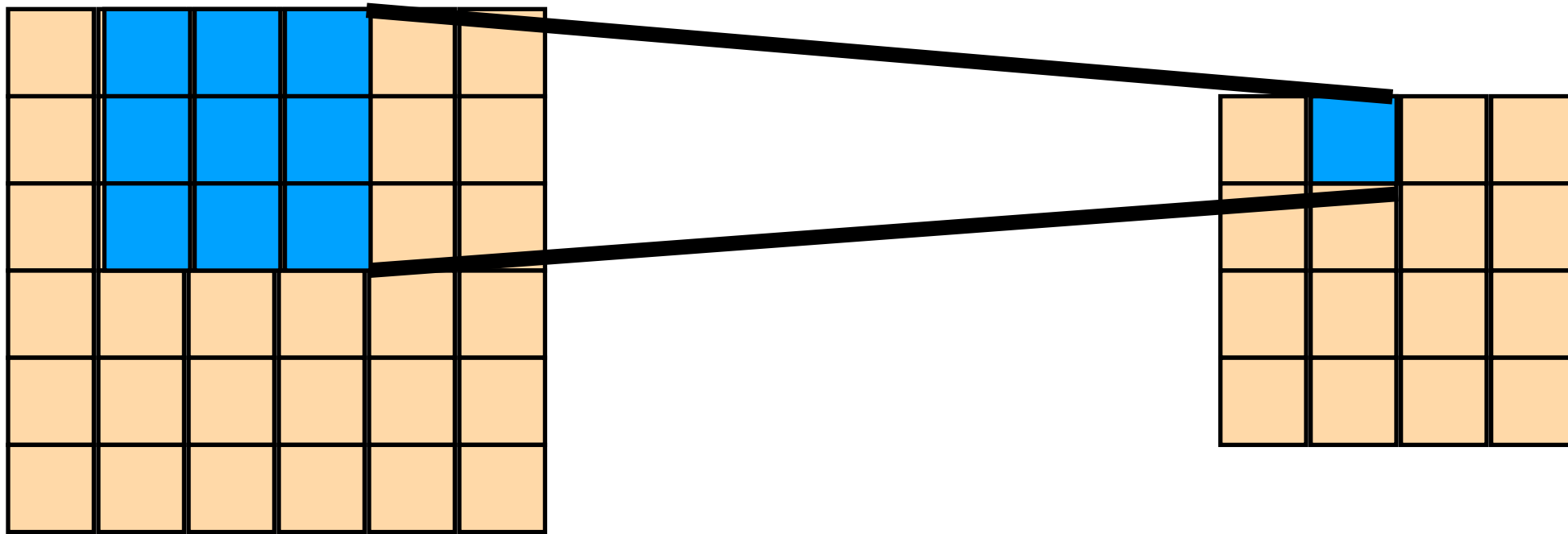
# CNN recap

No padding

Different strides

Input I

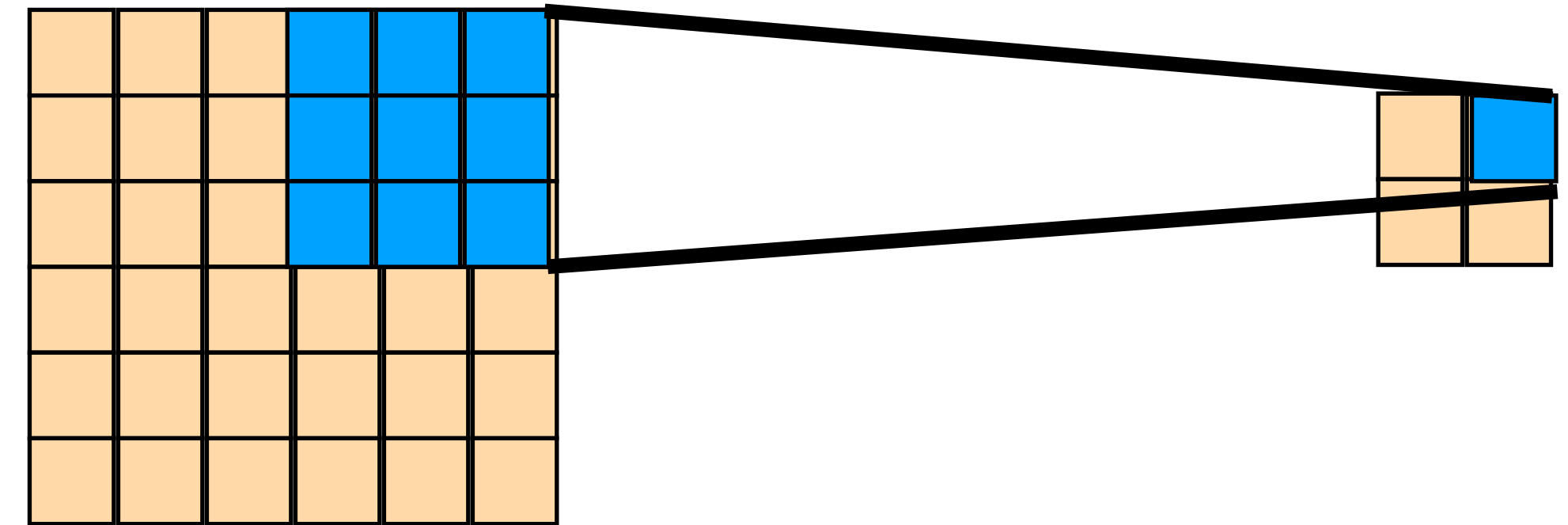
Output O



Stride = 1

Input I

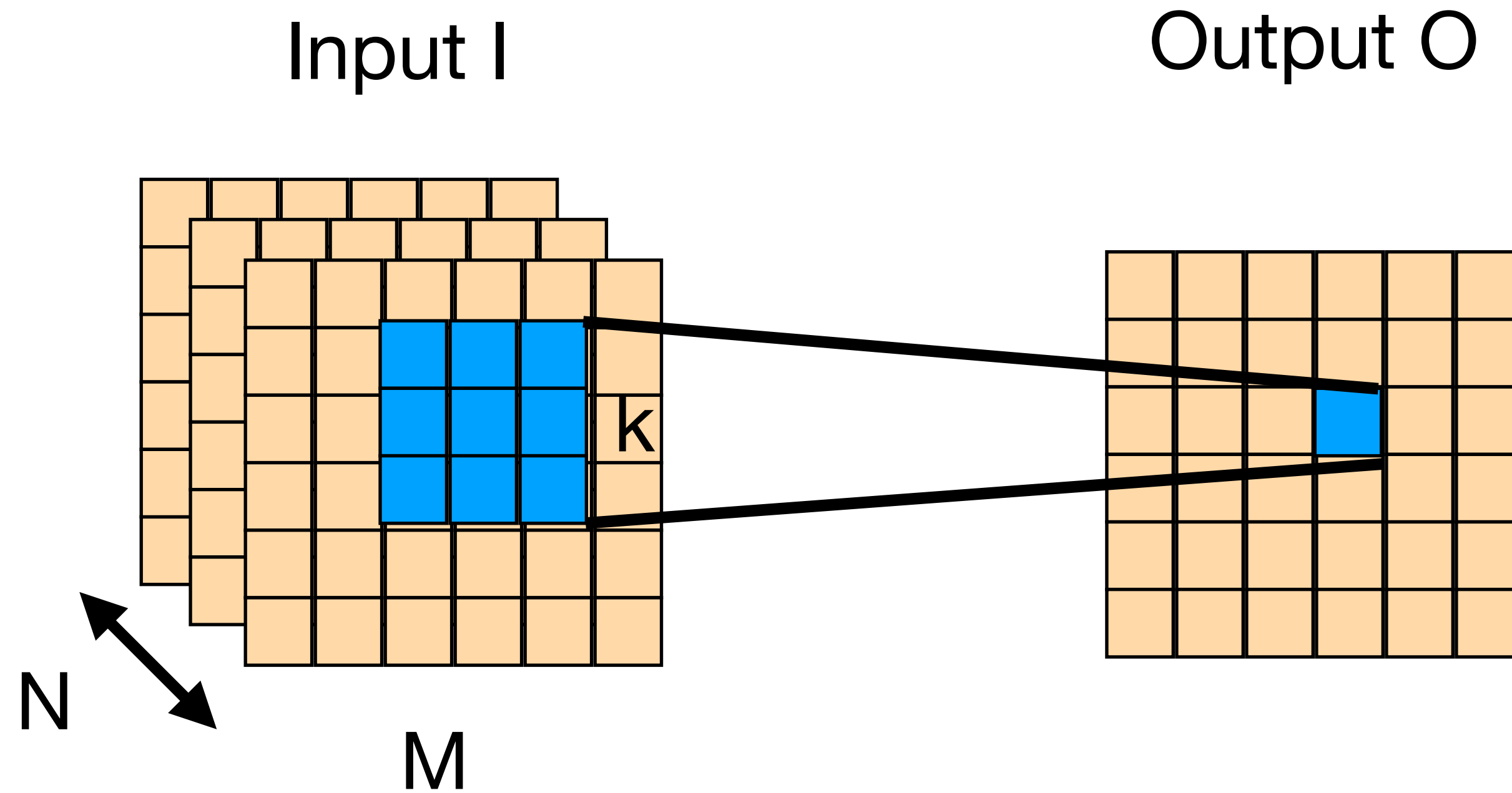
Output O



Stride = 3

# CNN recap

Multiple channels: N input channels 1 output channel



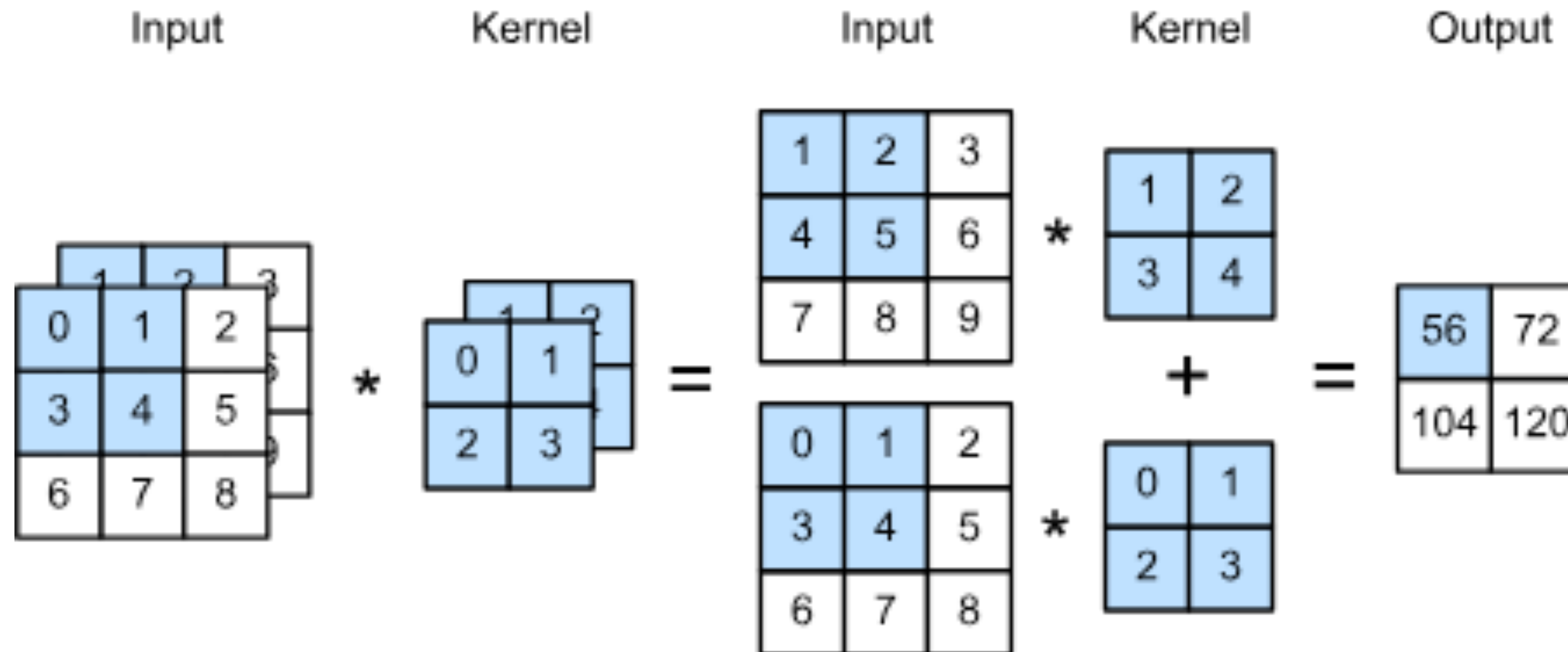
$I_{abc}$  is pixel value of channel a at position (b,c)

$$O_{i,j} = \sum_{c=0}^{N-1} \sum_{n=0, m=0}^{k-1, k-1} K_{c,n,m} I_{c,i-k//2+n, j-k//2+m}$$

# CNN recap

Multiple channels: N input channels 1 output channel

Example with two channels



# CNN recap

Multiple channels: N input channels L output channel

Just take L independent kernels (indexed by  $l$ )

$$O_{l,i,j} = \sum_{c=0}^{N-1} \sum_{n=0, m=0}^{k-1, k-1} K_{l,c,n,m} I_{c,i-k//2+n, j-k//2+m}$$

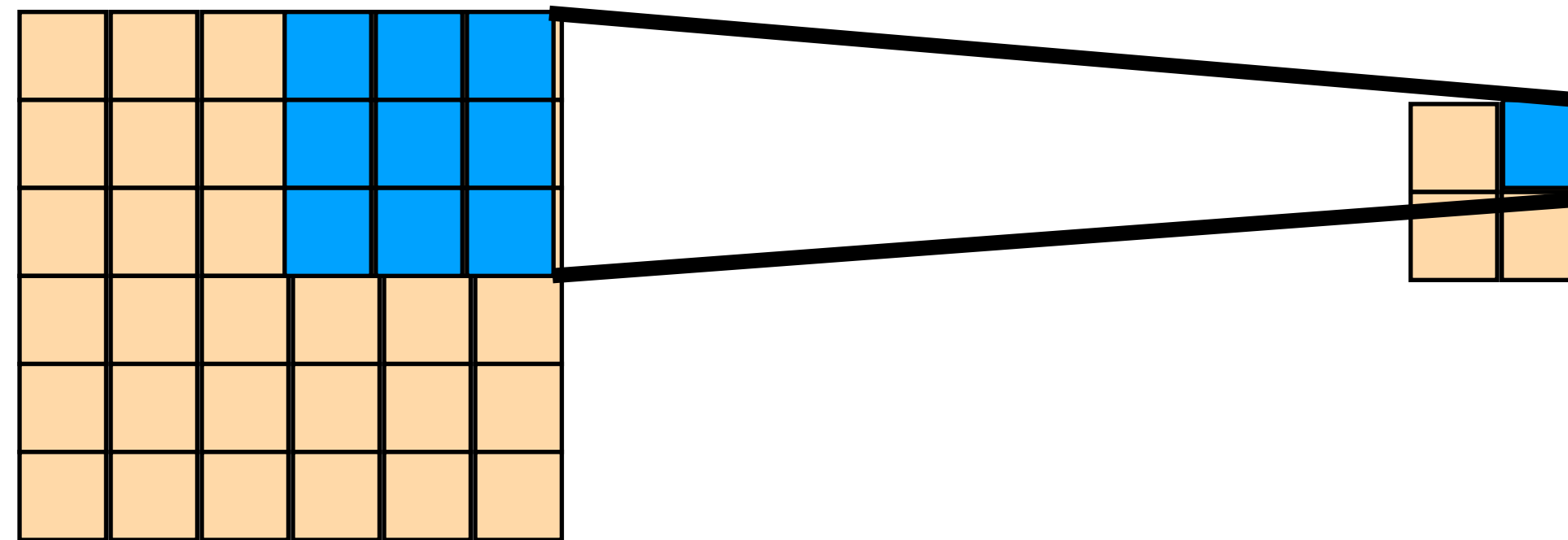
# CNN recap

## Downsampling

Operation e.g. max or average

Input I

Output O

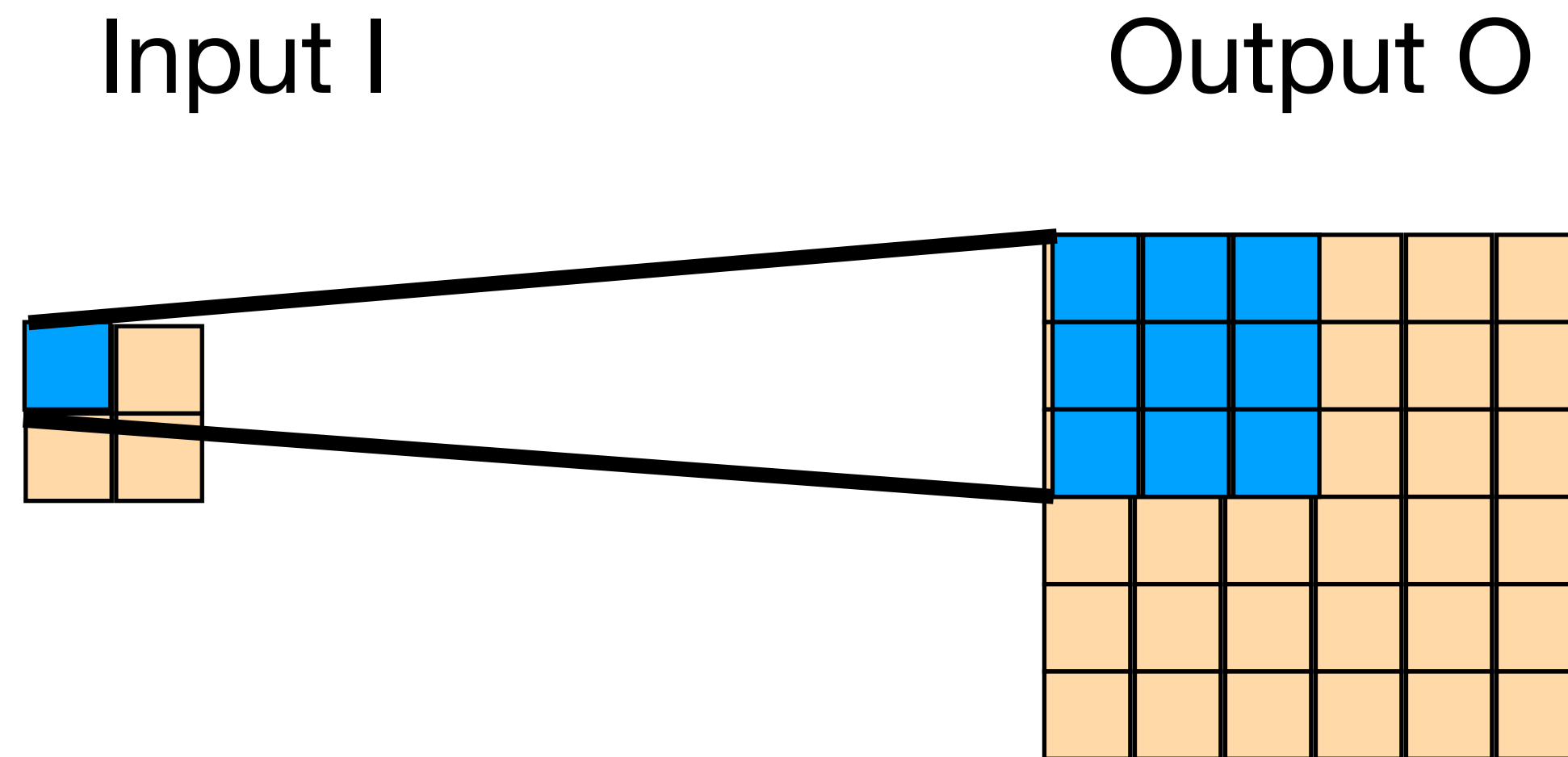


Stride = 3



# CNN recap

## Upsampling



E.g. All values equal or transposed convolution

# Categorical cross-entropy

$$C = - \sum_j y_j^{\text{target}} \ln y_j^{\text{out}}$$

$j$  goes over all possible labels