

# CNNs (Convolutional Neural Networks) Convolutions.

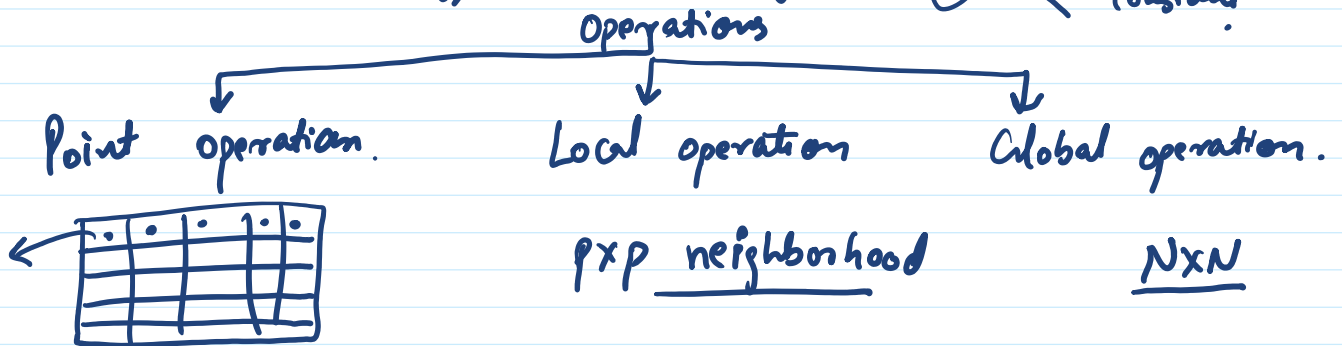
⇒ Image is represented as a matrix.

Gray-scale image 0 - 255 intensity values.

As a function:-  $(x, y) \leftarrow I$

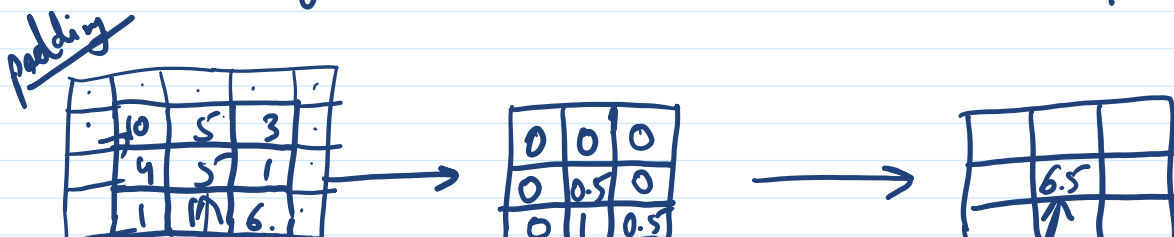
A digital image:- discrete (quantized version of this function)

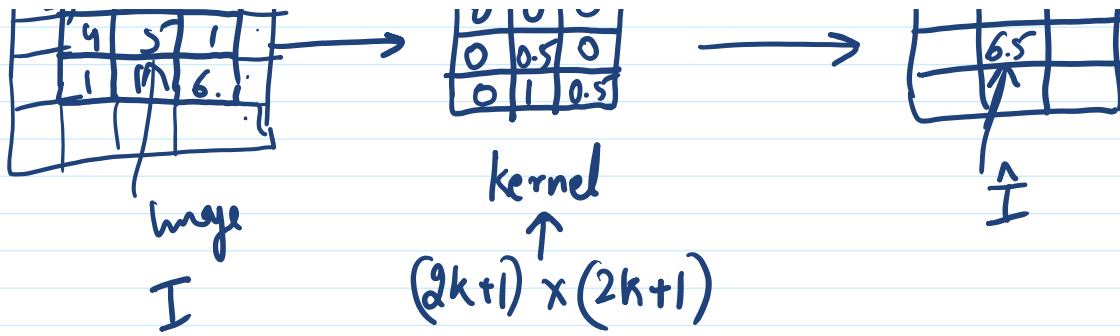
$$\hat{I}(x, y) = I(x, y) + \text{a} \leftarrow \text{constant}$$



Let  $I$  be the image with  $N \times M$  pixels,  
 $I_{max}$  be the max. intensity value (255)

filter  $\rightarrow$   $M \times N$



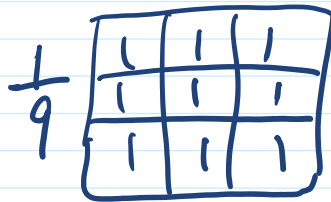


Linear Filter:-

Replacing each pixel by weighted sum of neighbors.

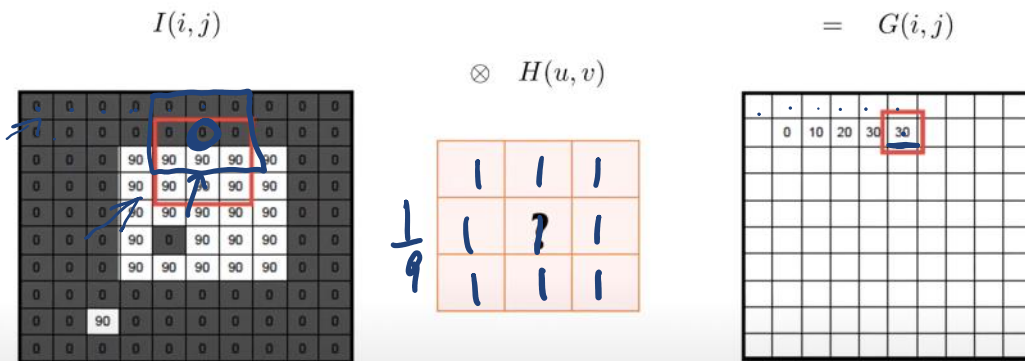
kernel / mask / filter

Average filter / Box filter:-



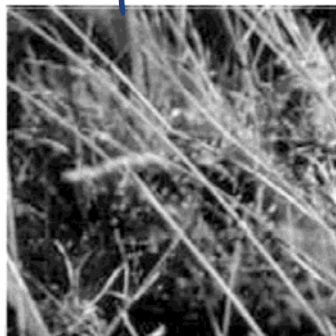
$$\frac{1}{9} (9 \cdot 0 + 9 \cdot 0 + 9 \cdot 0)$$

$$= G(i, j)$$

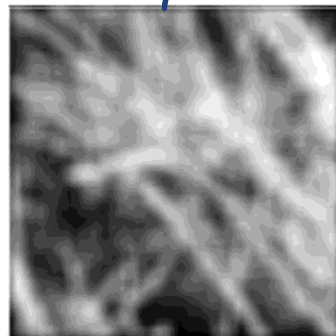


Credit: K Grauman, Univ of Texas Austin

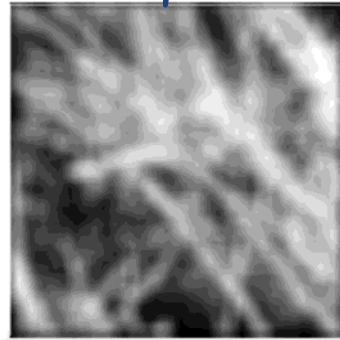
Input



Output



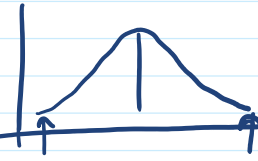
Credit: K Grauman, Univ of Texas Austin



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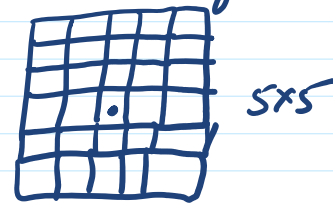
### Gaussian Average Filter :-

$$h(u,v) = \frac{1}{2\pi\sigma^2} \exp\left(\frac{-u^2-v^2}{2\sigma^2}\right)$$



$$\frac{1}{16} \begin{bmatrix} 1/4 & 2/4 & 1/4 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$$

2D gaussian filter.



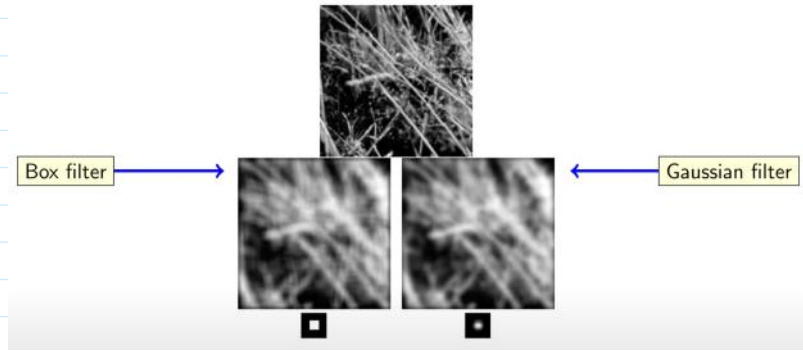
$I(i,j)$

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	90	0	90	90	90	0	0
0	0	0	90	90	90	90	90	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

$\otimes H(u,v)$

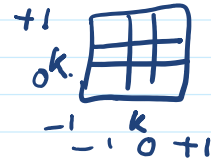
$$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix} =$$

Credit: K Grauman, Univ of Texas Austin



Given a kernel size  $(2k+1) \times (2k+1)$ :

$\Rightarrow$  cross-correlation:-



$$G(i, j) = \frac{1}{(2k+1)^2} \sum_{u=-k}^k \sum_{v=-k}^k I(i+u, j+v)$$

uniform weights  
to each pixel

$I(i, j)$   
Loop over pixels  
in considered  
neighbourhood around  
 $I(i, j)$

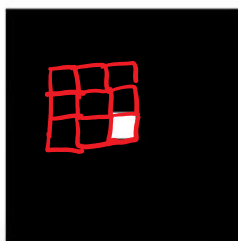
$$G(i, j) = \sum_{u=-k}^k \sum_{v=-k}^k H(u, v) I(i+u, j+v)$$

$$G = H \otimes I$$

non-uniform  
weights / filter

dot product b/w local neighborhood  
and kernel for each pixel  $H(u, v)$

Input image  
 $I(i, j)$

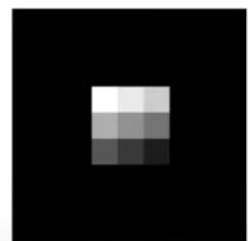


$\otimes H(u, v)$



a	b	c
d	e	f
g	h	i

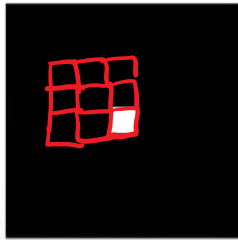
$G(i, j)$



1	4	5
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Imp

$I(i, j)$

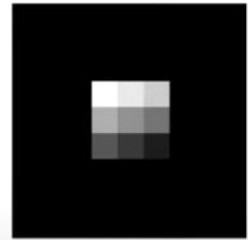


$\otimes H(u, v)$



a	b	c
d	e	f
g	h	i

$G(i, j)$



l	q	o
j	e	p
c	q	e

(cross-correlation does not maintain an identity with an impulse function.

Convolution :-

$$G(i, j) = \sum_{u=-k}^k \sum_{v=-k}^k H(u, v) \underbrace{I(i-u, j-v)}$$

double flipping the filter so that output turns out to be the filter itself.

$\otimes$  ← correlation.

$$G = H * I$$

↑  
convolution.

(cross-correlation is sliding a filter across an image.

convolution is sliding a flipped filter across an image.