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# Filtering (II)

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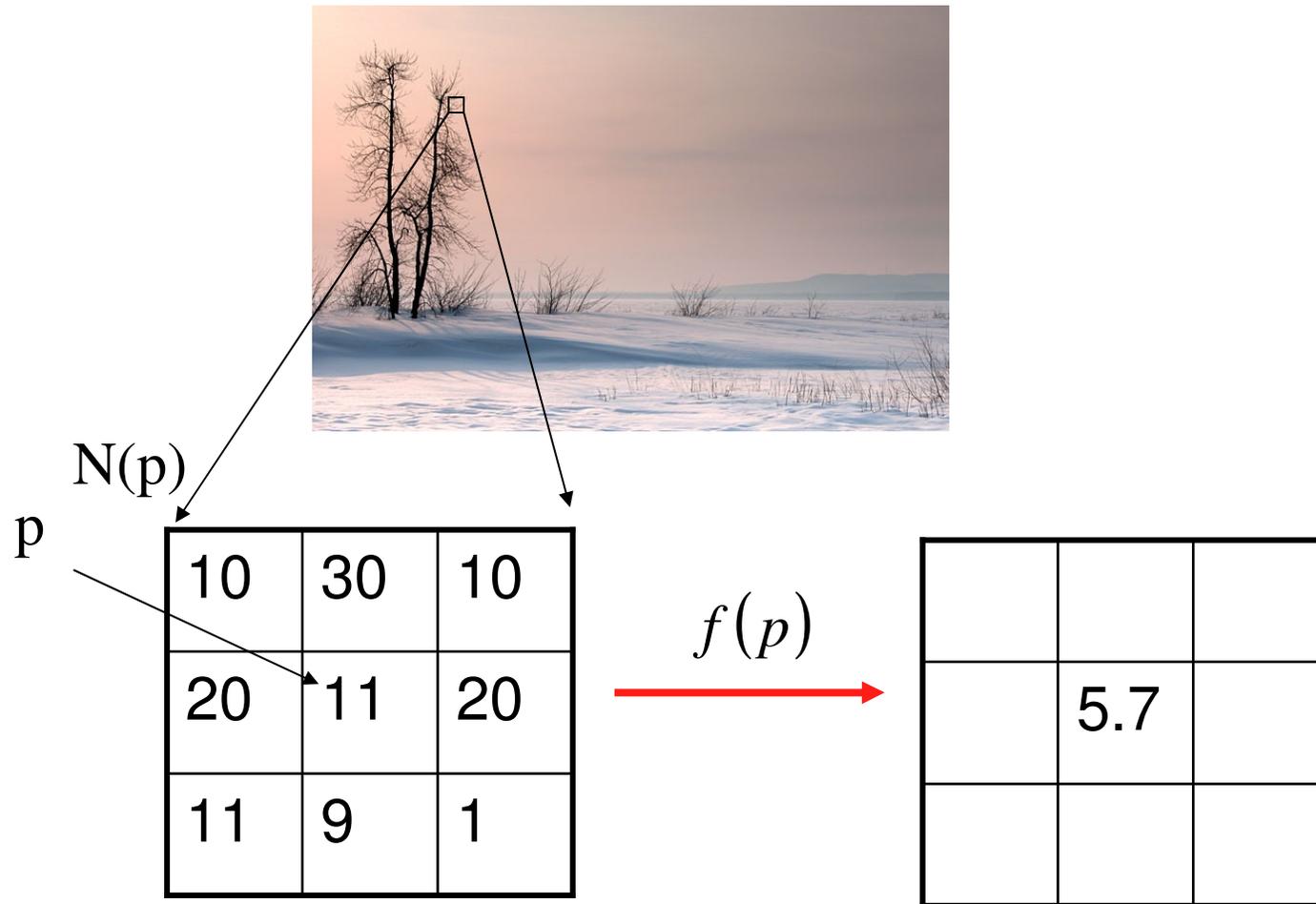
COMP 4900C

Winter 2008

# Image Filtering

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Modifying the pixels in an image based on some functions of a local neighbourhood of the pixels



# Linear Filtering – convolution

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The output is the linear combination of the neighbourhood pixels

$$I_A(i, j) = I * A = \sum_{h=-m/2}^{m/2} \sum_{k=-m/2}^{m/2} A(h, k) I(i - h, j - k)$$

The coefficients come from a constant matrix A, called kernel. This process, denoted by ‘\*’, is called (discrete) convolution.

1	3	0
2	10	2
4	1	1

Image

\*

1	0	-1
1	0.1	-1
1	0	-1

Kernel

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	5	

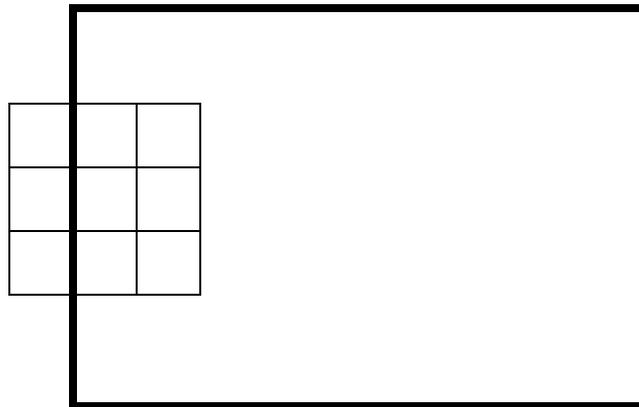
Filter Output

# Handle Border Pixels

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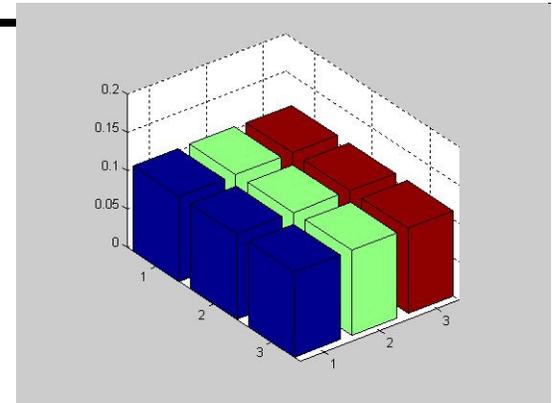
Near the borders of the image, some pixels do not have enough neighbours. Two possible solutions are:

- Set the value of all non-included pixels to zero.
- Set all non-included pixels to the value of the corresponding pixel in the input image.



# Smoothing by Averaging

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$$* \frac{1}{9}$$

1	1	1
1	1	1
1	1	1

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Convolution can be understood as weighted averaging.

# Gaussian Filter

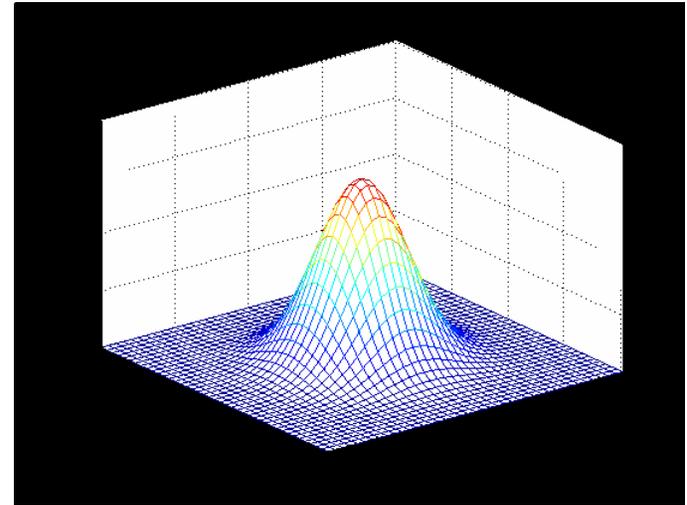
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$$G_{\sigma}(x, y) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{(x^2 + y^2)}{2\sigma^2}\right)$$

Discrete Gaussian kernel:

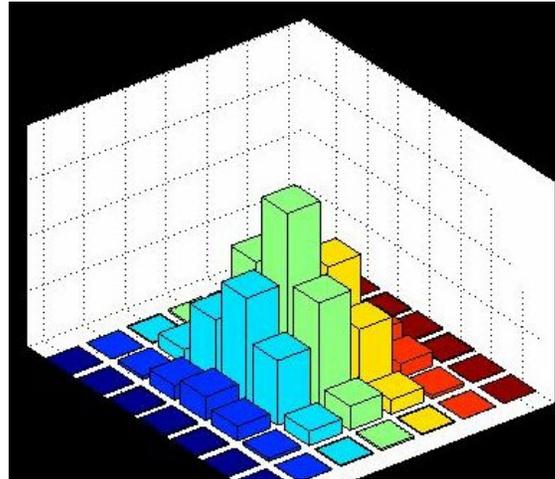
$$G(h, k) = \frac{1}{2\pi\sigma^2} e^{-\frac{h^2+k^2}{2\sigma^2}}$$

where  $G(h, k)$  is an element of an  $m \times m$  array



# Gaussian Filter

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$$* \frac{1}{273}$$

1	4	7	4	1
4	16	26	16	4
7	26	41	26	7
4	16	26	16	4
1	4	7	4	1

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$$\sigma = 1$$

# Gaussian Kernel is Separable

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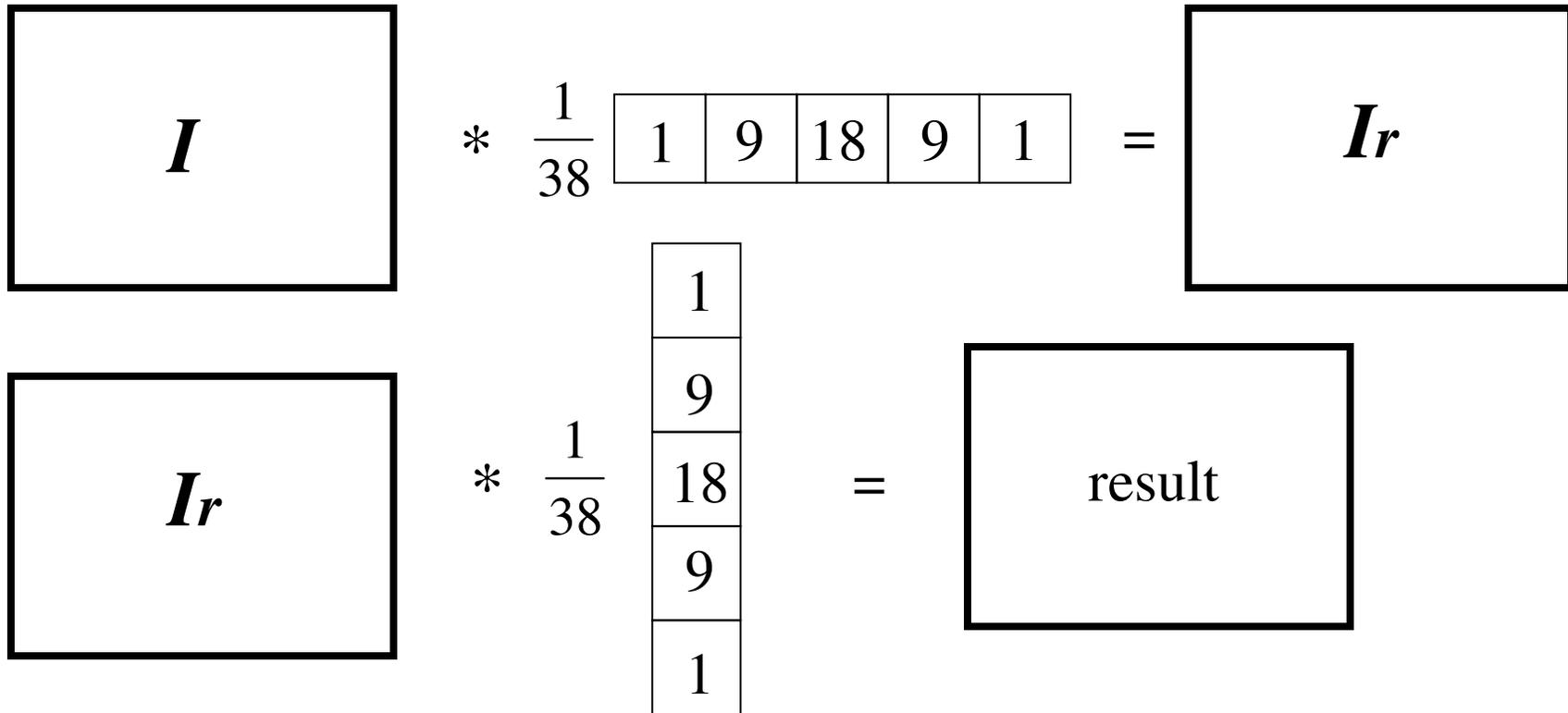
$$\begin{aligned} I_G &= I * G = \\ &= \sum_{h=-m/2}^{m/2} \sum_{k=-m/2}^{m/2} G(h, k) I(i-h, j-k) = \\ &= \sum_{h=-m/2}^{m/2} \sum_{k=-m/2}^{m/2} e^{-\frac{h^2+k^2}{2\sigma^2}} I(i-h, j-k) = \\ &= \sum_{h=-m/2}^{m/2} e^{-\frac{h^2}{2\sigma^2}} \sum_{k=-m/2}^{m/2} e^{-\frac{k^2}{2\sigma^2}} I(i-h, j-k) \end{aligned}$$

since 
$$e^{-\frac{h^2+k^2}{2\sigma^2}} = e^{-\frac{h^2}{2\sigma^2}} e^{-\frac{k^2}{2\sigma^2}}$$

# Gaussian Kernel is Separable

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Convolving rows and then columns with a 1-D Gaussian kernel.



The complexity increases linearly with  $m$  instead of with  $m^2$ .

# Gaussian vs. Average

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Gaussian Smoothing



Smoothing by Averaging

# Noise Filtering

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After Averaging



Gaussian Noise



After Gaussian Smoothing

# Noise Filtering

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Salt-and-pepper noise



After averaging



After Gaussian smoothing

# Nonlinear Filtering – median filter

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Replace each pixel value  $I(i, j)$  with the median of the values found in a local neighbourhood of  $(i, j)$ .

123	125	126	130	140
122	124	126	127	135
118	120	150	125	134
119	115	119	123	133
111	116	110	120	130

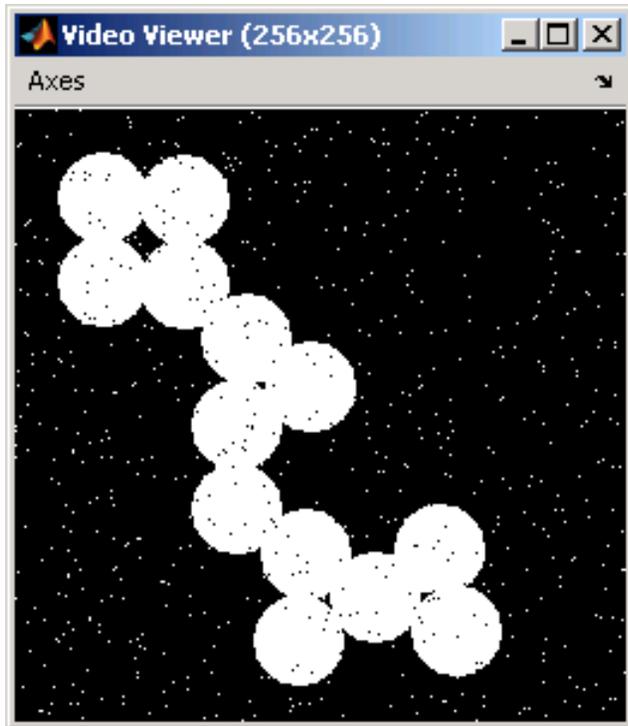
Neighbourhood values:

115, 119, 120, 123, 124,  
125, 126, 127, 150

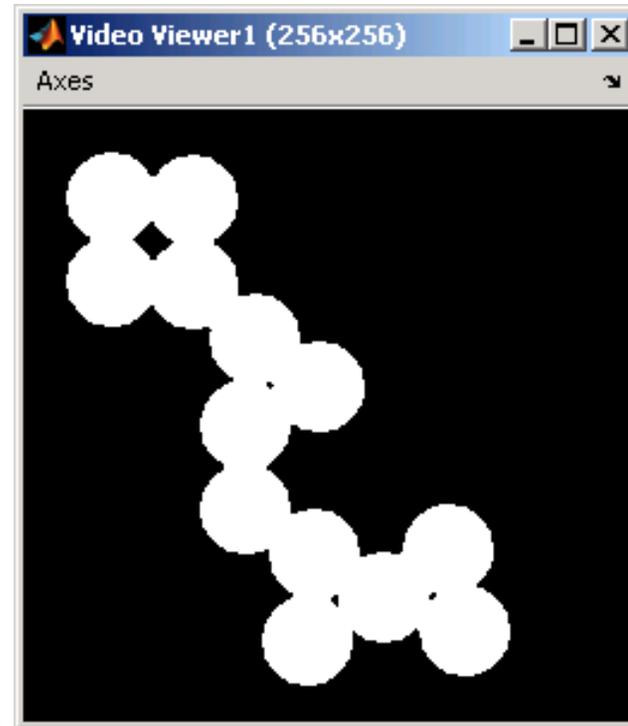
Median value: 124

# Median Filter

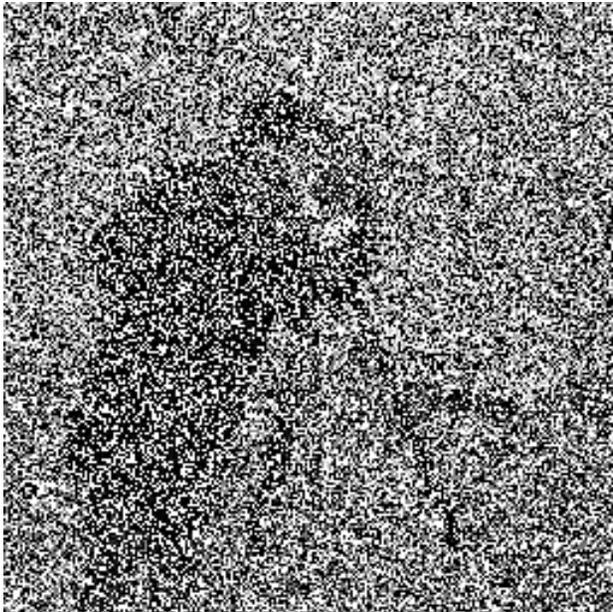
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Salt-and-pepper noise



After median filtering



**[Salt-and-Pepper Noise Removal by Median-type Noise Detectors and Edge-preserving Regularization](#)**

**Raymond H. Chan, Chung-Wa Ho, and Mila Nikolova**

*IEEE Transactions on Image Processing, 14 (2005), 1479-1485.*