

# Computational Photography

- \* Study the basics of computation and its impact on the entire workflow of photography, from capturing, manipulating and collaborating on, and sharing photographs.

# Image Processing and Filtering, via Convolution and Cross-Correlation

- \* Point-process and Neighboring Pixels  
Computations on an Image using  
Cross-Correlation and Convolution



## Lesson Objectives

1. Cross-Correlation
2. Convolution
3. Difference between Cross-Correlation and Convolution
4. Properties of these methods!



Recall A Mathematical Representation for Smoothing

$$G[3, 3] = \frac{1}{9}(A + B + C + D + E + F + G + H + I)$$

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

a	b	c
d	e	f
g	h	i

20	20	10	20	10	20	10	10	13
30	0	0	0	0	0	0	0	30
20	0	A	B	C	90	90	0	20
20	0	D	E	F	90	90	0	20
10	0	G	H	I	90	90	0	10
10	0	90	90	90	90	90	0	10
10	0	90	90	90	90	90	0	10
20	0	0	0	0	0	0	0	20
20	20	10	20	10	20	10	10	13

More Generally,

$$G[3, 3] = a * A + b * B + c * C + d * D + e * E + f * F + g * G + h * H + i * I$$

$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] F[i + u, j + v]$$

Referred to as Cross-correlation, which we will cover later

# Cross-Correlation Method

- ★ In signal processing, cross-correlation is a measure of similarity of two waveforms as a function of a time-lag applied to one of them.
- ★ Also known as a sliding dot product or sliding inner-product.

# Cross-Correlation Method

$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] F[i + u, j + v]$$

Denoted by

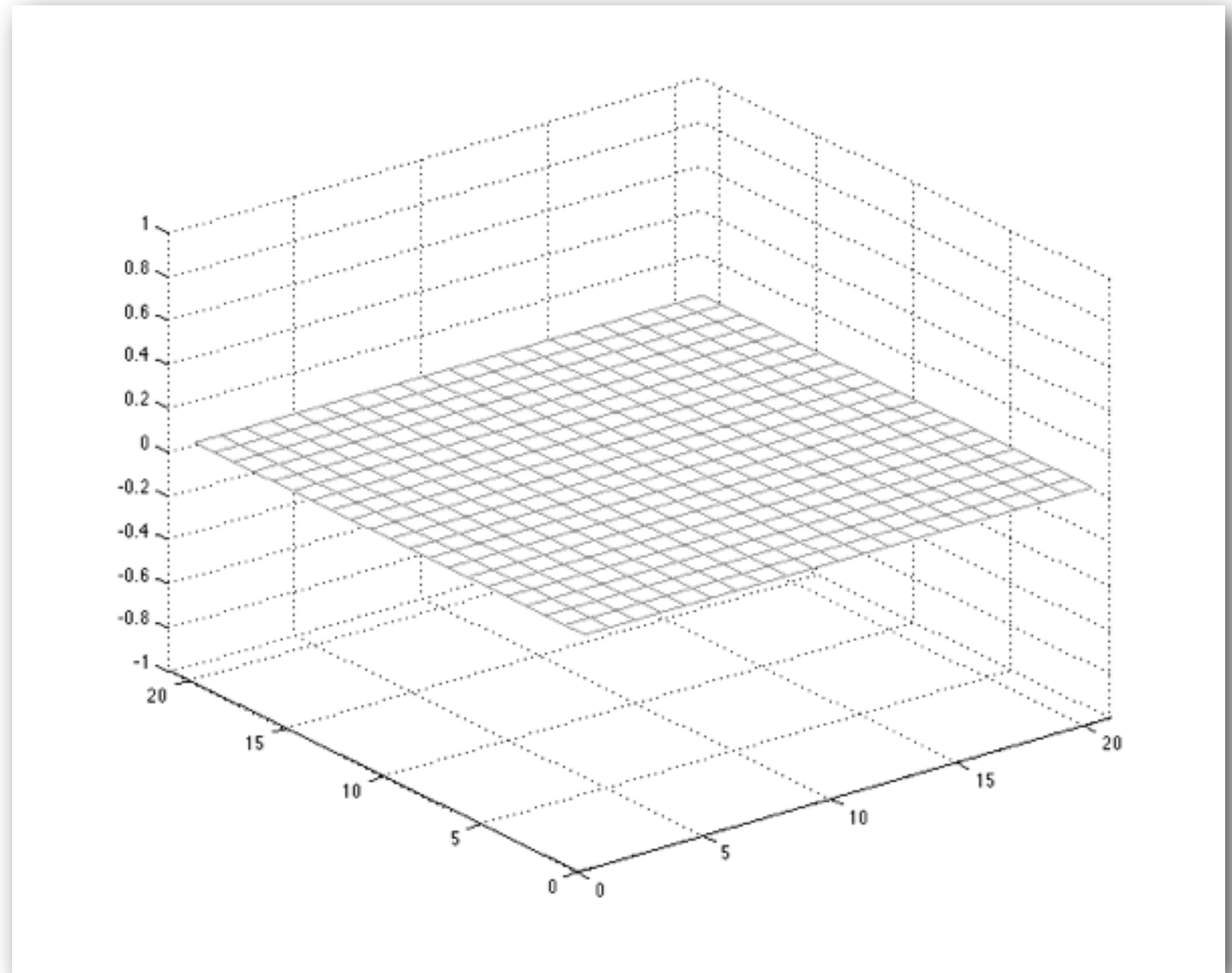
$$G = h \otimes F$$

- \* Filtering an image:
- \* Replace each pixel with a linear combination of its neighbors
- \* Filter "kernel" or "mask"
- \*  $h[u, v]$  is the prescription for weights in the linear combination

# Example: Box Filter

## Box/Average Filter

- Size:  $2 \times 2$
- Values: Uniform

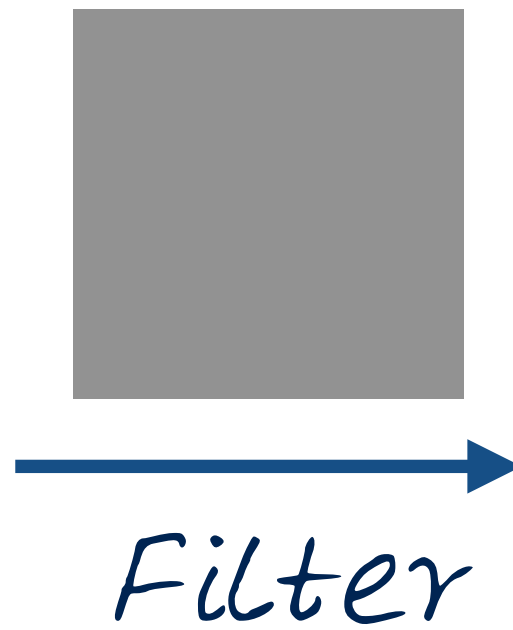




# Example: Box Filter



Original



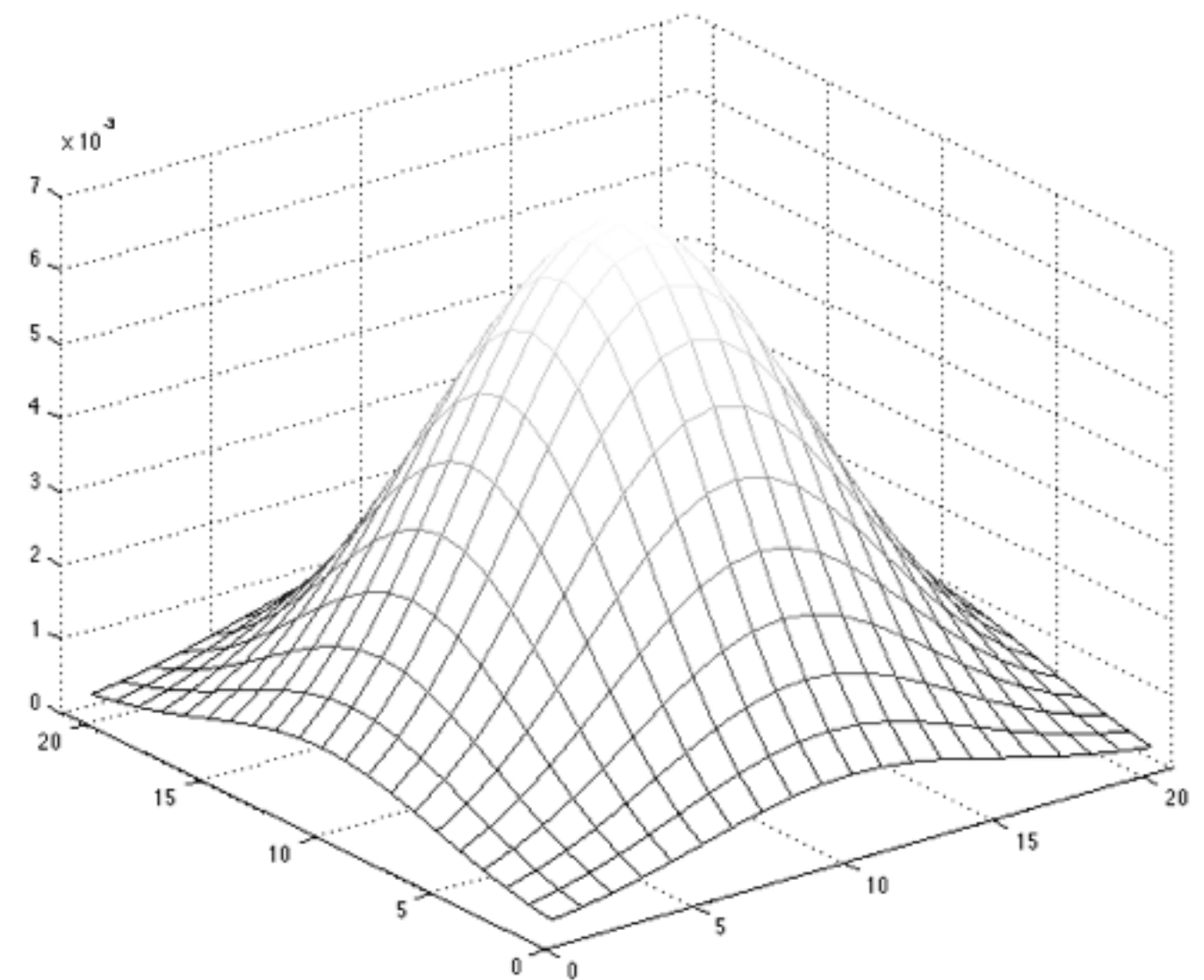
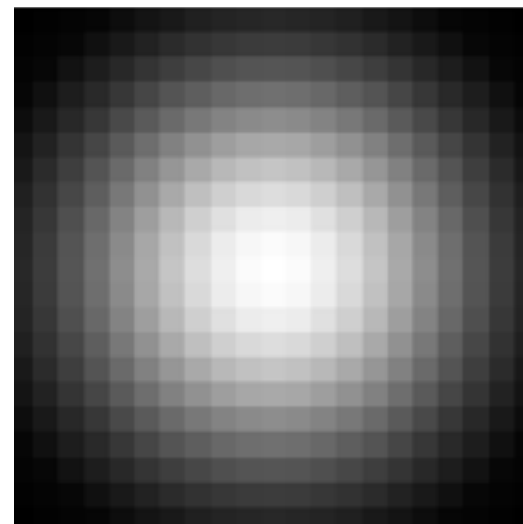
Result



# Example: Gaussian Filter

## Gaussian Filter

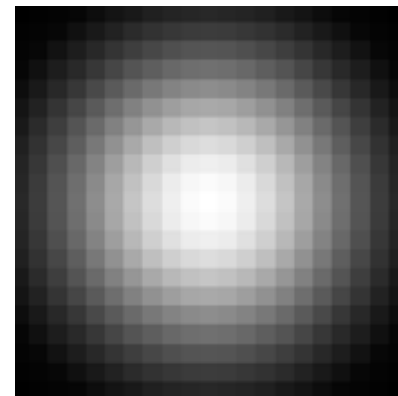
- Size:  $21 \times 21$
- Values: Gaussian or Normal distribution



# Example: Gaussian Filter



Original



Filter



Result

# Compare: Average vs. Gaussian

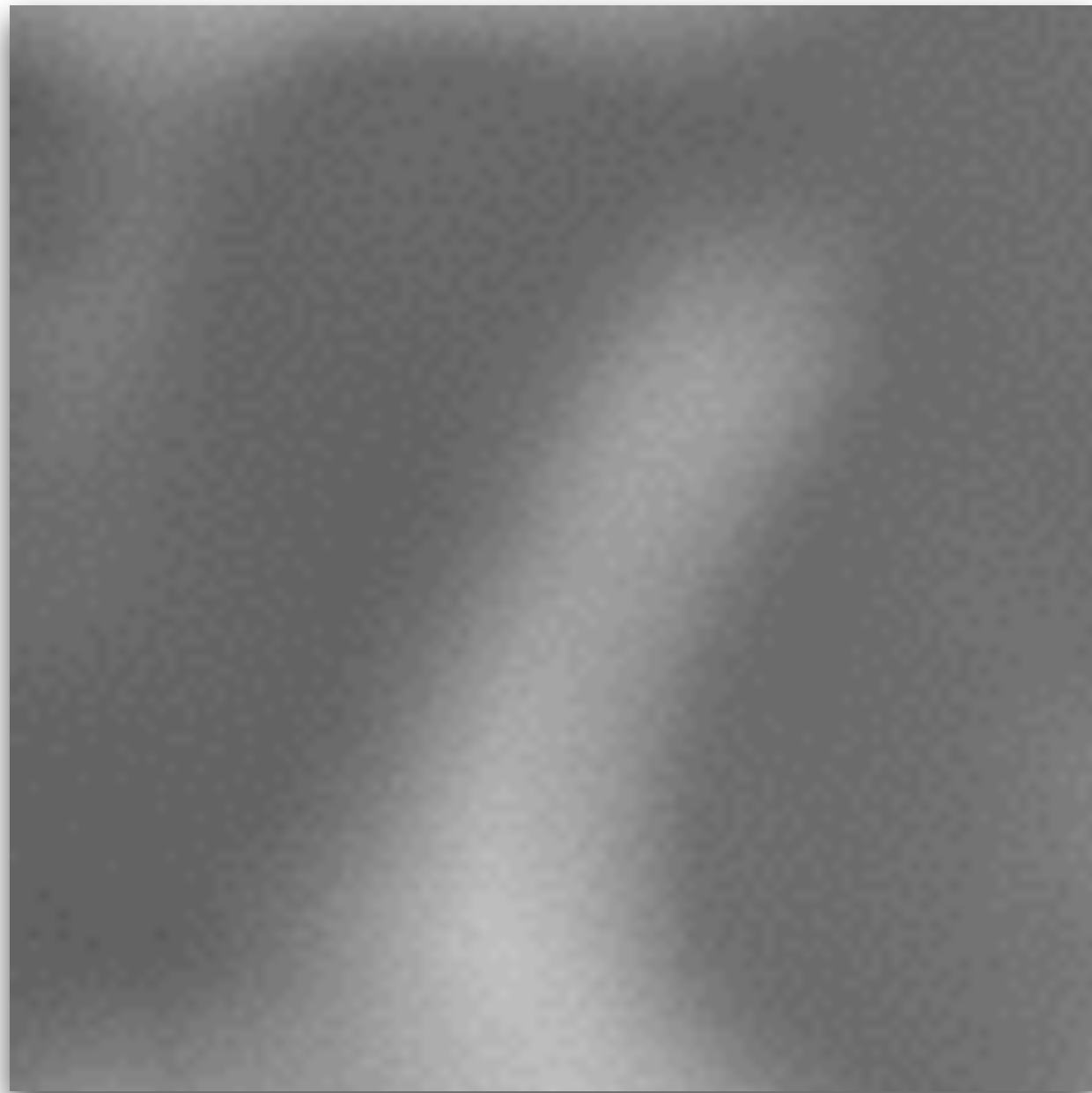


Average filter result

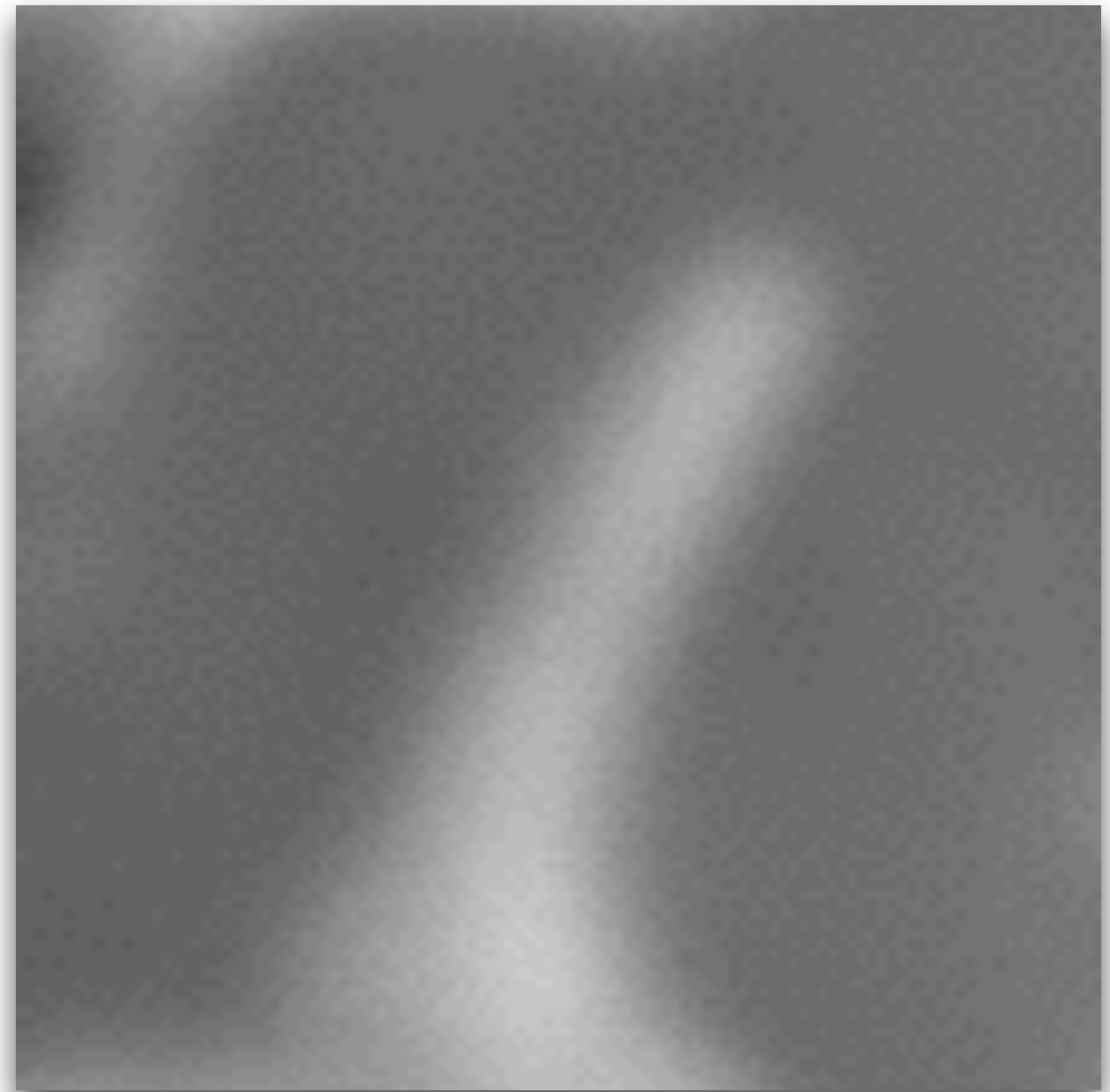


Gaussian filter result

# Compare: Average vs. Gaussian



Average filter result

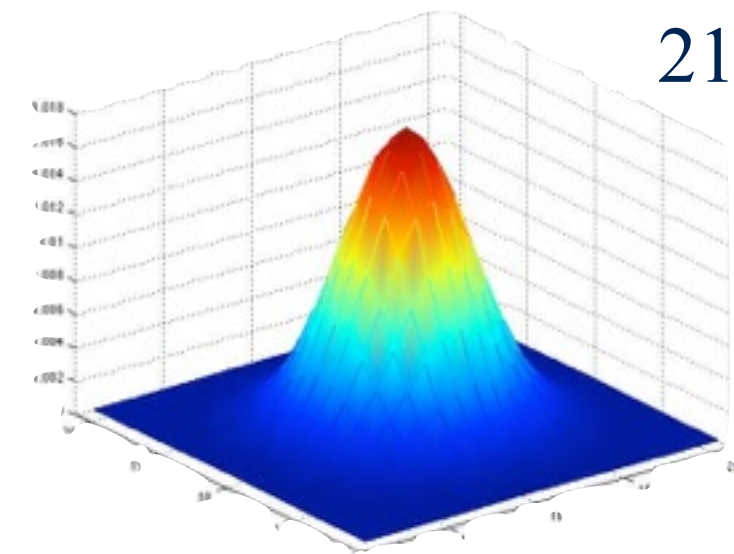


Gaussian filter result

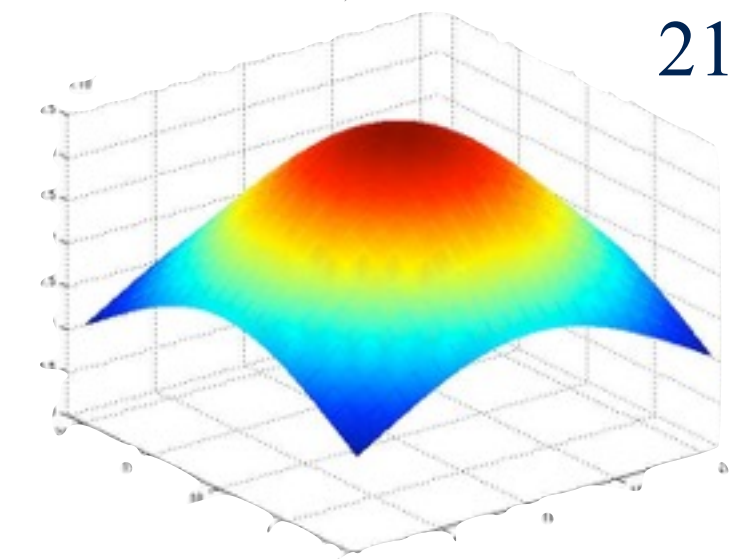


# Using Gaussian Filters?

- \* Square kernels are NOT smooth
- \* Average filter not eq. to a defocussed lens
- \* A single point of light viewed in a defocussed lens looks like a fuzzy blob; the averaging process is square.
- \* Gaussian function in 2D, with  $\sigma$  as the Variance, centered at (0,0):



21x21,  $\sigma=3$



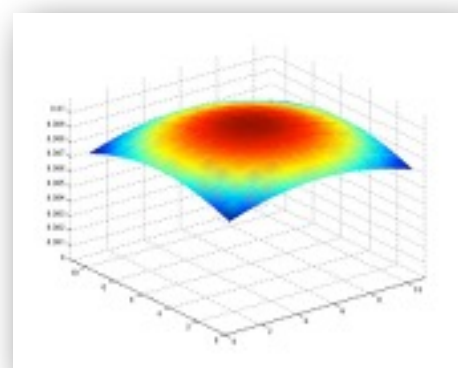
21X21,  $\sigma=9$

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

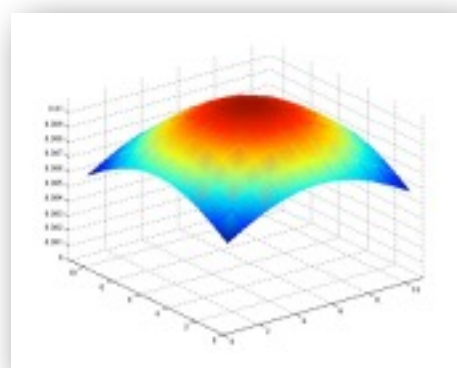
# Using Gaussian Filters for Smoothing



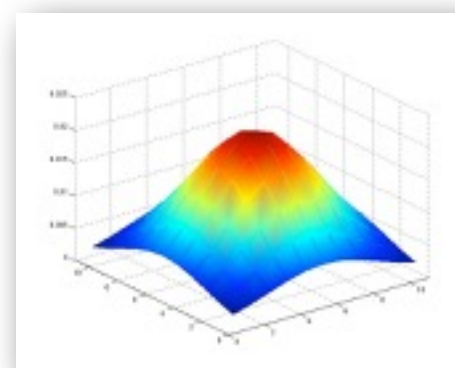
Original, 256 X 256



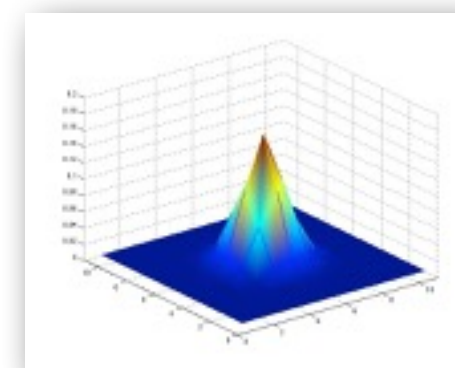
$\sigma=9$



$\sigma=6$



$\sigma=3$

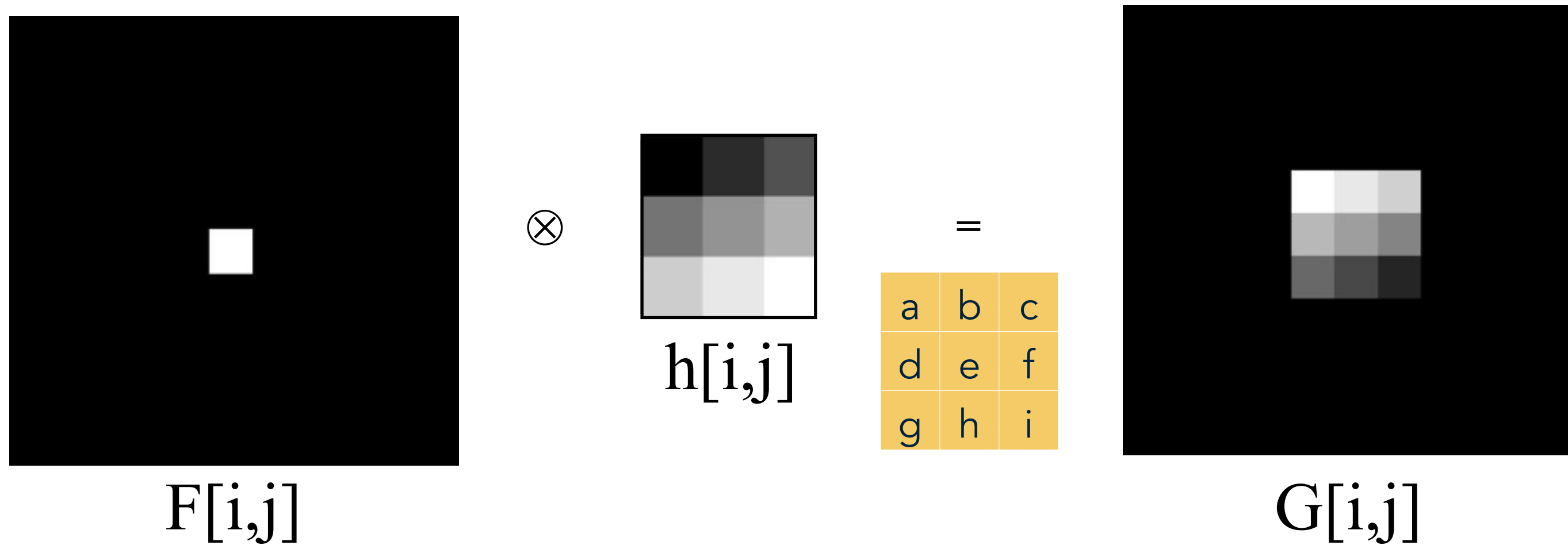


$\sigma=1$

$\sigma$  determines extent of smoothing



# Filtering by a kernel (defining Convolution)



- \* 'Filter' means to slide the kernel over the image
- \* Results in a reversed response

# Convolution Method

- \* Convolution is a mathematical operation on two functions **F** and **h**,
- \* Produces a third function that is typically viewed as a modified version
- \* Gives the area of overlap between the two functions
- \* In a form of the amount that one of the original functions is translated.

# Convolution Method

$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] F[i - u, j - v]$$

Denoted by  $G = h * F$

\* Flip filter in both dimensions.

\* Bottom to top

\* Right to left

\* Then apply cross-correlation

a	b	c
d	e	f
g	h	i

g	h	i
d	e	f
a	b	c

i	h	g
f	e	d
c	b	a

# Convolution vs. Cross-Correlation

Cross-Correlation:

$$G = h \otimes F$$

a	b	c
d	e	f
g	h	i

$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] F[i + u, j + v]$$

g	h	i	i	h	g
d	e	f	f	e	d
a	b	c	c	b	a

Convolution:

$$G = h * F$$

$$G[i, j] = \sum_{u=-k}^k \sum_{v=-k}^k h[u, v] F[i - u, j - v]$$

# Properties of Convolution

- \* Linear and Shift Invariants.
- \* Behaves the same everywhere (i.e., the value of the output depends on the pattern in the image neighborhood, not the position of the neighborhood).
- \* Commutative:  $F * G = G * F$
- \* Associative:  $(F * G) * H = F * (G * H)$

# Properties of Convolution

- \* Identity: Unit Impulse

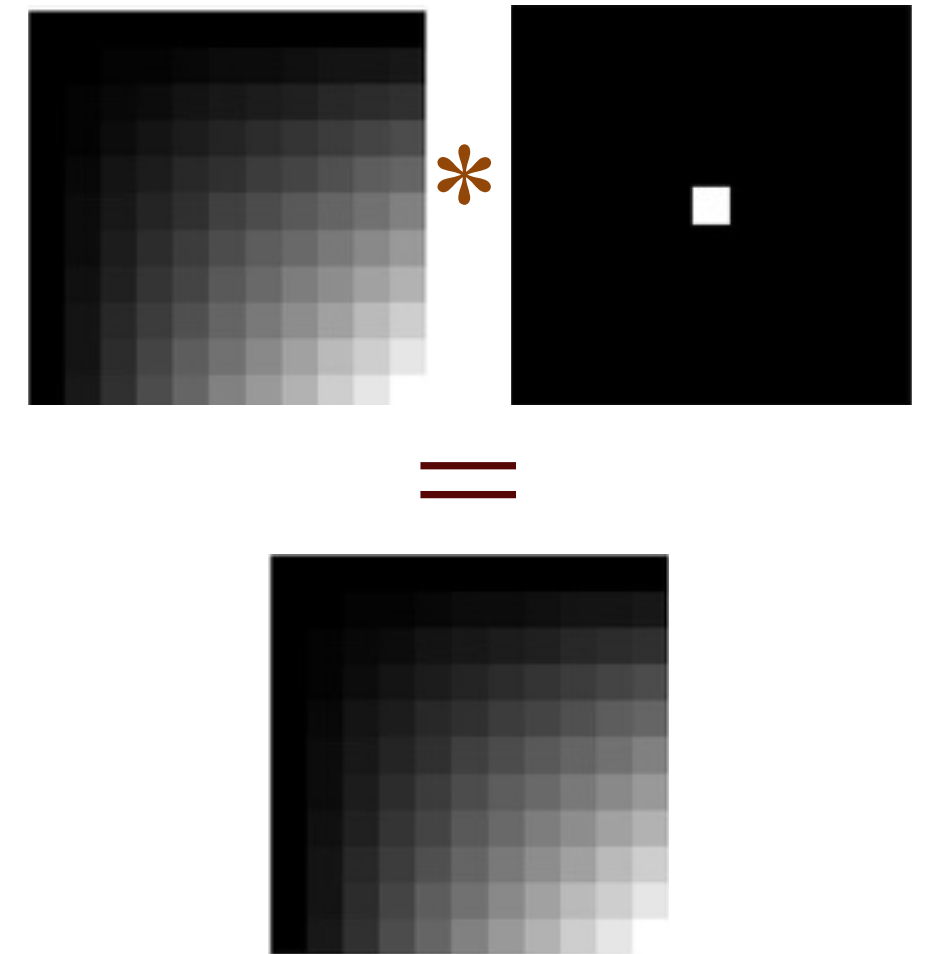
- \*  $E = [...0,0,1,0,0...]$ ,

- \*  $F * E = F$

- \* True of Cross-Correlation?

- \* Separable:

- \* If the filter is separable, convolve all rows, then convolve all columns.



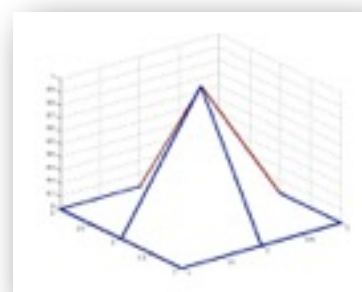


# Linear Filters

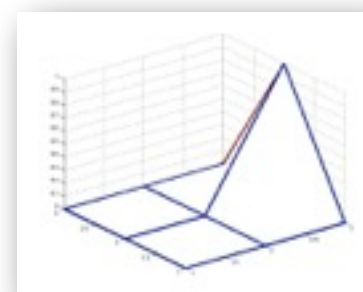
original, 64 X 64



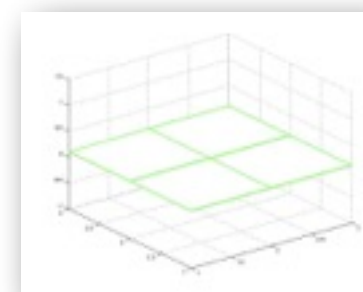
0	0	0
0	1	0
0	0	0



0	0	0
0	0	1
0	0	0



1/9	1/9	1/9
1/9	1/9	1/9
1/9	1/9	1/9



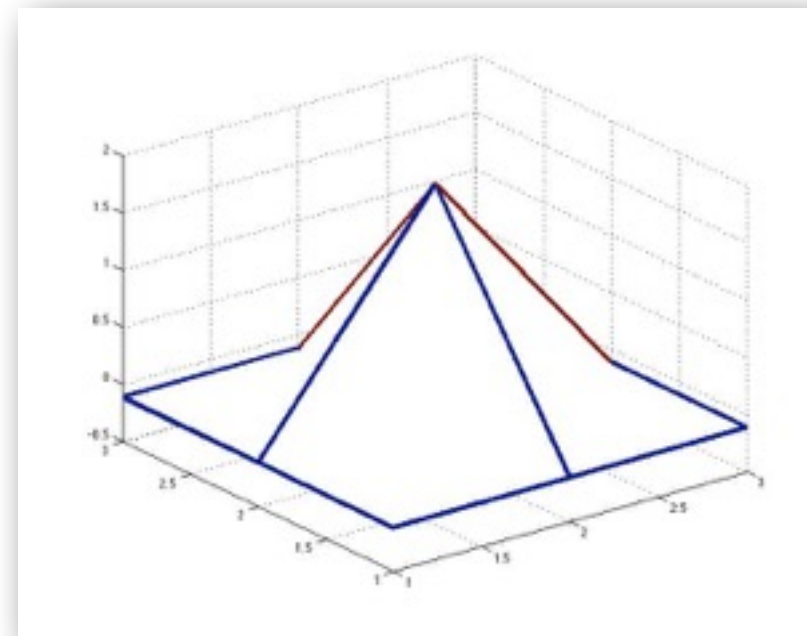
# Linear Filters

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \times 2 - \begin{bmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{bmatrix} =$$



original,  
64x64

-0.11	-0.11	-0.11
-0.11	1.9	-0.11
-0.11	-0.11	-0.11



# Summary



- \* Cross-Correlation
- \* Convolution
- \* Differences between the Cross-Correlation and Convolution methods for filtering images.
- \* Properties of the Convolution method for filtering images.

# Next Class

Image Analysis:  
Edge Detection



# Credits

- \* matlab software by mathworks Inc.
- \* Some Slides adapted from Aaron Bobick, Steve Seitz, Steve Marschner & David Forsyth.
- \* Images used from USC's Signal and Image Processing Institute's Image Database



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