

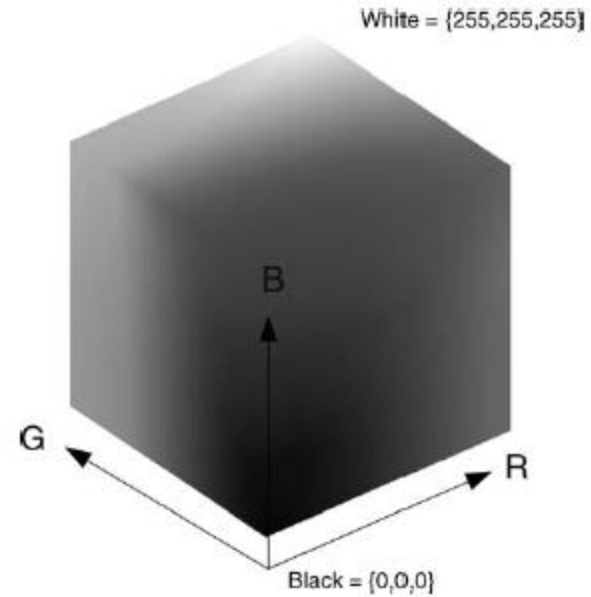
Matlab Tutorial

Derivatives, Filtering, Pyramids

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UCF

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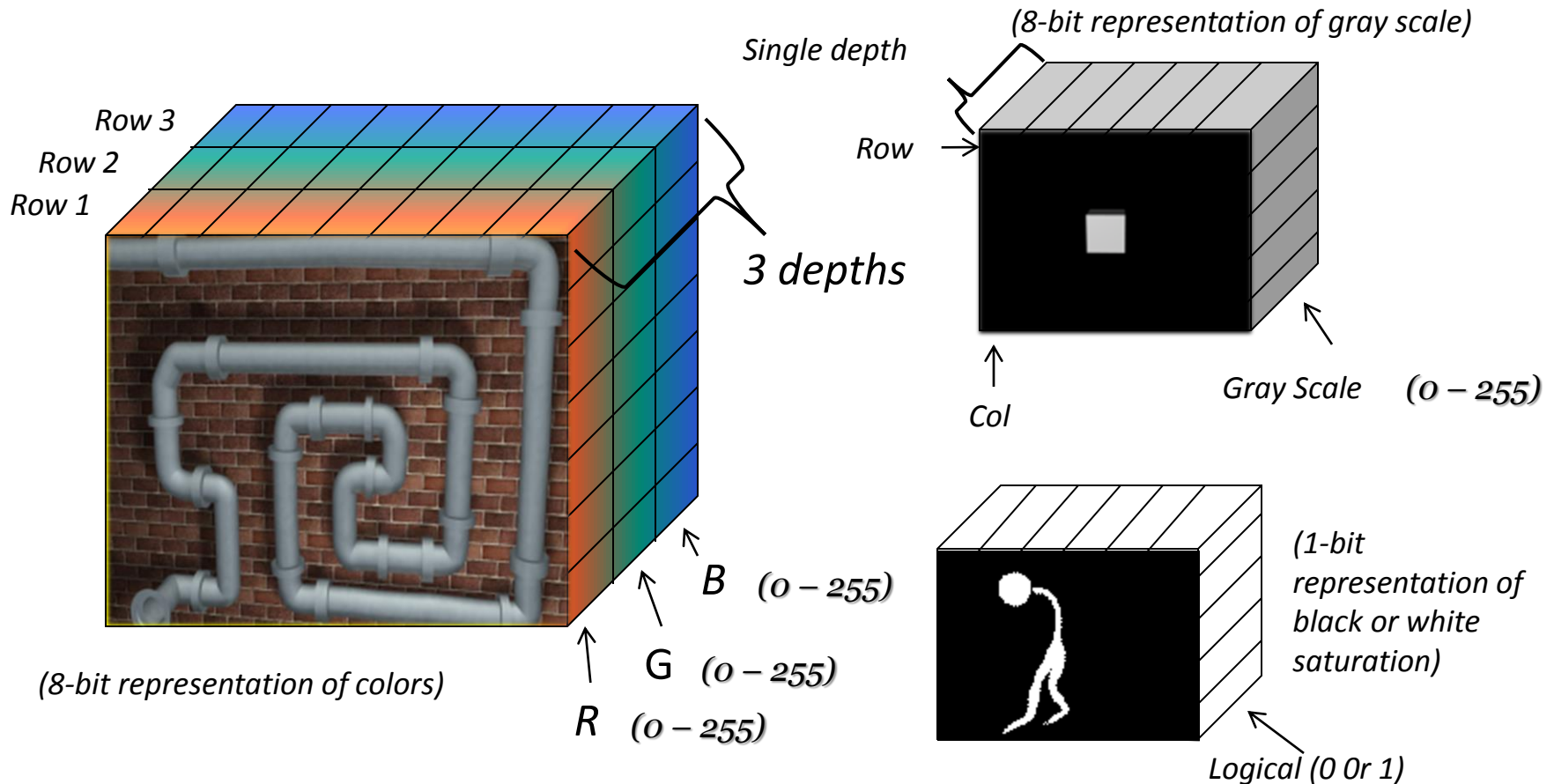


Matlab Tutorial. Session 1

COLOR SPACES: RGB, HSV, ETC

Image Architecture

- Raster Images (RGB, Gray Scale, Logical)



RGB space

- `>> I = imread('board.tif')`
- Displaying image:
 - `>> imshow(I)`
- Check image size
 - `>> size(I)`
- Convert color image to black and white image:
 - `>> rgb2gray(I)`
 - $\% \text{ Gray} = 0.2989 * R + 0.5870 * G + 0.1140 * B$

Other Color spaces

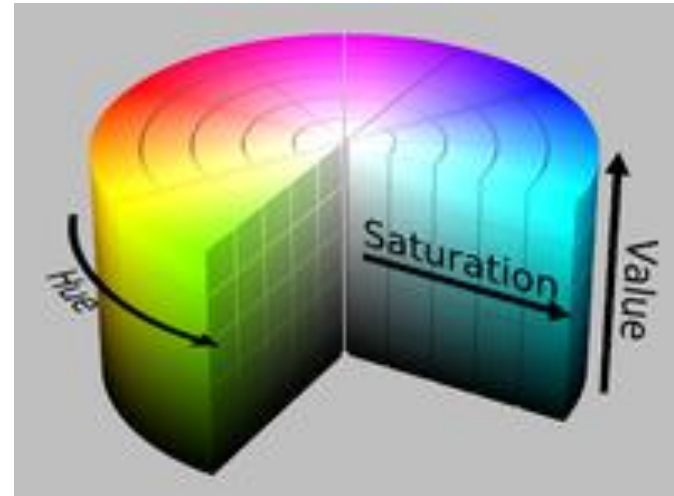
http://en.wikipedia.org/wiki/HSL_and_HSV

- `rgb2hsv(I)`
- `rgb2ycbcr(I)`
- `rgb2ntsc (I)`

CIELAB or CIECAM02

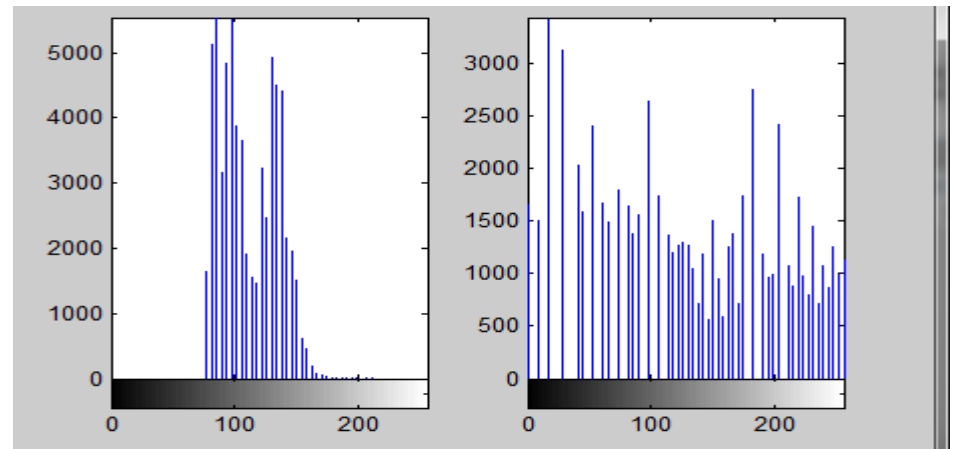
Color Space Conversions

<code>applycform</code>	Apply device-independent color space transformation
<code>iccfind</code>	Search for ICC profiles
<code>iccread</code>	Read ICC profile
<code>iccroot</code>	Find system default ICC profile repository
<code>iccwrite</code>	Write ICC color profile to disk file
<code>isicc</code>	True for valid ICC color profile
<code>lab2double</code>	Convert $L^*a^*b^*$ data to double
<code>lab2uint16</code>	Convert $L^*a^*b^*$ data to uint16
<code>lab2uint8</code>	Convert $L^*a^*b^*$ data to uint8
<code>makecform</code>	Create color transformation structure
<code>ntsc2rgb</code>	Convert NTSC values to RGB color space
<code>rgb2ntsc</code>	Convert RGB color values to NTSC color space
<code>rgb2ycbcr</code>	Convert RGB color values to YCbCr color space
<code>whitepoint</code>	XYZ color values of standard illuminants
<code>xyz2double</code>	Convert XYZ color values to double
<code>xyz2uint16</code>	Convert XYZ color values to uint16
<code>ycbcr2rgb</code>	Convert YCbCr color values to RGB color space



Histogram Equalization

```
%Histogram Equalization
I = imread('pout.tif');
J = histeq(I);
figure(1);
subplot(1,2,1);
imshow(I);
subplot(1,2,2);
imshow(J);
figure(2);
subplot(1,2,1);
imhist(I,64);
subplot(1,2,2);
imhist(J,64);
```



DERIVATIVES

Derivatives (Filters)

- In a continuous 1d Signal, derivative is:
 - $\lim_{dx \rightarrow 0} \frac{f(x+dx) - f(x)}{dx}$
- In a discrete 1d signal, derivative is:
 - $f(x+1) - f(x)$
 - It is a convolution with the filter:
 - Other popular filter is:

-1
1

Sobel Filter

-1
0
1

Derivatives in 2D

Table 4.1 Derivative operators: their formal (continuous) definitions and corresponding discrete approximations

2 D derivative measure	Continuous case	Discrete case	KERNEL									
$\frac{\partial f}{\partial x}$	$\lim_{\Delta x \rightarrow 0} \frac{f(x + \Delta x, y) - f(x, y)}{\Delta x}$	$f(x + 1, y) - f(x, y)$	[1 -1]									
$\frac{\partial f}{\partial y}$	$\lim_{\Delta y \rightarrow 0} \frac{f(x, y + \Delta y) - f(x, y)}{\Delta y}$	$f(x, y + 1) - f(x, y)$	<table border="1" style="display: inline-table; vertical-align: middle;"><tr><td>-1</td></tr><tr><td>1</td></tr></table>	-1	1							
-1												
1												
$\nabla f(x, y)$	$\left[\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$	$[f(x + 1, y) - f(x, y), f(x, y + 1) - f(x, y)]$										
$\frac{\partial^2 f}{\partial x^2}$	$\lim_{\Delta x \rightarrow 0} \frac{(\partial f / \partial x)(x + \Delta x, y) - (\partial f / \partial x)f(x, y)}{\Delta x}$	$f(x + 1, y) - 2f(x, y) + f(x - 1, y)$	[1 -2 1]									
$\frac{\partial^2 f}{\partial y^2}$	$\lim_{\Delta y \rightarrow 0} \frac{(\partial f / \partial x)(x, y + \Delta y) - (\partial f / \partial x)(x, y)}{\Delta y}$	$f(x, y + 1) - 2f(x, y) + f(x, y - 1)$	[1 -2 1]'									
$\nabla^2 f(x, y)$	$\frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2}$	$f(x + 1, y) + f(x - 1, y) - 4f(x, y) + f(x, y + 1) + f(x, y - 1)$	<table style="display: inline-table; vertical-align: middle;"><tr><td>0</td><td>1</td><td>0</td></tr><tr><td>1</td><td>-4</td><td>1</td></tr><tr><td>0</td><td>1</td><td>0</td></tr></table>	0	1	0	1	-4	1	0	1	0
0	1	0										
1	-4	1										
0	1	0										

Convolution

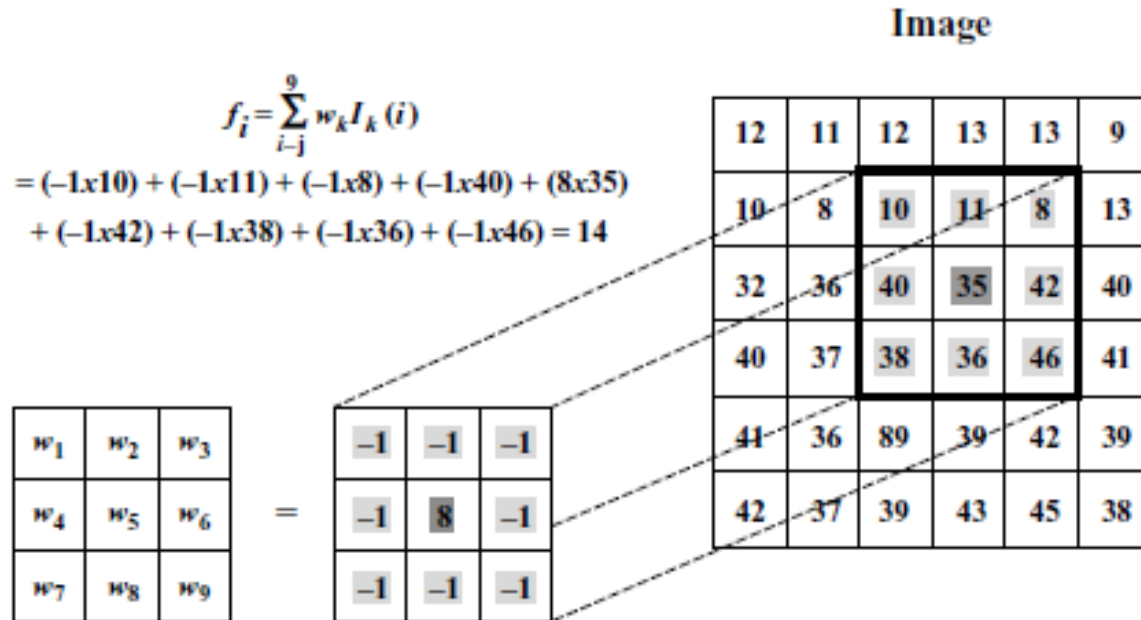
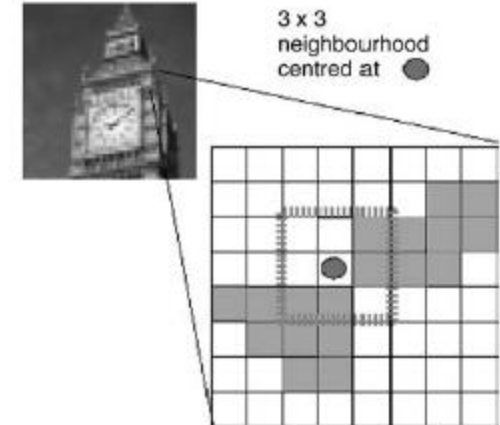
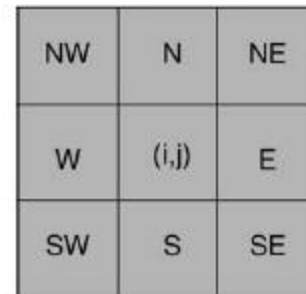


Figure 2.12 Discrete convolution. The centre pixel of the kernel and the target pixel in the image are indicated by the dark grey shading. The kernel is 'placed' on the image so that the centre and target pixels match. The filtered value of the target pixel is then given by a linear combination of the neighbourhood pixels, the specific weights being determined by the kernel values. In this specific case the target pixel of original value 35 has a filtered value of 14

Filtering in 2D Arrays (Convolution)



Image

$$f_i = \sum_{i,j} w_k I_k(i)$$

$$= (-1 \times 10) + (-1 \times 11) + (-1 \times 8) + (-1 \times 40) + (8 \times 35) + (-1 \times 42) + (-1 \times 38) + (-1 \times 36) + (-1 \times 46) = 14$$

w_1	w_2	w_3
w_4	w_5	w_6
w_7	w_8	w_9

=

-1	-1	-1
-1	8	-1
-1	-1	-1

12	11	12	13	13	9
10	8	10	11	8	13
32	36	40	35	42	40
40	37	38	36	46	41
41	36	89	39	42	39
42	37	39	43	45	38

Alternative derivatives in 2D

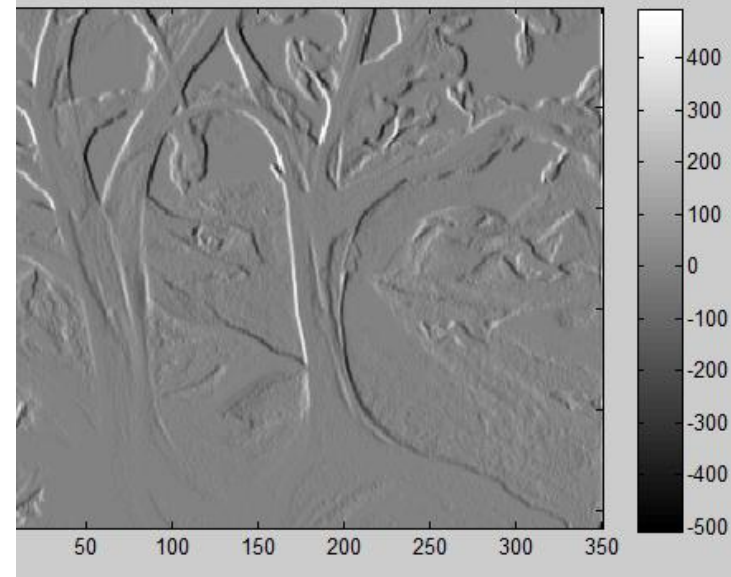
Roberts		Prewitt	Sobel																						
<table border="1"><tr><td>0</td><td>-1</td></tr><tr><td>1</td><td>0</td></tr></table>	0	-1	1	0	X derivative	<table border="1"><tr><td>1</td><td>0</td><td>-1</td></tr><tr><td>1</td><td>0</td><td>-1</td></tr><tr><td>1</td><td>0</td><td>-1</td></tr></table>	1	0	-1	1	0	-1	1	0	-1	<table border="1"><tr><td>1</td><td>0</td><td>-1</td></tr><tr><td>2</td><td>0</td><td>-2</td></tr><tr><td>1</td><td>0</td><td>-1</td></tr></table>	1	0	-1	2	0	-2	1	0	-1
0	-1																								
1	0																								
1	0	-1																							
1	0	-1																							
1	0	-1																							
1	0	-1																							
2	0	-2																							
1	0	-1																							
<table border="1"><tr><td>-1</td><td>0</td></tr><tr><td>0</td><td>1</td></tr></table>	-1	0	0	1	Y derivative	<table border="1"><tr><td>1</td><td>1</td><td>1</td></tr><tr><td>0</td><td>0</td><td>0</td></tr><tr><td>-1</td><td>-1</td><td>-1</td></tr></table>	1	1	1	0	0	0	-1	-1	-1	<table border="1"><tr><td>1</td><td>2</td><td>1</td></tr><tr><td>0</td><td>0</td><td>0</td></tr><tr><td>-1</td><td>-2</td><td>-1</td></tr></table>	1	2	1	0	0	0	-1	-2	-1
-1	0																								
0	1																								
1	1	1																							
0	0	0																							
-1	-1	-1																							
1	2	1																							
0	0	0																							
-1	-2	-1																							

Figure 4.9 First-order edge-detection filters

Derivatives in Matlab

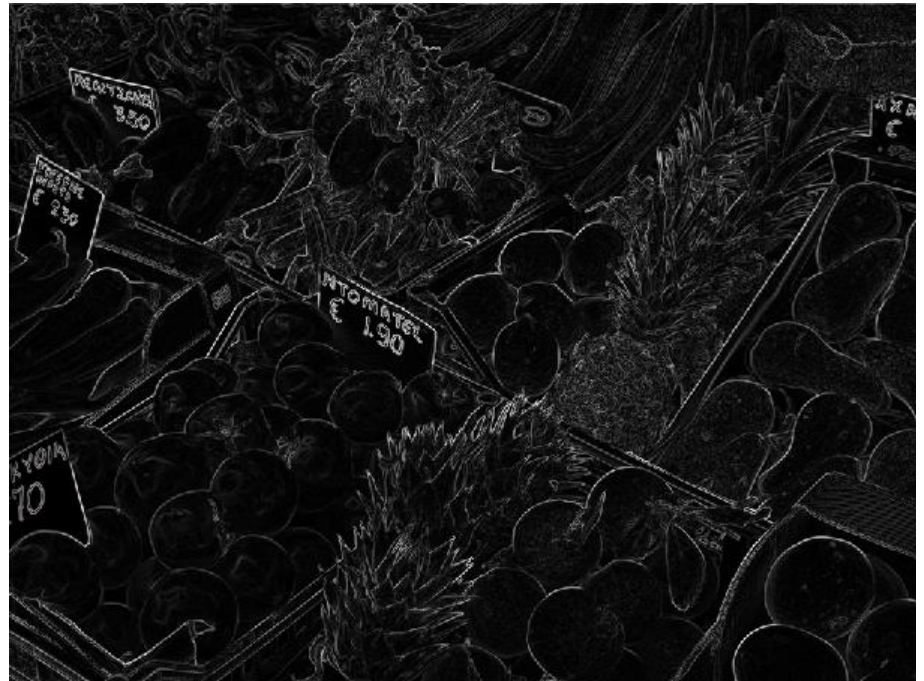
```
I = imread('trees.tif');  
imshow(I)  
k1=[ 1 0 -1;2 0 -2; 1 0 -1]  
o=imfilter(double(I),k1,'same');  
figure; imagesc(o)  
colormap gray
```

QUESTION: Why imagesc
instead of imshow ?



Gradient Magnitude and direction

```
% - Display derivative.  
horgradI2=abs(conv2(double(I2),hcent,'same'));  
figure;imshow(uint8(horgradI2));  
vergradI2=abs(conv2(double(I2),hcent','same'));  
figure;imshow(uint8(vergradI2));  
  
% - Compute gradient magnitude and direction.  
gradMag=sqrt(vergradI2.^2+horgradI2.^2);  
figure;imshow(uint8(gradMag));  
gradDir=atan2(vergradI2,horgradI2);
```



FILTERING

Mean filter

```
% - Box kernel for image smoothing  
I1 = imread('groceries.jpg');  
I2 = rgb2gray(I1);  
h=ones(7)/49;  
I6=uint8(conv2(I2,h,'same'));  
figure;imshow(I6);
```



Special filters

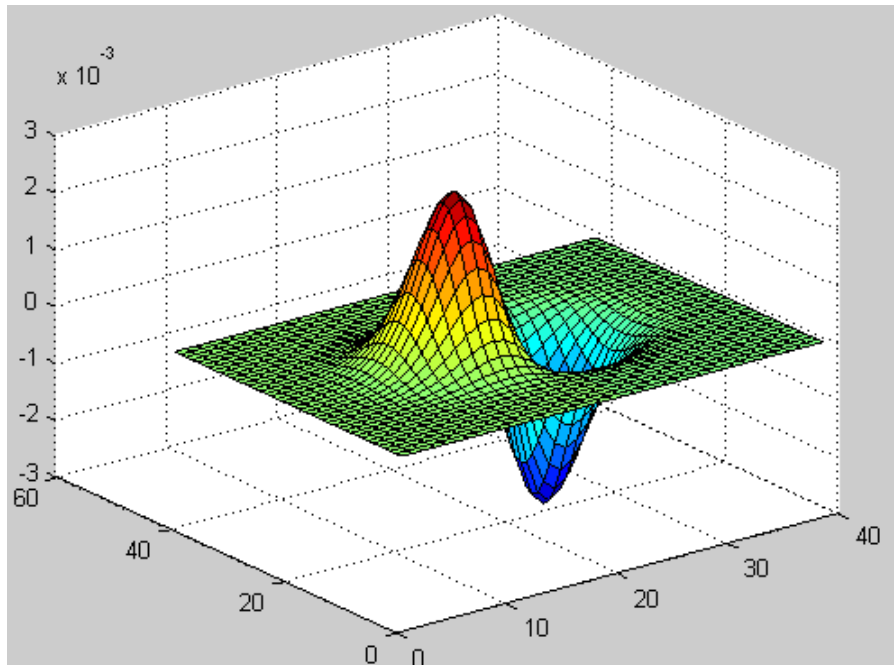
```
% - fspecial. Create arbitrary sized Gaussian kernel
% with different sizes and variances and their effects on smoothing (over/under smooth)
hg = fspecial('gaussian', [7 7], 1);

% - Use Gaussian kernel for smoothing
I3=conv2(double(I2),hg,'same');
imshow(uint8(I3));
```



Gaussian Derivative

```
% - Gaussian derivative. Write a one line code/formula for creating a Gaussian derivative kernel, for gradient computa  
% - I emphasized Gaussian derivative and that, in general, this should be used for gradient computation of images.  
gaussder = conv2(fspecial('gaussian',[7 7],4),[1 0 -1],'valid');  
figure;surf(gaussder);  
I4= conv2(double(I2),gaussder);  
imshow(I4);
```



Separable Filters

```
% - Separable kernels
% - Computing derivatives
close all;
keyboard;
M=magic(7)
f1=[1;-1] % vertical gradient
f2=[-1 1] % horizontal gradient
a=conv2(double(M),f1,'same')
b=conv2(double(a),f2,'same')
f3=f1*f2
c=conv2(double(M),f3,'same')
sum(sum(abs(b-c)))
```

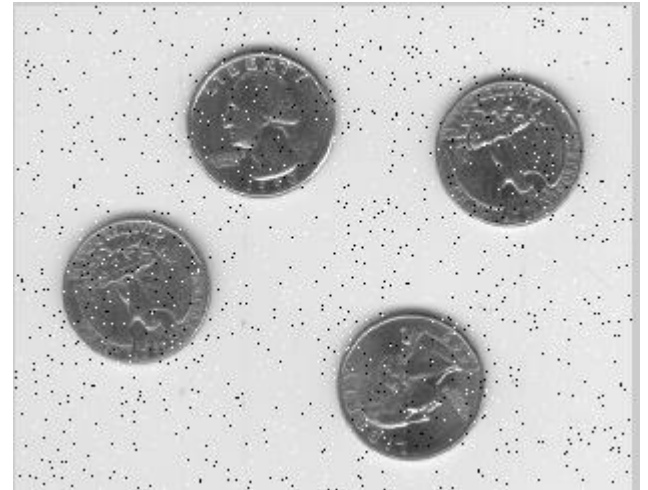
Ans = 0

Example: Removing noise

```
% Create a noisy image  
I = imread('eight.tif');  
imshow(I)  
J = imnoise(I,'salt & pepper',0.02);  
figure, imshow(J)
```

```
% Mean filter  
K = filter2(fspecial('average',3),J)/255;  
figure, imshow(K)
```

```
%Median filter  
L = medfilt2(J,[3 3]);  
figure, imshow(L)
```

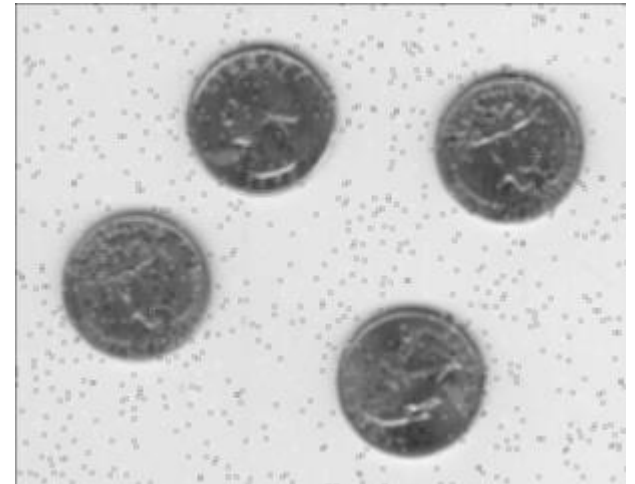


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```
%Median filter  
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```



Example: Removing noise

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I = imread('eight.tif');  
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J = imnoise(I,'salt & pepper',0.02);  
figure, imshow(J)  
  
% Mean filter  
K = filter2(fspecial('average',3),J)/255;  
figure, imshow(K)
```

```
%Median filter  
L = medfilt2(J,[3 3]);  
figure, imshow(L)
```



Smoothing

$$L(x, y, \sigma) = G(x, y, \sigma) * I(x, y)$$

$$G(x, y, \sigma) = \frac{1}{2\pi\sigma^2} e^{-(x^2+y^2)/2\sigma^2}$$

L is a blurred image

- G is the Gaussian Blur operator

- I is an image

- x,y are the location coordinates

- σ is the “scale” parameter. Think of it as the amount of blur. Greater the value, greater the blur.

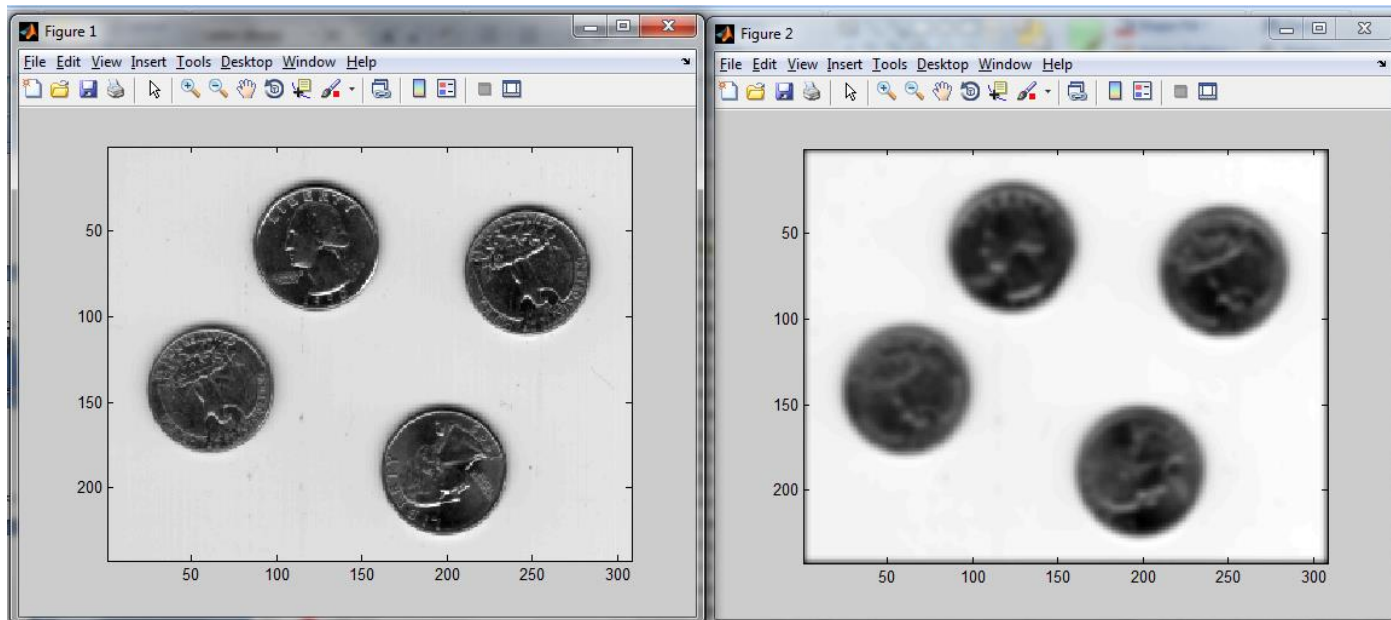
- The * is the convolution operation in x and y. It “applies” gaussian blur G onto the image I

Smoothing function

```
% gauss_filter: Obtains a smoothed gaussian filter image
%   - Output:
%       smooth: image filter by a gaussian filter
%   - Input:
%       image: Matrix containing single band image.
%       sigma: Value of the sigma for the Gaussian filter
%       kernel_size: Size of the gaussian kernel (default: 6*sigma)
function [smooth]= gauss_filter(image,sigma,kernel_size)
if nargin < 2
    sigma=1;
end
if nargin < 3
    kernel_size=6*sigma;
end
gaussian_radio=floor(kernel_size/2); %radio of the gaussian
x=[-gaussian_radio : gaussian_radio]; % x values (gaussian kernel size)
gaussiano= exp(-x.^2/(2*sigma^2))/(sigma*sqrt(2*pi) ); %calculate the unidimensional gaussian
temp=conv2(double(image),double(gaussiano),'same');
smooth=conv2(double(temp),double(gaussiano),'same');
end
```


Example smoothing

```
I = imread('eight.tif'); imagesc(I); colormap gray;  
antialiassigma=2;  
Ismooth=gauss_filter(I,antialiassigma,11);  
figure; imagesc(Ismooth); colormap gray;
```



Resize

`%Resizing an image`

```
%Y=imresize(X,rows,cols,method); %method -nearest bilinear, bicubic
```

```
[X,map]=imread ('groceries.jpg');
```

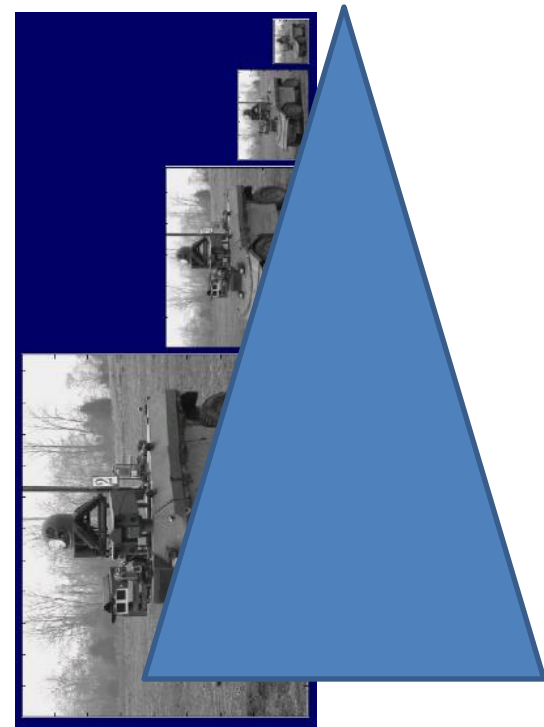
```
Y=imresize(X,[100,200],'bicubic'); %method -nearest bilinear, bicubic
```

```
imshow(Y);
```



Pyramid

- Definition:
 - is a type of multi-scale signal representation in which a signal or an image is subject to repeated smoothing and subsampling



Pyramid

```
I0 = imread('cameraman.tif');
```

```
I1 = impyramid(I0, 'reduce');
```

```
I2 = impyramid(I1, 'reduce');
```

```
I3 = impyramid(I2, 'reduce');
```

```
imshow(I0)
```

```
figure, imshow(I1)
```

```
figure, imshow(I2)
```

```
figure, imshow(I3)
```