

# Image inpainting

# Outline

- Why inpaint?
- Image filtering (convolution), simplest inpainting
- Heat flow, diffusion process, diffusion equation
- Anisotropic diffusion
- Further development

# Inpainting applications

- Image restoration
- Object removal
- Text removal
- Special effects

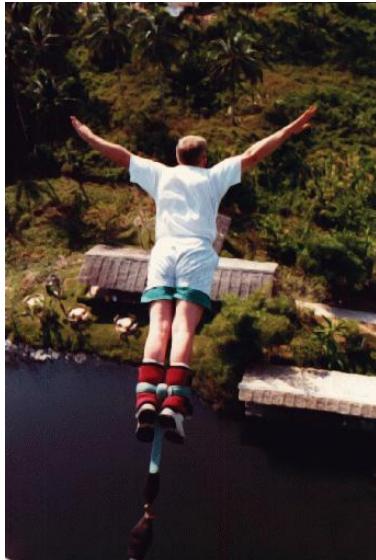
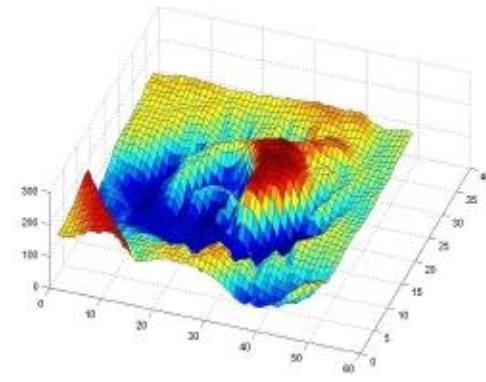
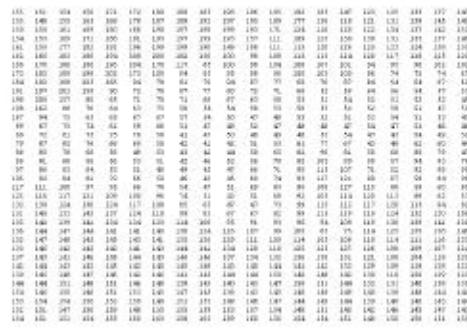
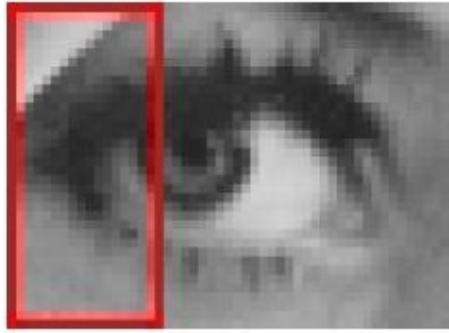


Image courtesy Andrea Bacynski

# Simplest inpainting

- Image is 2D discrete function



- Form an iterative algorithm
  - Each update gets you closer to the desired result

$$I_{i,j}^{n+1} = I_{i,j}^n + \Delta t I_{t,i,j}^n \quad \forall (i,j) \in \Omega$$

# Simplest inpainting

- Smooth image (using Gaussian filter)
- Replace missing pixels by smoothed version
- Repeat until the result stops changing

(MATLAB demo)



conv\_intro.m

# Convolution

Convolution of two functions  $f(x)$  and  $g(x)$

$$h(x) = f(x) \otimes g(x) = \int_{-\infty}^{+\infty} f(r)g(x-r)dr$$

Diagram illustrating the components of convolution:

- Output filtered image
- Image
- convolution operator
- Filter (mask/kernel)
- Support region of filter where  $g(x-r)$  is nonzero

The diagram shows the convolution equation  $h(x) = f(x) \otimes g(x) = \int_{-\infty}^{+\infty} f(r)g(x-r)dr$ . Arrows point from the labels to the corresponding parts of the equation: 'Output filtered image' points to  $h(x)$ , 'Image' points to  $f(x)$ , 'convolution operator' points to the integral symbol, 'Filter (mask/kernel)' points to  $g(x)$ , and 'Support region of filter where  $g(x-r)$  is nonzero' points to the limits of integration.

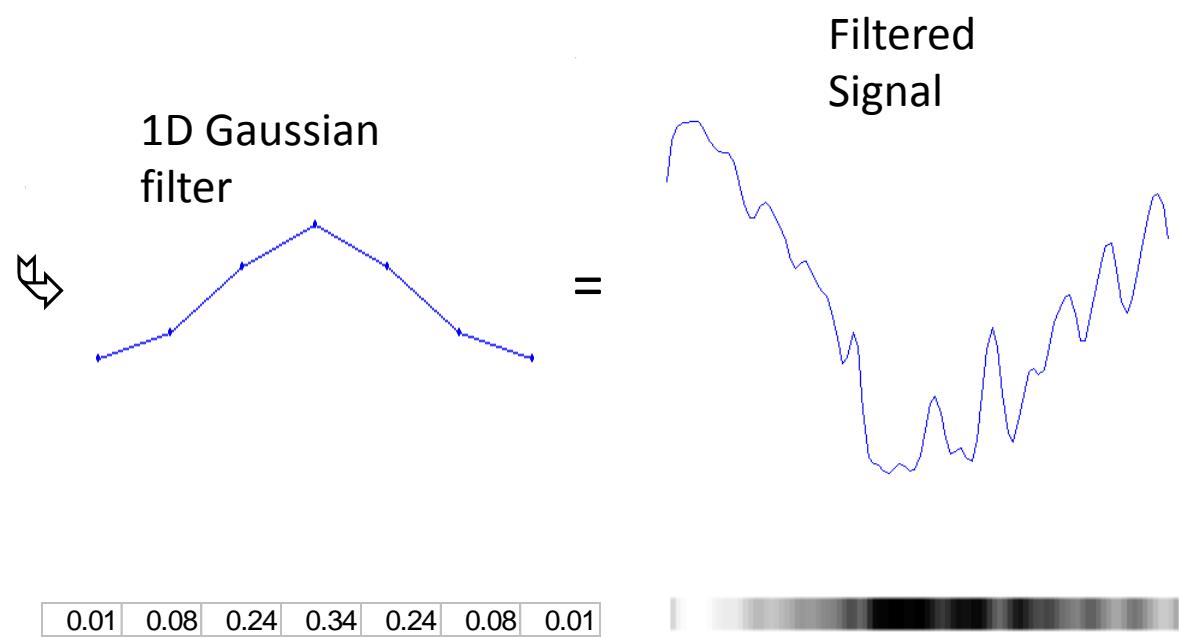
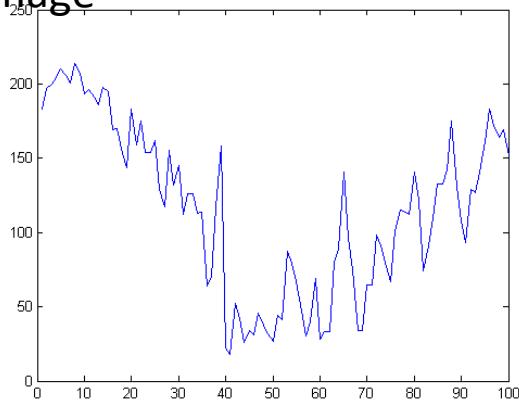
Discrete image processing 2D form

$$H(x, y) = \sum_{j=1}^{\text{height}} \sum_{i=1}^{\text{width}} I(i, j)M(x-i, y-j)$$

Compute the convolution where there are valid indices in the kernel

# Convolution example in 1D

Horizontal slice from Mandrill  
image



# Common convolution kernels

0.11	0.11	0.11
0.11	0.11	0.11
0.11	0.11	0.11

Arithmetic mean  
filter (smoothing)  
`>>fspecial('average')`

-0.17	-0.67	-0.17
-0.67	3.33	-0.67
-0.17	-0.67	-0.17

Laplacian (enhance edges)  
`>>fspecial('laplacian')`

-0.17	-0.67	-0.17
-0.67	4.33	-0.67
-0.17	-0.67	-0.17

Sharpening filter  
`>>fspecial('unsharp')`

0.01	0.08	0.01
0.08	0.62	0.08
0.01	0.08	0.01

Gaussian filter  
(smoothing)  
`>>fspecial('gaussian')`

1	0	-1
2	0	-2
1	0	-1

Sobel operators (edge  
detection in x and y directions)  
`>>fspecial('sobel')`  
`>>fspecial('sobel')`

1	2	1
0	0	0
-1	-2	-1



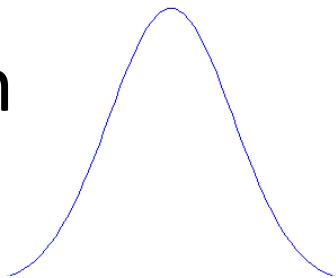
Investigate the listed kernels in Matlab by performing convolutions on the Mandrill and Lena images. Study the effects of different kernel sizes (3x3, 9x9, 25x25) on the output.

The median filter is used for noise reduction. It works by replacing a pixel value with the median of its neighbourhood pixel values (vs the mean filter which uses the mean of the neighbourhood pixel values). Apply Matlab's median filter function medfilt2 on the Mandrill and Lena images. Remember to use different filter sizes (3x3, 9x9, 16x16).

# Useful functions for convolution

- Generate useful filters for convolution

```
>>fspecial('gaussian',[kernel_height kernel_width],sigma)
```

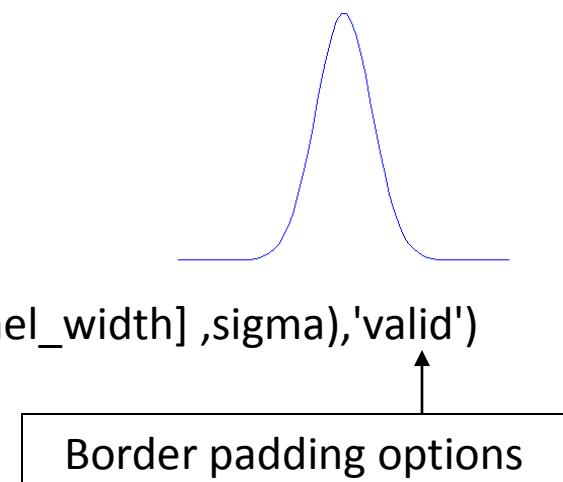
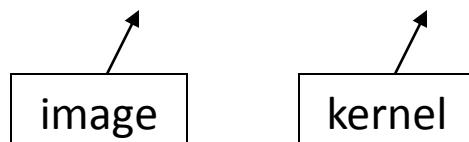


- 1D convolution

```
>>conv(signal,filter)
```

- 2D convolution

```
>>conv2(double(I(:,:2)),fspecial('gaussian',[kernel_height kernel_width] ,sigma),'valid')
```

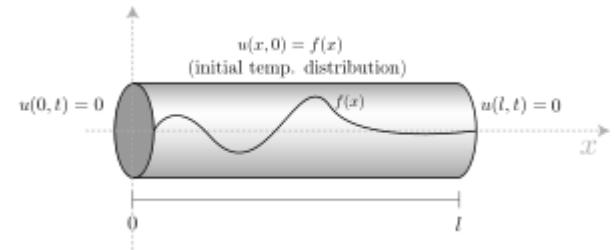


Perform the convolution of an image using Gaussian kernels with different sizes and standard deviations and display the output images.

# Heat (diffusion) equation

- 1822 Joseph Fourier: heat propagation in rod

$$\frac{\partial u(x, t)}{\partial t} = a^2 \frac{\partial^2 u(x, t)}{\partial x^2}$$



- Same in image (2D)

$$I_t(t, x, y) = a^2(I_{xx} + I_{yy}) = a^2 \Delta I$$

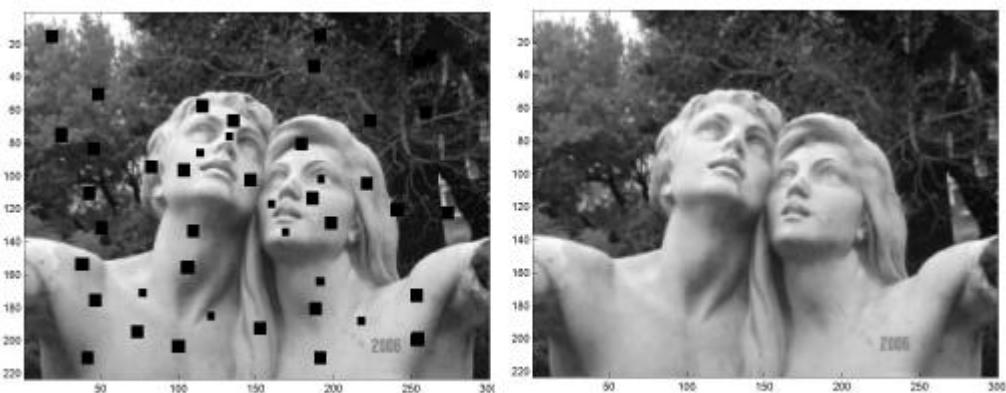
- Model intensity propagation as diffusion

# Diffusion equation solution

- Left: initial distribution, right:  $t=10$



- Damaged image:



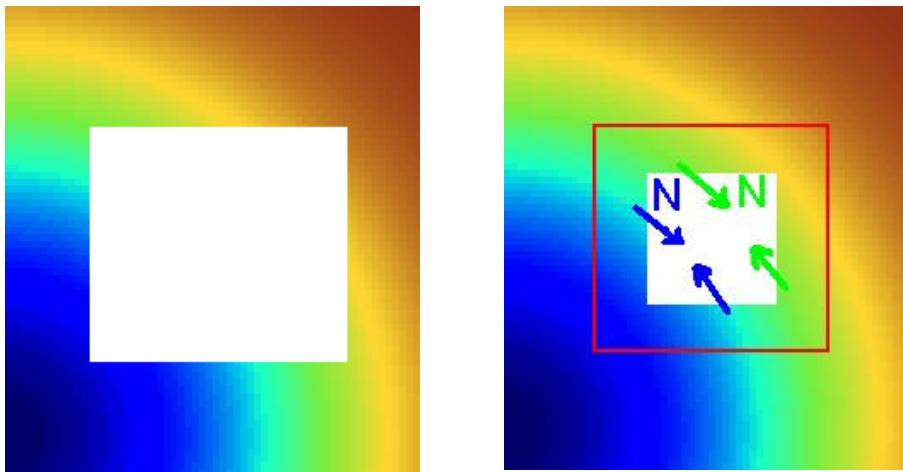
# Anisotropic diffusion

- Problem: edges are not preserved
- Do as professional restorator: continue along edges
- Solution: use anisotropic diffusion equation



# Anisotropic diffusion (2)

- Fill the hole by isophote direction

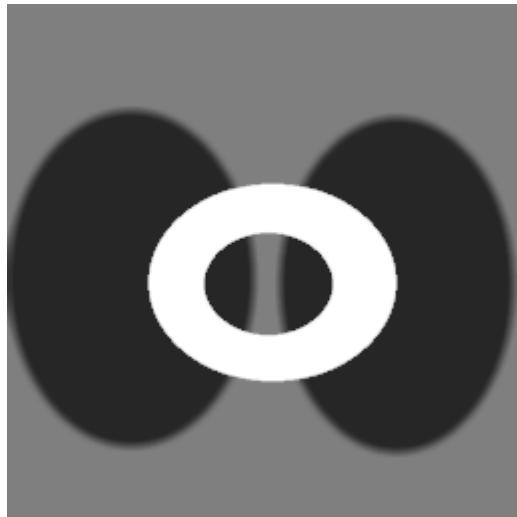


$$\vec{N} = [-I_y, I_x]$$

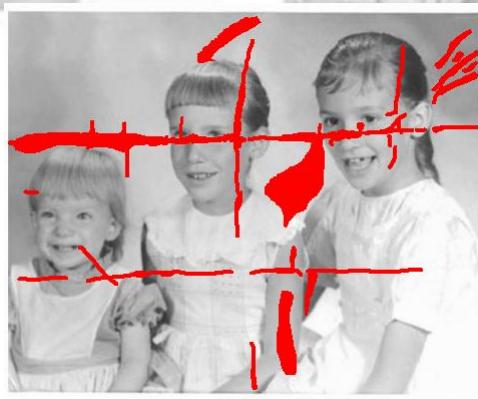
- Anisotropic equation

$$I_t = \nabla(\Delta I) \cdot \vec{N}$$

# PDE inpainting examples



# Example: photo restoration



# Further development

- Higher order equations for curvature minimization (smooth contour joining)
- Texture synthesis for large inpainting areas
- Combined approaches