

## Lecture 21:

# Continuous-Scale Morphology II: Shock Filters and Nonflat Morphology

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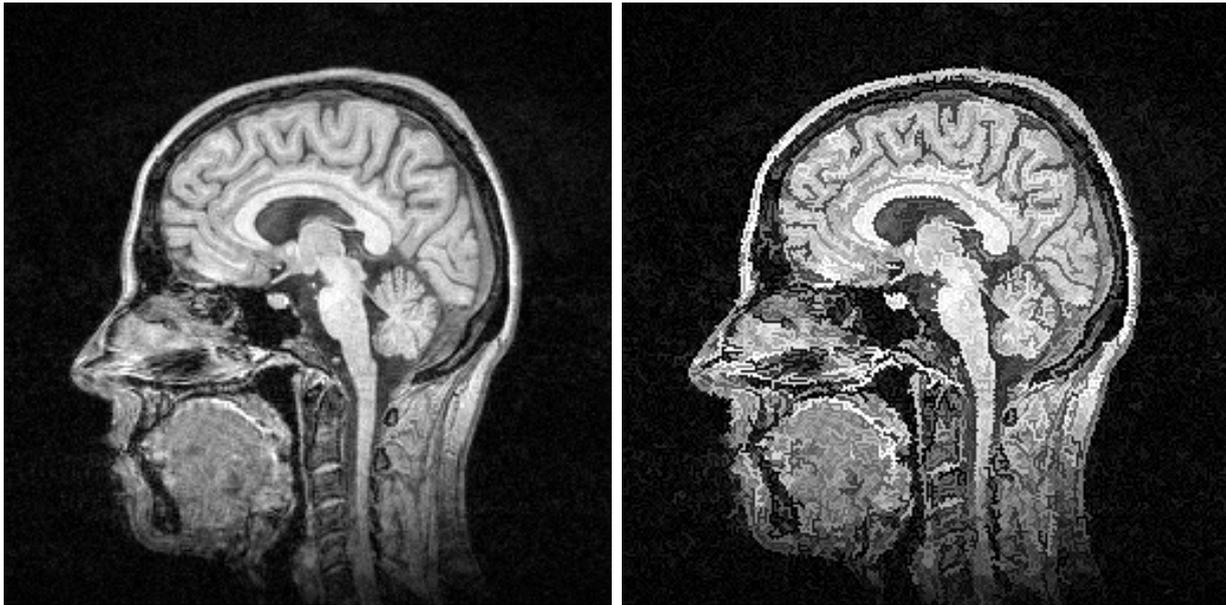
## Shock Filtering (1)

### Shock Filtering

- ◆ morphological deblurring technique
- ◆ first discrete formulation by Kramer/Bruckner (1975),  
 first PDE-based description by Osher/Rudin (1990)
- ◆ dilation around maxima (regions where  $\Delta u < 0$ ),  
 erosion around minima (regions where  $\Delta u > 0$ )
- ◆ simplest representative:
 
$$\partial_t u = -\text{sgn}(\Delta u) |\nabla u|$$
- ◆ creates shocks at edge locations where  $\Delta u = 0$ .
- ◆ no well-posedness theory known
- ◆ satisfies a discrete extremum principle
- ◆ Experiments show that segmentation-like nontrivial steady states exist.
- ◆ main disadvantage: noise sensitive

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## Shock Filtering (2)



(a) **Left:** Original image,  $\Omega = (0, 256)^2$ . (b) **Right:** Result when applying a shock filter,  $t = 100$ .

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## Shock Filtering (3)

◆ Osher/Rudin 1990:

- suggested that  $u_{\eta\eta}$  with  $\eta \parallel \nabla u$  can be a better edge detector than  $\Delta u$ :

$$\partial_t u = -\text{sgn}(u_{\eta\eta}) |\nabla u|$$

◆ Alvarez/Mazorra 1994:

- replaced  $u_{\eta\eta}$  by  $v_{\eta\eta}$  with a Gaussian-smoothed image  $v := K_\sigma * u$ :

$$\partial_t u = -\text{sgn}(v_{\eta\eta}) |\nabla u|$$

- more robust under noise

◆ Weickert 2003:

- replaced  $v_{\eta\eta}$  by  $v_{\mathbf{w}\mathbf{w}}$  where  $\mathbf{w}$  is the normalised dominant eigenvector of the structure tensor  $J_\rho(\nabla u)$ :

$$\partial_t u = -\text{sgn}(v_{\mathbf{w}\mathbf{w}}) |\nabla u|$$

- enhances line-like structures (*coherence-enhancing shock filtering*).

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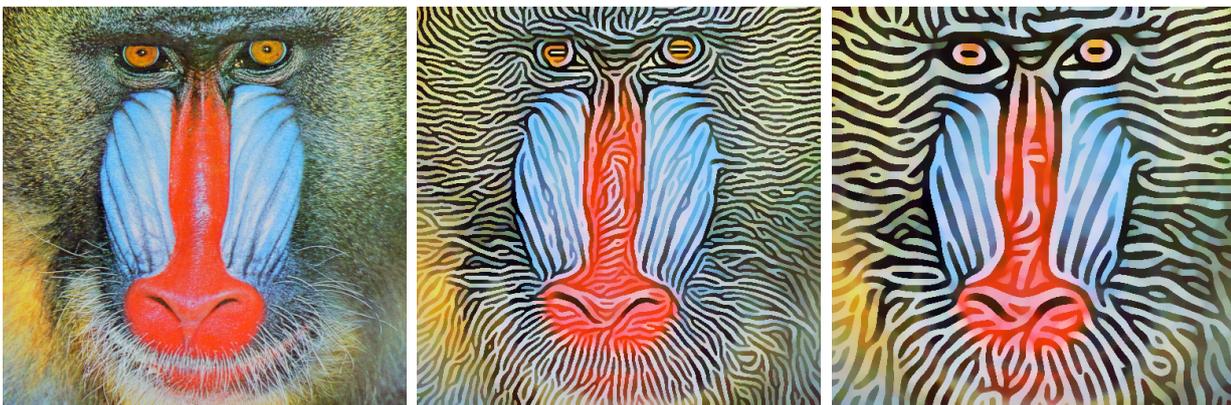
## Shock Filtering (4)



(a) **Left:** Fingerprint image,  $186 \times 186$  pixels. (b) **Middle:** Steady state using the Alvarez–Mazorra shock filter with  $\sigma = 1.5$ . (c) **Right:** Steady state using coherence-enhancing shock filtering with  $\sigma = 1.5$  and  $\rho = 5$ .

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## Shock Filtering (5)



Increasing the structure scale  $\sigma$  in coherence-enhancing shock filtering may create artistic effects. (a) **Left:** Mandrill,  $512 \times 512$  pixels. (b) **Middle:** Coherence-enhancing shock filtering,  $\sigma = 2$ ,  $\rho = 5$ ,  $t = 10$ . (c) **Right:** Ditto with  $\sigma = 4$ .

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(a) **Left:** Original image of Professor Albrecht,  $480 \times 640$  pixels. (b) **Right:** Coherence-enhancing shock filtering,  $\sigma = 2$ ,  $\rho = 4$ ,  $t = 50$ .

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## Nonflat Morphology

### ◆ Motivation:

- Morphological dilation/erosion with a structuring element resembles linear convolution with a box function.
- However, convolution also allows nonflat functions as kernels.
- Is there a morphological equivalent, such as a nonflat structuring element ?

### ◆ Consider the *structuring function (SF)*

$$b(\mathbf{x}) : \mathbb{R}^2 \rightarrow \mathbb{R}$$

instead of a structuring element  $B$ .

### ◆ Morphological operations on some image $f(\mathbf{x})$ with a structuring function $b(\mathbf{x})$ :

$$\text{dilation:} \quad (f \oplus b)(\mathbf{x}) := \sup \{f(\mathbf{x}-\mathbf{y}) + b(\mathbf{y}) \mid \mathbf{y} \in \mathbb{R}^2\}.$$

$$\text{erosion:} \quad (f \ominus b)(\mathbf{x}) := \inf \{f(\mathbf{x}+\mathbf{y}) - b(\mathbf{y}) \mid \mathbf{y} \in \mathbb{R}^2\}.$$

### ◆ The use of structuring functions renounces morphological invariance.

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## Nonflat Morphology (2)



### Connections to Flat Morphology and Structuring Elements

- Identifying a structuring element  $B$  with the “flat” structuring function

$$b(\mathbf{x}) := \begin{cases} 0 & (\mathbf{x} \in B), \\ -\infty & (\mathbf{x} \notin B). \end{cases}$$

transfers the novel definition

dilation:  $(f \oplus b)(\mathbf{x}) := \sup \{f(\mathbf{x}-\mathbf{y}) + b(\mathbf{y}) \mid \mathbf{y} \in \mathbb{R}^2\}.$

erosion:  $(f \ominus b)(\mathbf{x}) := \inf \{f(\mathbf{x}+\mathbf{y}) - b(\mathbf{y}) \mid \mathbf{y} \in \mathbb{R}^2\}.$

to the original definition:

dilation:  $(f \oplus b)(\mathbf{x}) := \sup \{f(\mathbf{x}-\mathbf{y}) \mid \mathbf{y} \in B\},$

erosion:  $(f \ominus b)(\mathbf{x}) := \inf \{f(\mathbf{x}+\mathbf{y}) \mid \mathbf{y} \in B\}.$

*This generalisation does not harm, but is useful?*

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## Nonflat Morphology (3)



### The Slope Transform

- The new definition of dilation is commutative:

$$f \oplus b = b \oplus f$$

Similar to convolution:  $f * b = b * f$

- Dilation in morphology resembles convolution in linear system theory (LST). In LST, the Fourier transform maps convolution to multiplication. Is there a similar transformation for dilation in morphology?

- Yes, if one considers *tangential dilation*:

$$(f \check{\oplus} b)(\mathbf{x}) := \operatorname{stat}_{\mathbf{y}} \left( f(\mathbf{x}-\mathbf{y}) + b(\mathbf{y}) \right)$$

where the *stationary values*  $\operatorname{stat}_{\mathbf{z}} f(\mathbf{z}) := \{f(\mathbf{z}) \mid \nabla f(\mathbf{z}) = \mathbf{0}\}$  contain the grey values at extrema.

- Tangential dilation is equivalent to usual dilation if  $f$  and  $b$  are concave (since the extrema are maxima then).

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## Nonflat Morphology (4)



- ◆ Dorst/van den Boomgaard and Maragos (1994) showed:  
The morphological equivalent to the Fourier transform is the *slope transform*

$$\mathcal{S}[f](\omega) := \operatorname{stat}_{\mathbf{x}} \left( f(\mathbf{x}) - \omega^{\top} \mathbf{x} \right)$$

Note the almost logarithmic connection to the Fourier transform (FT):

$$\mathcal{F}[f](\omega) := \int \left( f(\mathbf{x}) \exp(-i\omega^{\top} \mathbf{x}) \right) d\mathbf{x}$$

- ◆ The slope transform maps tangential dilation to addition:

$$\mathcal{S}[f \oplus b] = \mathcal{S}[f] + \mathcal{S}[b]$$

while the Fourier transform maps convolution to multiplication:

$$\mathcal{F}[f * b] = \mathcal{F}[f] \cdot \mathcal{F}[b]$$

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## Nonflat Morphology (5)



- ◆ The backtransformation to the slope transformation is given by

$$f(\mathbf{x}) = \operatorname{stat}_{\omega} \left( \mathcal{S}[f](\omega) + \omega^{\top} \mathbf{x} \right).$$

Backtransformation to the Fourier transformation:

$$f(\mathbf{x}) = \frac{1}{2\pi} \int \left( \mathcal{F}[f](\omega) \exp(i\omega^{\top} \mathbf{x}) \right) d\omega.$$

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## Nonflat Morphology (6)

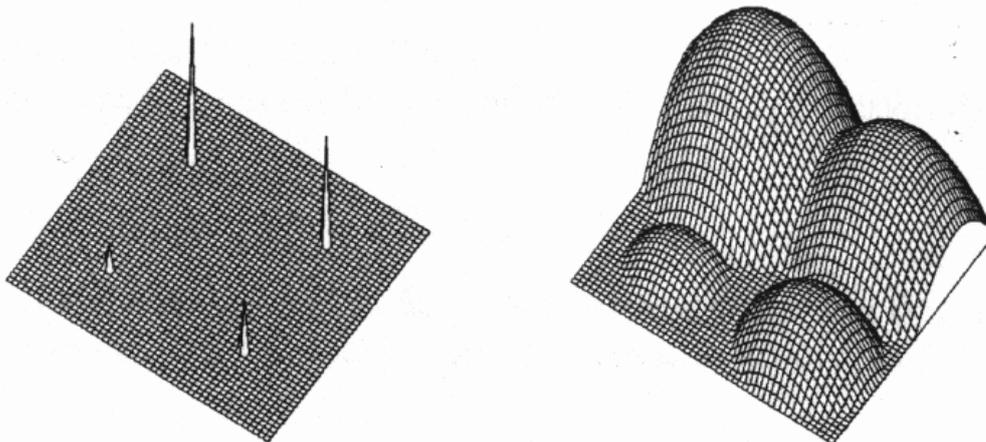


### Is There a Morphological Equivalent to Gaussian Convolution?

- ◆ In LST, convolution with Gaussians plays a fundamental role:
  - Gaussians remain Gaussians under the Fourier transform.
  - Gaussians are the only separable and rotationally invariant convolution kernels.
- ◆ Similar results in morphology (van den Boomgaard 1992):
  - Parabolas remain parabolas under the slope transform.
  - Parabolas are the only separable and rotationally invariant structuring functions.

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## Nonflat Morphology (7)



A 2-D signal and its dilation with a paraboloid as structuring function. **Author:** van den Boomgaard (1992).

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## Nonflat Morphology (8)

### PDE Formulation

- ◆ van den Boomgaard (1992) showed:  
Dilation of  $f(\mathbf{x})$  with a *quadratic structuring function (QSF)*

$$b(\mathbf{x}, t) = -\frac{|\mathbf{x}|^2}{4t} \quad (t > 0)$$

is equivalent to solving the differential equation

$$\partial_t u = |\nabla u|^2$$

with initial value  $f(\mathbf{x})$ .

- ◆ Erosion with a QSF is governed by

$$\partial_t u = -|\nabla u|^2.$$

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## Nonflat Morphology (9)



(a) **Top row:** Evolution under linear diffusion. From left to right:  $t = 0, 5, 10, 15$ . (b) **Middle row:** Dilation with a QSF:  $t = 0, 0.25, 1, 4$ . (c) **Bottom row:** Erosion with a QSF:  $t = 0, 0.25, 1, 4$ .

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## Nonflat Morphology (10)



### A Simple Algorithm for Dilations with QSFs

- ◆ separability: only 1D dilations necessary
- ◆ does not require to solve a PDE in an iterative manner
- ◆ uses direct implementation of the definition of dilation via the sup-operator
- ◆ exact, noniterative method for arbitrarily large times  $t$

```
for (i=0; i<=N-1; i++)
{
  max = f[i];
  for (j=0; j<=N-1; j++)
  {
    help = f[j] - (i - j) * (i - j) / (4.0 * t);
    if (help > max) max = help;
  }
  u[i] = max;
}
```

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## Nonflat Morphology (11)



### Application of QSFs: Euclidean Distance Transformation

(van den Boomgaard 1992)

- ◆ given: binary image  $f(\mathbf{x})$
- ◆ set foreground to 0 and background to  $\infty$
- ◆ erode with a quadratic structuring function  $b(\mathbf{x}, t) = -\mathbf{x}^2$
- ◆ gives the squared distance function
- ◆ noniterative, “exact” method
- ◆ tip for better visualisation:  
periodic repetition of the grey scale after a few pixels

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(a) **Top left:** Image with a single object pixel in the centre. (b) **Top right:** Euclidean distance transformation of (a). The greyscale of the distances is repeated periodically after 16 pixels. (c) **Bottom left:** Test image. (d) **Bottom right:** Euclidean distance transformation of (c).

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## Relations to Linear System Theory

(Burgeth/W. 2005)

There is an almost logarithmic connection between

*linear system theory* and *morphological system theory*

$$\mathcal{F}[f](u) \longleftrightarrow \mathcal{S}[f](u)$$

$$\mathcal{F}[f * b] = \mathcal{F}[f] \cdot \mathcal{F}[b] \longleftrightarrow \mathcal{S}[f \oplus b] = \mathcal{S}[f] + \mathcal{S}[b]$$

$$\frac{1}{(2\pi\sigma^2)^{n/2}} e^{-\frac{\mathbf{x}^\top \mathbf{x}}{2\sigma^2}} \longleftrightarrow -\frac{\mathbf{x}^\top \mathbf{x}}{4t}$$

What is the reason for this connection?

Is there a common basis for the two system theories?

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## Relations to Linear System Theory (2)

### The (max,+)- and the (min,+)-Algebra

Define (max,+)-algebra  $\mathbb{R}_{max}$  and (min,+)-algebra  $\mathbb{R}_{min}$  via the table

name	set	addition	multiplication
standard algebra $\mathbb{R}$	$\mathbb{R}$	+	$\times$
(max,+)-algebra $\mathbb{R}_{max}$	$\mathbb{R} \cup \{-\infty\}$	max	+
(min,+)-algebra $\mathbb{R}_{min}$	$\mathbb{R} \cup \{+\infty\}$	min	+

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## Relations to Linear System Theory (3)

Regarding the signal  $f$  as a mapping into these algebras replaces the conventional definition of convolution

$$(f * g)(\mathbf{x}) = \int_{\mathbb{R}^n} f(\mathbf{x}-\mathbf{y}) g(\mathbf{y}) d\mathbf{x}$$

by *new definitions of morphological convolutions*:

$$(f *_d g)(\mathbf{x}) := \sup_{\mathbf{y} \in \mathbb{R}^n} (f(\mathbf{x}-\mathbf{y}) + g(\mathbf{y})),$$

$$(f *_e g)(\mathbf{x}) := \inf_{\mathbf{y} \in \mathbb{R}^n} (f(\mathbf{x}-\mathbf{y}) + g(\mathbf{y})).$$

They are closely connected to dilation  $\oplus$  and erosion  $\ominus$  (with  $\bar{g}(\mathbf{x}) := -g(-\mathbf{x})$ ):

$$(f \oplus g)(\mathbf{x}) = \sup_{\mathbf{y} \in \mathbb{R}^n} (f(\mathbf{x}-\mathbf{y}) + g(\mathbf{y})) = (f *_d g)(\mathbf{x}),$$

$$(f \ominus g)(\mathbf{x}) = \inf_{\mathbf{y} \in \mathbb{R}^n} (f(\mathbf{x}+\mathbf{y}) - g(\mathbf{y})) = (f *_e \bar{g})(\mathbf{x}).$$

*Thus, morphology is linear system theory in another algebra.*

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## Summary

- ◆ Shock filters are morphological image enhancement methods that use dilation around maxima and erosion around minima.
- ◆ They may be modified in many ways, allowing even to enhance coherent, flow-like structures.
- ◆ The classical “flat” morphology with structuring elements may be generalised to nonflat morphology with structuring functions.
- ◆ Morphology with quadratic structuring functions provides an elegant way to compute Euclidean distance transformations.
- ◆ The resulting morphological system theory resembles linear system theory:
  - The morphological equivalent to convolution is (tangential) dilation.
  - The morphological equivalent to the Fourier transform is the slope transform.
  - The morphological equivalent to Gaussian convolution kernels are quadratic structuring functions.
- ◆ Morphology may be regarded as linear system theory in another algebra: the maxplus algebra resp. minplus algebra.

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*(introduced the slope transform together with Dorst / and den Boomgaard)*
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*(maxplus algebra and Cramer transform to unify linear and morphological systems)*

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