

Medical Image Processing with Orientation Scores

Erik Franken*, Remco Duits,
Markus van Almsick,
Bart ter Haar Romeny

*E-mail: e.m.franken@tue.nl

Eindhoven University of Technology
Department of Biomedical Engineering

Mathematica Technology Conference 2006
October 13th 2006, Champaign

1

Outline

- About our Research Group
- Orientation Scores
- Diffusion in Orientation Scores
- Stochastic Completion Fields
- Using Mathematica

2

The Biomedical Image Analysis group



3

Mathematica in the BioMIM lab



Starring:
Petr Šereda (Pilsen, Czech Republic)
Tim Peeters (Echt, The Netherlands)

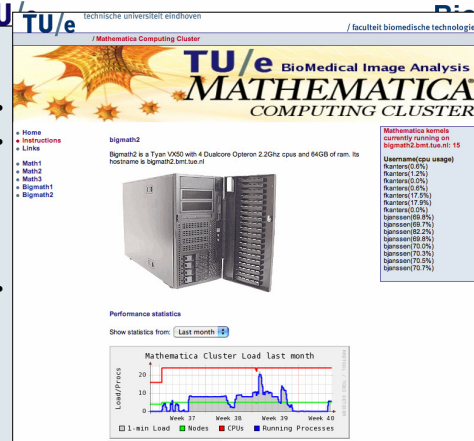
4

Our Mathematica Infrastructure

- Full campus license for Mathematica
- The need for “bigmath” kernel servers
 - Bigmath1: Tyan TX46, 4x Opteron 2.2Ghz, 32GB
 - Bigmath2: Tyan VX50, 4x Dualcore Opteron 2.2Ghz, 64GB
 - + 3 older servers
- Use of ParallelMathematica



5



6

TU/e BioMIM

MathVisionTools

Computer Vision Library for Mathematica:

- Gaussian derivatives
- Geometry driven diffusion
- Orientation score functions
- Image transformations
- DICOM import/export

www.mathvisiontools.net

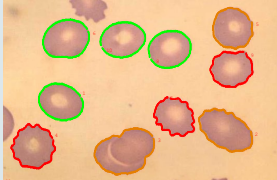
7

TU/e BioMIM

Mathematica in Bachelor Education

Image Analysis for Pathology.

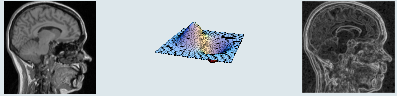
- Groups of 8 2nd year students
- “Invent” image analysis algorithms in *Mathematica*
- Competitive element
- 6 weeks project



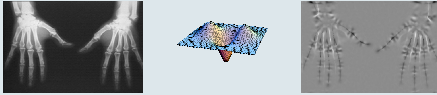
8

TU/e BioMIM

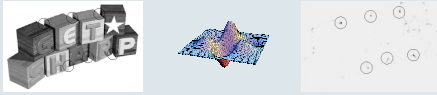
Example: Differential invariants



1st order
(edges)



2nd order
(ridges)



3rd order
(T-junctions)

For example
Rotation invariant
T-junction detection:

$$\frac{1}{(\frac{1}{2} + \frac{1}{3})^2} (-1 \cdot 1 + 1 \cdot 1 + 1 \cdot 1 + 1 \cdot 1 - 1 \cdot 1 - 1 \cdot 1 - 1 \cdot 1 - 1 \cdot 1) + \dots$$

9

TU/e BioMIM

Outline

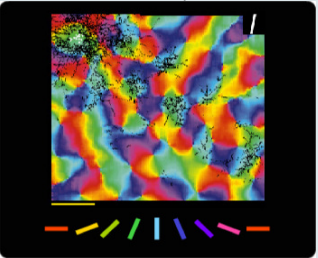
- About our Research group
- **Orientation Scores**
- Diffusion in Orientation Scores
- Stochastic Completion Fields
- Using Mathematica

10

TU/e BioMIM

Biological Inspiration

1. The retina contains receptive fields of varying sizes → *multi-scale* sampling device
2. Primary visual cortex is *multi-orientation* Measurement in Primary Visual Cortex



- Cells in the primary visual cortex are orientation-specific
- Strong connectivity between cells that respond to (nearly) the same orientation

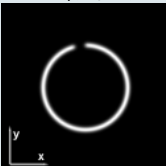
Bosking et al., J. Neuroscience 17:2112-2127, 1997

11

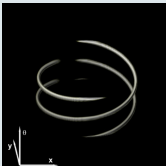
TU/e BioMIM

Orientation Scores

From 2D image $f(x,y)$ to orientation score $U_f(x,y,\theta)$ with position (x,y) and orientation θ



$f(x,y) \xrightarrow{W_\psi} U_f(x,y,\theta)$



$$(W_\psi[f])(\mathbf{x}, \theta) = U_f(\mathbf{x}, \theta) = \int_{\mathbb{R}^2} \psi(R_\theta^{-1}(\mathbf{x}' - \mathbf{x})) f(\mathbf{x}') d\mathbf{x}'$$

$$R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

An orientation score is a function on the *Euclidean motion group*

12

TU/e BioMIM

Our approach: Image Processing via Orientation Scores

13

TU/e BioMIM

Invertible Orientation Score Transformation

Design considerations: reconstruction, directional, spatial localization, quadrature

```
os1 = CKOrientationScoreTransform[img1, k, so, q, t, ss];
img1back = Plus@@os1;
```

14

TU/e BioMIM

Outline

- About our Research group
- Orientation Scores
- Diffusion in Orientation Scores**
- Stochastic Completion Fields
- Using Mathematica

15

TU/e BioMIM

The Diffusion Equation on Images

$$\begin{cases} \frac{\partial}{\partial t} u = \nabla \cdot \mathbf{D} \nabla u & f = \text{image} & \nabla = \left(\frac{\partial}{\partial x} \quad \frac{\partial}{\partial y} \right)^T \\ u(\mathbf{x}; 0) = f(\mathbf{x}) & u = \text{scale space of image} \\ \mathbf{D} = \text{diffusion tensor} & \mathbf{D} = \text{diffusion tensor} \end{cases}$$

Linear diffusion: $\mathbf{D} = \mathbf{I}$

Perona&Malik: $\mathbf{D}(\mathbf{x}) = g(|\nabla u(\mathbf{x})|) \mathbf{I}$

Coherence-enhancing diff.: $\mathbf{D}(\mathbf{x}) = g(\mathbf{x}) \mathbf{v}(\mathbf{x}) \mathbf{v}^T(\mathbf{x}) + \alpha \mathbf{I}$

$t = 0$

16

TU/e BioMIM

The Diffusion Equation on Images

$$\begin{cases} \frac{\partial}{\partial t} u = \nabla \cdot \mathbf{D} \nabla u & f = \text{image} & \nabla = \left(\frac{\partial}{\partial x} \quad \frac{\partial}{\partial y} \right)^T \\ u(\mathbf{x}; 0) = f(\mathbf{x}) & u = \text{scale space of image} \\ \mathbf{D} = \text{diffusion tensor} & \mathbf{D} = \text{diffusion tensor} \end{cases}$$

Linear diffusion: $\mathbf{D} = \mathbf{I}$

Perona&Malik: $\mathbf{D}(\mathbf{x}) = g(|\nabla u(\mathbf{x})|) \mathbf{I}$

Coherence-enhancing diff.: $\mathbf{D}(\mathbf{x}) = g(\mathbf{x}) \mathbf{v}(\mathbf{x}) \mathbf{v}^T(\mathbf{x}) + \alpha \mathbf{I}$

$t = 10$

17

TU/e BioMIM

Diffusion in orientation scores

curvature

Evolution of orientation score

$$\partial_t u = \begin{pmatrix} \partial_\theta & \partial_\xi & \partial_\eta \end{pmatrix} \begin{pmatrix} D_{11}' + D_{22}\kappa^2 & D_{22}\kappa & 0 \\ D_{22}\kappa & D_{22} & 0 \\ 0 & 0 & D_{33} \end{pmatrix} \begin{pmatrix} \partial_\theta \\ \partial_\xi \\ \partial_\eta \end{pmatrix} u$$

Left-invariant derivatives: $\partial_\xi = \cos \theta \partial_x + \sin \theta \partial_y$, $\partial_\eta = -\sin \theta \partial_x + \cos \theta \partial_y$

$\partial_\xi, \partial_\eta,$ and ∂_θ are left-invariant derivatives on Euclidean motion group, i.e. $\mathcal{L}_g \partial_i U = \partial_i \mathcal{L}_g U, i \in \{\xi, \eta, \theta\}$

Rotating tangent space coordinate basis

18

TU/e BioMIM

Example diffusion kernels

$$\partial_t u = (\partial_\theta \ \partial_\xi \ \partial_\eta) \begin{pmatrix} D'_{11} + D_{22}\kappa^2 & D_{22}\kappa & 0 \\ D_{22}\kappa & D_{22} & 0 \\ 0 & 0 & D_{33} \end{pmatrix} \begin{pmatrix} \partial_\theta \\ \partial_\xi \\ \partial_\eta \end{pmatrix} u$$

19

TU/e BioMIM

How to Choose Conductivity Coefficients

$$\partial_t u = (\partial_\theta \ \partial_\xi \ \partial_\eta) \begin{pmatrix} D'_{11} + D_{22}\kappa^2 & D_{22}\kappa & 0 \\ D_{22}\kappa & D_{22} & 0 \\ 0 & 0 & D_{33} \end{pmatrix} \begin{pmatrix} \partial_\theta \\ \partial_\xi \\ \partial_\eta \end{pmatrix} u$$

- Oriented regions: D'_{11} and D_{33} small, D_{22} large and κ according to estimate
- Non-oriented regions: D'_{11} large, $D_{22}=D_{33}$ large, $\kappa = 0$

20

TU/e BioMIM

Results

Diffusion in orientation score $t = 0$ Coherence enhancing diffusion

Size: 128 x 128 x 64

21

TU/e BioMIM

Results

Diffusion in orientation score $t = 10$ Coherence enhancing diffusion

Size: 128 x 128 x 64

22

TU/e BioMIM

Collagen image

Size: 200 x 200 x 64

$t = 0$ Diffusion in orientation score Coherence enhancing diffusion

23

TU/e BioMIM

Collagen image

Size: 200 x 200 x 64

$t = 30$ Diffusion in orientation score Coherence enhancing diffusion

24

TU/e BioMIM

Outline

- About our Research group
- Orientation Scores
- Diffusion in Orientation Scores
- Stochastic Completion Fields**
- Using Mathematica

25

TU/e BioMIM

Other PDE: the stochastic completion field

Resolvent of linear PDE
 $\partial_t u = (A - \lambda I)u$ with $A = (-\partial_x^2 + D_{1,1} \partial_{pp})$

It renders probability density field for line continuation based on random walker prior

26

TU/e BioMIM

Convolution on Orientation Scores

An image is a function on the translation group

An orientation score is a function on the Euclidean motion group

Normal convolution (on translation group)

$$[f * g](x) = \int_{\mathbb{R}^n} f(x-y) g(y) dy$$

G-convolution, where G is the Euclidean motion group

$$[K *_G U_f](x, \theta) = \int_{\mathbb{R}^2} \int_0^{2\pi} K(R_\theta^{-1}(x-x'), \theta - \theta') U_f(x', \theta') d\theta' dx'$$

27

TU/e BioMIM

Filling Gaps in Curves

(From M.Sc. thesis by Renske de Boer)

28

TU/e BioMIM

Enhancing edges in Medical images

Source image → Local information → Result

29

TU/e BioMIM

Outline

- About our Research group
- Orientation Scores
- Diffusion in Orientation Scores
- Stochastic Completion Fields
- Using Mathematica**

30

Using Mathematica

- Mathematica is helpful in solving the math (e.g. non-commuting operators)
- NDSolve in *mathematica* is not usable for our type of PDEs as far as I know
- PDE solver is written in C++, linked with Mathlink
- Typical problems of our PDE
 - Highly anisotropic, not aligned with grid
 - Non-commuting operators
 - Convection + diffusion

31

Acknowledgements

- Remco Duits
- Markus van Almsick
- Bart ter Haar Romeny
- Bart Janssen
- Arjen Ricksen
- Renske de Boer

For questions / more references about this work,
contact e.m.franken@tue.nl

32