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Image Processing and Computer Vision

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Winter Term 2007 / 2008
www.mia.uni-saarland.de/Teaching/ipcv07.shtml

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Organisational Issues (1)

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Organisational Issues

Welcome to this class!

Please do not hesitate to pose questions (in English or German).

Why is this Class Important?

- ◆ broad introduction into mathematically well-founded areas of image processing and computer vision
- ◆ important in numerous applications, e.g. medical imaging, computer-aided quality control, robotics, computer graphics, bioinformatics, multimedia and artificial intelligence.
- ◆ nice application area for almost all branches of mathematics
- ◆ qualify for starting a bachelor's thesis in our group

Organisational Issues (2)

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How Can I Make Use of These Classes?

- ◆ 4 hours classroom lectures, 2 hours tutorials (9 ECTS points)
- ◆ can be used in numerous study programmes:
 - **Visual Computing:**
 - class in the core area visual computing / subarea image analysis
 - **Computer Science:**
 - (theoretical) core class (Theorie-Stammvorlesung)
 - applied mathematics class if your minor is mathematics
 - **Mathematics:**
 - applied mathematics class
 - computer science class if your minor is computer science
 - **Computers and Communications:**
 - master core lecture
 - **Bioinformatics:**
 - as core lecture

Organisational Issues (3)

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What Prerequisites are Required?

- ◆ undergraduate mathematics (e.g. Mathematik für Informatiker I–III)
- ◆ elementary C knowledge (for the programming assignments)
- ◆ passive knowledge of English
(solutions in assignments or exam can be submitted in German)

Tutorials

- ◆ theoretical and programming assignments, alternating each week
- ◆ coordinated by Dr. Andrés Bruhn
- ◆ Tutors: Thomas Bühler, Oliver Demetz, Luis Pizarro, Sebastian Zimmer.
- ◆ 6 groups, 4 time slots: Tue + Wed, 2–4 pm, 4–6 pm
(1 German group, 1 group for honour's programme: Tue 4–6 pm)
- ◆ start: next week

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Grading Policy

- ◆ **Written exam: Friday, February 29, 2008 from 2:00 to 5:00 pm.**
Second chance: Wednesday, April 9, 2008 from 2:00 to 5:00 pm.
 You can take the lecture notes and the tutorial notes with you, but no books.
 The better grade counts.
- ◆ In order to qualify for both exams you need 50 % of all points from the assignments. Theoretical and practical assignments count for the final score.
- ◆ Working in groups of up to 3 people is permitted, but they must be in the same tutorial group.

Registration and Lecture Notes

- ◆ Please register for tutorials between Tue, Oct. 23, 3 pm and Fri, Oct. 26, 3 pm:
www.mia.uni-saarland.de/Teaching/ipcv07.shtml
- ◆ You can also download the lecture notes from there (password-protected).

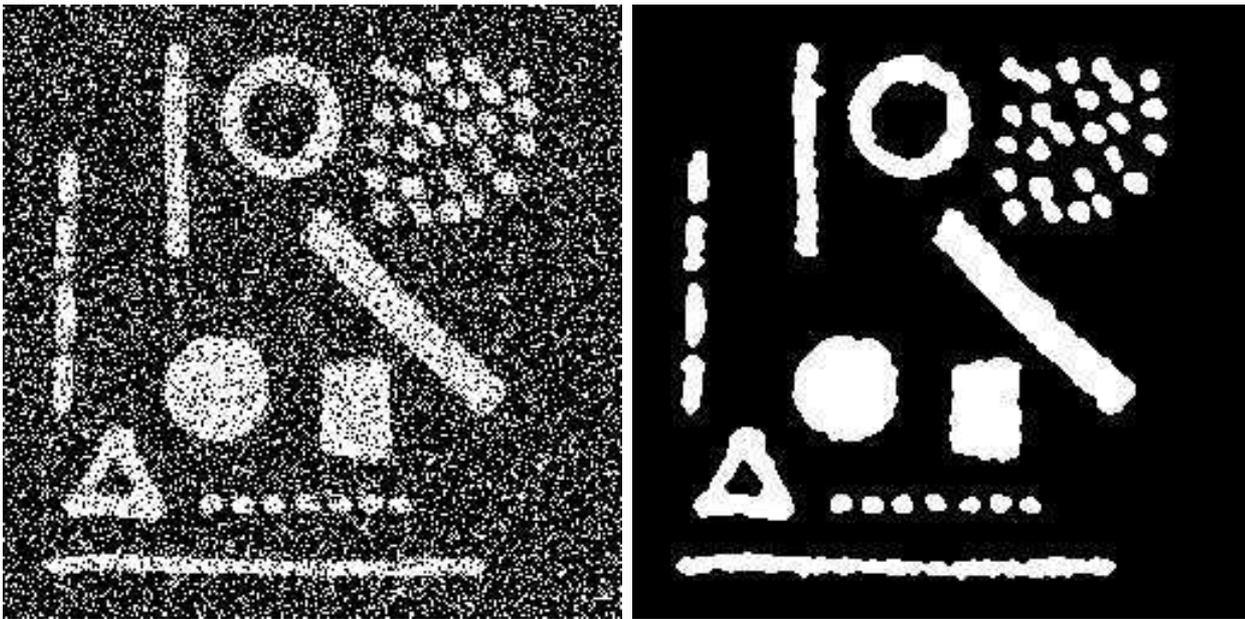
Important Areas within Visual Computing (1)

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Important Areas within Visual Computing

- ◆ **Image Processing (Bildverarbeitung):**
 A digital image is transformed into another digital image, that allows a better interpretation by humans or computers
Example: Denoising
- ◆ **Computer Vision, Image Understanding (Rechnersehen, Bildverstehen):**
 Extraction of information about the 3-D world from 2-D images.
Example: Stereographic reconstruction
- ◆ **Pattern Recognition (Mustererkennung):**
 Labelling of image structures to certain classes.
Examples: Character recognition, cell classification

Important Areas within Visual Computing (2)



Example for an image processing application. (a) **Left:** Noisy original image. (b) **Right:** Filtered. Author: J. Weickert (1999).

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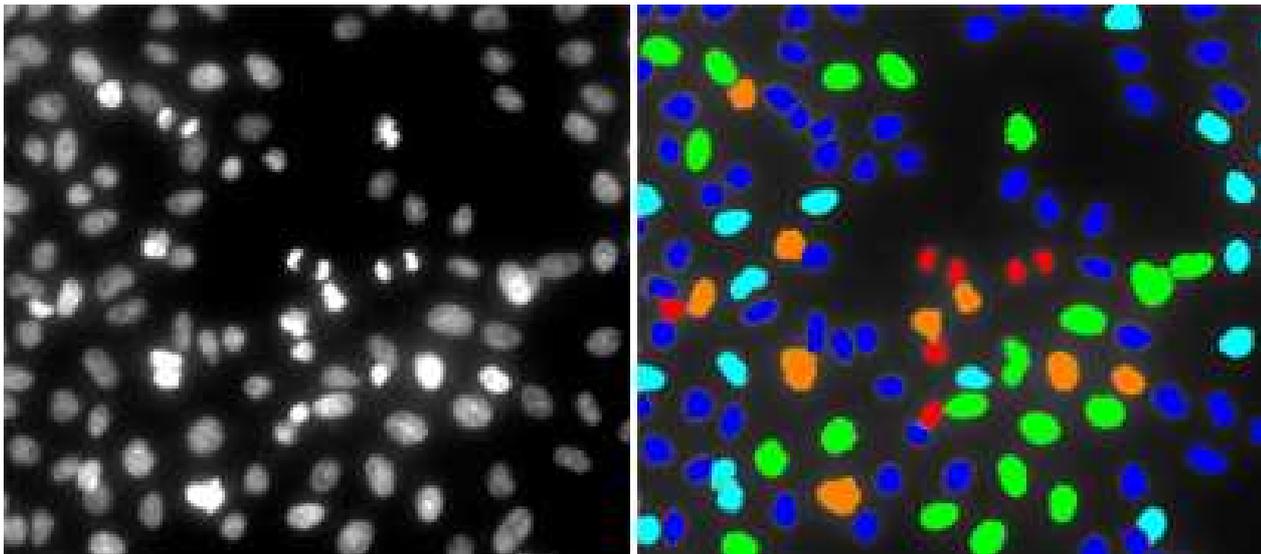
Important Areas within Visual Computing (3)



Example for a computer vision application. (a) **Left:** Stereo image pair. (b) **Right:** 3-D reconstruction. Authors: L. Alvarez, R. Deriche, J. Sánchez, J. Weickert (2002).

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Important Areas within Visual Computing (4)



Example for a pattern recognition application. (a) Left: Cells. (b) Right: Classification into different cell cycle phases. Source: MetaMorph Analysis, <http://www.moleculardevices.com/>.

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Important Areas within Visual Computing (5)

◆ Computer Graphics (Computergrafik):

Synthesis of a digital image that is supposed to look like a (usually realistic) image of a 3-D scene.

Example: Computer games

◆ Geometric Modelling (Geometrische Modellierung):

Mathematical description of curves and surfaces for synthesising 2-D and 3-D objects on a computer

Example: CAD-based design of a car coachwork

◆ Imaging (Bildgebende Verfahren):

deals with all aspects of image creation

Example: Tomographic reconstruction

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Important Areas within Visual Computing (6)

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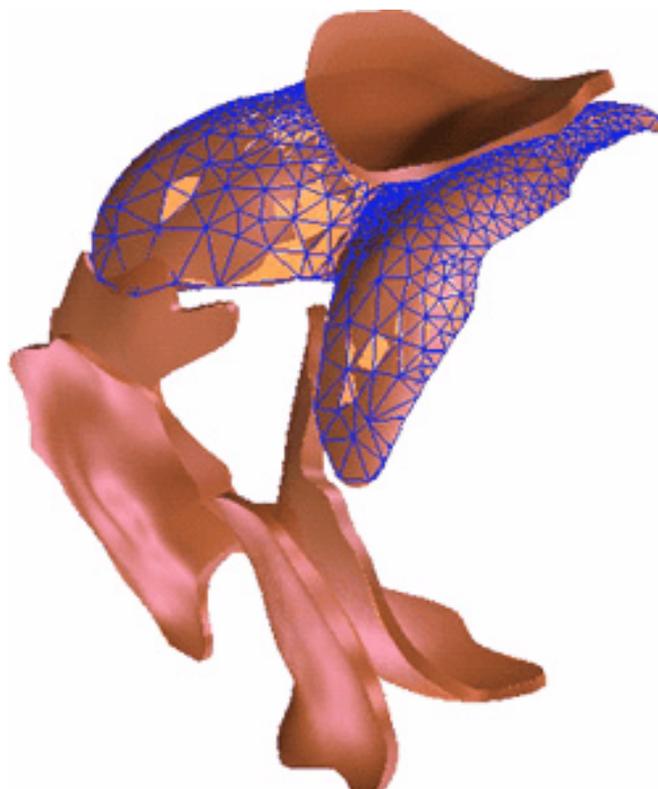


Example for a computer graphics application: Realistically looking simulation of a forest using ray tracing. Author: P. Slusallek.

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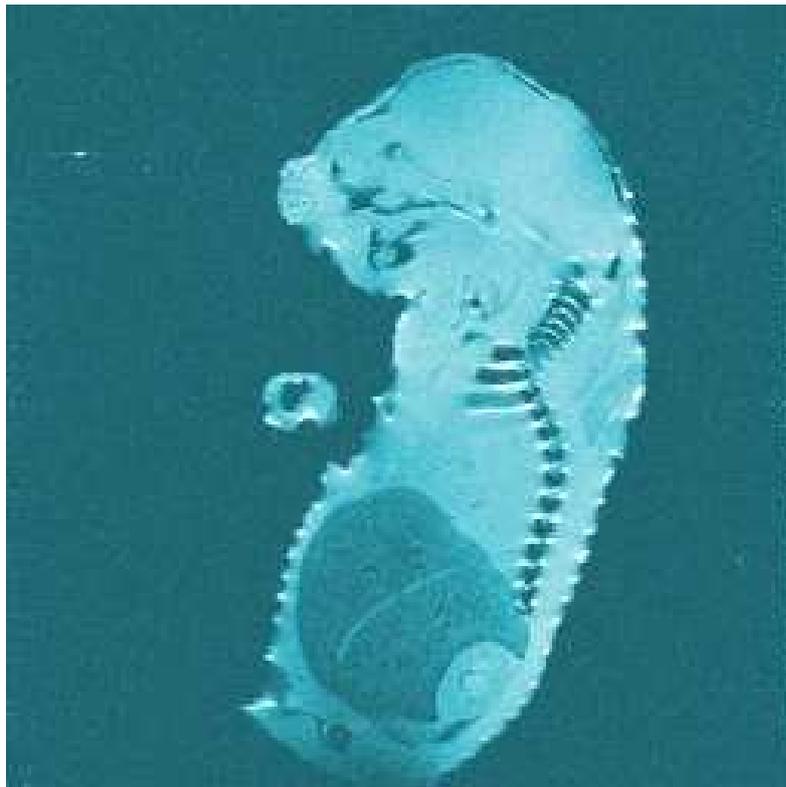
Important Areas within Visual Computing (7)

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Example for a geometric modelling application: A model of a tissue in a human knee. Source: http://www.scorec.rpi.edu/research_biomechanical.html.

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Example for an imaging application: Magnetic resonance image of a mouse embryo. Author: F. Volke.

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Collective Terms

◆ **Image Analysis (Bildanalyse):**

includes image processing, computer vision, pattern recognition

◆ **Image Synthesis (Bildsynthese):**

includes computer graphics and geometric modelling

◆ **Visual Computing (Computervisualistik):**

comprises everything related to imaging and the analysis and synthesis of digital images

This class focuses on image analysis, in particular on image processing and computer vision.

There is a growing confluence between image analysis and synthesis. In problems such as human motion capturing they even appear simultaneously.

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Planned Contents

- 1. Basics**
 - 1.1 Image Types and Discretisation
 - 1.2 Degradations in Digital Images
- 2. Image Transformations**
 - 2.1 Fourier Transform
 - 2.2 Image Pyramids
 - 2.3 Wavelet Transform
- 3. Colour Perception and Colour Spaces**
- 4. Image Compression**
- 5. Image Interpolation**
- 6. Image Enhancement**
 - 6.1 Point Operations
 - 6.2 Linear Filtering
 - 6.3 Wavelet Shrinkage, Median Filtering, M-Smothers
 - 6.4 Mathematical Morphology
 - 6.5 Diffusion Filtering
 - 6.6 Variational Methods
 - 6.7 Deblurring

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- 7. Feature Extraction**
 - 7.1 Edges
 - 7.2 Corners
 - 7.3 Lines and Circles
- 8. Texture Analysis**
- 9. Segmentation**
 - 9.1 Classical Methods
 - 9.2 Variational Methods
- 10. Image Sequence Analysis**
 - 10.1 Local Methods
 - 10.2 Variational Methods
- 11. 3-D Reconstruction**
 - 11.1 Camera Geometry
 - 11.2 Stereo
 - 11.3 Shape-from-Shading
- 12. Object Recognition**
 - 11.1 PCA Methods
 - 11.2 Invariances

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References

We do **not** focus on a specific textbook:
 Our didactic concept differs from all of them.
 However, here are some of the more useful books:

Textbooks on Image Processing

- ◆ J. Bigun: *Vision with Direction*. Springer, Berlin, 2006.
One of most systematic books on image processing, with a clear focus on matrix representations. My favourite book on image processing.
- ◆ R. C. Gonzalez, R. E. Woods: *Digital Image Processing*. Addison-Wesley, Reading, Second Edition, 2002.
A classical book on image processing. Comprehensive and fairly well readable.
- ◆ K. D. Tönnies: *Grundlagen der Bildverarbeitung*. Pearson Studium, München, 2005.
One of the better German books on image processing. Contains no computer vision.

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Textbooks on Computer Vision

- ◆ E. Trucco, A. Verri: *Introductory Techniques for 3-D Computer Vision*. Prentice-Hall, Upper Saddle River, 1998.
Good selection of important aspects.
- ◆ R. Jain, R. Kasturi, B. G. Schunck: *Machine Vision*. McGraw-Hill, New York, 1995.
Simple introduction to computer vision (and a bit of image processing), but sometimes too simple.
- ◆ R. Klette, K. Schlüns, A. Koschan: *Computer Vision: Three-Dimensional Data from Images*. Springer, Singapore, 1998.
Good for 3-D reconstruction (stereo, shape-from-shading).

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Useful Internet Resources

◆ CV Online

<http://homepages.inf.ed.ac.uk/rbf/CVonline/>
 Online compendium on numerous image processing and computer vision topics.
 Often very useful.

◆ Computer Vision Homepage

<http://www-cgi.cs.cmu.edu/afs/cs/project/cil/ftp/html/vision.html>
 useful links for many computer vision problems, but needs some update

◆ Keith Price's Annotated Computer Vision Bibliography

<http://iris.usc.edu/Vision-Notes/bibliography/contents.html>
 excellent, when searching for specific references

◆ CiteSeer

<http://citeseer.ist.psu.edu/cs>
 numerous citations and online articles in all areas of computer science

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35	

Lecture 1:

Image Types and Discretisation

Contents

1. Sampling and Quantisation
2. Types of Images

M	I
A	
1	2
3	4
5	6
7	8
9	10
11	12
13	14
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Sampling and Quantisation

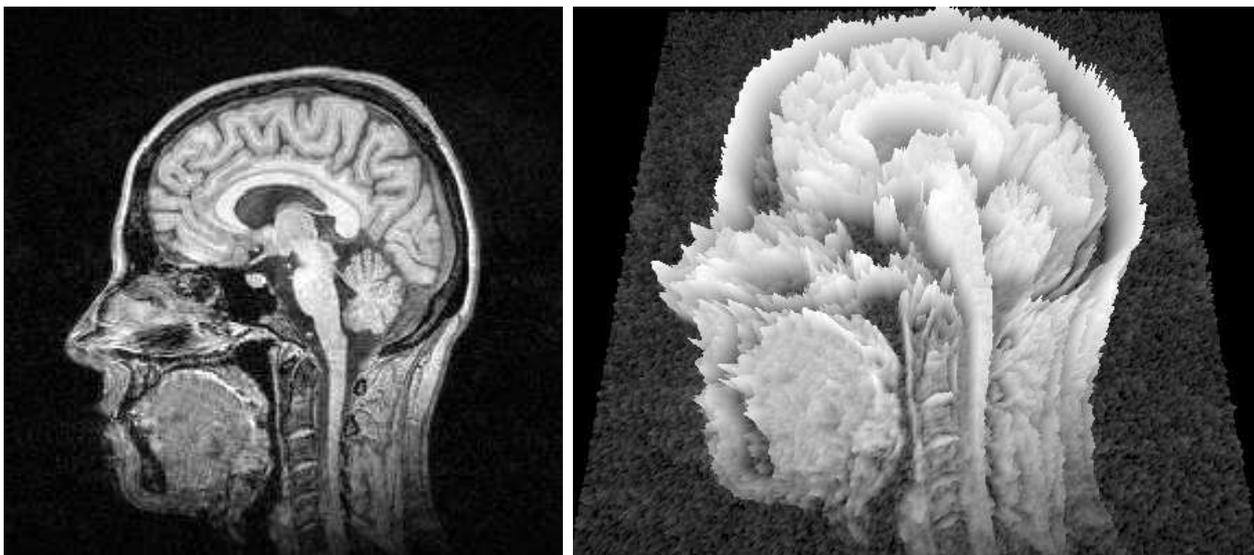
Continuous Greyscale Image (Kontinuierliches Grauwertbild):

- ◆ mapping f from a rectangular domain (Definitionsbereich) $\Omega = (0, a_1) \times (0, a_2)$ to a co-domain (Wertebereich) \mathbb{R} :

$$f : \mathbb{R}^2 \supset \Omega \rightarrow \mathbb{R}$$

- ◆ domain Ω is called *image domain* or *image plane*
- ◆ co-domain specifies *grey value*
- ◆ Usually low grey values are dark and high grey values bright.

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(a) **Left:** Magnetic resonance (MR) image of a human head. (b) **Right:** Representation as a function $f(x, y)$ over a rectangular image domain Ω . Authors: J. Weickert, C. Schnörr (2000).

Sampling and Quantisation (3)

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Sampling:

- ◆ discretisation of the *domain* Ω
- ◆ Image data are only given on a rectangular point grid within the image domain Ω .
- ◆ creates a digital image

$$\{f_{i,j} \mid i = 1, \dots, N; j = 1, \dots, M\}$$

- ◆ The grid point / cell (i, j) is called *pixel* (picture element).
- ◆ 2-D images usually have equal pixel distances in both directions.
- ◆ Image processing people often normalise these grid sizes to 1.
- ◆ If the sampling rate is too low, the image quality degrades severely.

Sampling and Quantisation (4)

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Digital test image with different sampling rates. **(a) Top left:** Sampled with 256×256 pixels. **(b) Top right:** 128×128 pixels. **(c) Bottom left:** 64×64 pixels. **(d) Bottom right:** 32×32 pixels. Author: J. Weickert (2000).

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Quantisation (Quantisierung):

- ◆ discretisation of the *co-domain*
- ◆ saves disk space
- ◆ if a grey values is coded by a single byte, the discrete co-domain is given by $\{0, 1, \dots, 255\}$
- ◆ Binary images have the co-domain $\{0, 1\}$.
- ◆ Humans can distinguish only 40 greyscales.
They are even very good in recognising the content of binary images.

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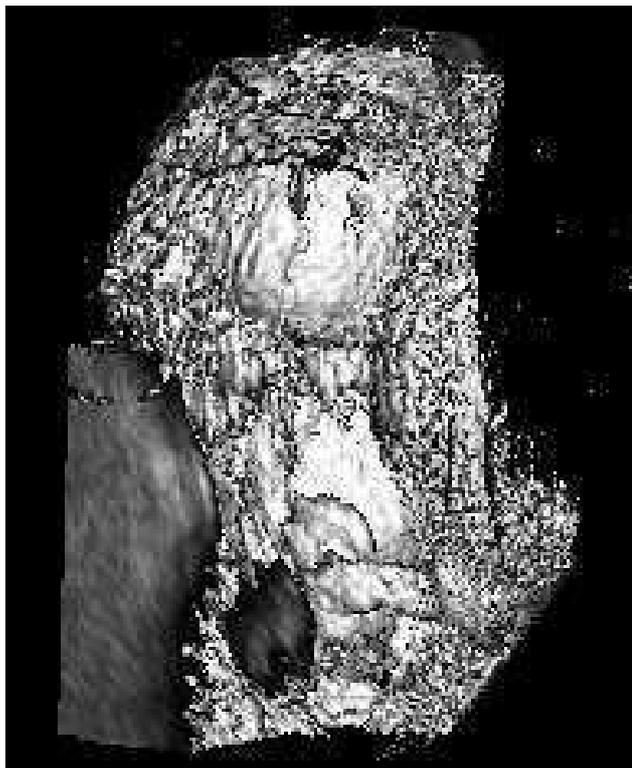


Digital test image (256×256 pixels) with different quantisation rates. **(a) Top left:** 256 greyscales. **(b) Top right:** 32 greyscales. **(c) Bottom left:** 8 greyscales. **(d) Bottom right:** 2 greyscales. Author: J. Weickert (2000).

Types of Images

m -Dimensional Images

- ◆ domain in \mathbb{R}^m
- ◆ $m = 1$: signals
- ◆ $m = 2$: (two-dimensional) images
- ◆ $m = 3$: three-dimensional images
 - important in medical imaging, e.g. computerised tomography (CT, Computertomographie), magnetic resonance imaging (MRI, Kernspintomographie).
 - image points/cells in 3-D are called *voxels* (volume elements).
 - voxel dimensions usually differ in different directions!



Rendering of a 3-D ultrasound image of a human fetus in its 10th week. Authors: J. Weickert, K. Zuiderveld, B.M. ter Haar Romeny, W. Niessen (1997).

Types of Images (3)

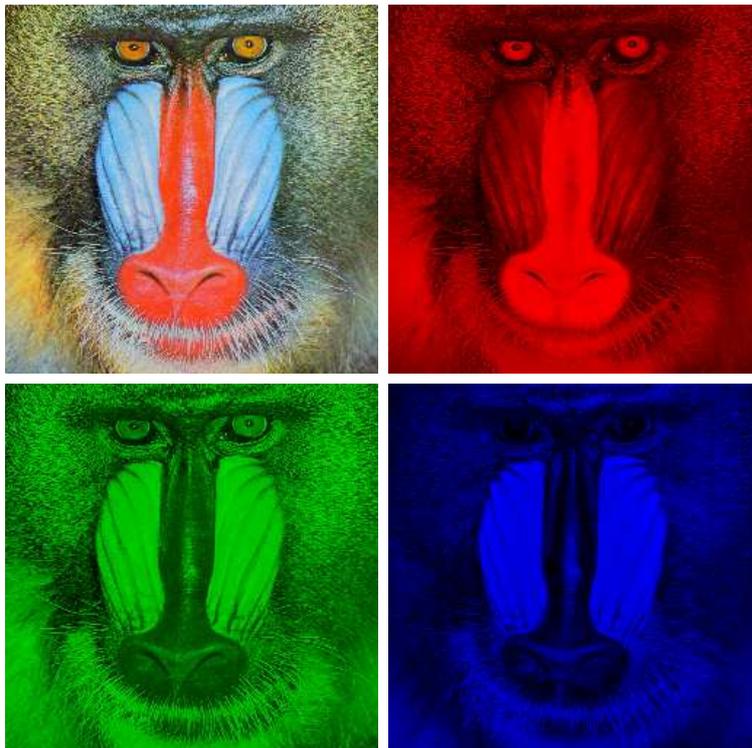
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Vector-Valued Images

- ◆ co-domain in \mathbb{R}^n , containing n channels
- ◆ Example 1: Colour images
Three channels: R (red), G (green), B (blue).
Humans can distinguish 2,000,000 colours!
- ◆ Example 2: Multispectral image (e.g. satellite images)
numerous channels (4–30) that represent different frequency bands

Types of Images (4)

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Colour image as example for a vector-valued image. (a) **Top left:** Original image. (b) **Top right:** Red channel. (c) **Bottom left:** Green channel. (d) **Bottom right:** Blue channel. Author: J. Weickert (2000).

Types of Images (5)

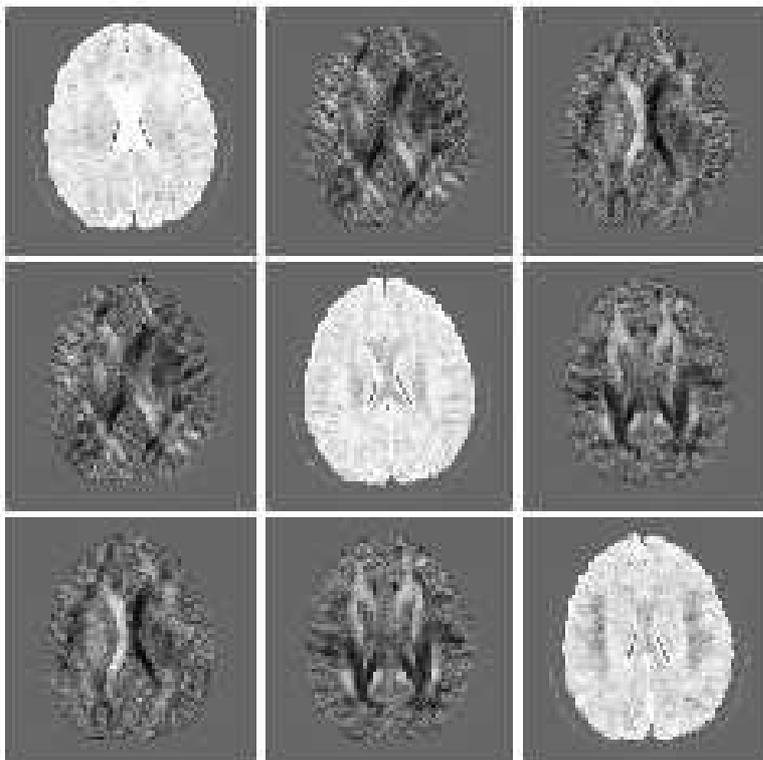
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Matrix-Valued Images

- ◆ co-domain in $\mathbb{R}^{n \times n}$
- ◆ Example: diffusion tensor MRI (DT-MRI, Diffusionstensor-Kernspintomographie): measures in each voxel of a 3-D image domain a symmetric positive definite 3×3 matrix
- ◆ may create additional constraints, e.g.:
A reasonable filter should not destroy relevant properties such as positive definiteness of the matrix field.

Types of Images (6)

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Images of the nine coefficients of a DT-MRI data set. Since the diffusion matrix is a symmetric 3×3 matrix, only 6 out of 9 images differ. Authors: D. Weinstein, G. Kindlmann, E. Lundberg (1999).

Types of Images (7)

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Image Sequences (Bildfolgen)

- ◆ One can consider image sequences for any of the above mentioned types of images.
- ◆ This increases the dimensionality of the domain from m to $m + 1$.
- ◆ Example: 3-D echo cardiography (Echokardiographie) creates a sequence of 3-D scalar-valued images (can be regarded as 4-D image)

Relevant Images in This Class

- ◆ We mainly focus on 2-D scalar images and their image sequences.
- ◆ Many methods can be generalised to other types of images.

Types of Images (8)

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Form left to right and from top to bottom: Four subsequent frames of an image sequence (the famous Hamburg taxi scene, 256×190 pixels). Can you recognise what has moved in which direction?
Source: http://i21www.ira.uka.de/image_sequences/

Summary

- ◆ Digital images have a discrete domain (sampling) and a discrete co-domain (quantisation)
- ◆ generalisation of the domain:
m-dimensional images, image sequences
- ◆ generalisation of the co-domain:
vector-valued images, matrix-valued images

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