

# Chapter 0

## Organization and Introduction

*Variational Methods for Computer Vision*  
WS 16/17

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# Organizational Stuff

# Requirements, or “is this something for me?”

## Necessary

- Interest in mathematical theory
- Solid background in analysis and linear algebra
- Numerics (Matlab)

## Nice to know

- Image processing and computer vision
- Optimization
- Functional analysis

## Exercises

- Exercise sheets covering the content of the lecture will be passed out every Monday
- Exercises contain theoretical as well as programming problems
- You have one week for the exercise sheets and can turn in your solution in the following week during Monday's lecture
- You may work on the exercises in groups of two
- **Reaching at least 50% of the total exercise points is a requirement for being admitted to the examination**
- If solutions have obviously been copied, both groups will get 0 points
- You will discuss the solution to the exercises with Jonas Geiping on Wednesday (14-16 o'clock, H-C 7326).



## Questions within the lecture

The more we discuss in the lecture, the more interesting the course will be! Please don't be shy to say something!

## Examination

- The final exam will be oral
- The lecture is worth 10 credits

## Miscellaneous

- My office: H-A 7106
- Jonas' office: H-A 7116
- Office hours: Please write an email.
- Lecture: Starts at 16:00 sharp. Short break in between.
- Course website:  
<http://www.vsa.informatik.uni-siegen.de/en/variational-methods-computer-vision>
- To access the course material: username "student"  
password "100%brain"

# Overview

## Chapter 1: Basics and necessary Tools

### Topics:

- Types of images and signals
- Discrete vs. continuous representations
- How simple things can fail - ill-posedness
- The general idea of variational methods
- A simple optimization method

### Goals:

- Establish a common ground
- Understand the motivation for variational methods
- Know a first way to implement them numerically

## Chapter 2: Linear inverse imaging problems

- Denoising
- TV-Regularization
- MAP-estimates and different data terms
- Non-local regularization
- Deblurring
- Super resolution
- Demosaicking
- Inpainting
- Patch-based regularization
- Non-convex regularizations

### Goals:

- A thorough understanding of how to formulate energy minimization problems
- Get to know different regularization methods

## Chapter 3: Non-linear and non-convex problems

- Image segmentation
- Stereo depth reconstruction
- Optical flow estimation
- 3D reconstruction
- Blind hyperspectral unmixing and matrix factorization

### Goals:

- Learn different ways to "fight" non-linearities
- Study several exciting problems

## Denoising



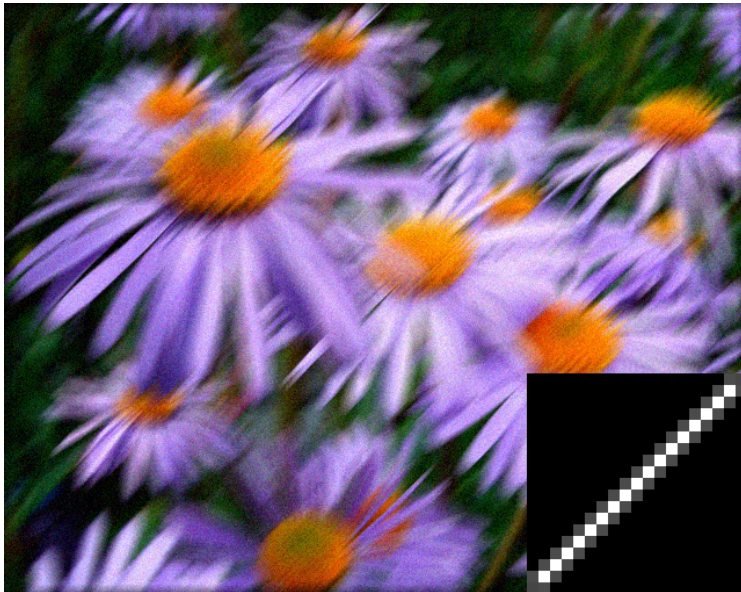
## Denoising





# An overview in images

## Deblurring



# An overview in images

## Deblurring



### Inpainting



## Inpainting

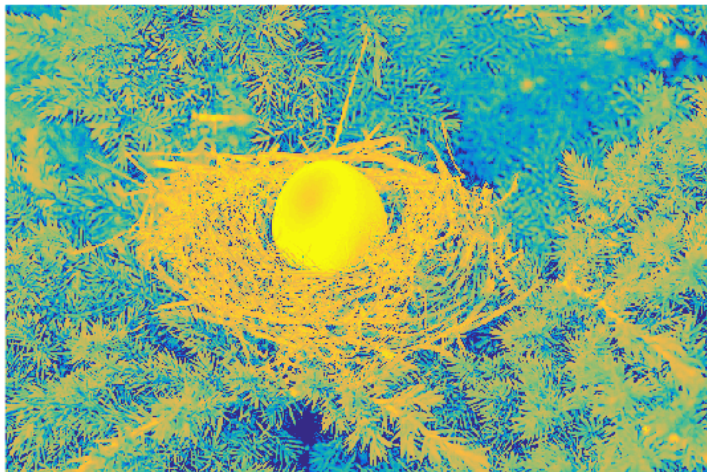


## Segmentation





## Segmentation



## Segmentation



## Stereo imaging



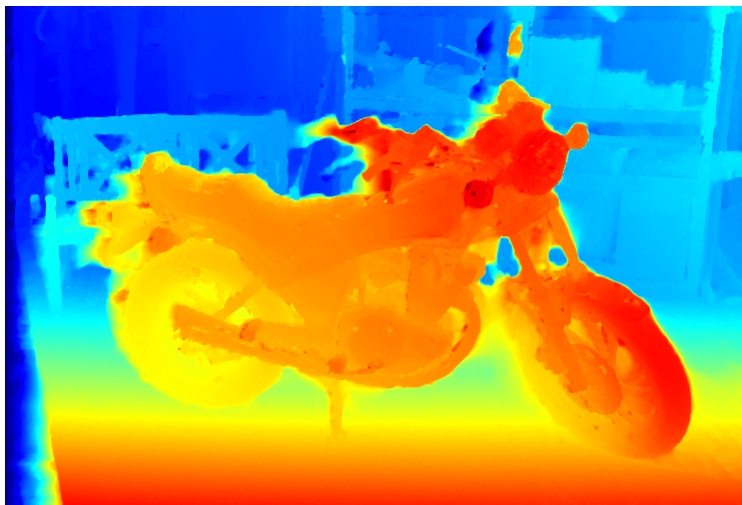


# An overview in images

## Stereo imaging



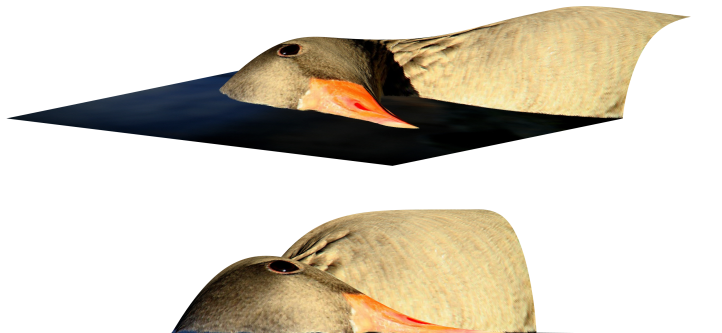
## Stereo imaging



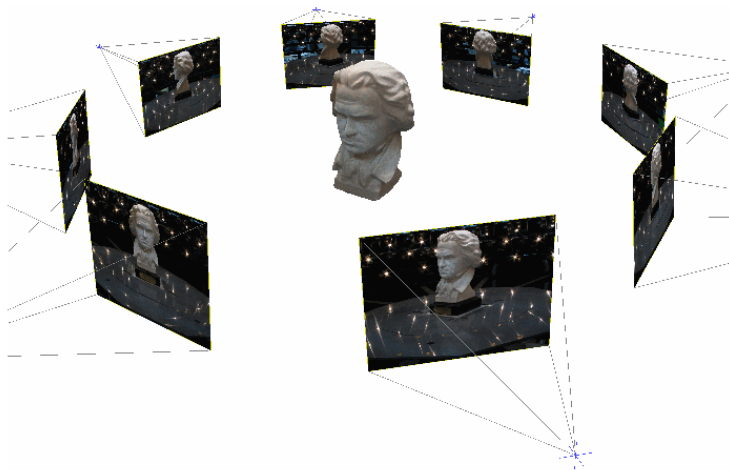
## 3D reconstruction



## 3D reconstruction



## 3D reconstruction



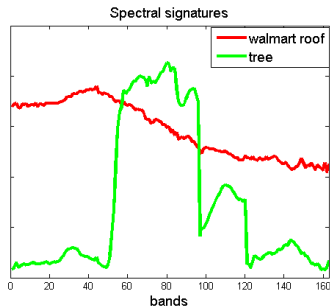
From

[http://vision.in.tum.de/research/image-based\\_3d\\_reconstruction/multiviewreconstruction](http://vision.in.tum.de/research/image-based_3d_reconstruction/multiviewreconstruction)

## Hyperspectral imaging



Hyperspectral cube with 163 bands



# An overview in images

## Hyperspectral imaging



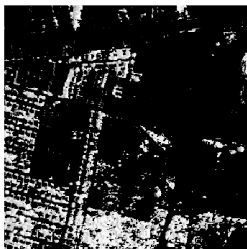
color image illustration



endmember "road"



endmember "roof"



endmember "trees"

- The lecture covers many different aspects
- This means we sometimes cannot go into many details
- This will be difficult for me!
- If you have the feeling I am covering more content by increasing the speed to an unreasonable level, stop me!

**Let's start!**