

# VISION-BASED ANALYSIS OF CONVENTIONAL SURGICAL PROCEDURES

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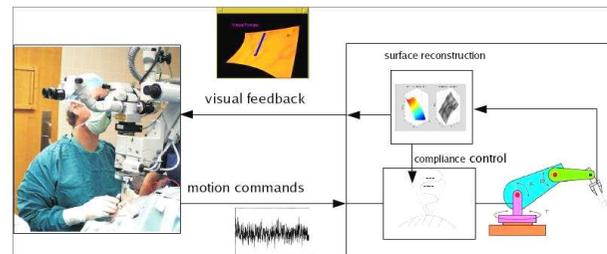
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## Virtual Fixtures (prevention of organ injuries)



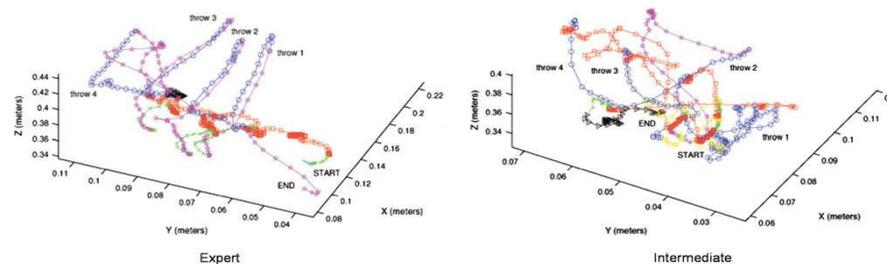
Example:  
eye surgery

## Task dependent Motion Constraints (compensation of human imperfections)



Hand tremor

## Skill Evaluation of the surgeons (certification of skills as estimate of success rate for a given procedure)



Source: Surgical Endoscopy, July 2010

All of these tasks (**Virtual Fixtures**, **Motion Constraints**, and **Skill Evaluation**) rely heavily on perception:

- **Virtual Fixtures:**  
Relies on **3D Reconstruction** and **Motion Capturing**.
- **Motion Constrains:**  
Relies on **3D Reconstruction**, **Motion Capturing** and **Procedure Modeling**.
- **Skill Evaluation:**  
Relies on **Motion Capturing**.
- There are three major building blocks to discuss:
  - Motion Capturing
  - Procedure Modeling
  - 3D Reconstruction

- **Motion capturing:** common systems are based on robot kinematics from robots like daVinci (e.g. the JHU system).  
**Problem:** What about standard procedures? They are still most common, thus should not be neglected.  
(→ *Motion Constraints, Skill Evaluation*)
- **Procedure Modeling:** movement constraints in a single step of the procedure are of interest. Operation-relevant properties and actions can be described by means of Object container, Location Map and Functionality Graph.  
(→ *Motion Constraints*)
- **3D Reconstruction:** since the anatomy details vary between patients, an accurate reconstruction to map the procedure into the given environment is required.  
**Problem:** many procedures still use monocular equipment, which generates low-quality image data. Photogrammetric enhancements to the reconstruction process are necessary. View planning algorithms can be used to improve accuracy.  
(→ *Virtual Fixtures, Motion Constraints*)



conventional

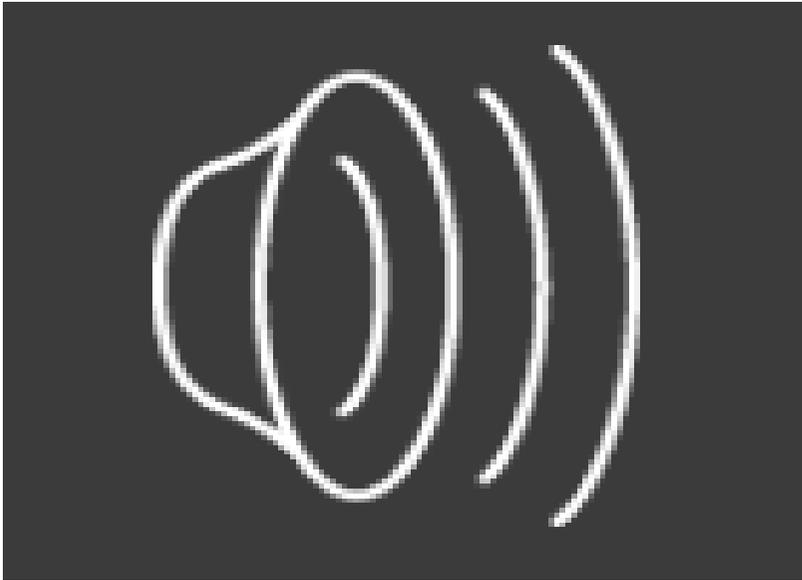


Source: MedGadget

Kinematics-based

We are interested in providing similar motion information about tools as from systems working with kinematics data

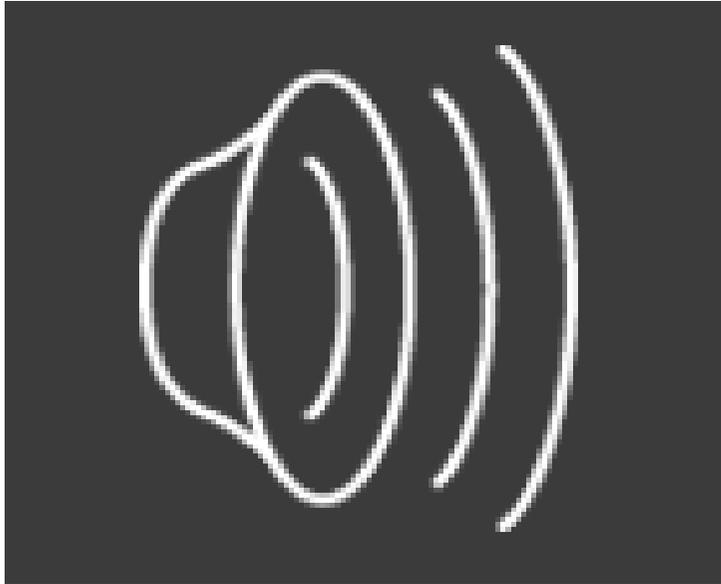
A new trend in computer-assisted surgery is usage of IMUs in videoendoscopes.



**Problem:** high dynamics of the human hand is too fast for the low frame-rates of standard video cameras



**Solution:** coupling of inertial sensors IMU with visual perception  
(Mair et al. IROS 2009)



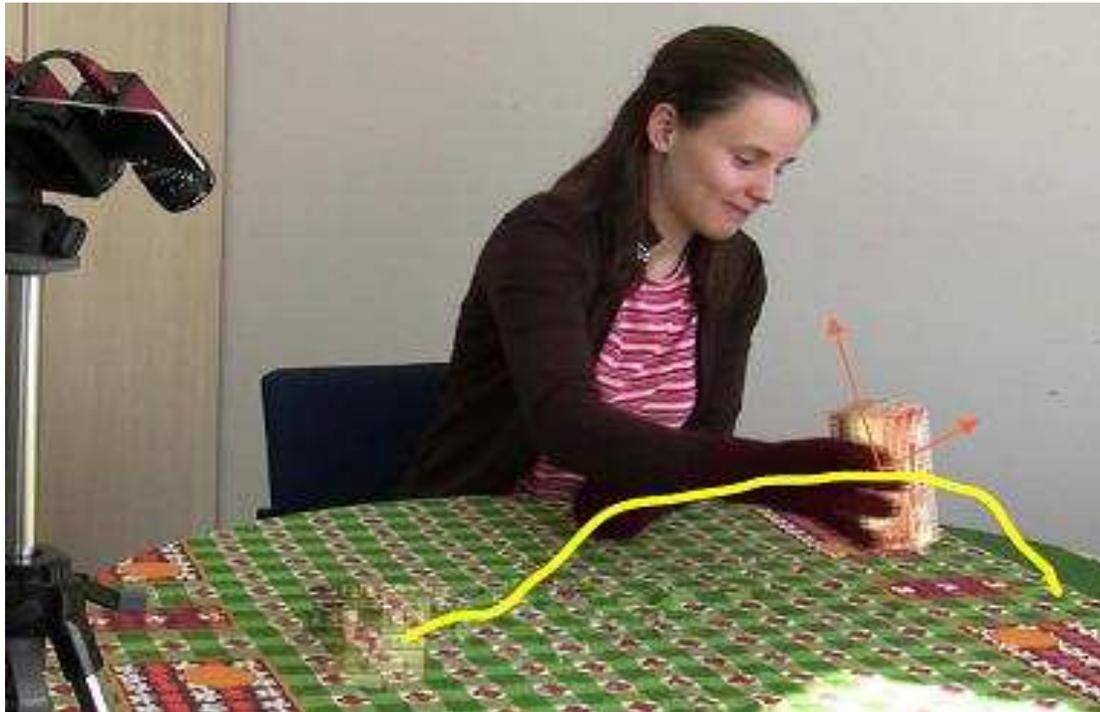
Template-based Tool Tracking  
in image space  
Application: guidance of the  
camera

(work with G.Hager, JHU)

6DoF pose estimation of tools  
manipulated in the scene  
The system estimates in real-time  
the pose of the moved object

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Problem: What about standard procedures? They are still most common, thus should not be neglected.  
(→ *Motion Constraints, Skill Evaluation*)
- **Procedure Modeling:** movement should be task-, object- and location-specific. Manipulation-relevant properties and actions can be described by means of Object container, Location Map and Functionality Graph.  
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# Learning of Action Attributes through Observation of Human Actions

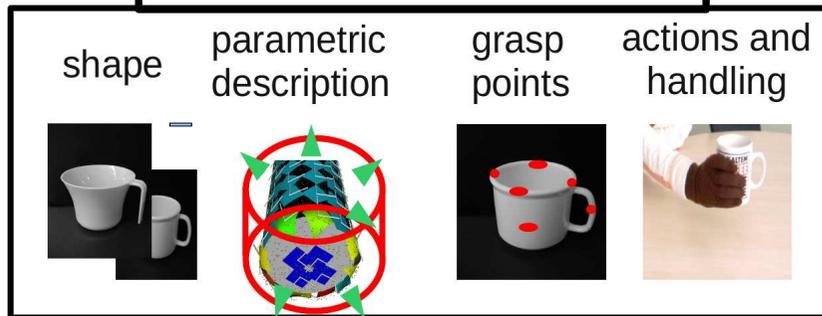


# Representation of Manipulation-Relevant Object Properties and Actions

- Observation of human demonstrations
- Extraction of knowledge:
  - New knowledge
  - Relevance
- Knowledge:
  - Object properties
  - Object functionalities in the environment



## Object container

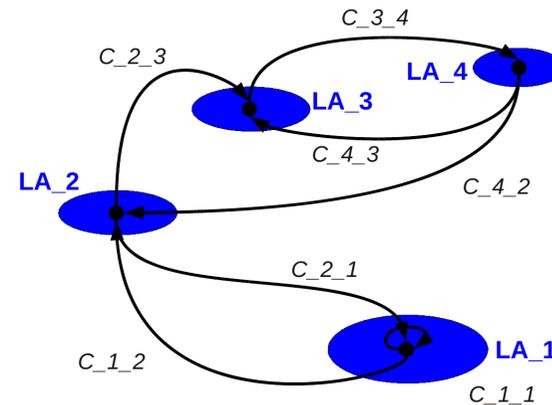


Each tool used in the procedure has its own container describing its shape, handling properties etc.

(Petsch/Burschka IROS2011)

## Functionality Map

Locations Area  $LA_i$

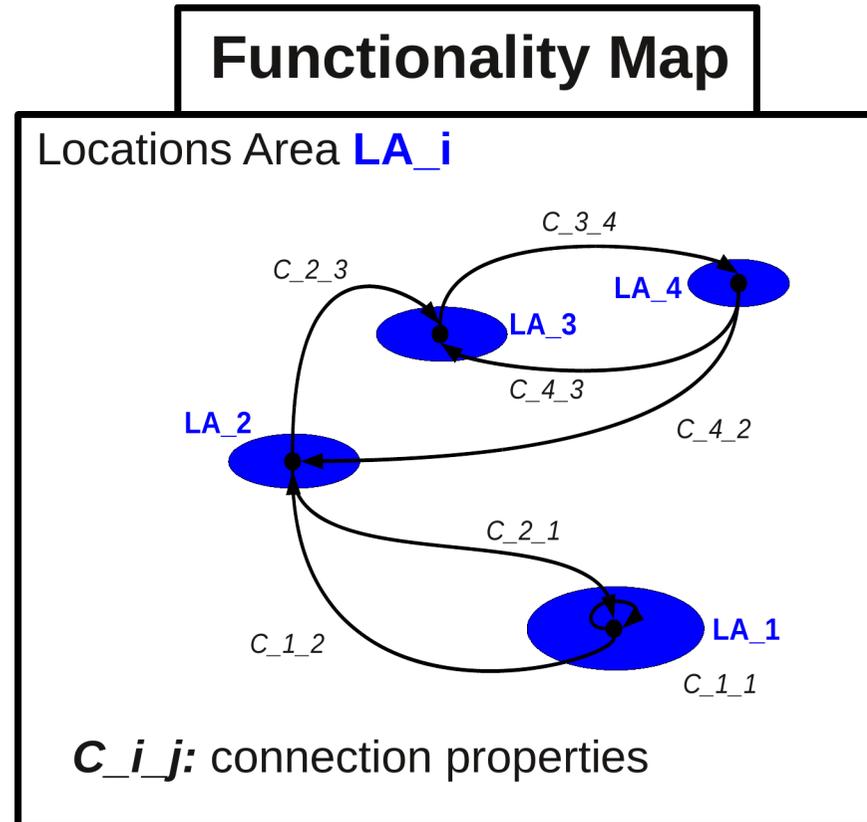


$C_{i_j}$ : connection properties

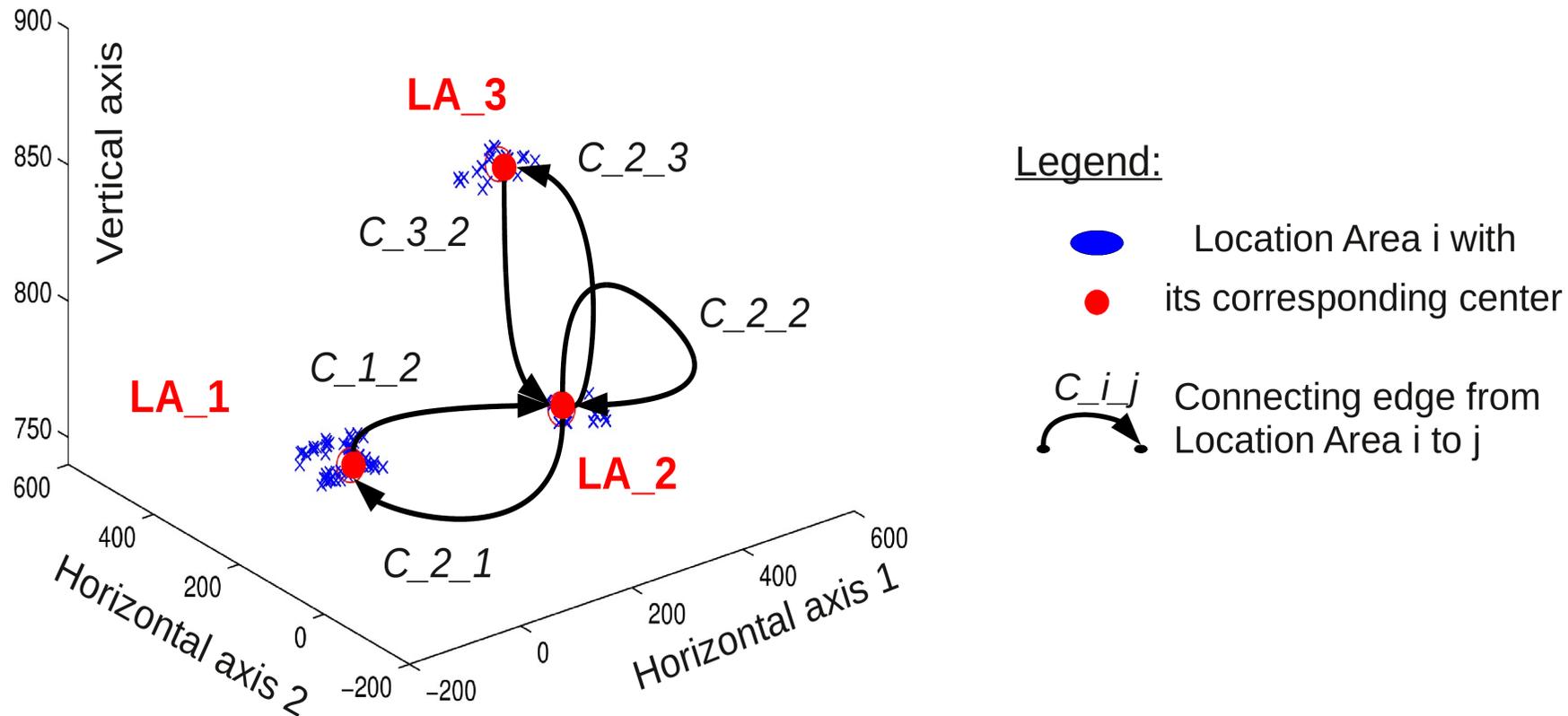
In addition, we store a functionality map for a specific procedure that describes the way how the tool was used during the procedure while moved between points in the body

# Functionality Map

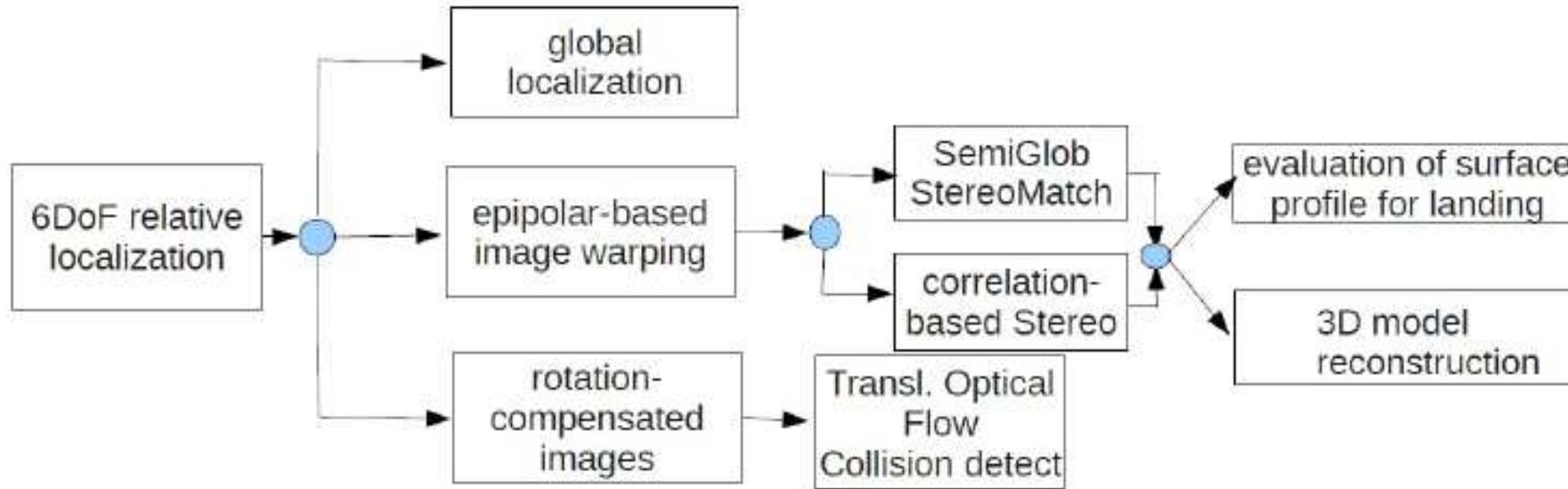
- Location Areas
- Connection Properties:
  - Arbitrary movement vs. constrained trajectory
  - Connection relevance
  - Velocity constraint during pick-up
  - Grasp taxonomy



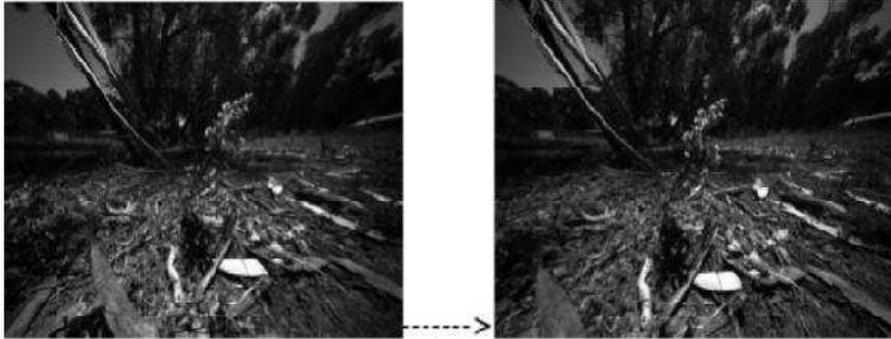
# Tracking Data: Location Areas



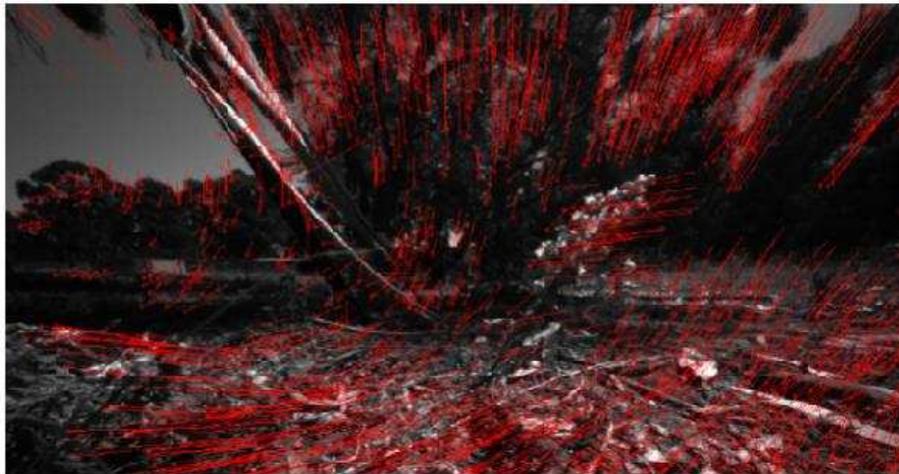
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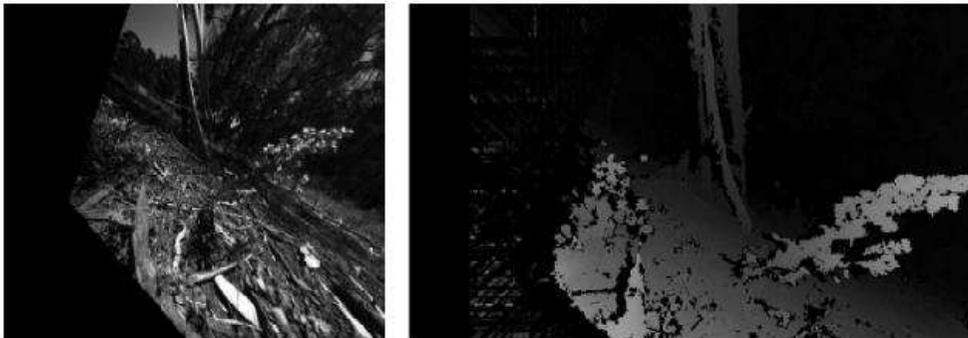
Modular reconstruction pipeline for an easy exchange of modules



Initial images

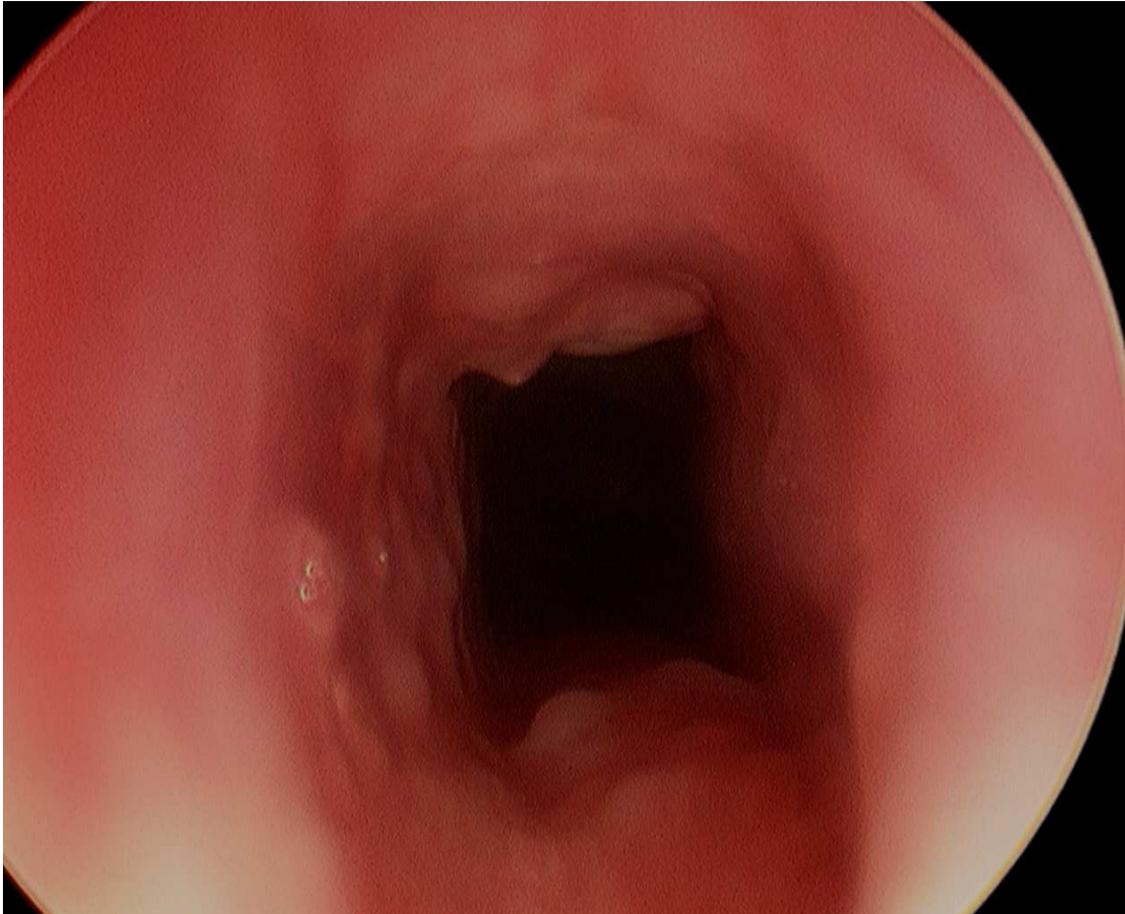


Motion estimation

Motion correction  
and reconstruction  
result

# Scene and Motion Recovery from Monocular Images

- A well-known problem in Computer Vision
- Algorithms for “simple” cases are abundant
- But: Videendoscopy is difficult!
- Problems:
  - Difficult lighting conditions
  - Bad image quality
  - Very little image structure

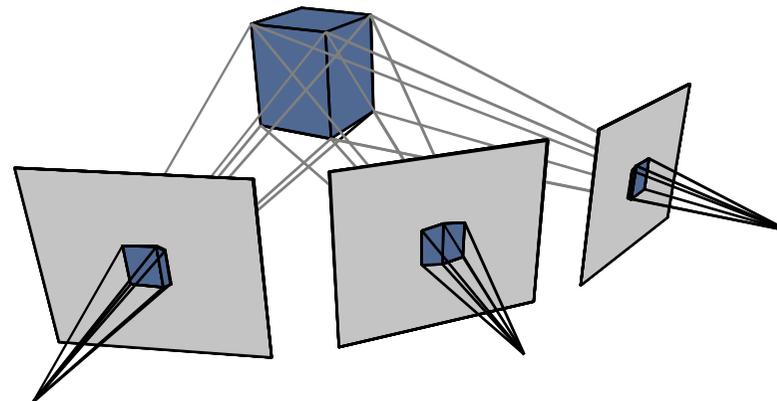


Typical difficulties: image blur, image noise, almost no texture

# The Basic Idea

- Bundle adjustment is a well-known algorithm for structure and camera position recovery
- Feature-based and intensity-based variants
- Toy example for feature-based reconstruction:
- Nonlinear optimization minimizes reprojection error of pixel coordinates.

$$\min_{a,b} \sum_{i=1}^n \sum_{j=1}^m d(Q(a_j, b_i), x_{i,j})$$



# Intensity-based methods

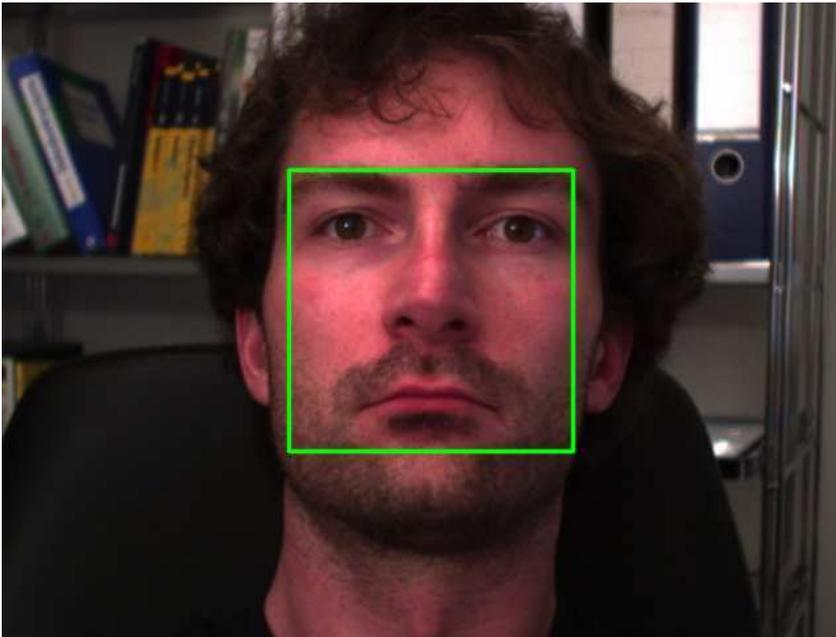
- In videoendoscopy, reliable feature detection is nearly impossible
- We use an intensity based approach
- Price to pay:
  - Computationally expensive
  - Less detailed models (regularization)
- Advantage:
  - Reconstruction is possible where feature-based methods fail!

# Reconstruction with constant lighting

- First idea: Apply a multi-view intensity-based reconstruction method
- Important advantage: Intensity-based methods use as much information as there is
- This approach uses the brightness constancy assumption
- Idea: Find parameters that explain images by nonlinear optimization

$$\min_{a,b} \sum_{i=1}^n \sum_{j=1}^m d(I_j(\mathbf{Q}(\mathbf{a}_j, \mathbf{b}_i)), I_0(\mathbf{p}_i))$$

- Works well for Lambertian surfaces

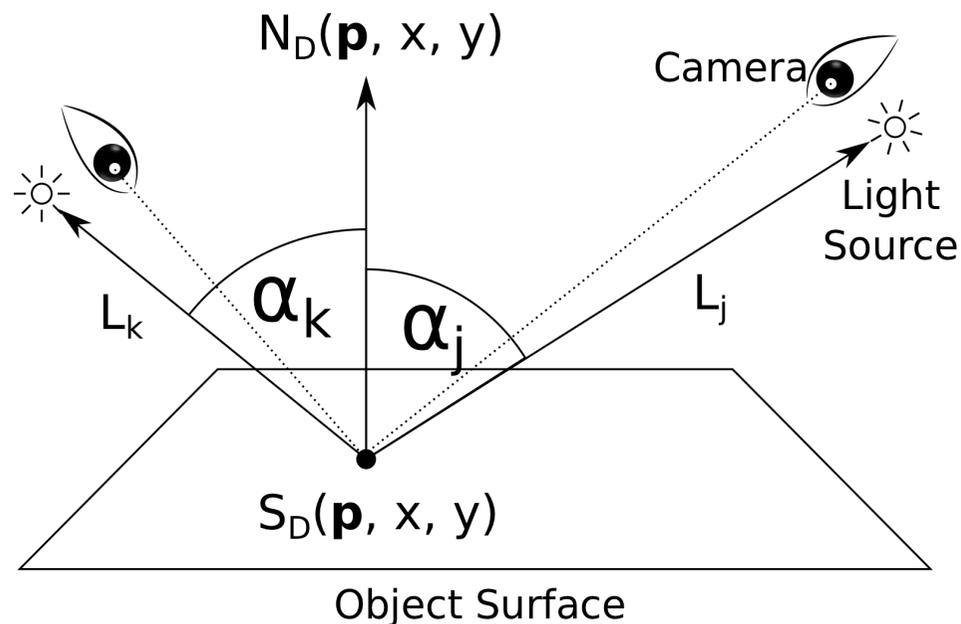


Ruepp and Burschka. Fast recovery of weakly textured surfaces from monocular image sequences. ACCV 2010

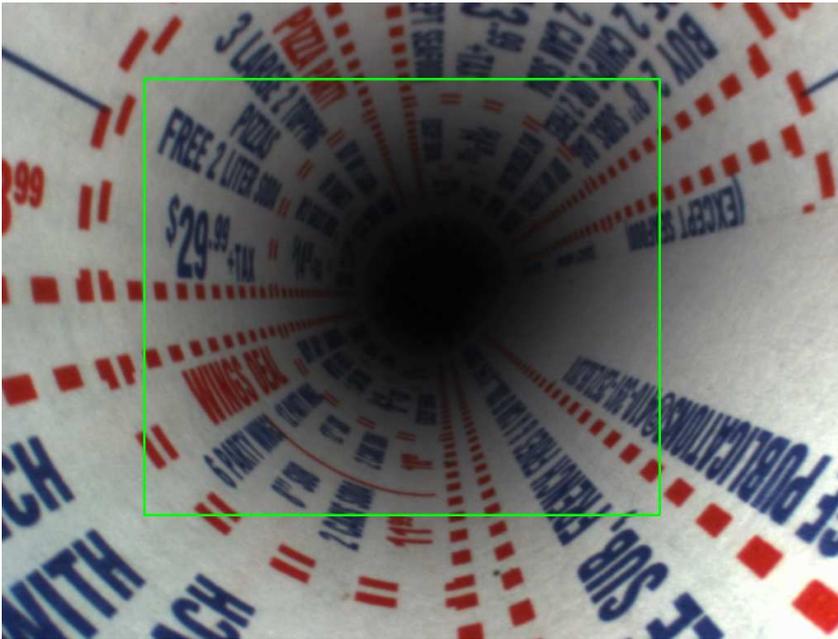
<http://mvp.visual-navigation.com/>

# Under changing lighting conditions

- In videoendoscopy, the brightness constancy assumption is violated!
- The light source is moving with the camera
- This can be modeled by the reconstruction algorithm



# Reconstruction results



# Current and Future Research

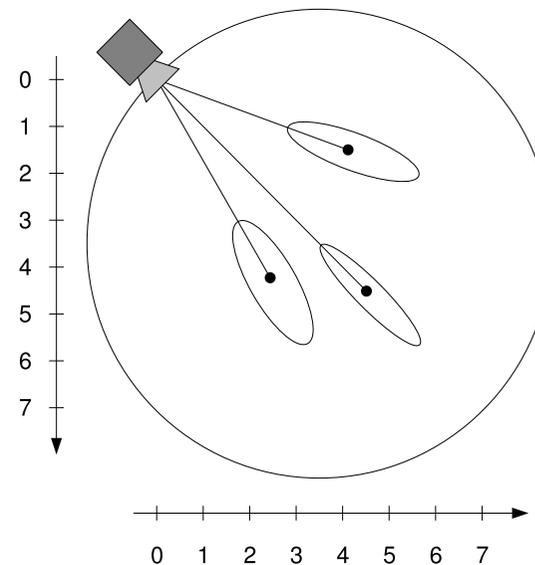
- The algorithm makes idealizing assumptions that lead to suboptimal results:
  - Linear camera response, no vignetting
  - Singular point light source
  - Uniform light source intensity
  - Light source coincident with camera position
- A calibration method is needed!
- Also: Performance can be improved
- Deal with partial occlusions

# View planning

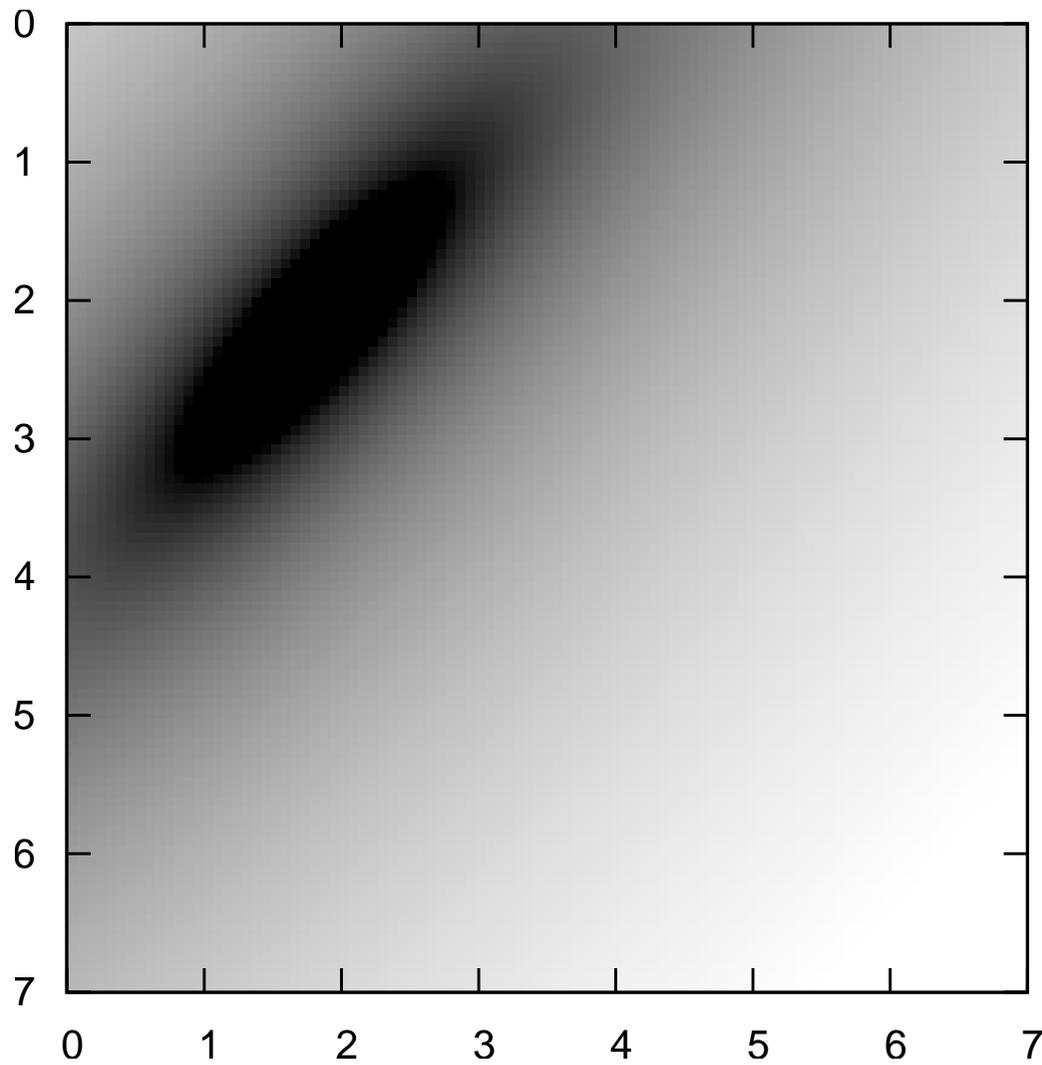
- Reconstruction accuracy is crucial for safety.
- How do we evaluate the accuracy?
- How can we improve it?
- View planning algorithms help to determine camera movement so as to maximize accuracy

# Schematic Overview

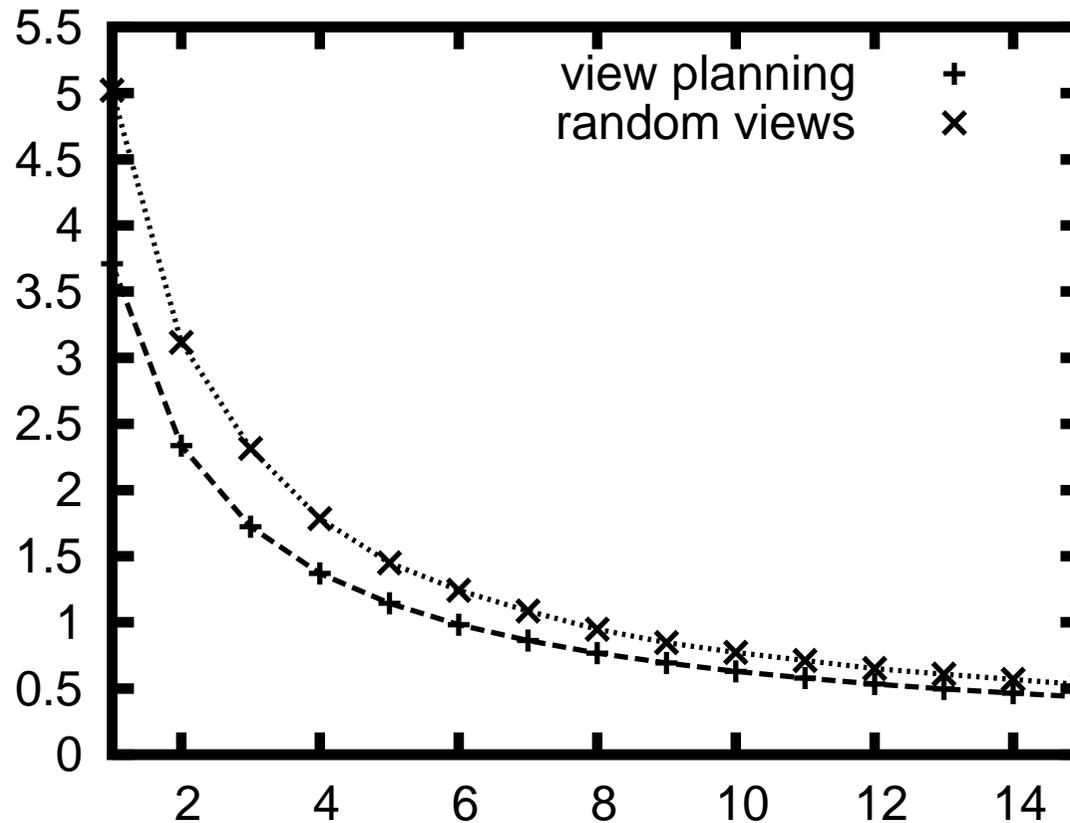
- Reconstruction results are typically affected by uncertainty
- Viewing direction should be orthogonal to maximum uncertainty
- This leads to an evaluation function for camera positions



# Camera Position Evaluation



<http://mvp.visual-navigation.com/>



Especially in the first few steps, uncertainty is reduced significantly

# Future Research

- Until now: Camera movement restricted to a circular path around target
- Needs to be adapted to videoendoscopy, where movement is more restricted
- Deal with partial occlusions

**Thank you for  
your attention!**

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