

## Object Recognition and Real-Time Tracking in Microscopy Imaging

Jan Wedekind

30th Aug – 1st Sep 2006

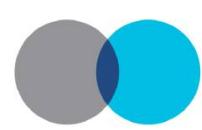


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MiCRoN	<a href="http://wwwipr.ira.uka.de/~micron/">http://wwwipr.ira.uka.de/~micron/</a>
Mimas	<a href="http://sourceforge.net/projects/mimas/">http://sourceforge.net/projects/mimas/</a>
MediaWiki	<a href="http://vision.eng.shu.ac.uk/mediawiki/">http://vision.eng.shu.ac.uk/mediawiki/</a>

People: J. Wedekind, M. Boissenin, B.P. Amavasai, F. Caparrelli, J. Travis

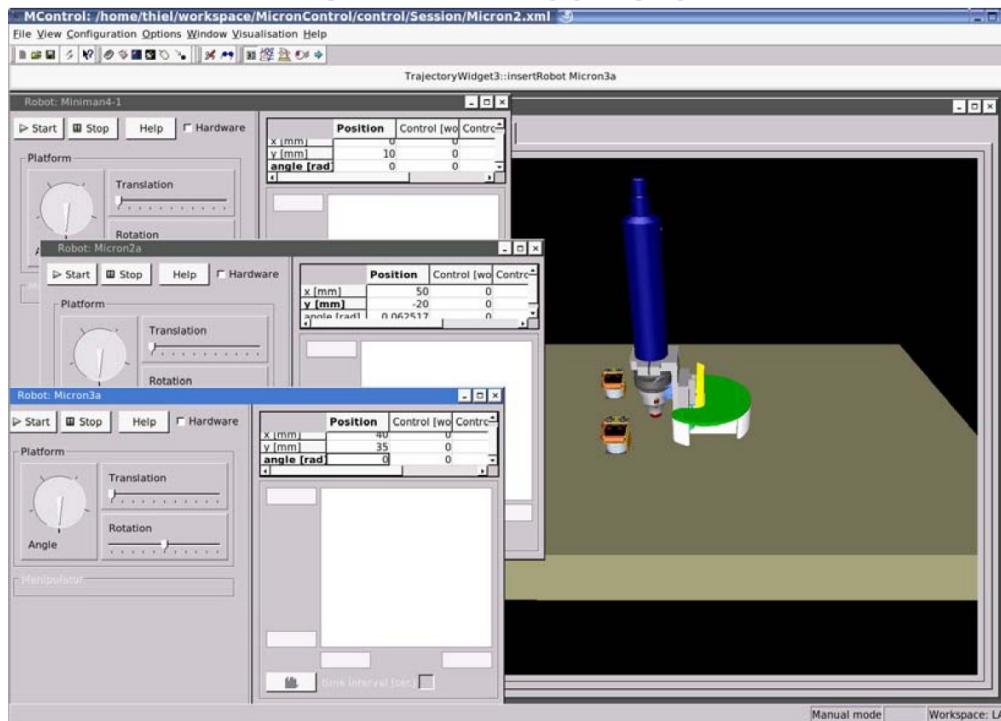


*Sheffield  
Hallam University*



Uppsala, Lausanne, St. Ingbert, Athens, Pisa, Barcelona, Karlsruhe

## Control & GUI



Universität Karlsruhe (Germany)

<http://wwwipr.ira.uka.de/~micron/>  
<http://www.cordis.lu/ist/>

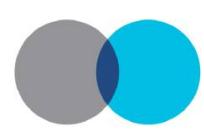
## Motivation

- prototype soldering/assembly
- cell manipulation
- manipulations inside vacuum chamber

## Project Goals

- Manipulate  $\mu\text{m}$ -sized objects
- Closed-loop control of robot
- 3D object recognition and tracking



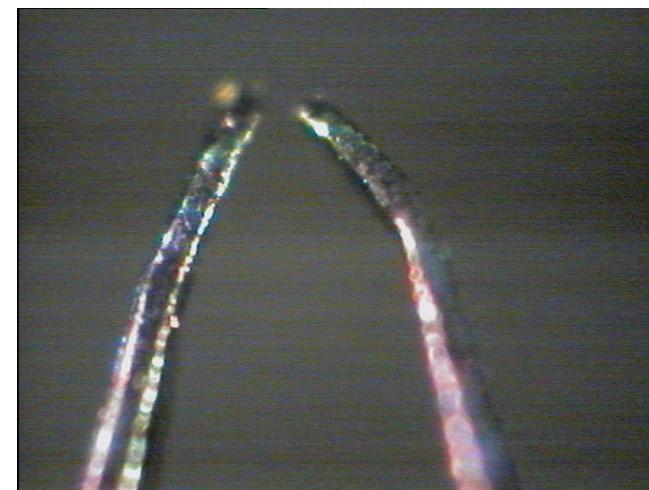


**Locomotion Platform**



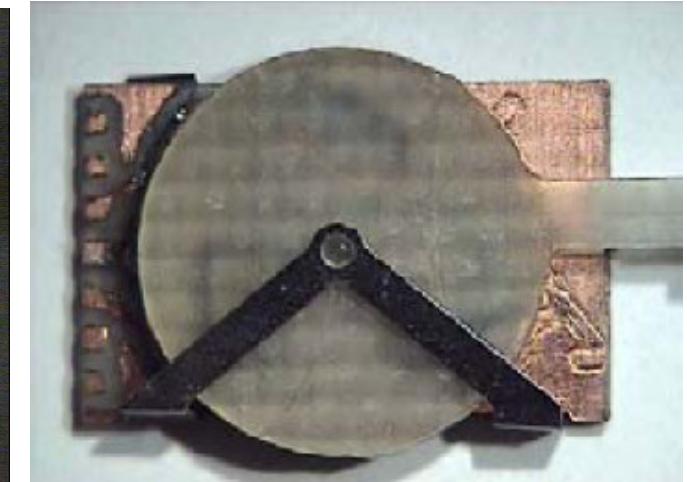
Ecole Polytechnique  
Fédérale de Lausanne  
(Switzerland)

**Gripper**



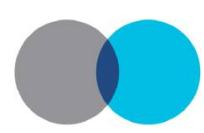
Scuola Superiore,  
Sant'Anna (Italy)

**Rotor**



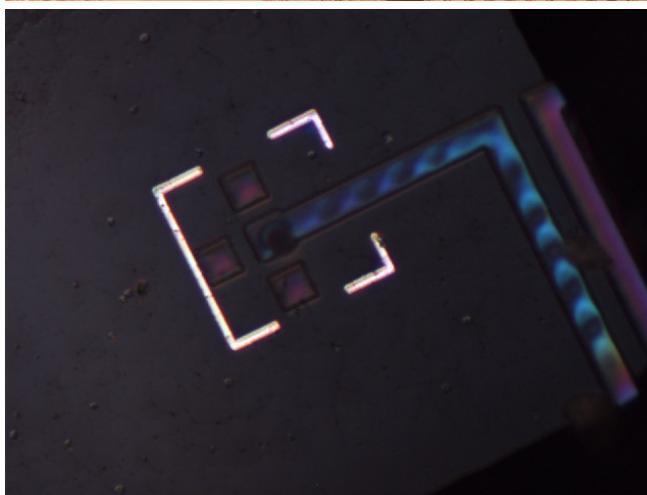
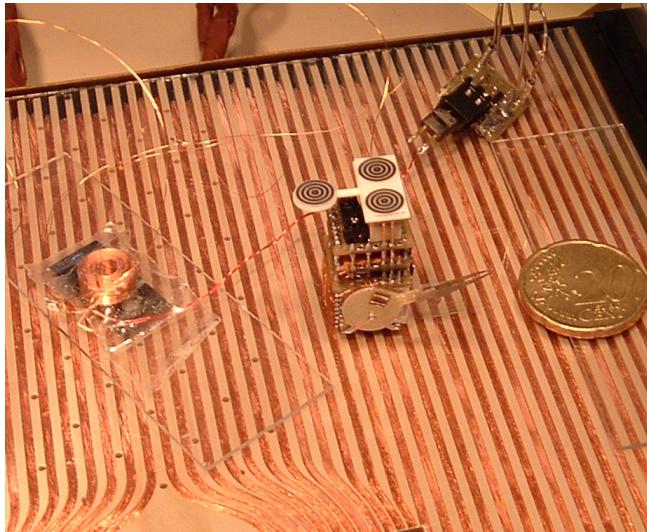
Uppsala University  
(Sweden)





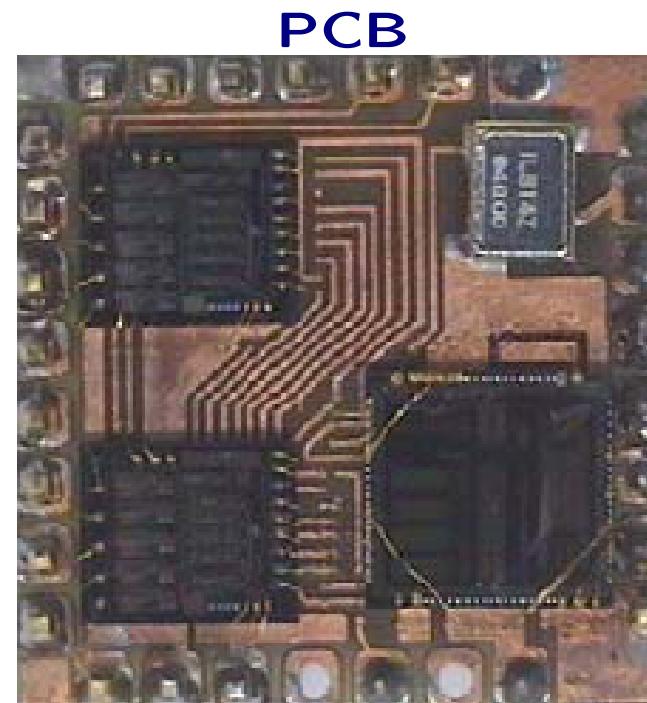
MATERIALS AND ENGINEERING  
RESEARCH INSTITUTE

## Power Floor, Syringe



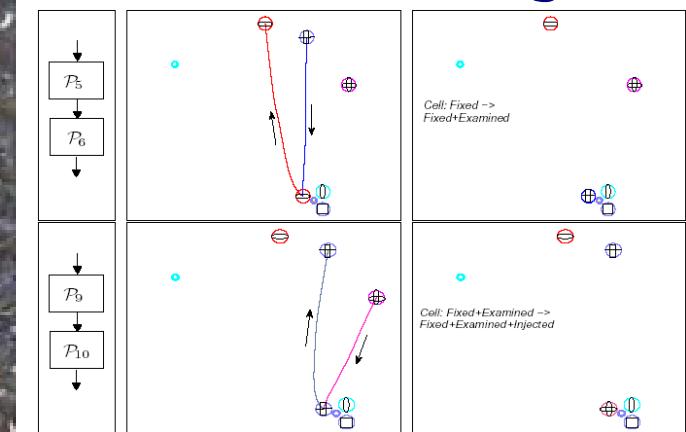
Fraunhofer Institute, St.  
Ingbert (Germany)

## Motivation MiCROn robot (ii)

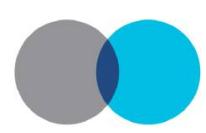


University of Barcelona  
(Spain)

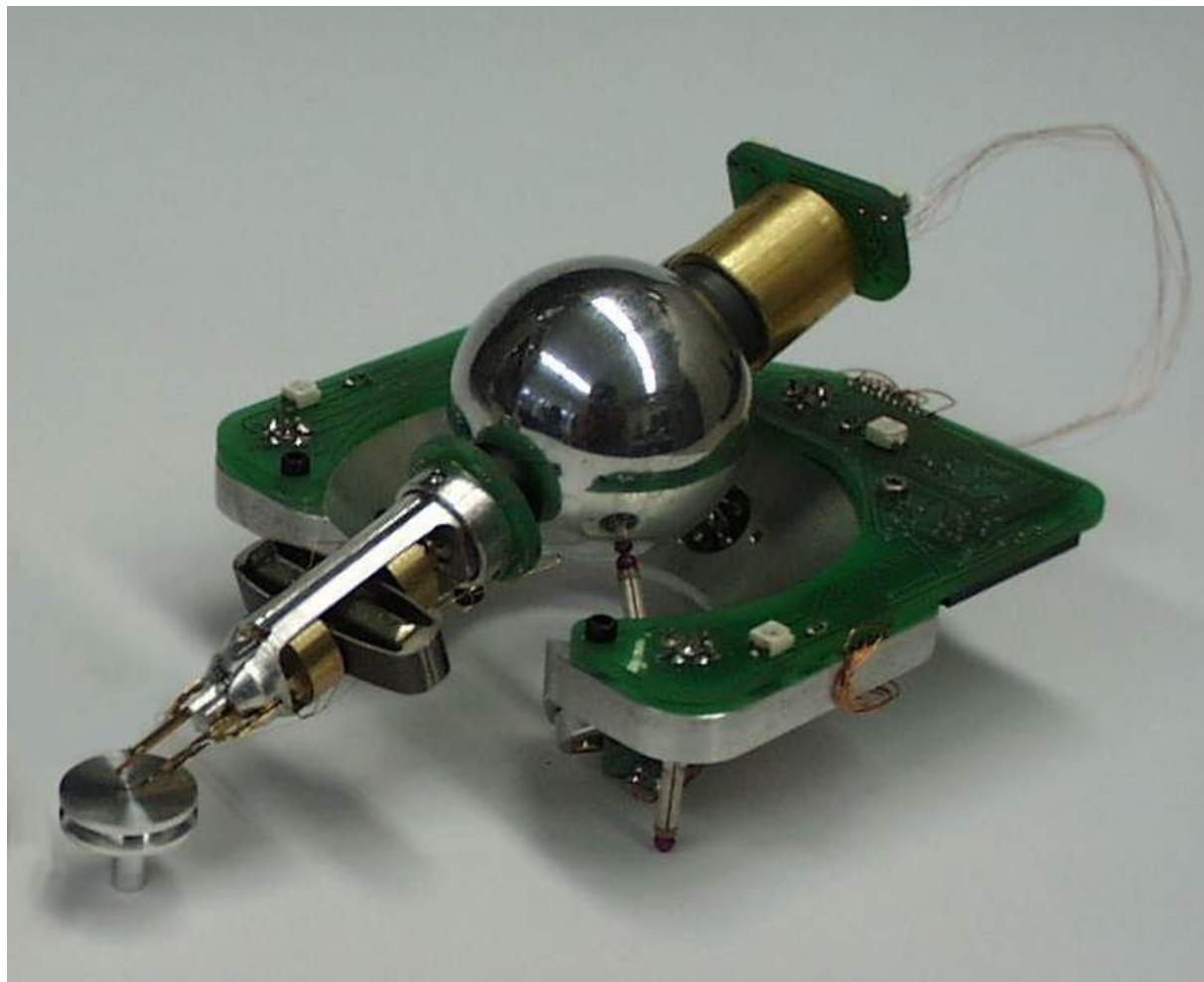
## PCB



National Technical  
University of Athens  
(Greece)



**MINIMAN III-2 (5 d.o.f.)**



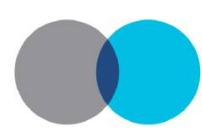
centre of sphere - end effector = 6.2 cm

**MiCROn (4 d.o.f.)**



fits on a 20 cent coin





## Geometric Hashing : A General and Efficient Model-Based Recognition Scheme

*Yechezkel Lamdan and Haim J.Wolfson*

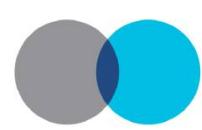
Robotics Research Laboratory  
Courant Inst. of Math., NYU  
715 Broadway, 12'th floor,  
New York, N.Y. 10003.

**Abstract:** A general method for model-based object recognition in occluded scenes is presented. It is based on *geometric hashing*. The method stands out for its efficiency. We describe the general framework of the method and illustrate its applications for various recognition problems both in 3-D and 2-D. Special attention is given to the recognition of 3-D objects in occluded scenes from 2-D gray scale images. New experimental results are included for this important case.

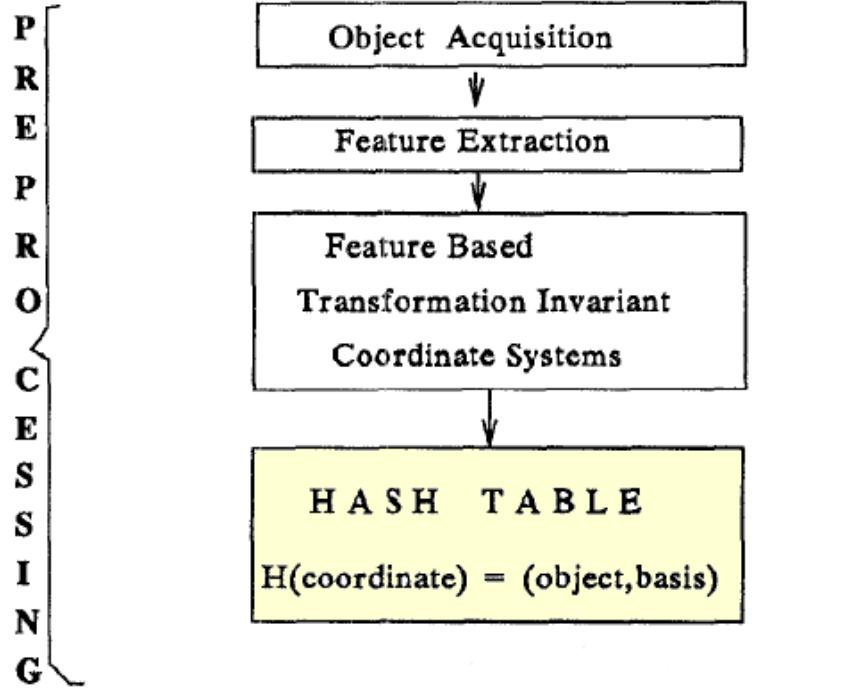
### 1. Introduction.

We present a unified approach to the *representation* and *matching* problems which applies to object recognition under various geometric transformations both in 2-D and 3-D. The objects are represented as sets of geometric features, such as points or lines, and their geometric relations are encoded using minimal sets of such features under the allowed object transformations. This is achieved by standard methods of *Analytic Geometry* invoking *coordinate frames* based on a minimal number of features, and representing other features by their coordinates in the appropriate frame. Our



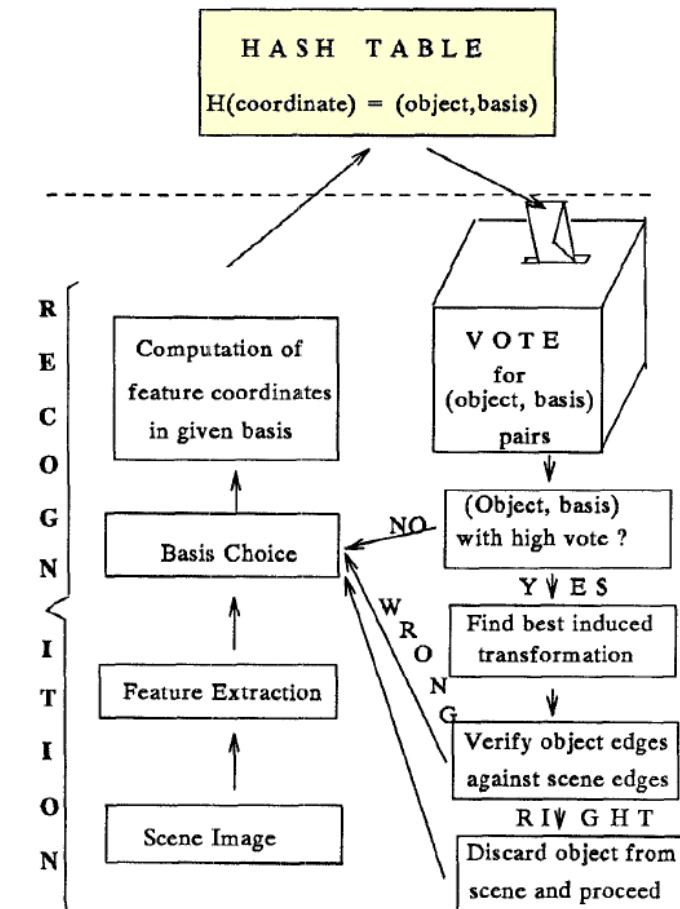


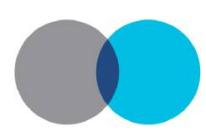
## Preprocessing



1988, Lamdan & Wolfson

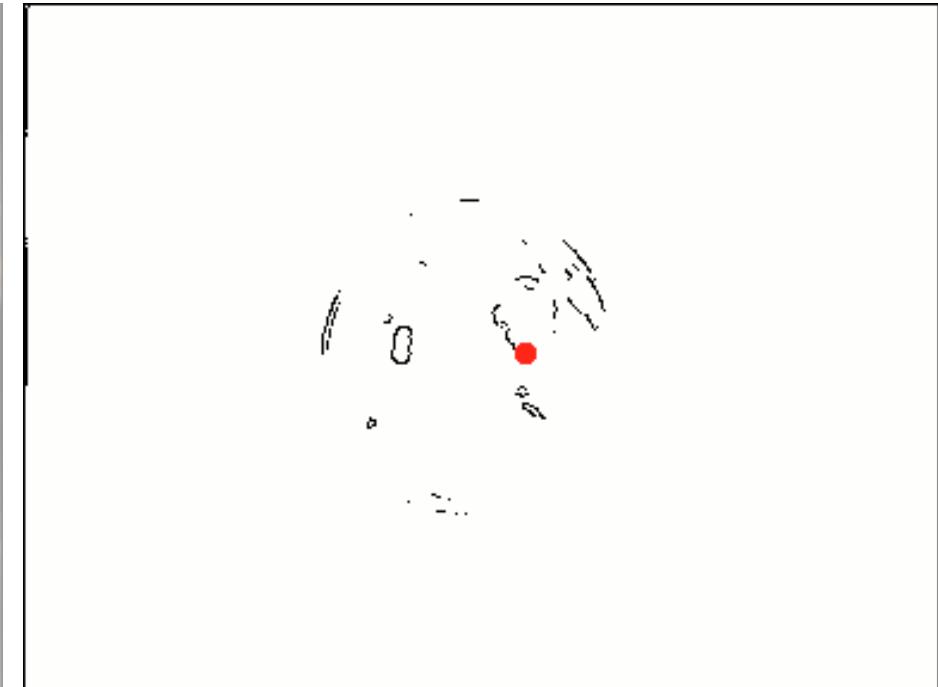
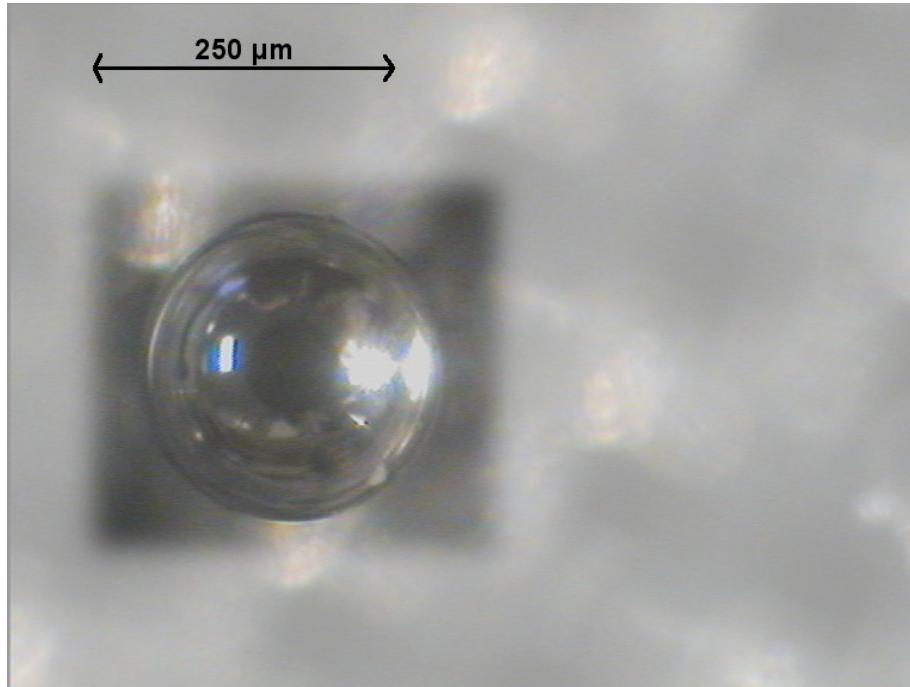
## Recognition

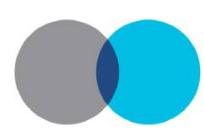




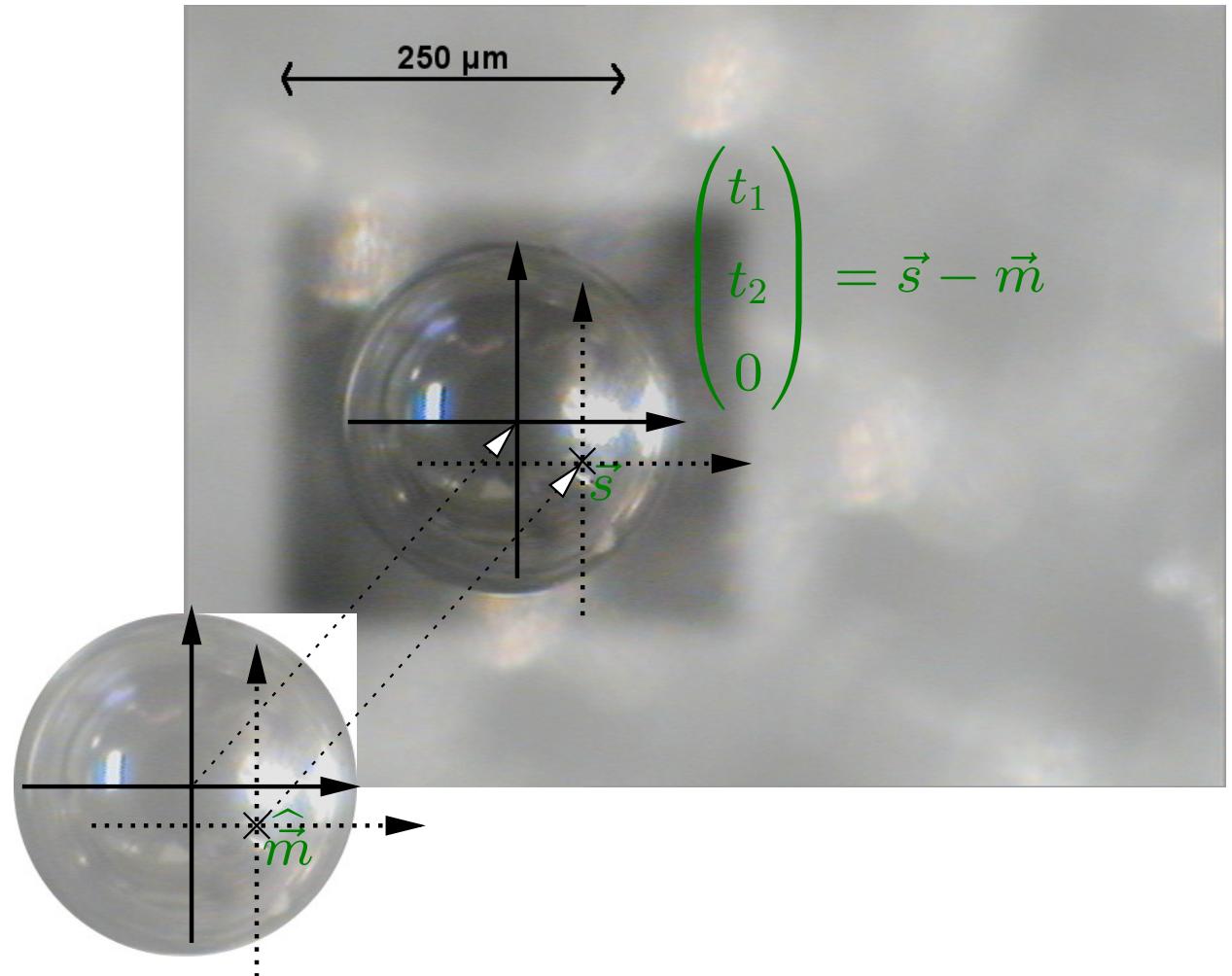
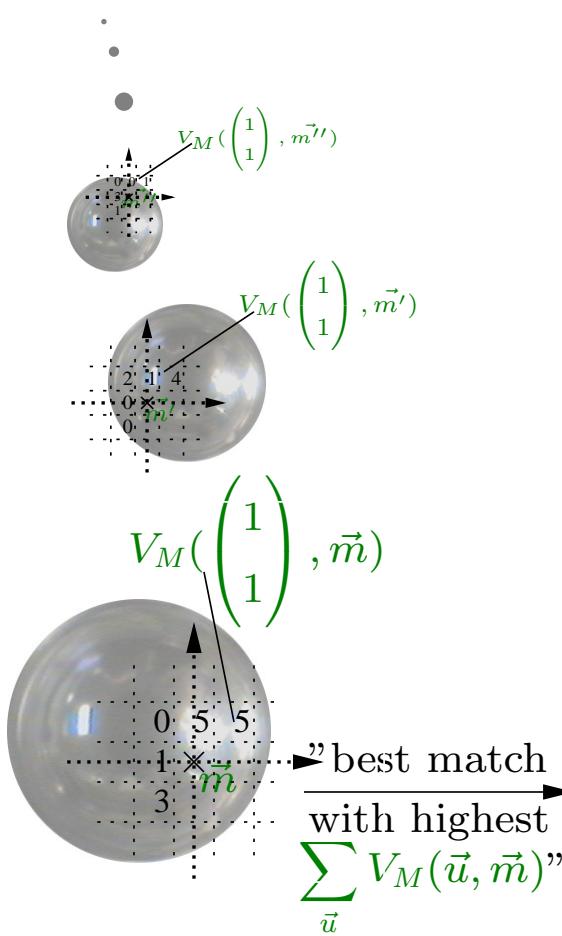
## Solder Sphere

- Extract Sobel edges
- Randomly select a feature location



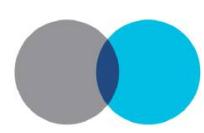


## Solder Sphere



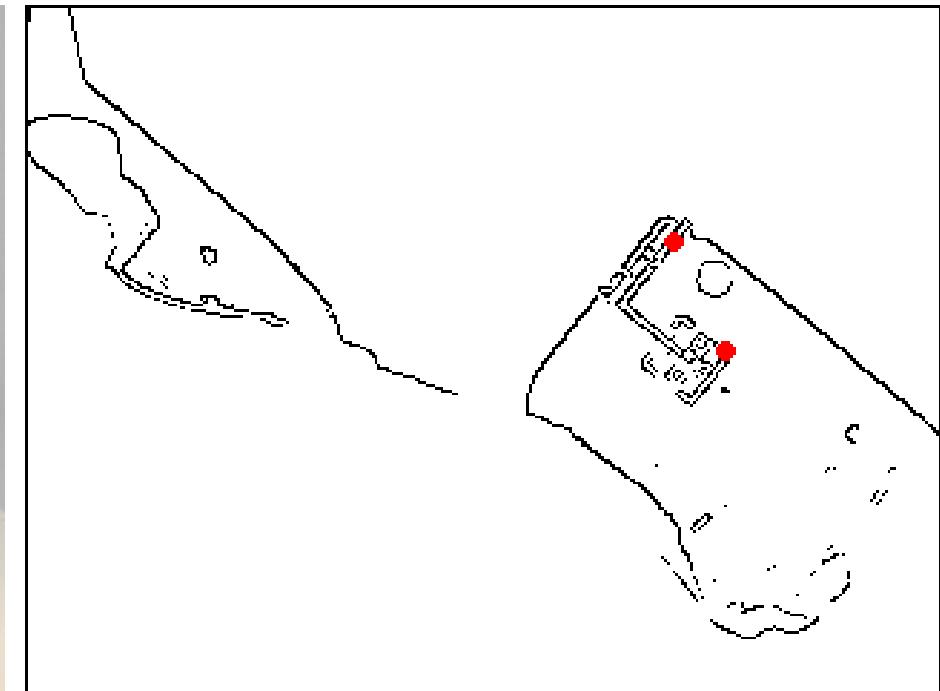
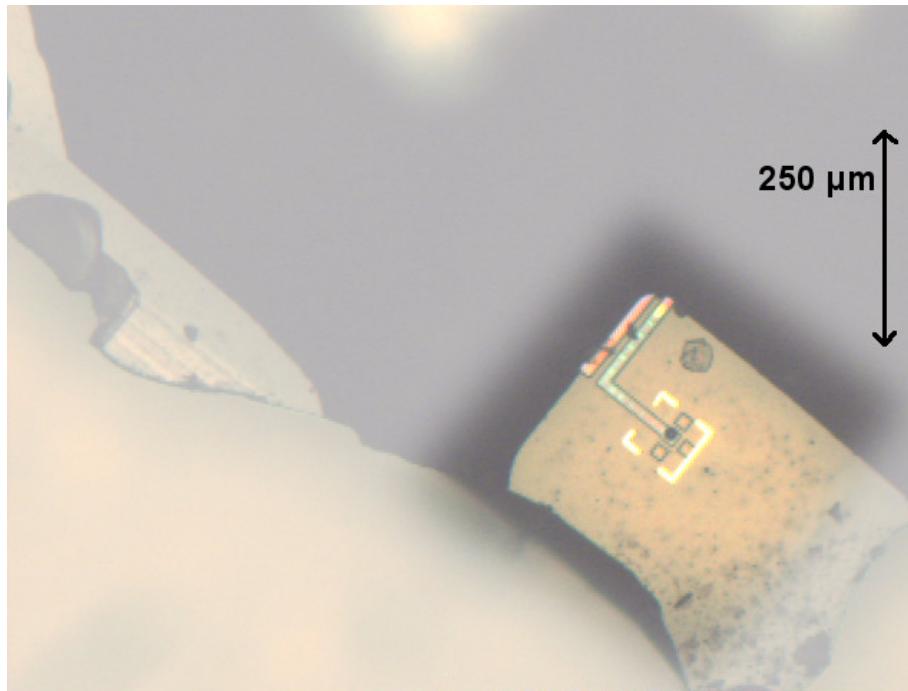
a single feature-correspondence reveals the object's pose

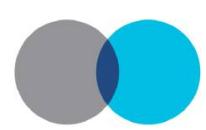




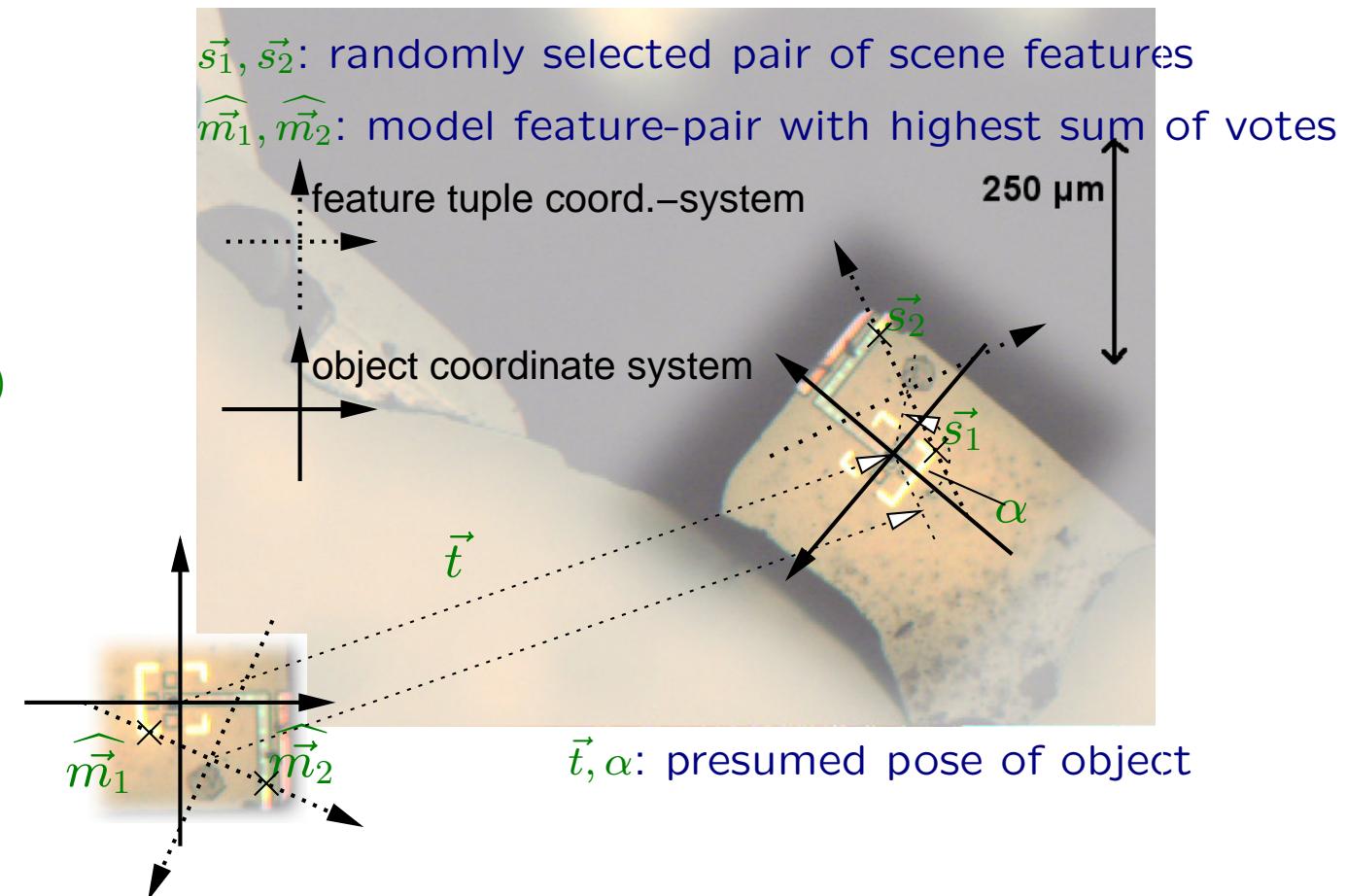
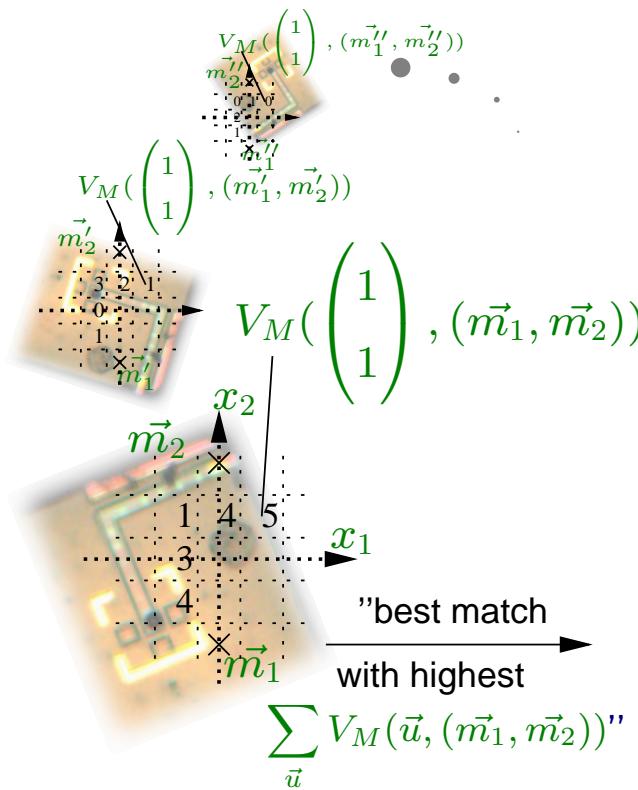
## Syringe Chip

- Extract Sobel edges
- Randomly select two feature locations





## Syringe Chip



two feature-correspondences are revealing the object's pose



## Efficient Tracking with the Bounded Hough Transform

Michael Greenspan<sup>1,2,4</sup> Limin Shang<sup>1</sup> Piotr Jasiobedzki<sup>3</sup>

<sup>1</sup>Dept. of Electrical & Computer Engineering, <sup>2</sup>School of Computing, Queen's University, Canada

<sup>3</sup>MDRobotics, 9445 Airport Rd., Brampton, Ontario, Canada

<sup>4</sup>corresponding author: michael.greenspan@ece.queensu.ca

### Abstract

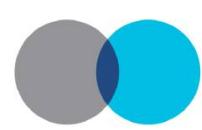
*The Bounded Hough Transform is introduced to track objects in a sequence of sparse range images. The method is based upon a variation of the General Hough Transform that exploits the coherence across image frames that results from the relationship between known bounds on the object's velocity and the sensor frame rate. It is extremely efficient, running in  $\mathcal{O}(N)$  for  $N$  range data points, and effectively trades off localization precision for runtime efficiency.*

*The method has been implemented and tested on a variety of objects, including freeform surfaces, using both simulated and real data from Lidar and stereovision sensors.*

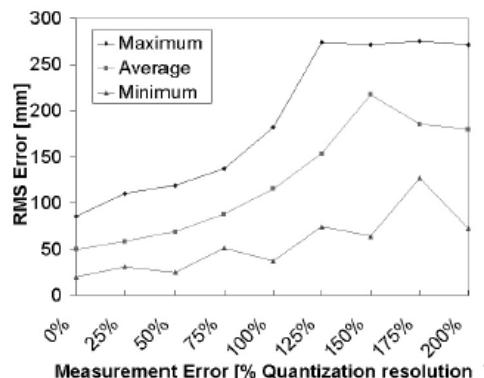
ing variants of the Iterative Closest Point Algorithm (ICP) [1]. This is primarily because range data is more expensive to collect, and so the images tend to be sparse, which makes it difficult to extract meaningful features. Examples of ICP-based tracking are [2, 3] and recently [4], which simultaneously reconstructs while tracking.

The Hough Transform is a well known and effective method of feature extraction and pose determination that has been explored thoroughly in the literature [5]. Many variations of the Hough Transform have been proposed [6], some of which are specifically tailored to tracking. The Velocity Hough Transform (VHT) [7] included a specific ve-

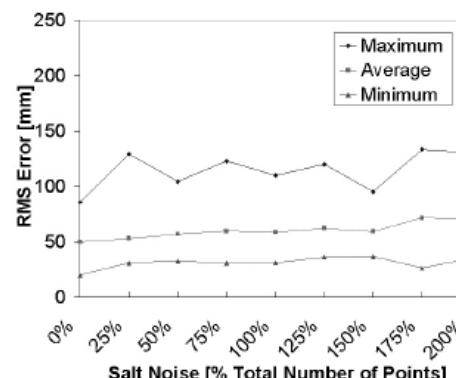




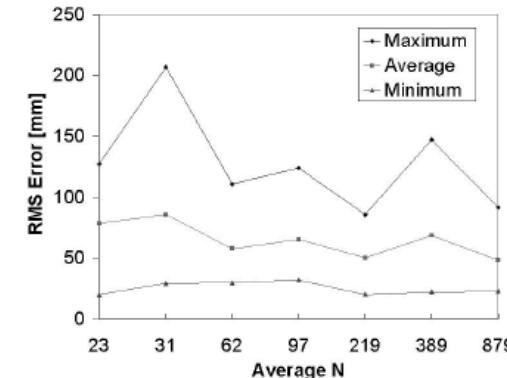
# Bounded Hough Transform 6 d.o.f. satellite tracking 2001/2004, Greenspan, S. & J.



a) measurement error



b) outliers



c) sparse data

Figure 6: Effects of Data Degradation



Figure 7: Robot Mounted Satellite Model  
of a Radarsat satellite was mounted on a 6 dof articulated

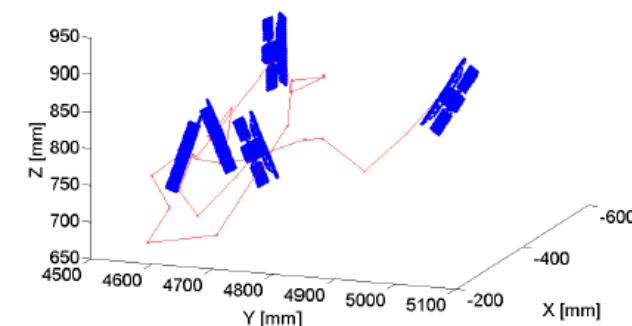
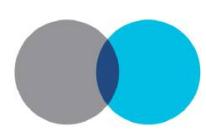


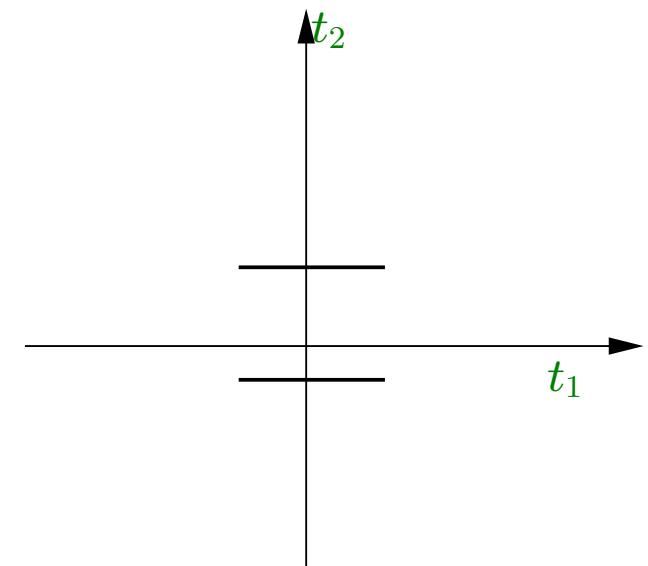
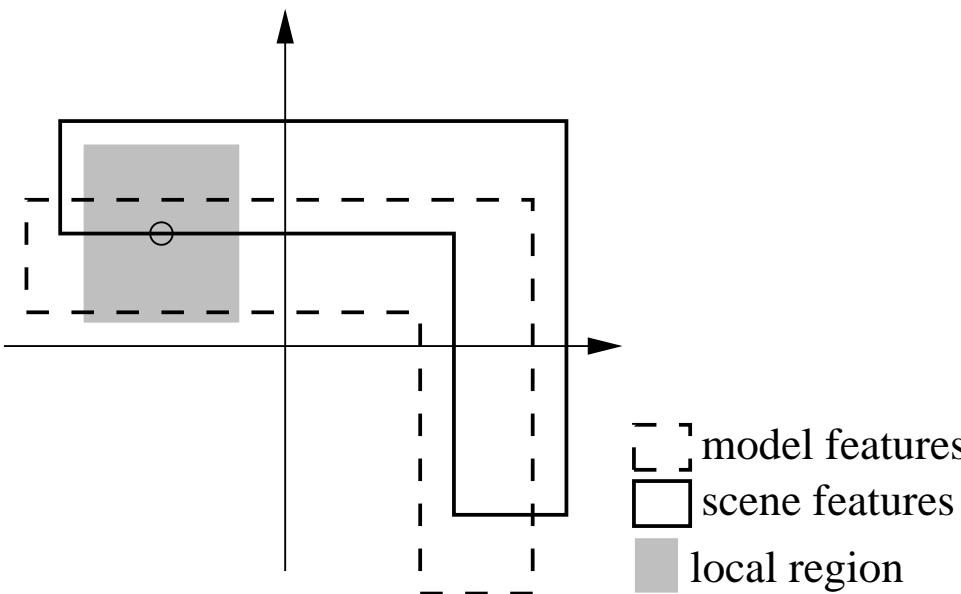
Figure 8: Satellite Trajectory, Lidar Data

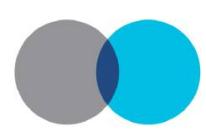
majority of which fell on the surface of the robot and the background and were therefore outliers. To demonstrate the effectiveness of the method at tracking in sparse as well as



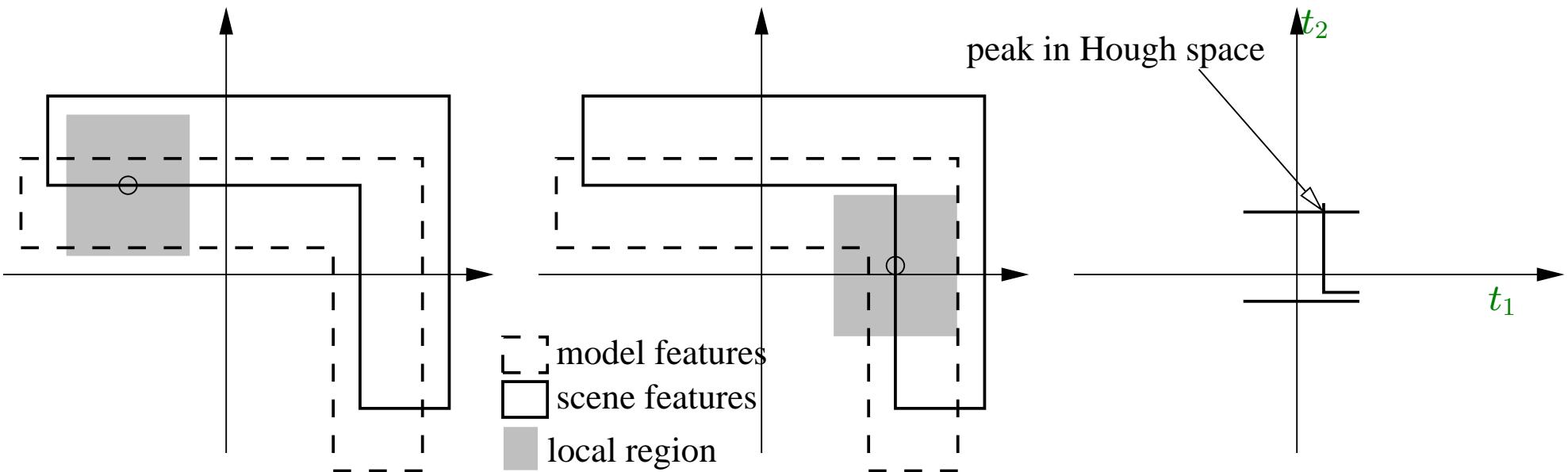


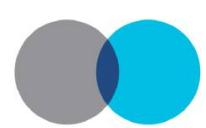
### Accumulate Votes



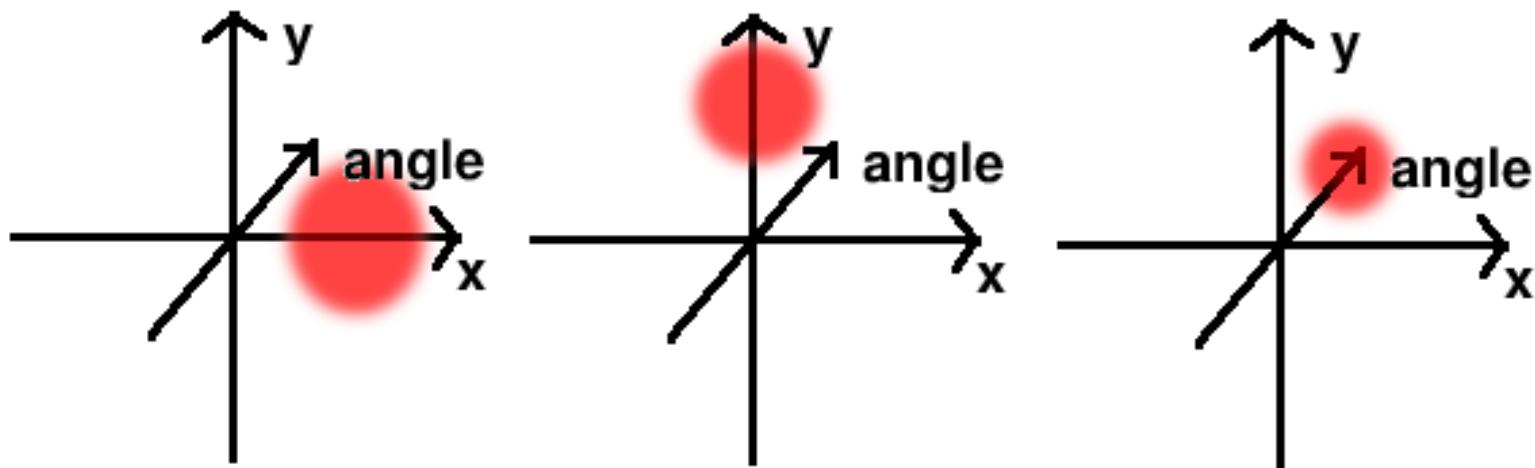
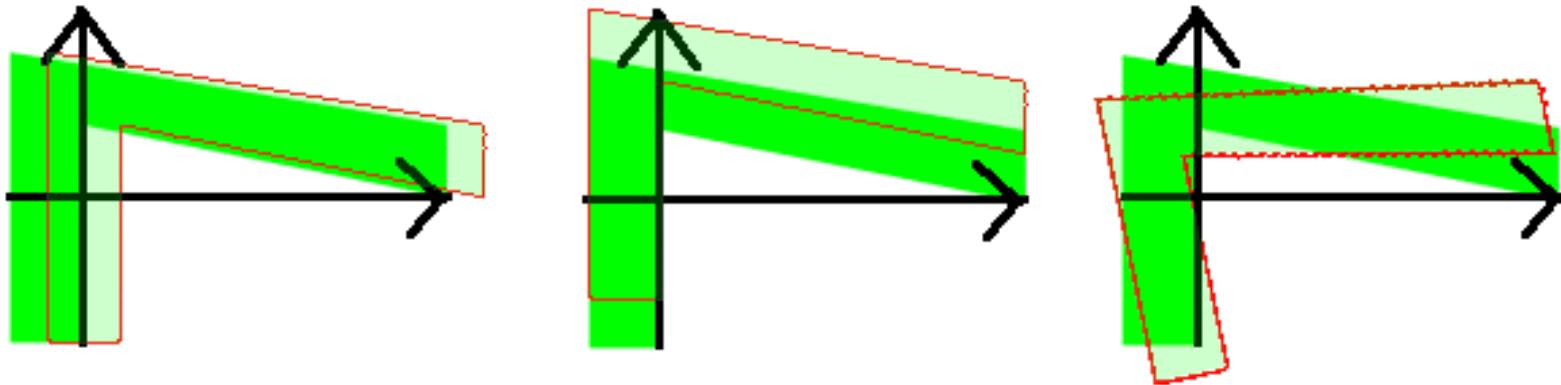


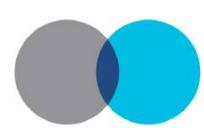
## Accumulate Votes





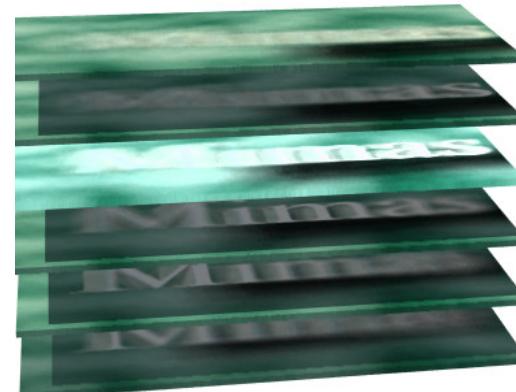
## Bounded Hough Transform 3 Degrees-Of-Freedom



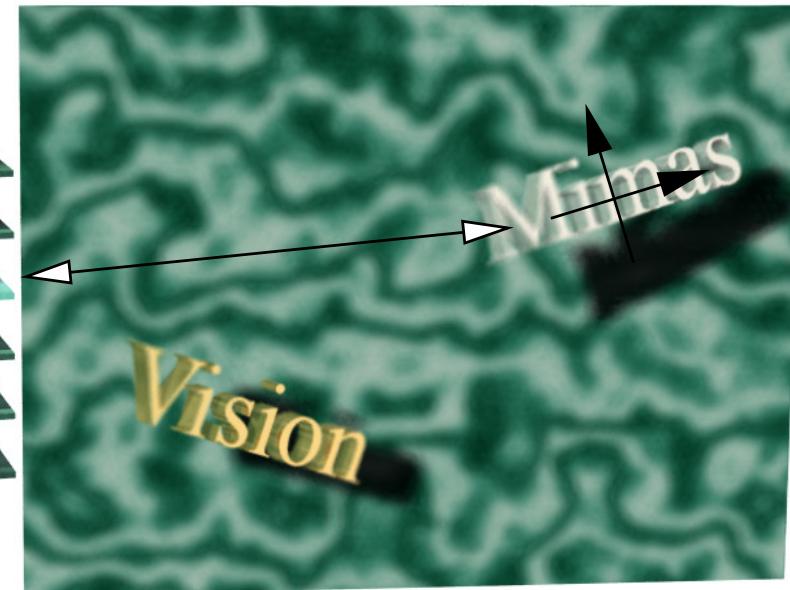


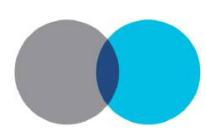
## 4 Degrees-Of-Freedom

Artificial Scene

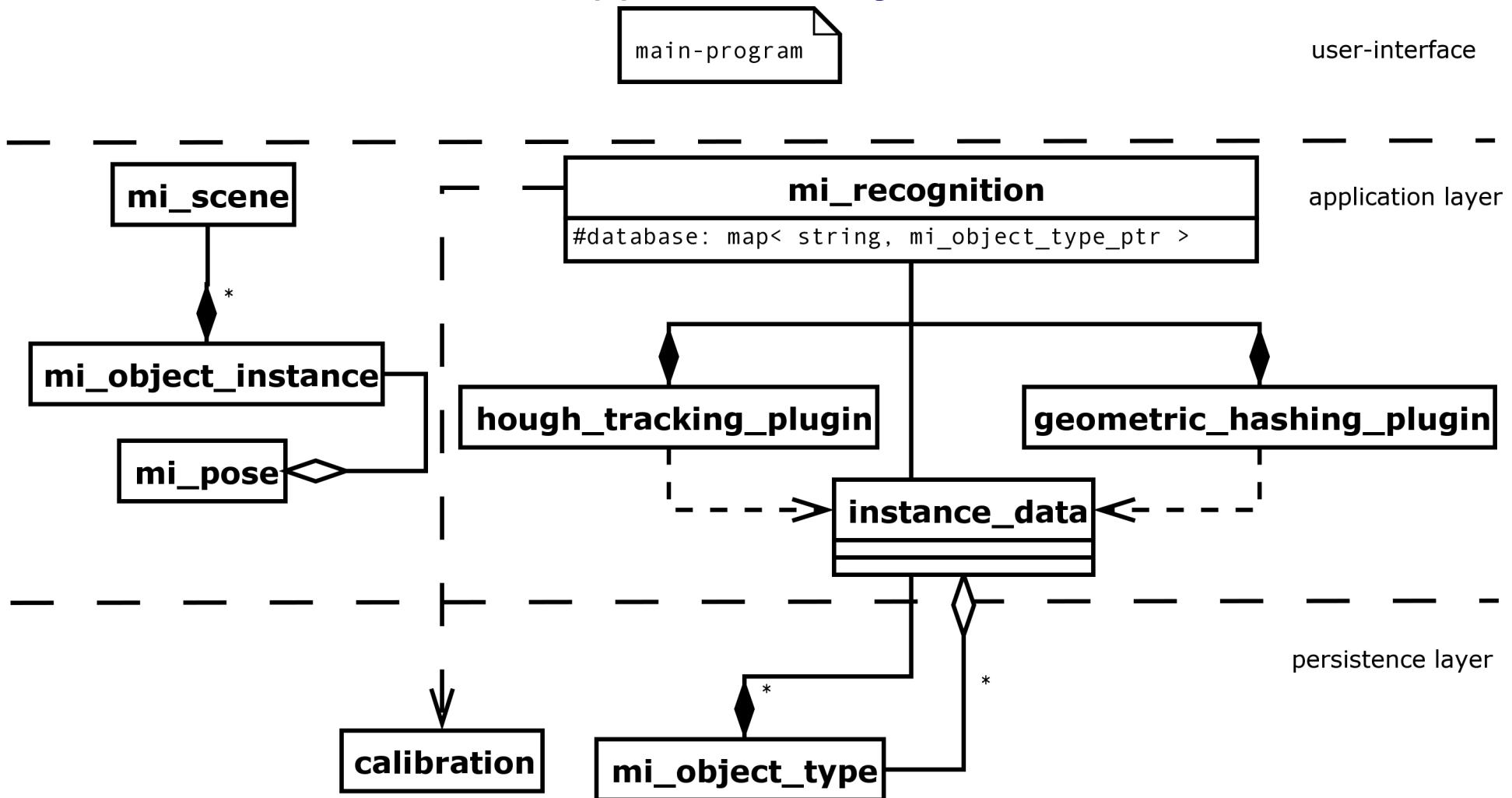


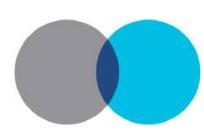
Focus Stack



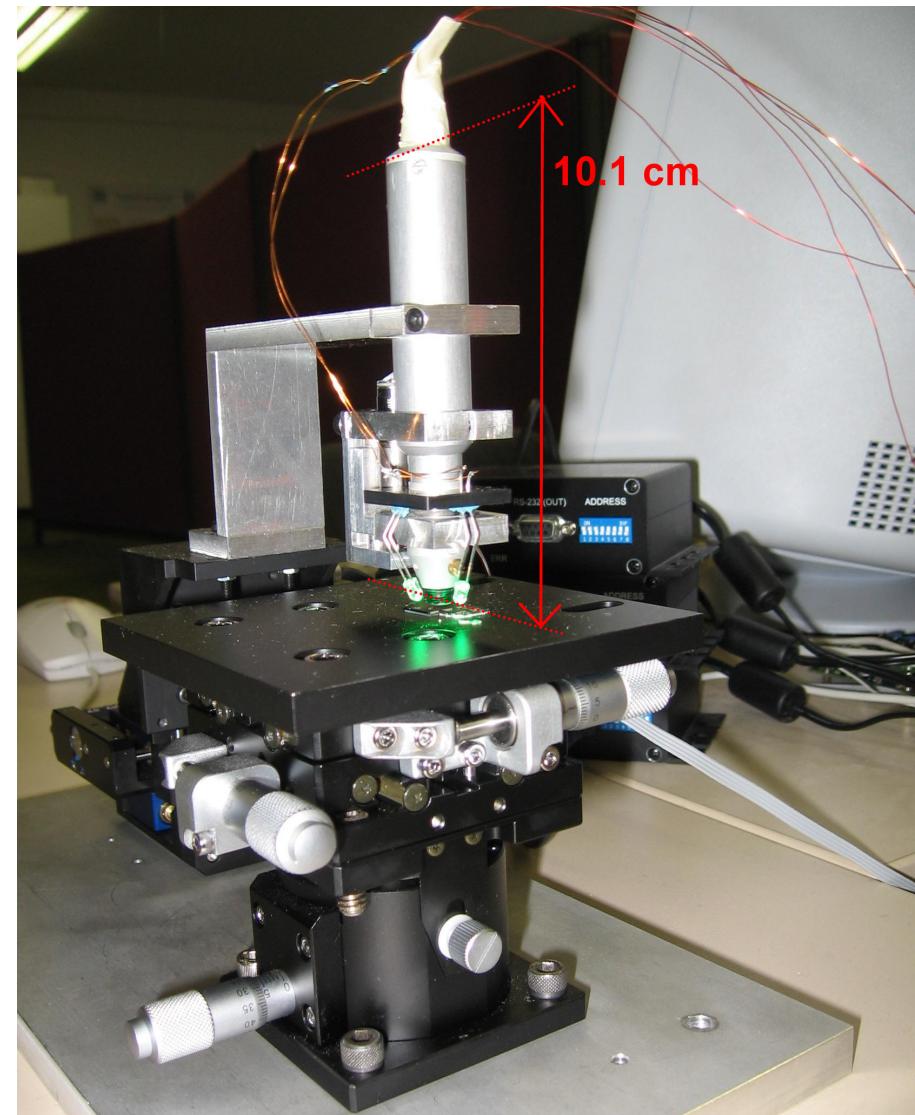


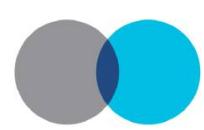
## Application Layers



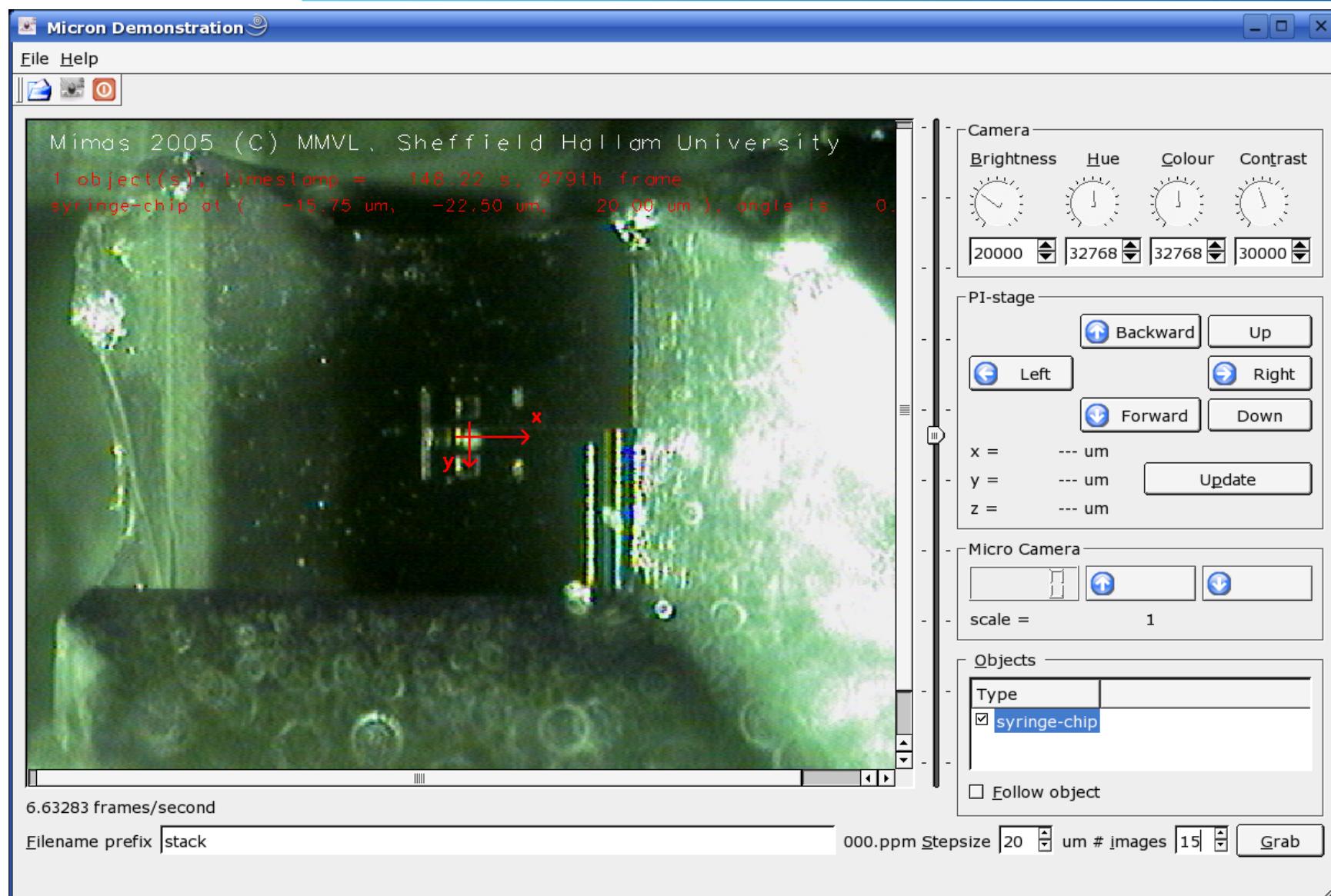


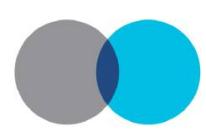
# Implementation Microstage with Custom-build Camera





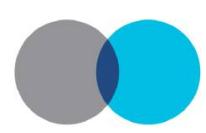
# Implementation Graphical User Interface



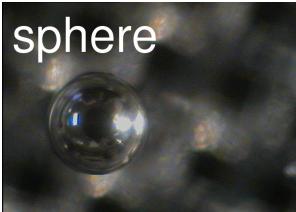


## Pushing Sugar

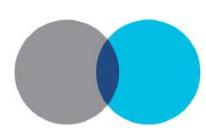




## Results (i)

video	reso- lution (down- sampled)	time per frame (recogni- tion)	stack size	degrees- of- freedom	recog- nition- rate	time per frame (tracking)
dry run (load frames only)	$384 \times 288$	0.0081 s	-	-	-	-
sphere	 $384 \times 288$	0.20 s	7	$(x, y, z)$	88%	0.020 s
syringe-chip	 $160 \times 120$	0.042 s	10	$(x, y, z, \theta)$	87%	0.016 s

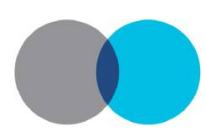




## Results (ii)

video	reso- lution (down- sampled)	time per frame (recogni- tion)	stack size	degrees- of- freedom	recog- nition- rate	time per frame (tracking)
	$384 \times 288$	0.27 s	16	$(x, y, z, \theta)$	88%	0.025 s
	$384 \times 288$	0.072 s	14	$(x, y, z, \theta)$	88%	0.018 s
	$192 \times 144$	0.32 s	9 1	$(x, y, z, \theta)$ $(x, y, \theta)$	35% 45%	0.022 s

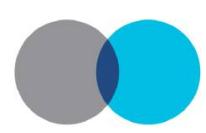




## Conclusion

- Depth estimation based on a single image is possible
- Real-time was achieved
  - Real-time recognition possible with low recognition-rate
  - Low recognition-rate much more tolerable than low frame-rate
  - Real-time tracking solves problem of low recognition-rate
- Focus stack must not be self-similar
- Rough surfaces are rich in features





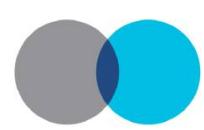
### Problems and Possible Solutions

Problem	Solution
Geometric Hashing scales badly with number of objects	Use RANSAC with Linear Model Hashing
High memory requirements, parametrisation for more than 2 objects is difficult	Use local feature context, use less features, use only salient features
Sub-sampling decreases accuracy	Implement non-uniform partitioning of Hough-space (adaptive accuracy)

### Research Topics

- more than 4 degrees-of-freedom
- develop micro-assembly for industrial application
- develop semi-automated supporting microscope tool for biological application





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**Microsystems and Machine Vision Laboratory**

**The Microsystems and Machine Vision Laboratory (MMVL)** is a research group within the [Materials and Engineering Research Institute \(MERI\)](#) at [Sheffield Hallam University](#).

Our main research activities involve the design, development and implementation of machine vision techniques targeted at a variety of real-time and non real-time applications which include microrobotic systems, biological applications, micromanipulation, microscope imaging, SEM/TEM applications and non-destructive testing of MEMS devices.

**Latest news:**

**NEW!** 20 March 2006  
Mimas Toolkit v2.0 released. See the [Mimas website](#) for details.

**NEW!** 9 March 2006  
PhD studentship available. See the [MMVL Wiki website](#) for details.

**NEW!** 9 March 2006  
Call for papers. See the [IEEE AICS2006 website](#) for details.

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 **Microscope Vision Software**

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- 1.3.2 Test Data

**2 See Also**

**3 External Links**

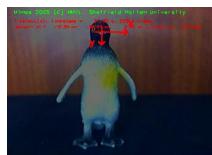
**MiCRoN Microscope Vision Software** [\[edit\]](#)

**Implementation** [\[edit\]](#)

Computer vision for microscopes has to deal with the problem of limited depth of field. But instead of trying to overcome this, one can actually use the depth information concealed in this images to achieve real-time object recognition for microscopes.

Employing algorithms based on **sobel edge detection**, **geometric hashing** and based on the **bounded hough transform**, it was possible to recognise

[www.shu.ac.uk/research/meri/mmvl/](http://www.shu.ac.uk/research/meri/mmvl/) [vision.eng.shu.ac.uk/mediawiki/](http://vision.eng.shu.ac.uk/mediawiki/)



Open source MiCRoN vision software + test data  
Open source Mimas real-time computer vision library

