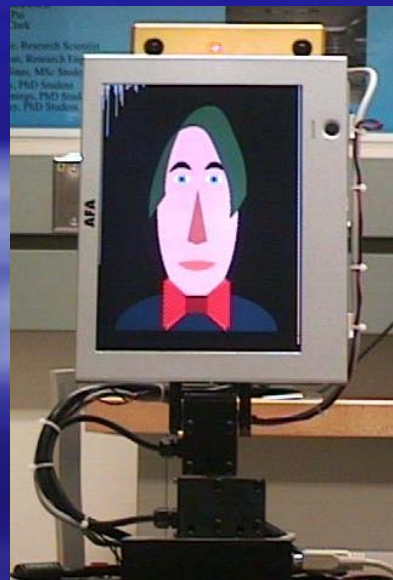




# Interaction with an autonomous agent

Jim Little  
Laboratory for  
Computational Intelligence  
Computer Science  
University of British  
Columbia  
Vancouver BC Canada



# Background

When we build systems that interact with the world, with humans, and with other agents, we rely upon all of the aspects of cognitive vision:

- knowledge representation
- descriptions of the scene and its constituent objects
- models of agents and their intentions
- learning
- adaptation to the world and other agents
- reasoning about events and about structures
- interpretation of other agents' and users' interactions
- recognition and categorization

# Background (cont.)

I will review the ongoing Robot Partners project at UBC which focuses on the design and implementation of visually guided collaborative agents, specifically interacting autonomous mobile robots. I will show how

- localization and mapping
- user modeling
- interpretation of gestures and actions
- interaction with human agents

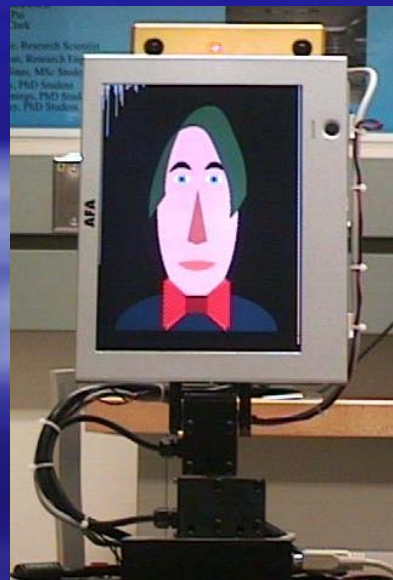
are achieved in the context of the project





# Cognitive Vision for Agents Robot Partners: Visually Guided Multi-agent Systems

Jim Little  
Laboratory for  
Computational Intelligence  
Computer Science  
University of British Columbia  
Vancouver BC Canada



# Personnel

- Jim Little (Project Leader), CS UBC
- James J. Clark, ECE McGill University
- Nando de Freitas, CS UBC
- David G. Lowe, CS UBC
- Alan K. Mackworth, CS UBC
- Dinesh K. Pai, CS UBC
- Stephen Se, MD Robotics

# Research Focus

To develop techniques for the specification, design and implementation of collaborative robotic systems, based on rich interaction between humans and sensor-based robotic systems.

New tools:

- stochastic constraint-based design
- real-time systems for vision
- control with event-based interaction with perception
- stereo-based shape and appearance models
- robot-human communication based on gestures, sounds, and spoken commands
- precise visually guided localization for mobile agents

# Benefits

- Theories and tools for modelling, developing and verifying constraint-based controllers for stochastic agent environments.
- Multi-agent algorithms for collaboration and competition in e-games with mixed initiative (human/robot) control.
- Location recognition capability for mobile agents
- Theories and implementations of human/robot communication through gestures and speech
- Mobile robot collaboration on mapping and model building
- Distributed surveillance and monitoring



# Applications

Autonomous and semi-autonomous systems that assist and partner with humans:

- warehouse and inventory control systems
- construction systems: teams of vehicles for surveying, collecting, and retrieval
- office assistants
- mapping and modeling; surveillance and monitoring
- remote agents for telepresence in meetings, tours and lectures

Our methods/technologies, such as pose estimation, tracking, object/world modeling, localization, and model learning, have direct application in all embedded systems, as well as space applications, including monitoring activity and autonomous and semi-autonomous exploration.

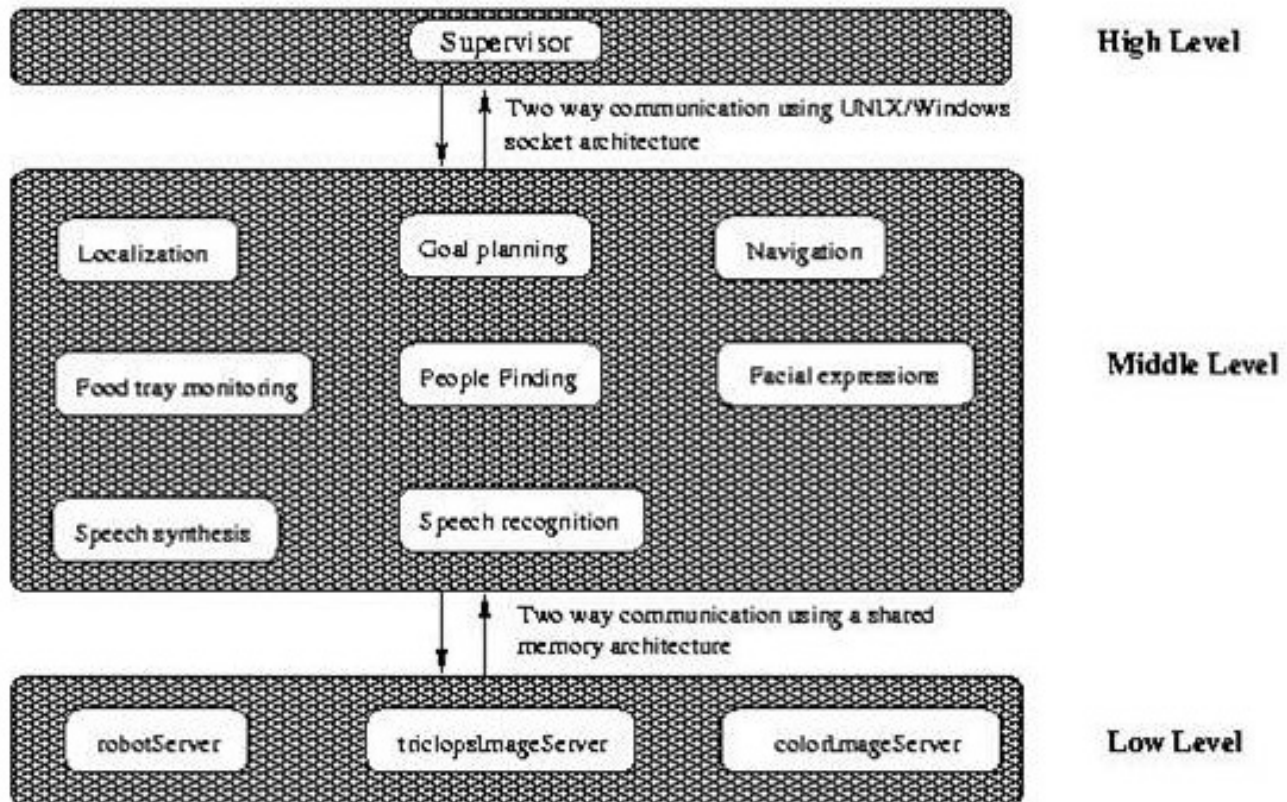


# José: Autonomous service

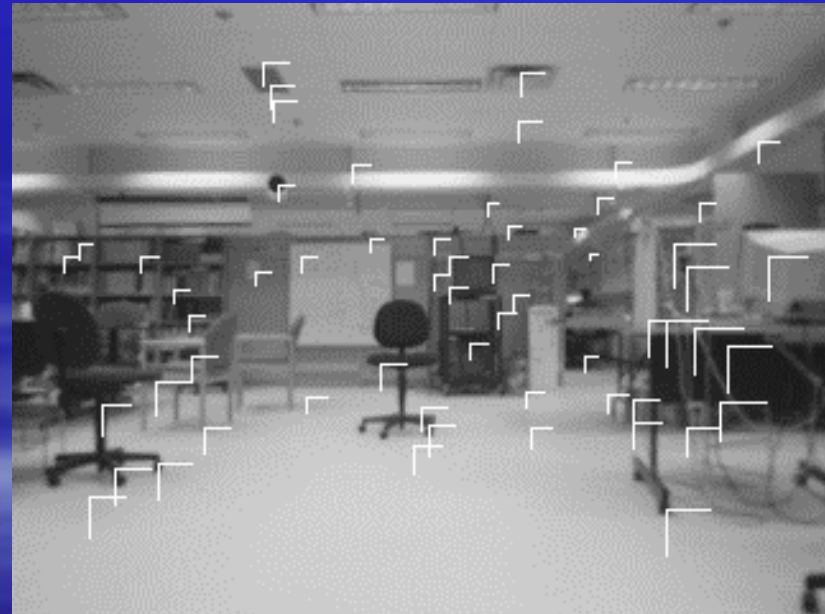
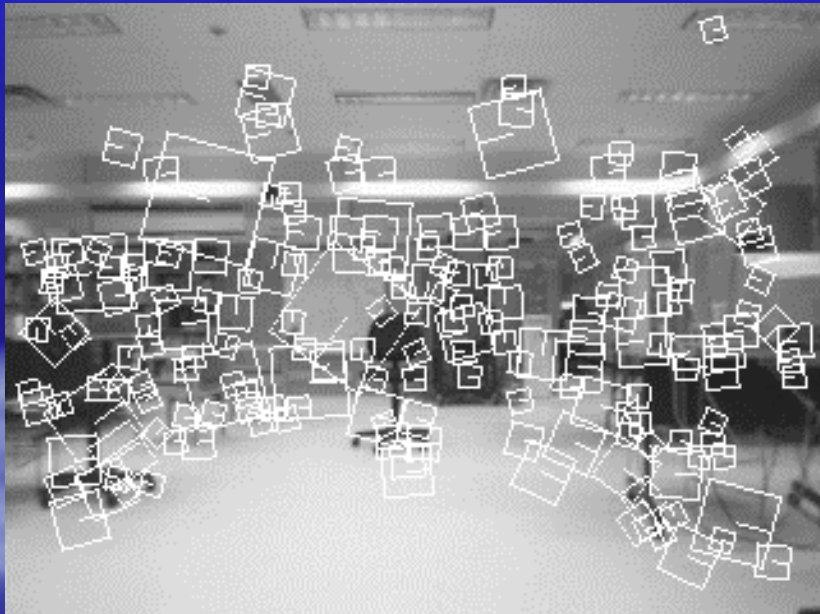
- Robot Control Architecture
- Localization
- Navigation - Avoiding dynamic obstacles
- People Finding
- Location Decision: Where to serve next?
- Food Tray Monitoring
- Face Modeling
- Eric's persona
- Speech Recognition and Synthesis
- Interaction



# Control Architecture



# SIFT-base localization



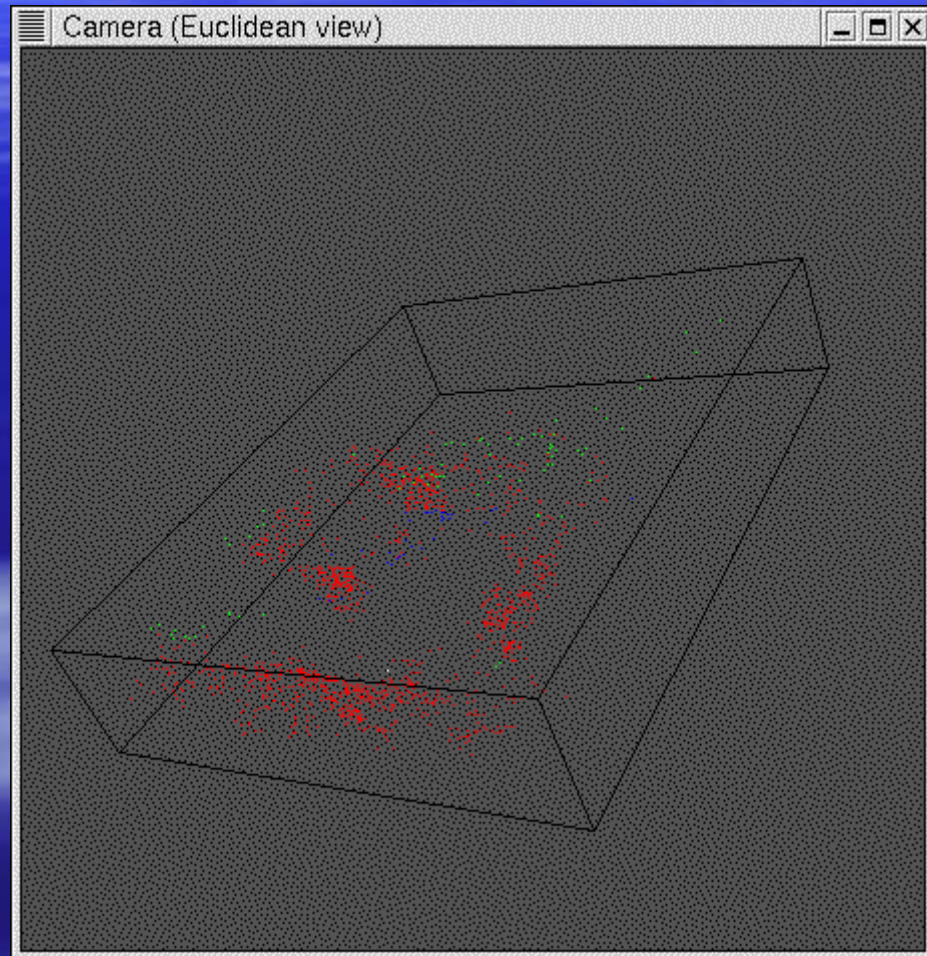


# 3D Map

**Red:** Features on the Walls

**Blue:** Features on the Floor

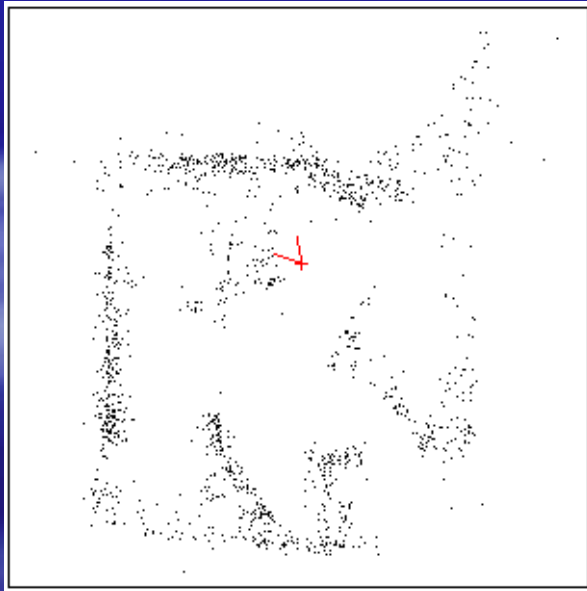
**Green:** Features on the Ceiling





# Global Localization

- Kidnapped robot problem
- Recognize robot pose relative to a pre-built map
- Hough Transform matching approach
- Vote bins for pose of each potential match
- Peak : pose with most matches



Measured Pose :  
(70, 300, -40°)

Estimated Pose :  
(75.8, 295.9, -41.1°)

# Face



Neutral



Surprised



Angry



Sad

# Finding People

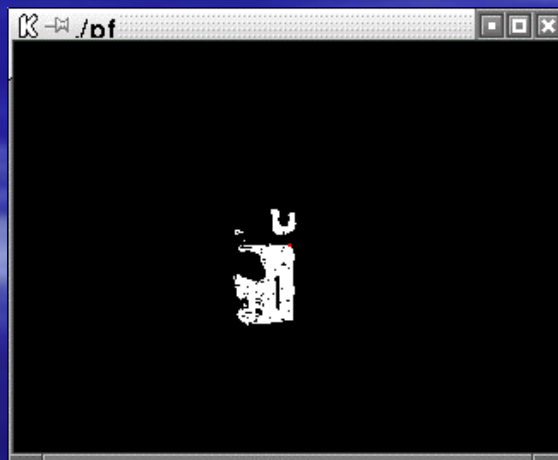
- Construct occupancy grid probability map of where people are standing
- Use the map to decide where to serve next
- Detect people using skin color segmentation
- Use stereo data to compute 3D position of people
- Project locations to floor plane
- Decrease the probabilities over time because people move around

# Finding People

- Use occupancy grid probability map to decide where to serve
- Detect people using skin color
- Use stereo data to compute 3D position of people
- Decrease the probabilities over time because people move



Color Image



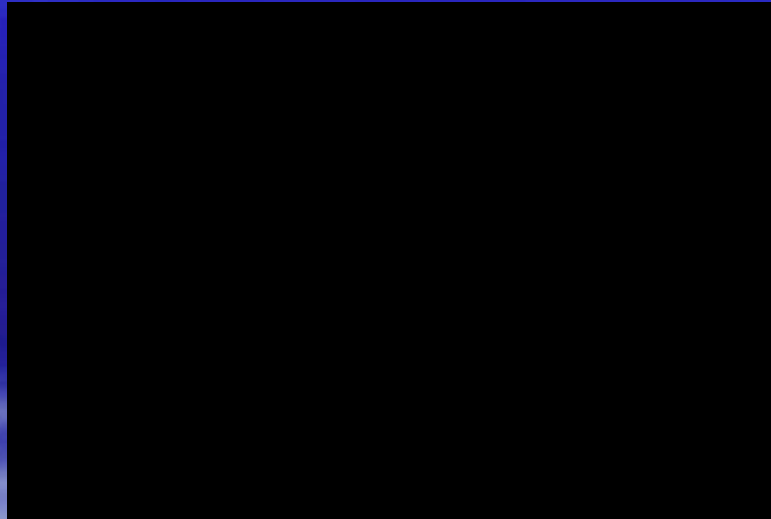
Skin Regions



Probability Map



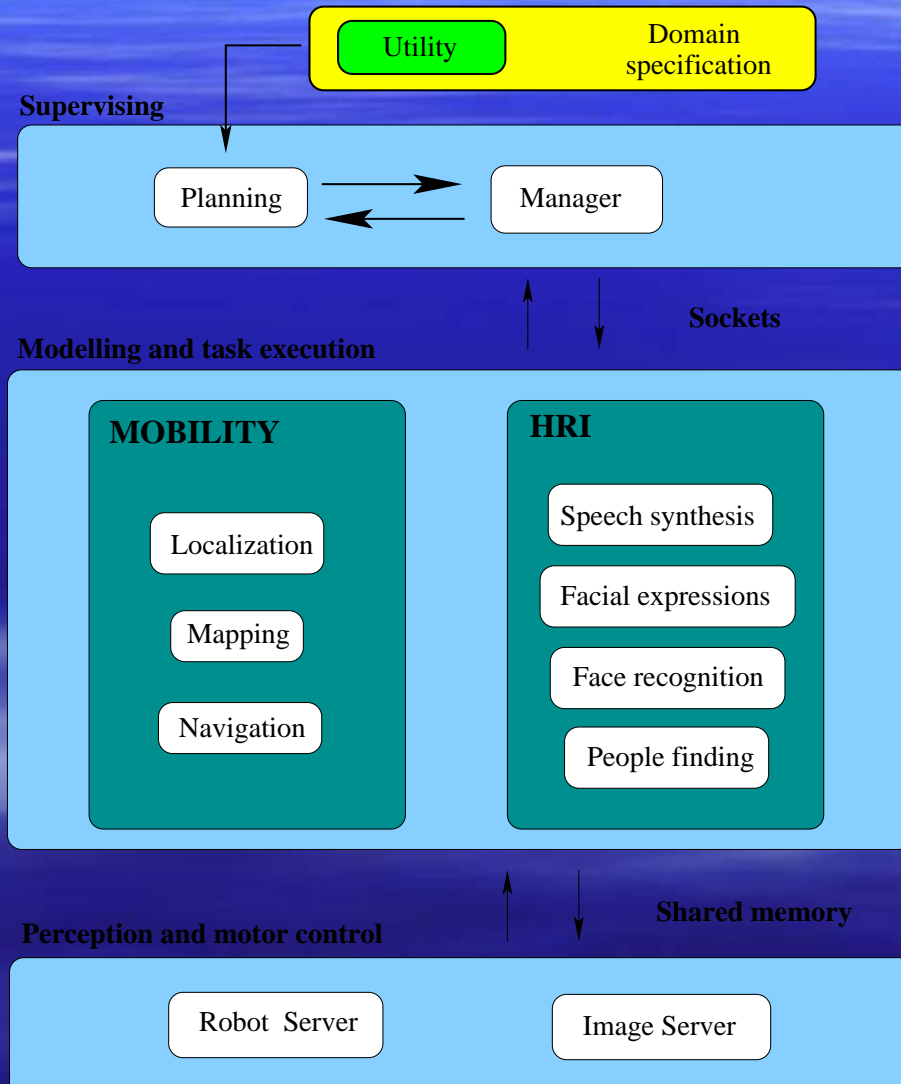
# Homer: Human Oriented Messenger Robot



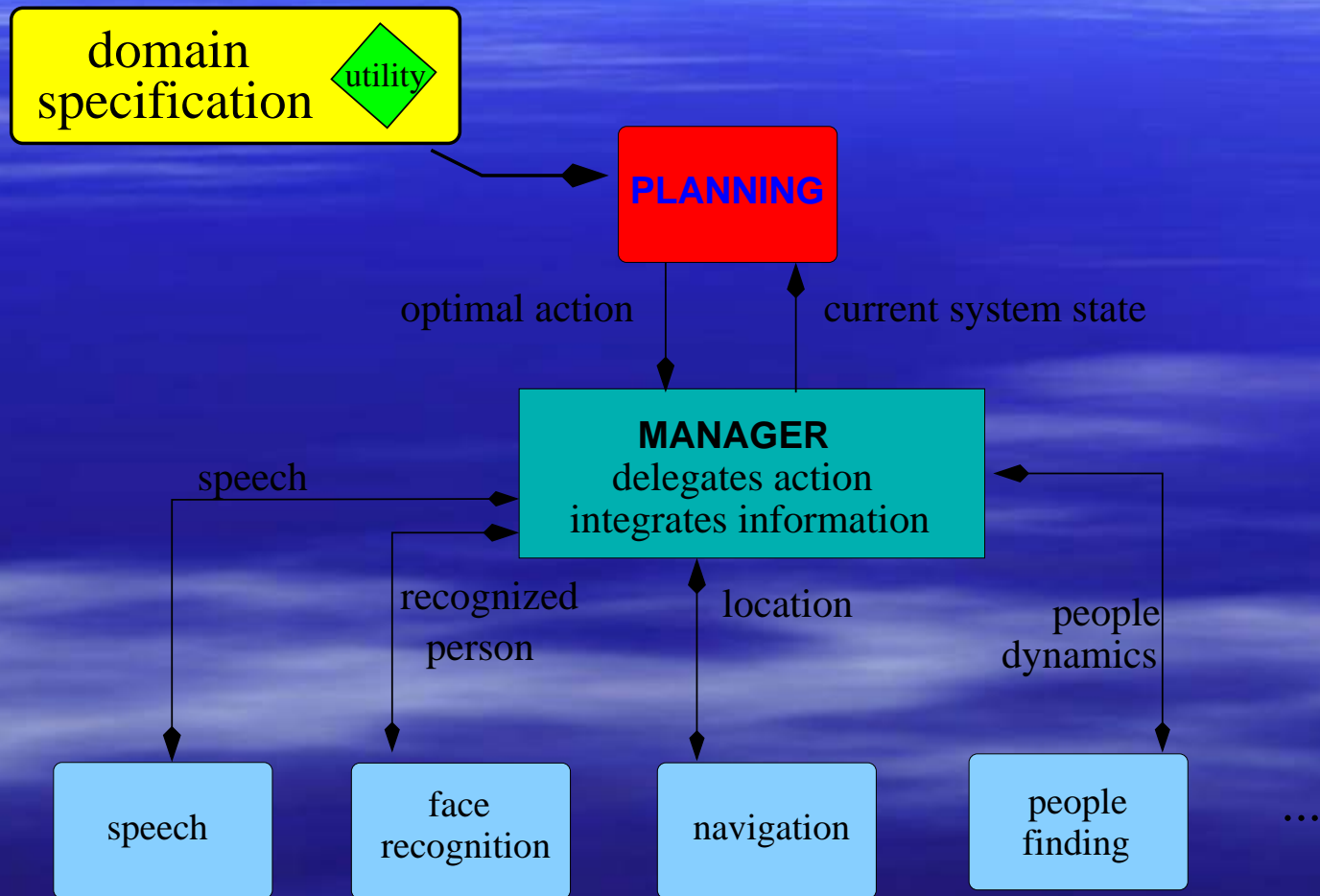
# Homer: Human Oriented Messenger Robot

- stereo-vision guided mobile robot for performing human-interactive tasks.
  - navigation, localization, map building and obstacle avoidance
  - human interaction capacities
    - person recognition
    - speech,
    - facial expression and gesture recognition
    - human dynamics modeling.
- capabilities
  - modular and independent,
  - integrated in a consistent and scalable fashion
  - controlled by a decision-theoretic planner to model the uncertain effects of the robot's actions
- planner uses factored Markov decision processes, allowing for simple specification of tasks, goals and state spaces.
- Task: message delivery task

# Control Architecture



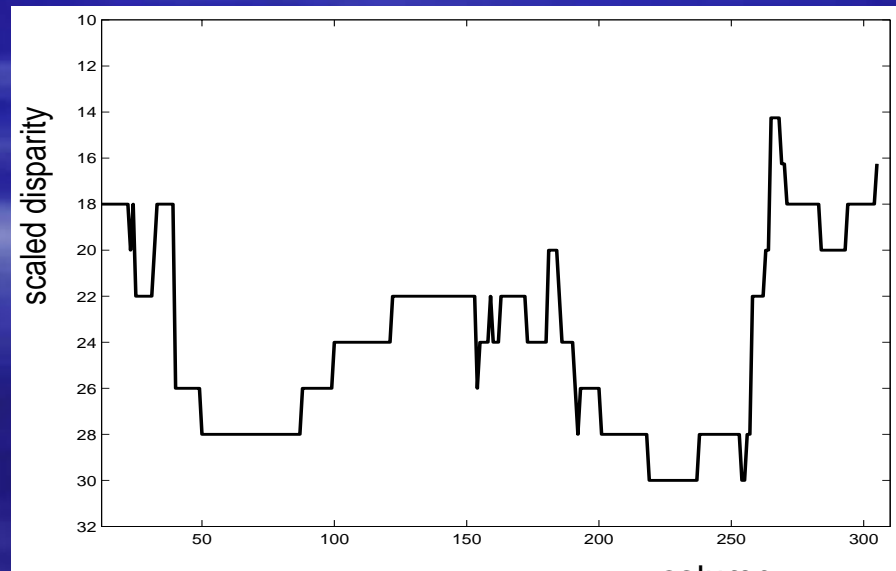
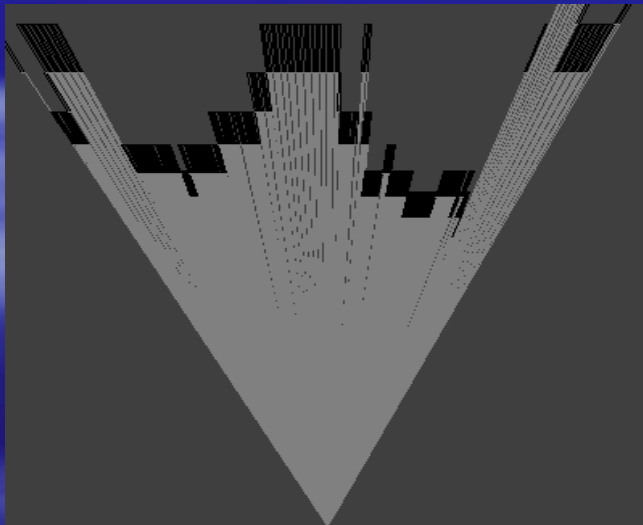
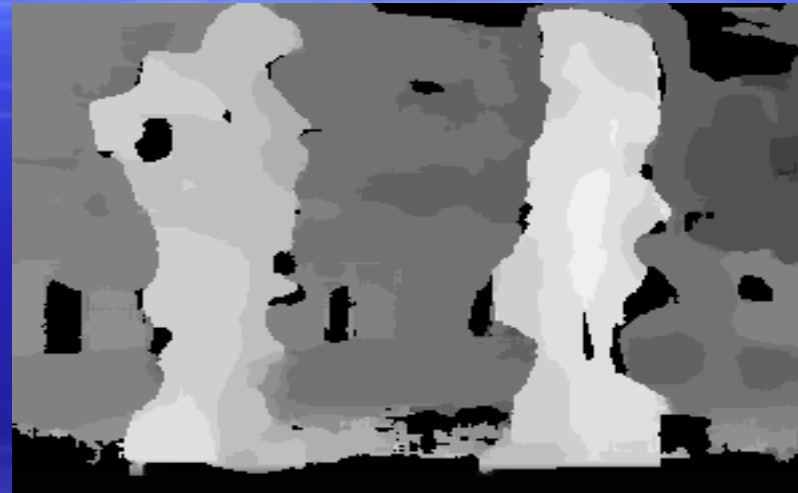
# Task Organization



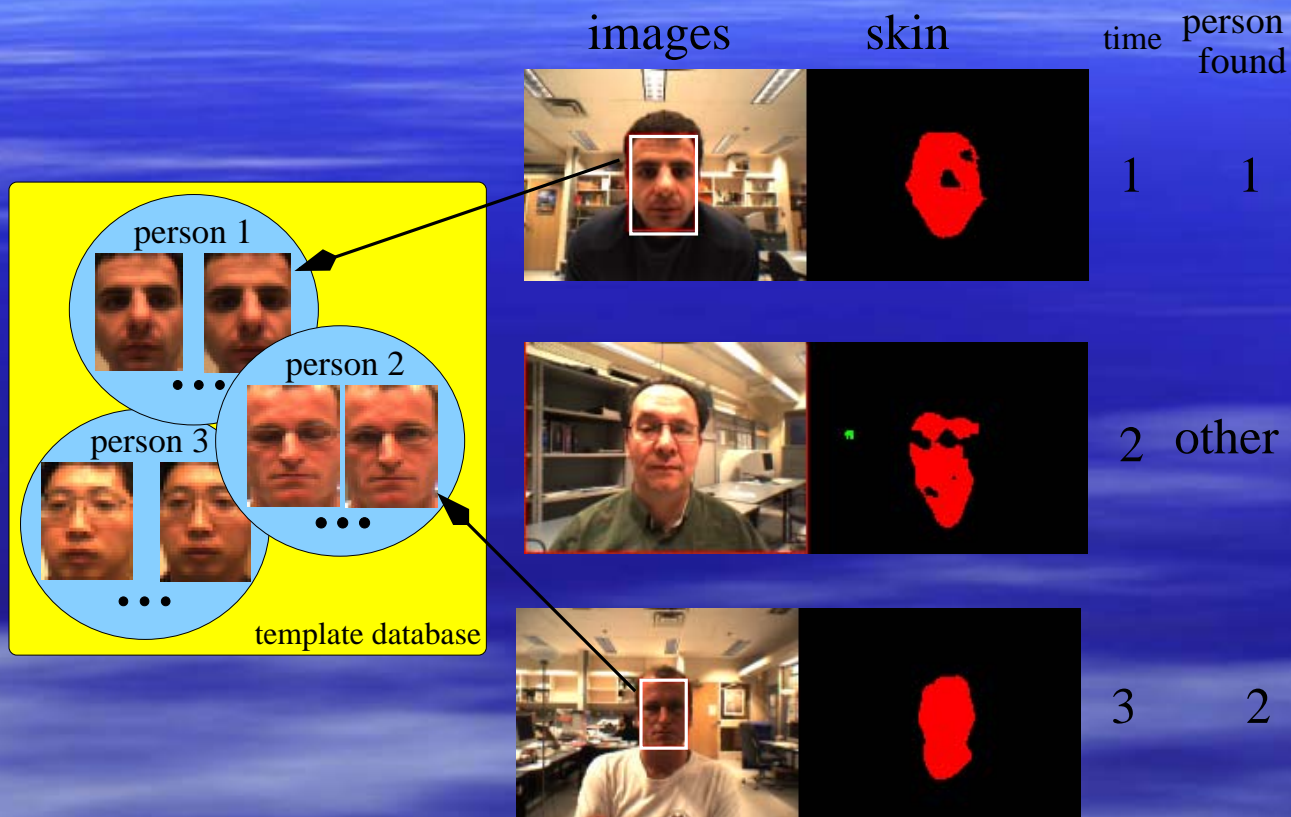
modules control actuators on robot



# From stereo to maps



# Face recognition



Exemplars for three persons, input images linked to their most likely and most likely match scores and reported person.

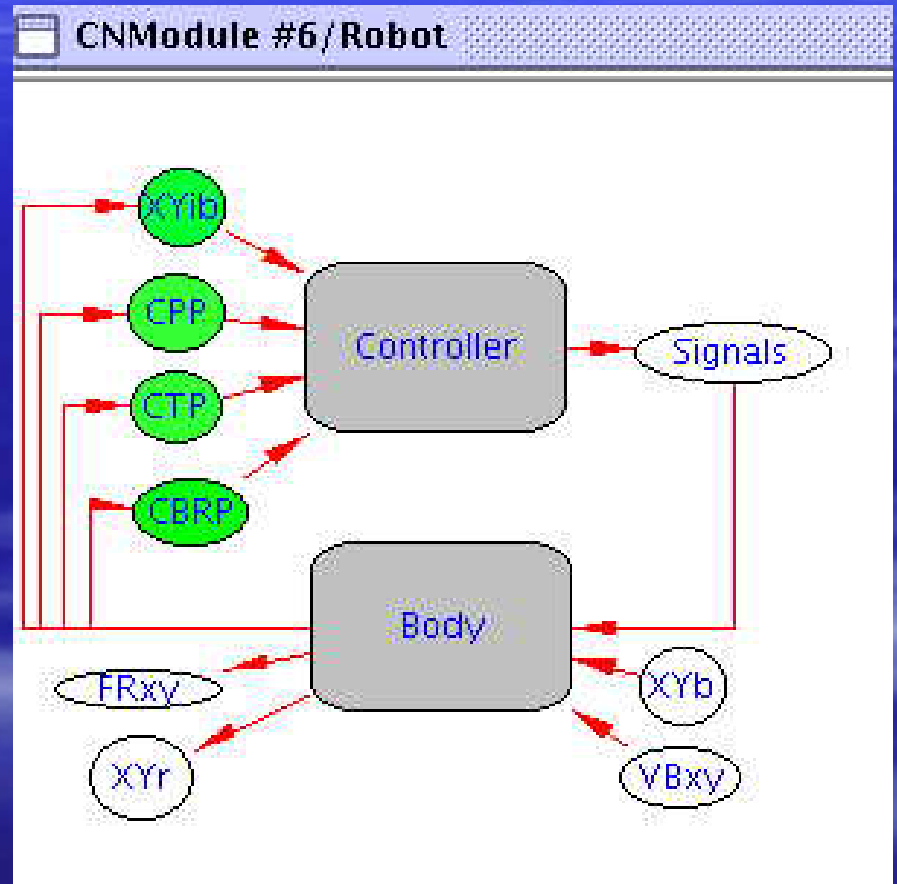


# Constraint Nets: theory, tools, applications

- Robert St-Aubin & Mackworth are designing and building Probabilistic Constraint Nets (PCN) for representing uncertainty in robotic systems.
- Song & Mackworth have designed and implemented CNJ, a visual programming environment for Constraint Nets, implemented in Java.  
The system includes a specification and implementation of CNML, an XML environment for Constraint Nets.
- CNJ is now being used as a tool by Pinar Muyan to develop a constraint-based controller for a simple robot soccer player.

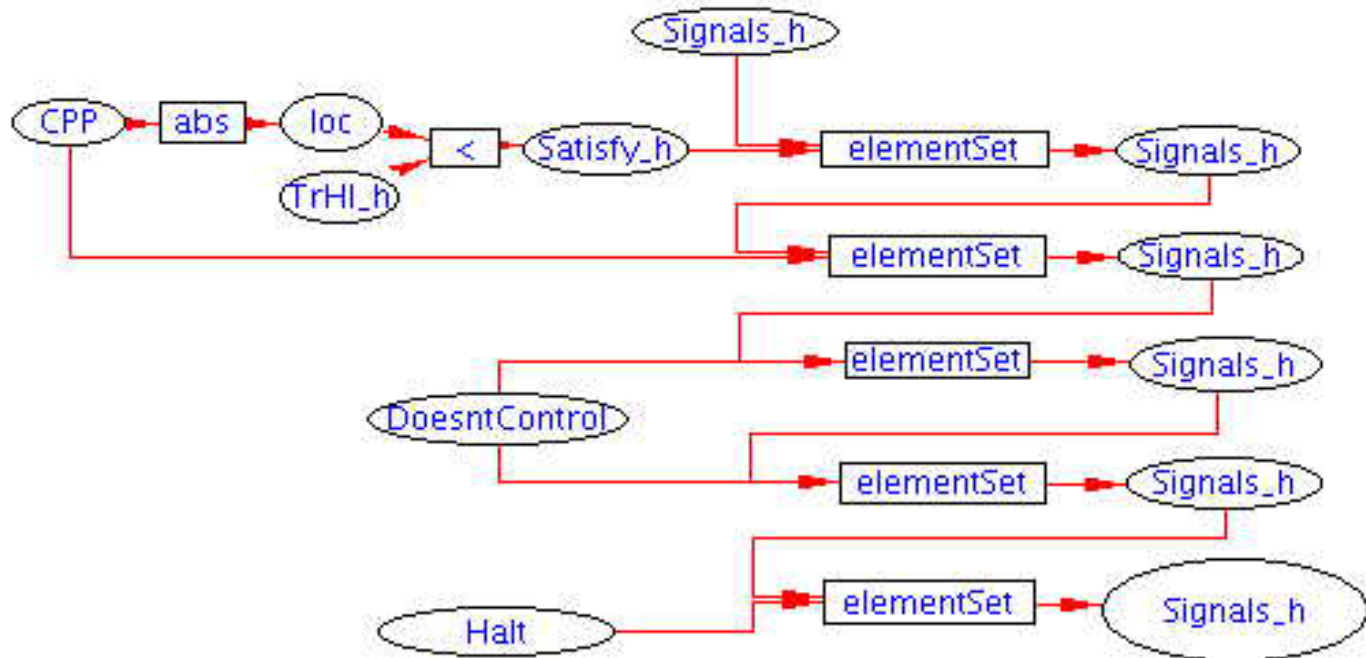


# CNJ: Robot in pursuit of ball

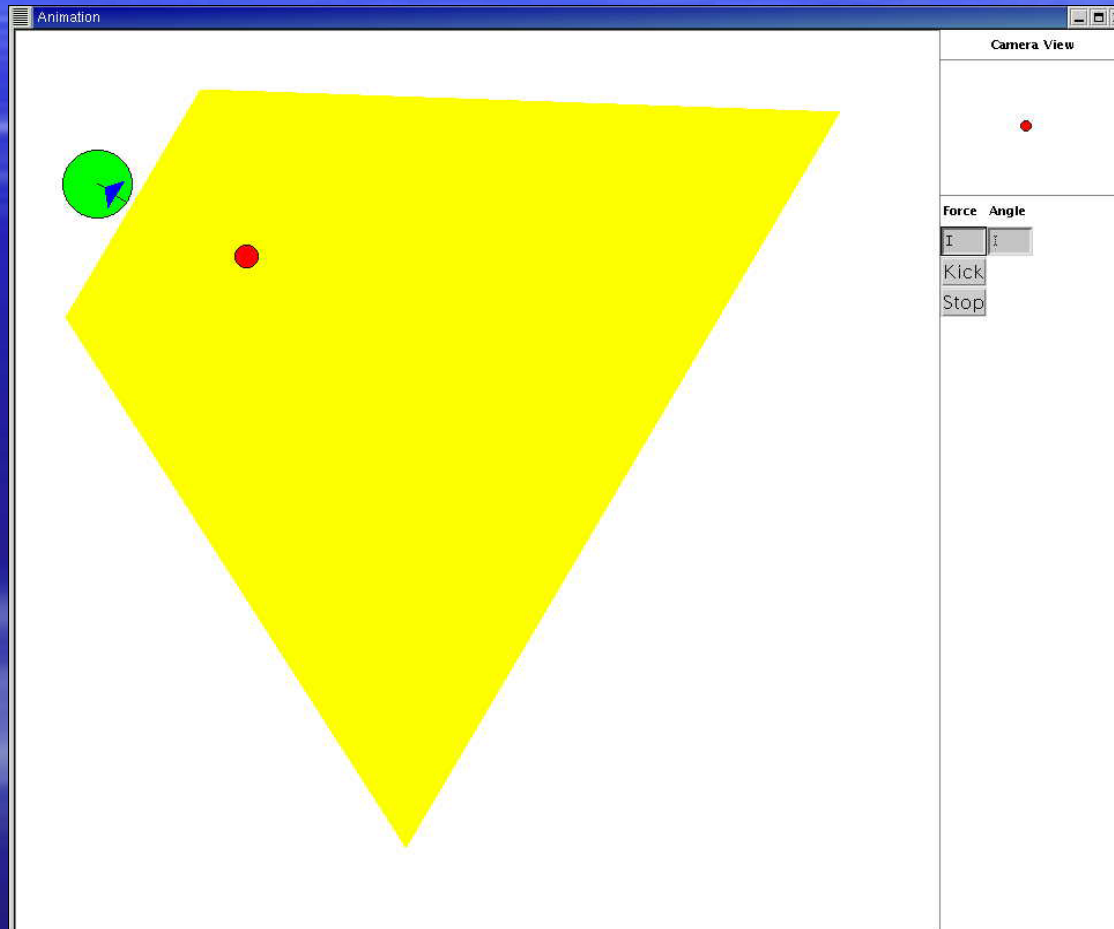


# CNJ: Controlling Rotation/Pan/Tilt

CNModule #13/BaseHeadingPan

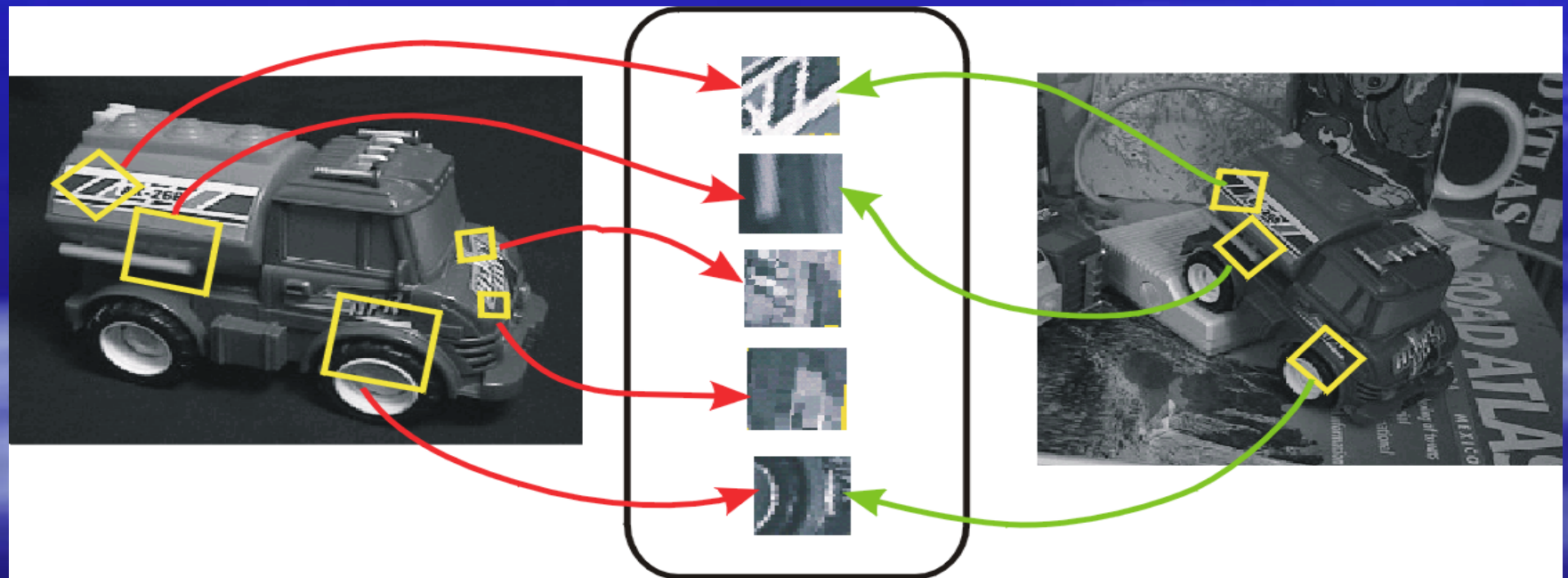


# CNJ: Animation of controller



# Scale Invariant Feature Transform (SIFT)

- Image content is transformed into local feature coordinates that are invariant to translation, rotation, scale, and other imaging parameters



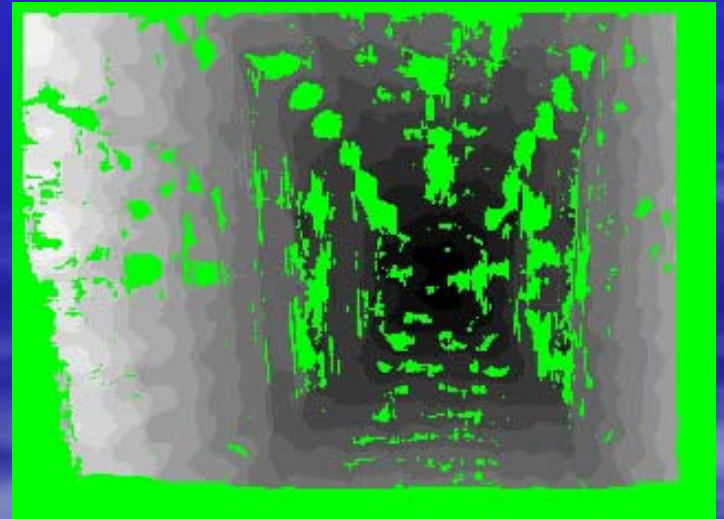
SIFT Features



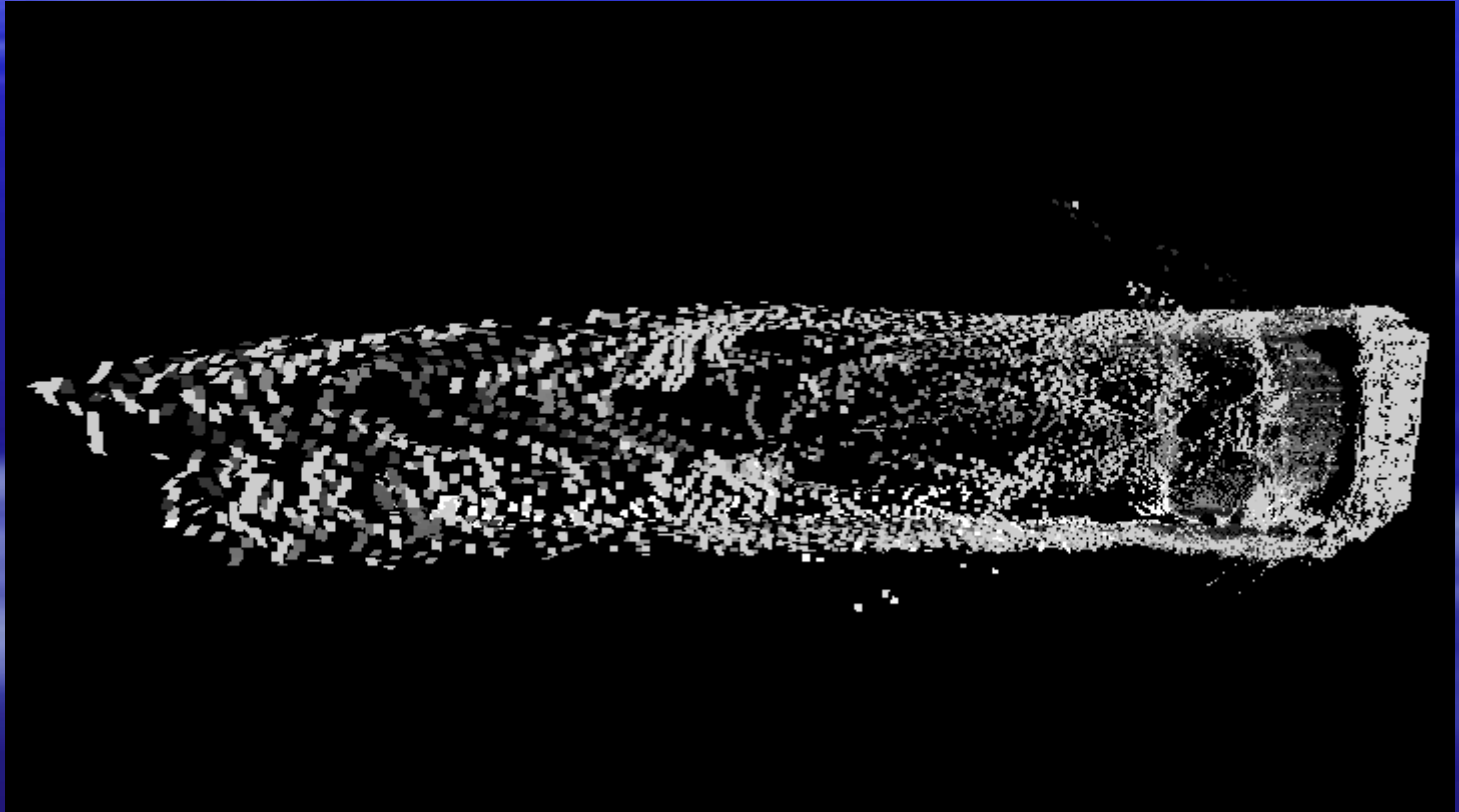
# Patchlet Surface Representation

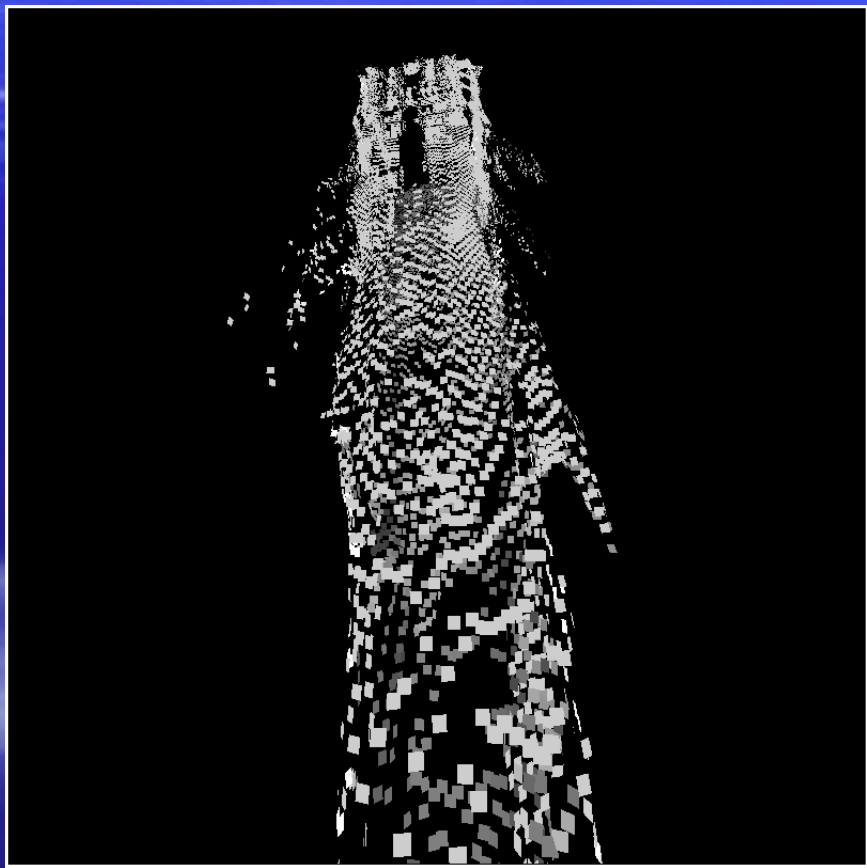
- Goal: to properly interpret the uncertainty of stereo measurements in surface reconstruction.
- The sensor elements considered are local patches in the stereo image that create patchlets.
- These patchlets are fit to a plane and the uncertainty of the plane in orientation and position is determined from the stereo 3d points.

# Brightness and Depth



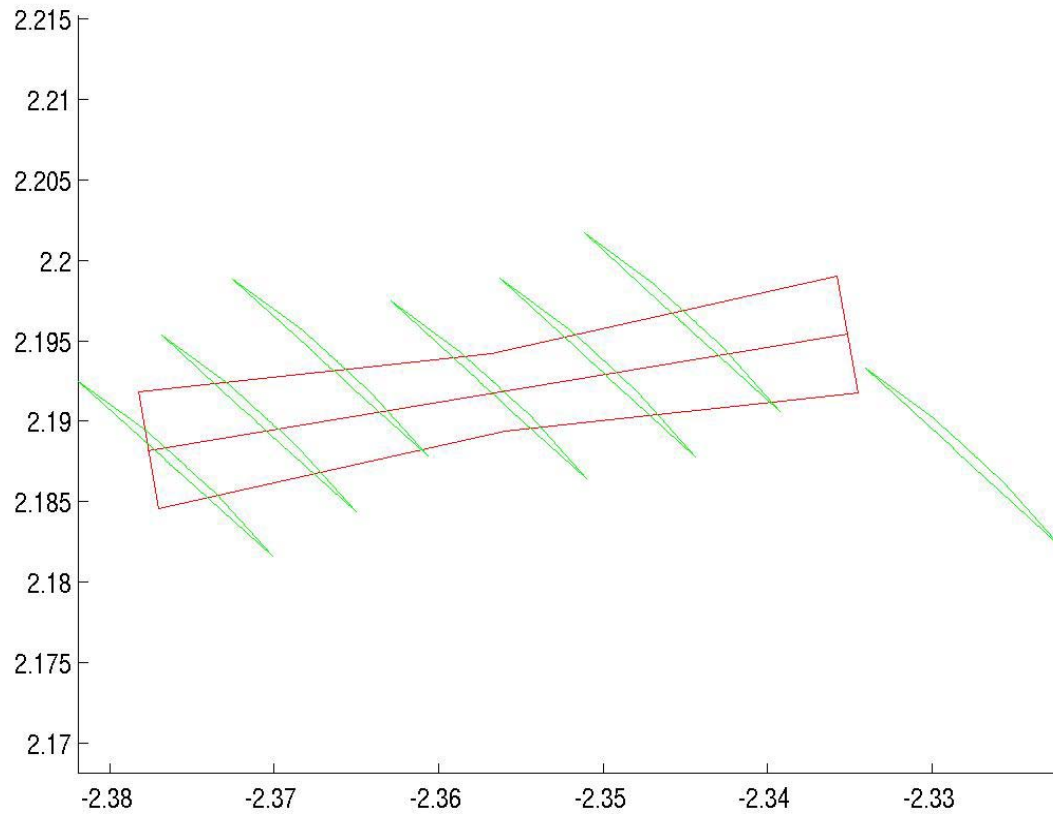
# Depth Pixels and Scale



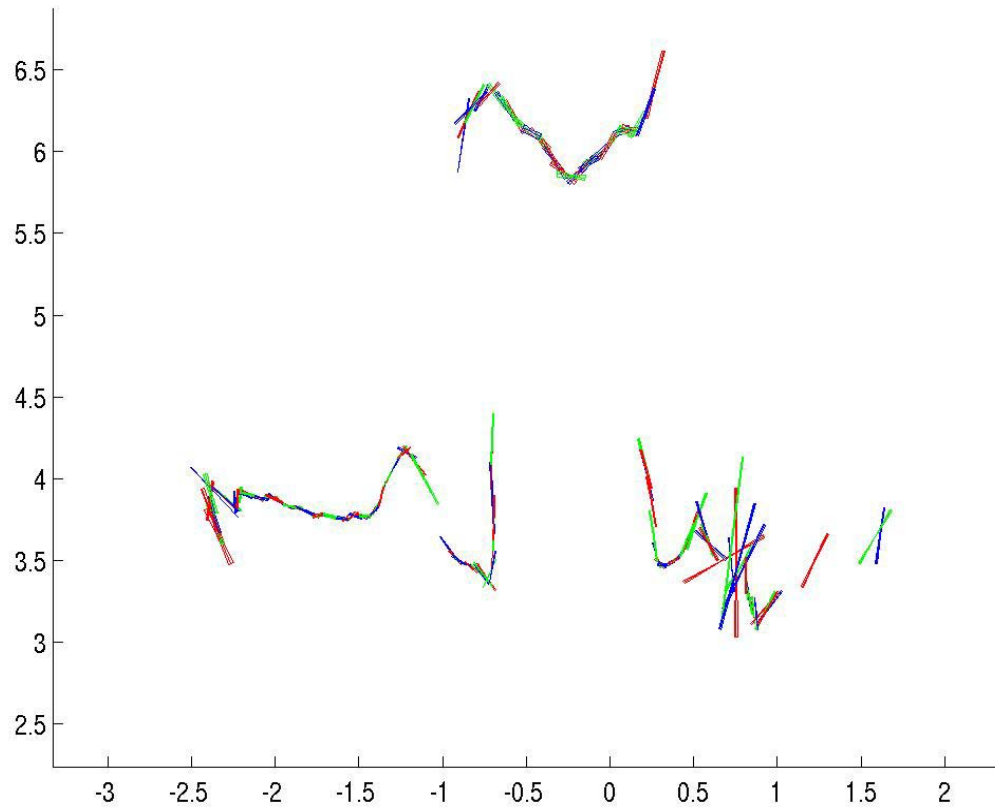




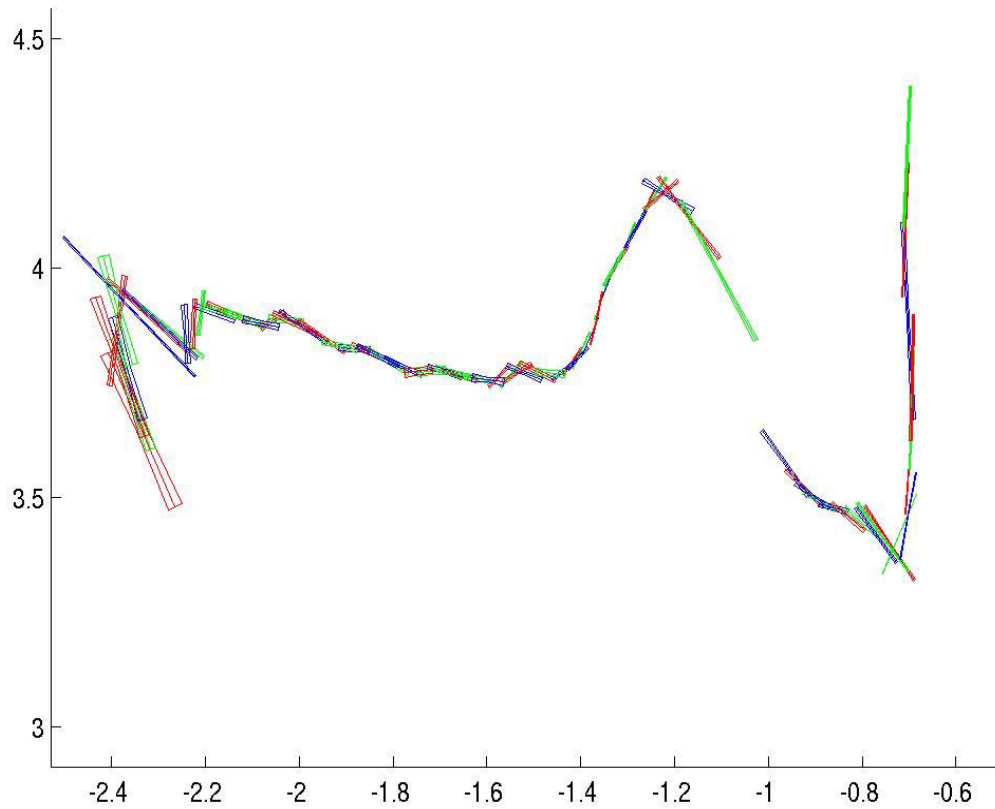
# Patchlet Uncertainty



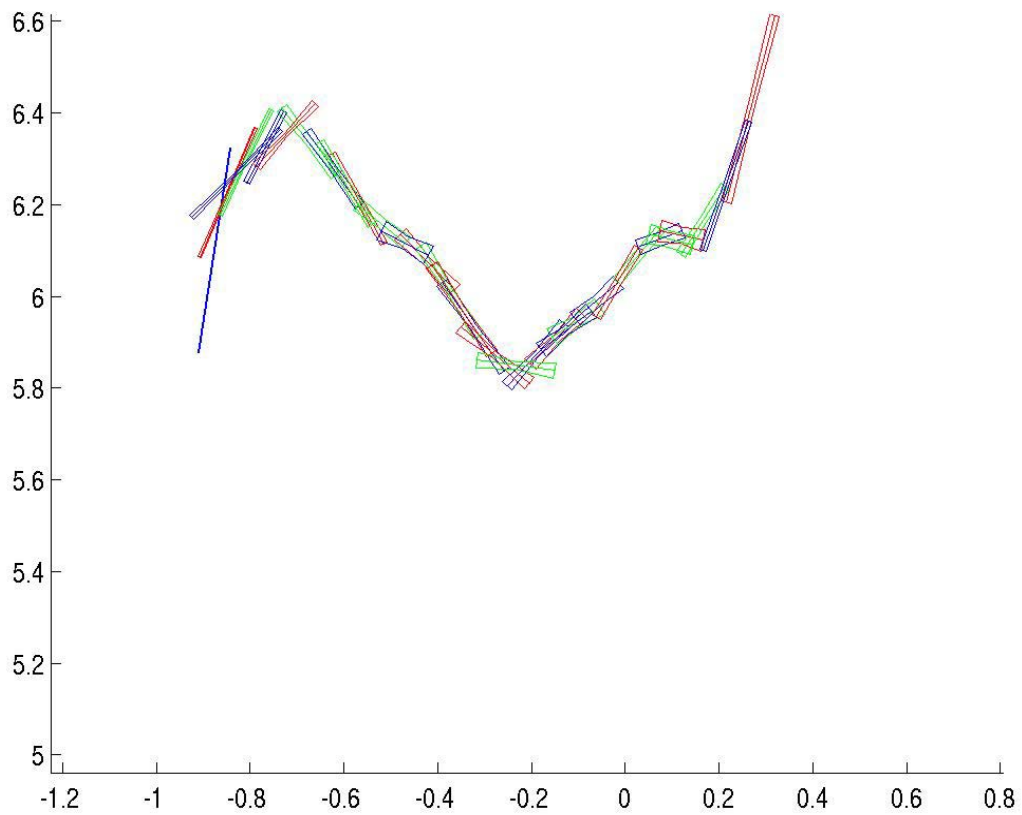
# Patchlets



# Near



# Far





# Representation and Recognition of Complex Human Motion

- Psychological research
- Video coding, search
- Human-computer interaction
- Articulated, non-rigid motion at many scales
- Find a general representation for any type of motion at any scale
- Show Zernike polynomials as ideal for this purpose

# Optical Flow

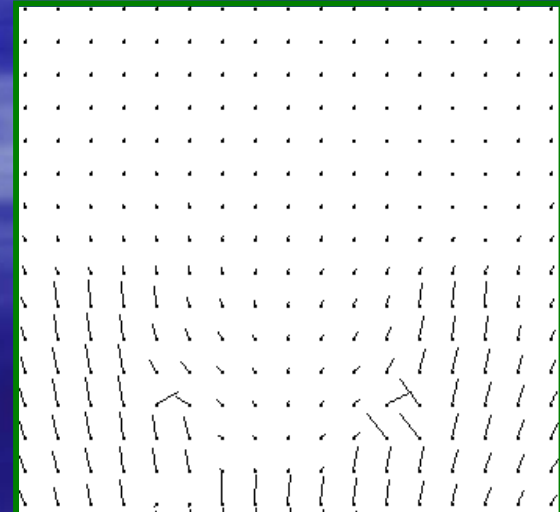
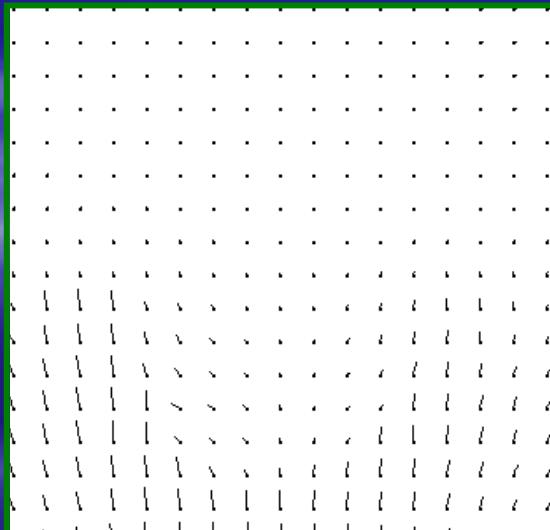
$t=0$



$t=1$

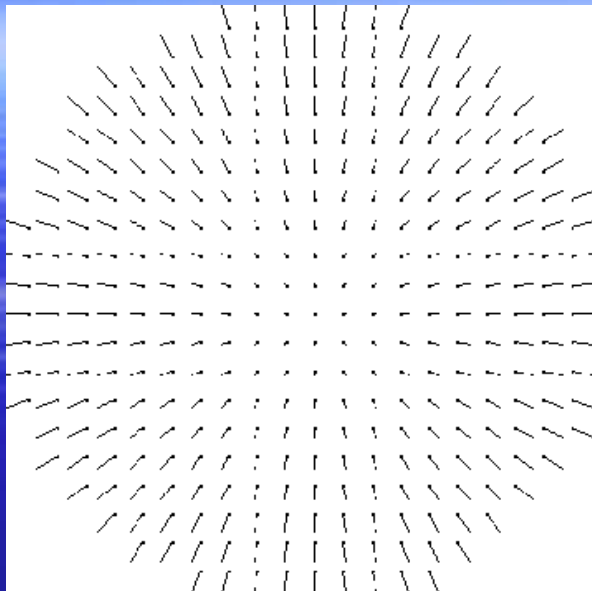


$t=2$

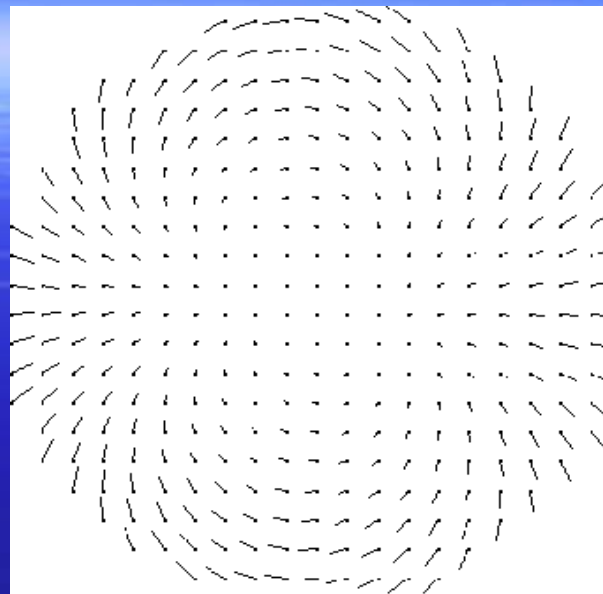


# Example Flows

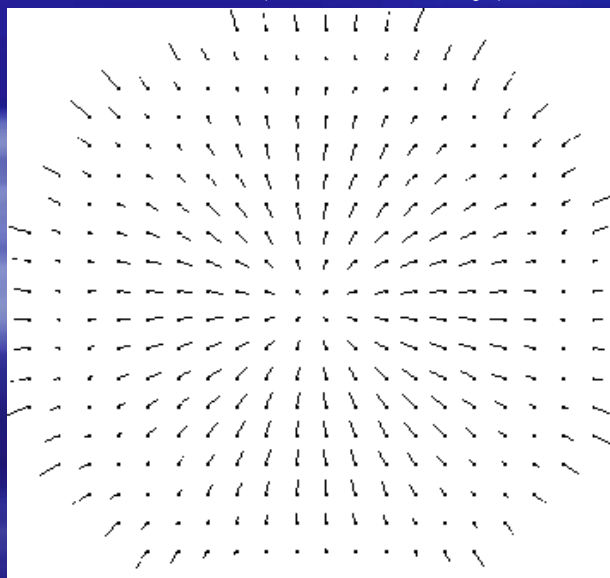
$n=1$   $m=1$  (affine)



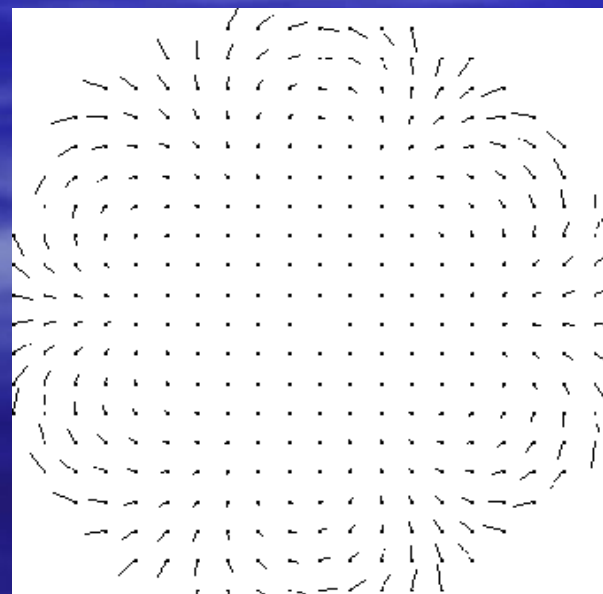
$n=2$   $m=2$



$n=4$   $m=0$  (radial only)

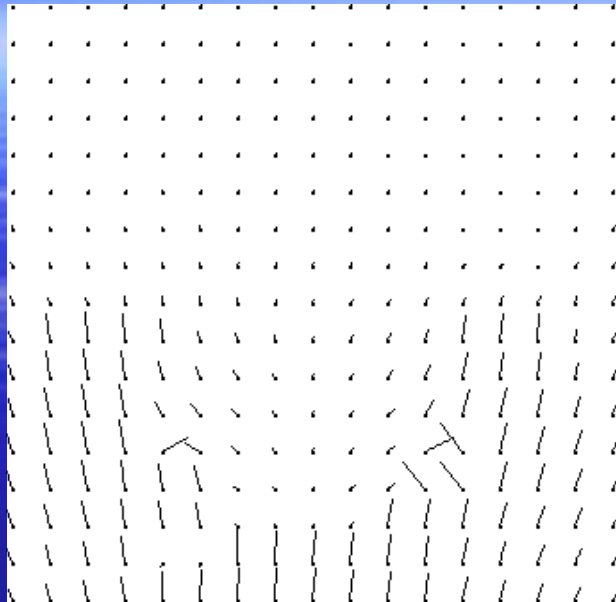


$n=4$   $m=2$

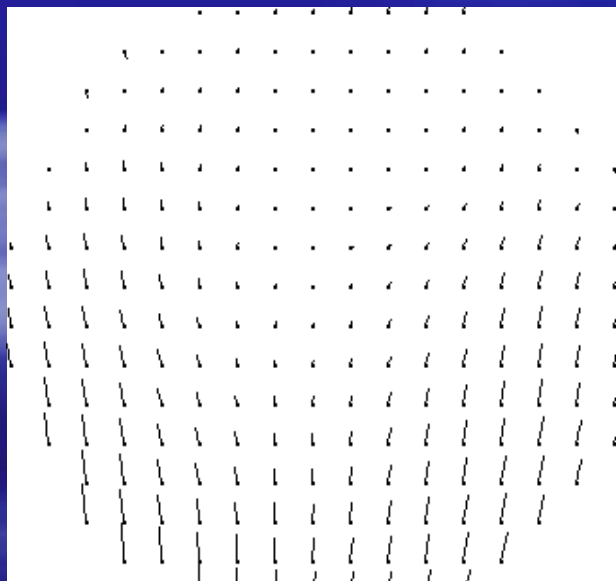


# Reconstructed Flows:

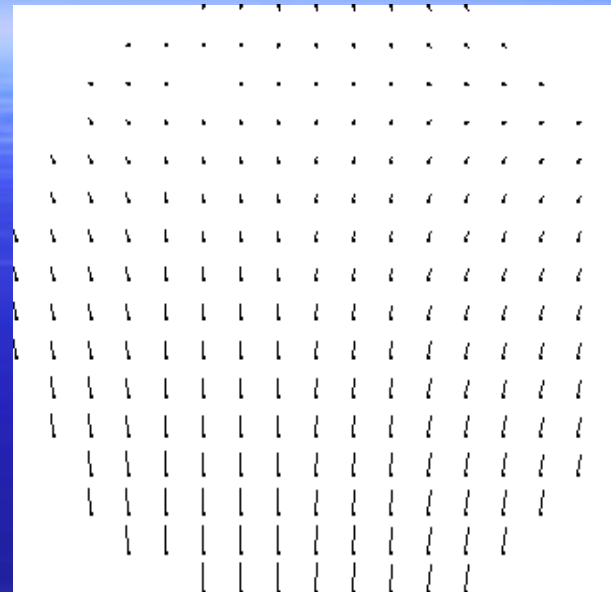
original flow field



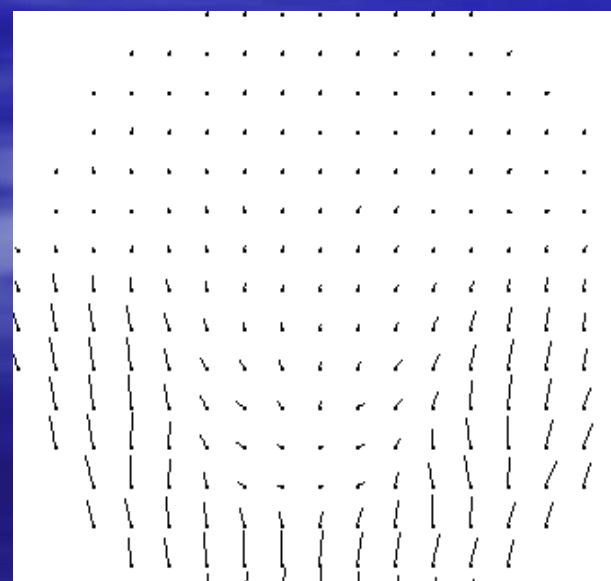
first 7 ZPs



affine flow



first 49 ZPs





# Facial expression

- no rigid head motion
- 72 subjects - 6 expressions\*

\*Cohn-Kanade Facial Expression Database



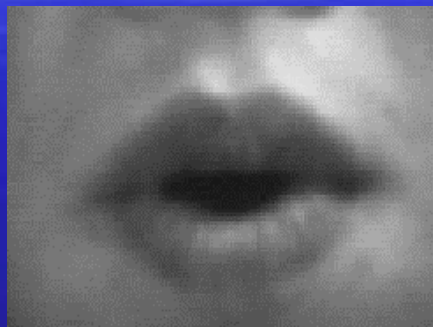
Affine (2 ZPs): 71%

(267 sequences, 4604 frames)

First 7 ZPs 90%

# Lip-Reading

- Tulips1 database
- 12 subjects - 4 words



Affine (2 ZPs): 66%

First 7 ZPs 76%

2,4,8,9,10,14,22: 79%

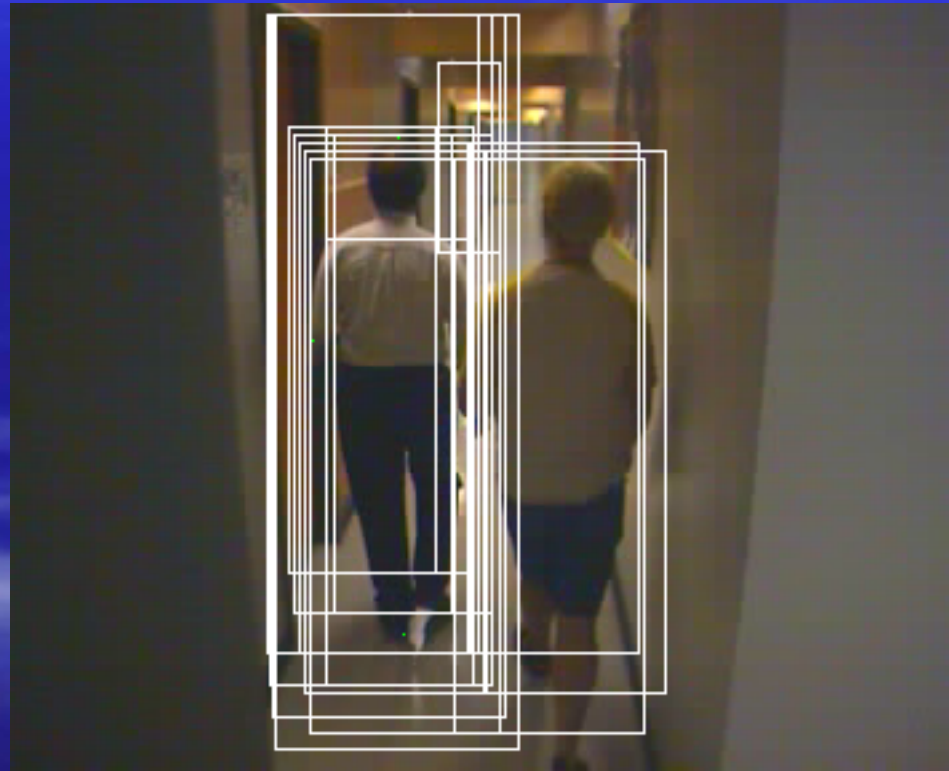
(96 sequences 835 frames)

# Multiple Camera Area Surveillance Techniques

- To distinguish "normal" people, objects, and activities from anomalous ones, and alert a security agent in the case of anomalous conditions.
- Our current projects are:
  - Fast people detection in corridor images, to be used as input for activity recognition.
  - Similarity filter development, for knowing when a given scene has been seen before.
  - Context-based object and scene recognition algorithms.

# People Detection Algorithm

- uses JPEG encoded images from a network camera. A *large* set of image features is computed, *without* the need for complete JPEG decompression.
- A support-vector machine (SVM) is used to classify image regions into people or non-people regions.
- A related project is developing an FPGA hardware implementation.





# Statistical Translation for Object Recognition

- statistical model for learning the probability that a word is associated with an object in a scene.
- learn these relationships without access to the correct associations between objects and words.



- a Bayesian scheme for automatic weighting of features (e.g., colour, texture, position) improves accuracy by preventing overfitting on irrelevant features.

# Contextual Translation

- Poor assumption: all the objects are independent in a scene.
- Our more expressive model takes context into account.
- Use loopy belief propagation to learn the model parameters. –
- On our Corel data set ([www.cs.ubc.ca/~pcarbo](http://www.cs.ubc.ca/~pcarbo)), we achieve almost 50% precision.



# Object Recognition for Robots

- Our scheme is not real-time because of the expensive segmentation step.
- BUT: our contextual translation model + a fast, crude segmentation results in equal or better precision!  
Moreover, object recognition is more precise because the segments tend to be smaller.



FIN