

**MICROMANIPULATION USING
VISION and FORCE FEEDBACK**

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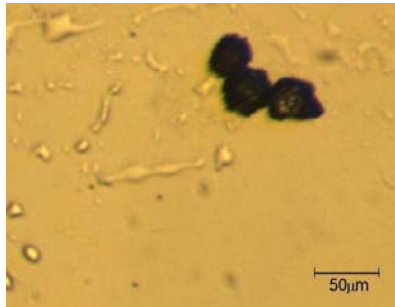
BILLFEST, Cancun, Mexico

December 7, 2008

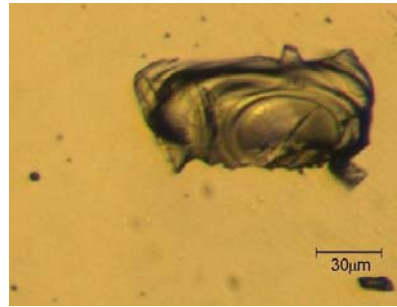
Dedicated to Bill Wolovich

Micromanipulation

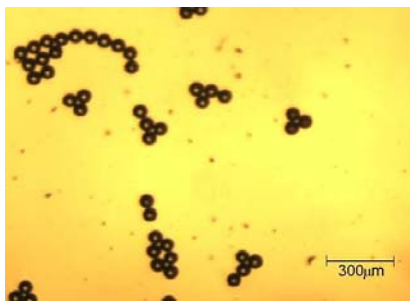
- Manipulation of objects at
 - microscale
 - mesoscale
 with micron/submicron tolerances



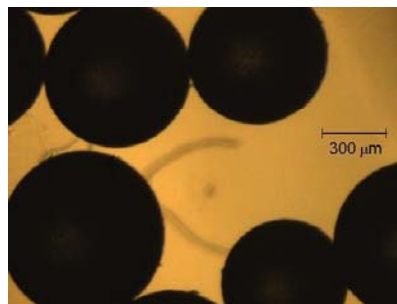
Pollens (~ 40µm)



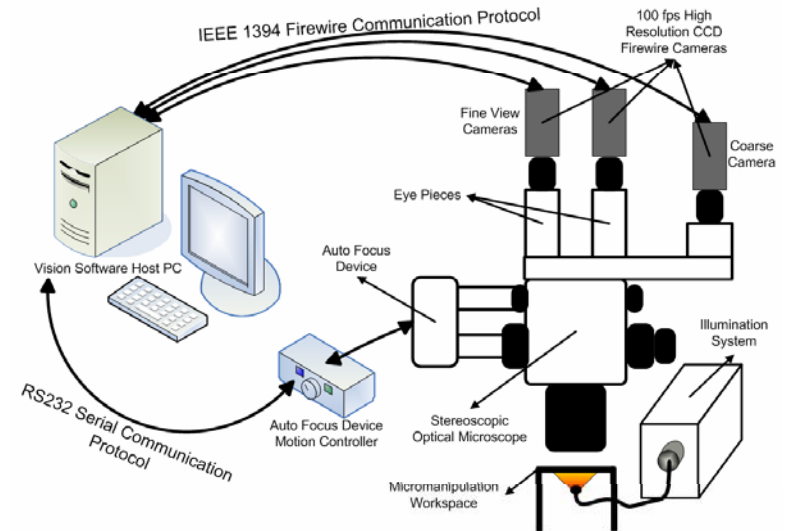
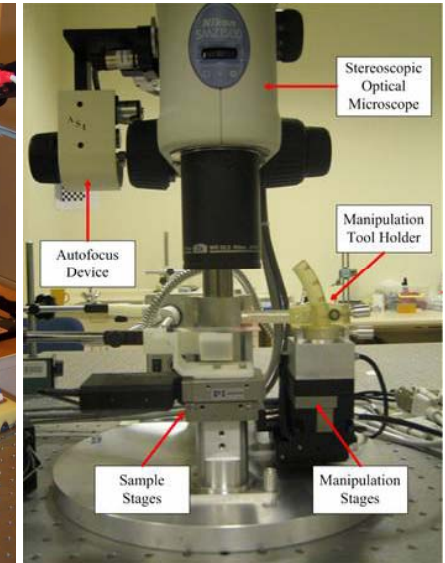
Glass Particle (~ 90µm)



Polymer Balls (~ 65 µm)



Polystyrene Balls (~ 450 µm)

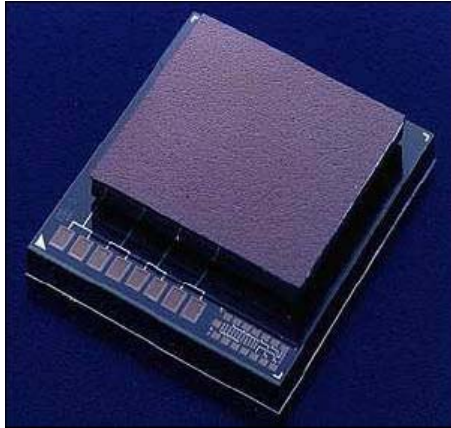


Relative accuracy of 1 - 5%

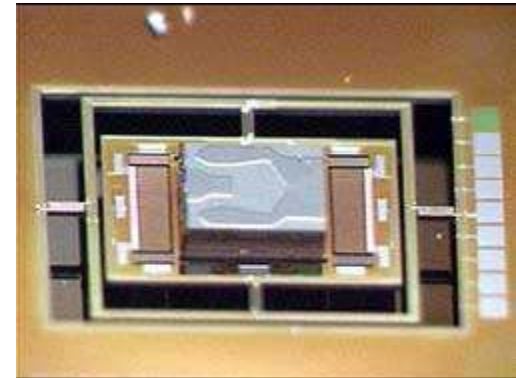
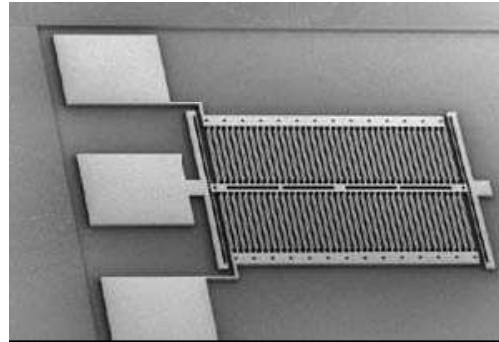


0.5 – 2.5 µm

Micro Devices



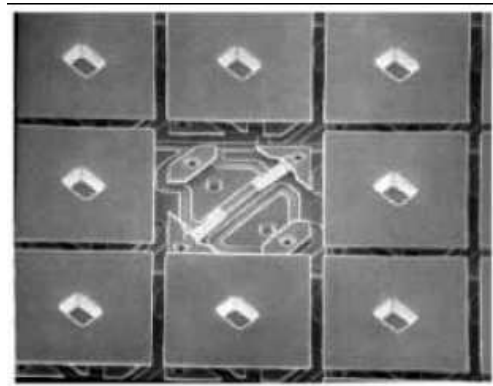
Low-g Accelerometer



Hard Drive's Read/Write Head



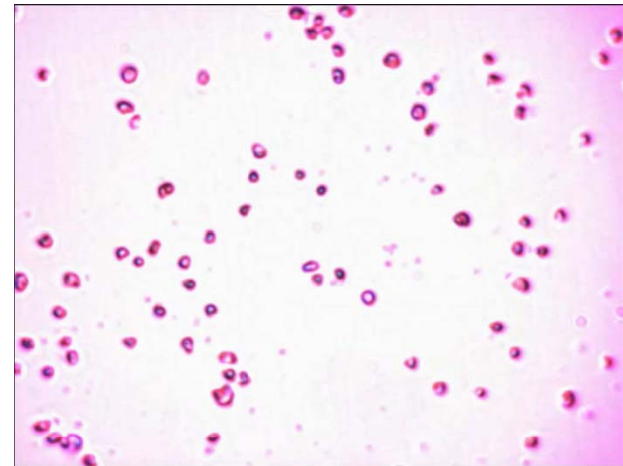
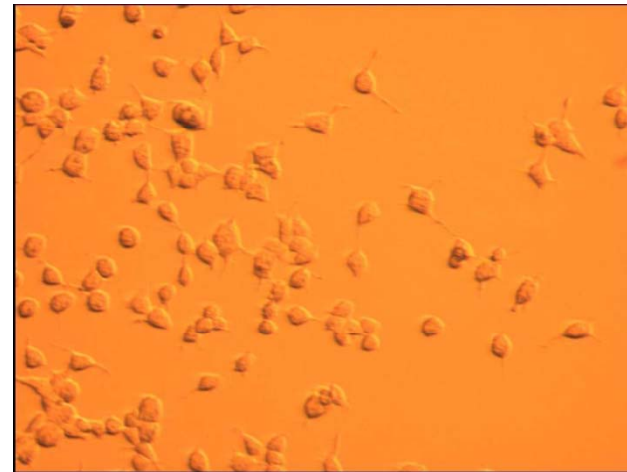
Digital Micromirror Device



Life Sciences and Medical Applications

Automated Tracking of Cell Populations

- measurements of a range of cell behaviors
 - *migration* (translocation),
 - *mitosis* (division),
 - *apoptosis* (death),
 - *lineage* (parent-daughter relations, e.g. Yeast model of ageing)
- valuable in medical research: genomics, proteomics, stem cell biology, and tissue engineering

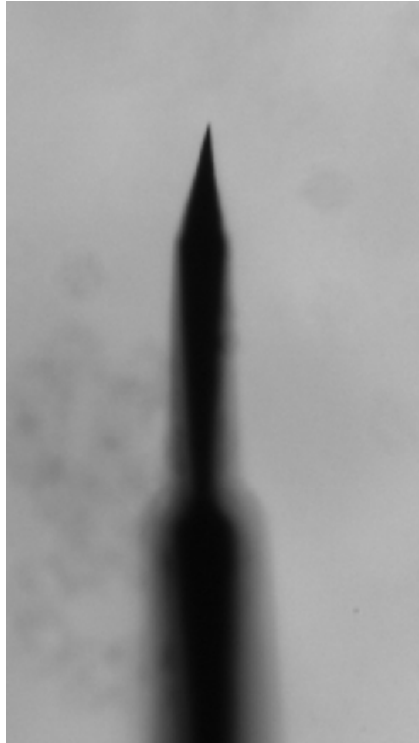


293T Human Kidney Cells and Jurkat Cancer Cells

Challenges

- Autofocusing
- Calibration of the Optical System
- Robust Feature Extraction and Real-Time Tracking
- Robust Visual Servoing Algorithms
- 3D Reconstruction
- Force Estimation Using Image Data for Teleoperations

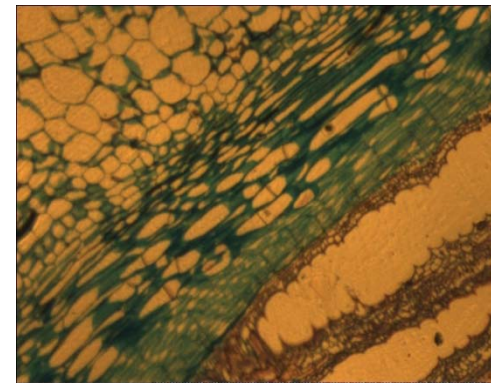
Autofocusing



Probe

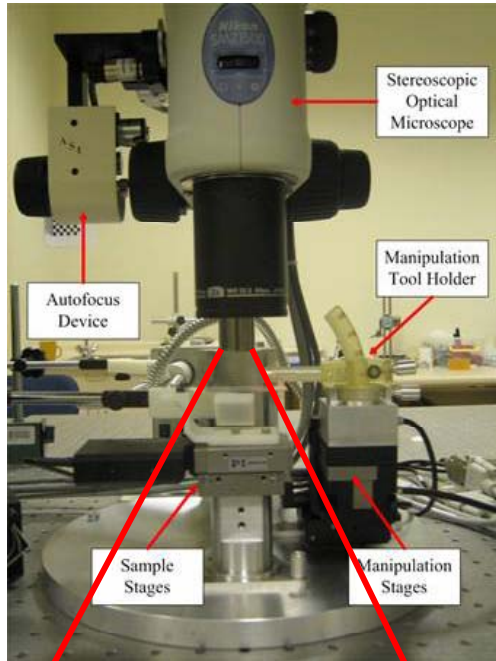


Defocused and Focused Images of Drosophila



Defocused and Focused Images of Pumpkin Cells

Optical System Calibration



Parameters

Objective focal length, Tube Length, Magnification etc.

Projection Model

$$u(1 + k_1 r^2) = (T_{op} + f) \frac{r_{11}X_g + r_{12}Y_g + r_{13}Z_g + T_x}{r_{31}X_g + r_{32}Y_g + r_{33}Z_g + T_z}$$

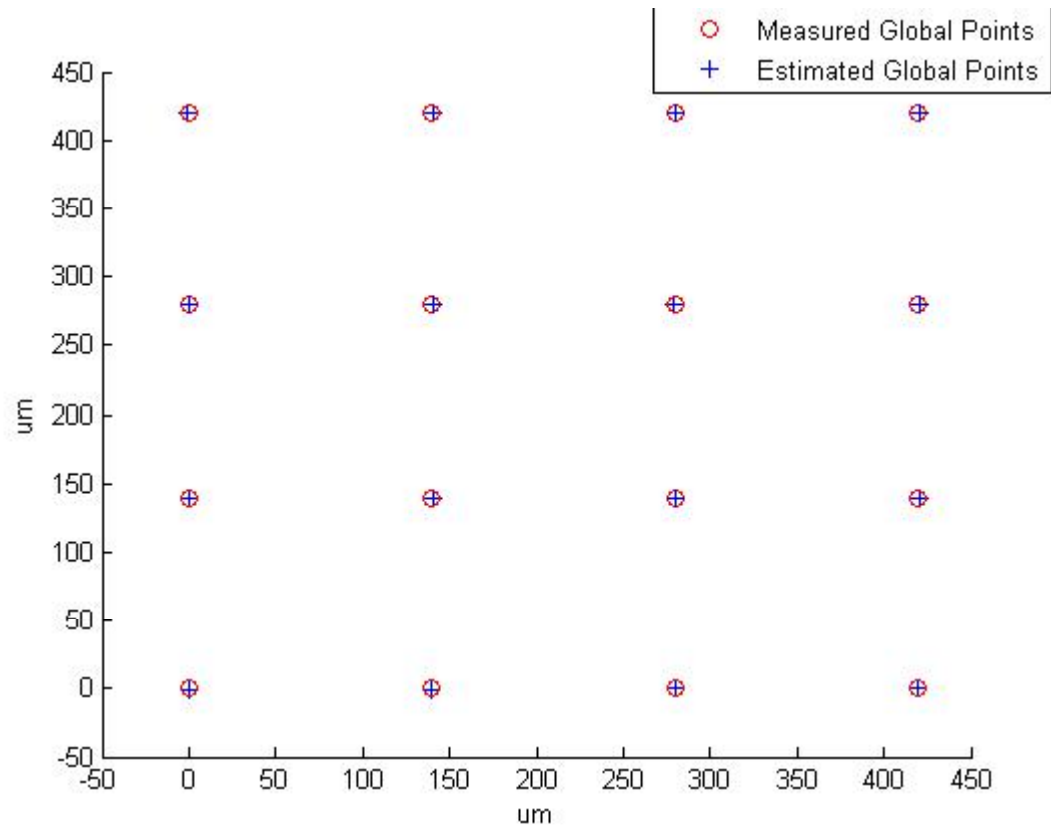
$$v(1 + k_1 r^2) = (T_{op} + f) \frac{r_{21}X_g + r_{22}Y_g + r_{23}Z_g + T_y}{r_{31}X_g + r_{32}Y_g + r_{33}Z_g + T_z}$$

$$T_z \approx f + d$$

**Nonlinear
Optimization**

**ESTIMATED
PARAMETERS**

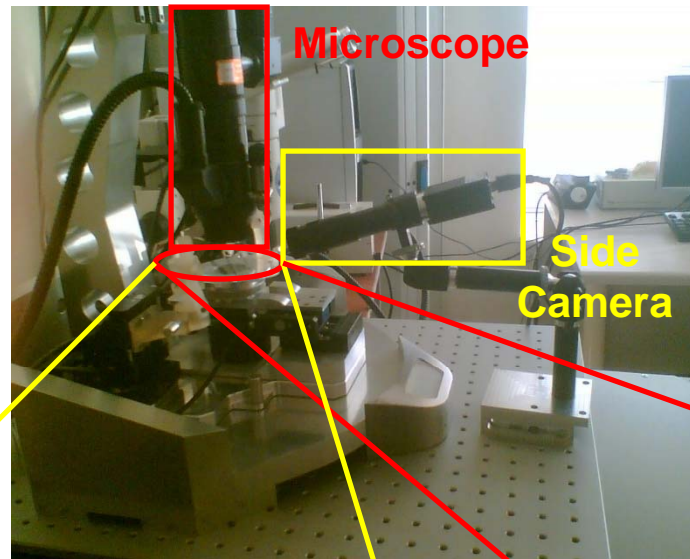
Reconstruction/Reprojection Errors



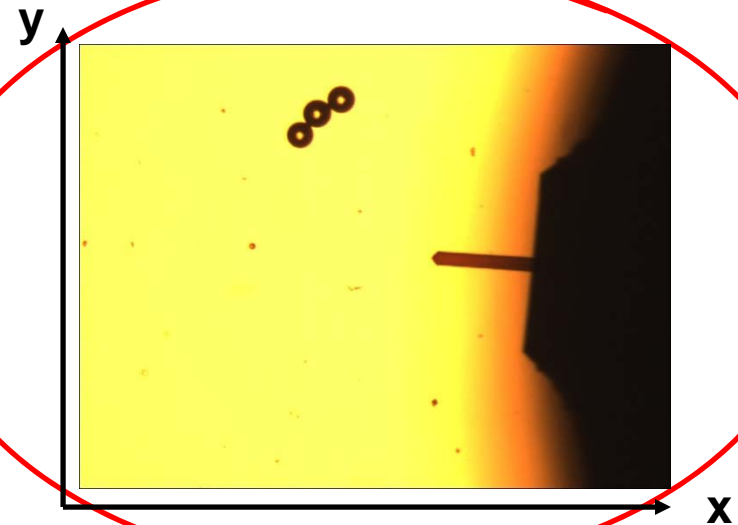
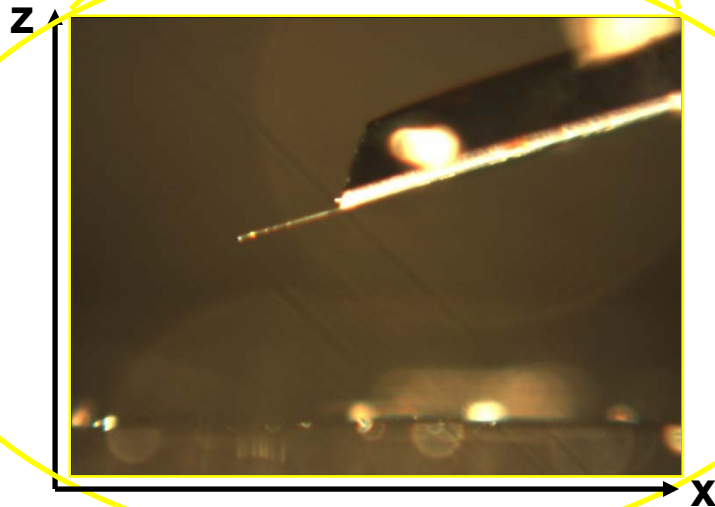
Reconstruction Errors

Magnification	1.6	6.4
Mean Error(μm)	0.220	0.063
Max Error (μm)	1.720	0.584

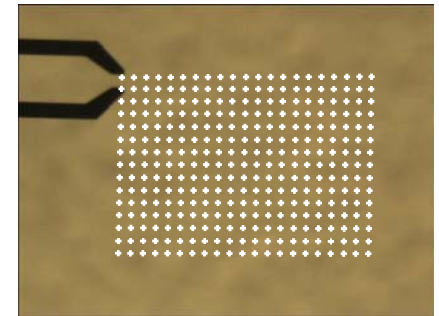
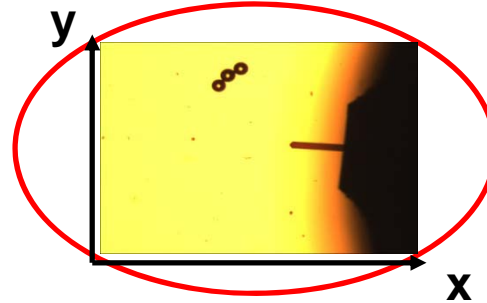
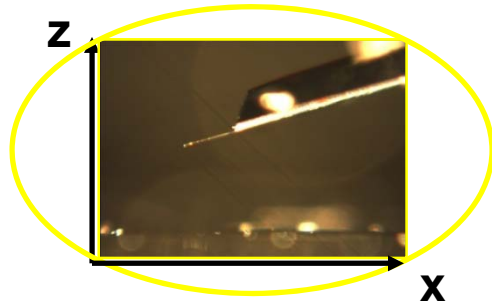
3D View by Side Camera



- x-z (depth) information from the side camera
- x-y (sample stage) information from the optical microscope



A Novel Online Calibration Technique

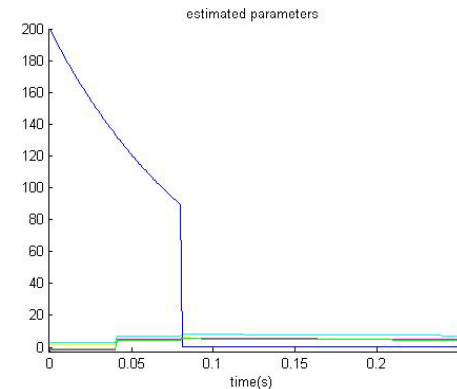
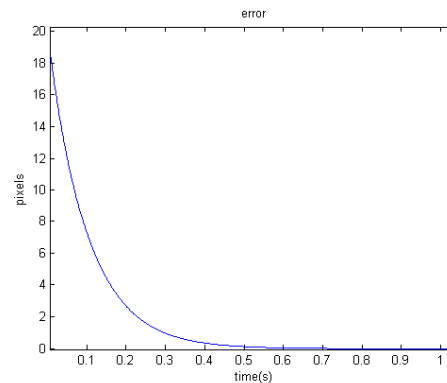


Virtual Pattern

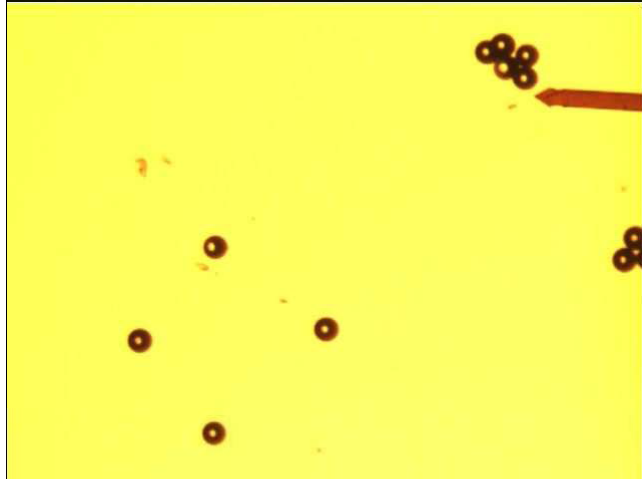
$$\begin{pmatrix} x_{im} \\ z_{im} \end{pmatrix} = \begin{pmatrix} p_{11} & p_{12} \\ p_{21} & p_{22} \end{pmatrix} \begin{pmatrix} x_w \\ z_w \end{pmatrix} + \begin{pmatrix} q_1 \\ q_2 \end{pmatrix} \quad \rightarrow \quad \begin{pmatrix} x_{im} \\ y_{im} \end{pmatrix} = \begin{pmatrix} k_{11} & k_{12} \\ k_{21} & k_{22} \end{pmatrix} \begin{pmatrix} x_w \\ y_w \end{pmatrix} + \begin{pmatrix} m_1 \\ m_2 \end{pmatrix}$$

$$\begin{pmatrix} x_{im} \\ y_{im} \\ z_{im} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{pmatrix} x_w \\ y_w \\ z_w \end{pmatrix} + \begin{pmatrix} t_x \\ t_y \\ t_z \end{pmatrix}$$

- adapted to different magnifications
- robust to noise
- high accuracy
- high convergence rate



Robust Feature Extraction



Top View of the Sample Stage

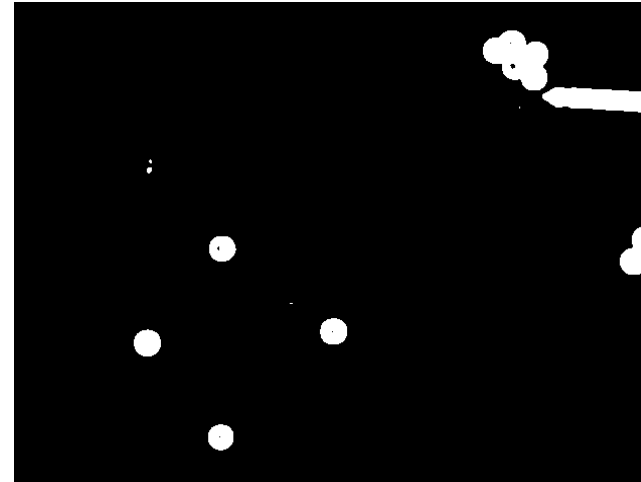
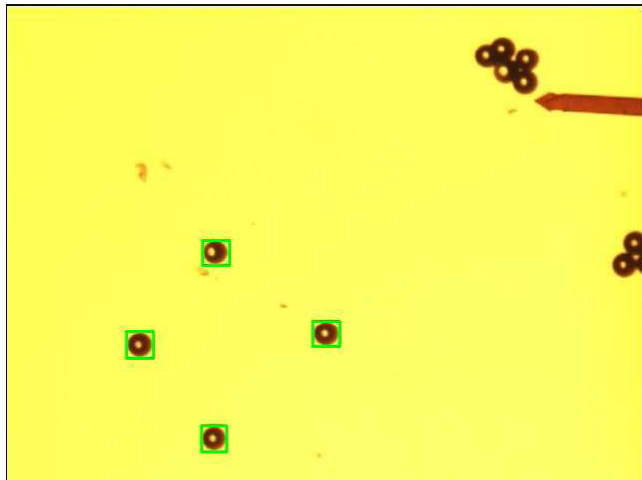
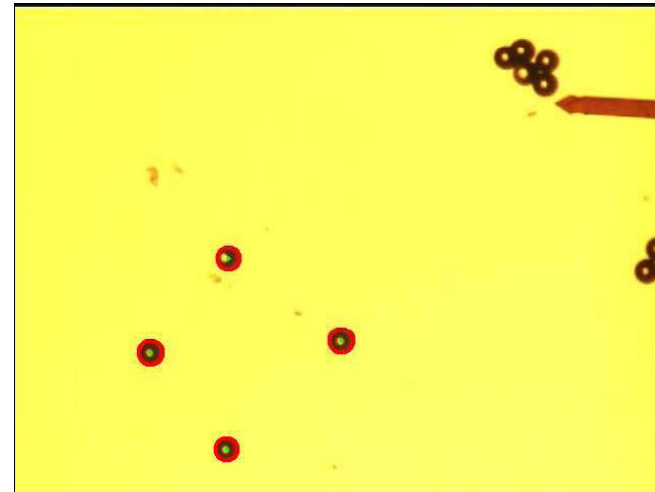


Image After Thresholding



Segmentation



Extracted Circles

Robust Feature Extraction

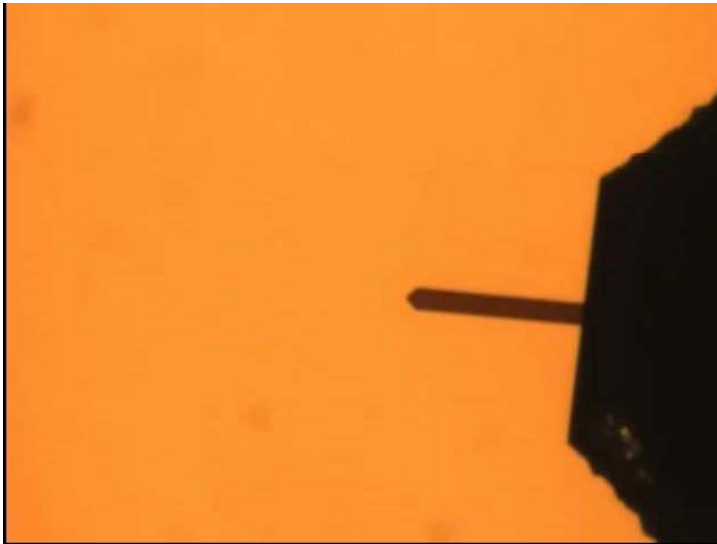
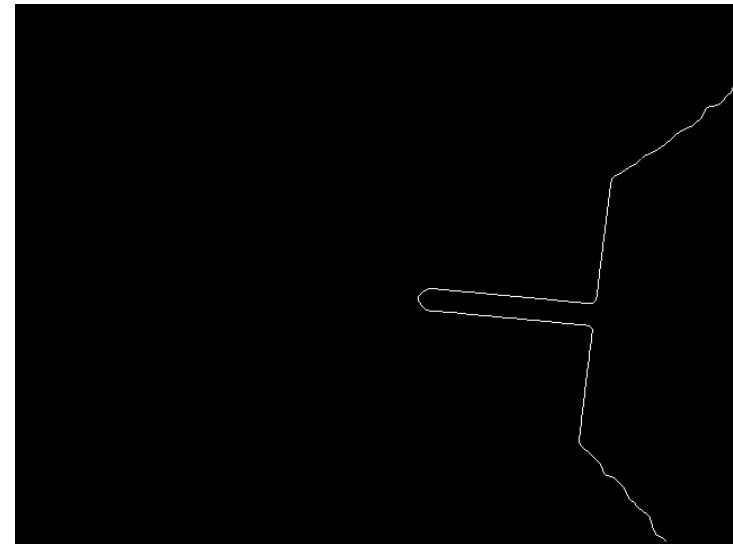


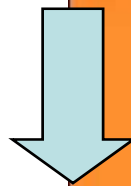
Image of the Probe



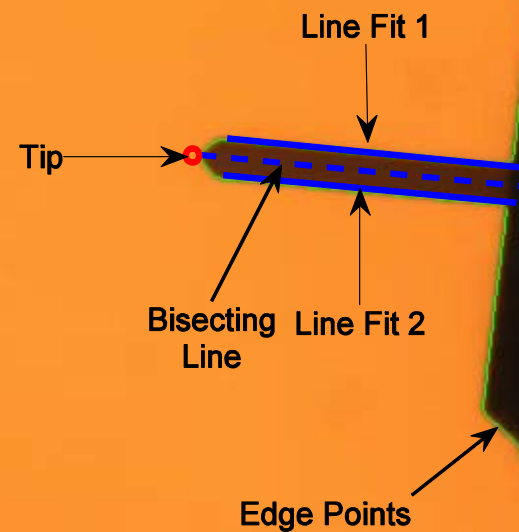
Edge Pixels of the Probe

Probe Tip Detection & Real-Time Tracking

Tracking at video rate (30 fps)



**Kalman Filtering
&
Template Matching**



Design of Visual Controllers

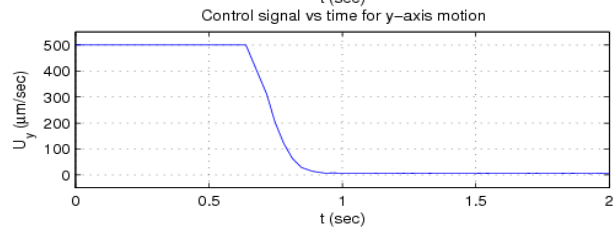
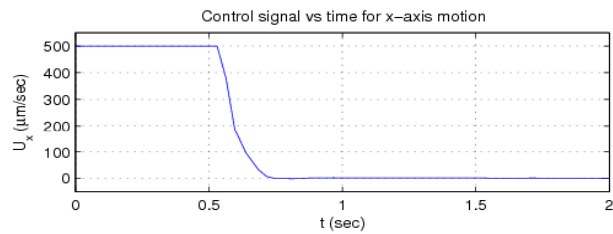
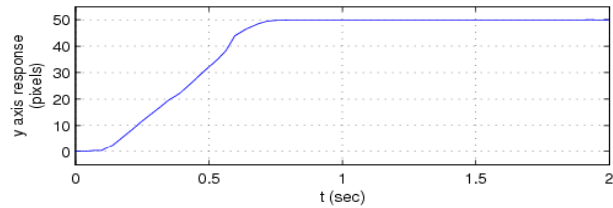
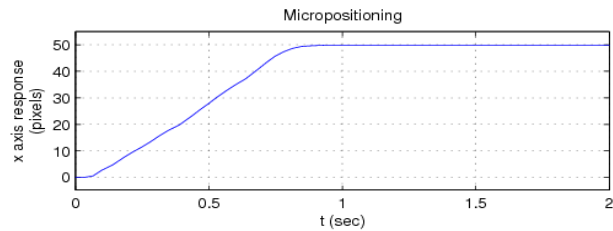
Having optical system parameters from calibration and real time image feature measurements from tracking, how do we implement vision based control algorithms to move the gripper?

Optimal Control

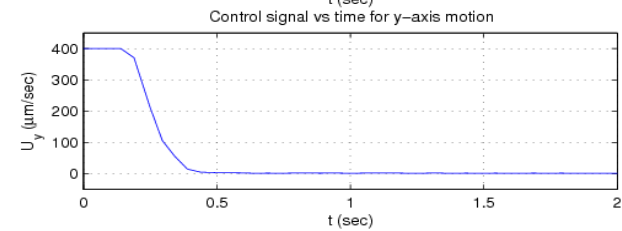
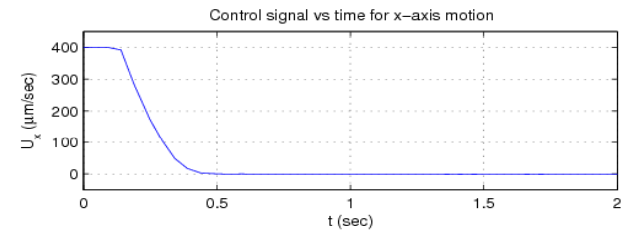
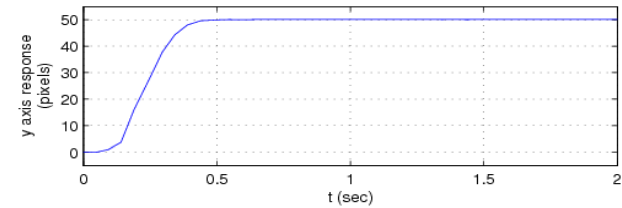
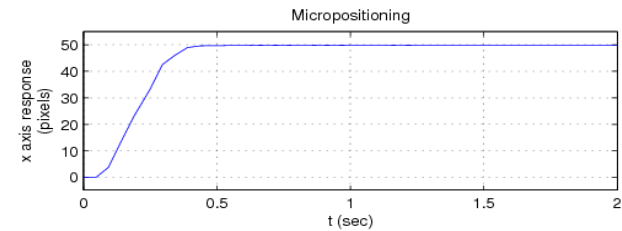
$$E(k+1) = (f(k+1) - f^*(k+1))^T Q (f(k+1) - f^*(k+1)) + u^T(k) L u(k)$$

$$u(k) = -(TJ^T(k) Q T J(k) + L)^{-1} T J^T(k) Q (f(k) - f^*(k+1))$$

Micropositioning Results

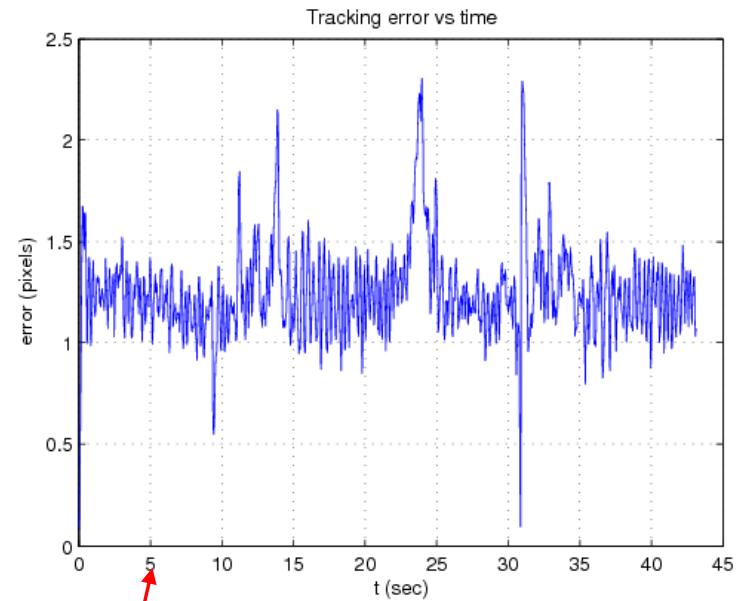
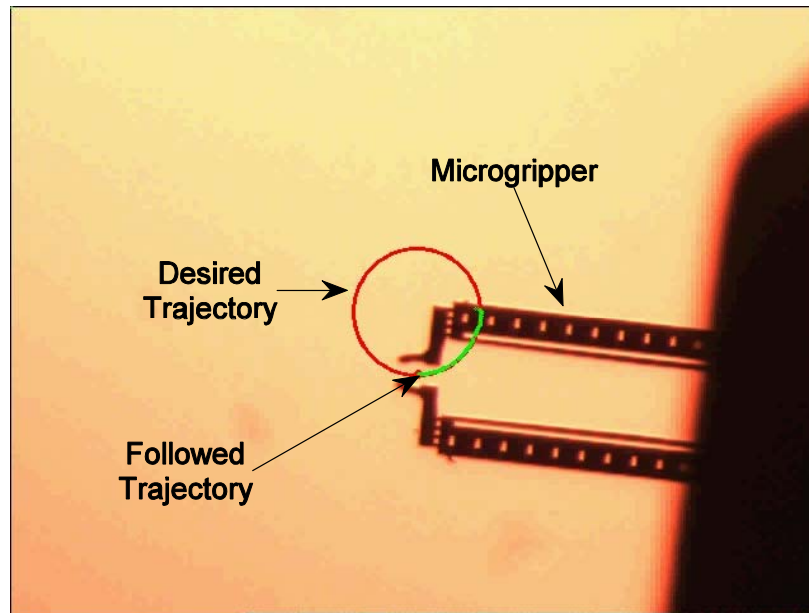


	Step (pixels)	t_s (sec)	Acc. (μm)	Prec. (μm)
1x	50	0.8	9.86	2.71



	Step (pixels)	t_s (sec)	Acc. (μm)	Prec. (μm)
4x	50	0.45	1.35	0.57

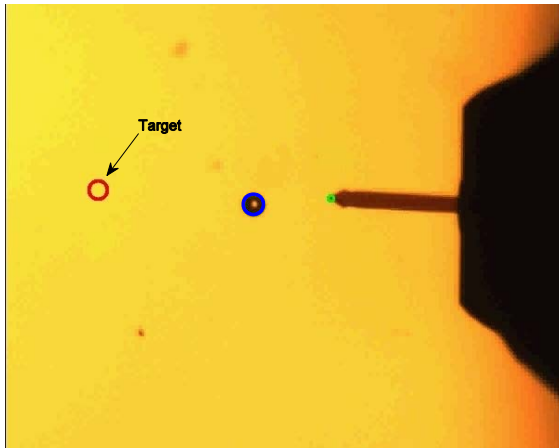
Trajectory Tracking Results



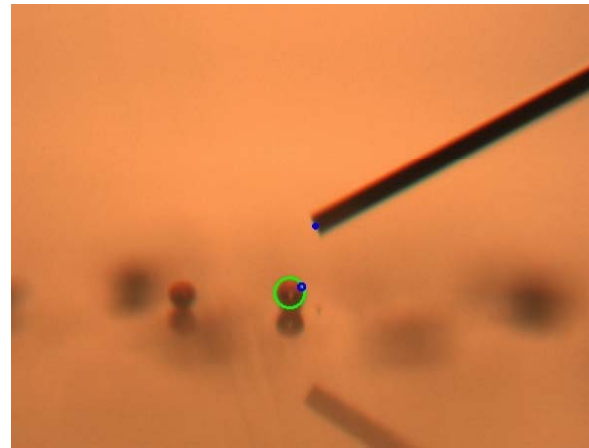
RESULTS OF TRAJECTORY TRACKING FOR CALIBRATED VISUAL SERVOING

	Square		Circular		Sinusoidal	
	Acc. (μm)	Prec. (μm)	Acc. (μm)	Prec. (μm)	Acc. (μm)	Prec. (μm)
1x	5.93	2.28	7.72	1.40	4.79	2.37
4x	1.47	1.19	1.57	0.95	1.12	1.31

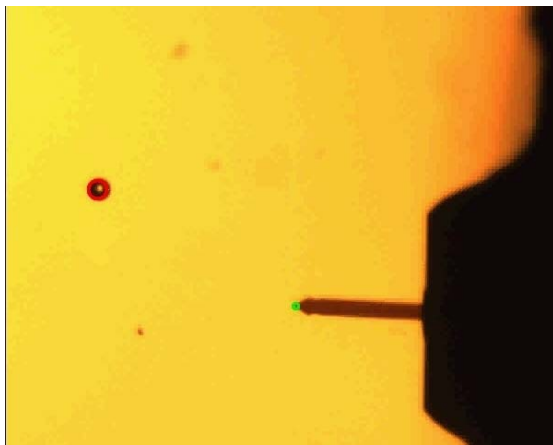
Automatic Manipulation by Pushing



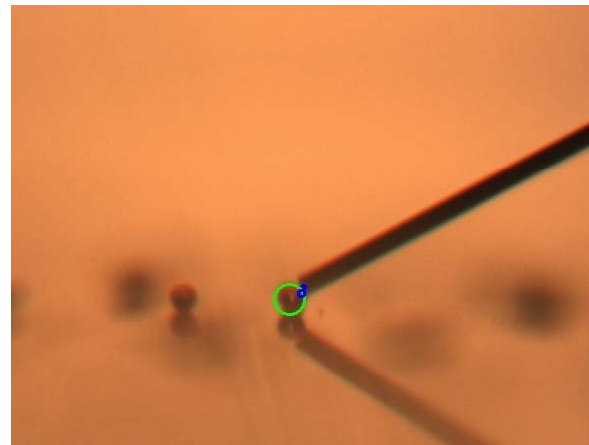
TOP VIEW



SIDE VIEW



TOP VIEW



SIDE VIEW

Mode 0. Explore

- Target Pose
- Probe Tip Detection
- Object (Ball) Detection
- Which Ball to Manipulate
- Tracking using Kalman Filters

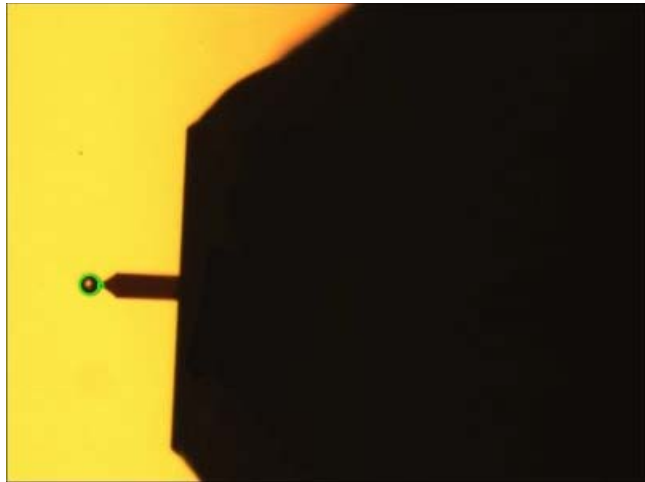
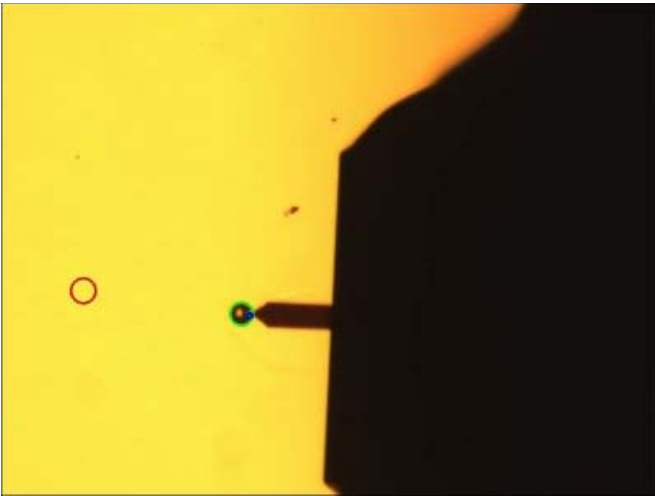
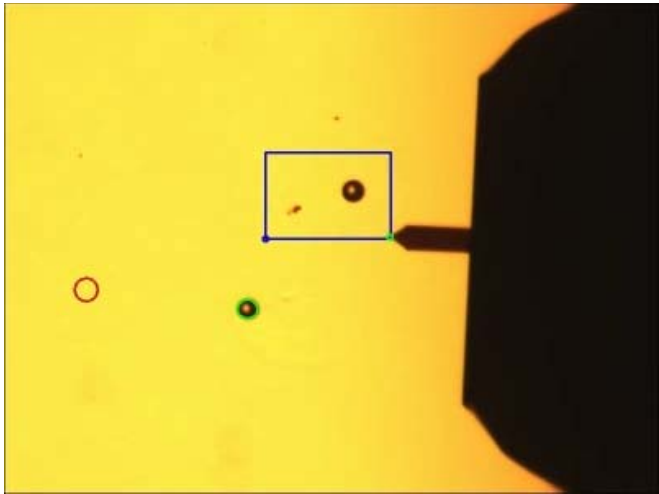
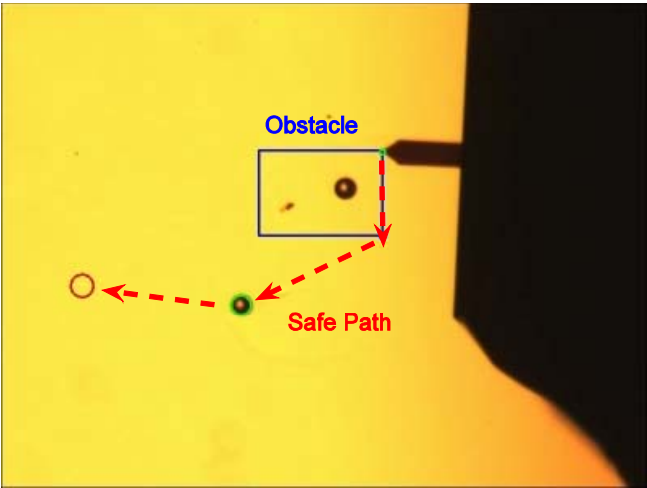
Mode 1. Obstacle Avoidance

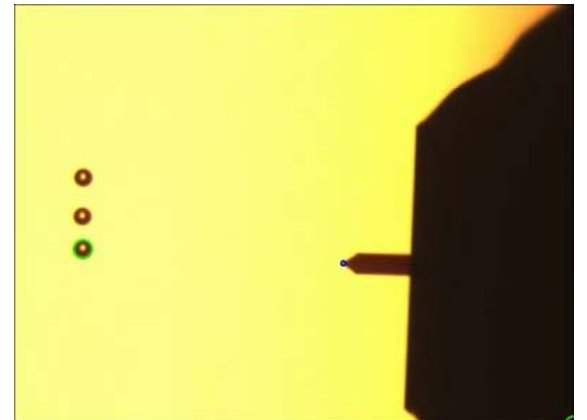
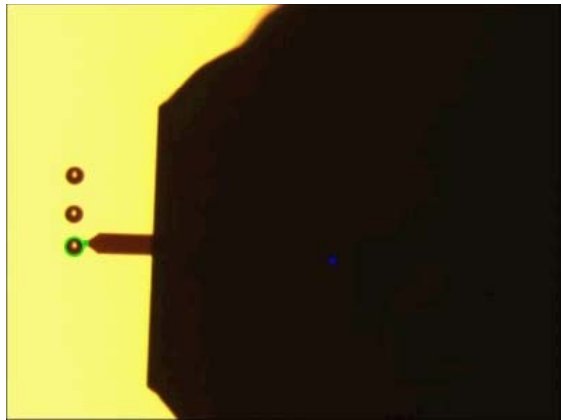
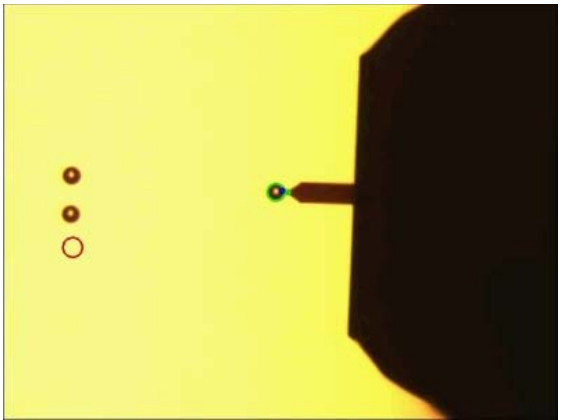
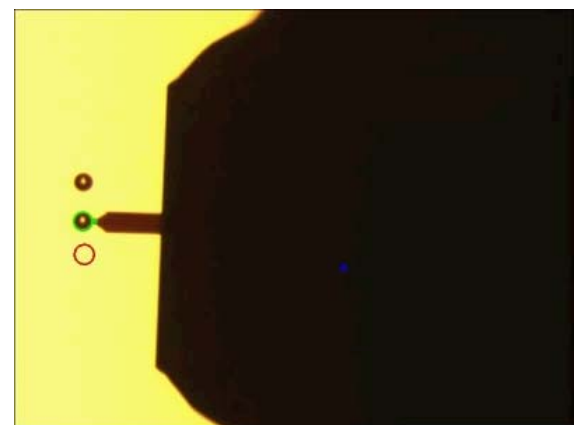
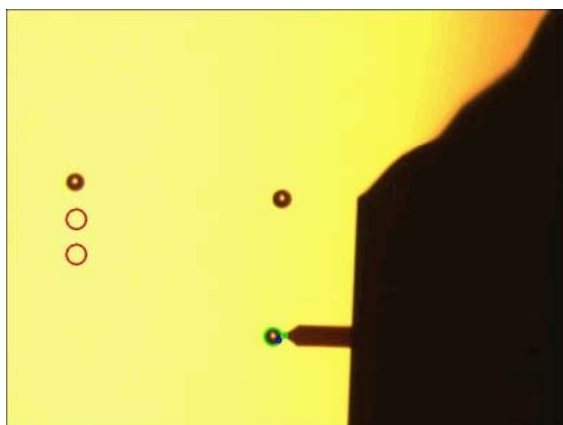
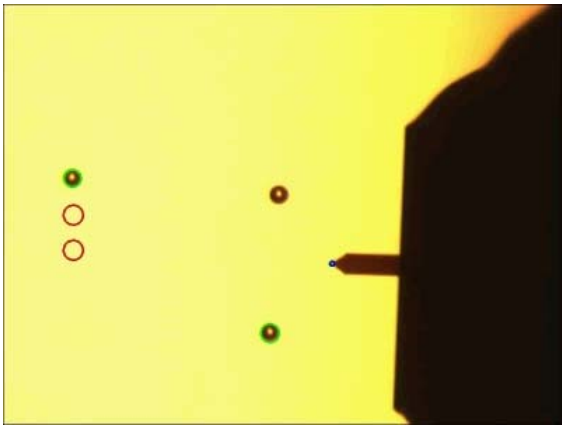
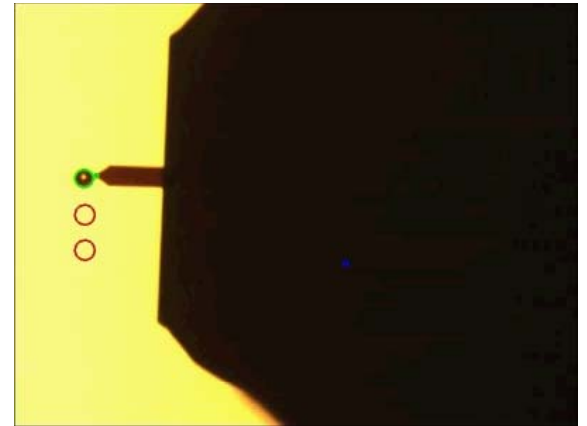
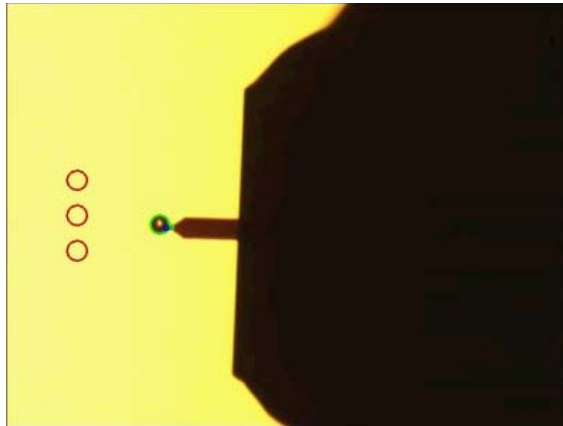
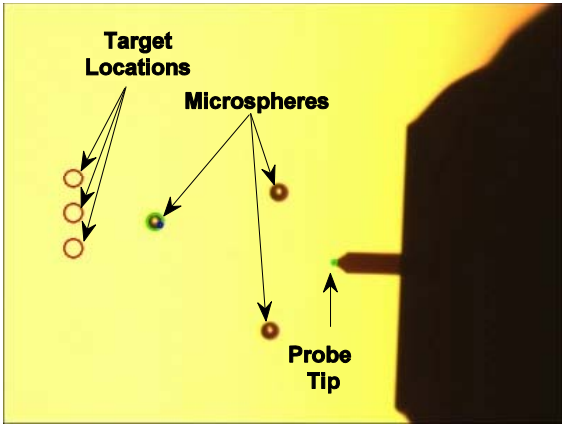
Mode 2. Automated Pushing

- Visual Servoing

Mode 3. Home

Obstacle Avoidance





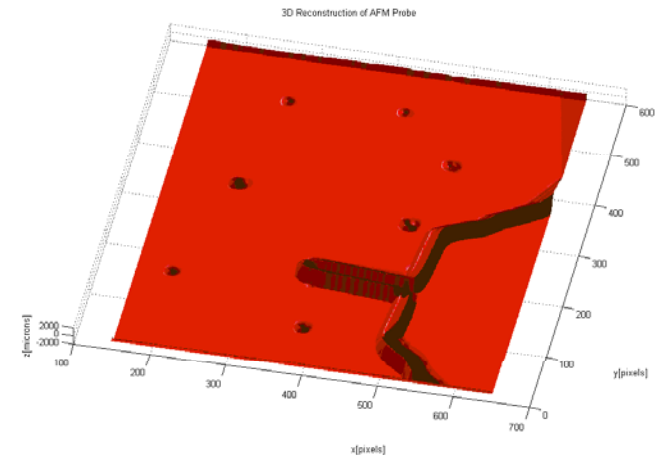
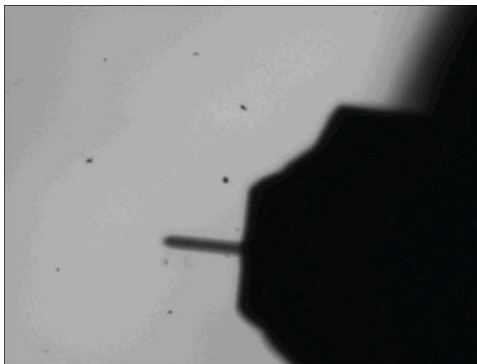
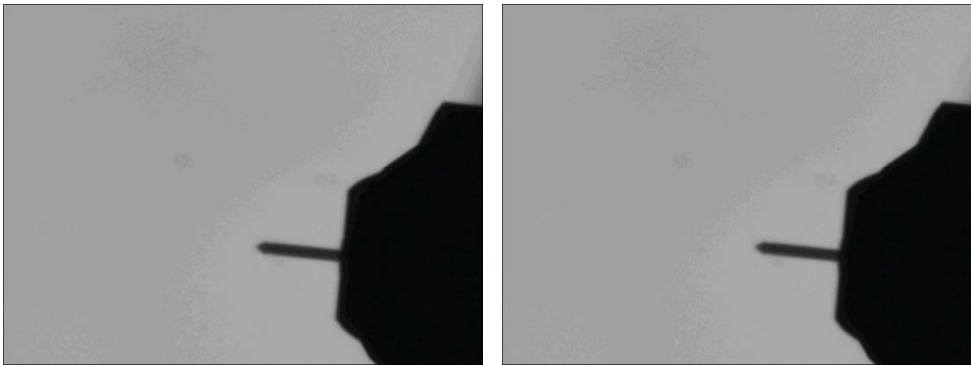
Active 3D Reconstruction

- **Shape from Focus**

2D Images



3D Model

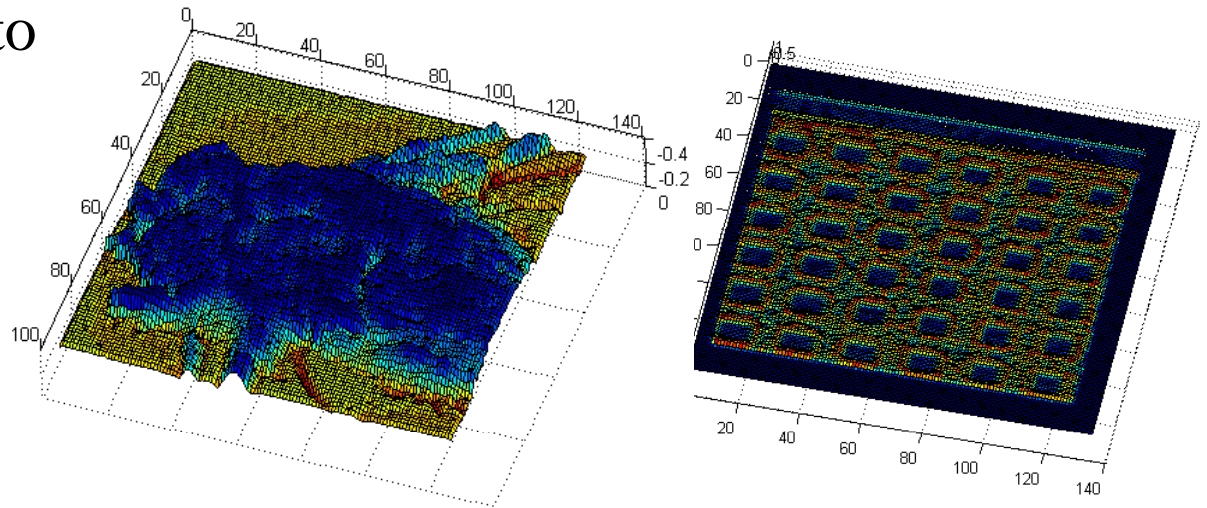
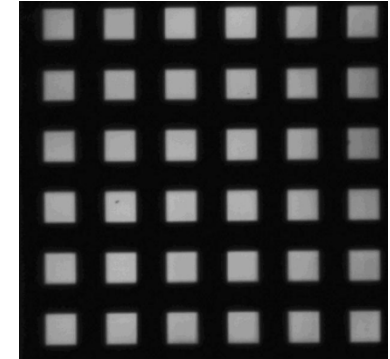


- 2D Images at different focus levels captured to create 3D model of AFM Probe

Passive 3D Reconstruction

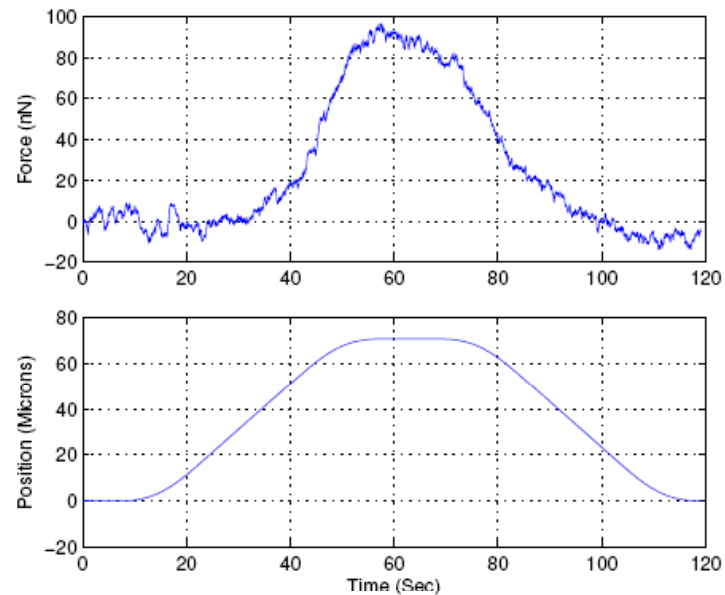
Shape From Defocus

- estimating the 3D surface of a scene from a set of two or more images of the scene
- exploiting defocus to recover the depth



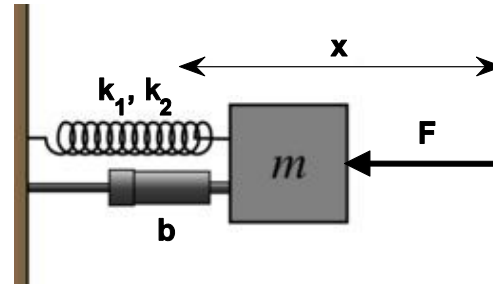
Force Sensing

- In order to achieve high dexterity in manipulation, control of the interaction forces is required.
- In micro-manipulation, control of interaction forces necessitates force sensing in mili-Newton range with nano-Newton resolution



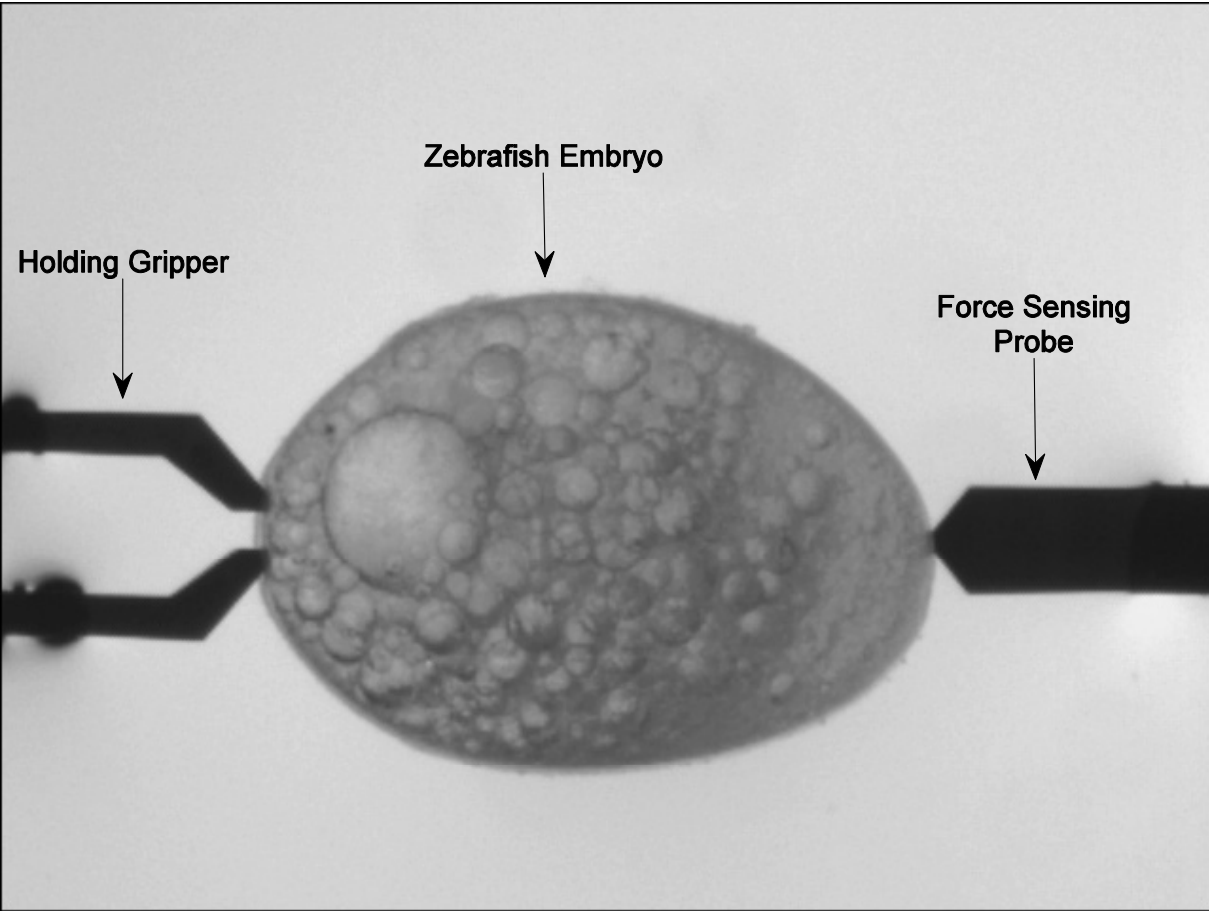
Pulling in-out for smooth step position references

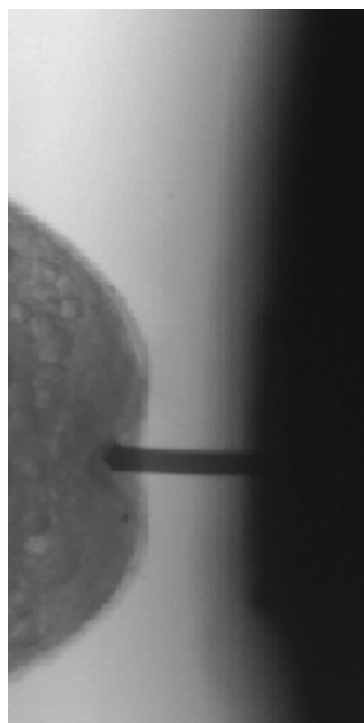
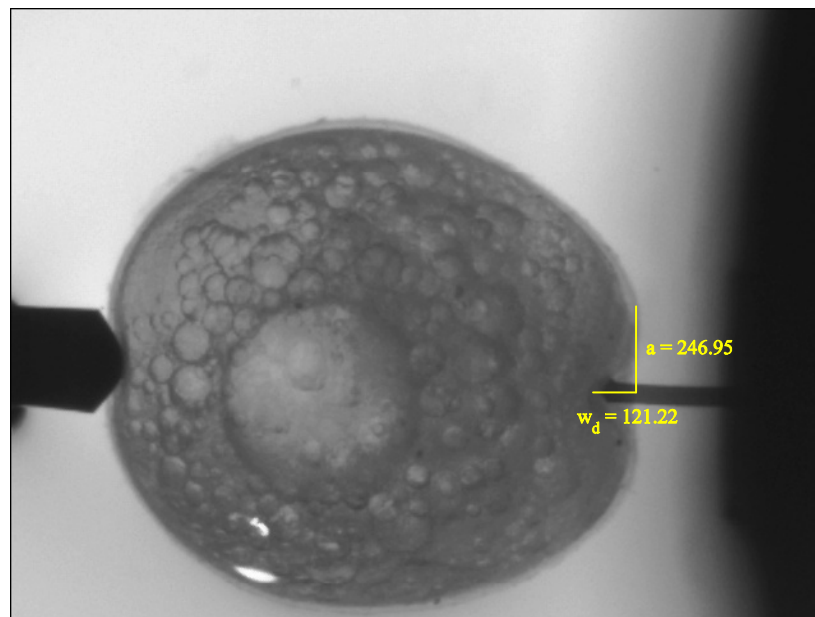
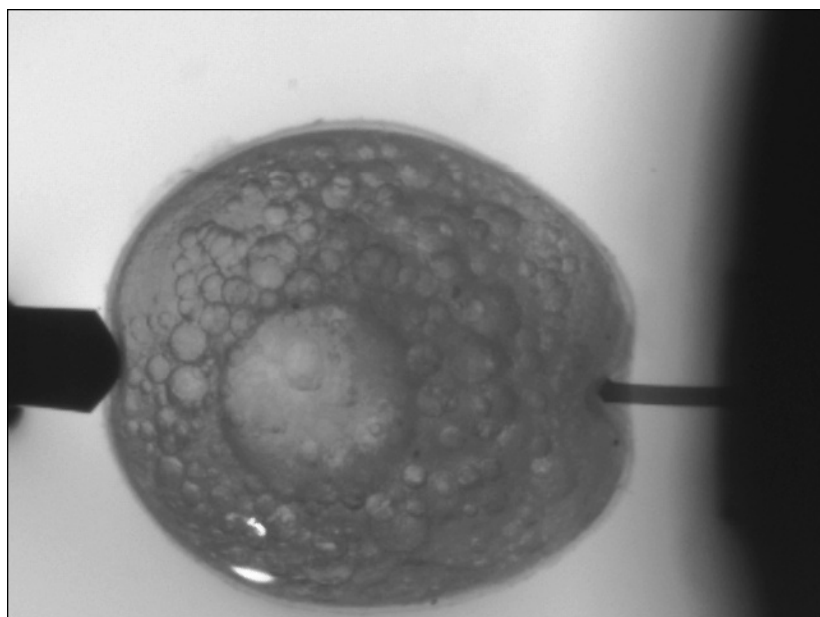
Force Estimation and Mechanical Characterization



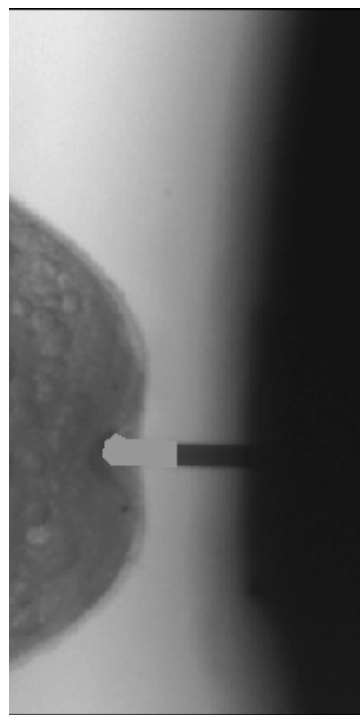
$$F = m\ddot{x} + b\dot{x} + k_1x + k_2x^3 \Rightarrow F_i = \underbrace{(\ddot{x}_i \quad \dot{x}_i \quad x_i \quad x_i^3)}_{\varphi_i^T} \underbrace{\begin{pmatrix} m \\ b \\ k_1 \\ k_2 \end{pmatrix}}_{\theta}$$

$$\hat{\theta} = \arg \min (\|F - \Phi \theta\|^2 + \delta \|\theta\|^2) = (\Phi^T \Phi + \delta I)^{-1} \Phi^T F$$





(a)



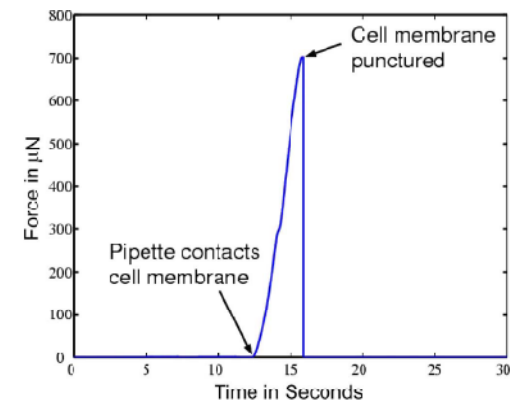
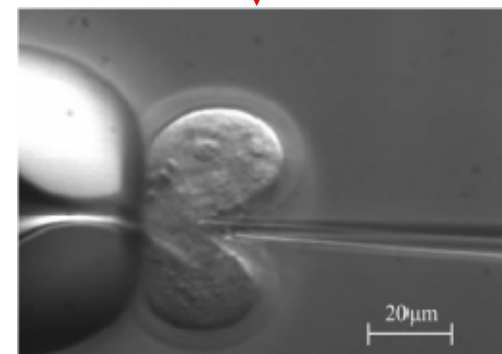
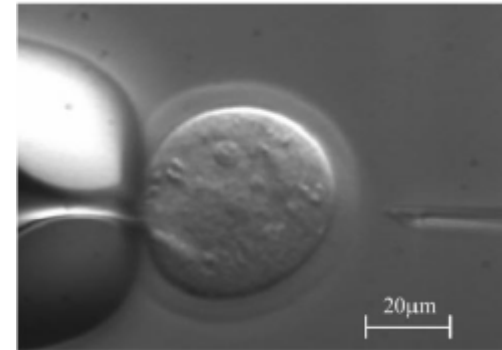
(b)



(c)

Applications

- Automated injection
 - intracytoplasmic sperm injection (ICSI), DNA injection, gene therapy
 - manual injection
 - low success rate and poor reproducibility
 - efficiency of injection can be improved by using visual feedback



Summary

- Robust and repeatable algorithms for optical system calibration
- Robust feature extraction and real-time tracking
- Path Planning, Visual servoing, Automatic Manipulation by Pulling
- 3D Reconstruction from Focus and Defocus
- Force Estimation and Mechanical Characterization

THANK YOU

&

BEST WISHES, BILL!