

# A 3D Laser Targeting System

## *Master Thesis*

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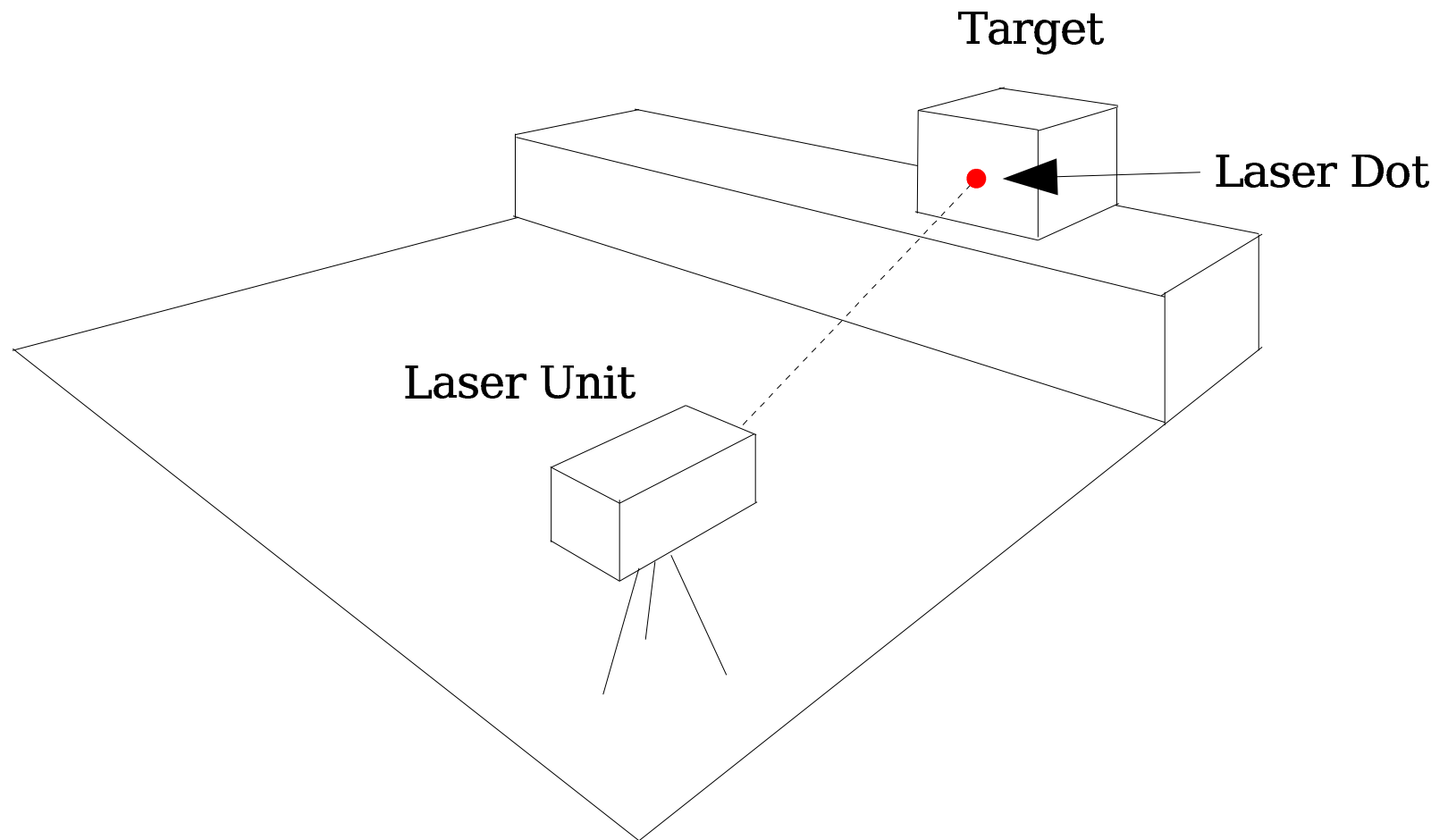
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# Overall Goal

Aim laser at a point in the environment using observations from a stereo camera



# Contribution

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Two methods of calibrating camera observations to laser controls

- > Theoretical justification
- > Implementation
- > Experimental analysis
  - > Overall system accuracy
    - > depth of target
    - > position of target
  - > Accuracy of different calibration methods

# Outline

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- > Related Work & Motivations
- > Background
- > Calibration Algorithms
  - > Theory
  - > Experimental Results
- > Improving Calibration
  - > Automatic Detection
  - > Point Selection
  - > More Experimental Results
- > Concluding Remarks

# Related Work: Visual Servoing

Iterative method to control robotic manipulator using camera observations

- > use error gradient to pick action that minimizes difference between target and observed position

Advantages

- > Analytic relationship not required
- > Can dynamically adapt to observed errors

Problems

- > convergent method, never \*exactly\* on
- > Requires consistent knowledge of laser dot position
  - > Laser dot detection not robust

# Current Method

Solve for transformation between laser and one plane in space.

- > requires only one camera
- > allows direct aiming of the laser
- > calibration possible with 4 corresponding points between laser & image

## Problems

- > Doesn't model full 3D geometry
- > targeting outside depth plane is inaccurate
- > must recalibrate to change it

# New Approach

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- > Stereo camera measures depth
- > Exact transformation allows direct aiming of laser

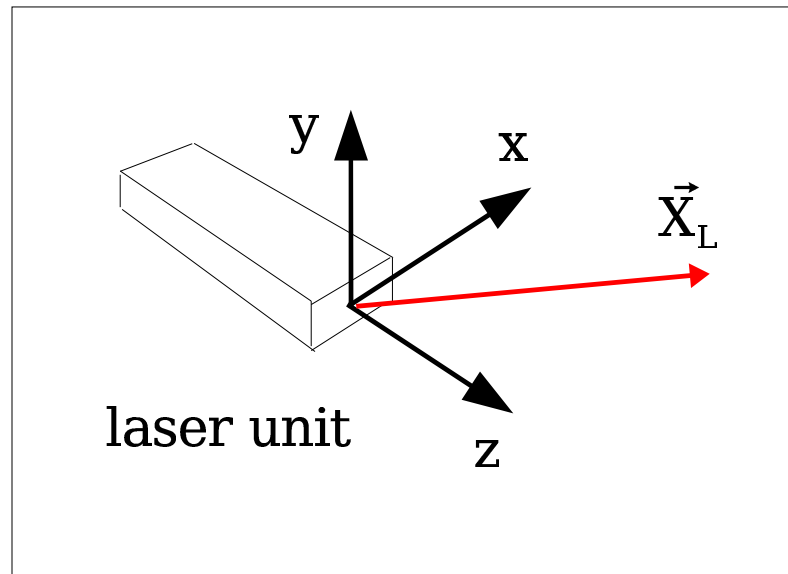
Two calibration methods

- > Direct (3D -> laser)
- > Epipolar (2D x 2 -> laser)

# Background: Laser

We model the laser as a black box:

- > Two inputs  $(u, v)$  control direction  $\vec{X}_L$  of the laser.
- > Fixed origin
- > Direction  $\vec{X}_L$  linear with  $x_L = (u, v)$ .





# Background: Laser

Direction  $\vec{X}_L$  linear with  $x_L$ .

$$w\mathbf{x}_L = \mathbf{A}_L\vec{X}_L$$

Where

- >  $w$  is a scale factor
- >  $\mathbf{A}$  is a  $3 \times 3$  laser projection matrix.

$x_L$  projects on a line of 3D points.

# Background: Depth Sensor

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Requirements:

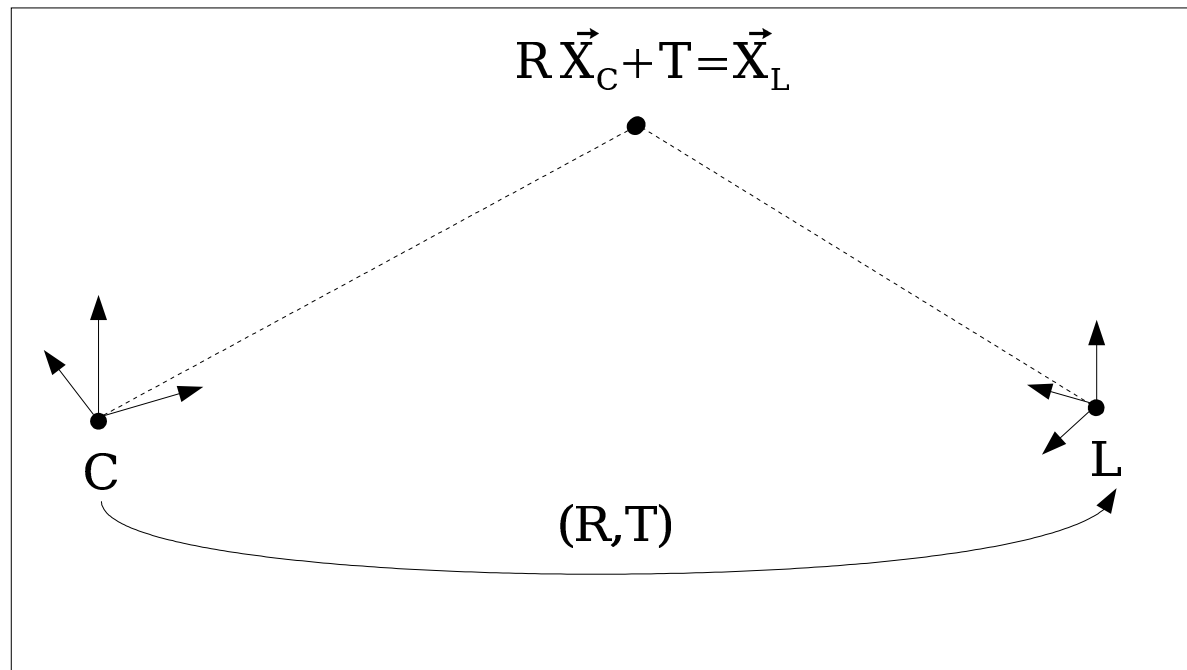
- > can sense laser dot
- > can report position relative to some 3D coordinate system

Tyzx Stereo Camera

- > dot is visible in dim lighting
- > report location relative to left camera center

# Coordinate system relationship

Camera and laser 3D coordinate systems are related by a rotation and translation.



- >  $R$  is a  $3 \times 3$  rotation matrix
- >  $T$  is a  $3 \times 1$  translation vector.

# Coordinate system relationship

Laser control and Camera coordinate related by

$$\mathbf{H}\mathbf{X}_C = \mathbf{x}_L$$

- > Where  $\mathbf{H} = \mathbf{A}_L[\mathbf{R}|\mathbf{T}]$ 
  - >  $\mathbf{A}_L$  is laser projection matrix
  - >  $[\mathbf{R}|\mathbf{T}]$  is  $3 \times 4$  augmented matrix of rotation and translation
- > Calibrate laser by solving for  $\mathbf{H}$
- > Control laser by multiplying  $\mathbf{H}$  and the desired target  $\mathbf{X}_C$

# Direct Calibration

- > Observe correspondance between laser, 3d coordinate of laser in camera image
- > Each correspondance provides three linear constraints on  $\mathbf{H}$ :

$$Xh_1 + Yh_2 + Zh_3 + h_4 = wu$$

$$Xh_5 + Yh_6 + Zh_7 + h_8 = wv$$

$$Xh_9 + Yh_{10} + Zh_{11} + h_{12} = w$$

- > Where  $h_i$  are the components of the matrix  $\mathbf{H}$

# Direct Calibration

- > Eliminating  $w$  maintains two linear constraints

$$Xh_1 + Yh_2 + Zh_3 + h_4 = u(Xh_9 + Yh_{10} + Zh_{11} + h_{12})$$

$$Xh_5 + Yh_6 + Zh_7 + h_8 = v(Xh_9 + Yh_{10} + Zh_{11} + h_{12})$$

- > Need 6 or more correspondences to solve for 12 degrees of freedom of  $\mathbf{H}$  using linear least squares

# Deriving Laser Controls with H

Given H:

- > Define 3D coordinate  $\vec{X}_C$  of target using Tyzx Stereo camera

- > Product  $H\vec{X}_C = \begin{bmatrix} wv \\ wu \\ w \end{bmatrix}$ .

- > Solve for laser controls  $(u, v)$  by dividing out  $w$ .

Results to come ...

# Epipolar Calibration

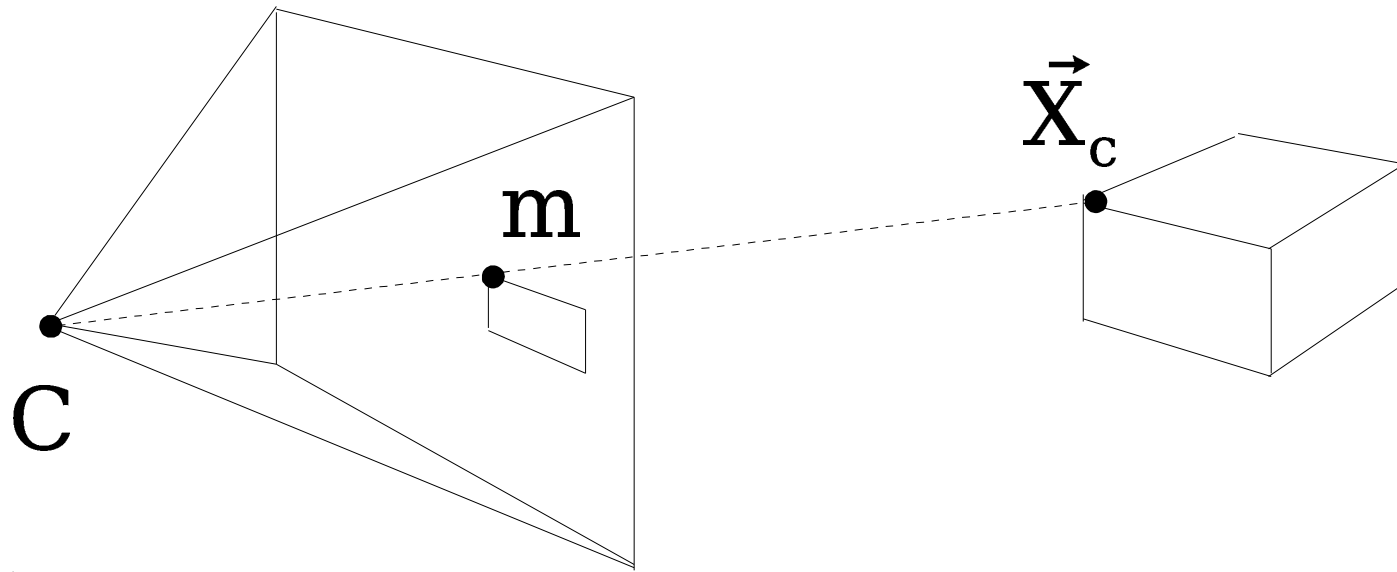
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- > 3D sensor not required
- > Requires two or more conventional cameras
- > Cameras can be uncalibrated



# Background: Camera

Pinhole Perspective projection model.



- >  $C$  = center of projection
- >  $\vec{X}_C$  = 3D point relative to  $C$
- >  $m$  = projection of  $\vec{X}_C$  on 2D image plane

# Camera Projection Equation

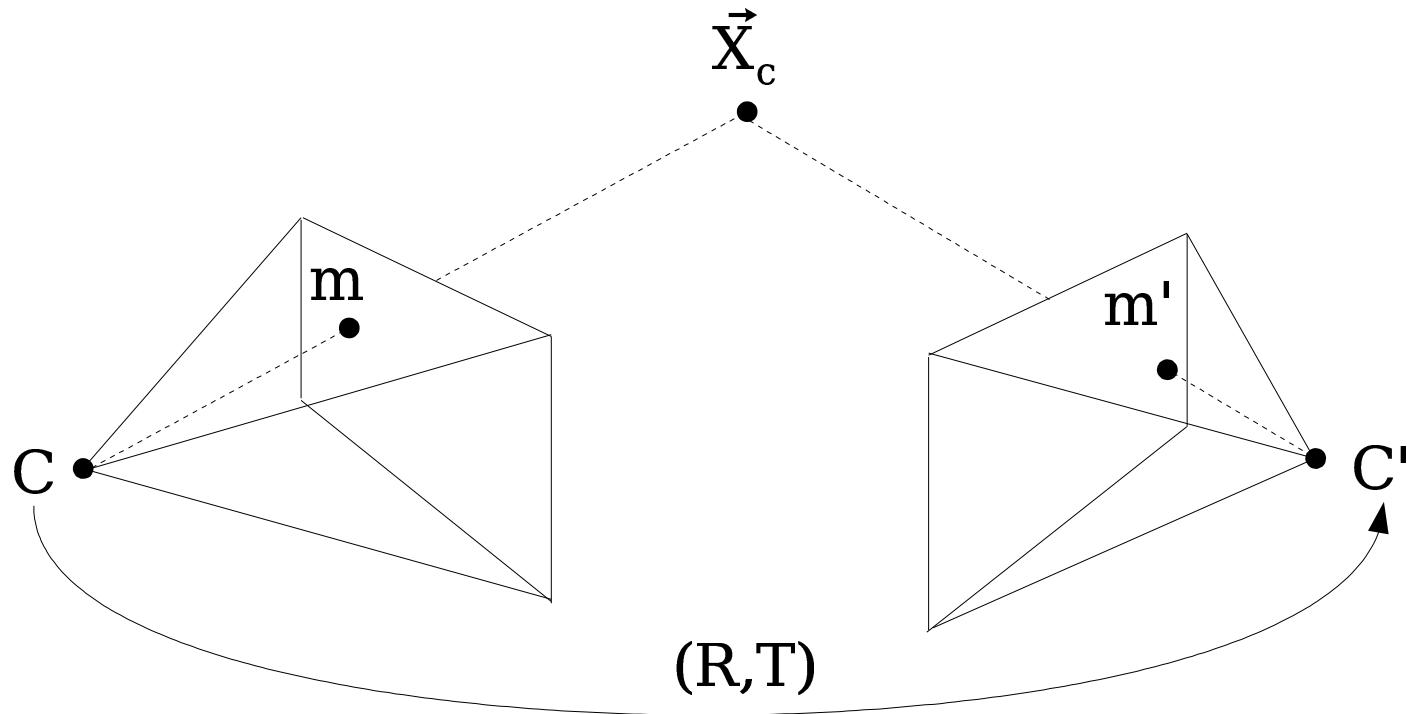
$$s\mathbf{m} = \mathbf{A}_C \vec{X}_C$$

Where:

- >  $\mathbf{m}$  = homogeneous 2D image coordinate  $\begin{bmatrix} x & y & 1 \end{bmatrix}^T$
- >  $\vec{X}_C$  = 3D point relative to camera center
- >  $\mathbf{A}_C = 3 \times 3$  camera calibration matrix encoding *intrinsic* parameters
- >  $s$  = the projective depth

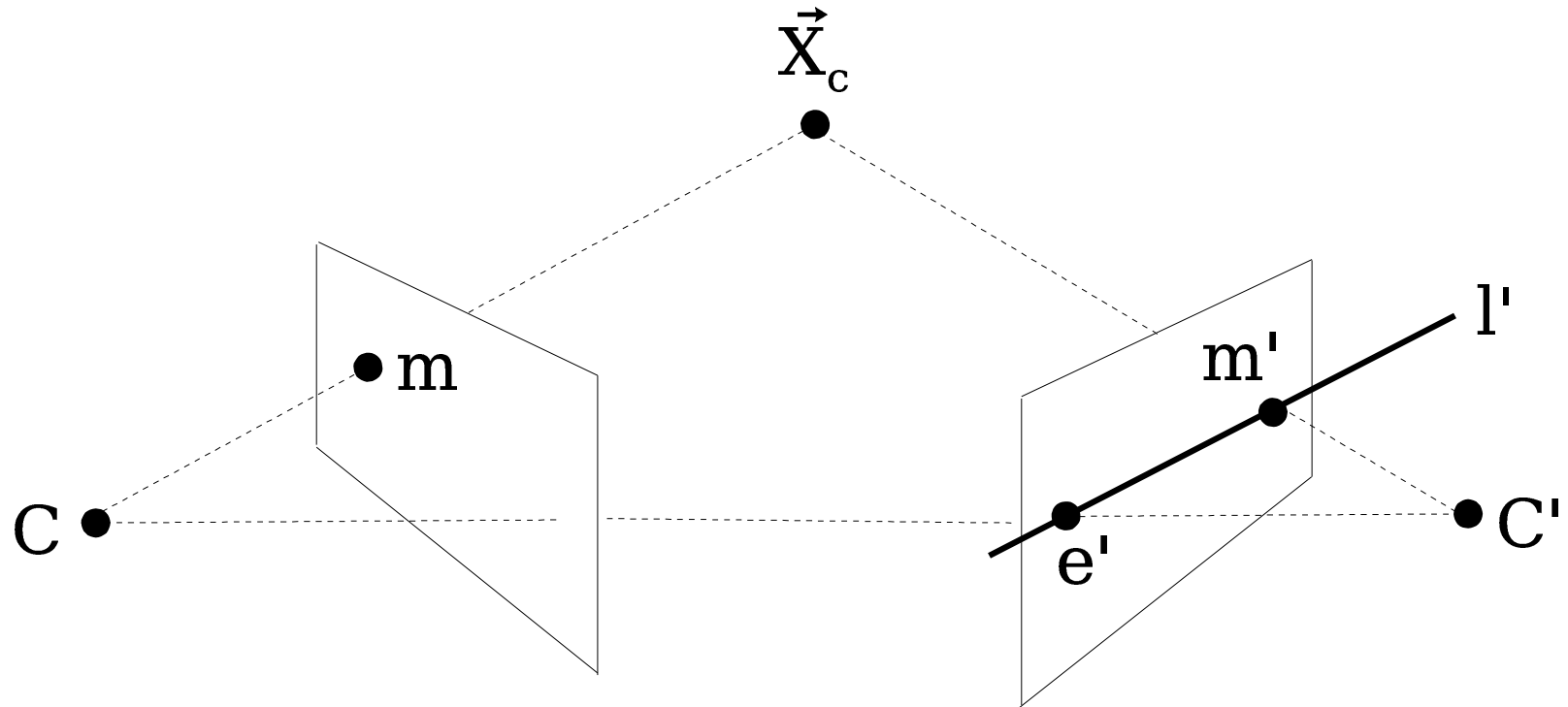
# Background: Stereo

Cameras related by rotation and translation



# Background: Epipolar Line

Point on camera image 1 constrained to lie on a line in camera image 2 (and vice versa).



# Background: Fundamental Matrix

Epipolar geometry encoded in the Fundamental Matrix:

$$\mathbf{m}^T \mathbf{F} \mathbf{m}' = 0$$

- >  $\mathbf{F}$  is a  $3 \times 3$  matrix.
- > Well studied in vision literature.
- > Given examples of corresponding  $\mathbf{m}, \mathbf{m}'$ , many techniques to solve for  $\mathbf{F}$ .

# Epipolar Calibration

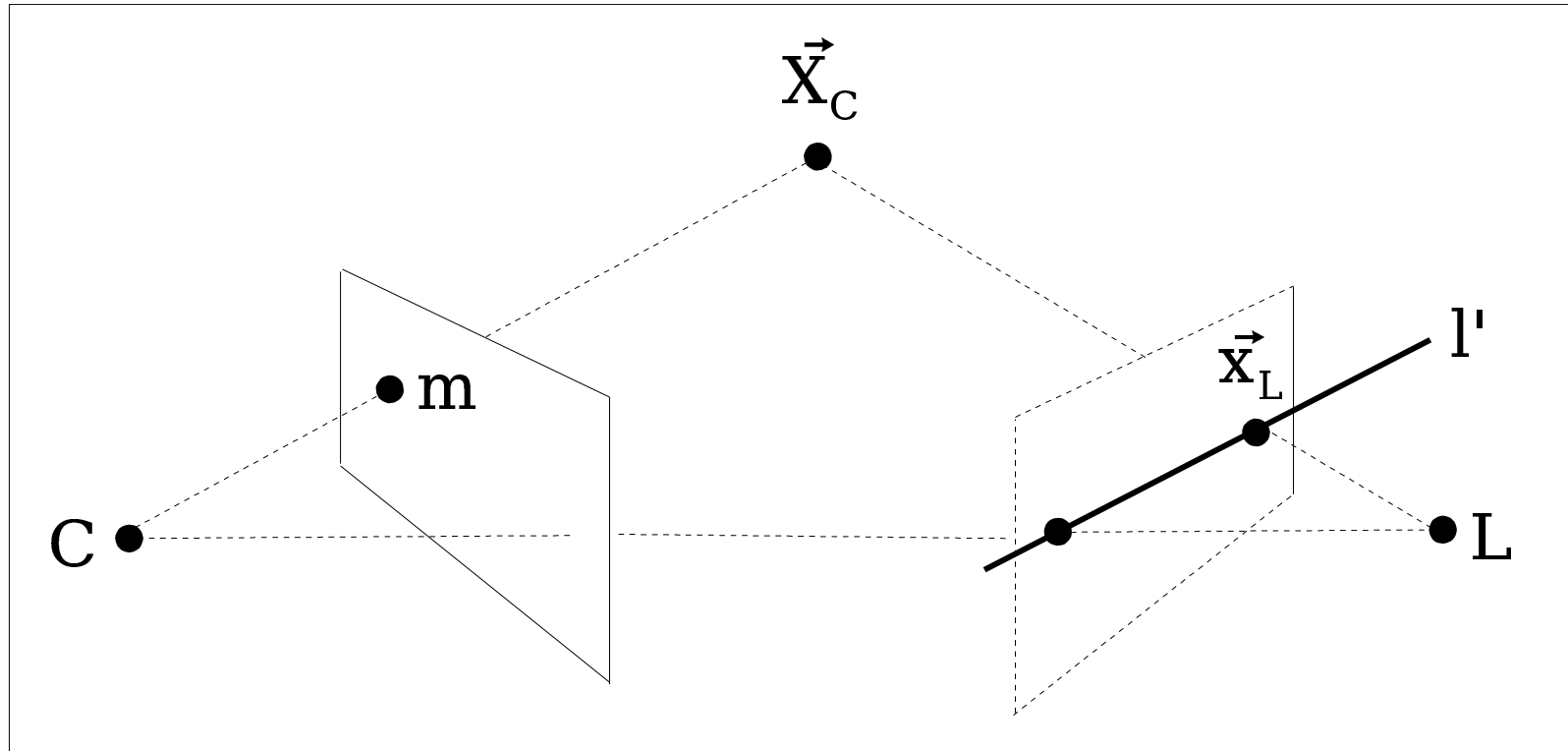
Key intuition: Laser is an inverted camera

- > Emits light instead of absorbing it
- > (u,v) laser controls congruent to (x,y) image coordinates.
- > Same linear relationship.

$$\underbrace{s\mathbf{m} = \mathbf{A}_C \vec{X}_C}_{Camera} \quad \underbrace{w\mathbf{x}_L = \mathbf{A}_L \vec{X}_L}_{Laser}$$

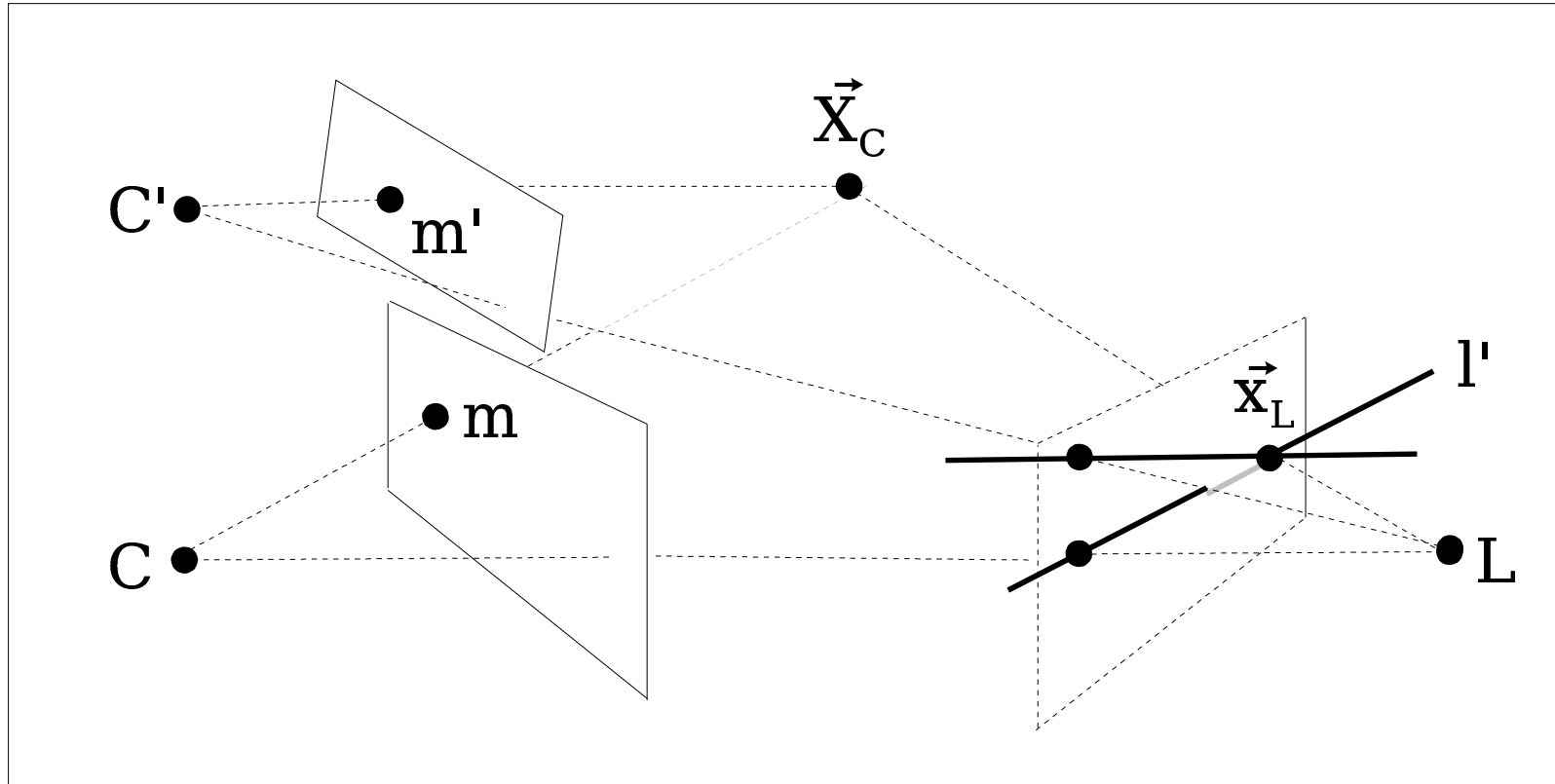
# Epipolar Calibration

One camera constrains laser control to a particular line in  $(u, v)$  space.



# Epipolar Calibration

Two cameras constrain laser control to the intersection of epipolar lines in  $(u, v)$  space.





# Epipolar Calibration

- > Fundamental matrix  $F$  encodes this geometric relationship
- > Each correspondance provides 1 constraint on  $F$
- > Utilize Hartley's Normalized 8 point algorithm to solve for  $F$
- > Need to solve for two  $F$ 's:
  - > Camera 1 and Laser
  - > Camera 2 and Laser

# Deriving Laser Control

Requires:

- > Two fundamental matrices acquired during calibration
- > Image coordinates of the target in each camera

Plugging these in yields:

- > Two linear constraints (one for each camera)
- > Two Unknowns  $(u, v)$

Solve directly for laser control  $(u, v)$ .

# Experimental Procedure

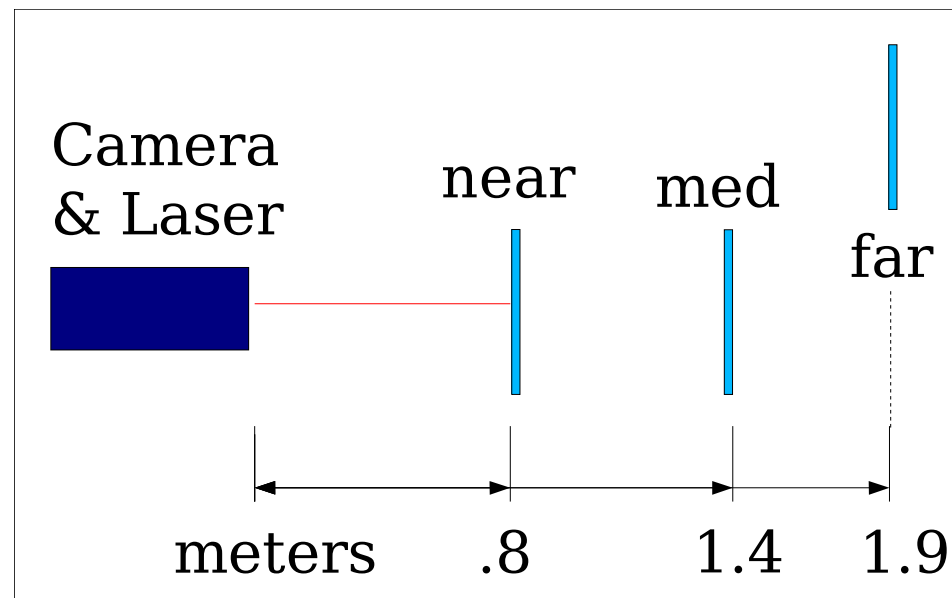
## Calibration

- > Move laser to an arbitrary  $(u, v)$  coordinate
- > Click on laser position in camera image
- > Laser position, clicked image position define corresponding points.
- > Laser moved in regular grid along image
- > Repeated at several different depth planes

# Experimental Procedure

## Targeting

- > Targets are the 4 extreme corners on a chessboard
- > Error is difference between actual position and target in mm
- > Test at 3 positions



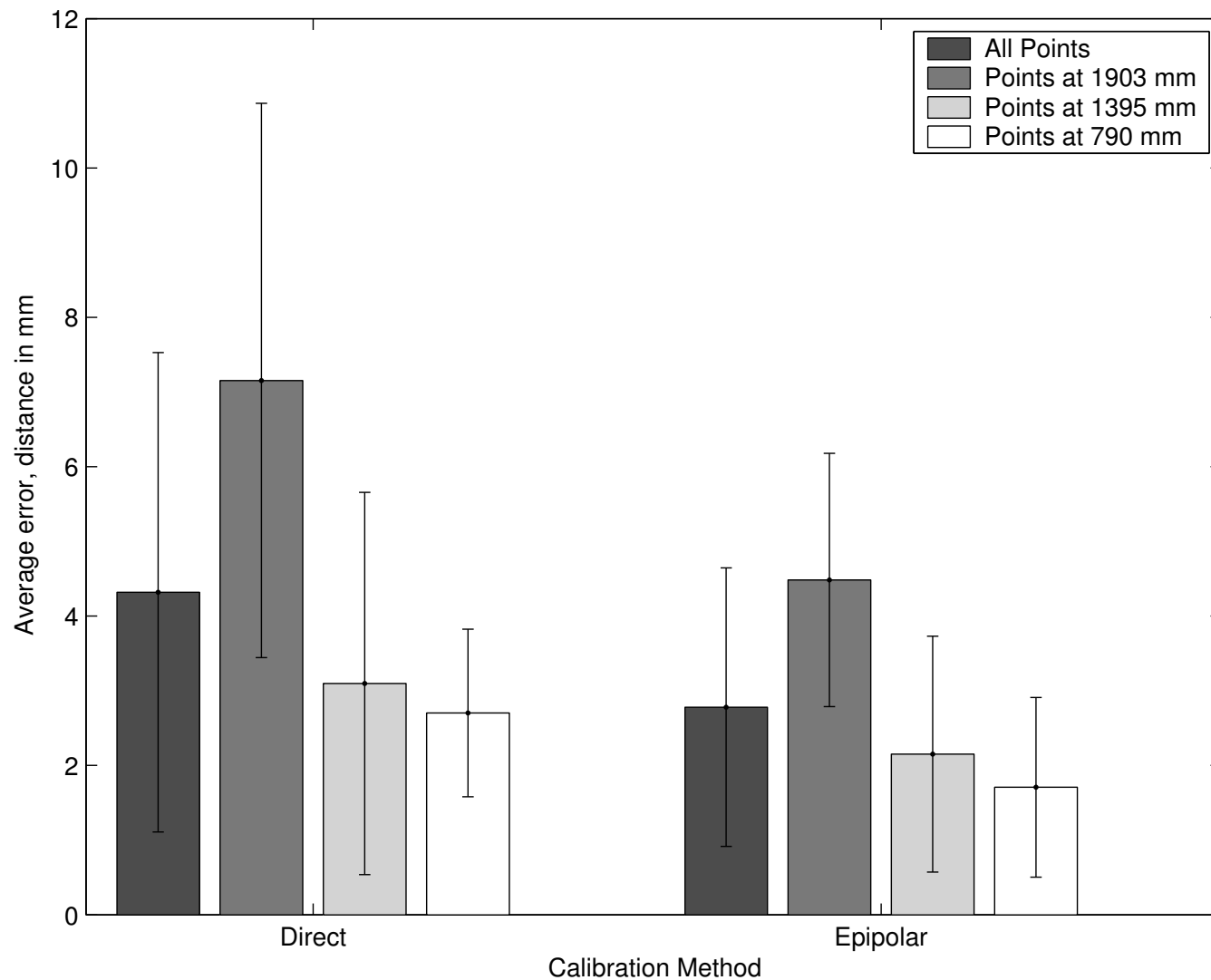
# Parameter optimization

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- > Number of calibration planes
- > Number of calibration points/plane
- > Maximum angle of laser

See paper for details.

# Results



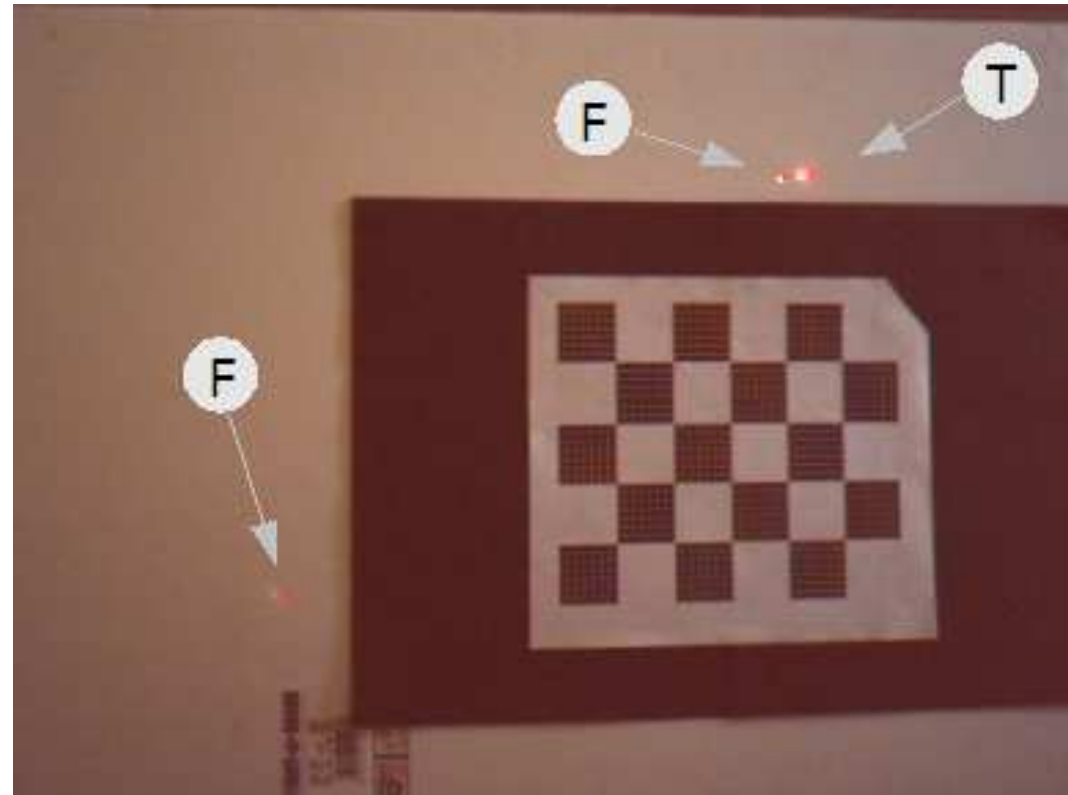
# Discussion

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- > Both methods accurate to within 3 – 4 mm on average
- > Epipolar method slightly better at all depths
- > Why?
  - > Maturity of fundamental matrix solution method.
  - > Noise in 3D sensor (epipolar method uses image coordinates directly)

# Automatic calibration

- > Mouse clicking is tiresome and prone to inaccuracies
- > Automatic detection must consider laser artifacts in camera image:

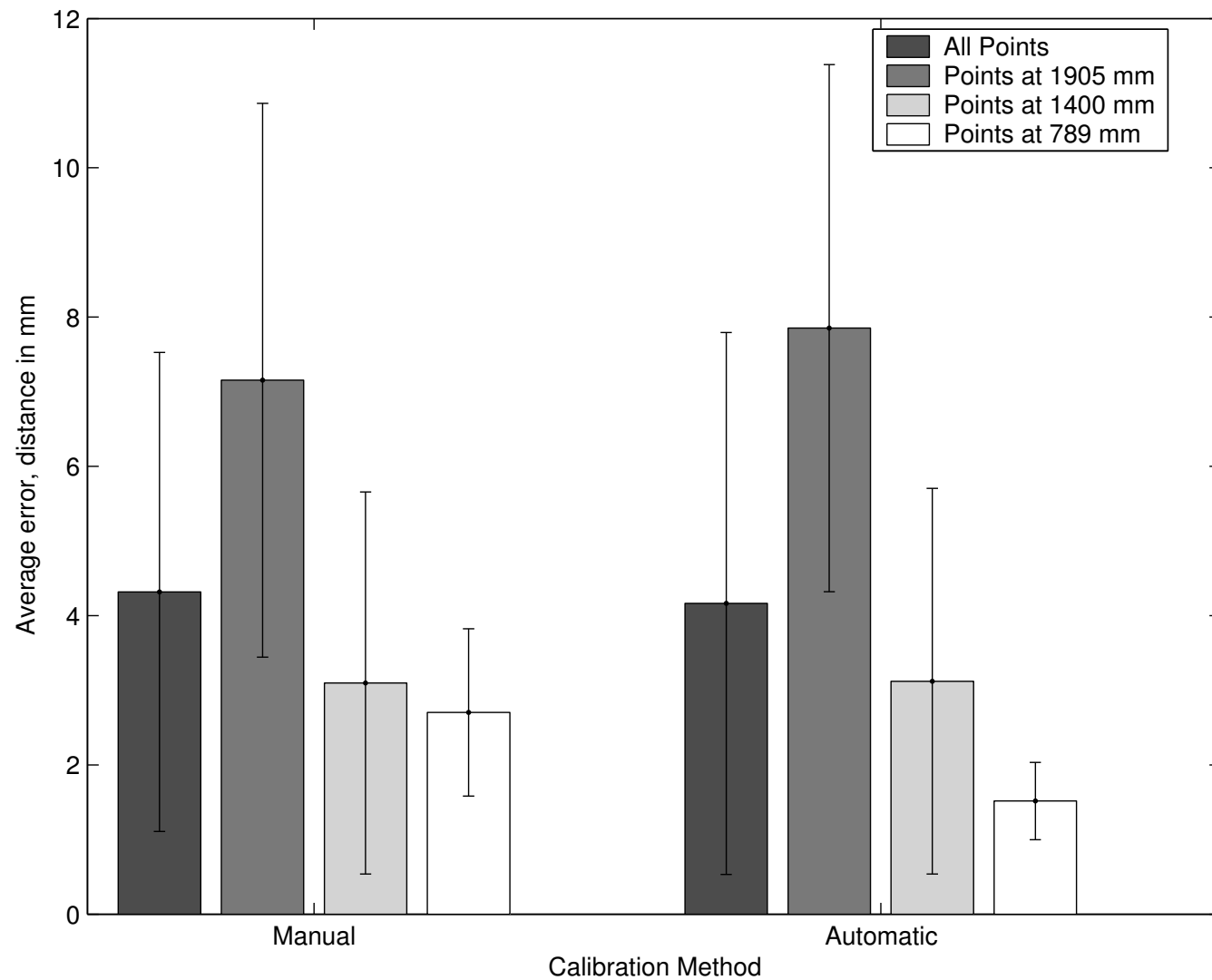




# Red dot detection algorithm

- > Capture background image (without laser)
- > Capture image with laser, subtract out background image
- > Keep red color channel only
- > Threshold pixels
- > Compute weighted center of mass  $(x, y)$  over entire image
- > Recompute using a window around  $(x, y)$

# Results



# Point Selection

- > Currently specify laser coordinates
  - > choose/detect corresponding image coordinate
- > Stereo camera only provides sparse depth
  - > Points without depth are thrown away during calibration
- > Can we specify image coordinates, then move laser to match?
  - > manual control laborious
  - > automatic control (chicken and egg problem?)

# Image Point Selection

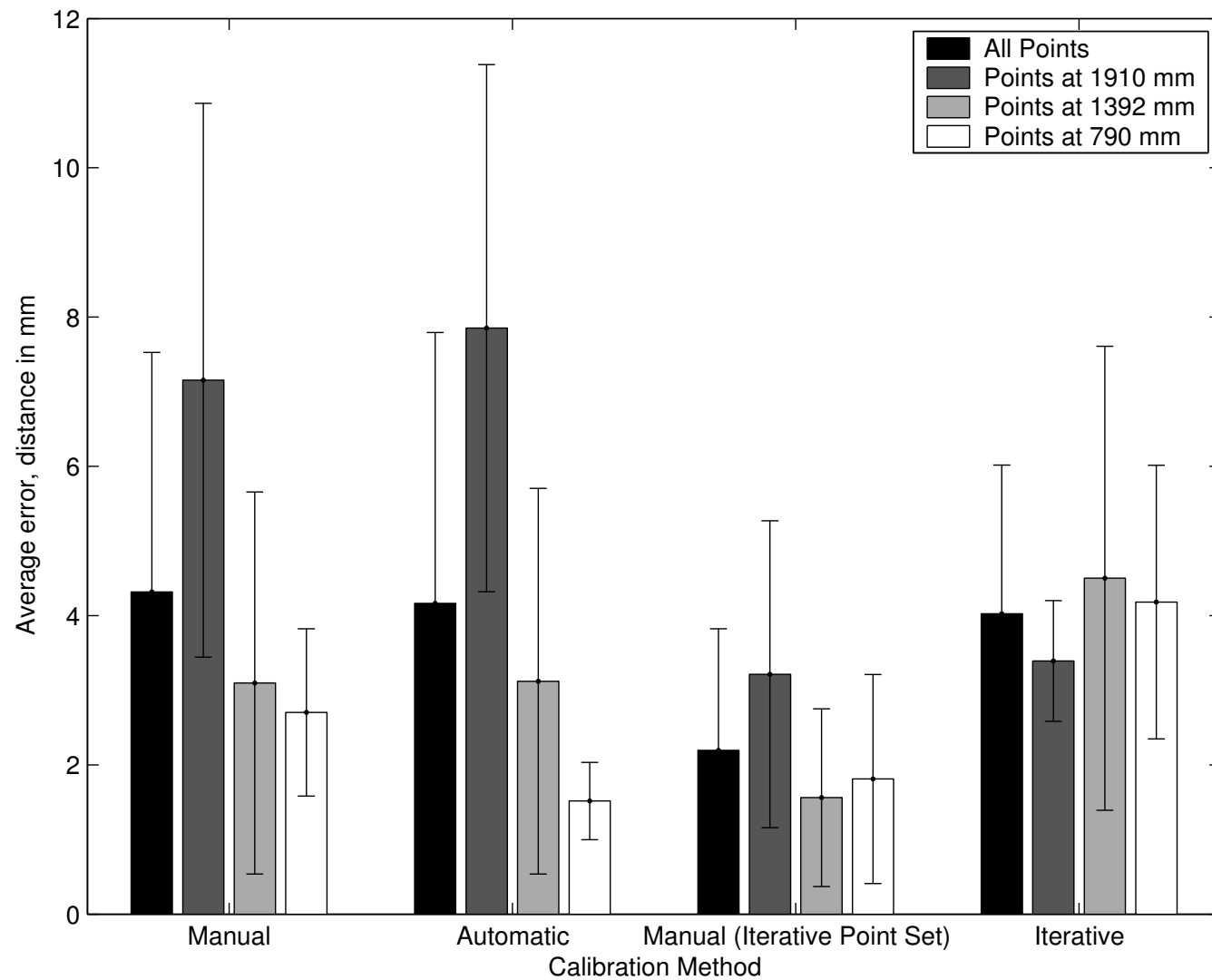
- > Algorithm:
  1. Measure distance between laser & target
  2. Move  $\alpha \cdot x.\text{distance}$ ,  $\beta \cdot y.\text{distance}$
  3. Repeat until distance = 0
  4.  $\alpha, \beta$  are constants determined empirically to minimize distance
- > Will probably only work if coordinate systems are roughly aligned.
- > Highly unsophisticated instance of visual servoing methodology.
  - > could be easily improved to be more robust

# Experimental Procedure

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- > Use chessboard corners as calibration points
  - > Take advantage of automatic corner detection
- > Repeat for 4 positions
  - > Use 8 points at each position

# Results



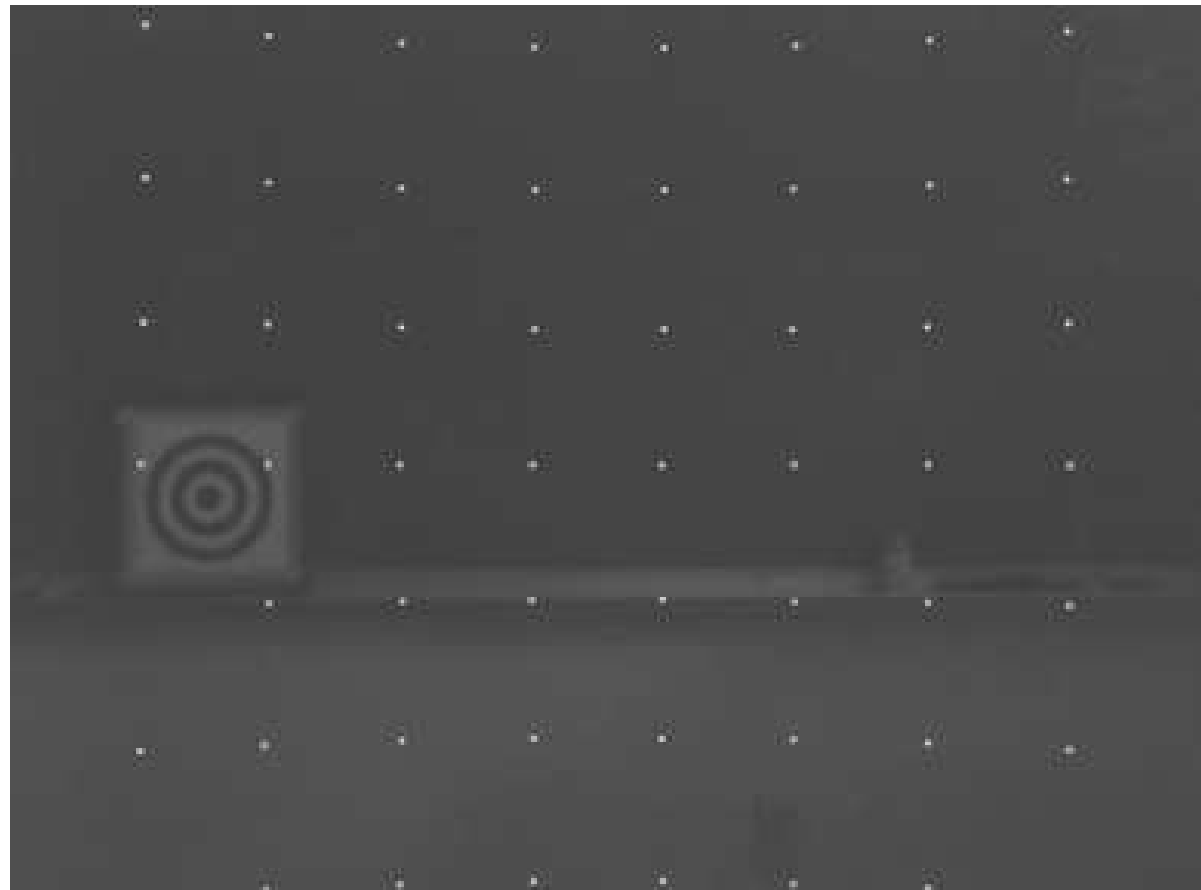
# Discussion

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- > Both manual and automatic calibration show improvement with inverse selection.
- > Point selection is more important than number of calibration points.
- > Improvement possibly due to sub-pixel accuracy of corner detection.

# Overall Discussion

- > The best overall average accuracy achieved is around 2.5 mm.
- > Good, but not perfect – bias.





# Future Work

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- > Identify and model non-linearity in laser unit.
- > Evaluate in comparison to visual servoing as an alternative targeting approach.

# Conclusion

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- > Two calibration methods
  - > Verified by experimental results to 3-4 mm accuracy
- > Automatic laser point detection
- > Image point correspondence
  - > Verified by experimental results to 2.5 mm accuracy

# Acknowledgements

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