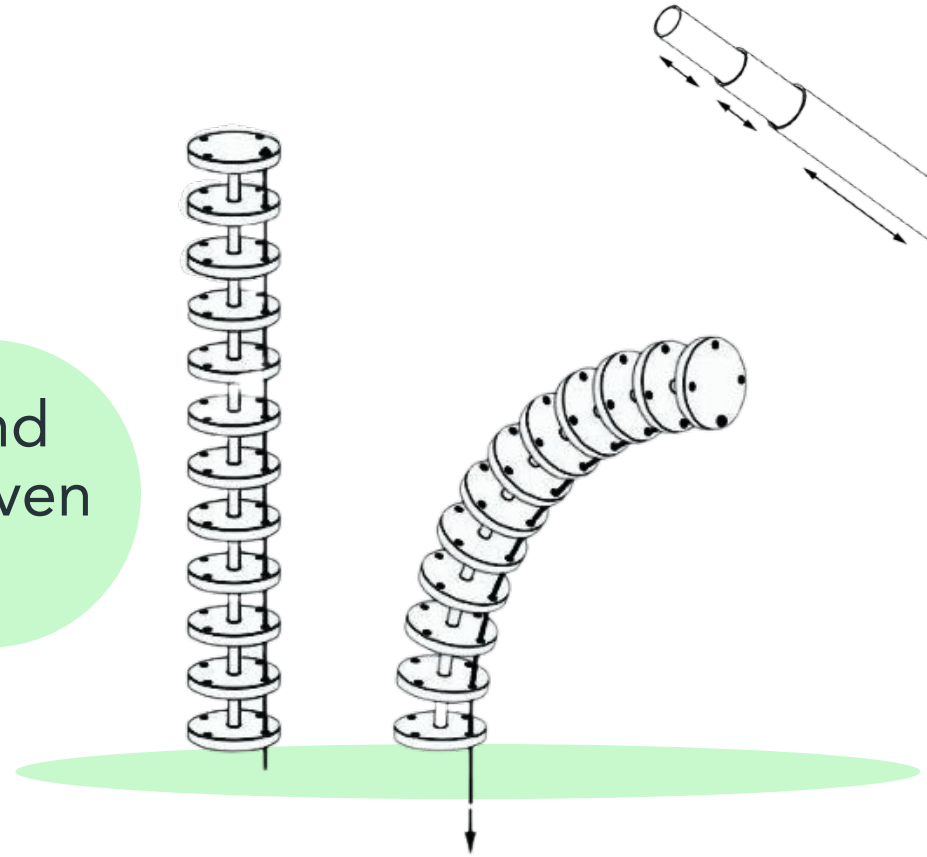




camera-based shape sensing and motion capturing of tendon-driven continuum robots

Supervisor: Priyanka Rao
Yasmeen Hmaidan





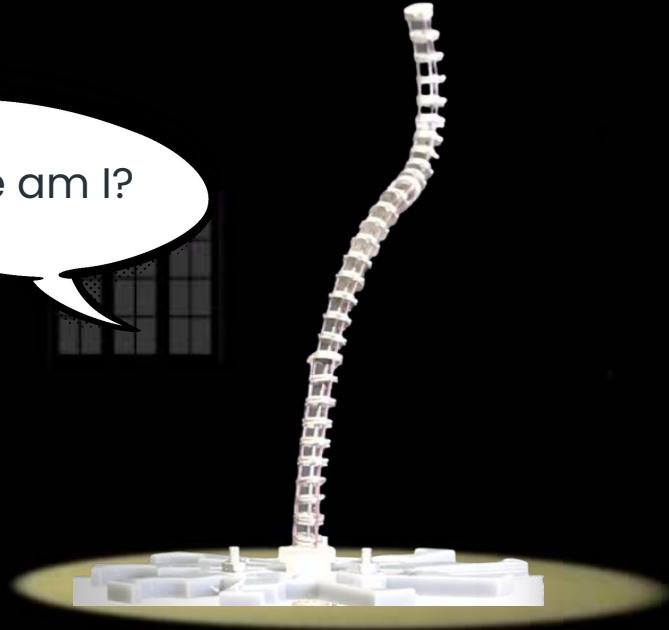
proximity

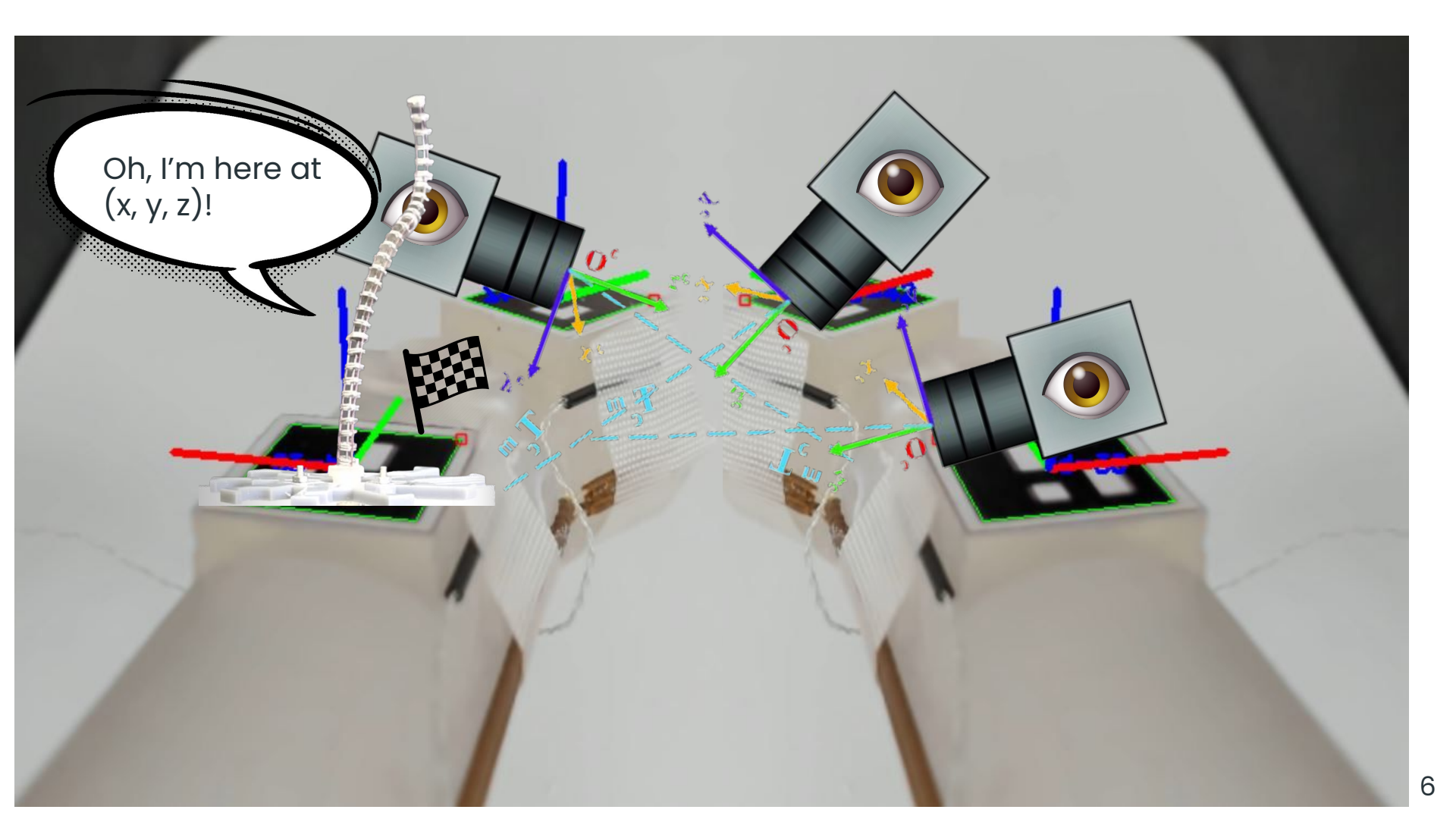
distance

object



Where am I?





Oh, I'm here at
(x, y, z)!

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- Continuum Robot
- TDCR elements

2

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- Purpose
- Different Strategies
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- Multiple-Camera system
- Computer Vision

4

Project Timeline

- Workflow
- Target Outcomes



1

Intro to TDCRs

Tendon Driven Continuum
Robots



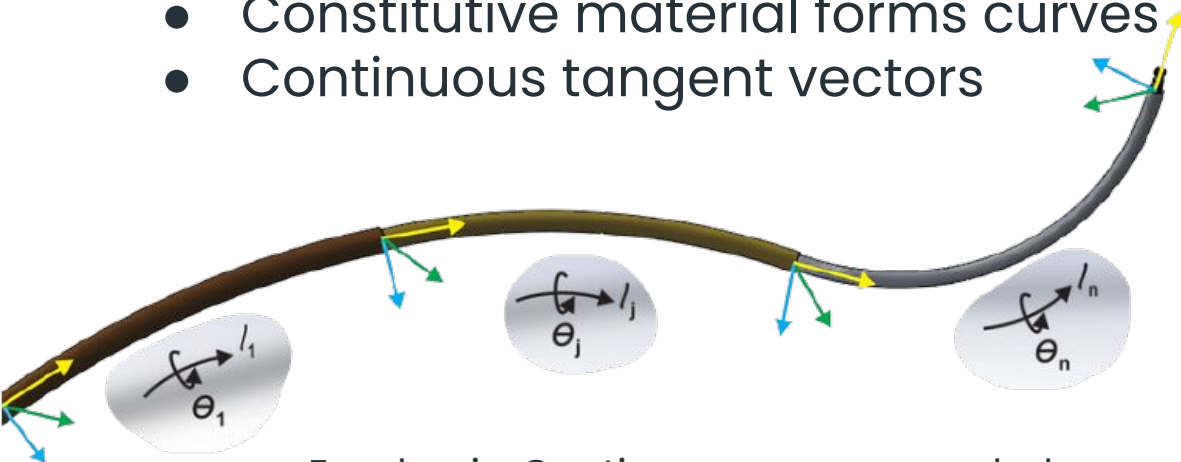
What is a Continuum Robot (CR)?

According to the Burgner-Kahrs, Rucker, & choset, 2015 definition:

- Actuatable structure
- Constitutive material forms curves
- Continuous tangent vectors

Note, no assumptions are made on:

- CR's actuation method
- Elasticity of composing materials



Emphasis: Continuous curve morphology

Pro: conformity
Con: less precise positioning (w/o discrete rigid links)

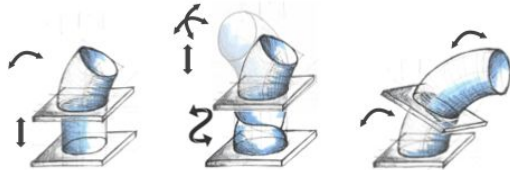
CR Elements

Ultimate Goal:

- Fully controllable continuously bending manipulator

Motion Primitives Set:

- Extension & Contraction
- Bending
- Twisting

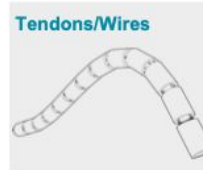


range of motion \propto # of stacked segments

Extrinsic Actuators:

- Tendon/Wire-Actuated
- Telescoping Pre-Curved Tubes

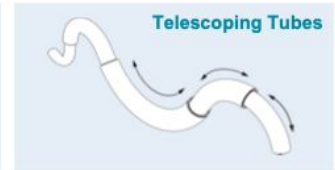
EXTRINSIC



DISCRETE



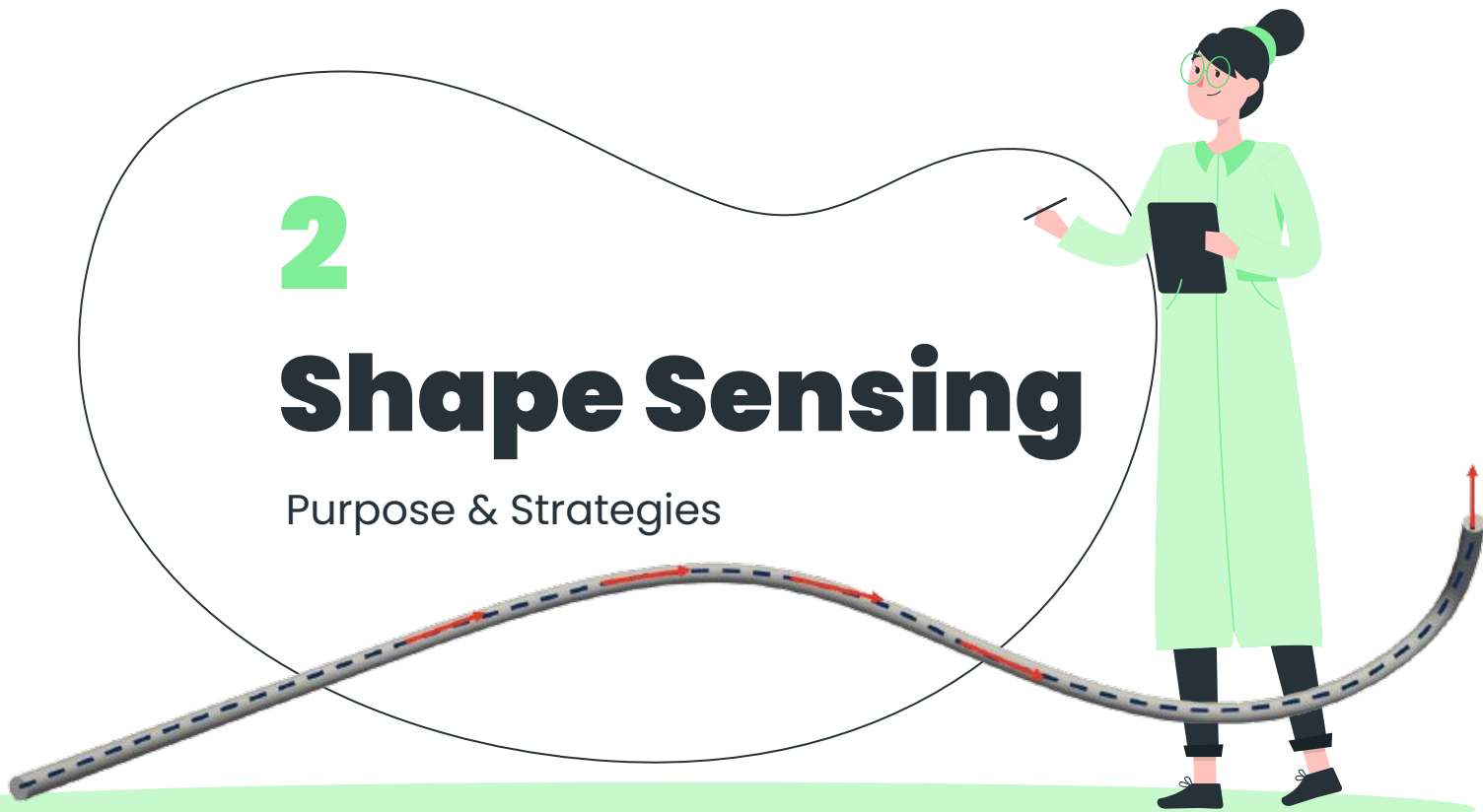
CONTINUOUS



2

Shape Sensing

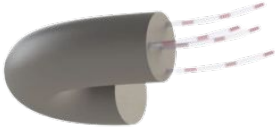
Purpose & Strategies



Shape Sensing Types

Embedded Sensors

- **Fibre-optic sensing**
 - Geometric fibres and sensor array
 - **Pro:** Real-time shape information
 - **Con:** Expensive



- **Electromagnetic sensing**
 - Locates objects with sensor coils
 - Real-time pose tracking
 - **Pro:** No line-of-sight restrictions
 - **Con:** Robot rigidity



(CSC476, 2020)

External Sensors

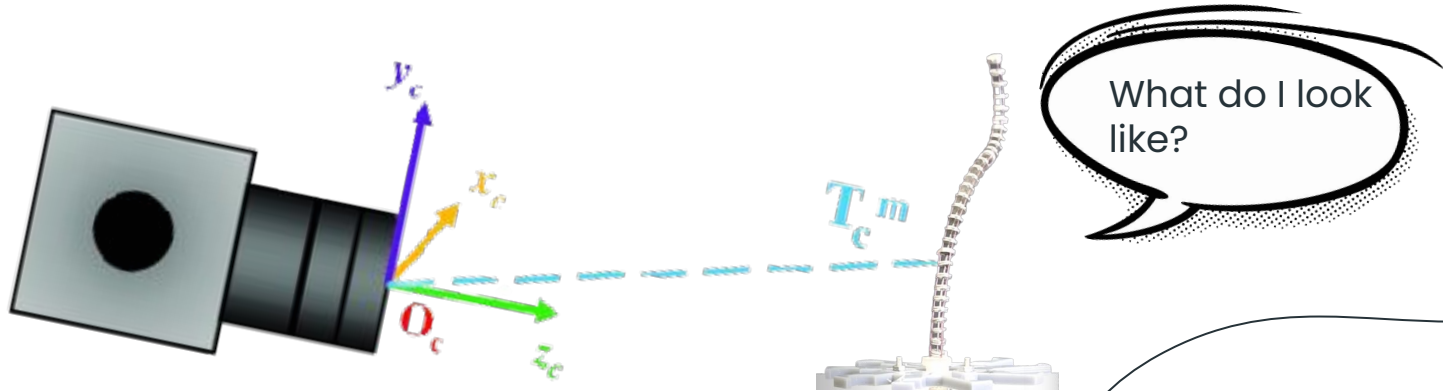
- **Mechanical probe**
 - Touches robot to measure tip position or shape
 - **Pro:** Measure multiple distinct locations
 - **Con:** Not contactless
- **Laser probe**
 - Emits laser line that uses images from the camera along robot to get object distance
 - **Pro:** Dense point cloud of robot shape
 - **Con:** Time-consuming



Shape Sensing Types

Ideal Choice: External image-based sensing

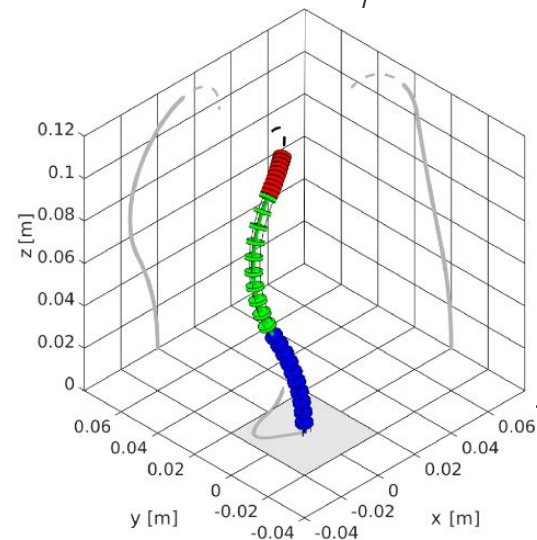
- Multiple **external cameras**
- **Contactless**
- Used in both **static** and **dynamic** TDCR applications
- Precise **3D reconstruction**
- **Challenge:** Direct **line-of-sight** (occlusions/partial views are processed)



3

Depth Estimation (methodology)

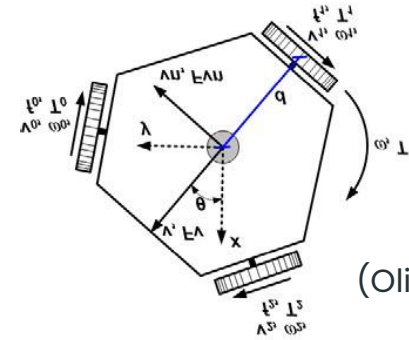
Multiple-Camera system &
Computer Vision



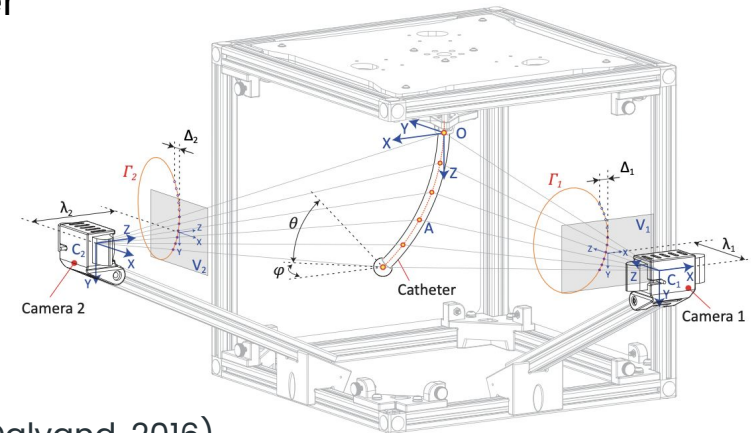
Multiple-Camera system

Main Project Goal:

- Orient **three cameras**
- Calculate **3D transformations** to 'global' **coordinates**
- **Input:**
 - **Pixel** coordinates of detected object (per camera)
- **Output:**
 - **3D** coordinates in decided frame + the three axes (x,y,z)



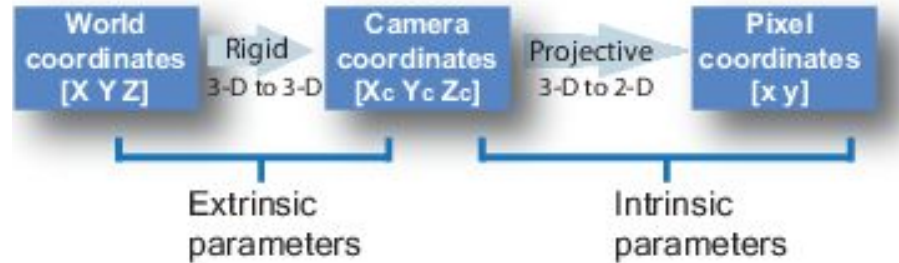
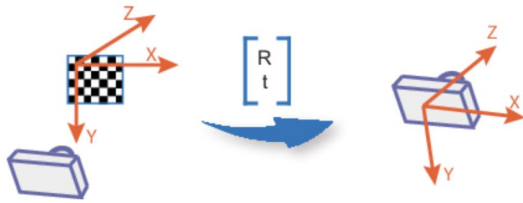
(Oliveria, 2008)



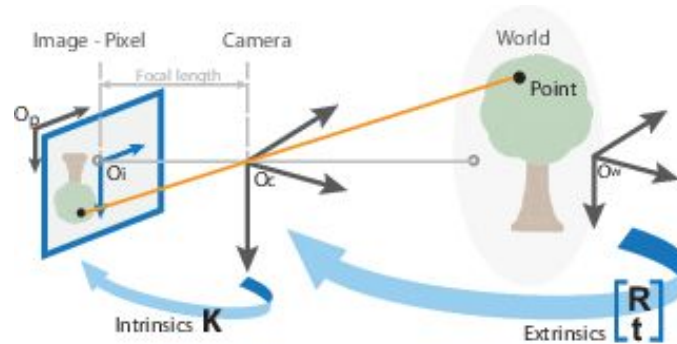
(Dalvand, 2016)

Camera Calibration Algorithm

- Calculates camera matrix (to detect object's global location, size, etc.)
 - Two **parameters**

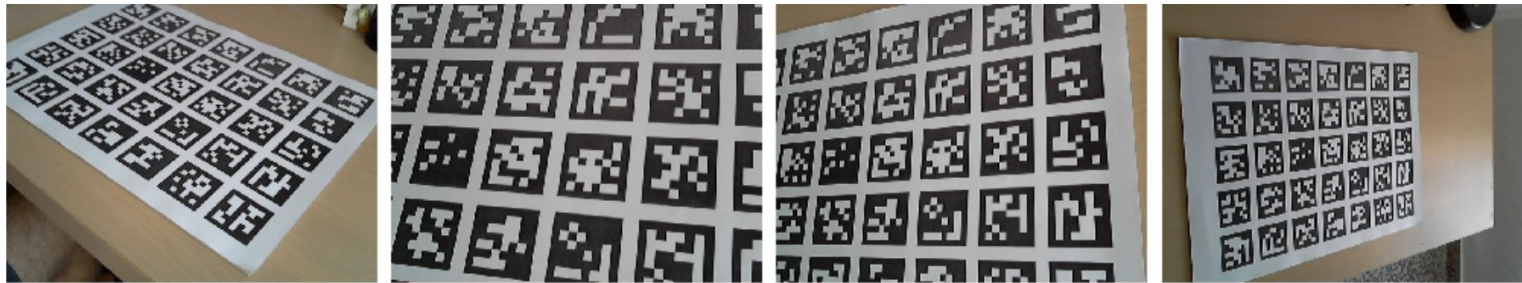


$$\begin{bmatrix} f_x & 0 & 0 \\ s & f_y & 0 \\ c_x & c_y & 1 \end{bmatrix}$$



(Mathworks, 2021)

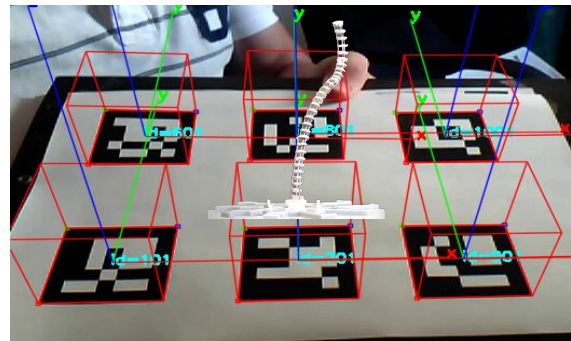
Camera Calibration: ArUco Module Markers



ArUco calibration viewpoints

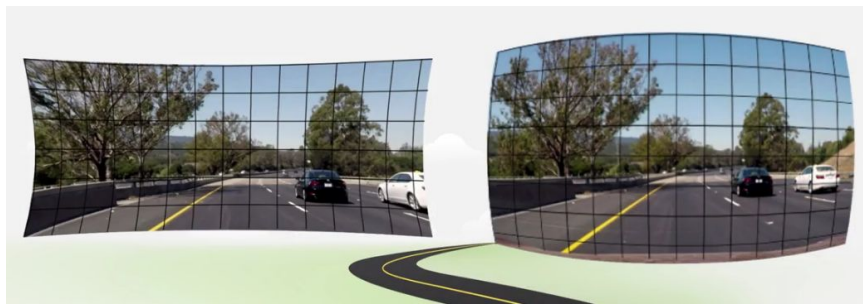
A type of **QR code**:

- **Black border** = fast image detection
- **Inner binary matrix** = identifier and detects + corrects distortion error
- **Output**: returns a list of detected markers



Calibration Challenges

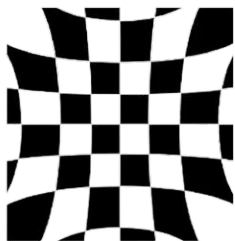
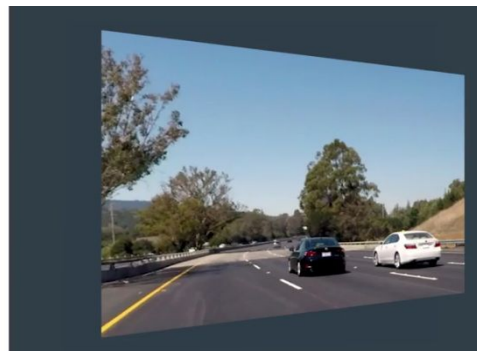
Problem: camera **distorts images** when not parallel to the imaging plane.



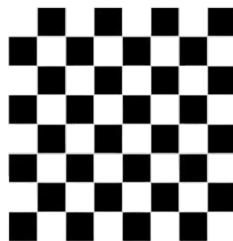
Radial Distortion: straight lines appear curved.



Tangential Distortion: some areas in the image look nearer.

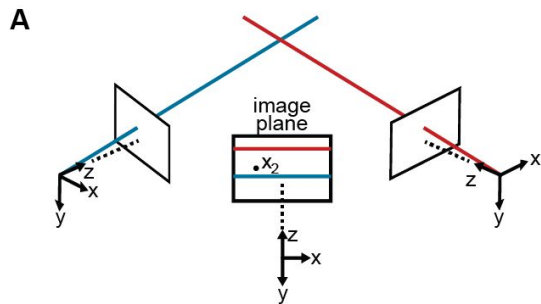


Pincushion Distortion

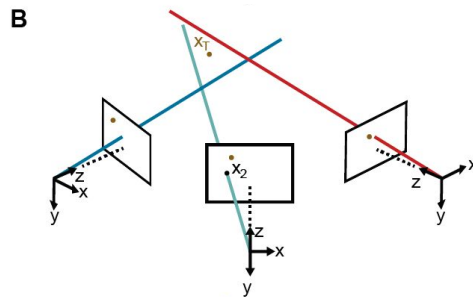


No Distortion

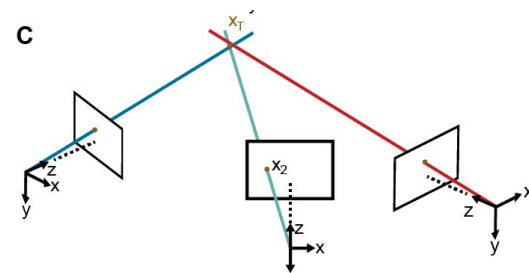
Calibration Challenges: Stereovision System



Correct 2D pose estimation by using epipolar relationships.



Triangulated point not synced and not accurate 3D position estimate.



Correct camera locations by **optimized orientation & location**.

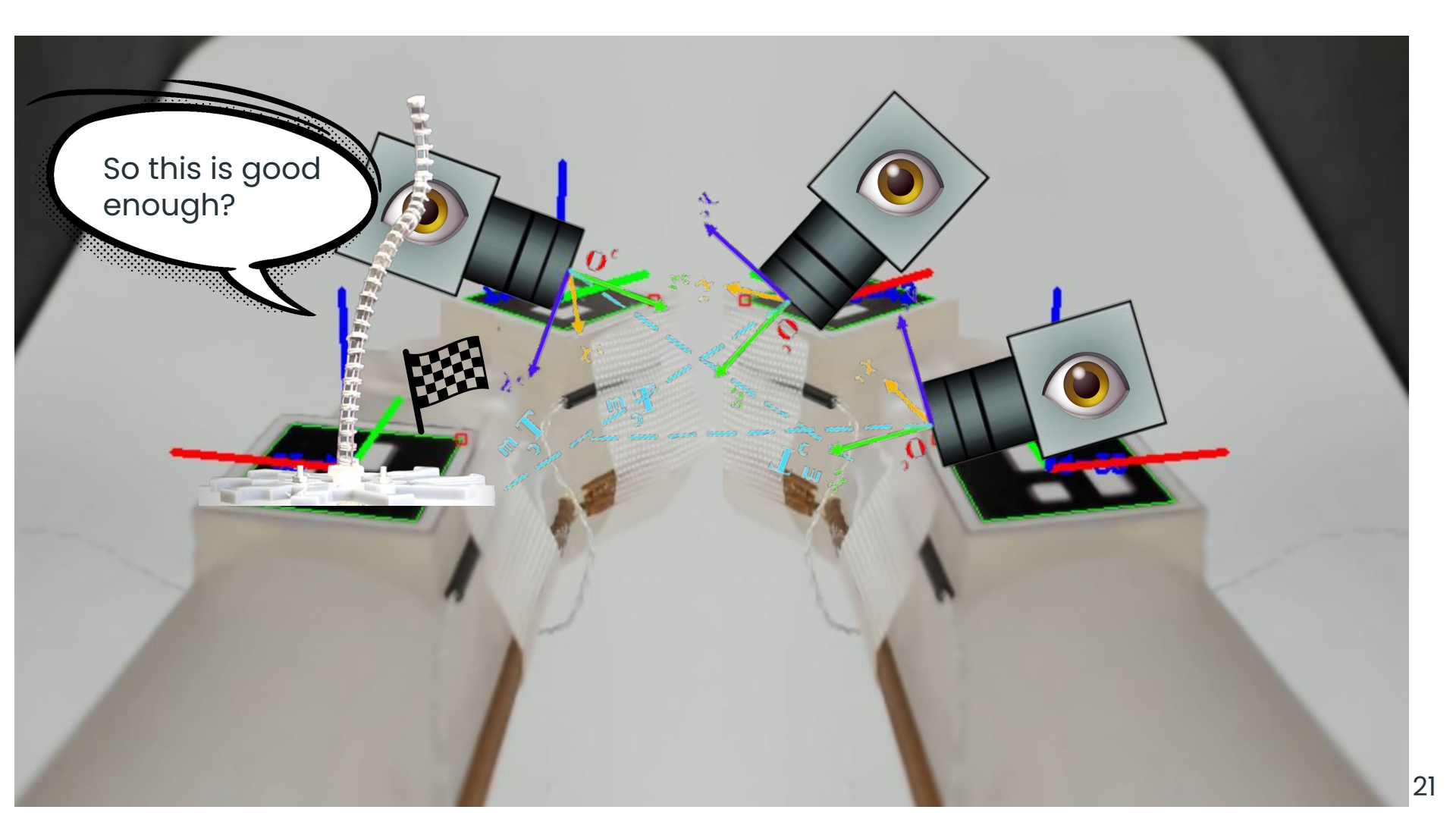
3D reconstruction accuracy of object: ∞ # of **cameras** and triangulated **angles** covered.

(DeepFly3D, 2019)

What does good camera calibration look like?

Good Camera Calibration = **accurate estimates** of objects in the **world** and where the **TDCR** is in this environment with no blind spots.

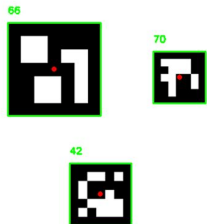
- **External** image-based sensing
- **Three-camera** system
- **Intrinsic** and **extrinsic** camera calibration **parameters**
 - ArUco Markers
- **OpenCV** distortion & calibration optimization methods



So this is good enough?

The image shows a 3D perspective of a race track layout on a light gray surface. Three white, rectangular blocks are positioned at the corners of the track. Each block has a black sensor array on its top surface and a square module with a large, realistic eye graphic mounted on top. A white, segmented, worm-like robot is positioned at the start of the track on the left. A black and white checkered flag is also at the start. The track is marked with red, blue, and green lines. Various colored arrows (red, blue, green, yellow) and small icons (circles, squares) are scattered across the track, possibly representing data or obstacles. A speech bubble from the leftmost eye module contains the text 'So this is good enough?'.

Research Plan



ArUco Markers

First 10 markers of
ARUCO_MIP_36h12

Camera Setup

Live video feed to
image collection

Computer Vision

OpenCV

Application

Rectangle robot
base setup

3 cameras in
tandem orientation

Image processing
pipeline

Goal

Establish global
coordinate system

Calculate extrinsic
& intrinsic params

3D Reconstruction
and transformation



TDCR Image Processing Pipeline

Pre-processing

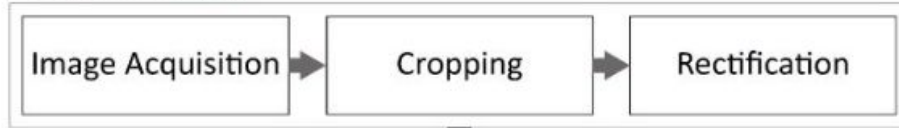
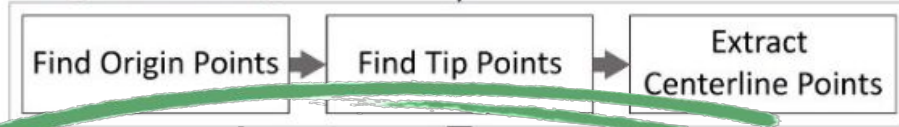
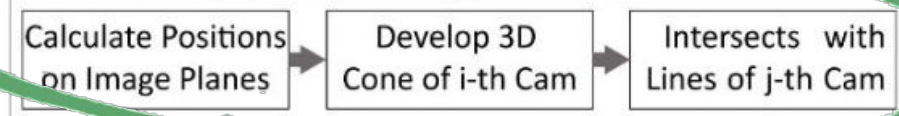


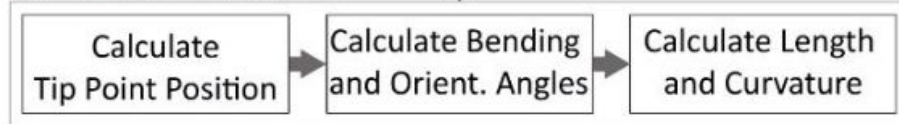
Image Processing



3D Reconstruction

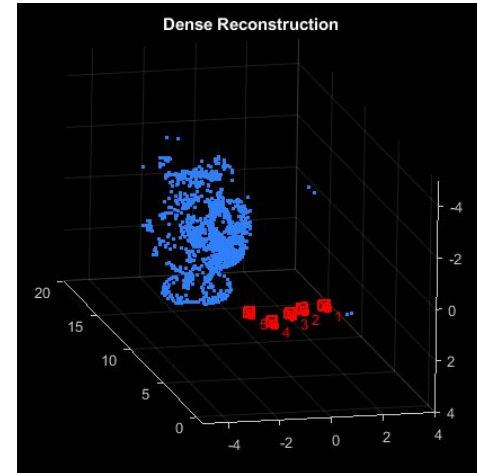


Parameter Calculation



3D OpenCV Reconstruction:

- Segmentation
- Epipolar geometry
- Distortion reduction



4

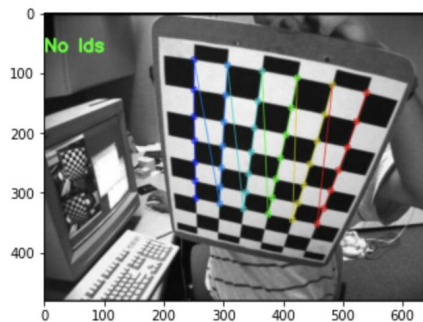
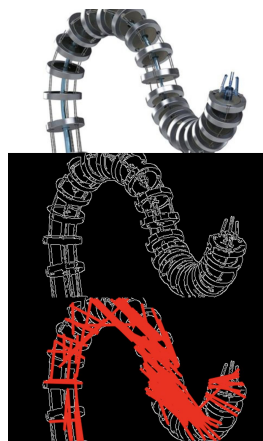
Project Timeline

Workflow & Target Outcomes

So what now?



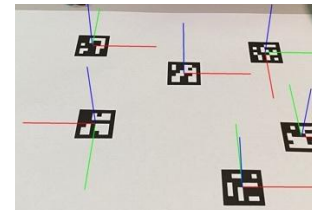
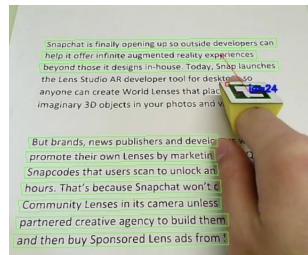
Project Timeline



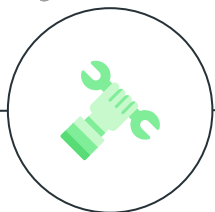
```
# Find the chess board (calibration pattern) corners
ret, corners = cv2.findChessboardCorners(gray, (7,6),None)

# If found, add detected corner points
if ret == True:
    objpoints.append(objp)
    corners2 = cv2.cornerSubPix(gray,corners,criteria)
    imgpoints.append(corners2)

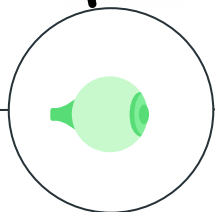
# Draw and display the corners
img = cv2.drawChessboardCorners(img, (7,6), corners2,ret)
```



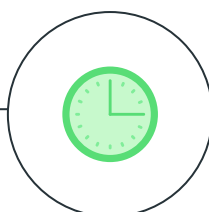
weeks 1-2
OpenCV
Tutorials
+ CSC476



weeks 3-5
Camera
calibration
+ Aruco
markers



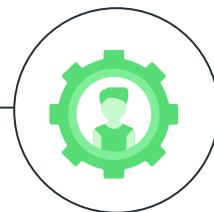
weeks 6-8
Set up camera
system + 3D
transforms



weeks 8-10
3D depth
mapping +
real-time
tracking

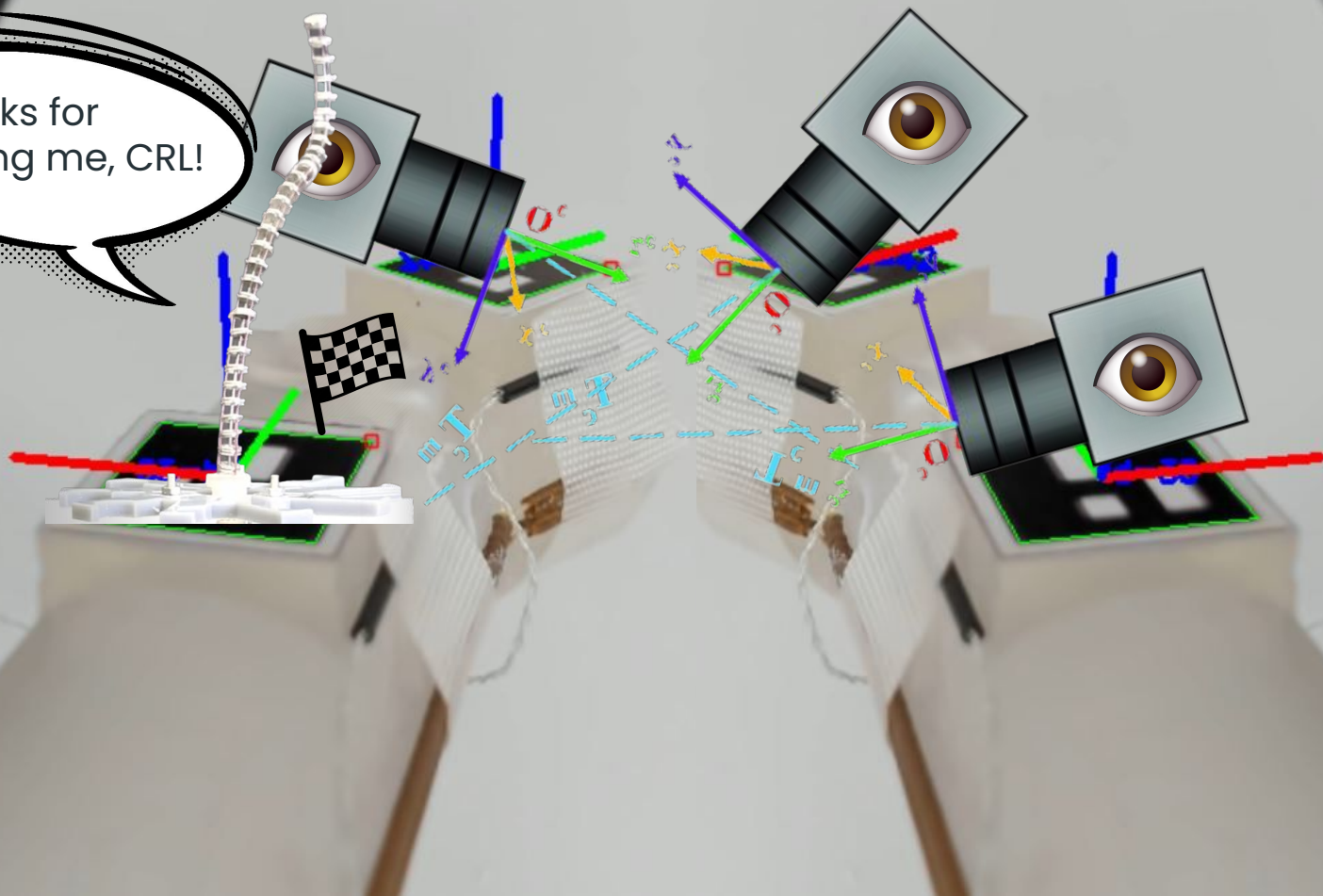


weeks 11-12
Extra
documentation +
report writing



weeks 13-16
TDCR robot trial +
GUI + ML joint to tip
position mapping

Thanks for
finding me, CRL!



References

https://www.researchgate.net/figure/Three-wheel-Omnidirectional-robot_fig10_256089781

<https://april.eecs.umich.edu/software/apriltag.html>

https://www.researchgate.net/figure/The-DeepFly3D-pose-estimation-pipeline-A-Data-acquisition-from-the-multi-camera_fig4_336273571

https://www.researchgate.net/figure/Three-wheel-Omnidirectional-robot_fig10_256089781

<https://robotics.stackexchange.com/questions/19901/apriltag-vs-aruco-markers>

<https://www.mathworks.com/help/vision/ug/camera-calibration.html>

https://www.youtube.com/watch?v=E9ka_2mAXvw

https://docs.opencv.org/master/da/d13/tutorial_aruco_calibration.html

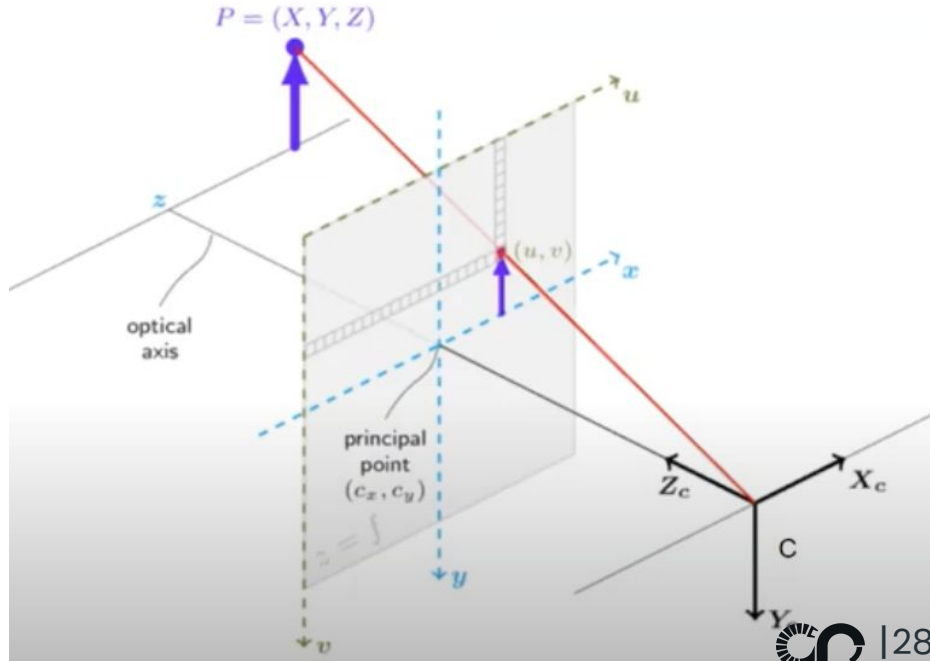
http://biorobotics.harvard.edu/pubs/2016/ref_conf/MDalvand_SMC2016.pdf

Calibration Types

Intrinsic Calibration

- Given a point $\mathbf{p} = (x, y, z)$ in camera frame
- Image coordinates calculated from top of frame in a live camera feed
- w.r.t camera center (principal point) and distance to image plane (focal length)
- PPM maps \mathbf{p} to image coordinates (u, v)
- **Goal:** camera parameters

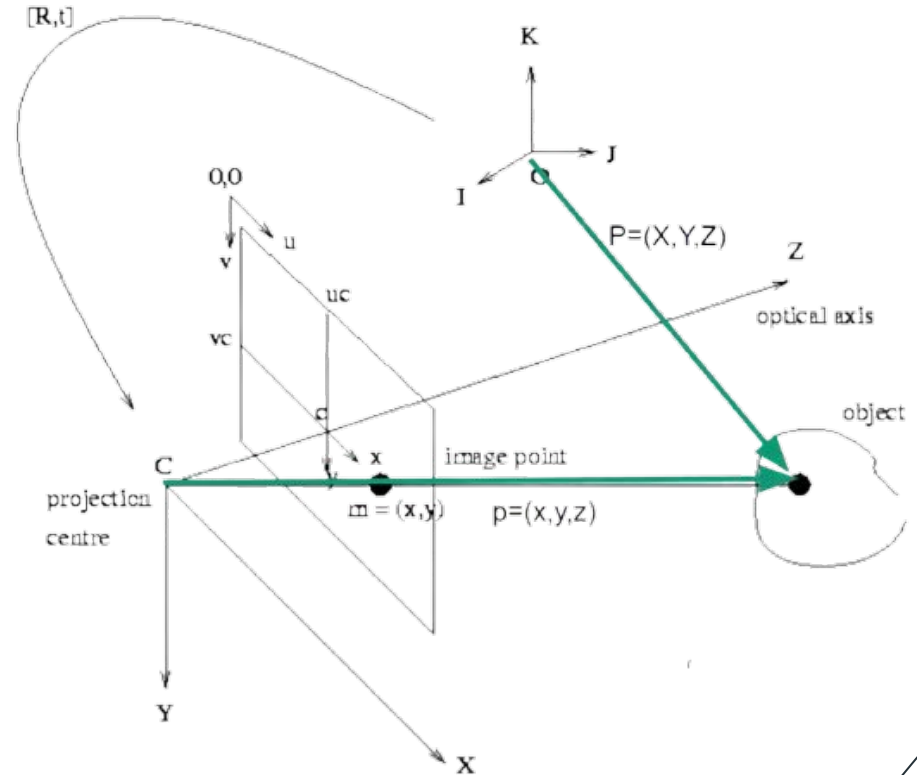
Perspective Projection Model (PPM)



Components

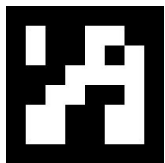
Extrinsic Calibration

- Maps where object is in **camera frame**
- Then, maps it to robot frame with **rotational translation**
- **Goal:** know where cameras are relative to TDCR in real-time



Pros & Cons

ArUco Markers



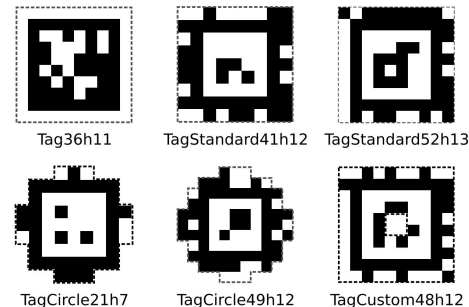
Pros

- Easier OpenCV implementation
- Available arUco marker generator
- Fewer false detection (w/ default)

Cons

- More computationally intensive
- Newer versions are GPL licensed, so opencv is older
- More tuning parameters

AprilTags

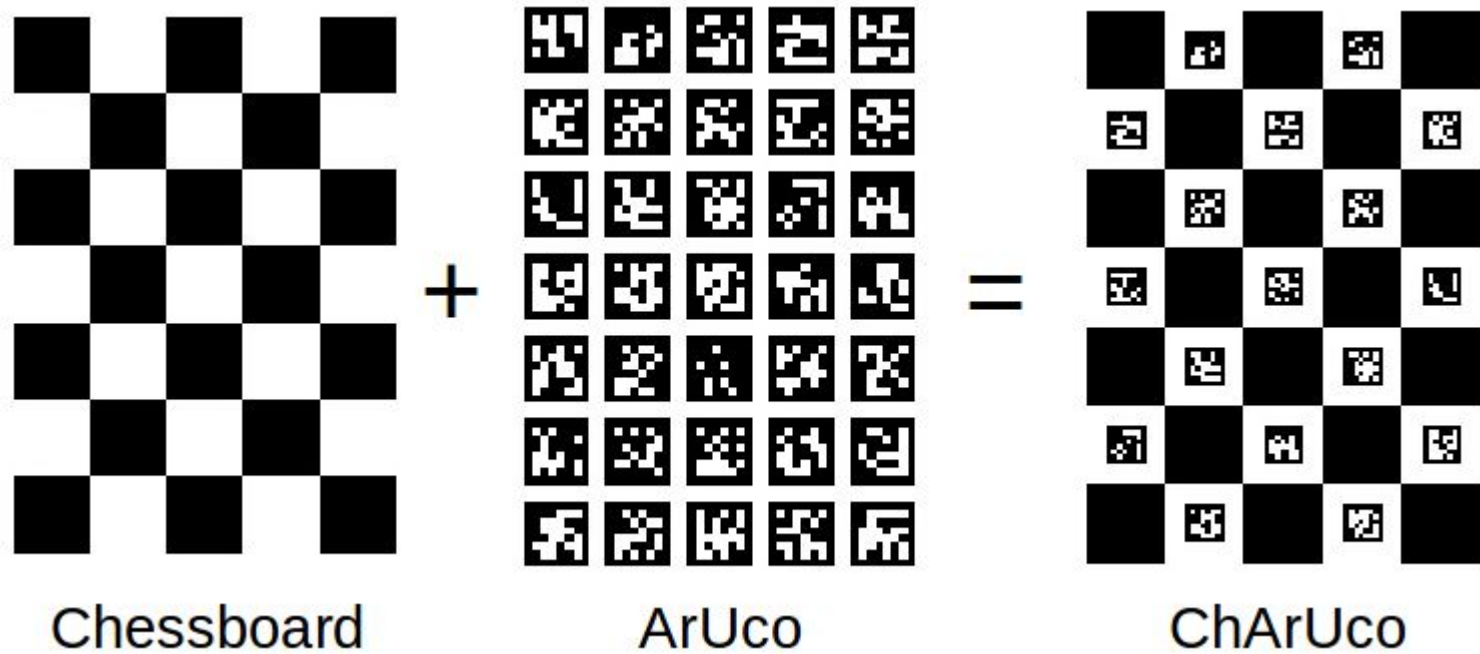


Pros

- BSD License
- Fewer tuning parameters
- Long distance compatibility
- More flexible marker design
- Less computation

Cons

- No opencv implementation
- More steps to obtain markers
- More false detection (with default parameters)

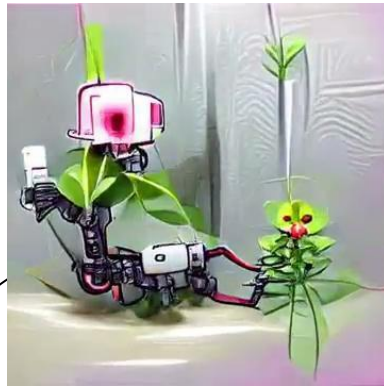
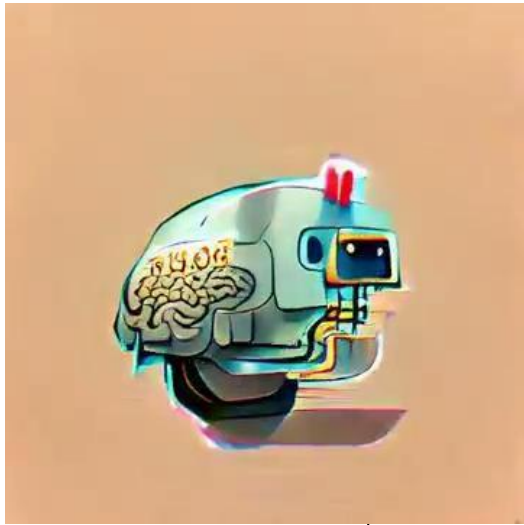


ChAruco boards > ArUco boards for camera calibration = more accurate marker corners.

Benefits: occlusions and partial views are allowed, and not all the corners need to be visible in all the viewpoints.



(OpenCV, 2019)



**thanks &
some CRL
art!**

