

# Perception devices and systems for monitoring humans and production process in Industry 4.0 scenarios

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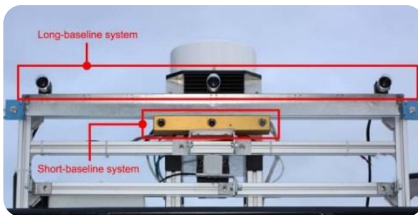
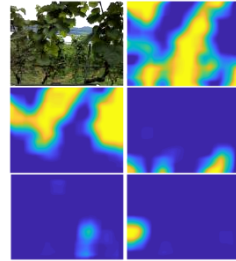
**Tutors:**

**Prof. Anna Gina Perri**

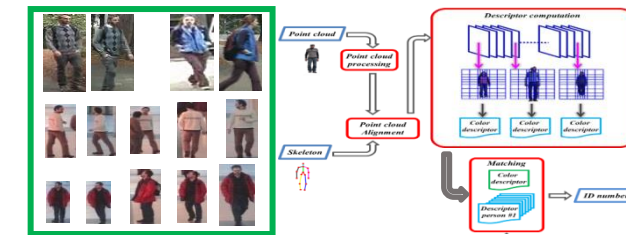
**Ing. Roberto Marani**

- Introduction
- System Overview
- Azure Kinect: Specifics and Analysis
- Setup Assessment:
  - Multi-vision System
  - Action Recognition
- Conclusion and Future Works

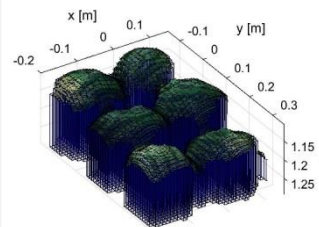
## Agricultural Robotics



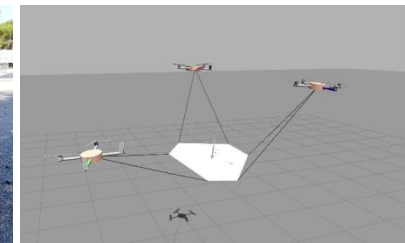
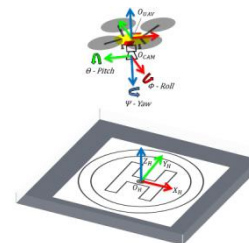
## Ambient Intelligence

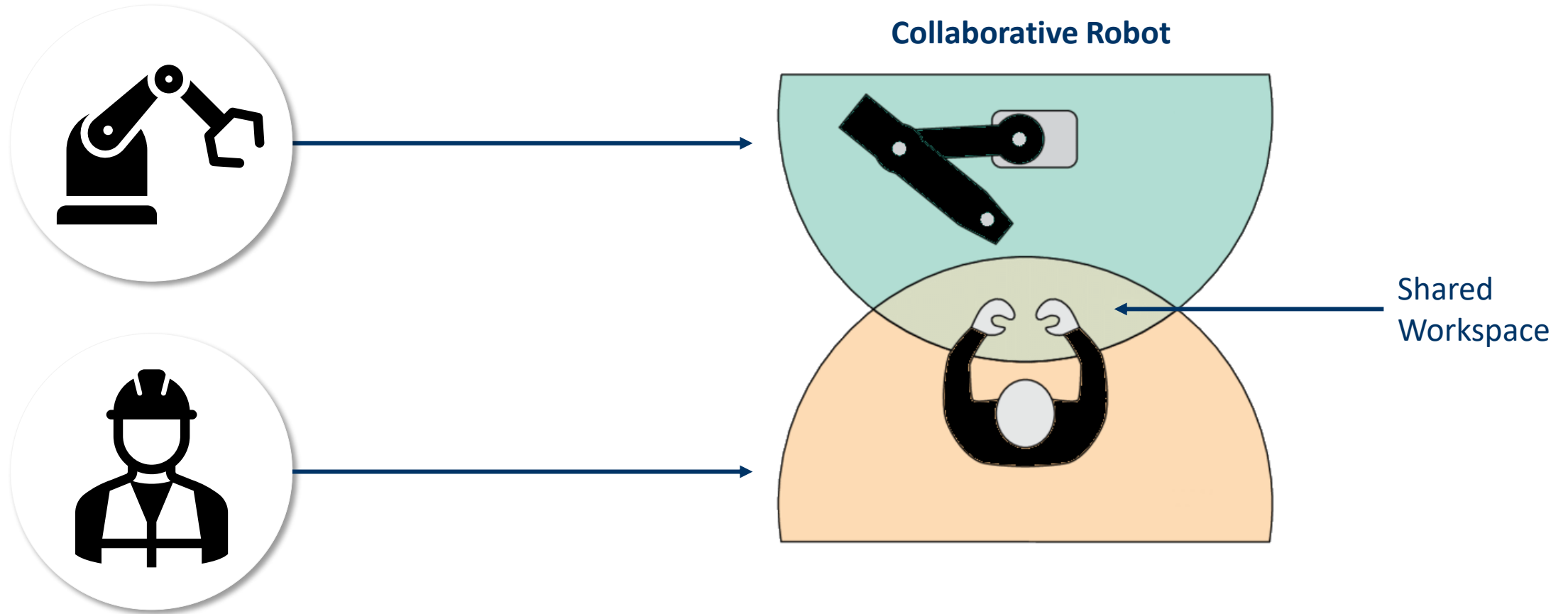


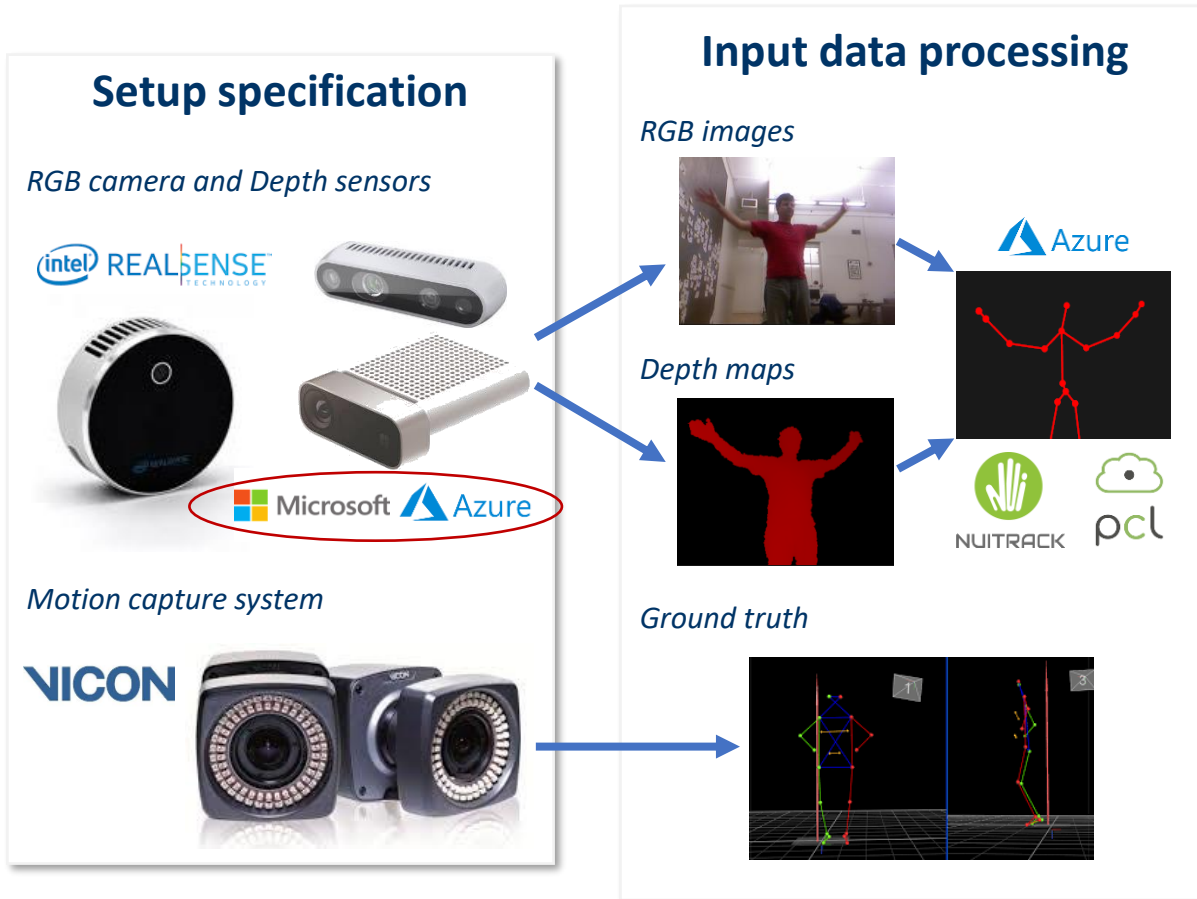
## Smart Retail



## Robot and Sensor Networks







## RGB camera:

- CMOS, CCD standard sensors

## Depth camera:

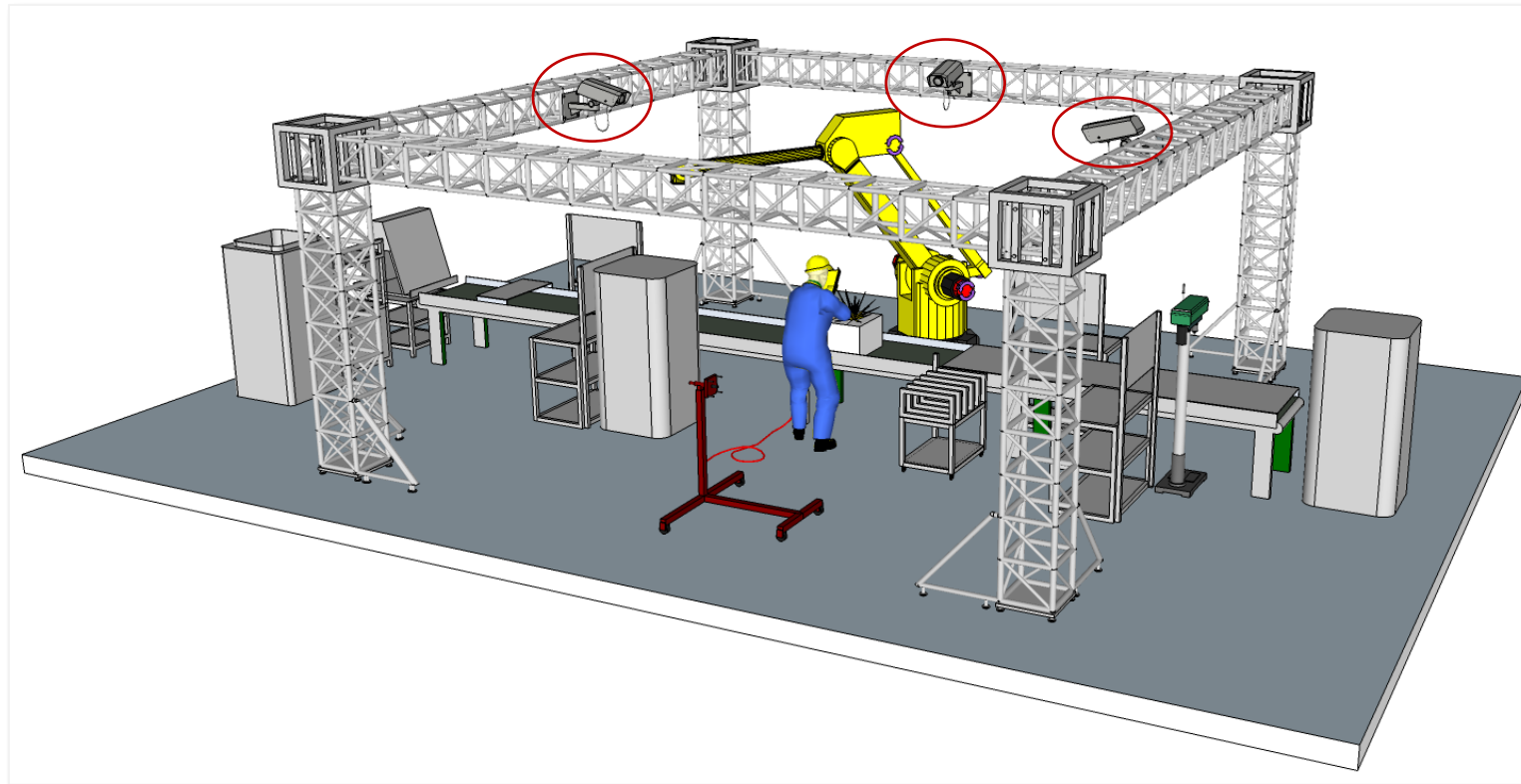
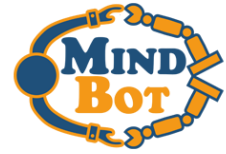
- Stereo
- Time of Flight (ToF)

## Skeleton extraction:

- VICON
- Azure Body Tracking
- Nutrack
- ...

## Aim:

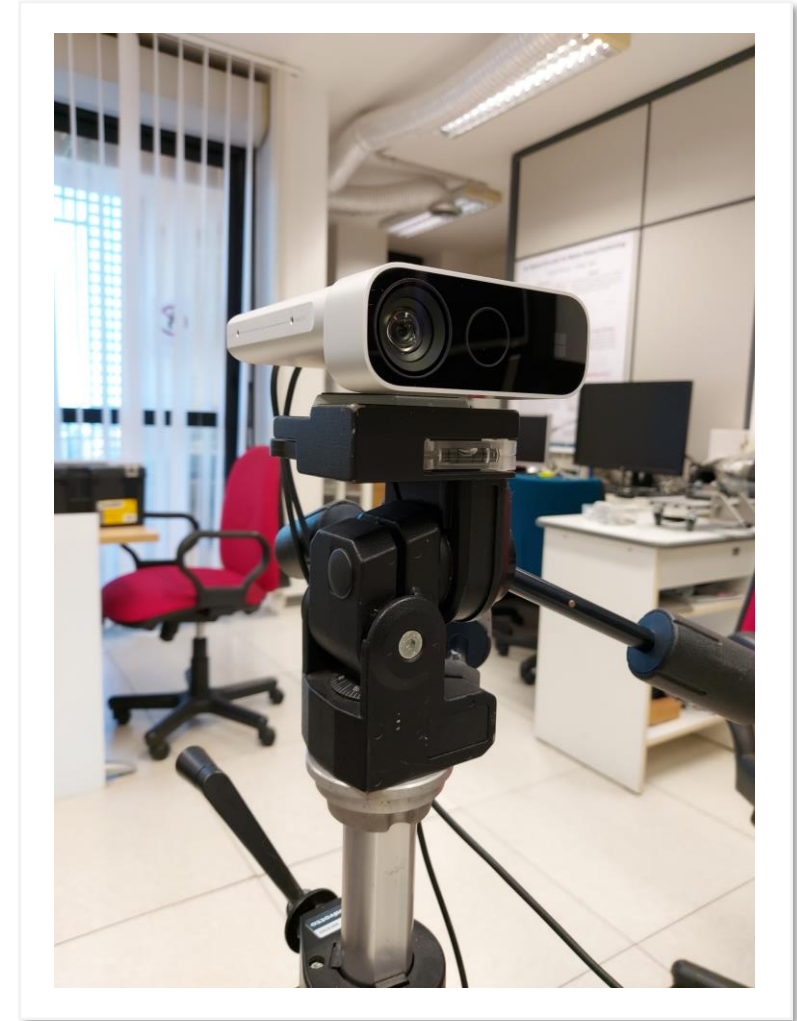
- Provide information about the dynamic occupancy (even predicting movements) of humans within the cell
- Establish a new way of communication between cobots and operators

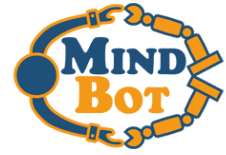


Romeo, L., Petitti, A., Marani, R., & Milella, A. (2020). "Internet of robotic things in smart domains: applications and challenges". *Sensors*, 20(12), 3355.

## Microsoft Azure Kinect SDK

- RGB camera with Depth sensor
- Spatial microphone
- Motion sensors (Gyroscope and Accelerometer)
- Advanced AI sensors that provide sophisticated computer vision

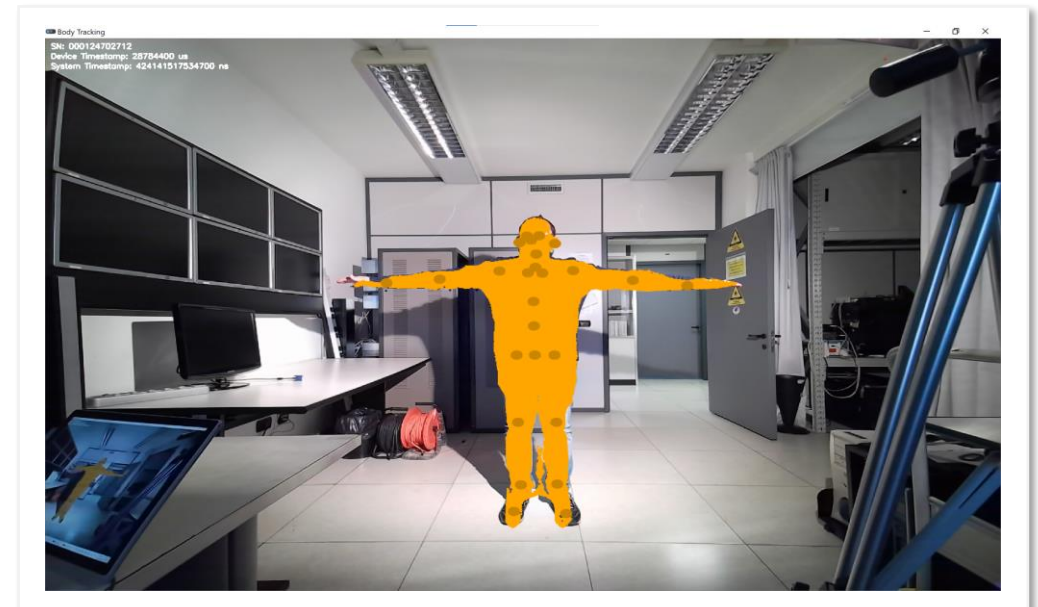




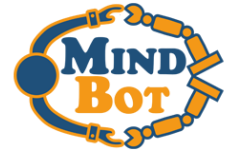
## Azure Kinect Body Tracking SDK

Tracking of bodies in 3D when used with the Azure Kinect DK hardware.

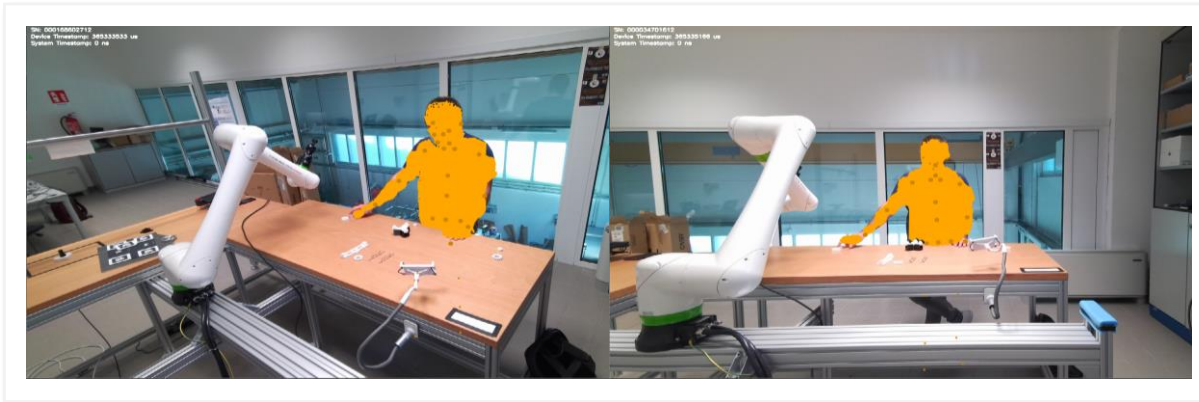
- Body segmentation
- Recognition of 32 skeletal joints for each body
- Extraction of anatomically correct skeleton for each partial or full body in FOV
- Unique identity for each body





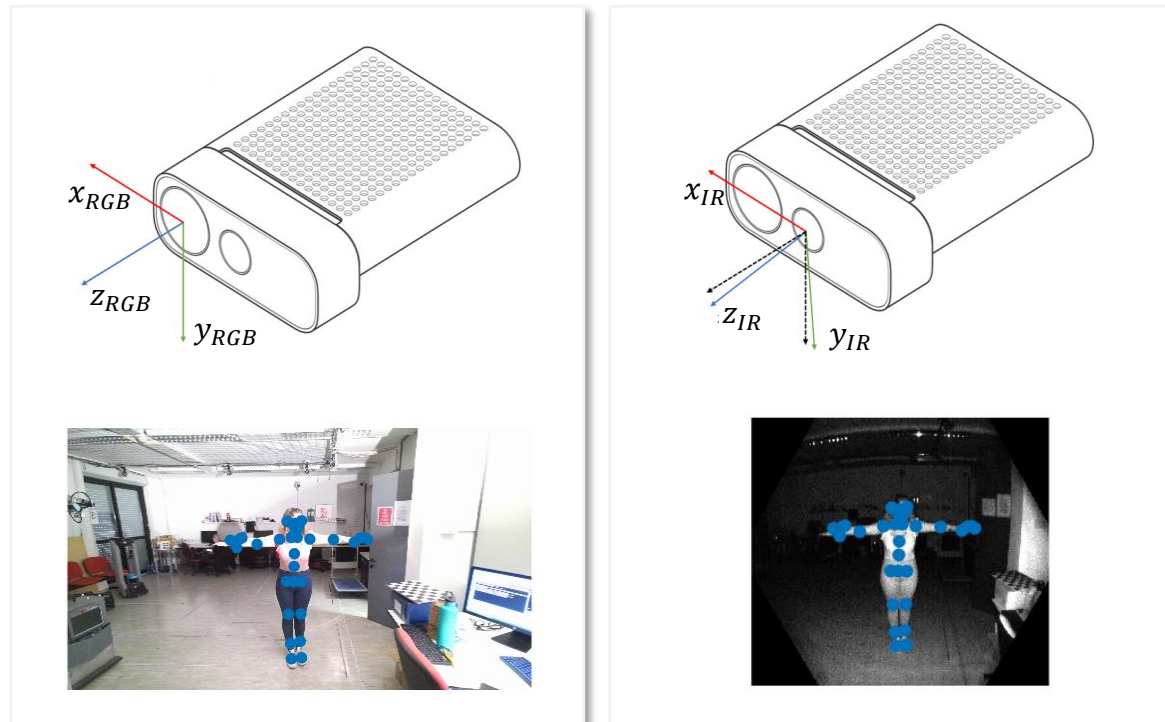


1. Azure Kinect analysis and performance evaluation
2. Multi vision system calibration
3. Image and Skeleton acquisition for action recognition

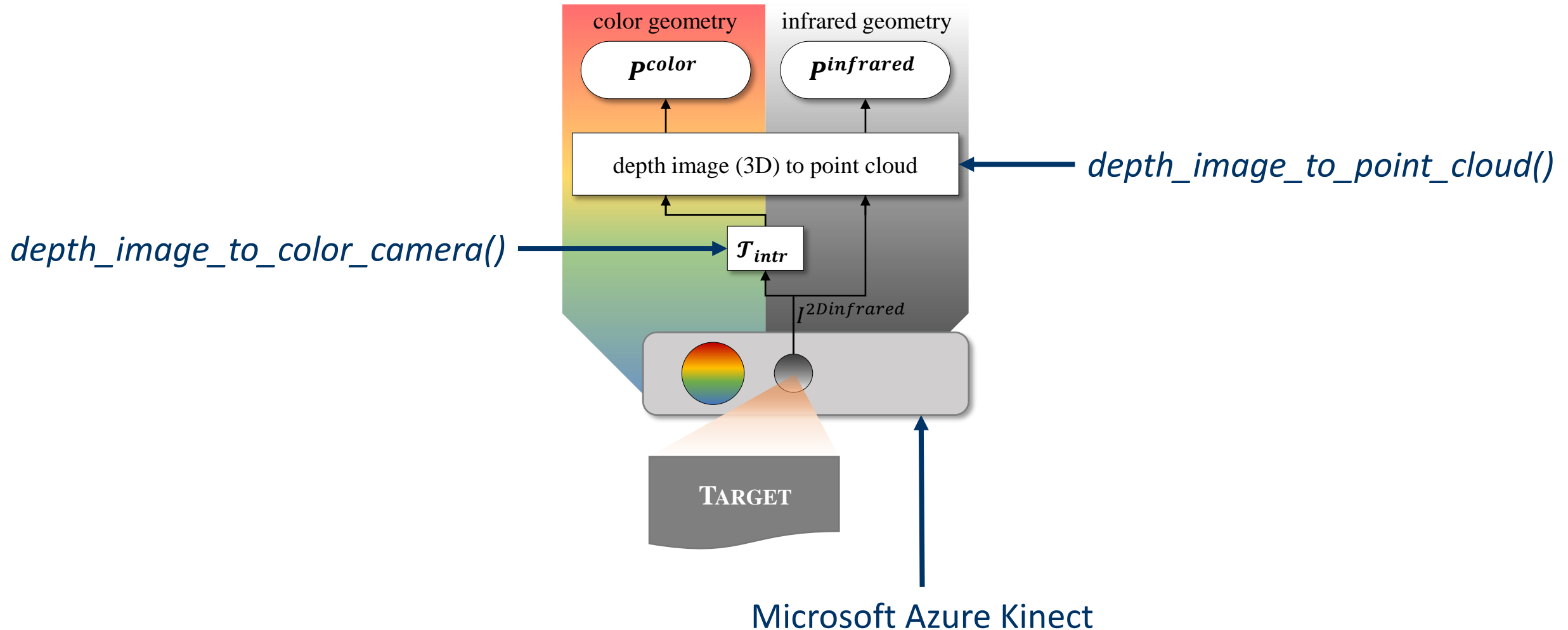


Skeletal joints coordinates **extracted** using the IR sensor.

- *k4a\_calibration\_3d\_to\_2d()*: converts the 3D coordinates in 2D pixel coordinates
- *k4a\_transformation\_depth\_image\_to\_color\_camera()*: converts the 3D coordinates from depth resolution to color resolution
- *k4a\_calibration\_2d\_to\_2d()*: converts the 2D pixel coordinates from depth resolution to color resolution, and vice-versa



## Point Cloud realization

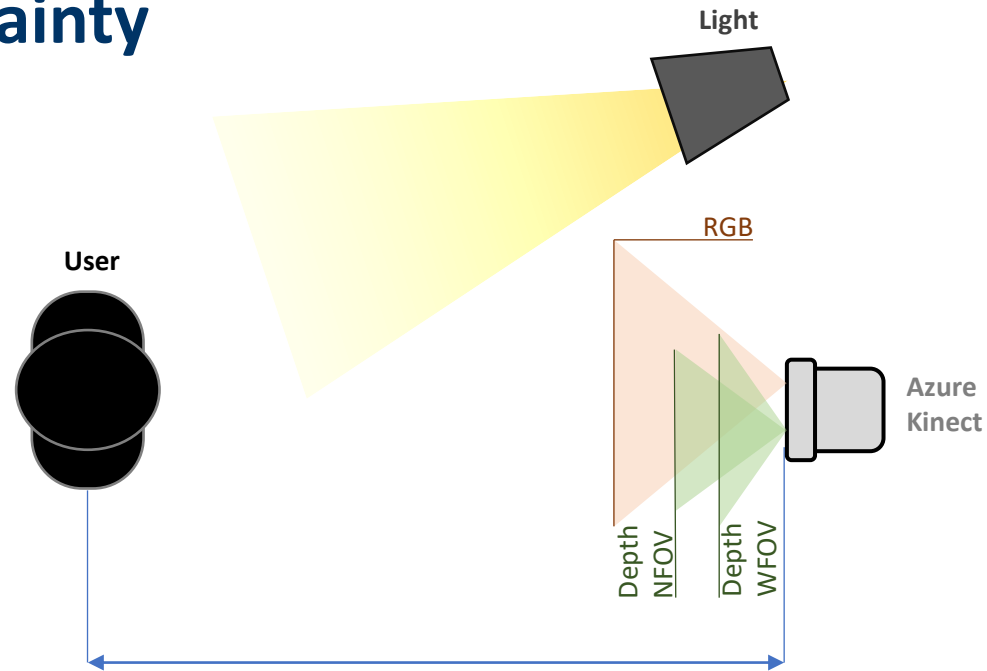


## Point Cloud realization

- Elaboration from the IR sensor
- Set of 3D coordinate points in a specific coordinate system
- Color, depth, intensity information for each point



## Skeletal joints uncertainty



Standard Deviation computed by changing:

*Infrared Resolution*

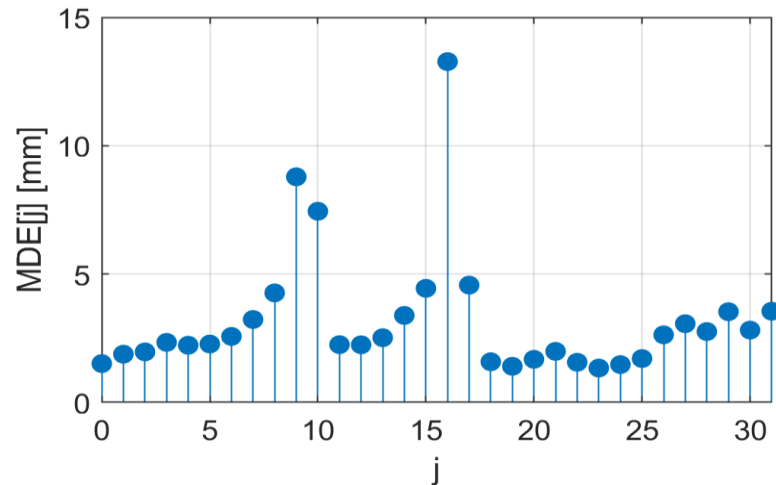
*Ambient light conditions*

*Body occlusions*

*Subject-camera distance*

Example of MDE (Mean Distance Error) of the 32 joints computed by the Azure Kinect and its body-tracking library with the following configuration:

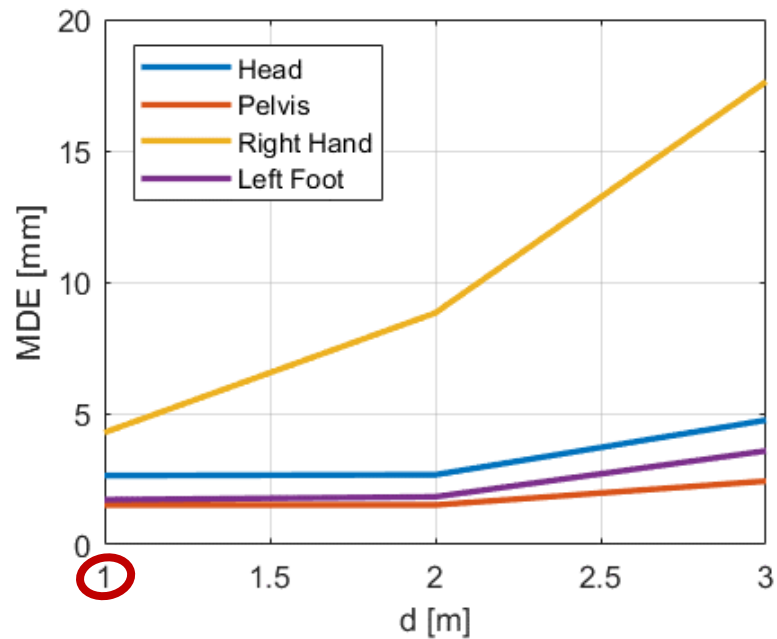
- Narrow depth resolution
- No Occlusions
- Lights at 1750 Lux (direct illumination)
- User at 1-meter distance



Four representative Joints:

- Head joint (j = 26)
- Pelvis joint (j = 0)
- Left-hand joint (j = 8)
- Right-foot joint (j = 25)

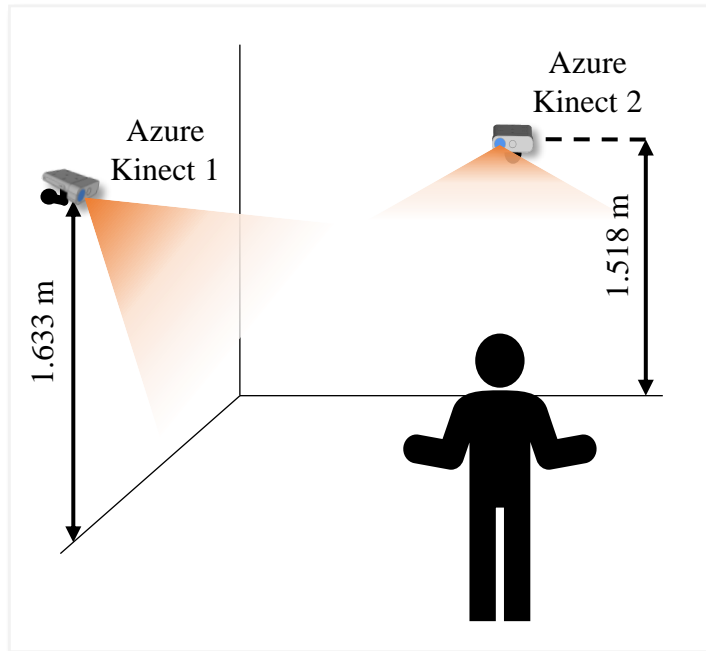
## Best configuration



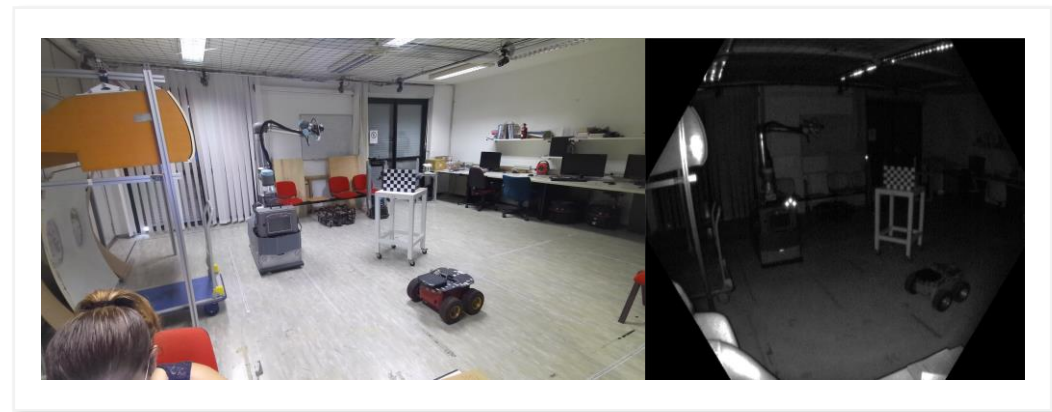
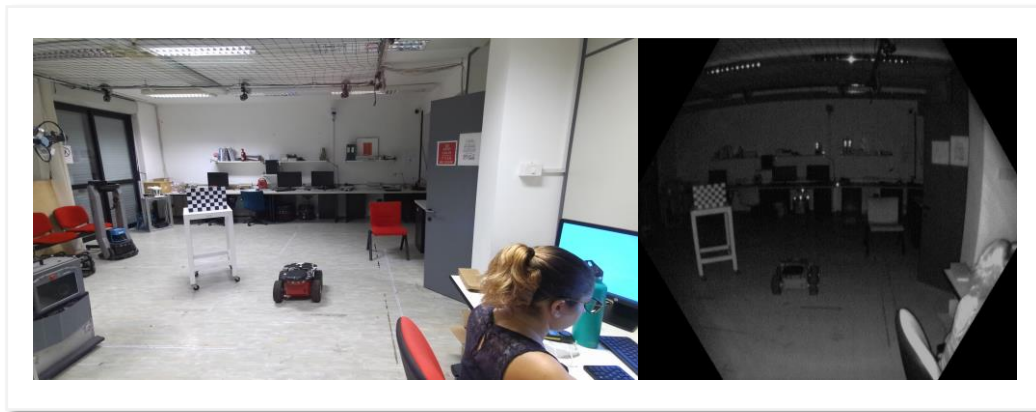
- Narrow Resolution
- No Occlusions
- 1750 lux
- 1 meter distance

- MDEs increase as the user-camera distance grows
- Wide Resolution increases the MDE, which is about tripled of the case with Narrow Resolution
- Direct illumination of 1750 lux generally downs the performance of the body tracking
- The joints show higher values of uncertainty when the skeleton is partially occluded
- Left-hand joint shows the worst MDEs under all testing conditions

$1 \text{ mm} < \text{MDE} < 53 \text{ mm}$   
MEAN = 8 mm  
STD = 6 mm



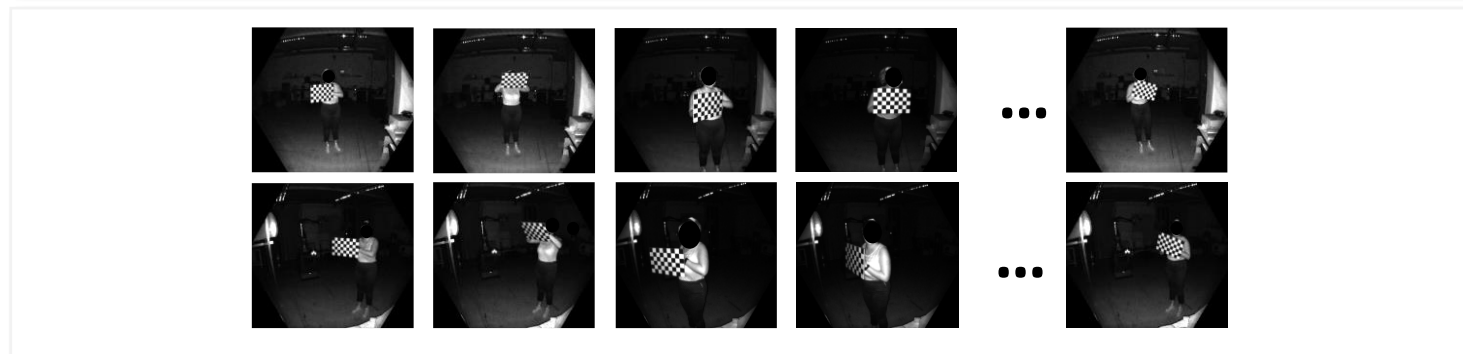
- Vision system of two cameras
- Calibration assessed depending on the use-case:
  - ✓ Skeletal joints
  - ✓ Point Cloud with color geometry
  - ✓ Point Cloud with infrared geometry

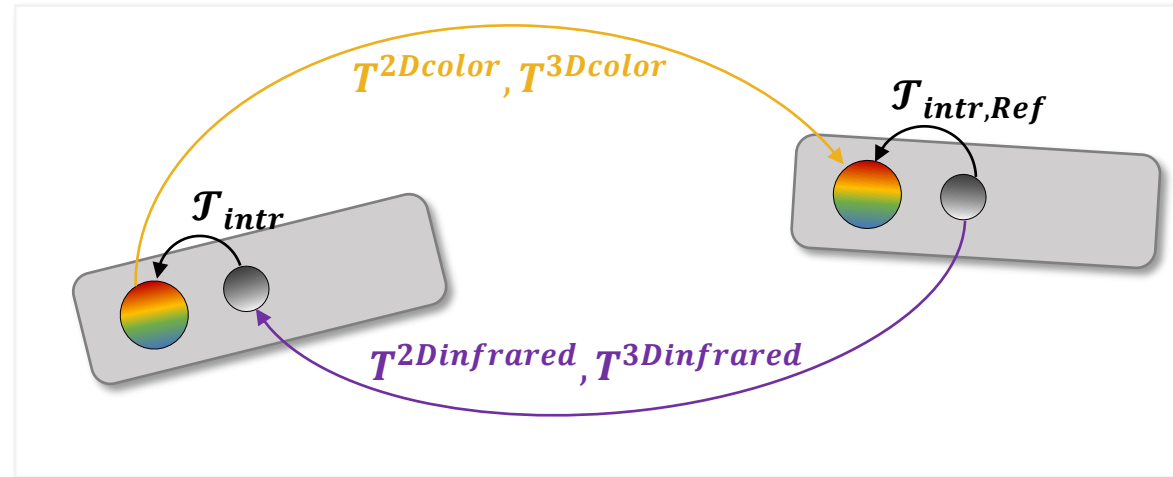




## Camera Calibration

- Interface between 3D world and 2D camera images
- Minimization of metric distance inaccuracy
- Methodologies based on a feature target
- Chessboard acquisition:
  - ✓ black and white squares
  - ✓ frames with different positions and orientations
  - ✓ RGB and IR images





- **2D calibration (RGB and IR separately):**
  - ✓ Chessboard frames from both cameras
  - ✓ Matching chessboard square corners among frames
  - ✓ Transformation matrix as output

- **3D calibration (RGB and IR separately):**
  - ✓ Chessboard frames from both cameras
  - ✓ SDK function for 3D projection
  - ✓ 3D geometric transformation estimation
  - ✓ Transformation matrix as output

Romeo, L., Marani, R., Perri, A.G., & D'Orazio, T. "Multiple Azure Kinect Calibration for three-dimensional dense Point Clouds and reliable Skeletons". *IEEE Transaction on Instrumentation and Measurement*. (to be submitted)

## Point Cloud Calibration (RGB)

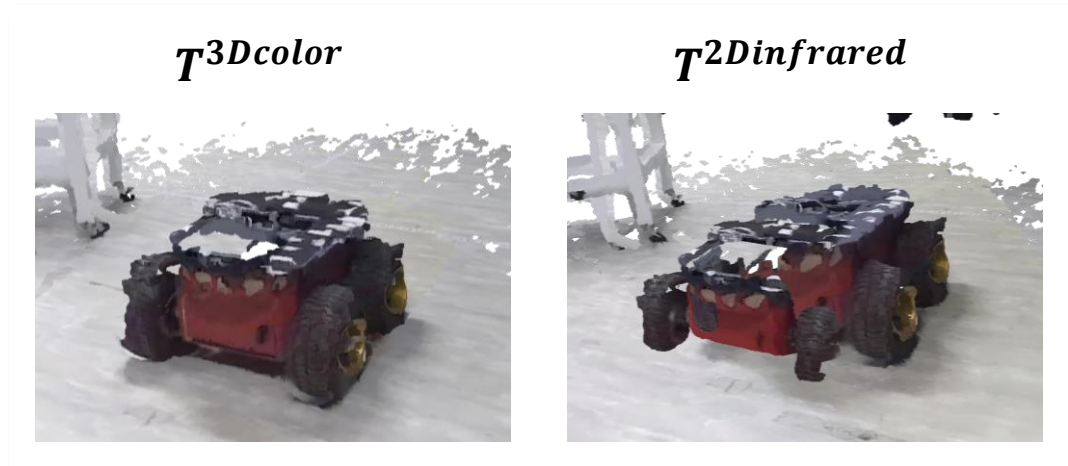
- Best result: 3D calibration with RGB images
- Worst result: 2D calibration with IR images

### Pros:

- Large set of coordinates
- High point cloud density

### Cons:

- Low coordinate precision
- Massive data storage



## Point Cloud Calibration (IR)

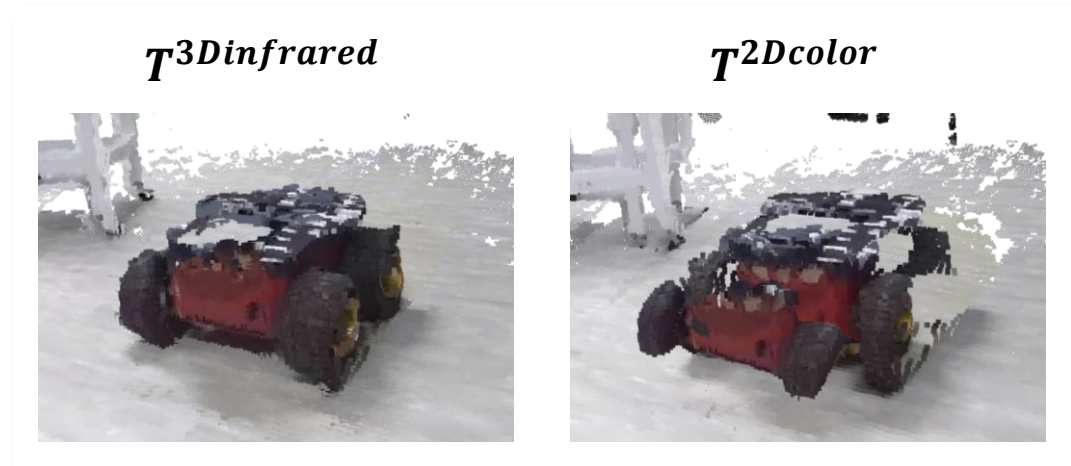
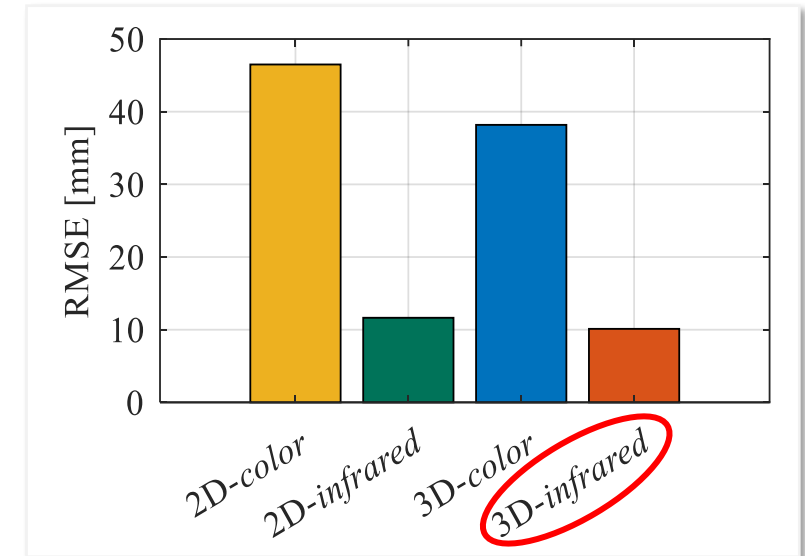
- Best result: 3D calibration with IR images
- Worst result: 2D calibration with RGB images

### Pros:

- High coordinate precision
- Low data storage

### Cons:

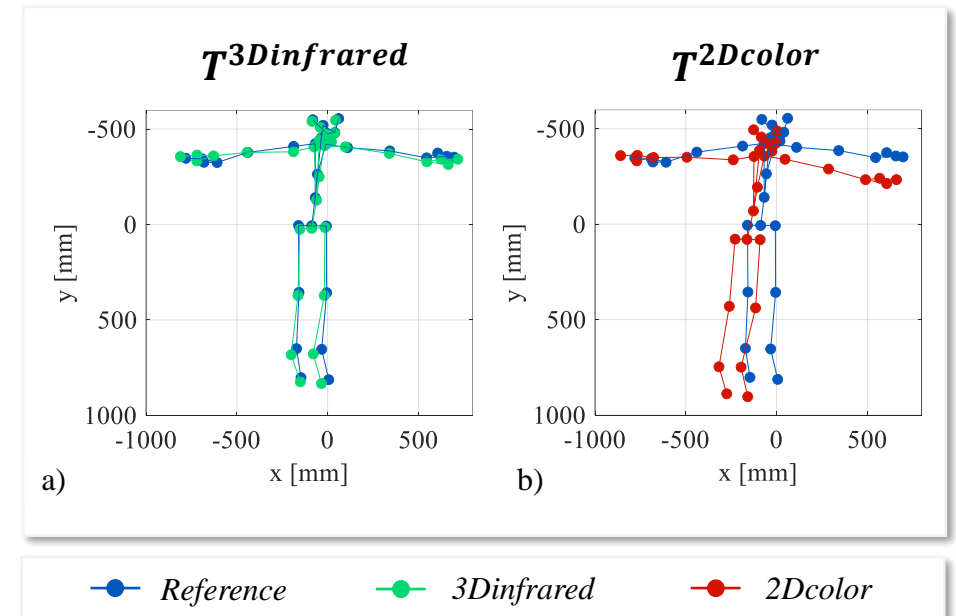
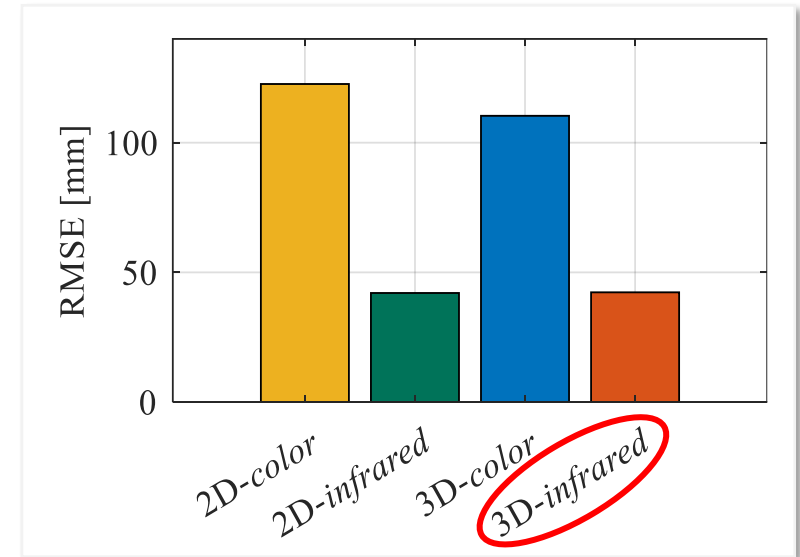
- Small set of coordinates
- Low point cloud density



## Skeleton Calibration (IR)

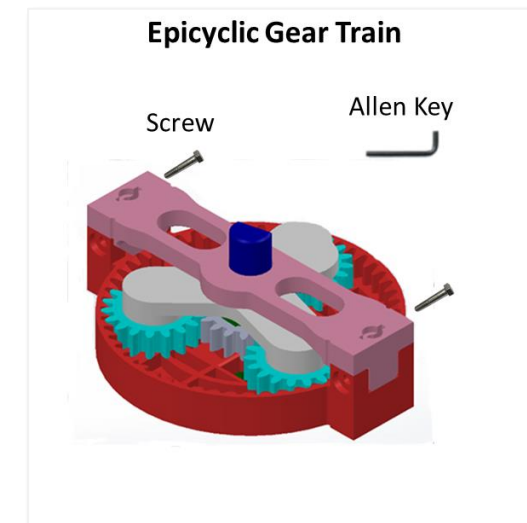
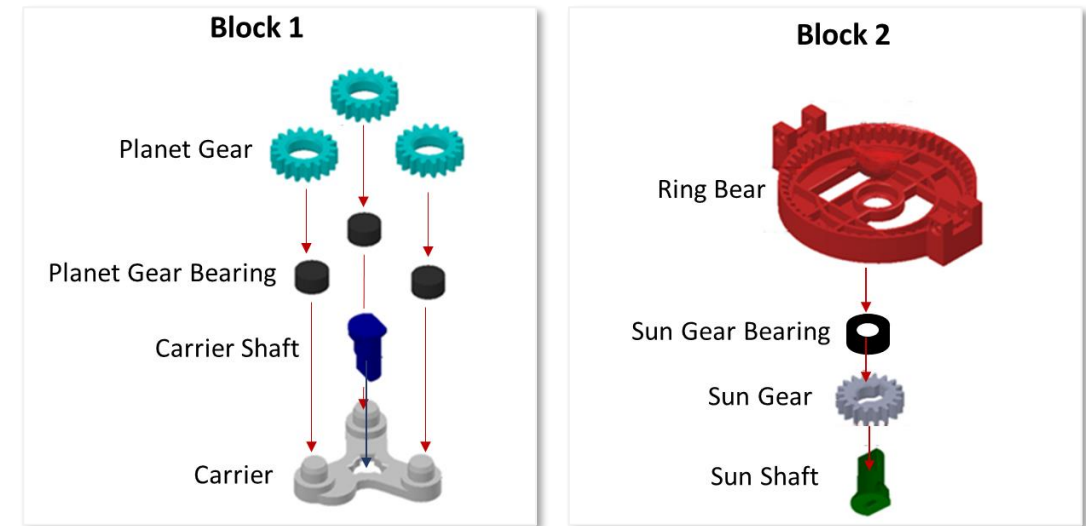
- Best result: 3D calibration with IR images
- Worst result: 2D calibration with RGB images

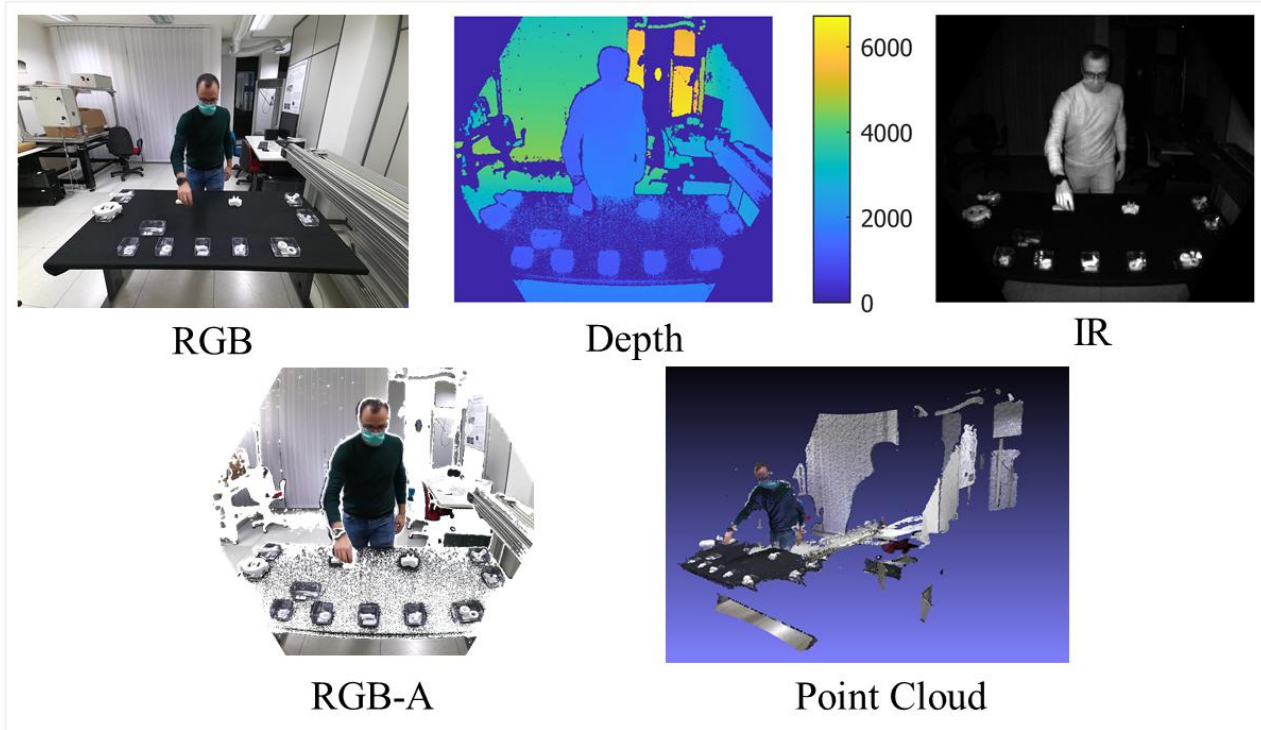
- Results highly comparable with IR point cloud calibration
- Uncertainty due to the pseudo-random error from Azure Kinect Body Tracking SDK



## Epicyclic Gear Train (EGT)

- Visual recognition of human performing a specific task
- Epicyclic Gear Train to be assembled by the operator
- Action recognition information sent to the collaborative robot

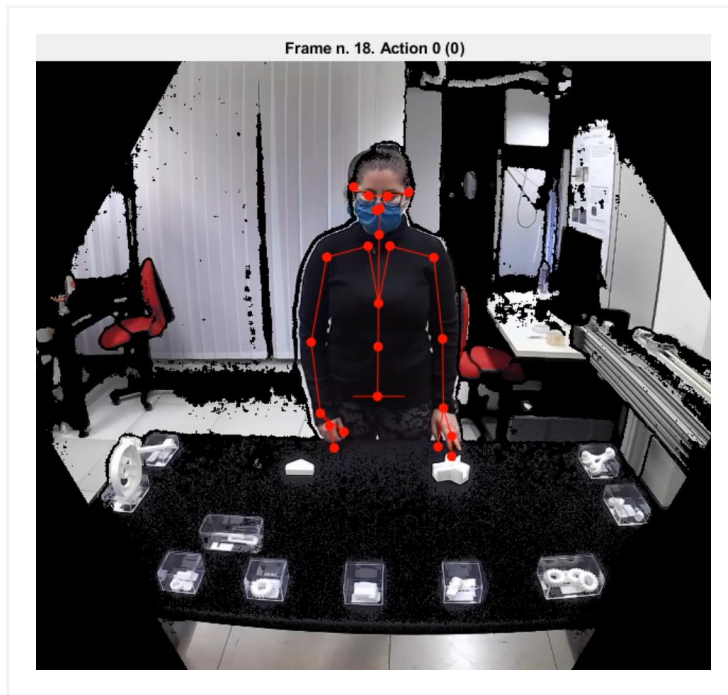




- Creation of a dataset for action recognition
- Recording of the assembling process using an Azure Kinect
- Dataset composed of:
  - ✓ RGB images
  - ✓ IR images
  - ✓ Depth images
  - ✓ Point Cloud with IR resolution
  - ✓ RGB images in IR environment
  - ✓ Skeleton information

# Setup assessment: action recognition

- Image Segmentation
- Analysis of the dataset
- Extraction of relevant information
- Deep learning implementation (CNN, GCNN, LSTM, ...)



	Quantity	Components	Class Action
<b>Block 1</b>	1	Carrier	1
	3	Planet Gear Bearing	2
	3	Planet Gear	3
	1	Carrier Shaft	4
<b>Block 2</b>	1	Sun Shaft	5
	1	Sun Gear	6
	1	Sun Gear Bearing	7
	1	Ring Bear	8
<b>Block 3</b>	1	Block 2 on Block 1	9
	1	Ring Shaft	10
	2	Screws	11
	1	Allen Key	12



- Implementation of a monitoring system in a real-case scenario
- Analysis of video devices and their uncertainties
- Calibration methodologies for the estimation of 3D coordinates in a multi-vision system
- Action Segmentation and Recognition on a real Dataset

## Future Works:

- Development of the most suitable neural network for the segmented Dataset, for Action Recognition
- Extraction of the most relevant feature to better exploit the created dataset
- Implementation of the Action Recognition system also using a second camera, in a calibrated setup



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per il Manifatturiero Avanzato  
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di Bari



# Thanks for the attention!

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**Ing. Roberto Marani**

DEI Doctoral Research Seminars

Bari, 23/02/2022