

Improving Real-Time Omnidirectional 3D Multi-Person Human Pose Estimation With People Matching and Unsupervised 2D-3D Lifting

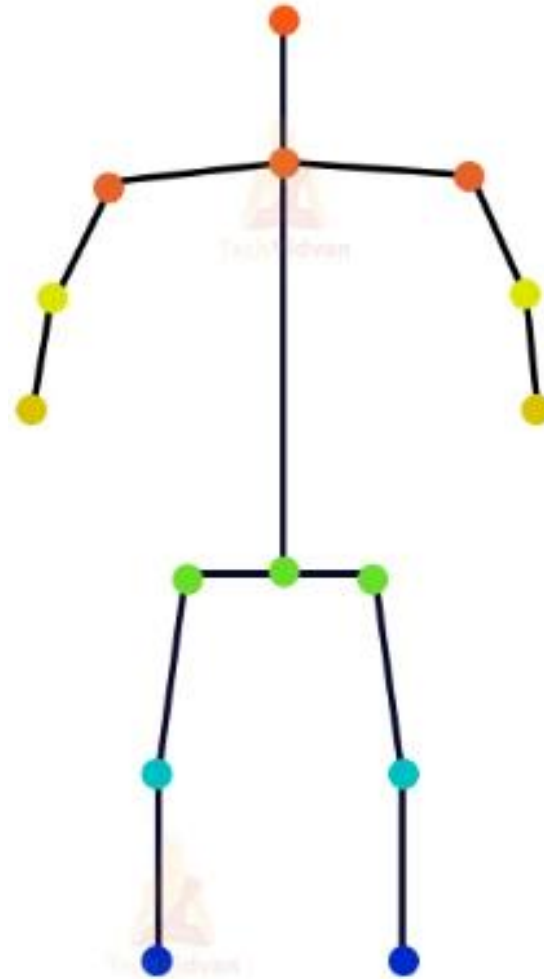
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1. University of Southampton, School of Electronics & Computer Science

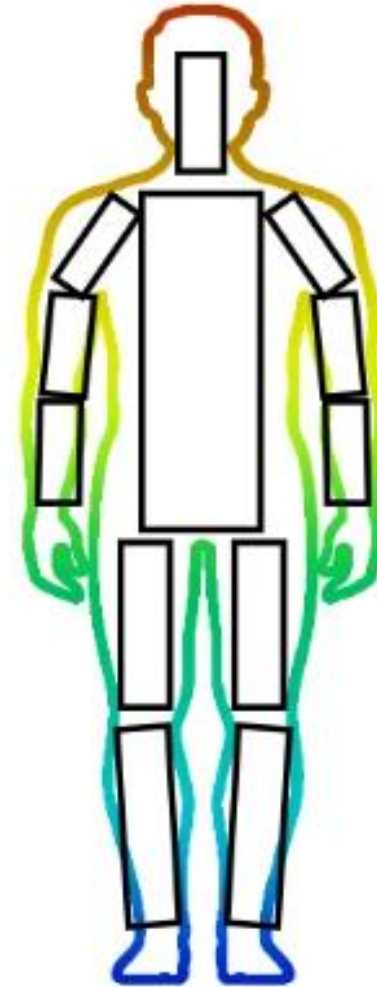
2. Korea Institute of Science and Technology

Human Pose Estimation

- Computer Vision tasks to identify and classify the poses of human body parts and joints in images or videos.
- Applications in healthcare, entertainment, surveillance, sports, education, and beyond



(a) Kinematic



(b) Planar



(c) Volumetric



2D Human Pose Estimation

- Locations of the body joints in 2D space
- Bottom-up, and top-down methods
- Classic approaches OpenPose and DeepPose

OpenPose: Realtime Multi-Person 2D Pose Estimation using Part Affinity Fields, Cao et al., 2019

DeepPose: Human Pose Estimation via Deep Neural Networks, Toshev & Szegedy, 2014

3D Human Pose Estimation

- Locations of the body joints in 3D space
- Classic approaches: mmMesh, XNect

Video Frame



Ground Truth

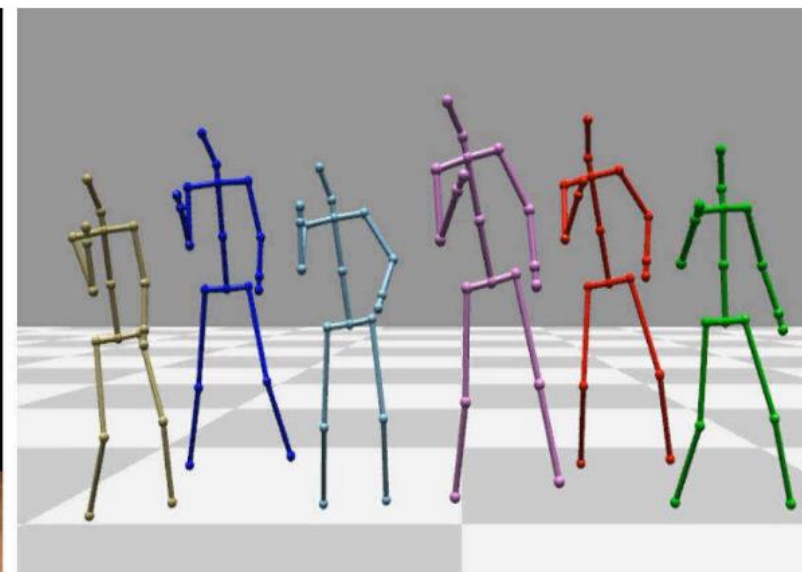


mmMesh



mmMesh: Towards 3D Real-Time Dynamic Human Mesh Construction Using Millimeter-Wave, Xue et al., 2021

XNect: Real-time Multi-Person 3D Motion Capture with a Single RGB Camera, Mehta et al., 2020



Sensors for Human Pose Estimation

- Monocular camera: Cheap, yet with limitations like occlusion and depth ambiguity.
- Set of cameras: Addresses monocular camera limitations but is costly and lacks transferability, restricting applications.
- RADARs: Effective in occlusion handling but produces sparse data.
- LIDARs: High-resolution output, but expensive with scarce data.
- IR-based sensors (Kinect): Faces challenges outdoors.
- Motion Capture Sensor Systems: Limited in applications



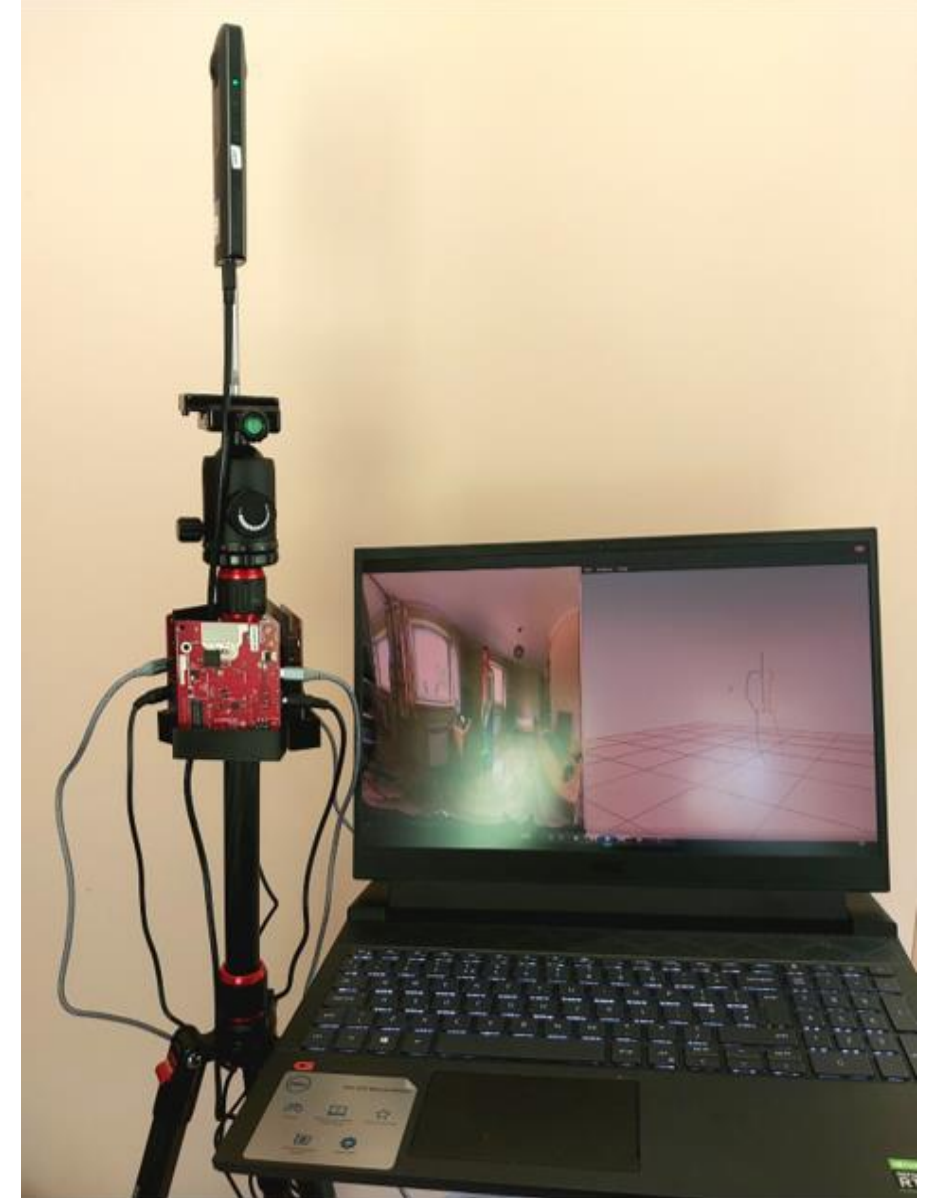
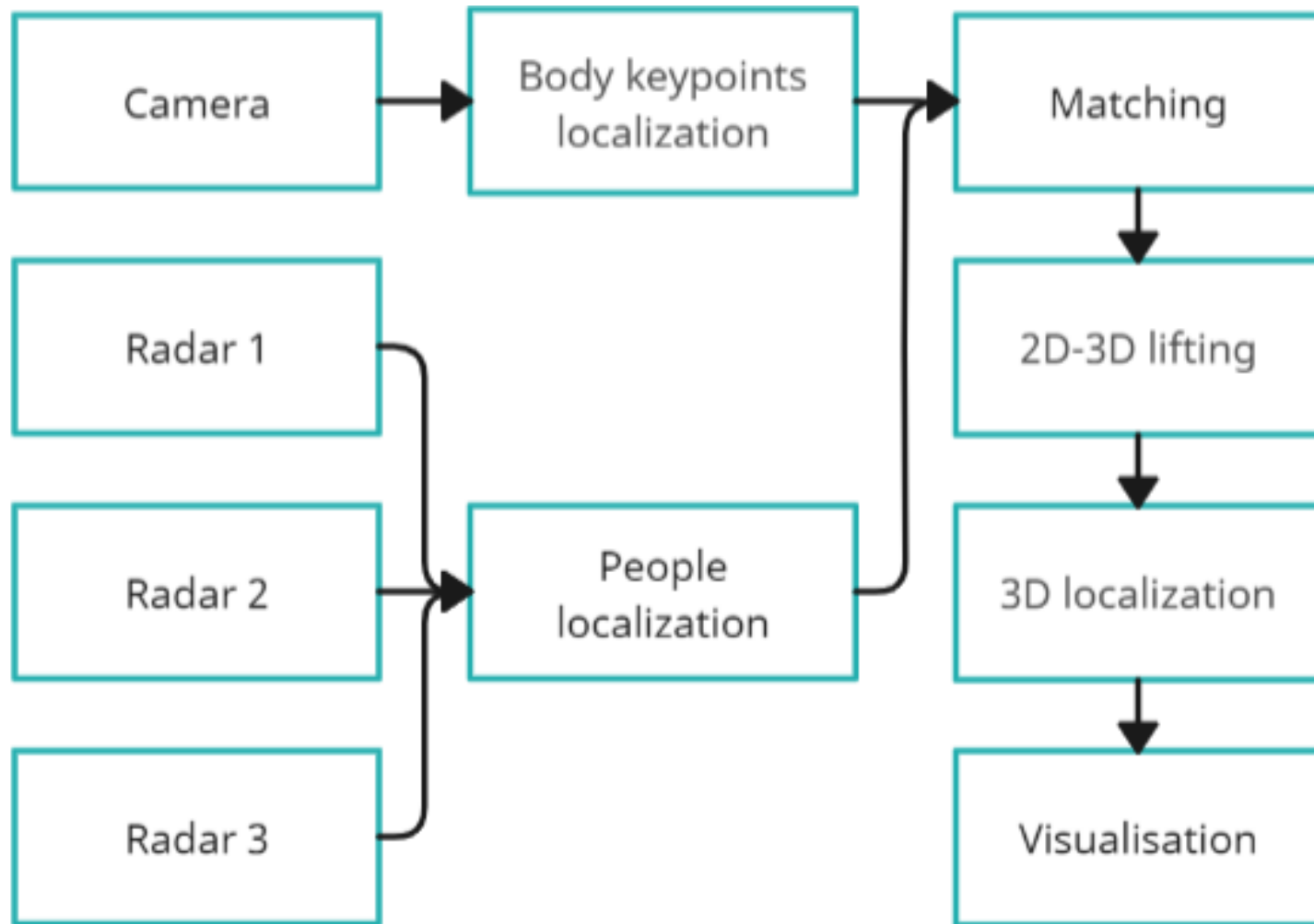
Human Pose Estimation – major problems

- Majority of solutions focus on single-person 3D HPE
- Multi-person solutions have limited range and limited occlusion handling capabilities [Mehta 2020, Carraro 2019]

Contributions

- Multi-person 3D pose estimation
- Omnidirectional real-time HPE systems for real-world localization – applications in robotics
- New design for systems using blend of sensors
- Cheap and robust real-time, multi-person system

System Overview



Keypoints detection

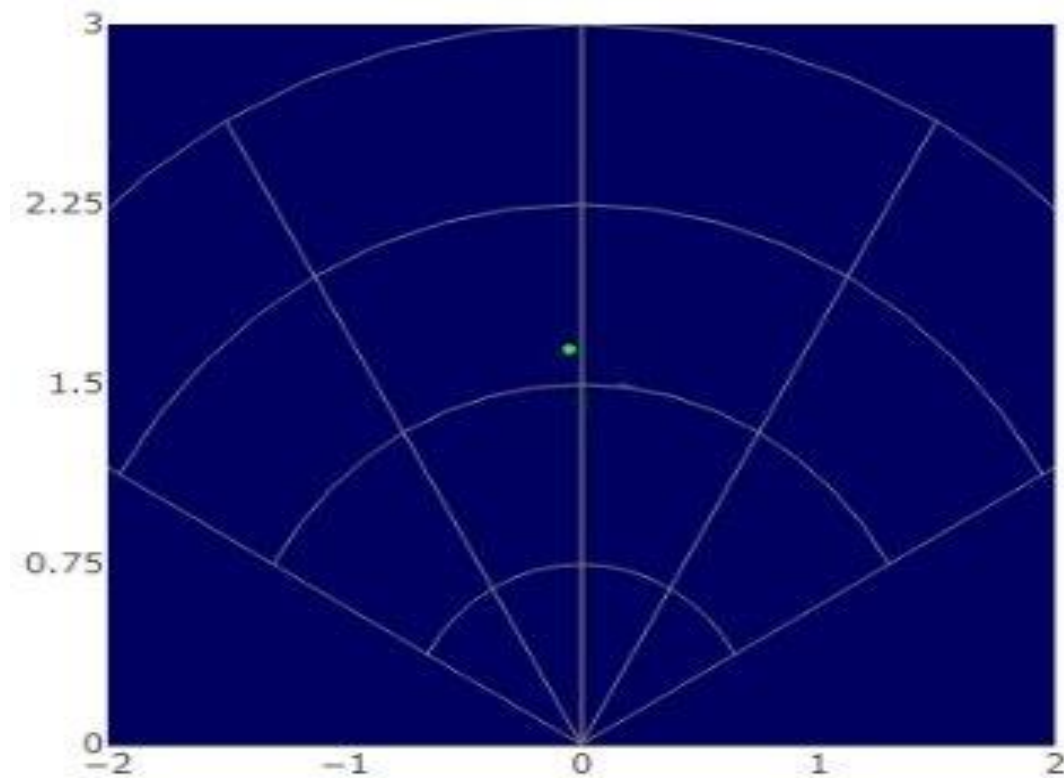
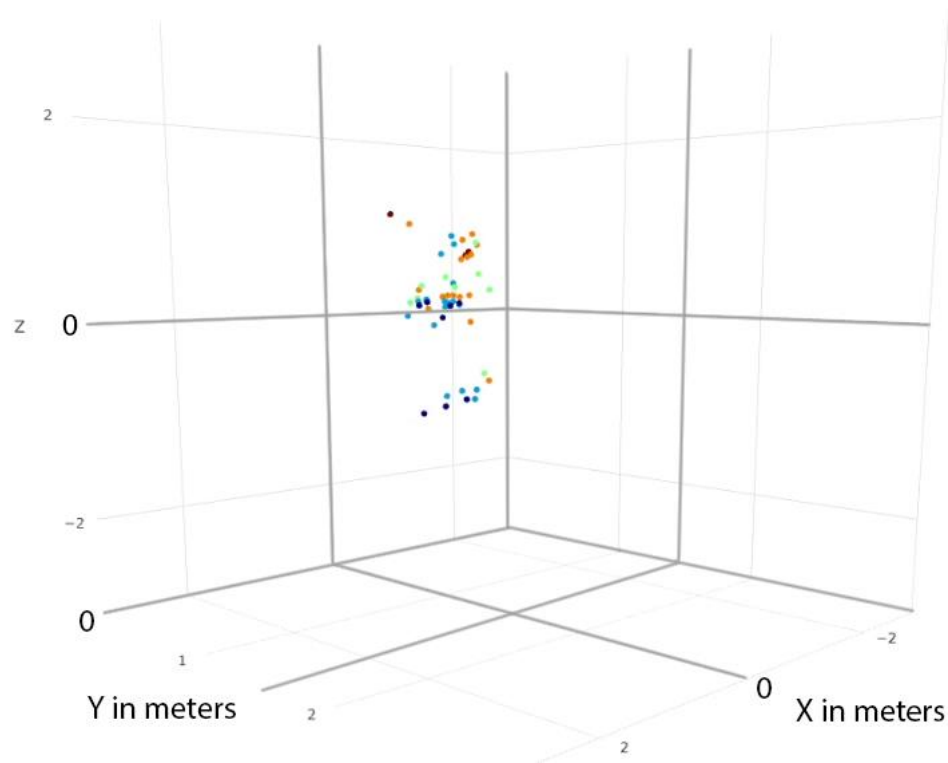
- Openpose for 2D keypoints detection
- It uses Part Affinity Fields
- Handle multiple-people
- No tracking. Detection happens in each frame



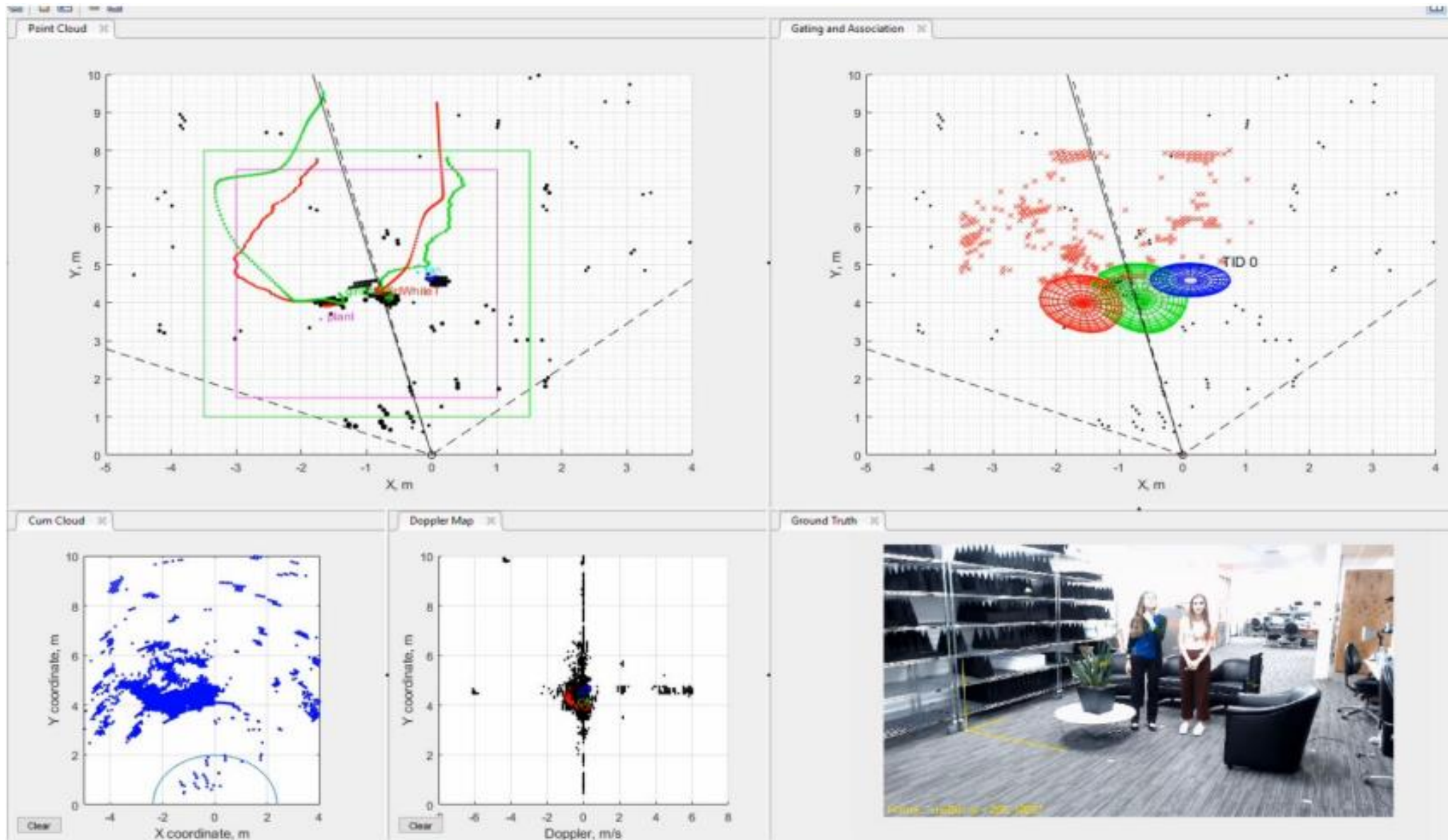
Z. Cao et al. "Openpose: Realtime multi-person 2d pose estimation using part affinity fields", 2019.

People counting algorithm

- People localisation done with people tracking algorithm from Texas Instruments (Garcia 2019)



People counting algorithm



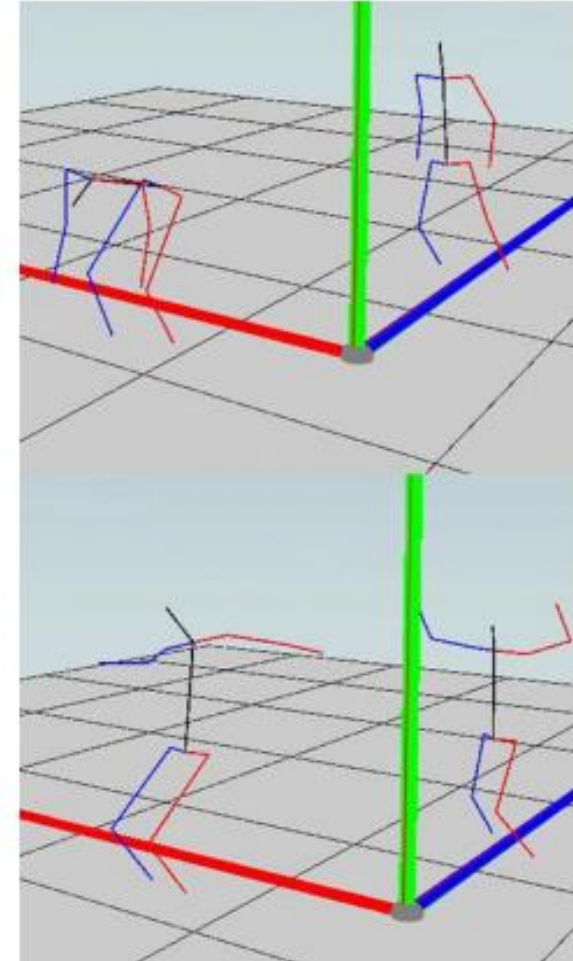
Matching

- Binary search tree
- Based on disparity between the average image x coordinate, and the radars coordinates transformed into the image coordinate space through a learned transform
- Transform based on Oh et al. - learned by Levenberg–Marquardt algorithm
- Radar data used to move poses to correct position in 3D space

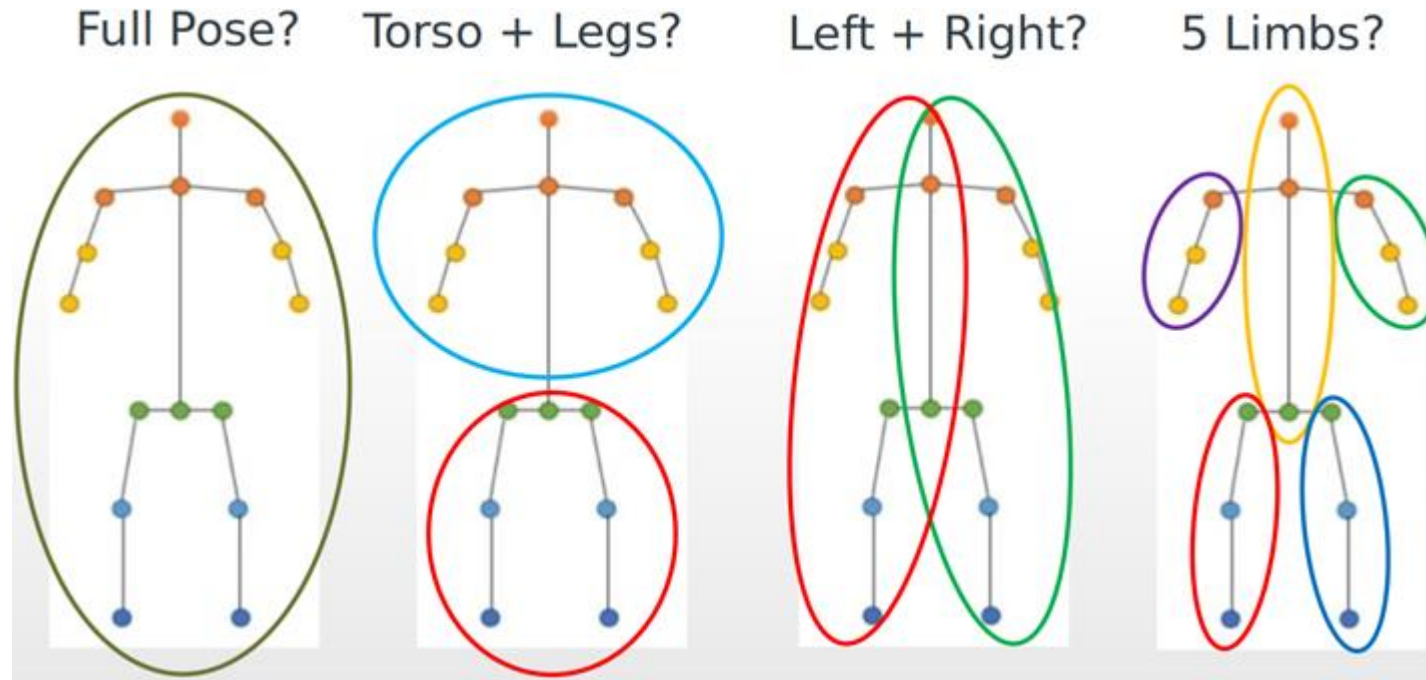
Input image



Reconstructed poses



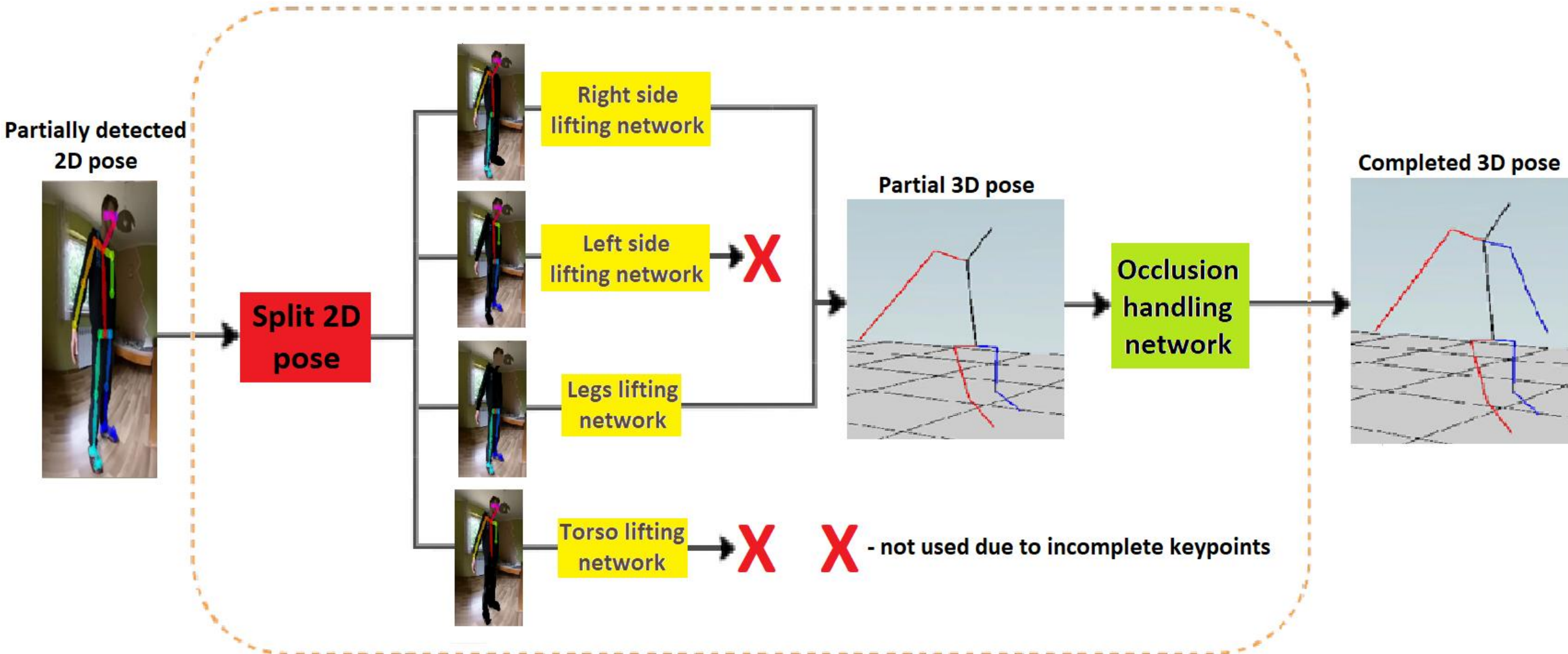
Overview of the lifting algorithm



| Pose Detected | Front Cameras | Rear Cameras | Avg. |
|-----------------|---------------|--------------|-------|
| Full 2D Pose | 54.8% | 35.5% | 45.1% |
| Partial 2D Pose | 92.4% | 84.8% | 88.6% |

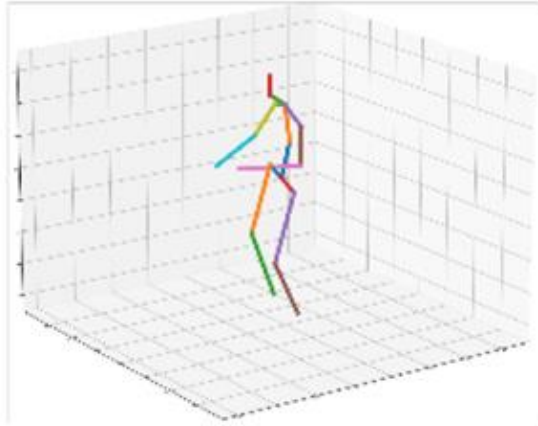
Overview of the lifting algorithm

LiNKs approach

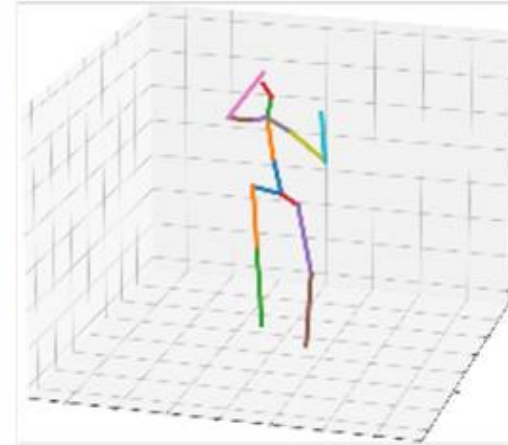
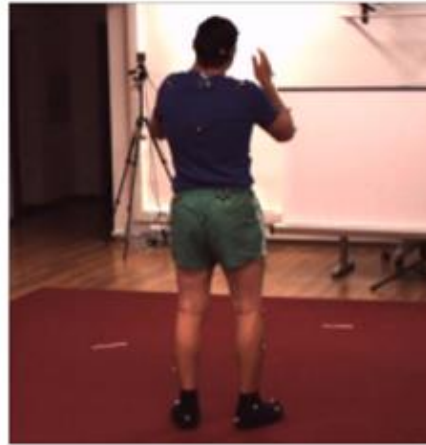


Qualitative Results of lifting algorithm

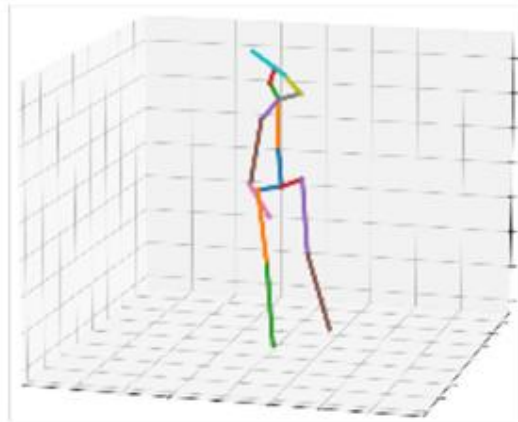
Right Arm Occluded



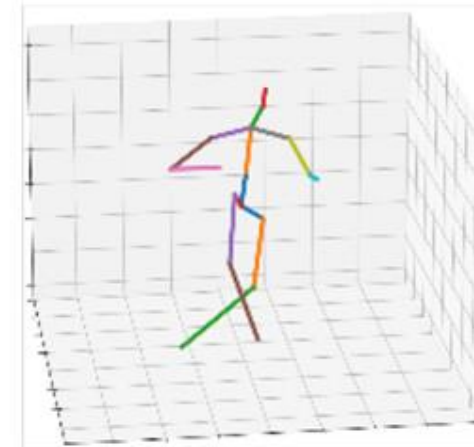
Left Arm Occluded



Right Leg Occluded



Left Leg Occluded

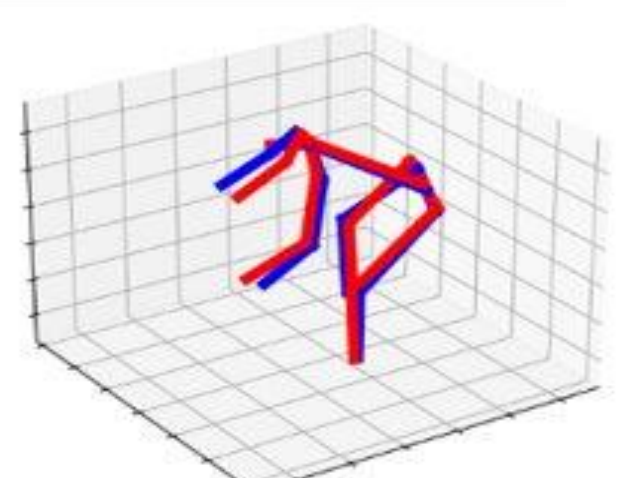
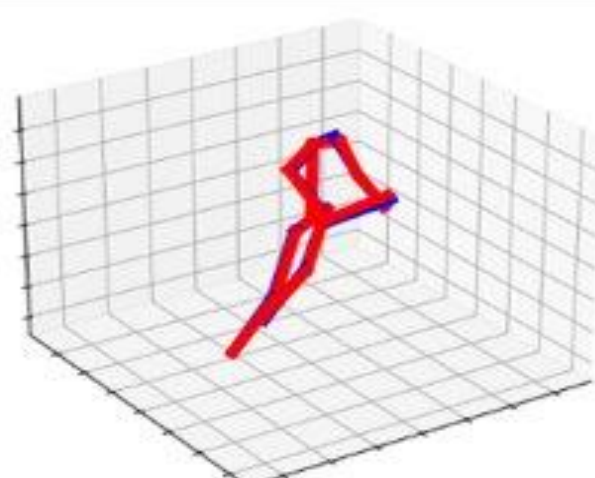
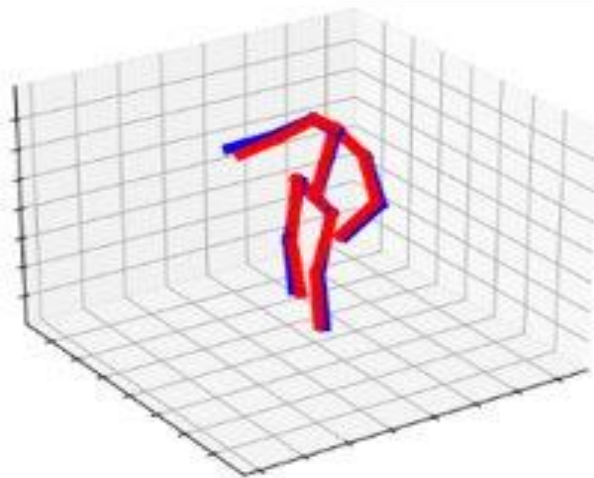
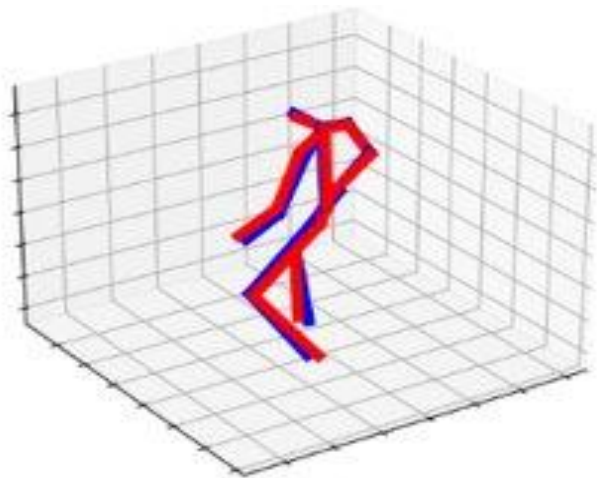


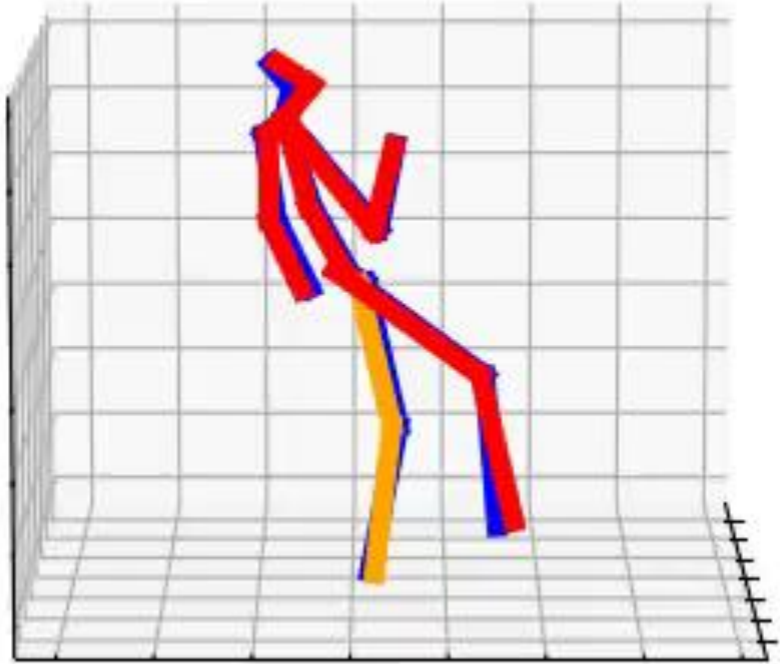
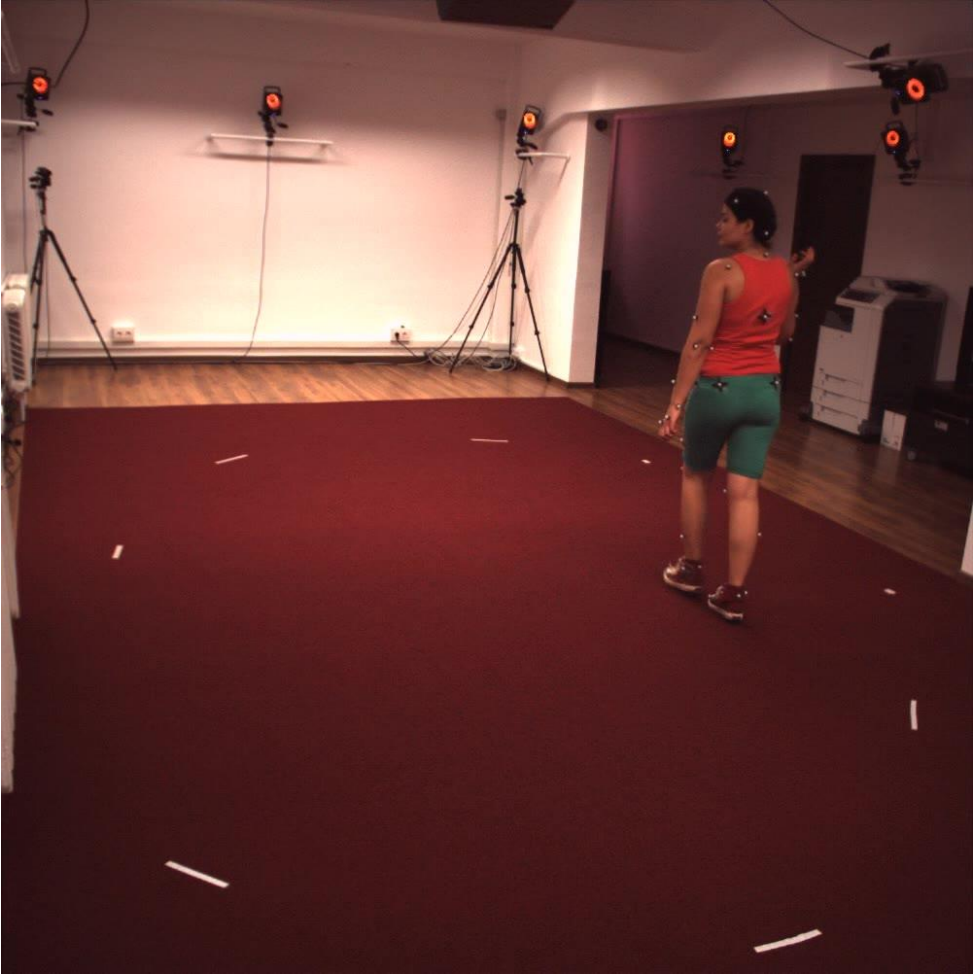
Pose reconstruction errors on the Human3.6M

TABLE III

- The Procrustes aligned mean per-joint position error (PA-MPJPE)
- Normalized mean per joint position error (N-MPJPE)
- The smaller, the better for both

| Method | Occlusion | PA-MPJPE | N-MPJPE |
|-------------------|-----------------|----------|---------|
| LInKs [2] | None | 33.8 | 61.6 |
| Ours (Recreation) | None | 37.2 | 61.7 |
| Ours (Recreation) | Left Arm | 52.1 | 78.1 |
| Ours (Recreation) | Left Leg | 46.0 | 73.2 |
| Ours (Recreation) | Right Arm | 49.8 | 75.7 |
| Ours (Recreation) | Right Leg | 44.5 | 71.6 |
| Ours (Recreation) | Left Arm & Leg | 62.0 | 86.0 |
| Ours (Recreation) | Right Arm & Leg | 60.2 | 83.7 |
| Ours (Recreation) | Both Legs | 69.3 | 99.8 |
| Ours (Recreation) | Torso | 88.4 | 122.0 |





Matching algorithm results

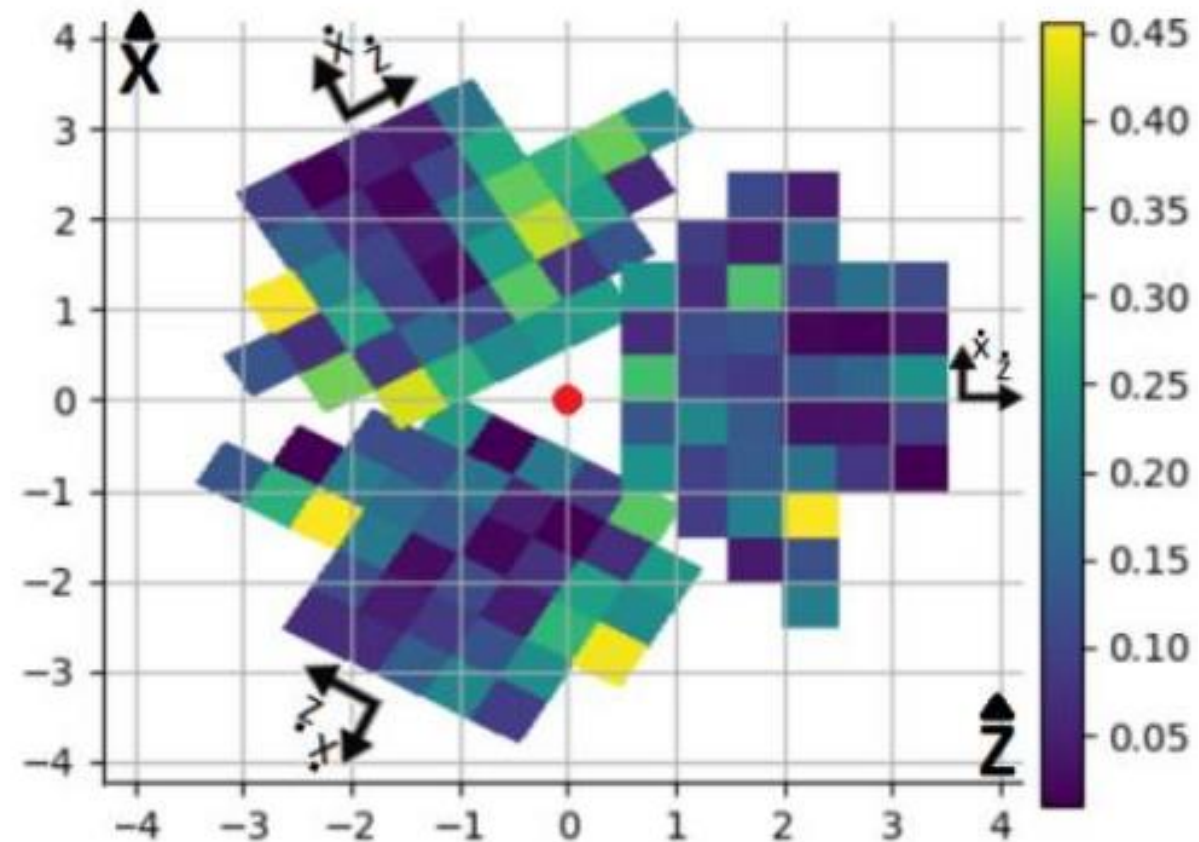
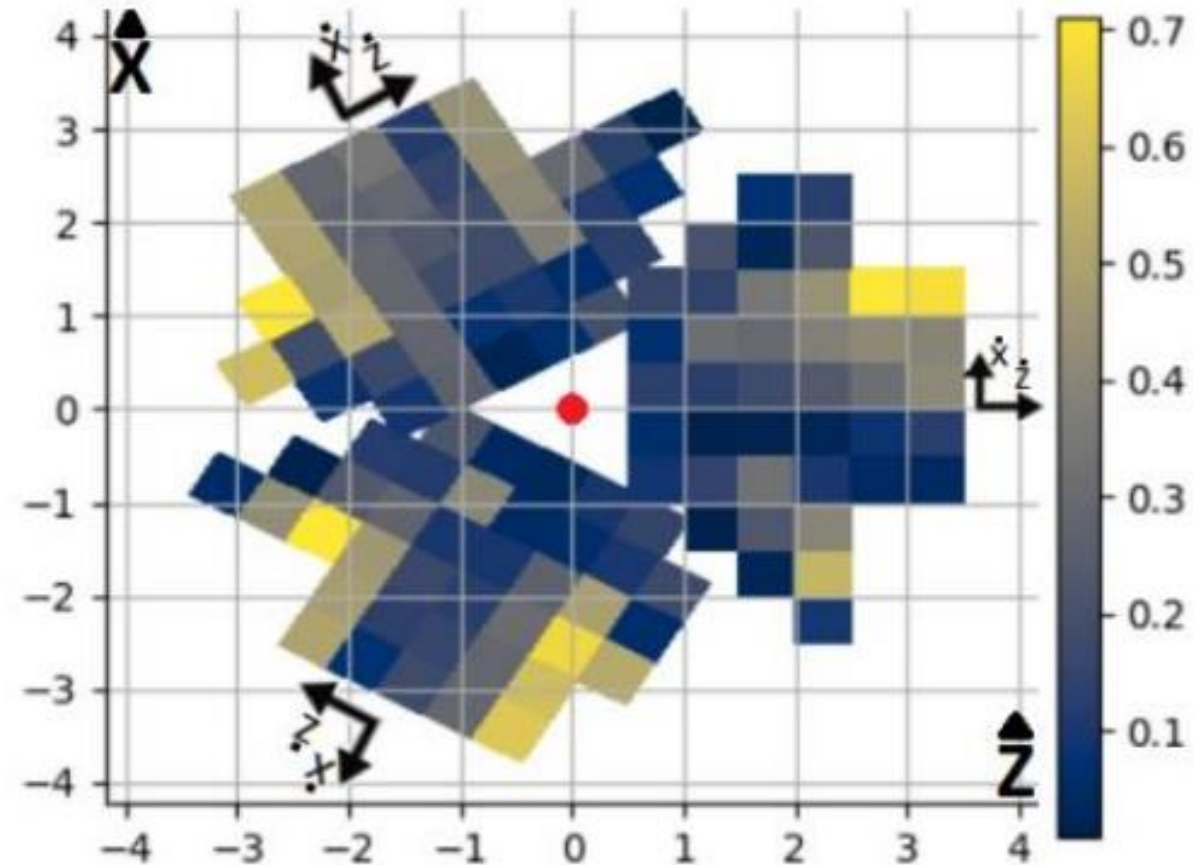
TABLE II

| | Radar 1 ↓ | Radar 2 ↓ | Radar 3 ↓ |
|----------------------|----------------|----------------|-----------------|
| Preliminary Work [3] | 23.89% ± 6.57% | 33.57% ± 50.55 | 66.89% ± 263.89 |
| Ours | 2.52% ± 2.51 | 9.44% ± 13.27 | 1.94% ± 1.52 |

- Low average matching error of 4.63%
- Error represents the absolute difference between the radar and camera matching values of an individual, divided by the camera values

Radar detection errors

- Localisation error in meters.
- The errors were evaluated in each radar's \hat{x} (left) and \hat{z} (right) directions.
- The figures represent these errors in the (\hat{X}, \hat{Z}) 2D global coordinate system.



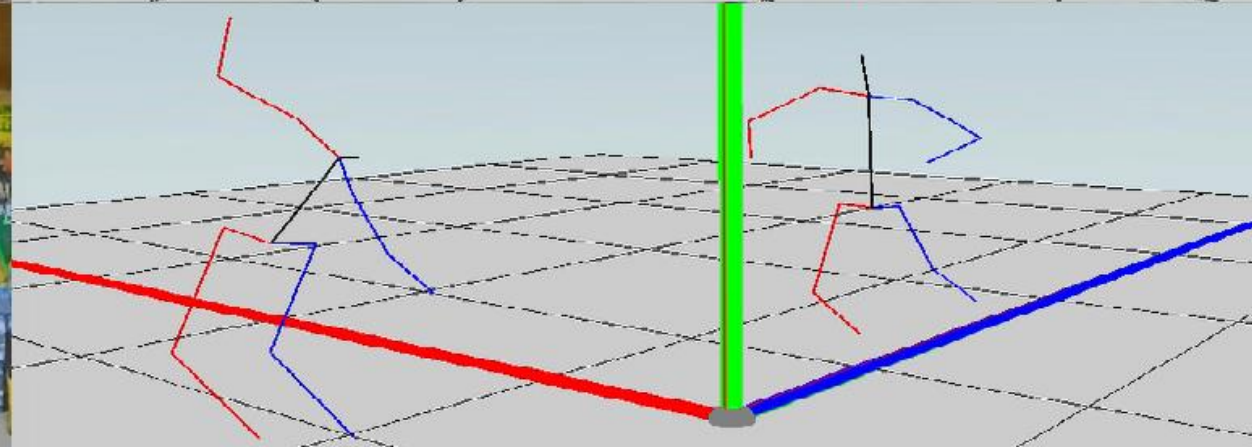
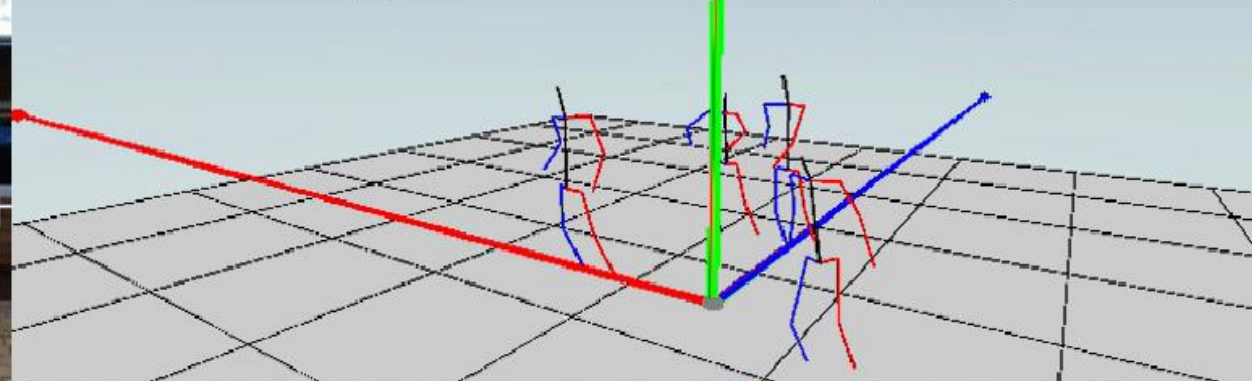
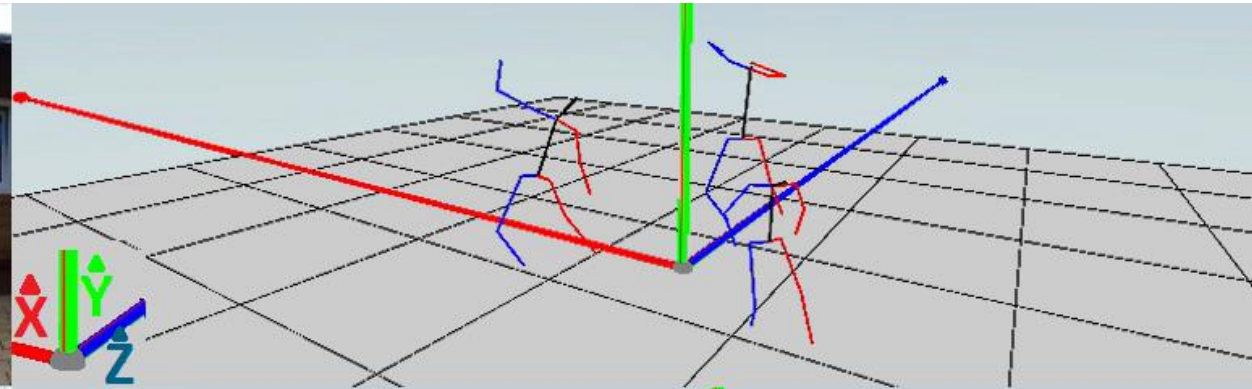
Radar detection errors

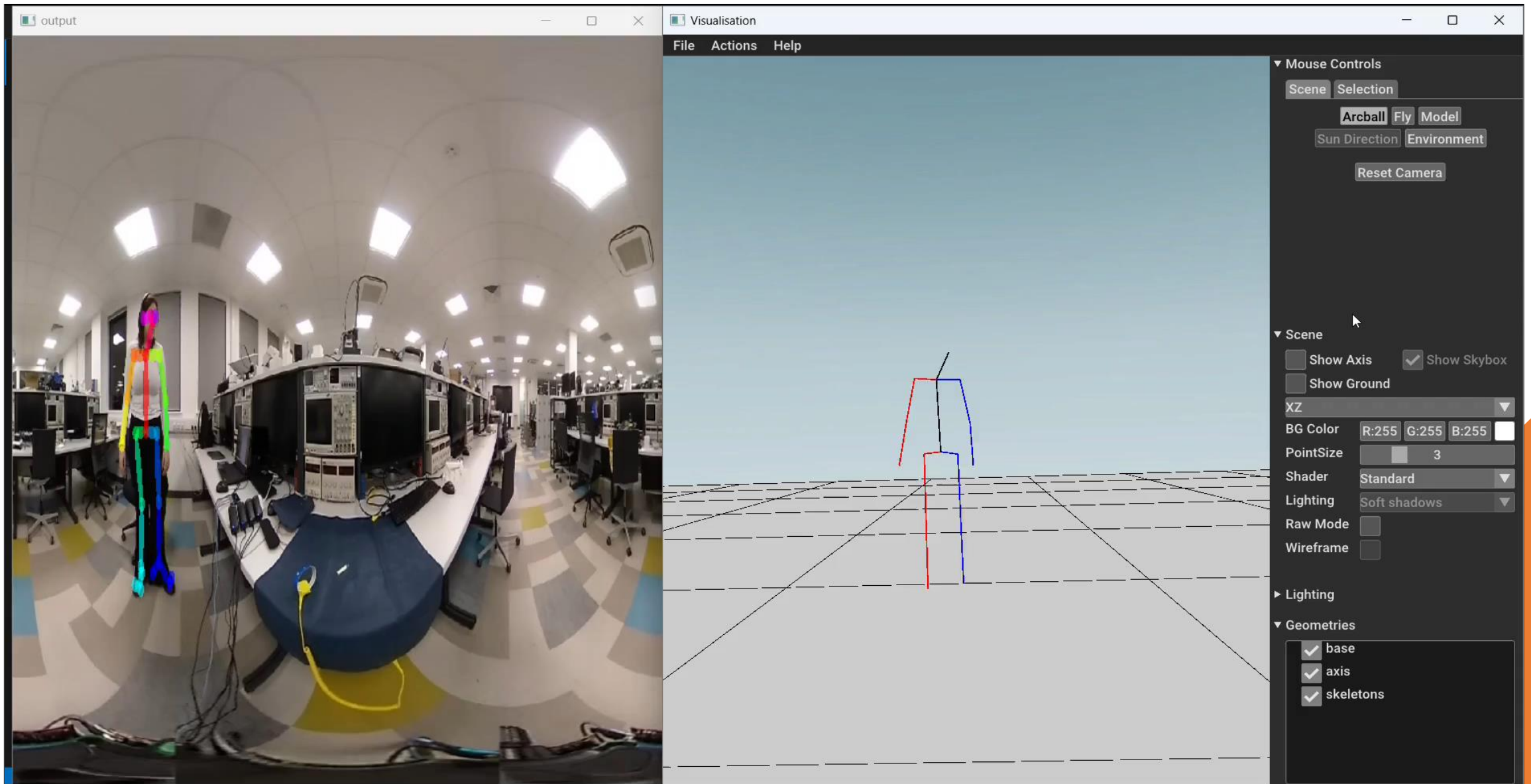
- Localization errors in centimeters were reduced with radar and camera calibrations

TABLE I

| Radar | Direction | Preliminary [3] | Ours |
|-------|-----------|-----------------|-------|
| 1 | x | 20.65 | 16.45 |
| | z | 11.41 | 11.45 |
| 2 | x | 26.19 | 24.86 |
| | z | 15.39 | 10.77 |
| 3 | x | 16.88 | 15.94 |
| | z | 13.83 | 13.46 |

Qualitative system's results





<https://youtu.be/FAFVYWSzu7Q>

Conclusions

- Robust detection system
- Performs consistently regardless of the number of individuals.
- Theoretically can handle any number of detected people.
- Another limitation is the system's inability to accurately detect when a person is facing away from it.

[1] Z. Cao et al. "Openpose: Realtime multi-person 2d pose estimation using part affinity fields", 2019.

[2] Peter Hardy and Hansung Kim. "LInKs - Lifting Independent Keypoints - Partial Pose Lifting for Occlusion Handling with Improved Accuracy In 2D-3D Human Pose Estimation", 2023.

[3] Aarti Amin et al. "Real-time 3d multi-person pose estimation using an omnidirectional camera and mmwave radars", 2023