

MVTracker: Multi-View 3D Point Tracking

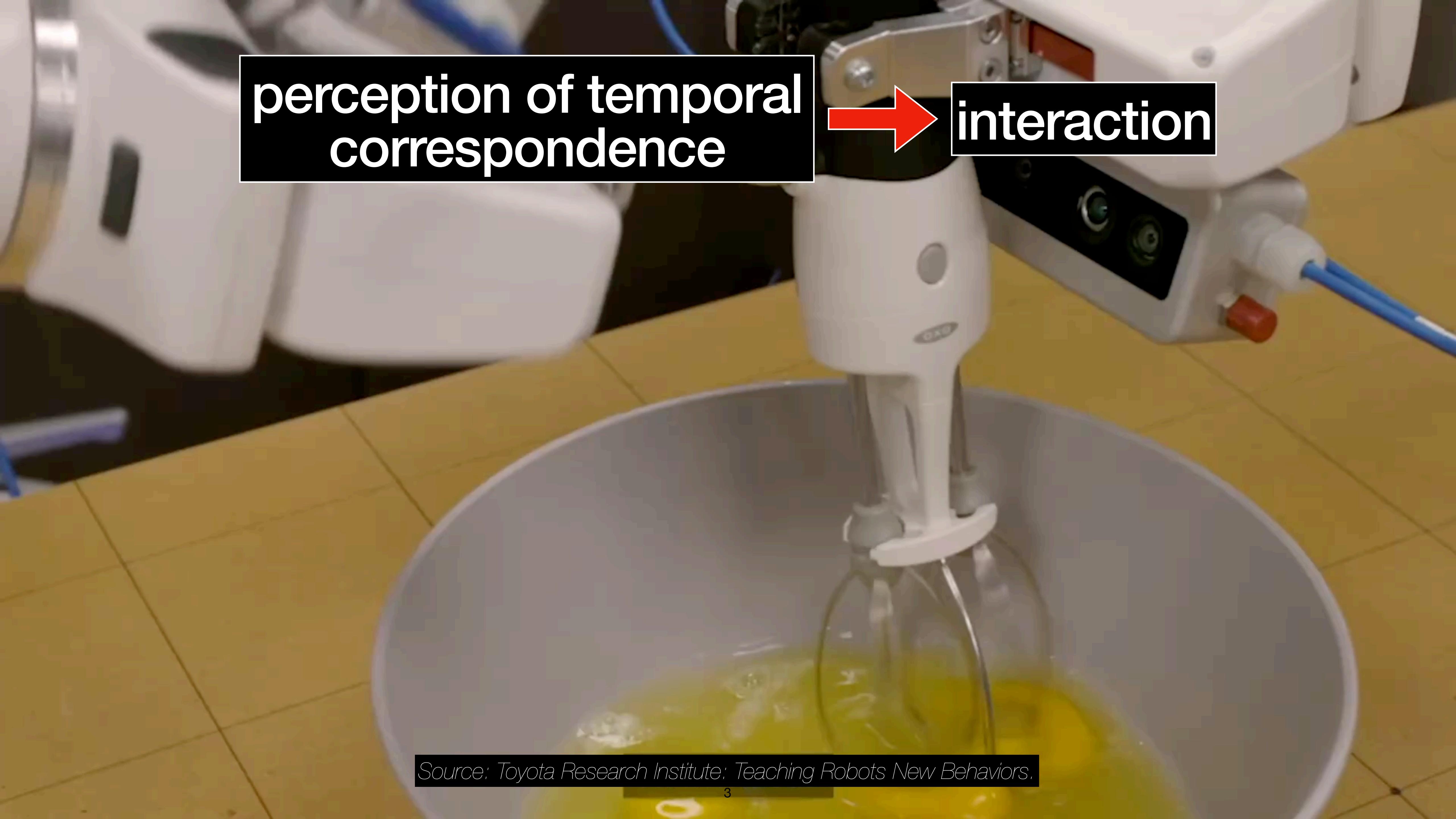
Frano Rajič¹, Haofei Xu¹, Marko Mihajlović¹, Siyuan Li¹, Irem Demir¹,
Emircan Gündoğdu¹, Lei Ke², Sergey Prokudin^{1,3}, Marc Pollefeys^{1,4}, Siyu Tang¹

¹ETH Zürich, ²Carnegie Mellon University, ³Balgrist University Hospital, ⁴Microsoft



the world is dynamic

Sources: Pexels, Pixabay.

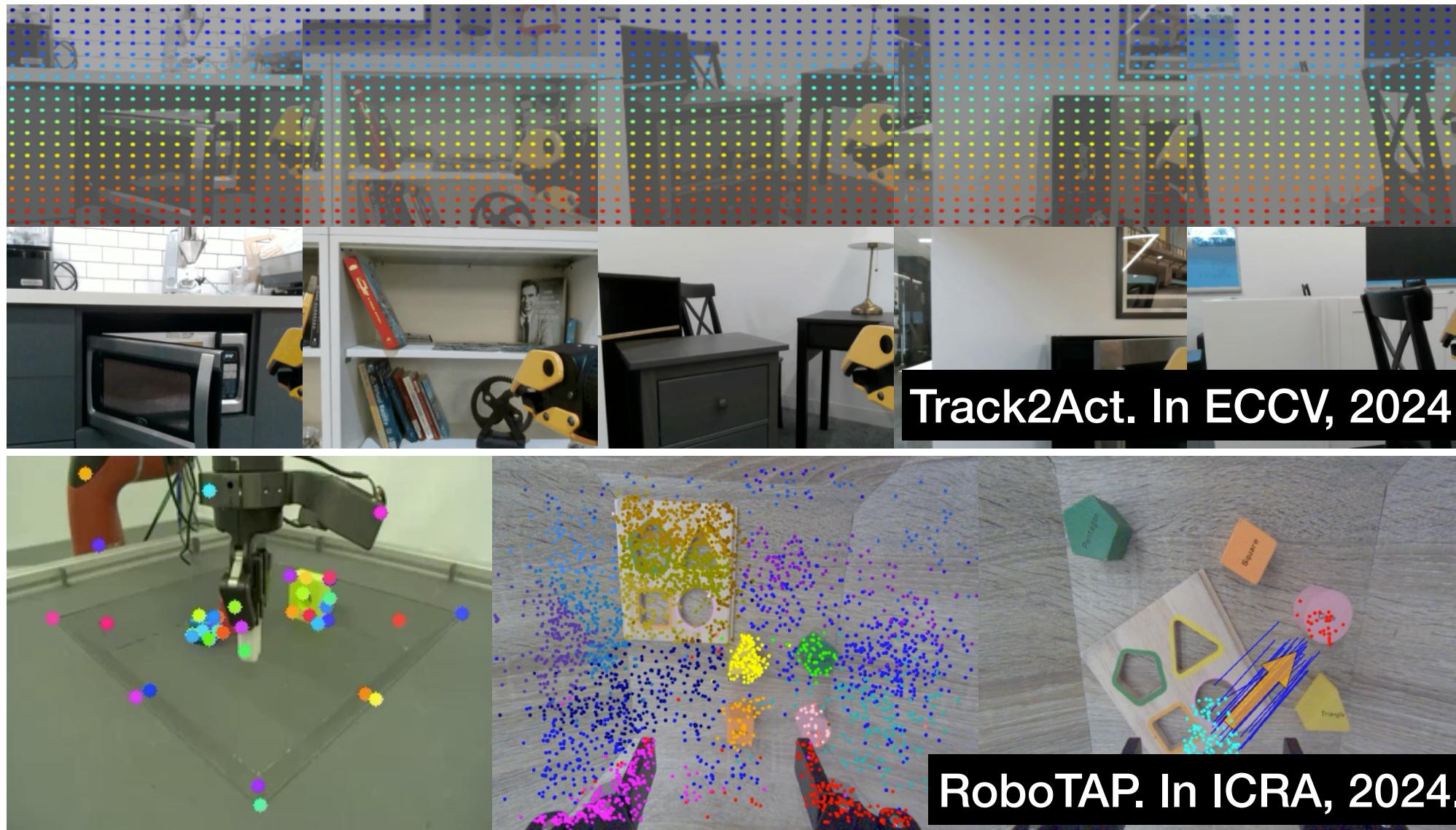


perception of temporal
correspondence

interaction

Source: Toyota Research Institute: Teaching Robots New Behaviors.

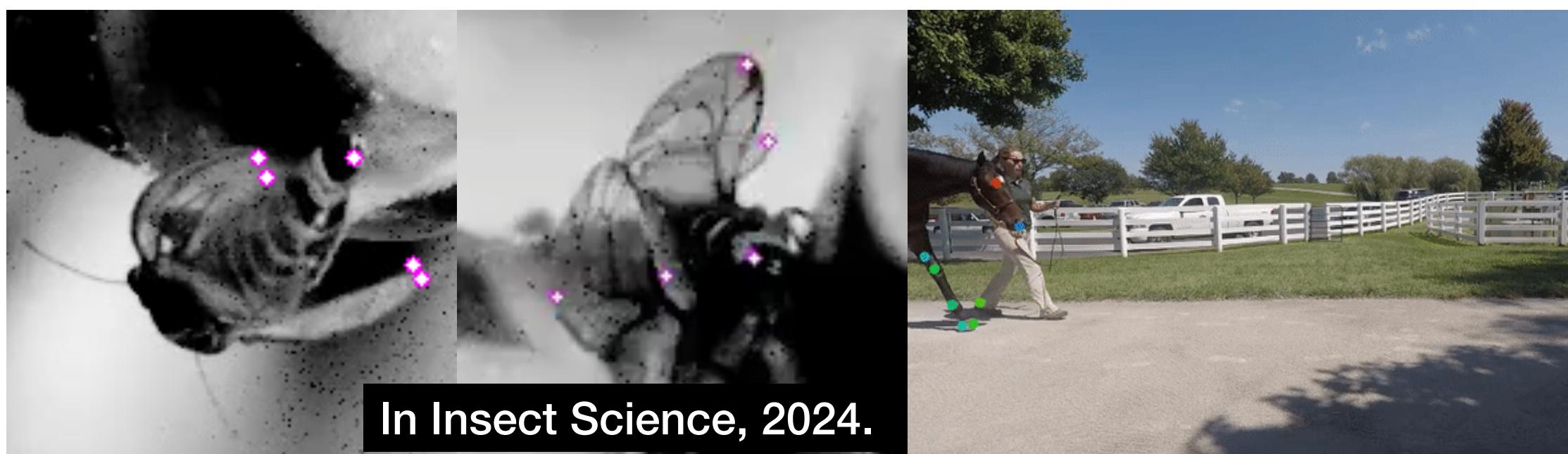
Dynamics perception in robotics and beyond



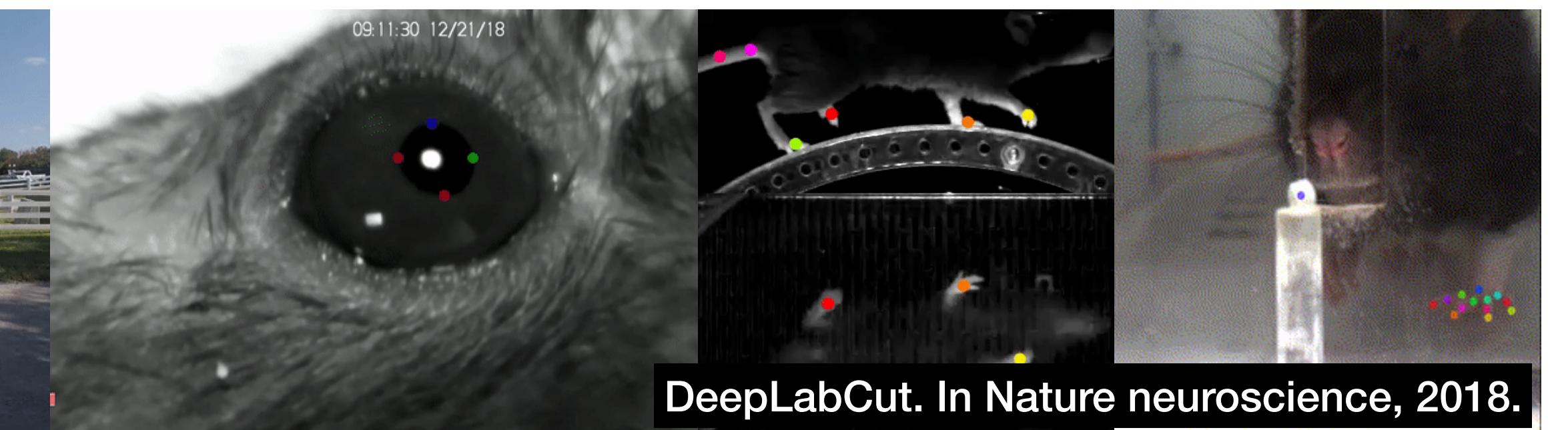
Robotics: Steering actions. Less data needed.



Surgical room: Intra-operative assistance.



Biology and neuroscience: Tracking and analysis of animal movements.



Dynamics perception in robotics and beyond

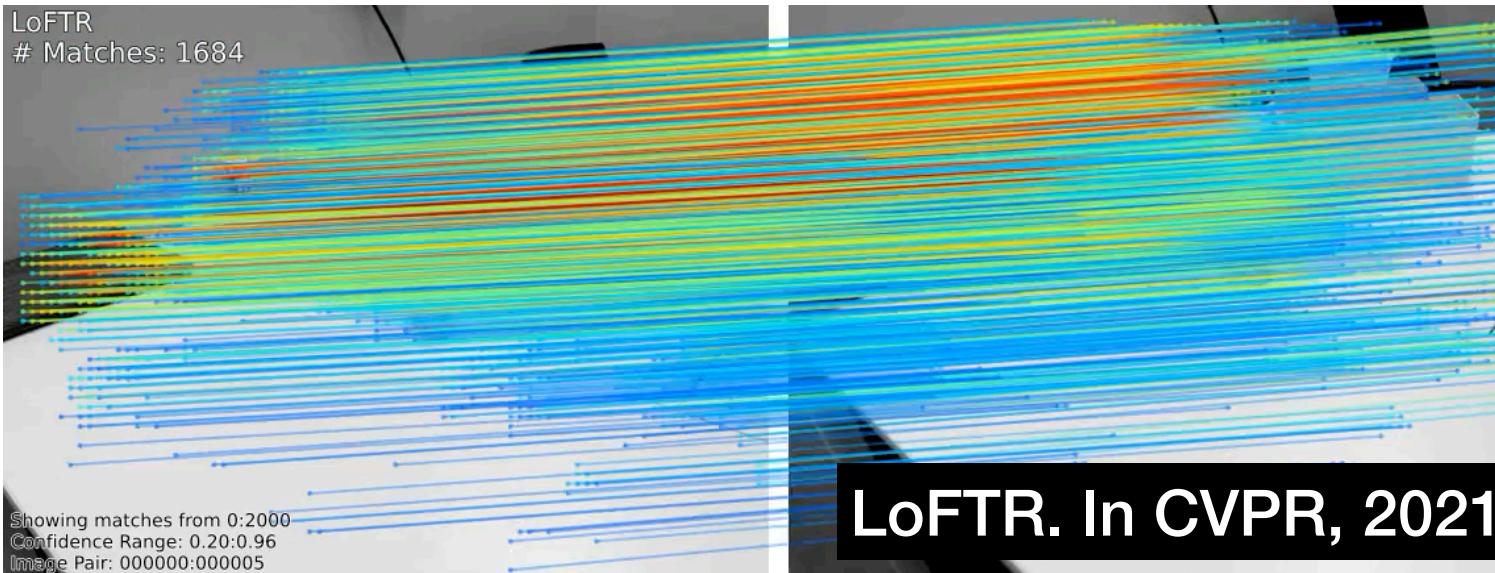
1. Efficiency.

2. Precision

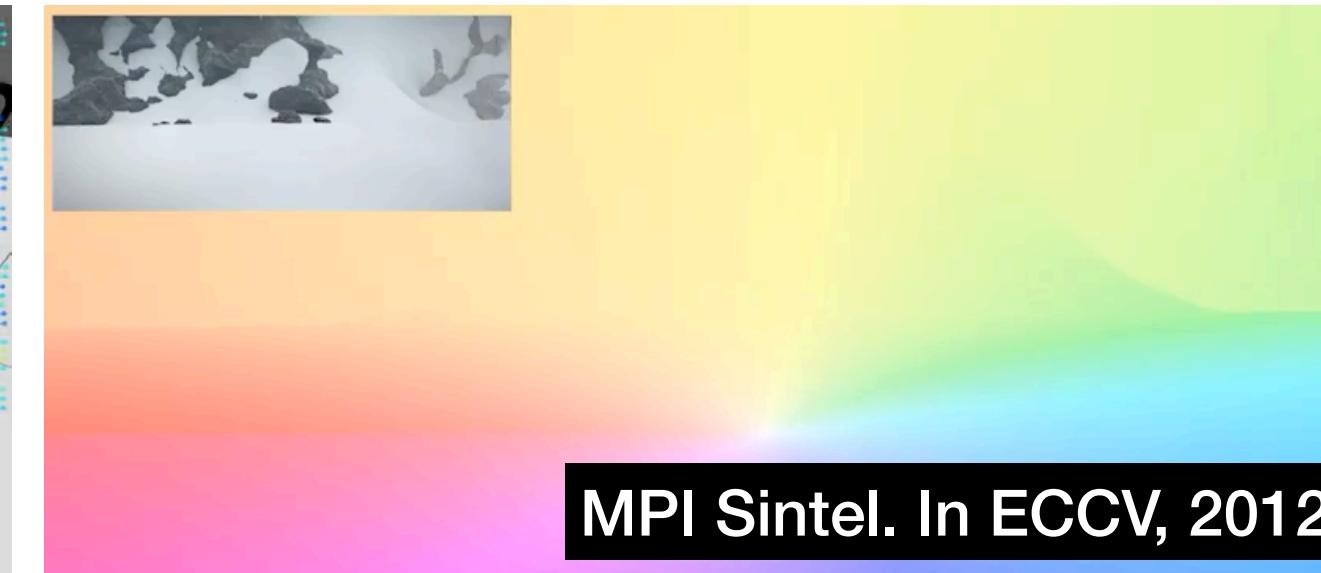
3. Robustness to camera setups (number, positions, etc.)

Overview of related problems

Point correspondences in 2D



Feature matching.

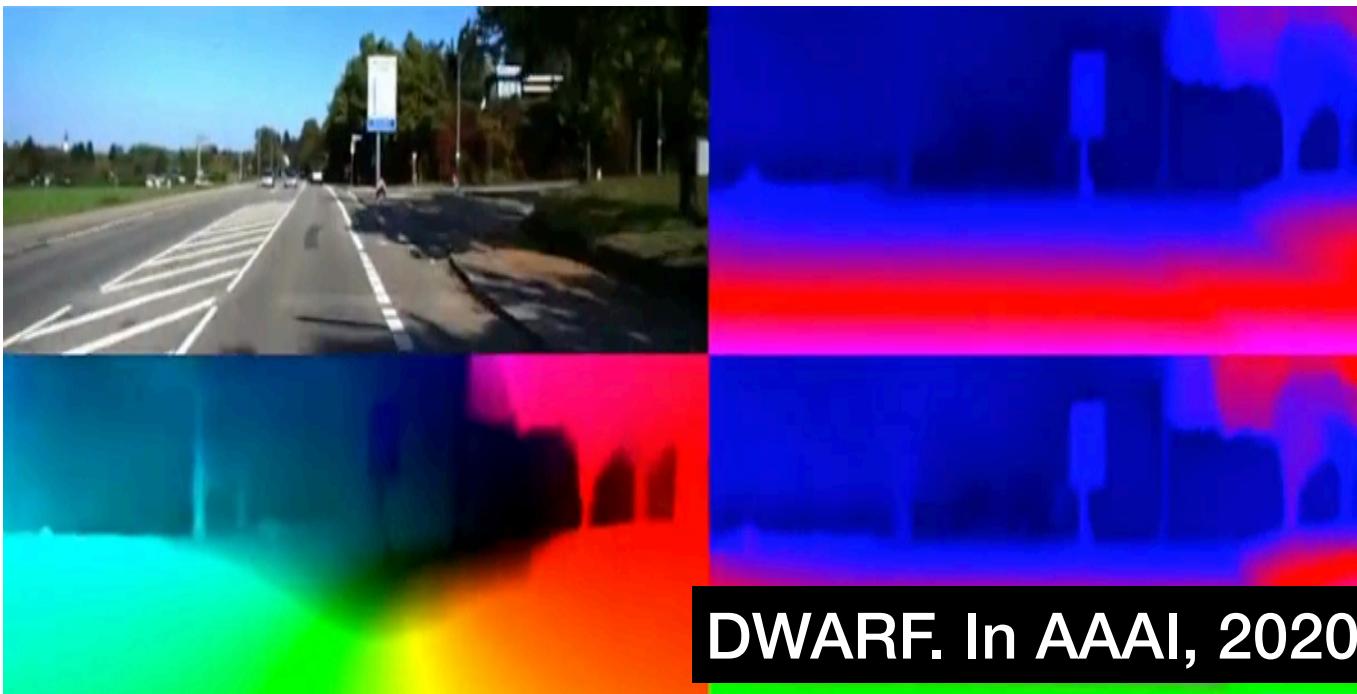


Optical flow.

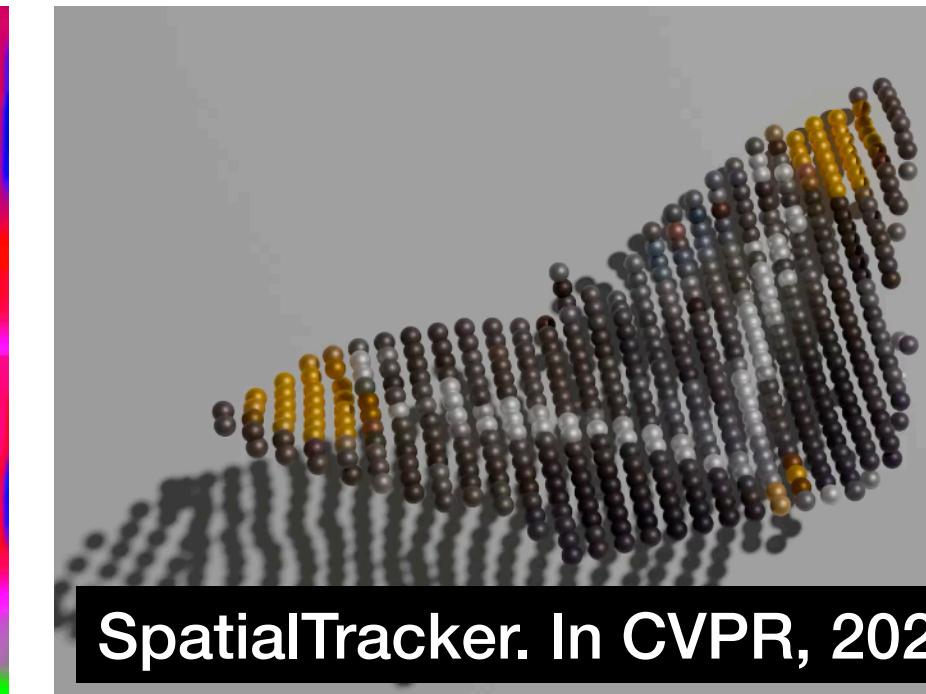


Point tracking in 2D.

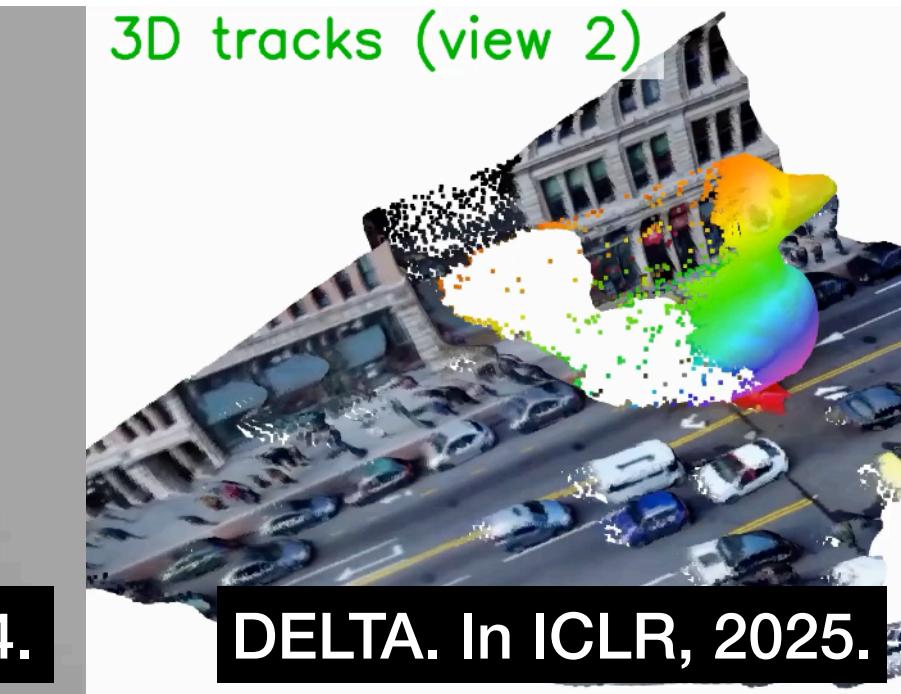
Point correspondences in 3D



Scene flow.

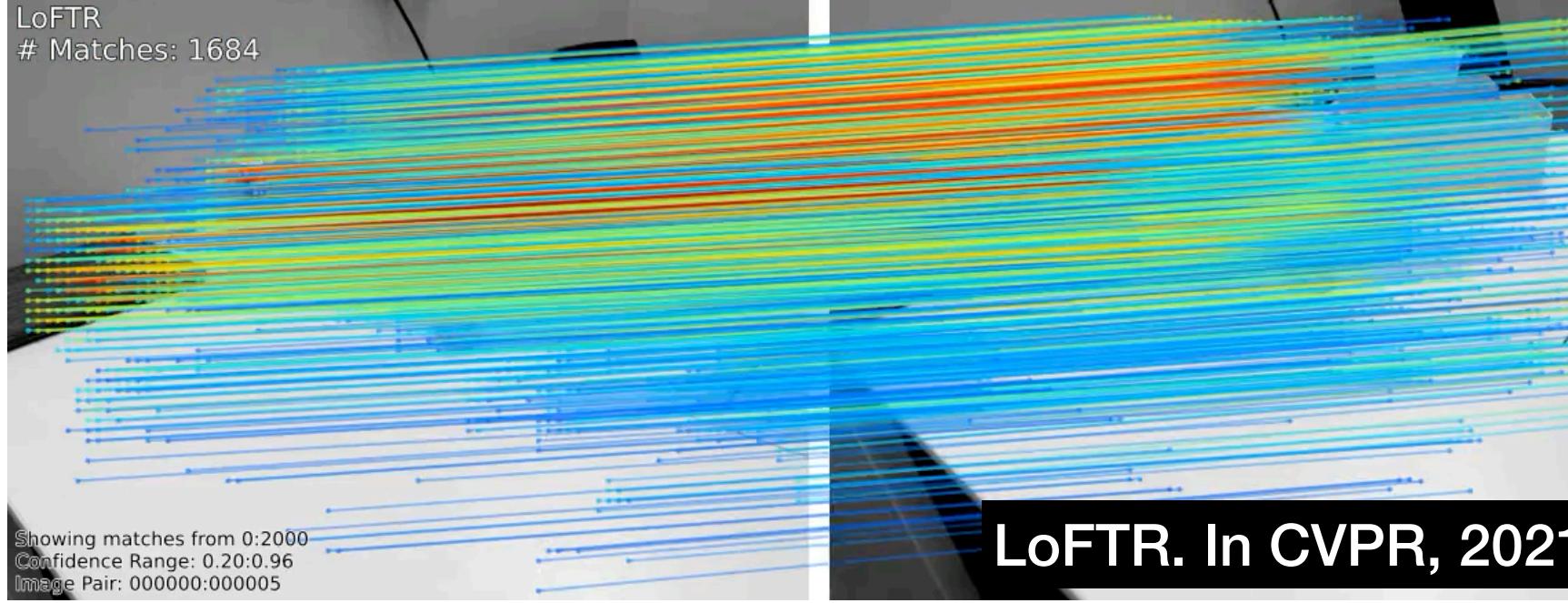


Monocular 3D point tracking.

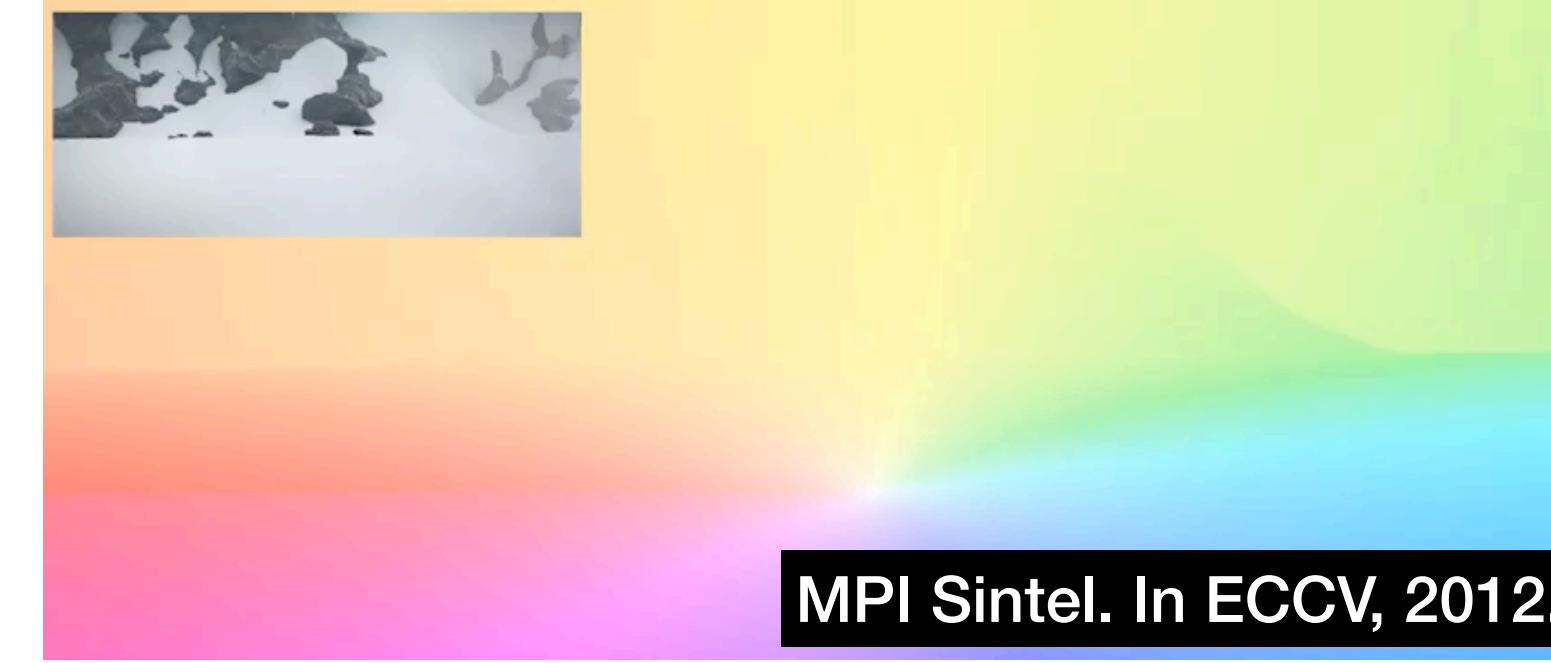


Multi-view 3D point tracking.

Point correspondences in 2D



LoFTR. In CVPR, 2021.



MPI Sintel. In ECCV, 2012.

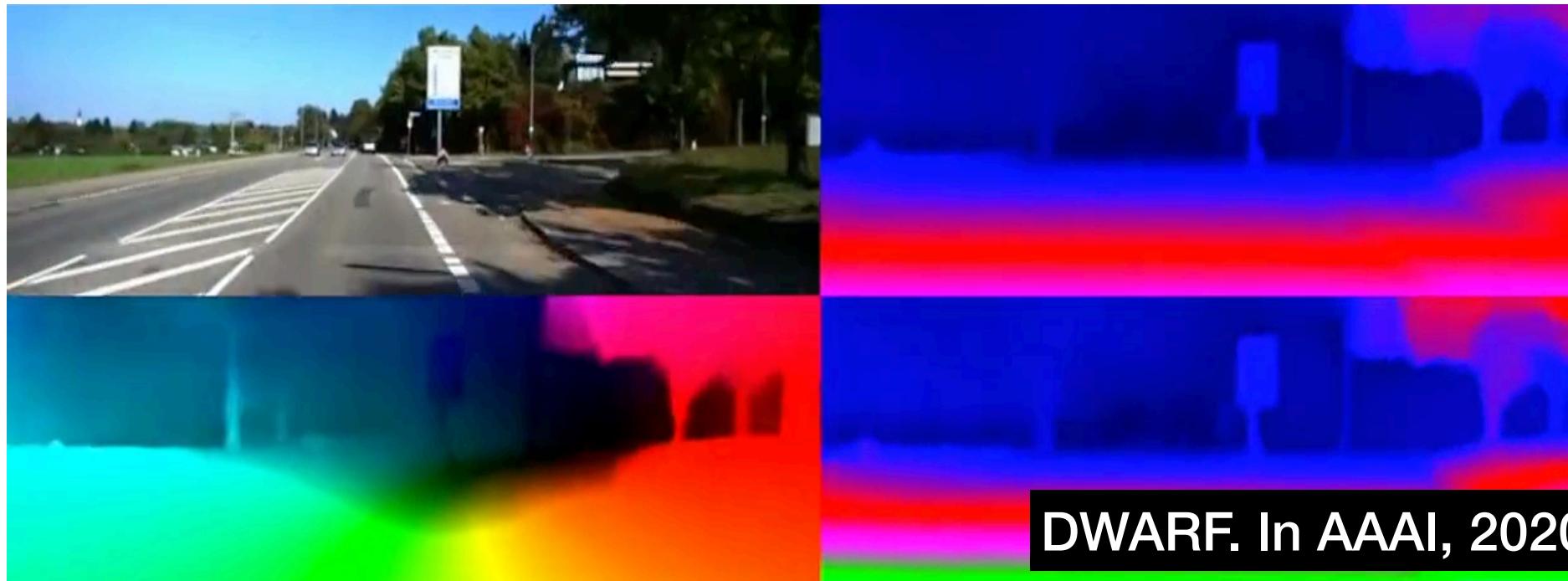
Feature matching: Tracking static points. **Optical flow:** Dynamic, short-term.



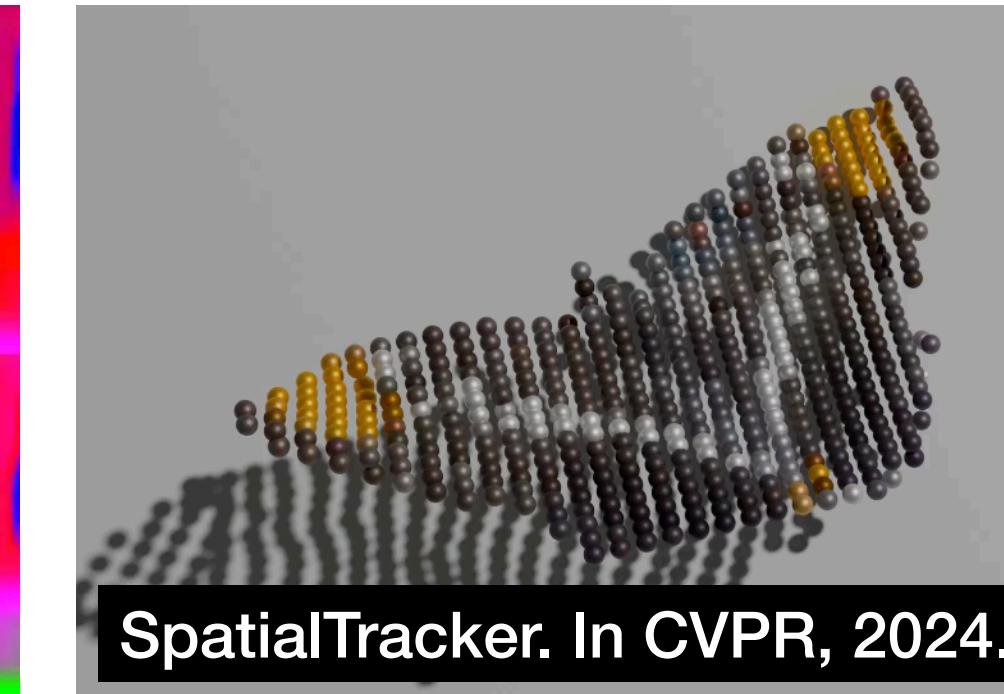
AllTracker. In ICCV, 2025.

Point tracking (2D): Dynamic points, long-term.

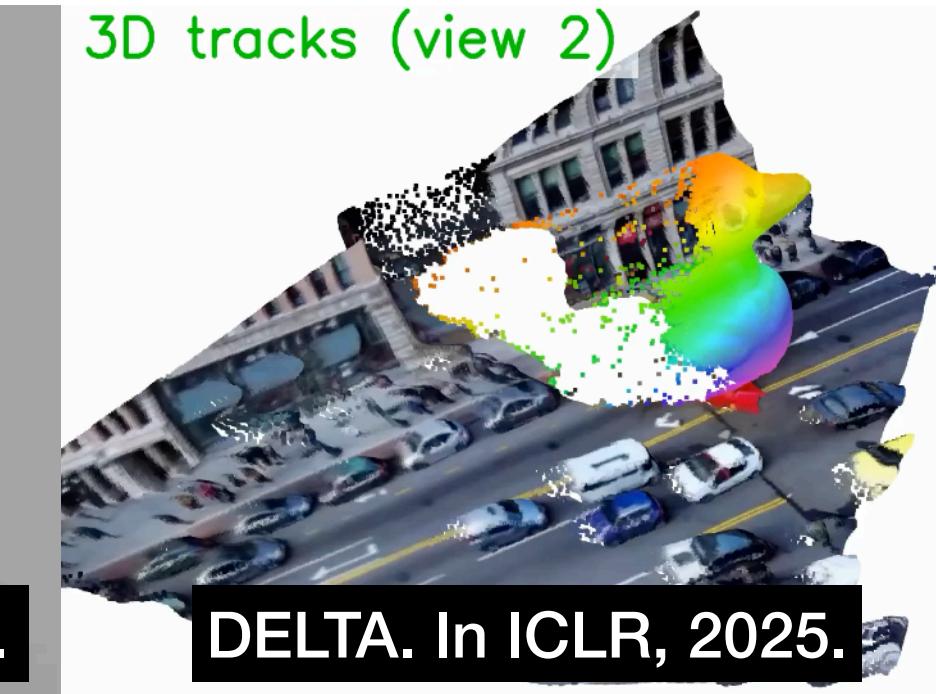
Point correspondences in 3D



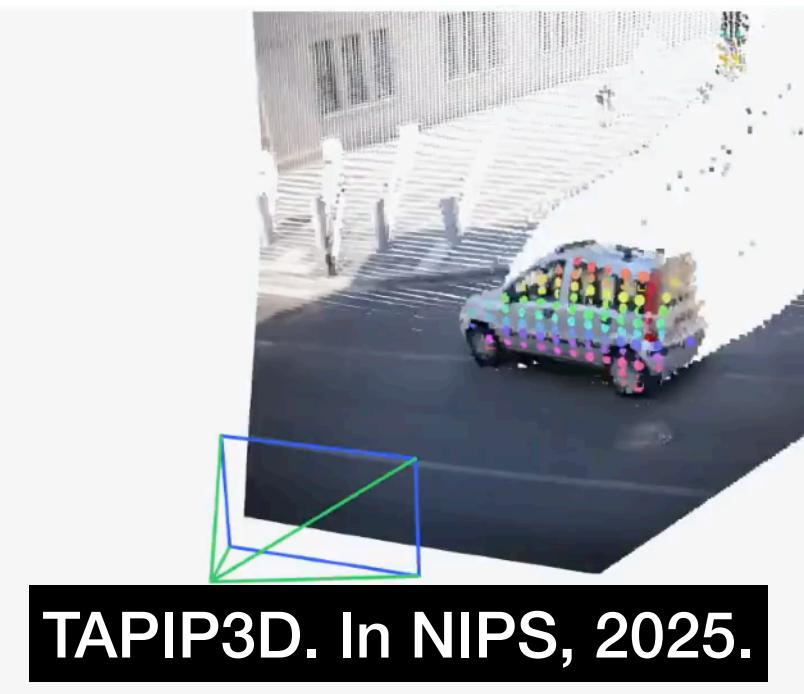
DWARF. In AAAI, 2020.



SpatialTracker. In CVPR, 2024.



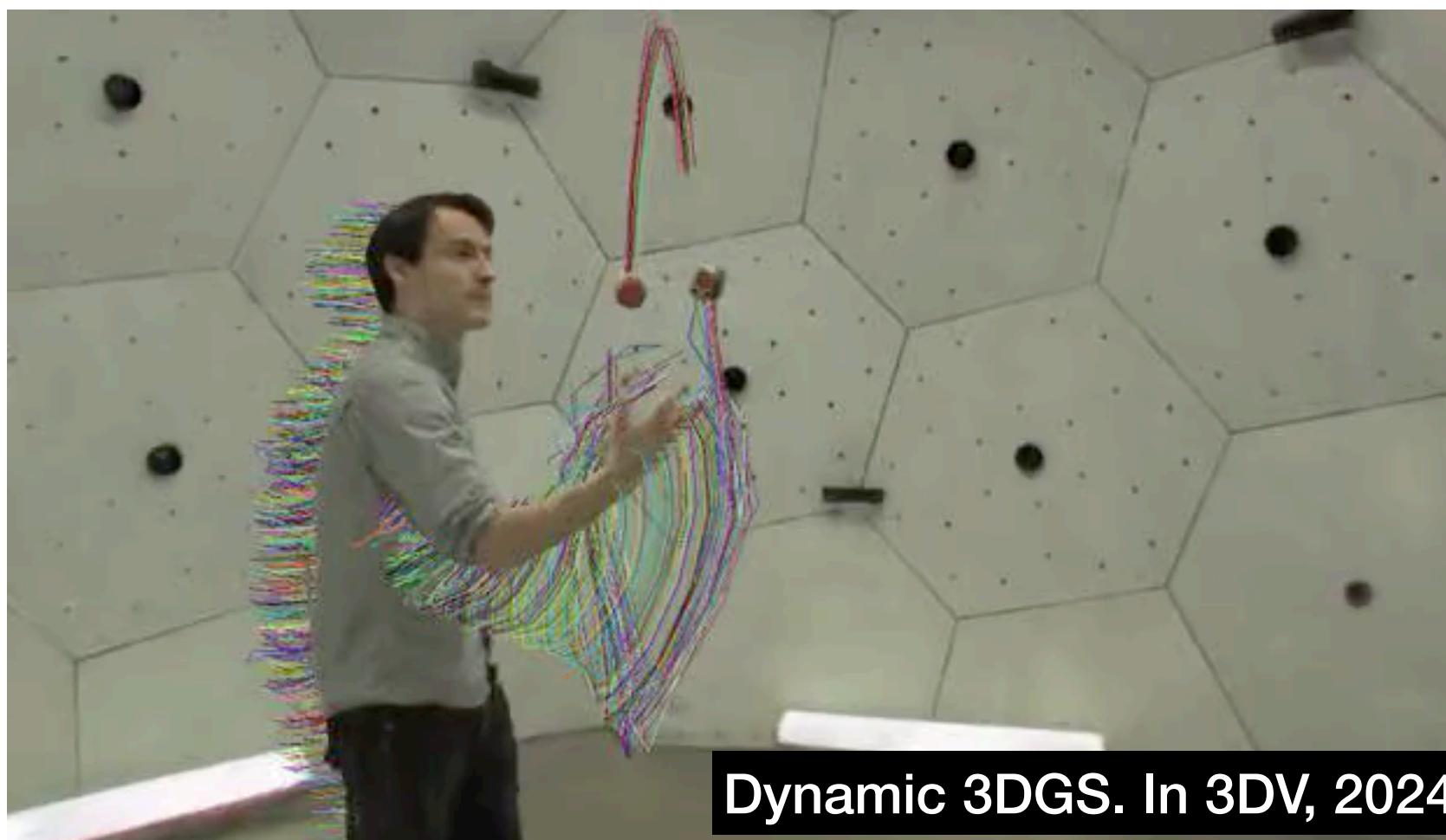
3D tracks (view 2)
DELTA. In ICLR, 2025.



TAPIP3D. In NIPS, 2025.

Scene flow: Extension of optical flow to 3D.

Monocular 3D point tracking: Can't leverage multi-view input.



Dynamic 3DGS. In 3DV, 2024.



GauSTAR. In CVPR, 2025.

Multi-view 3D point tracking: Better coverage. Only optimization-based.
Require +27 cameras or rely on monocular priors.

Point correspondences in 3D

MVTracker (ours):

1. Runs at 14.9 FPS¹.
2. Works with as few as 2–4 cameras.
3. Directly leverages multi-view input.
4. State-of-the-art performance.

¹Measured on an NVIDIA H200 for 512x384 inputs and 4 views.

Dynamic 3DGS. In 3DV, 2024.

GauSTAR. In CVPR, 2025.

Multi-view 3D point tracking: Better coverage. Only optimization-based.
Require +27 cameras or rely on monocular priors.

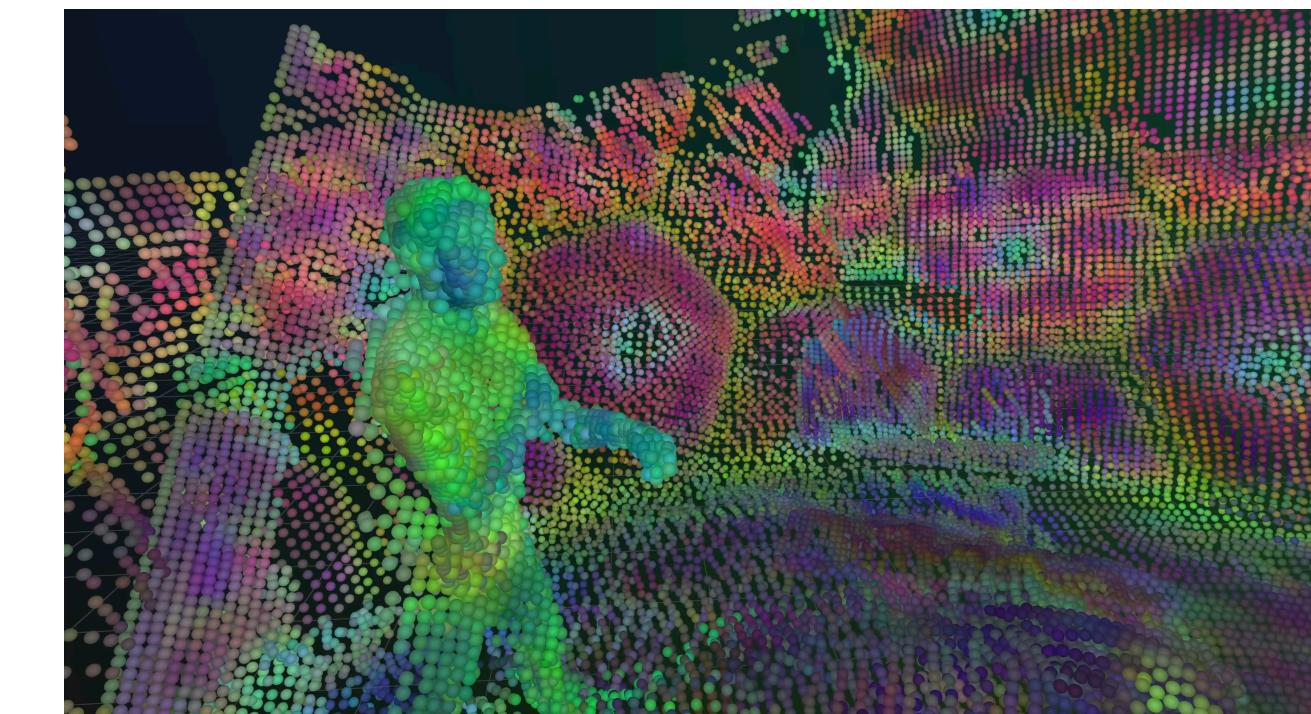
Fusing multi-view features into a point cloud

Colors visualize top 3 PCA components of the features.

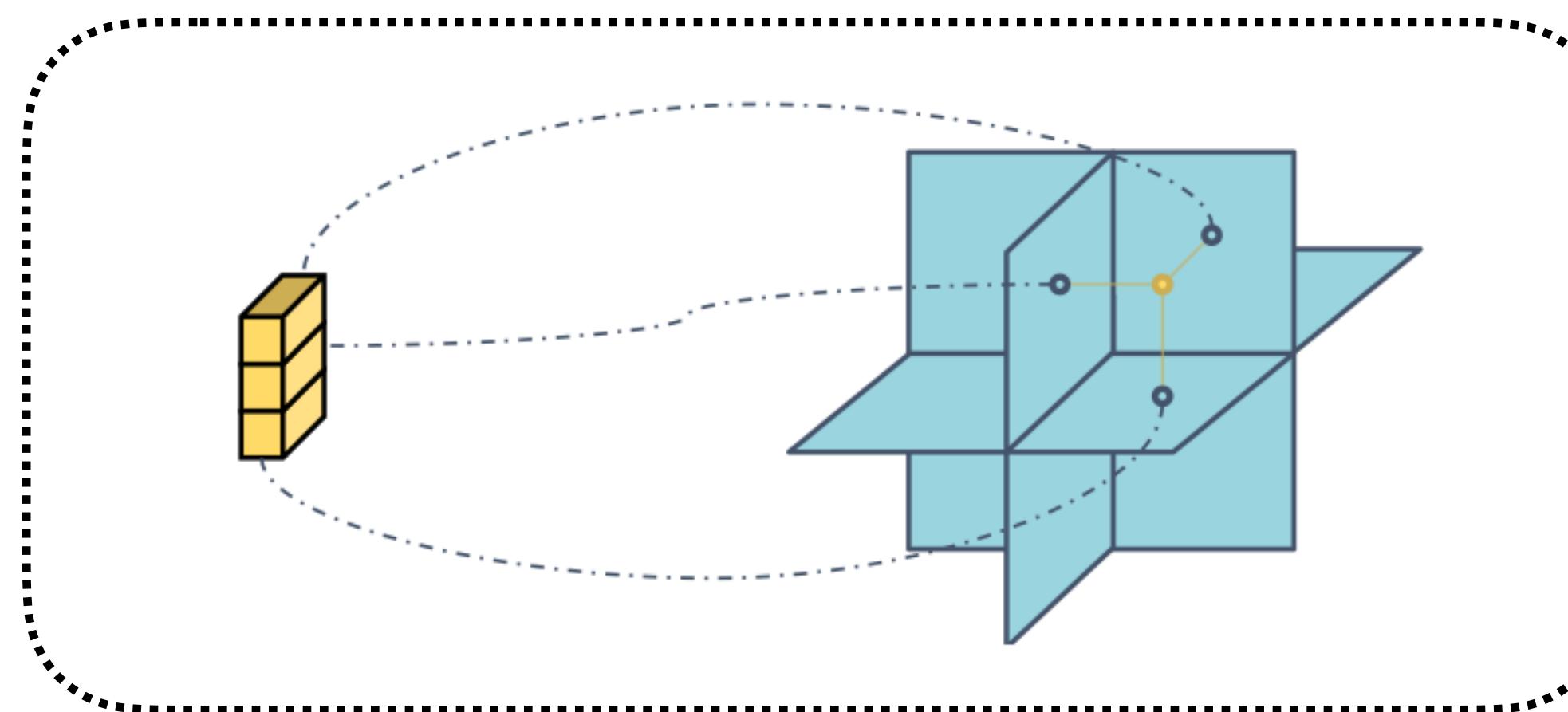
Fusing multi-view features into a point cloud

MVTracker (ours):

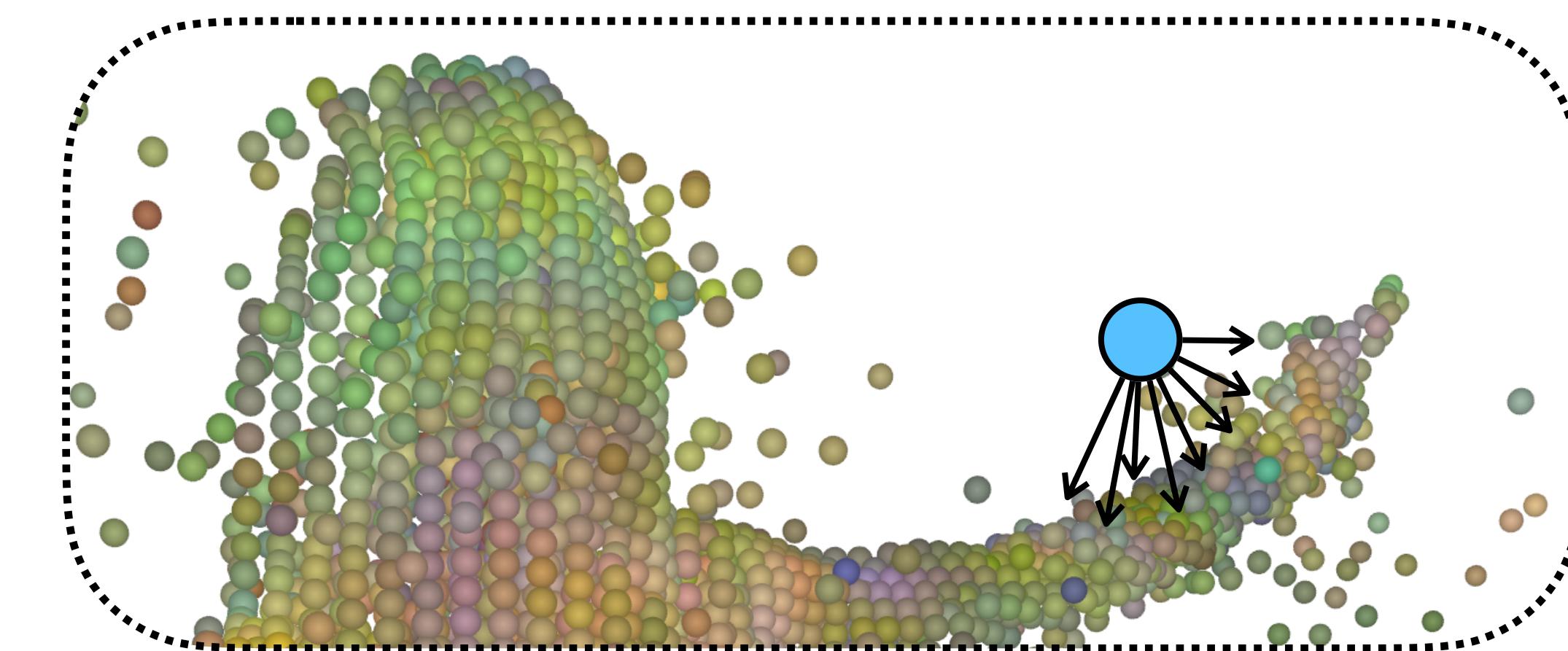
- **Data-driven** prior using a transformer.
- Contribute train. and eval. **datasets**.
- **Flexible** to a different number of cameras, their arrangements, and the depth source.



Fused 3D feature point cloud.



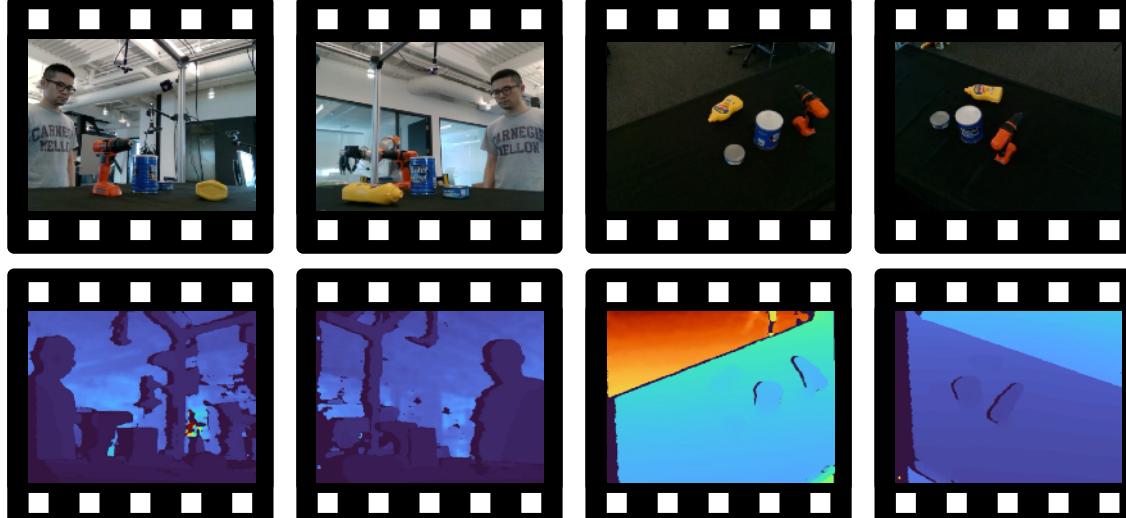
SpaTracker (CVPR'24): Triplane correlation.
Overlap and compression. \rightarrow Information loss.



Ours: Efficient kNN correlation in point cloud.

Multi-view 3D point tracking

4 Kinect Cameras
(RGB + Depth)

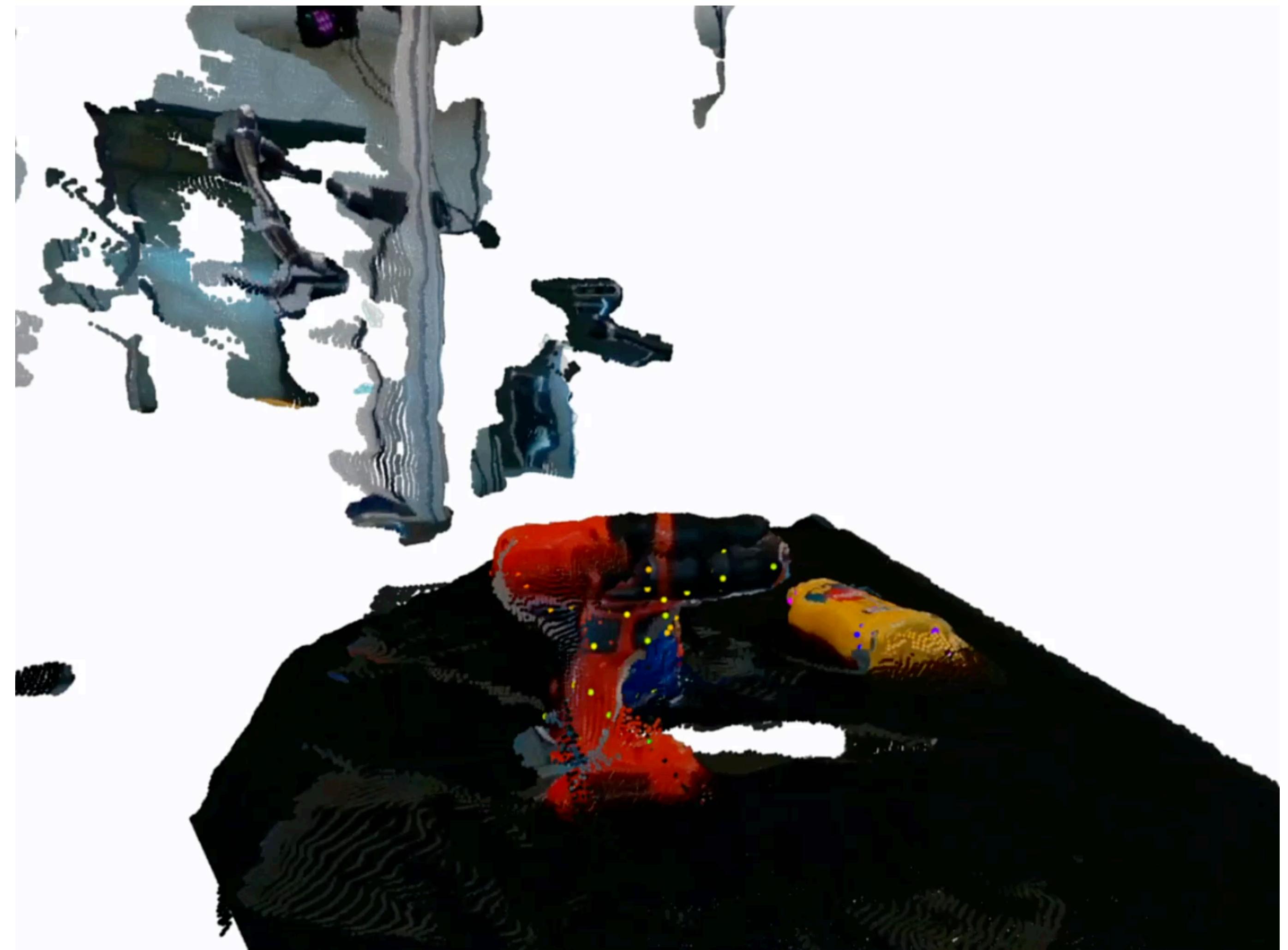
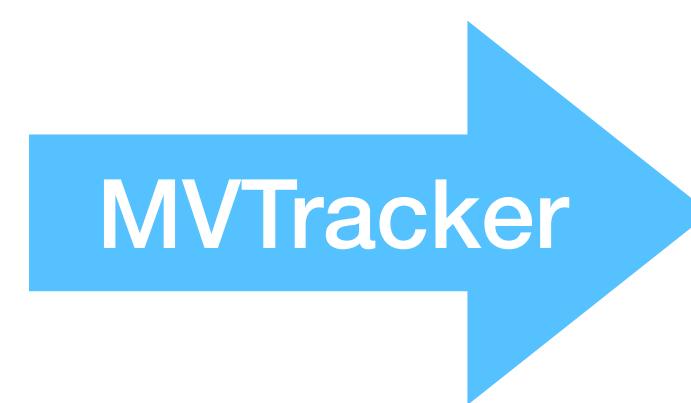


Dataset: DexYCB

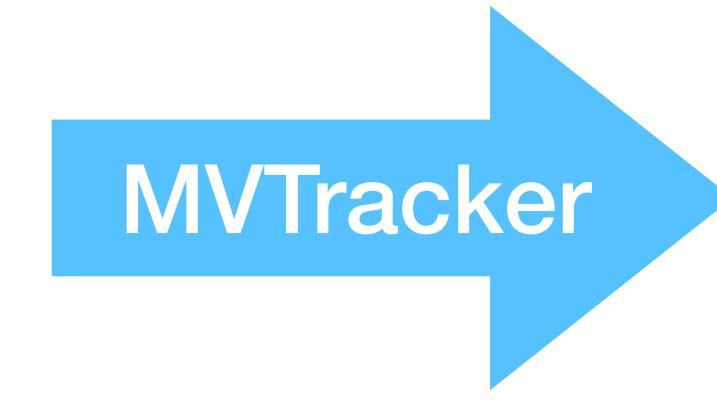
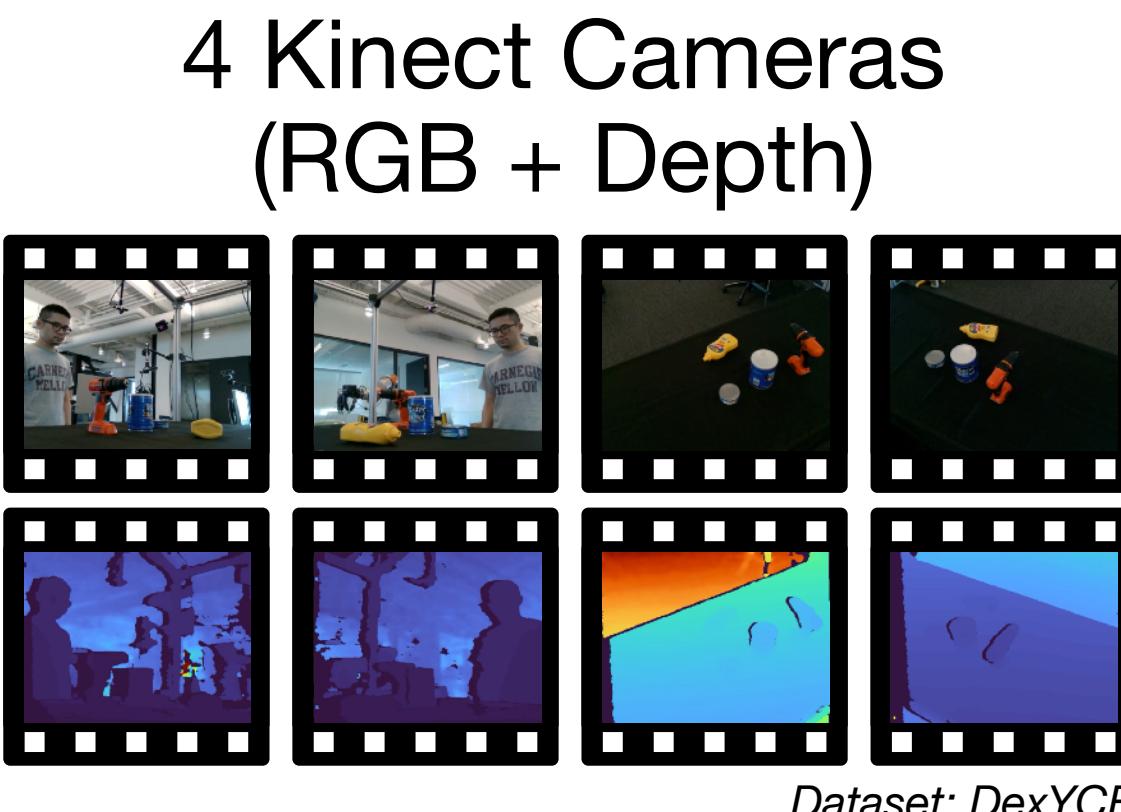
+

Camera Poses (calibrated)

MVTracker

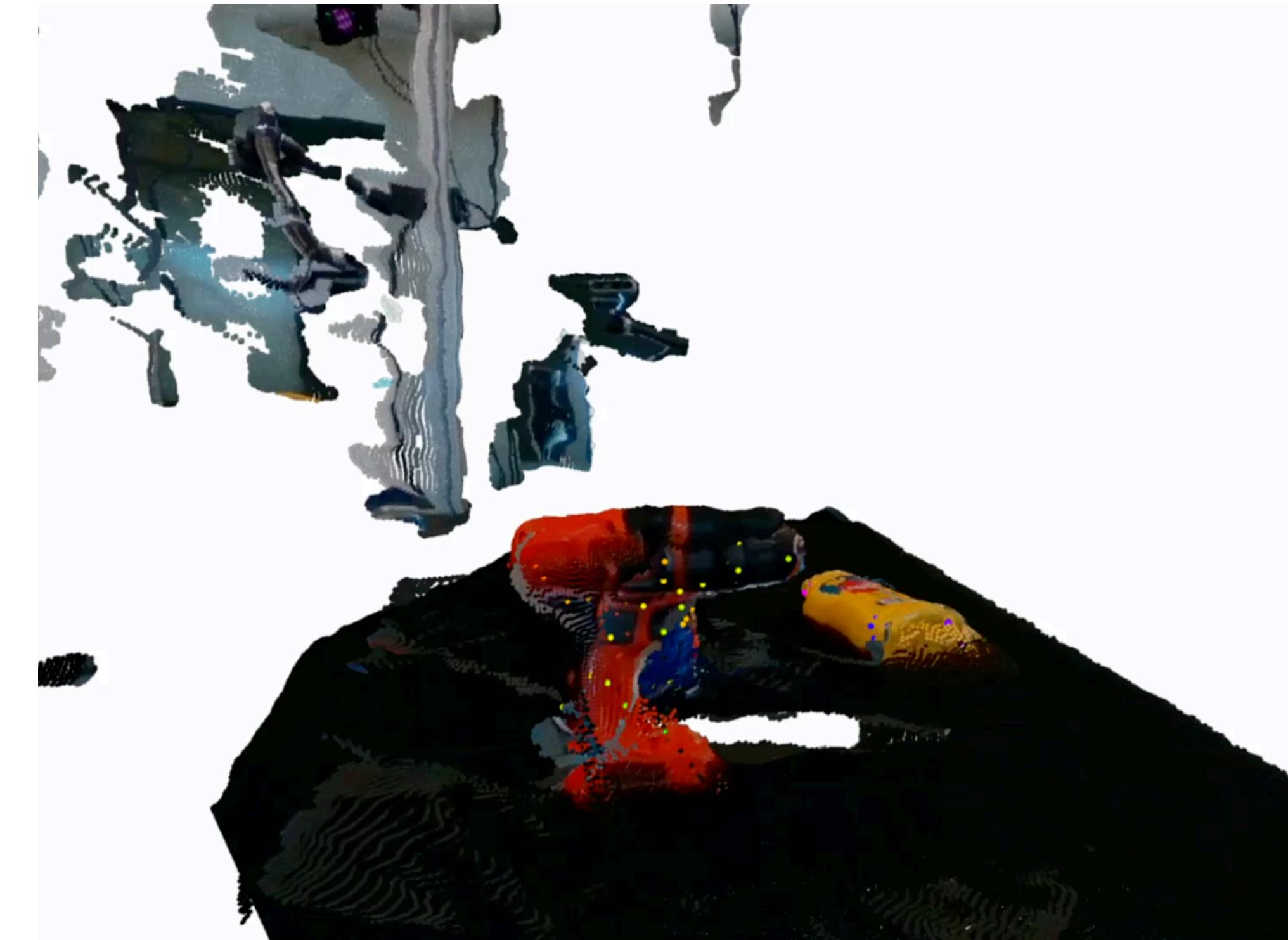


Multi-view 3D point tracking



+

Camera Poses (calibrated)



Inputs:

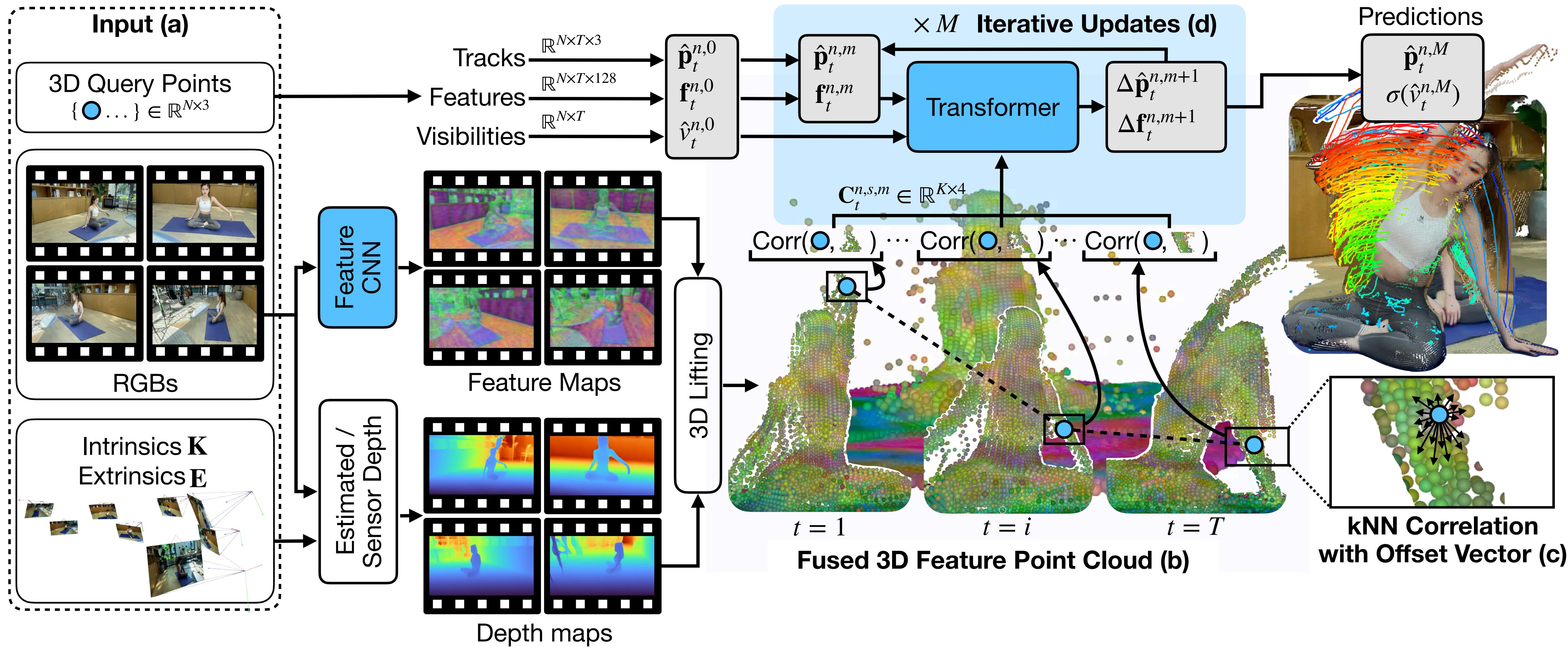
- Multi-view RGB: $(V, T, H \times W \times 3)$.
- Depth maps: $(V, T, H \times W \times 1)$.
- Intrinsics: $(V, T, 3, 3)$.
- Extrinsics: $(V, T, 3, 4)$.
- 3D query points: $(N, 4)$ as $txyz$.

Outputs:

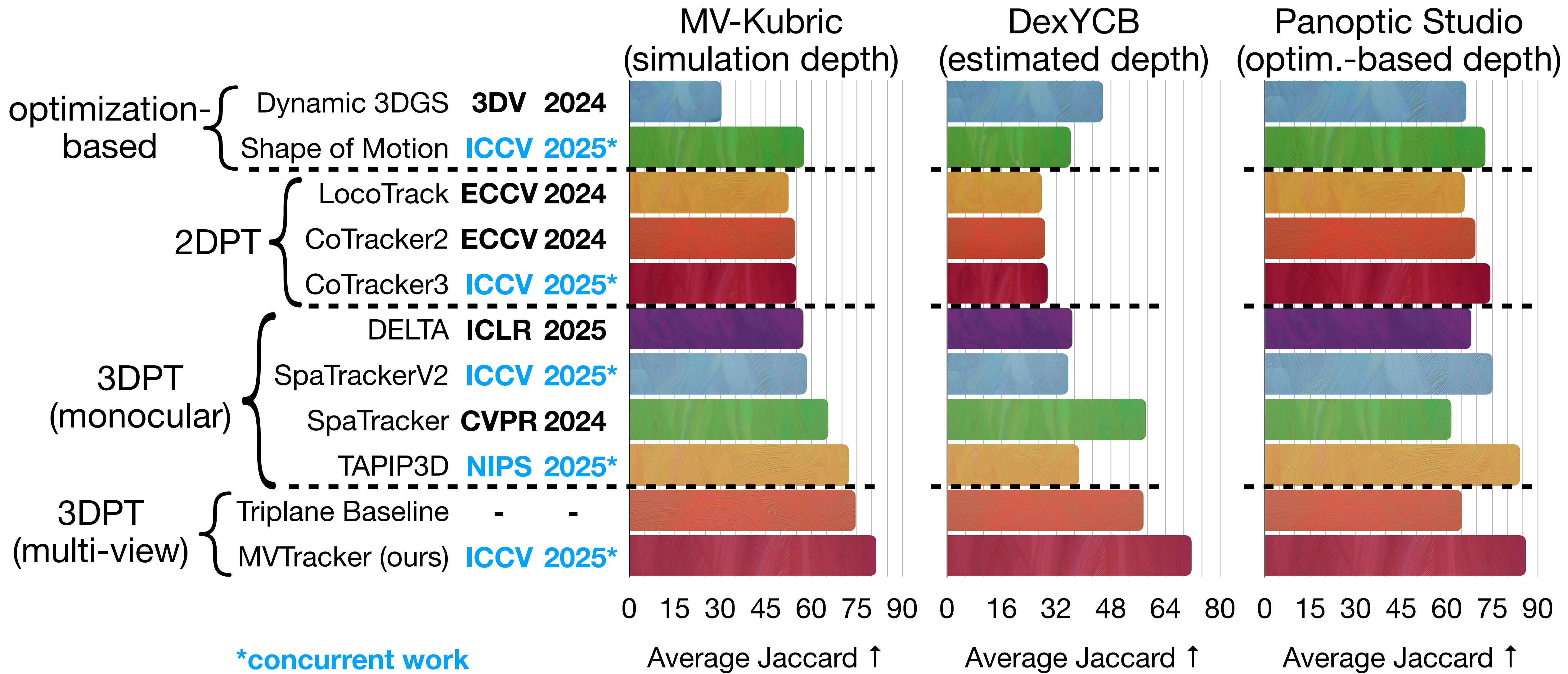
- Predicted tracks: $(N, T, 3)$.
- Predicted visibility: $(N, T, 1)$.

Notation: V views; T frames; N tracks.

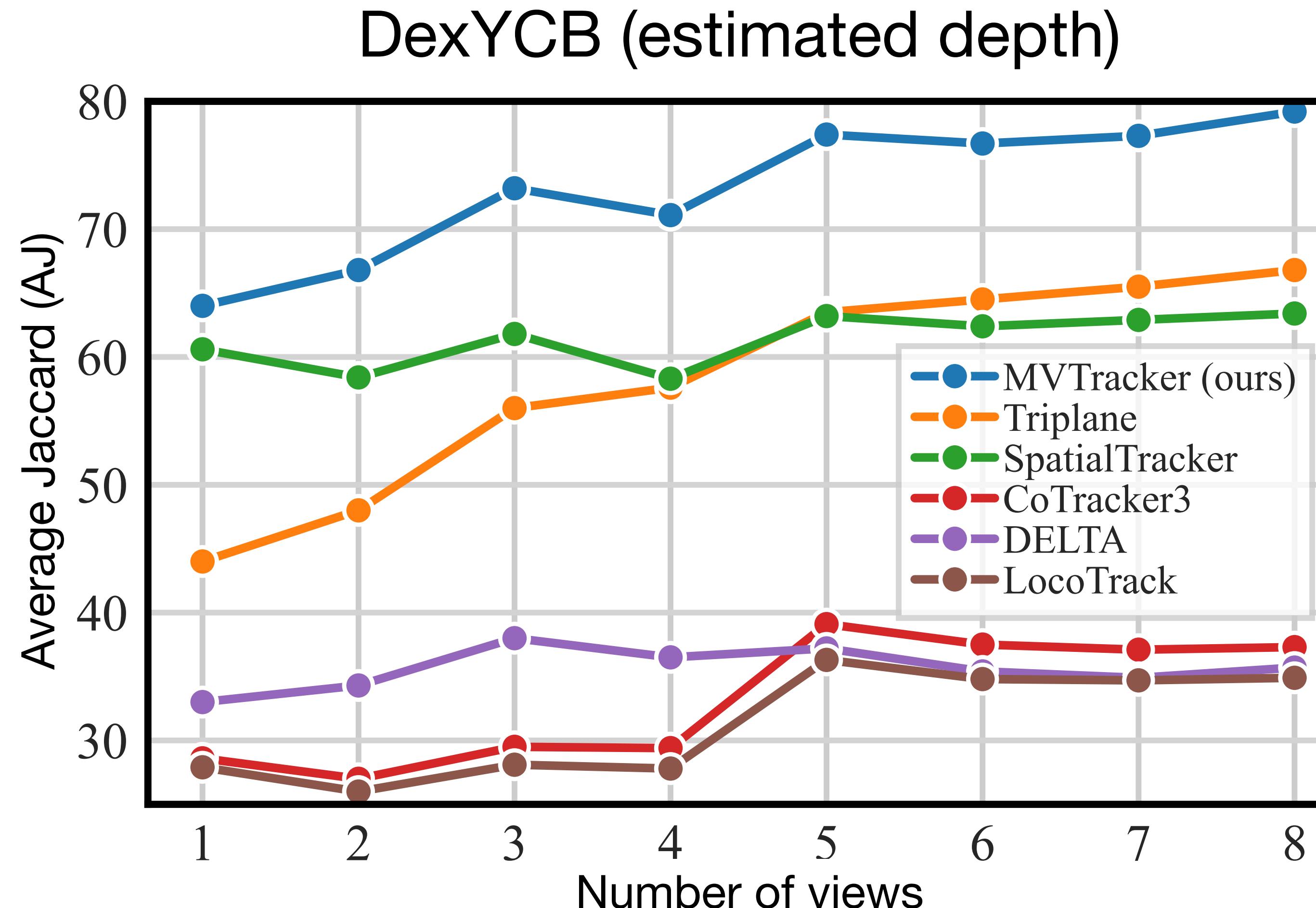
Fused 3D feature point cloud + transformer



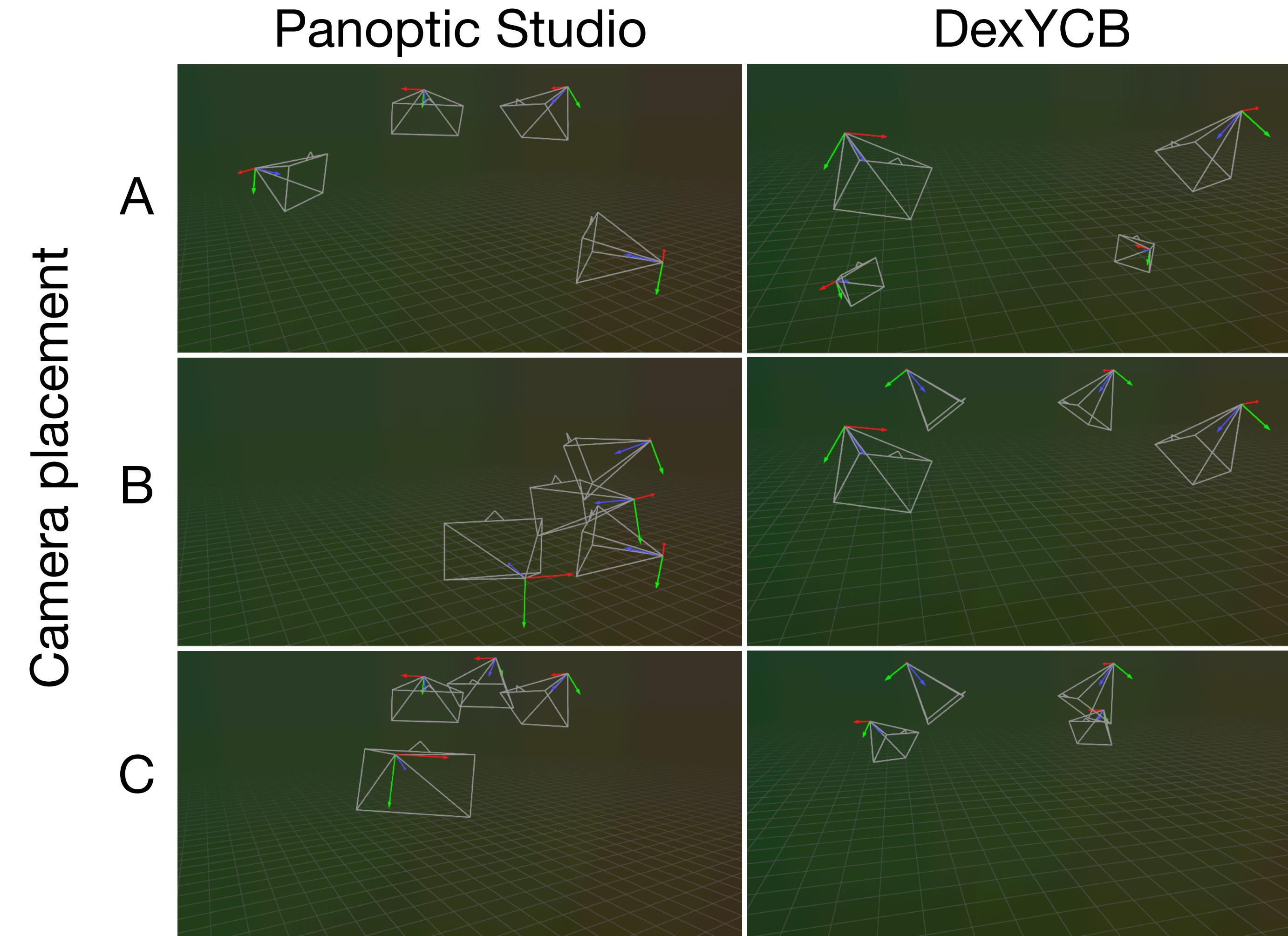
Main comparison (4 cameras)



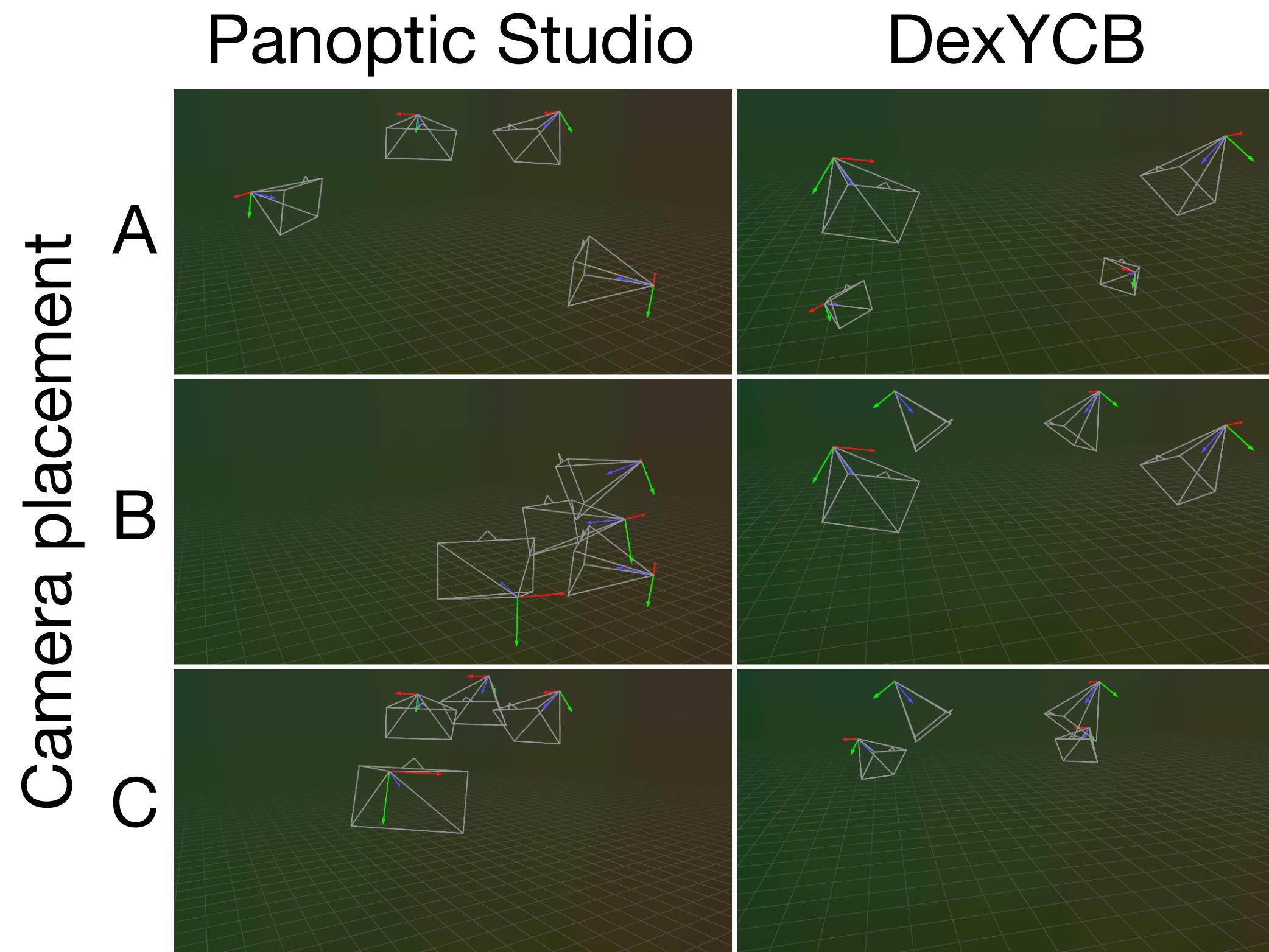
Accuracy improves with more cameras



Stable results across camera placements



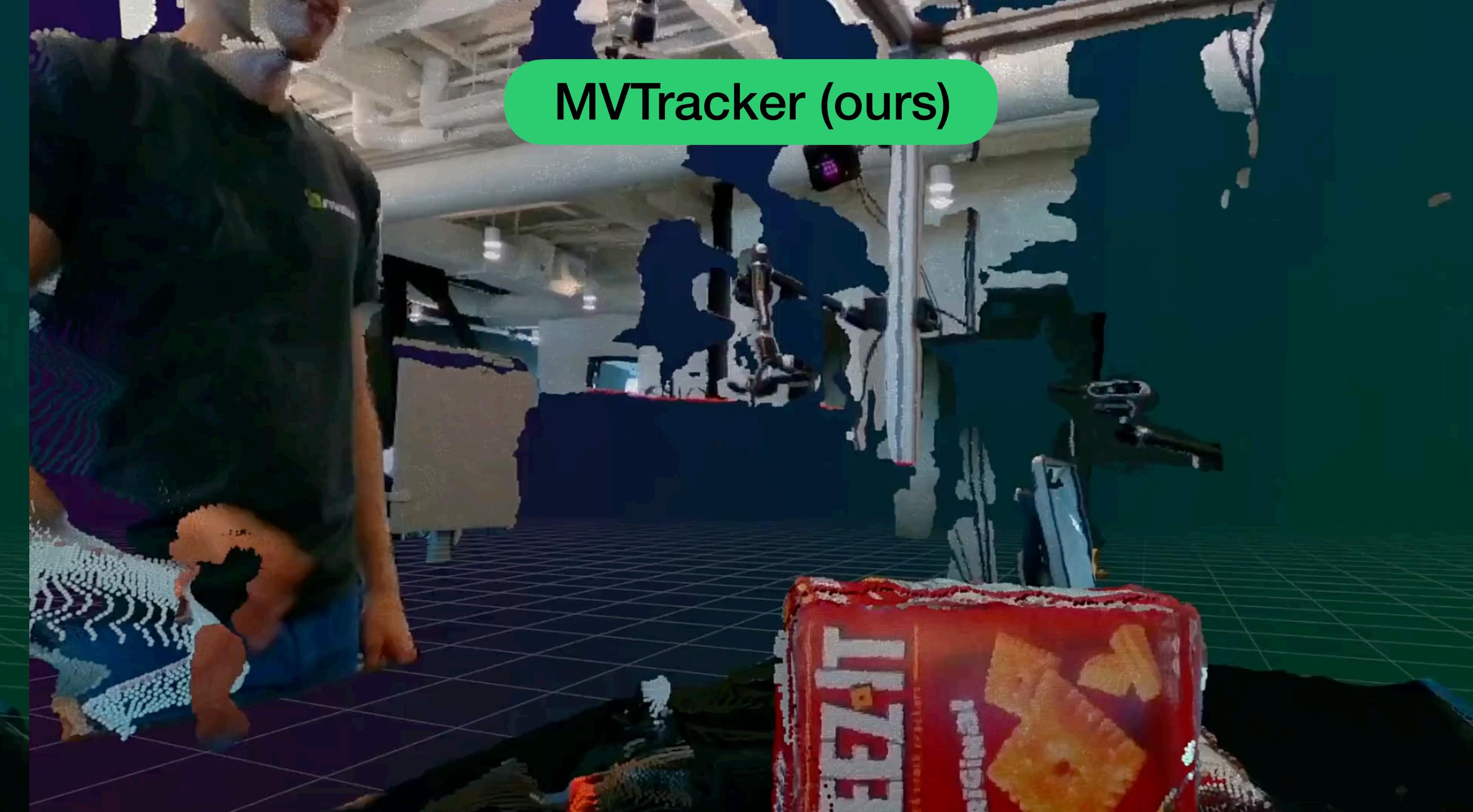
Stable results across camera placements



Method	PStudio [18]			DexYCB [3]		
	A	B	C	A	B	C
Dynamic 3DGS [23]	66.5	50.8	56.6	45.7	—	—
Shape of Motion [35]	72.6	64.3	66.8	36.2	—	—
LocoTrack [5]	65.8	57.9	63.7	27.8	40.9	42.9
DELTA [24]	68.1	61.1	65.9	36.5	43.3	47.6
CoTracker2 [16]	69.5	62.3	66.4	28.8	42.0	44.4
CoTracker3 [17]	74.5	66.3	70.9	29.4	43.8	46.3
SpaTracker [38]	61.5	54.8	57.8	58.3	57.9	63.8
Triplane Baseline	65.1	59.9	63.5	57.5	62.0	66.3
MVTracker (ours)	86.0	75.7	83.2	71.0	71.2	78.3

+15.4%

+18.1%



Red lines indicate distance to ground truth.

Ground Truth



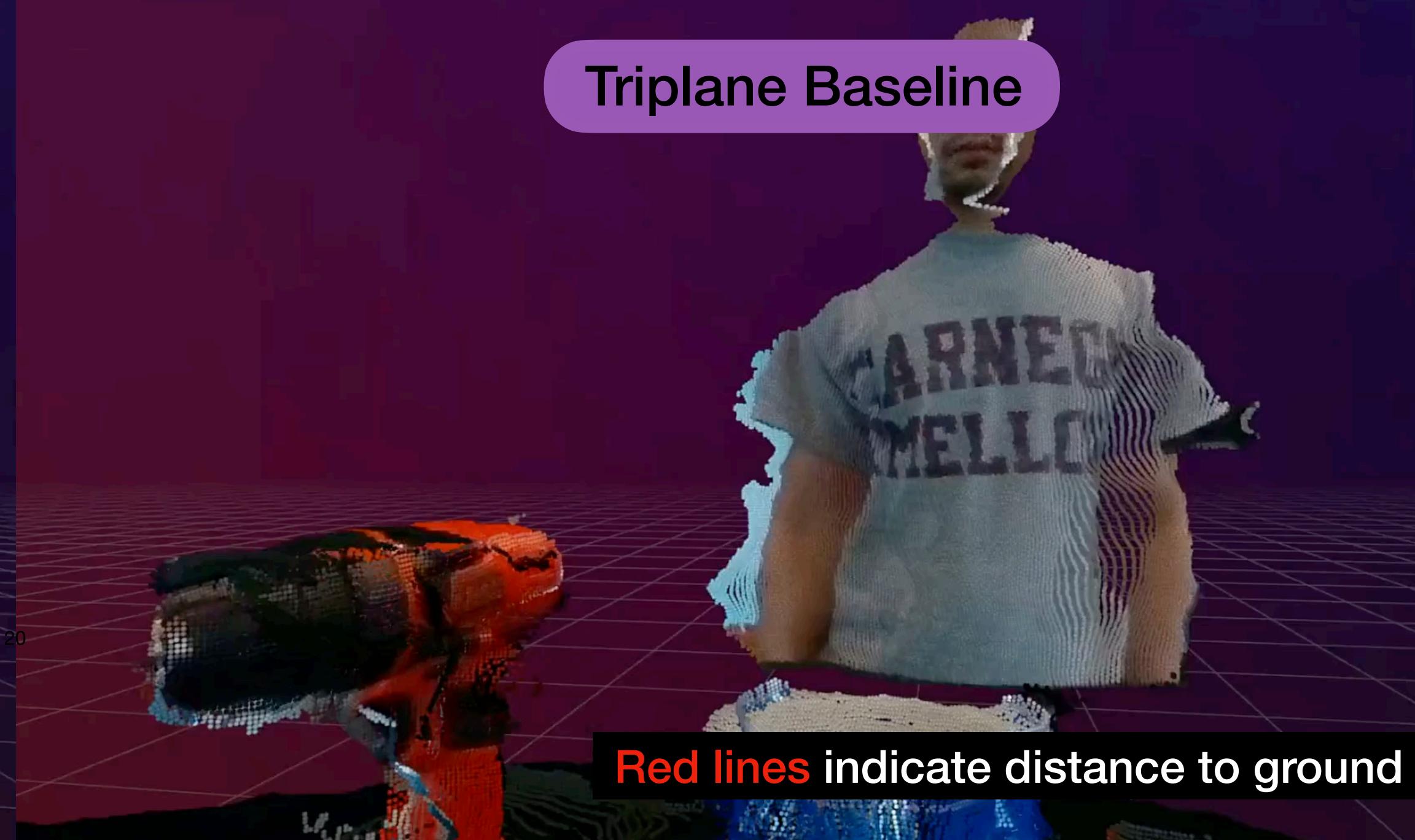
MVTracker (ours)



SpatialTrackerV1



Triplane Baseline



Red lines indicate distance to ground truth.

Depth source: Blender simulation

4 Cameras (synchronized)

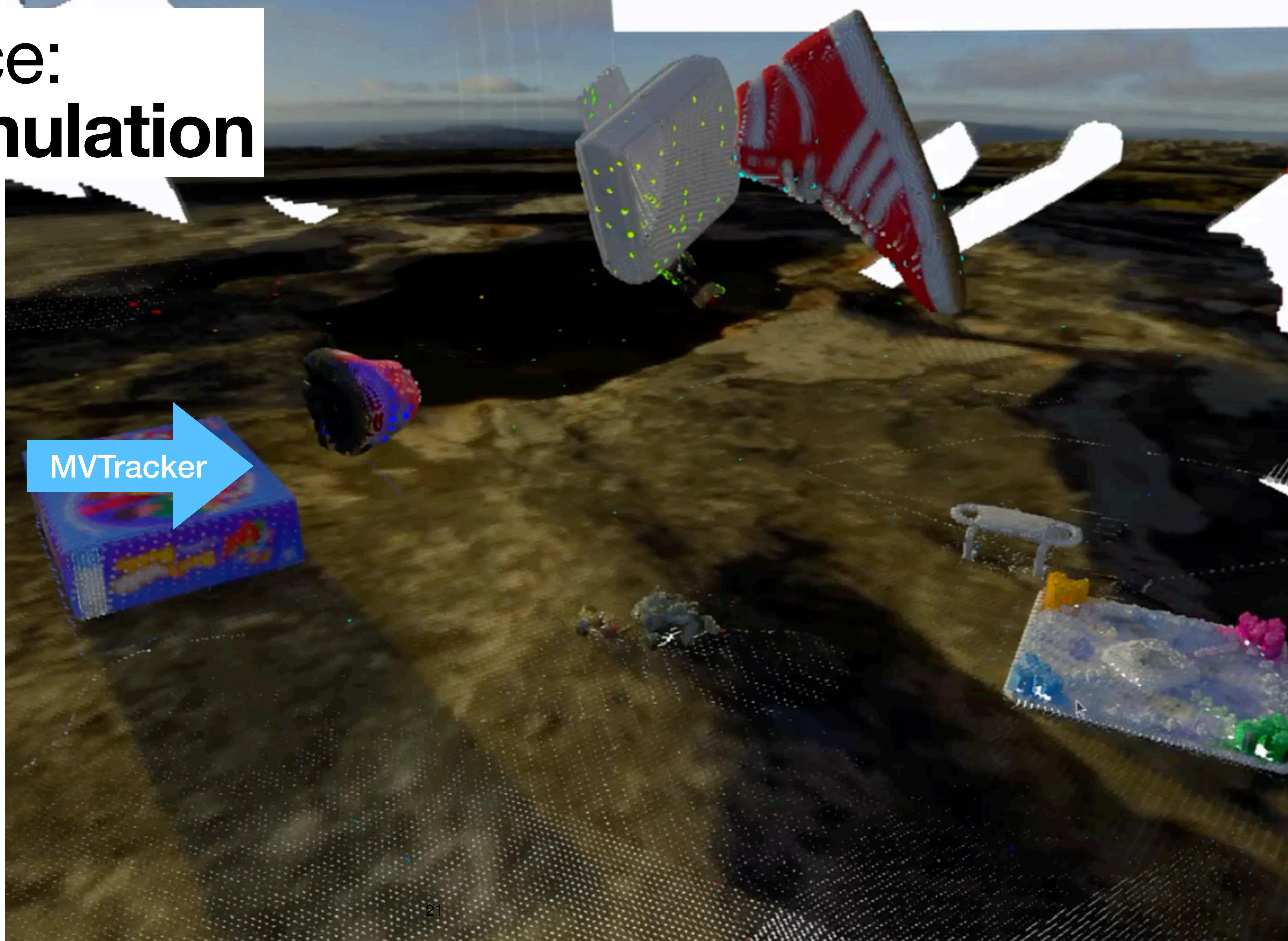
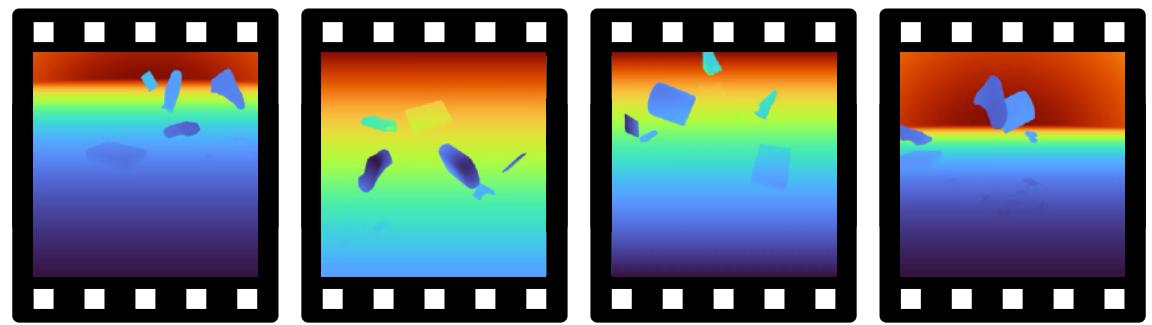


+

Camera Poses (calibrated)

+

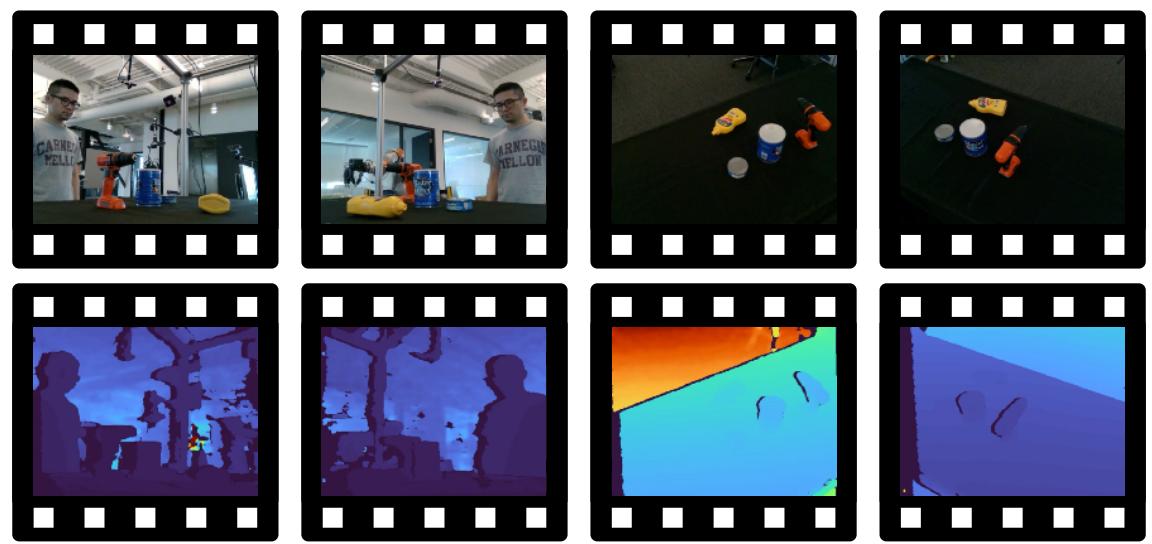
Blender-simulated Depth



Dataset: MV-Kubric

Depth source: Kinect camera

4 Kinect Cameras
(RGB + Depth)



MVTracker

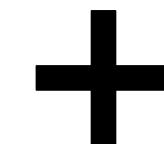
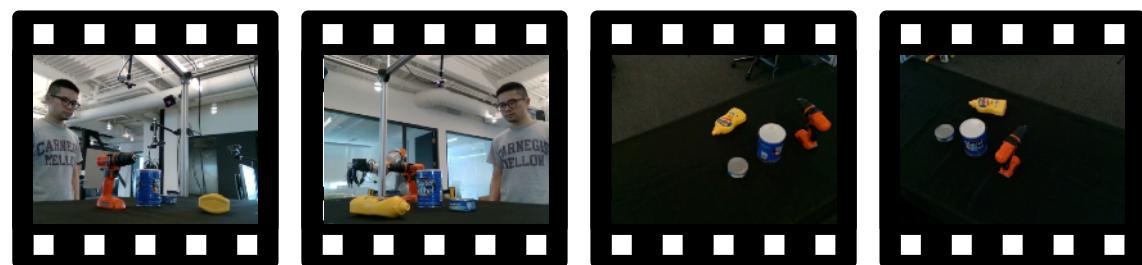
+

Camera Poses (calibrated)

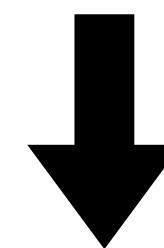
Dataset: DexYCB

Depth source: DUSt3R¹ estimates

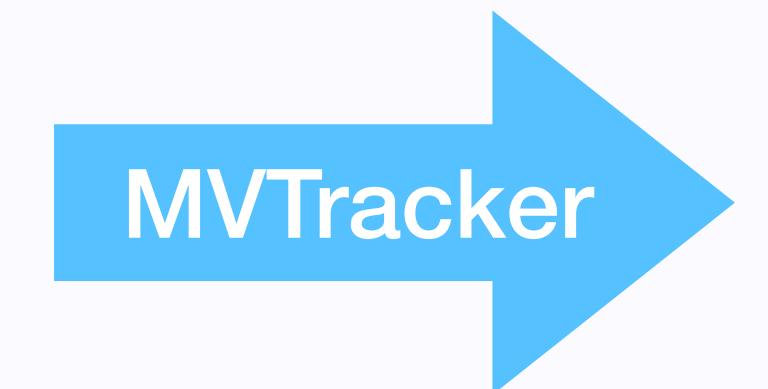
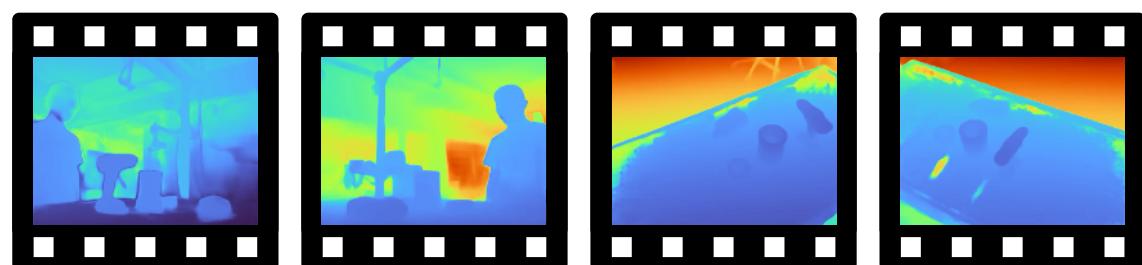
4 Cameras (synchronized)



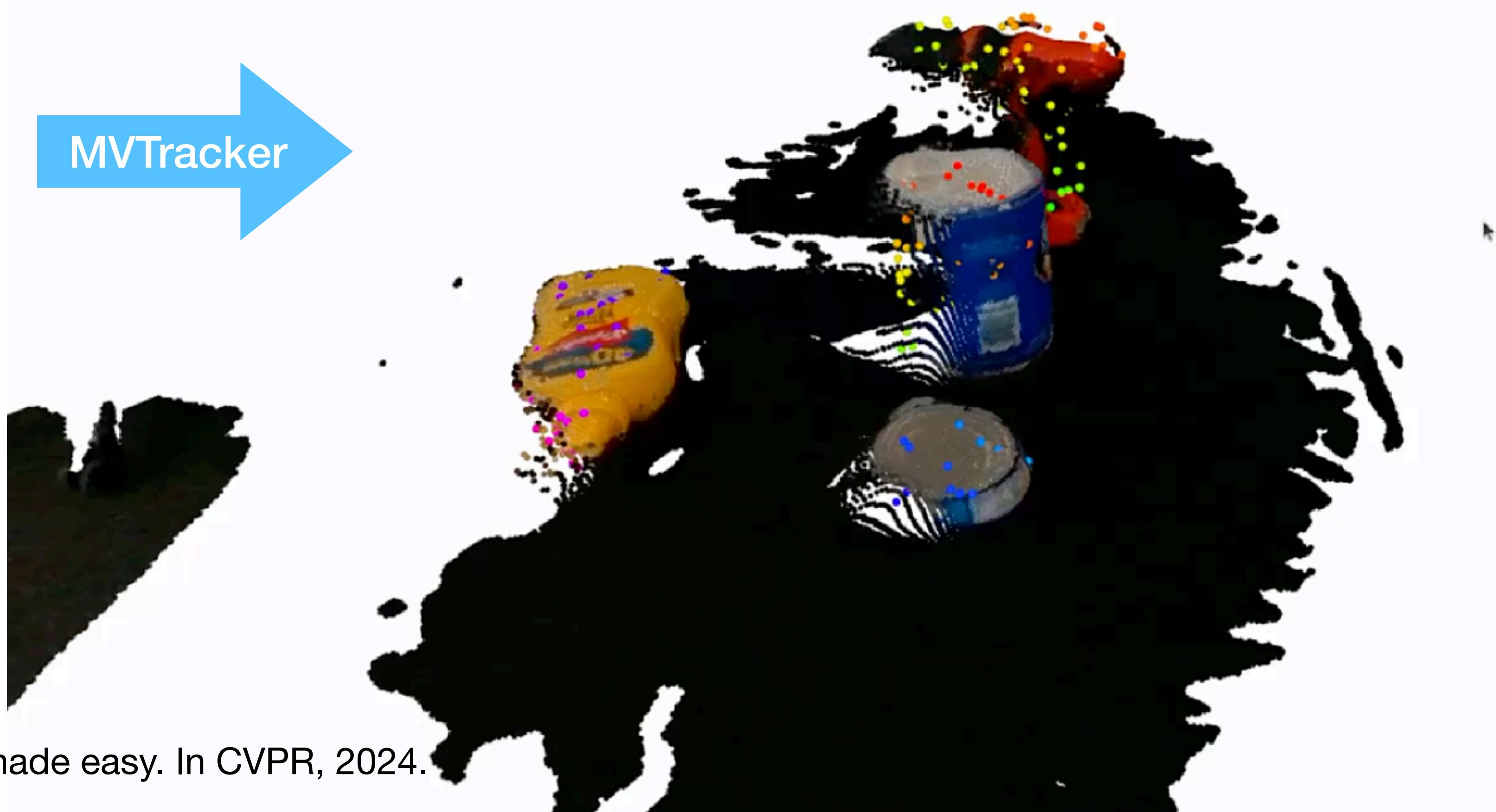
Camera Poses (calibrated)



DUSt3R¹-estimated Depth



Dataset: DexYCB



¹DUSt3R: Geometric 3D vision made easy. In CVPR, 2024.

Depth source: Studio capture

4 Cameras (synchronized)



+

Camera Poses (calibrated)

+

Studio Capture Depth



MVTracker



Dataset: 4D-DRESS

Conclusion and key takeaways

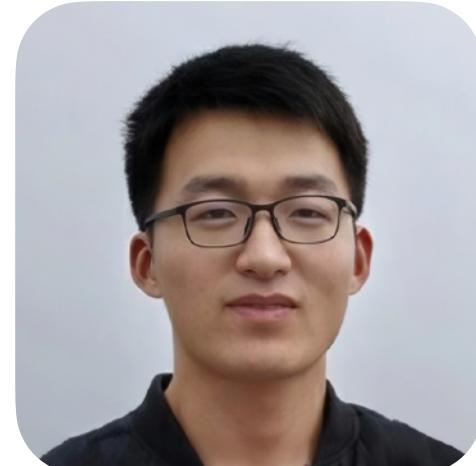
MVTracker (ours):

- First **data-driven multi-view** point tracker.
- Fuses multi-view features into a **point cloud**.
- **Robust** to number of cameras, camera rigging, depth source and noise.
- **Significant gains** over monocular and multi-view baselines.
- **Main limitations:**
 - Dependence on multi-view **depth estimators**.
 - Tested only within **bounded scenes**.
 - We need **more data**. More is more.

Frano
Rajić¹



Haofei
Xu¹



Marko
Mihajlović¹



Siyuan
Li¹



Irem
Demir¹



Emircan
Gündoğdu¹



Lei
Ke²



Sergey
Prokudin^{1,3}



Marc
Pollefeys^{1,4}



Siyu
Tang¹



¹ETH, ²CMU, ³Balgrist, ⁴Microsoft

Frequently asked questions

1. Can MVTracker run **online**? Yes, at 14.9 FPS using sliding windows.
2. Can tracked points only be added **at the first frame**? No, at any frame.
3. What is the **scale of the training dataset**? 5000 sequences of MV-Kubric.
4. Can MVTracker be used with a **stereo camera** / small baseline? Yes, but this would amount to monocular 3D point tracking (after stereo depth estimation).
5. Does MVTracker regress **invisible point locations**?
During training yes, but we didn't benchmark this.



<https://ethz-vlg.github.io/mvtracker/>