

# Pose Estimation Beyond Feature Matching

Eric Brachmann





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[ebrach.github.io](https://ebrach.github.io)



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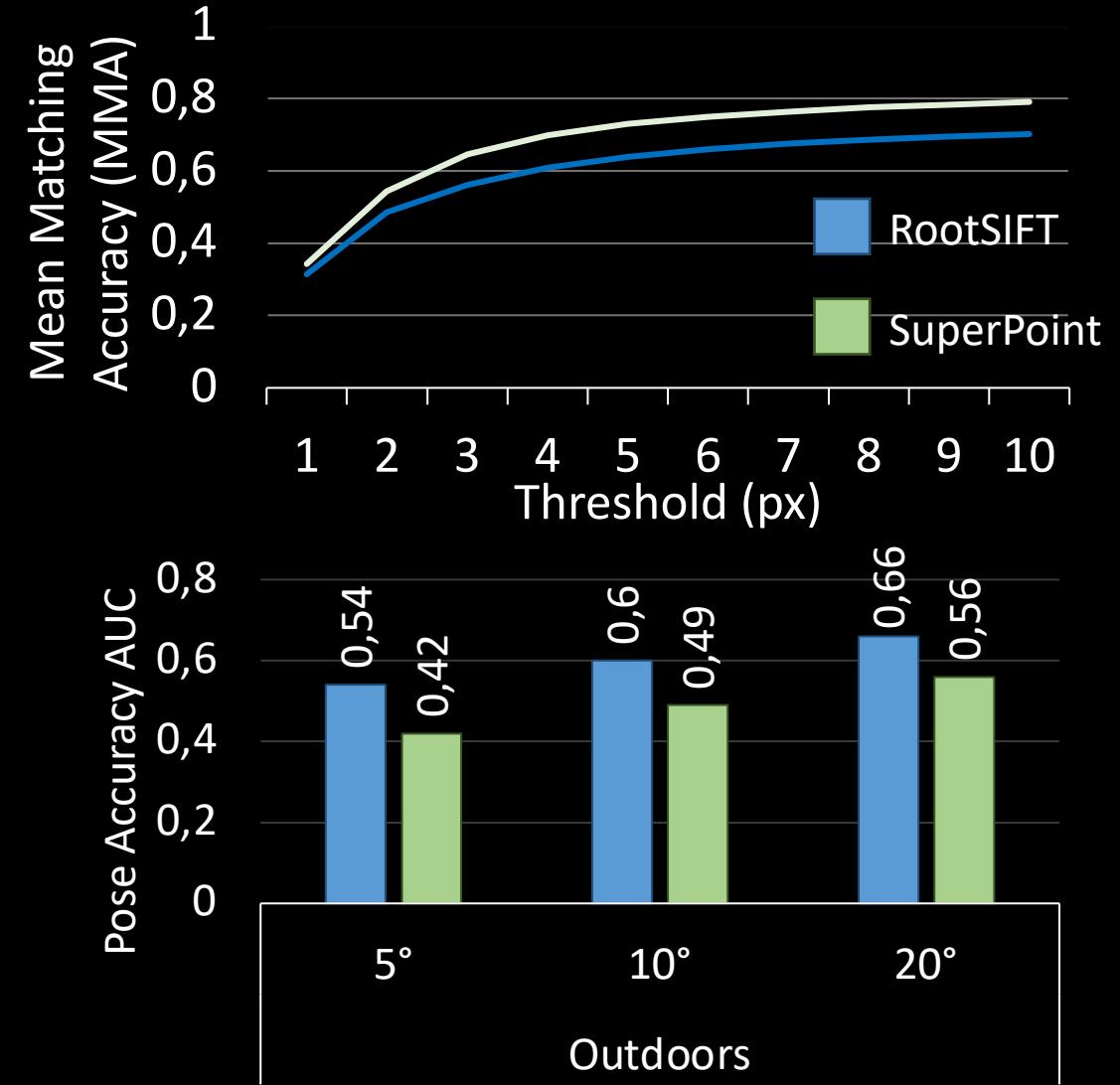


[ebrachmann@nianticlabs.com](mailto:ebrachmann@nianticlabs.com)



[linkedin.com/in/eric-brachmann](https://linkedin.com/in/eric-brachmann)

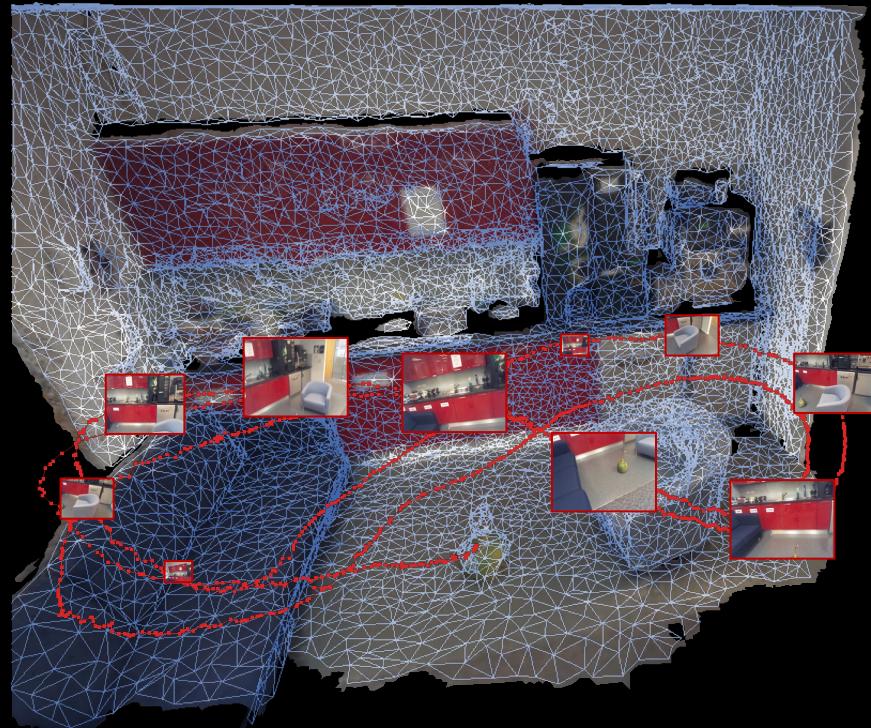




"Reinforced Feature Points: Optimizing Feature Detection and Description for a High-Level Task" Bhowmik et al., CVPR'20 (oral)

## Mapping

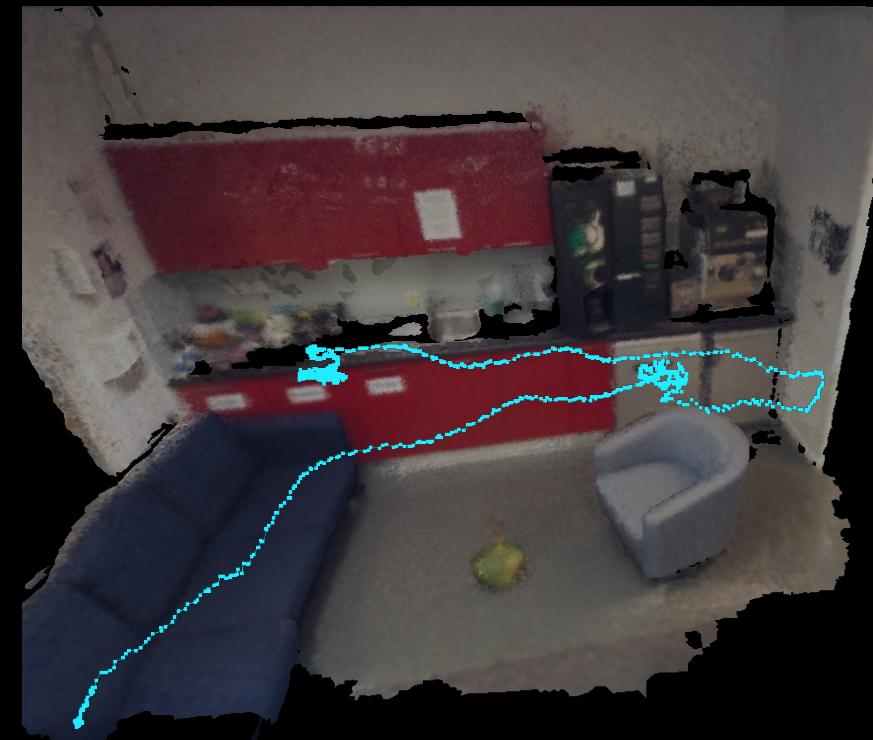
Build the Scene Representation



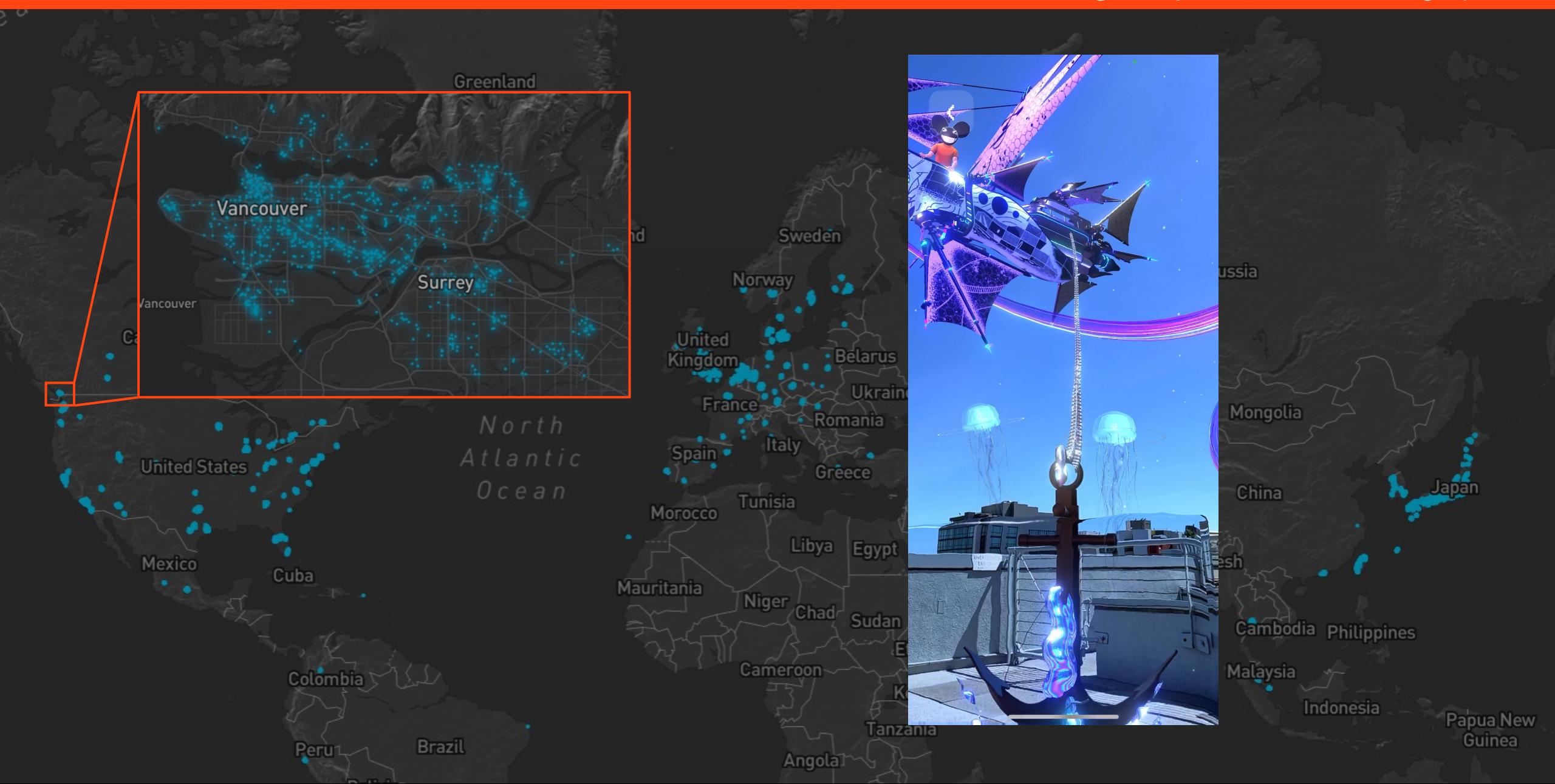
Obtain Posed Images  
Triangulate Scene  
Bundle Adjustment

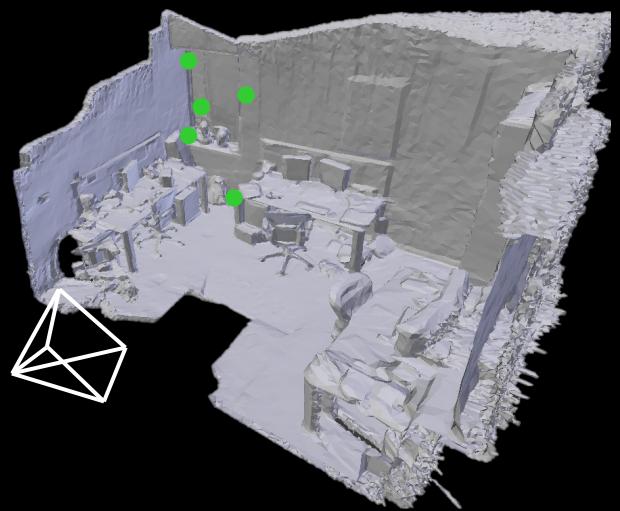
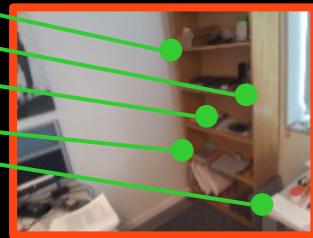
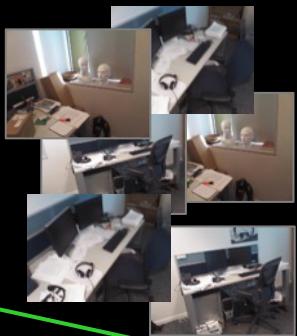
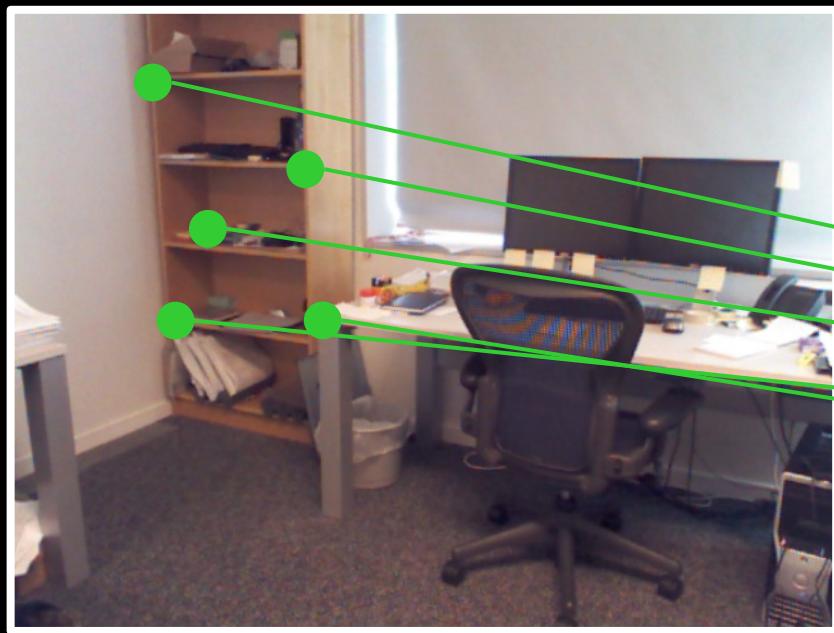
## Re-Localisation

Register Query Frames



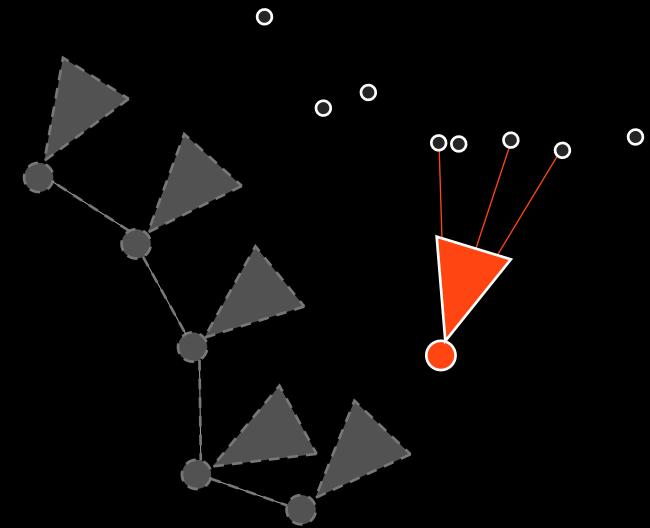
NN Retrieval  
Discrete Feature Matching  
Pose Optimization



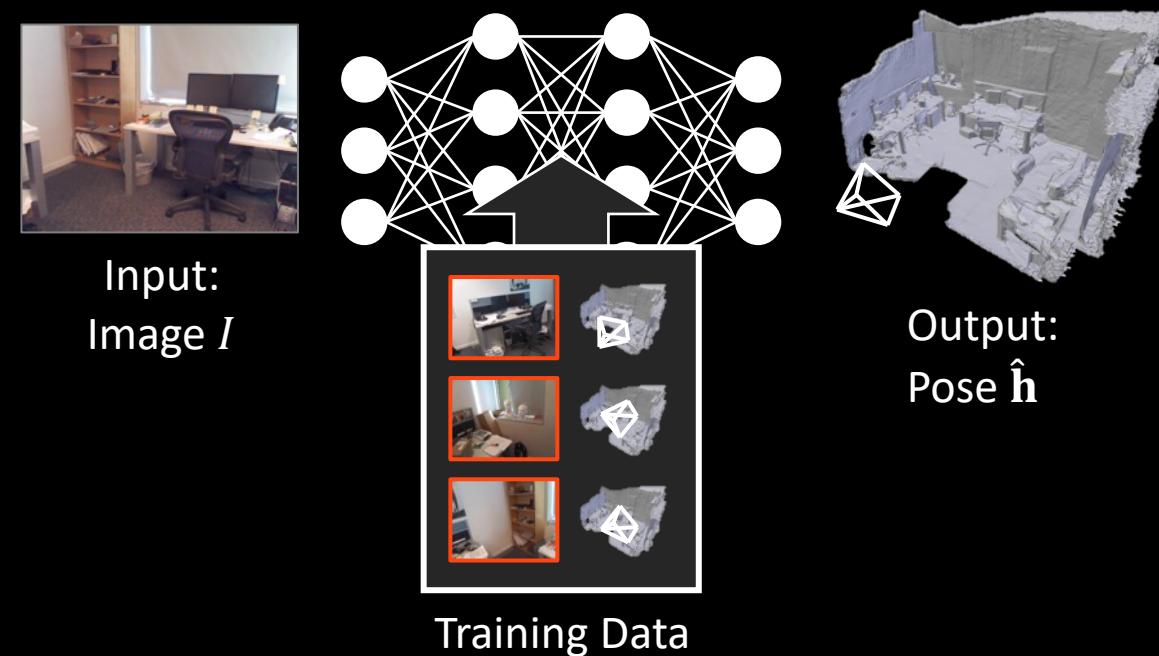
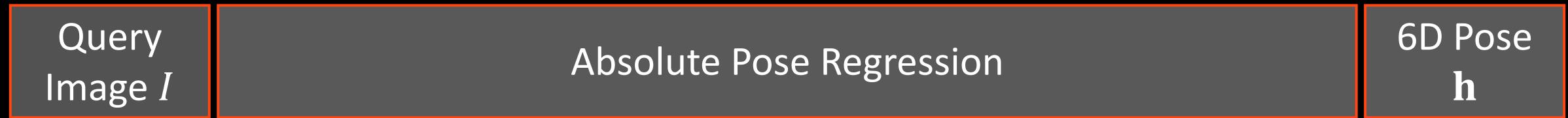
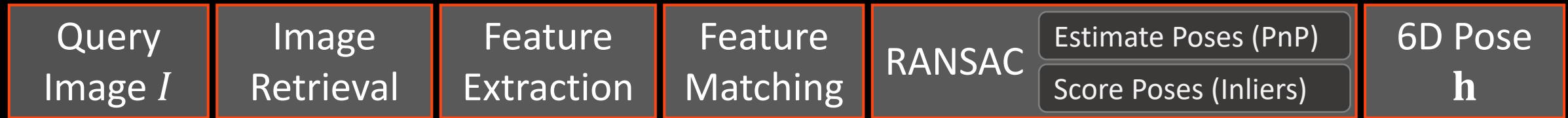


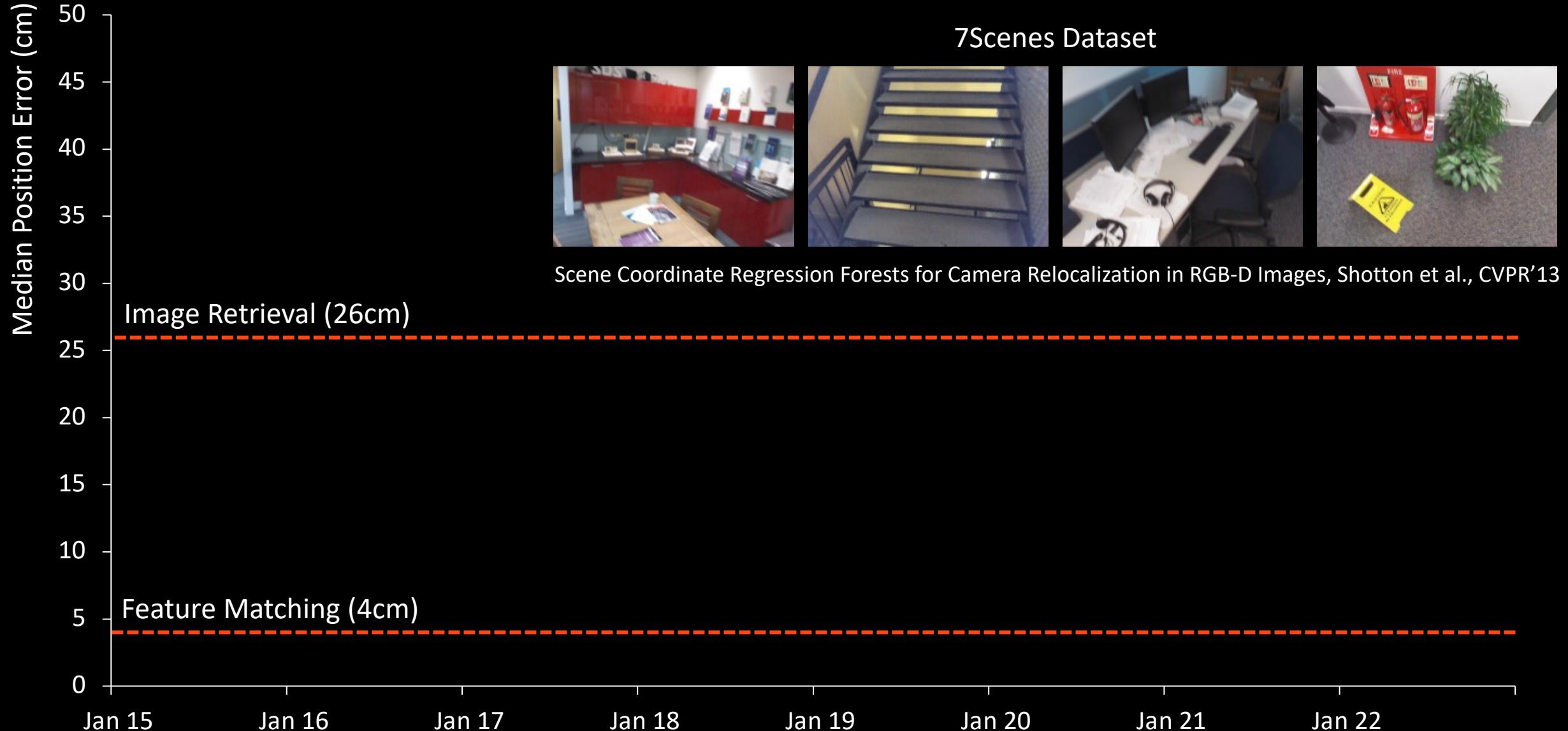
e.g. "From Coarse to Fine: Robust Hierarchical Localization at Large Scale", Sarlin et al., CVPR'19

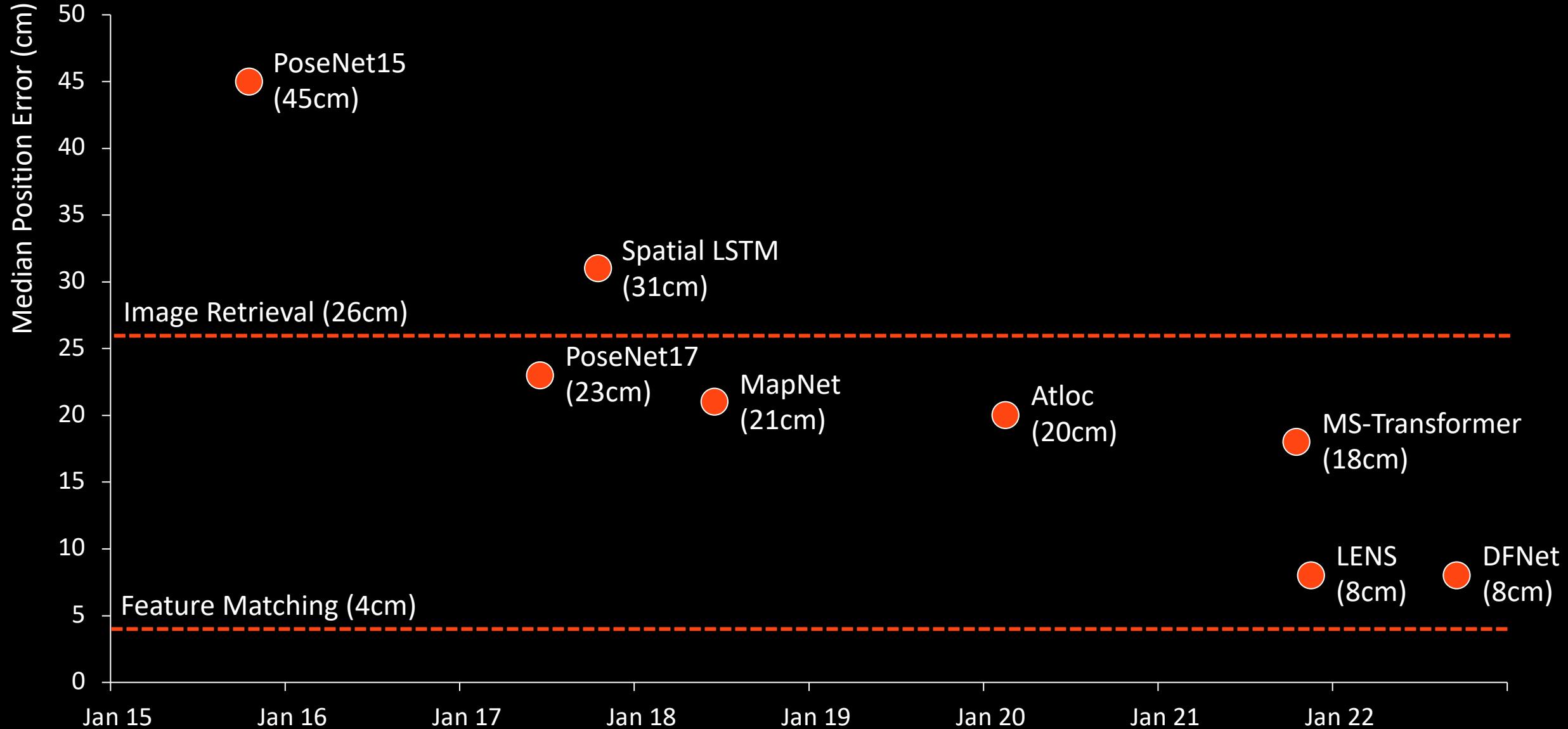
## Feature Matching



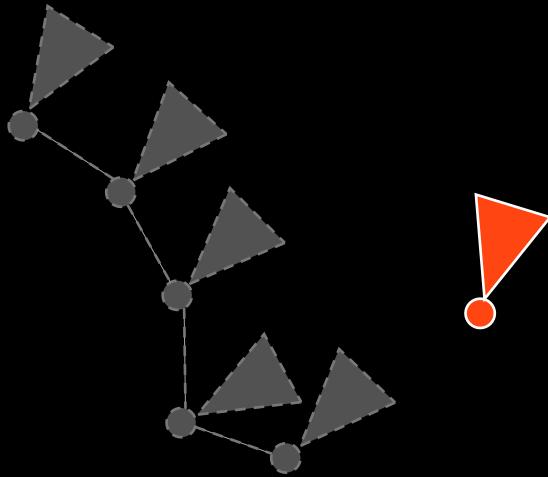
- Moderate query time
- Slow at mapping time
- High Memory Demand
- High Accuracy





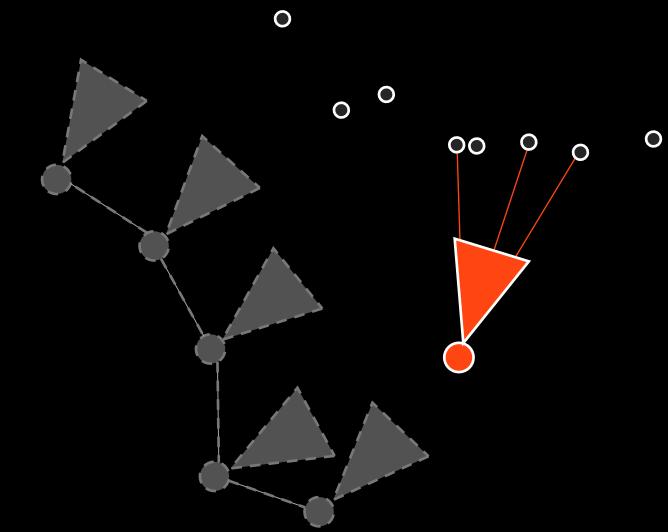


## Absolute Pose Regression

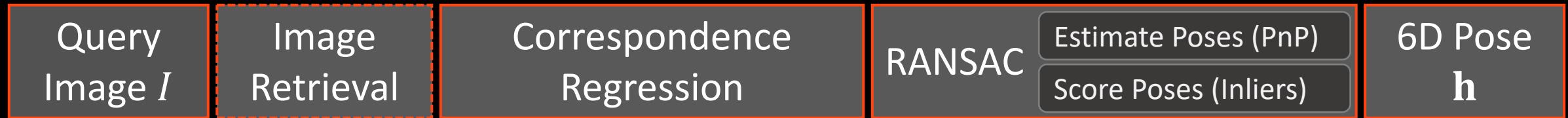
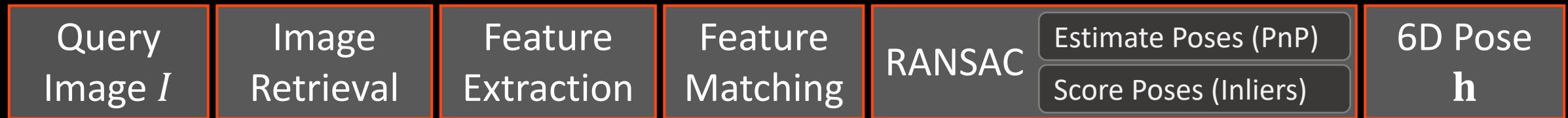


- Fast at query time
- Slow at mapping time
- Moderate memory demand
- Moderate accuracy

## Feature Matching



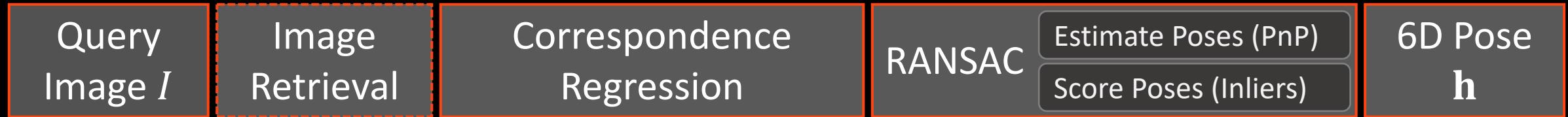
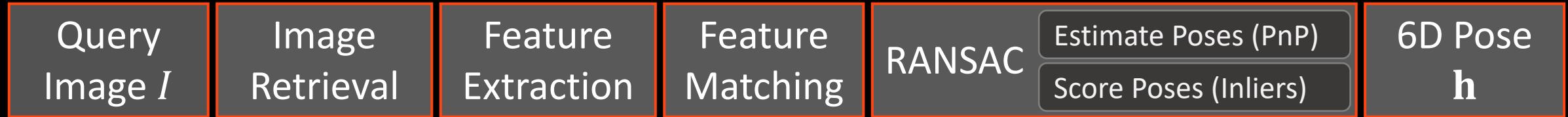
- Moderate query time
- Slow at mapping time
- High memory demand
- High accuracy



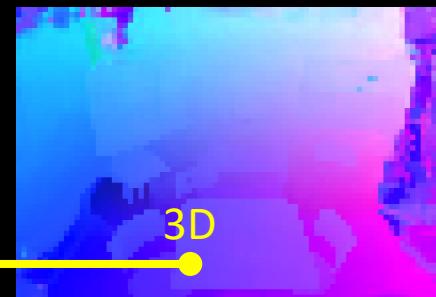
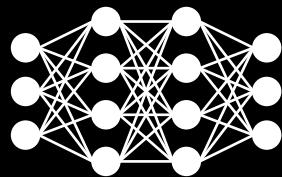
Query Image



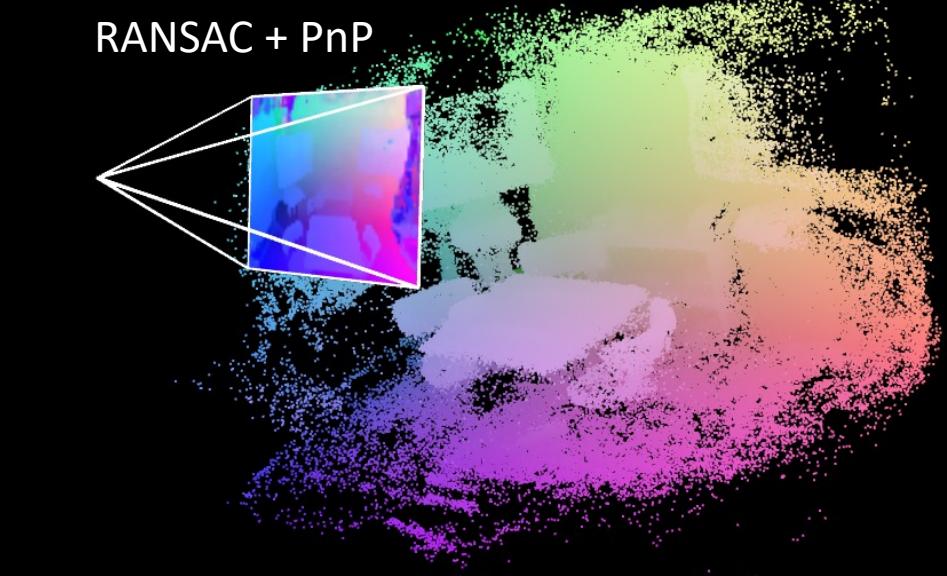
Scene Coordinate Regression Forests for Camera Relocalization in RGB-D Images, Shotton et al., CVPR'13



Query Image

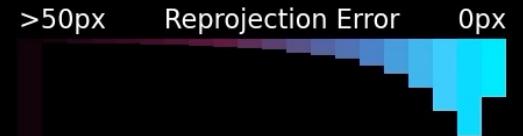


Scene Coordinate Prediction



Scene Coordinate Regression Forests for Camera Relocalization in RGB-D Images, Shotton et al., CVPR'13

Accelerated Coordinate Encoding: Learning to Relocalize in Minutes using RGB and Poses  
Brachmann, Cavallari, Prisacariu (CVPR23 Highlight) – TUE PM 86

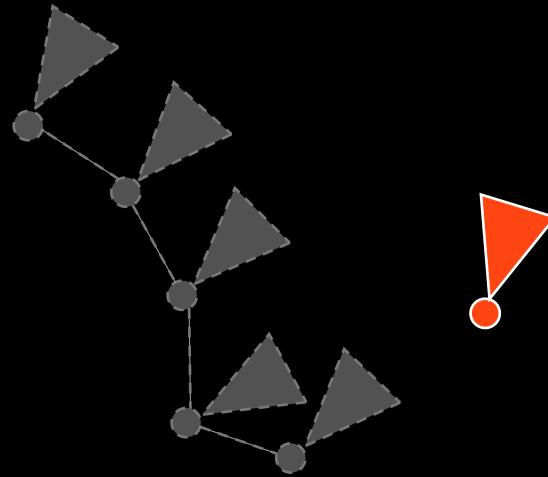


# Accelerated Coordinate Encoding

Training Time: 0:00:08.66

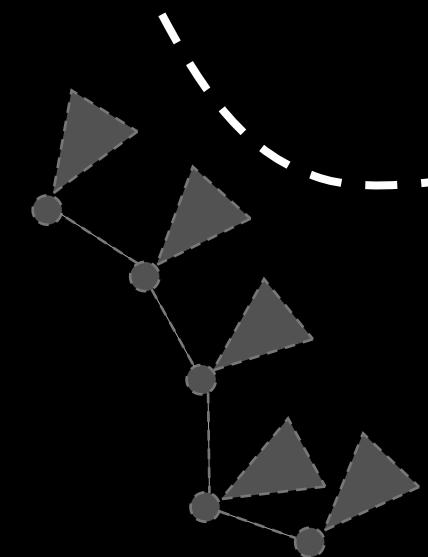


## Absolute Pose Regression



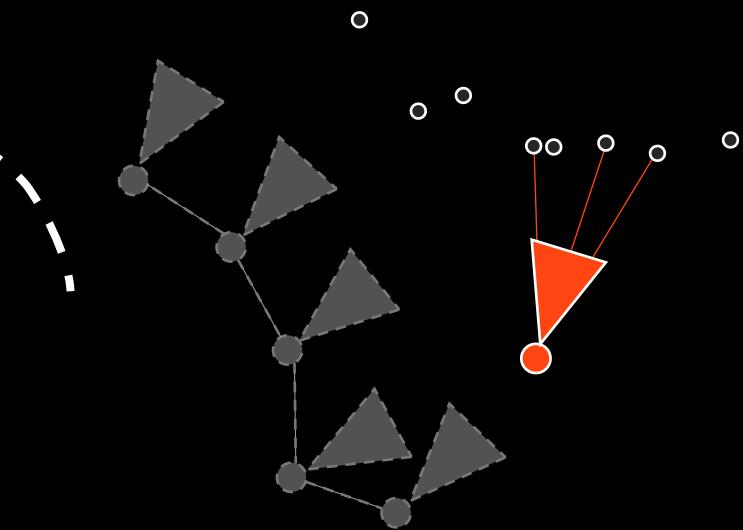
- Fast at query time
- Slow at mapping time
- Moderate memory demand
- Moderate accuracy

## Correspondence Regression

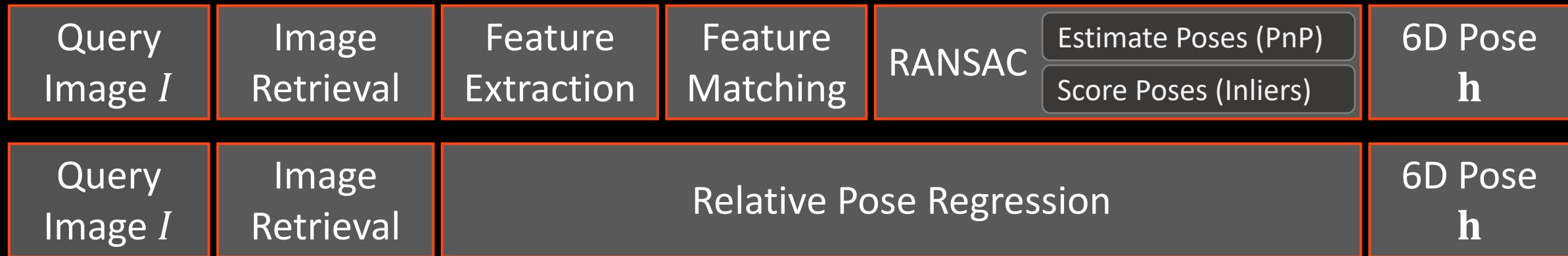


- Fast at query time
- Moderate mapping time
- Low memory demand
- High accuracy

## Feature Matching

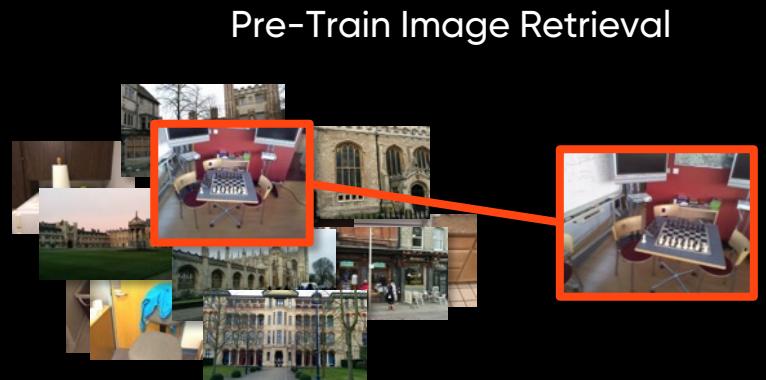


- Moderate query time
- Slow at mapping time
- High memory demand
- High accuracy



## Preparation

### Scene-Agnostic Training



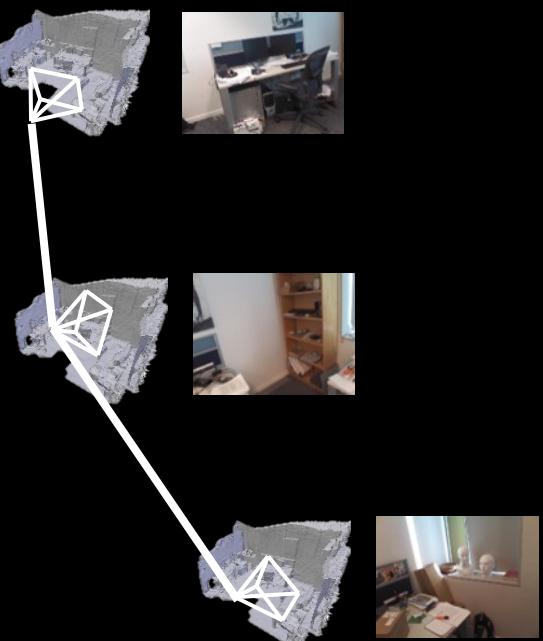
Pre-Train Relative Pose Regressor



## Mapping

### Scene-Specific Training

Obtain Posed Images  
Build Retrieval Index



**Mapping Time: Very Low**

## Re-Localisation

### Evaluation

Retrieve NN  
Refine pose



## Preparation

### Scene-Agnostic Training



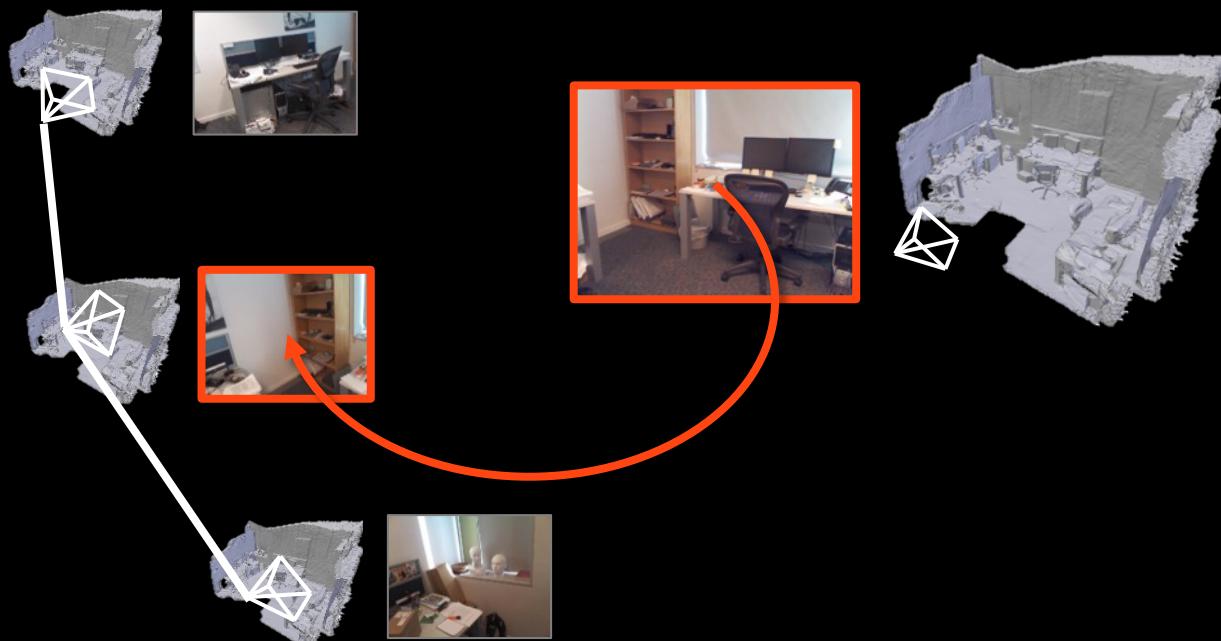
Pre-Train Relative Pose Regressor



## Mapping

### Scene-Specific Training

Obtain Posed Images  
Build Retrieval Index

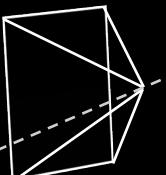


## Re-Localisation

### Evaluation

Retrieve NN  
Refine pose

Reference Image



Query Image

RelocNet

Median error on 7Scenes

Train on 7Scenes: 21cm

Train on ScanNet: 29cm

RelocNet: Continuous Metric Learning

Relocalisation using Neural Nets

Balntas et al., ECCV 2018

## Preparation

### Scene-Agnostic Training

Pre-Train Image Retrieval  
Pre-Train Relative Pose Regressor

## Mapping

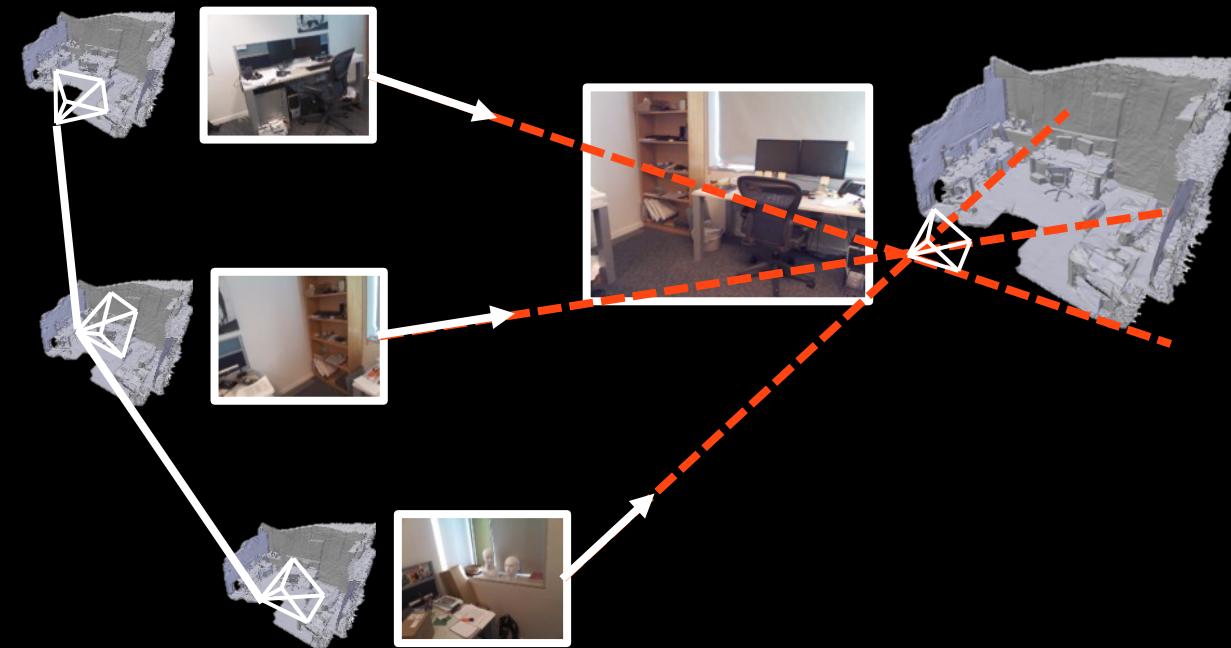
### Scene-Specific Training

Obtain Posed Images  
Build Retrieval Index

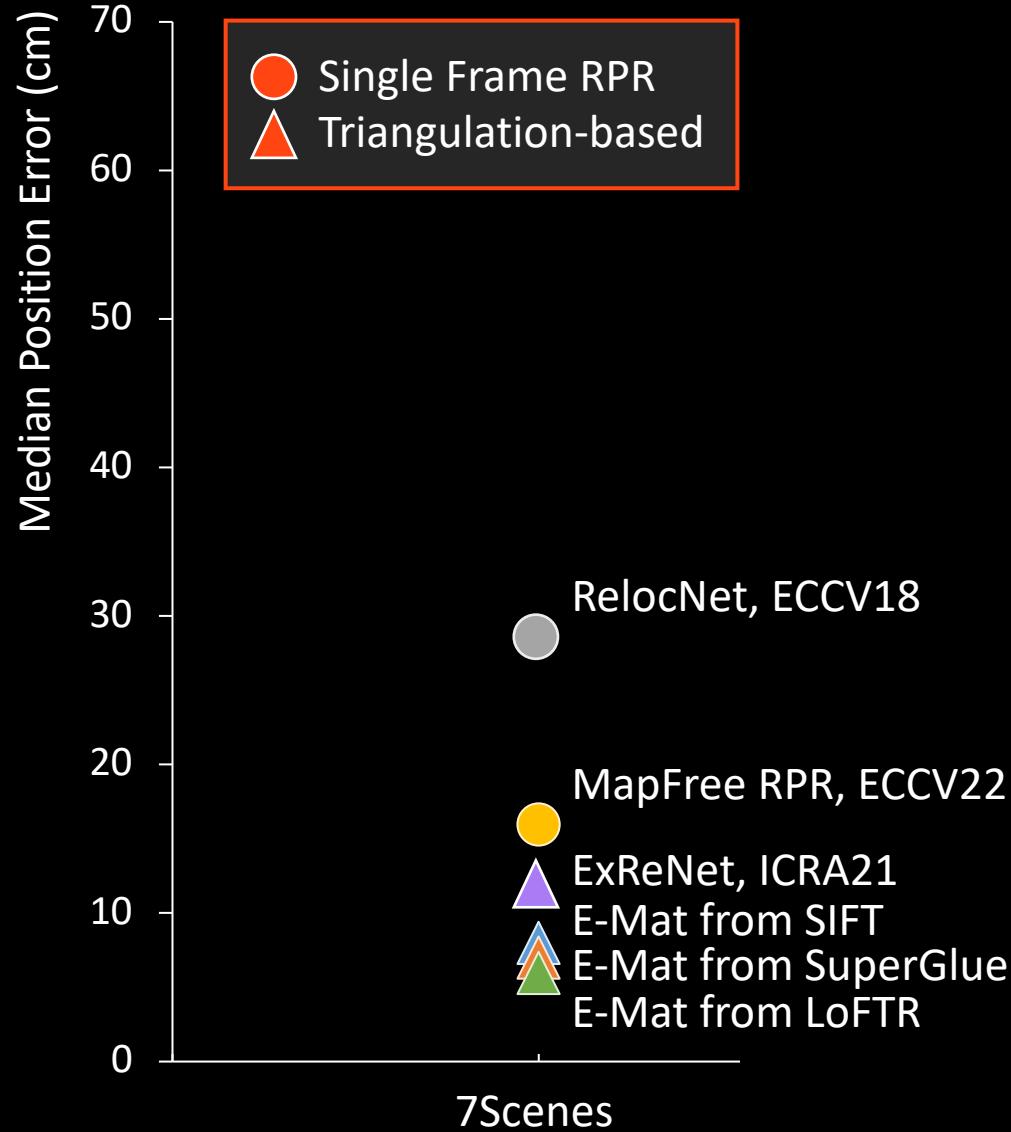
## Re-Localisation

### Evaluation

Retrieve NN  
Refine pose

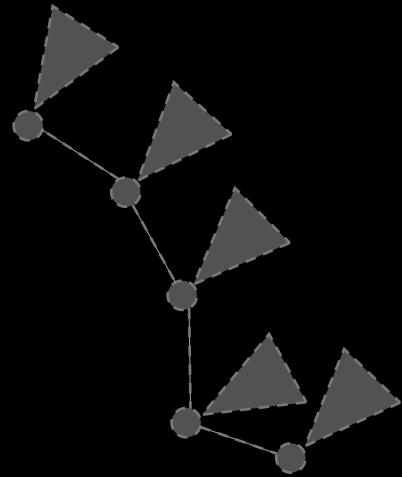


“To Learn or Not to Learn: Visual Localization from Essential Matrices”, Zhou et al., ICRA’20

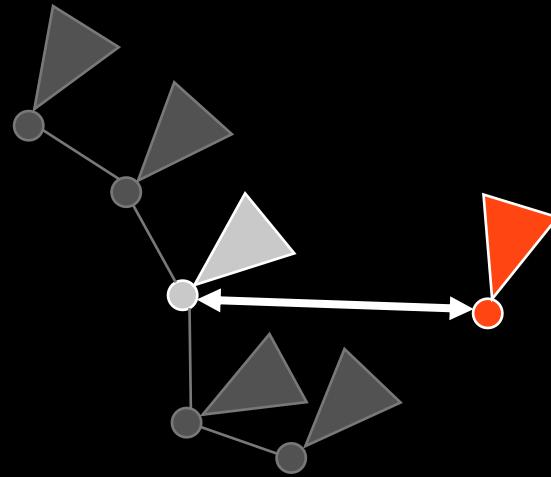


Data from Arnold et al., “Map-free Visual Relocalization: Metric Pose Relative to a Single Image”, ECCV22

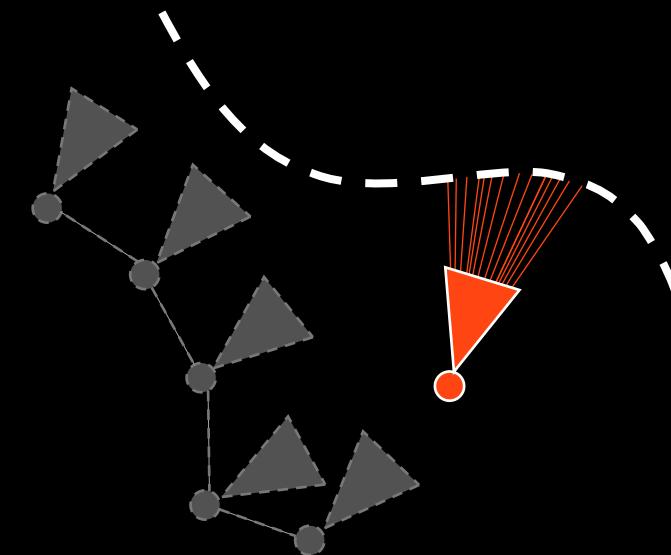
## Absolute Pose Regression



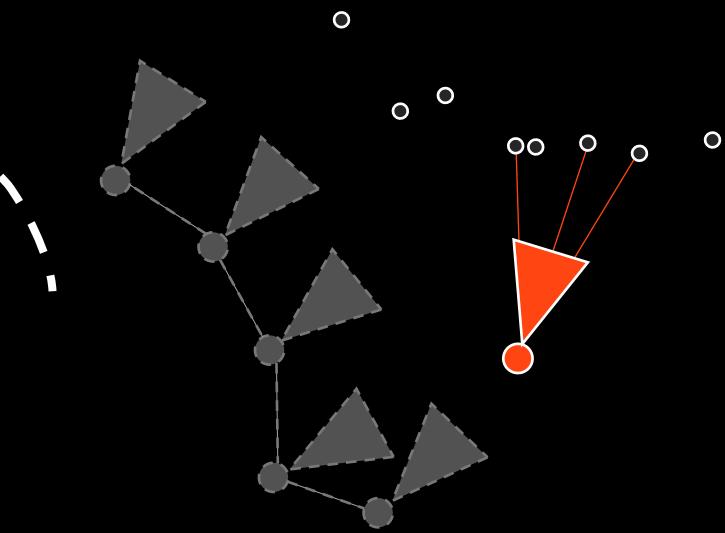
## Relative Pose Regression



## Correspondence Regression



## Feature Matching

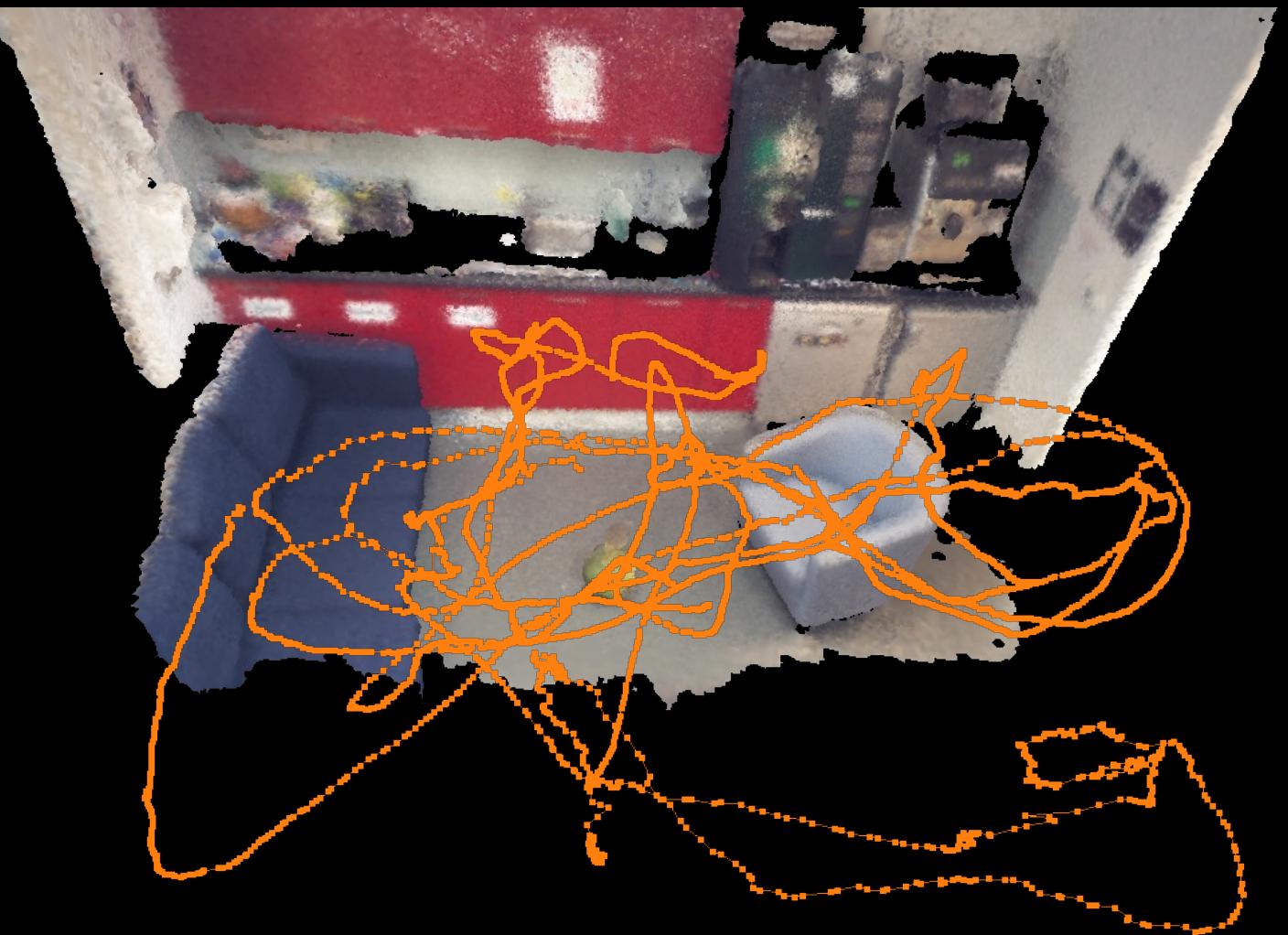
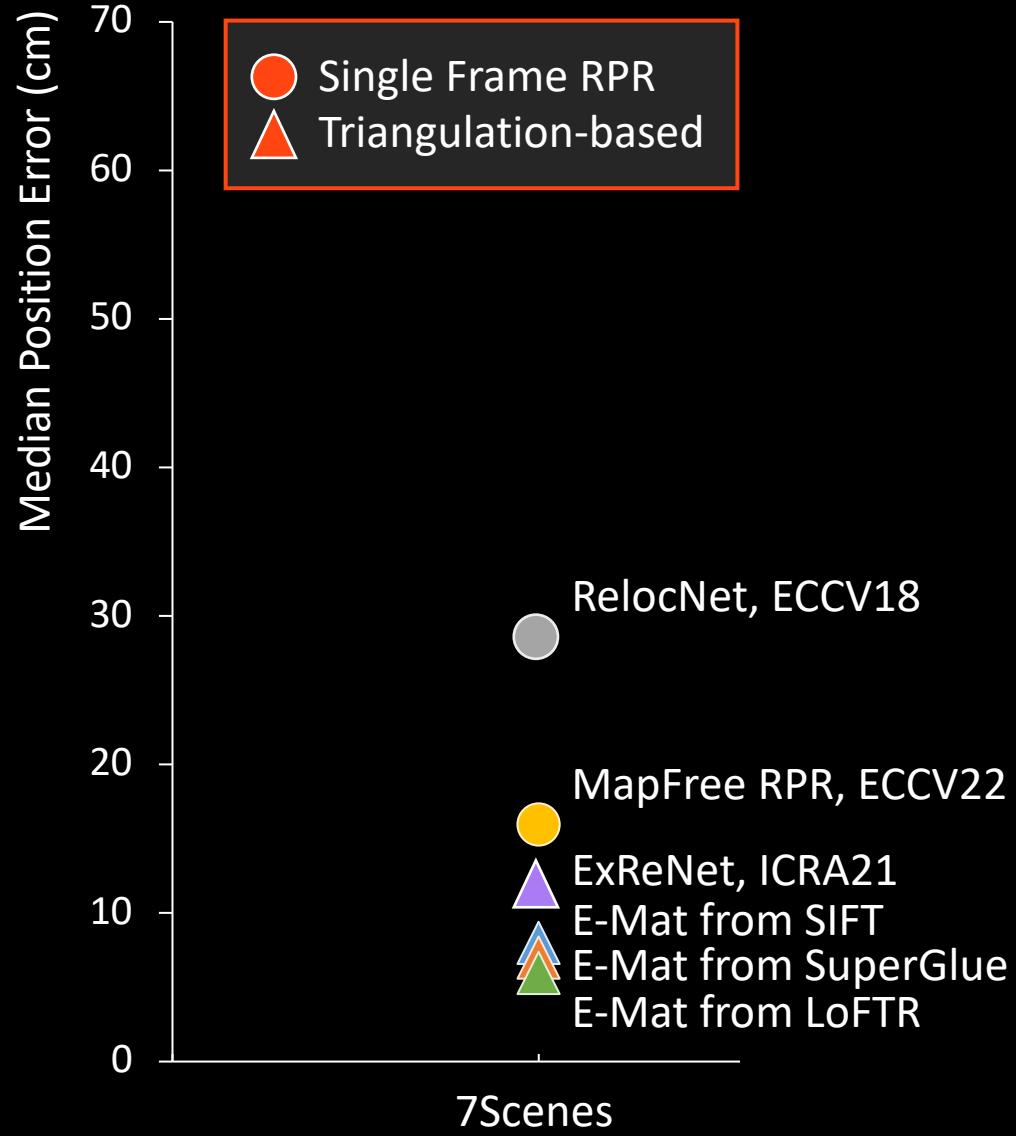


- Fast at query time
- Slow at mapping time
- Moderate memory demand
- Moderate accuracy

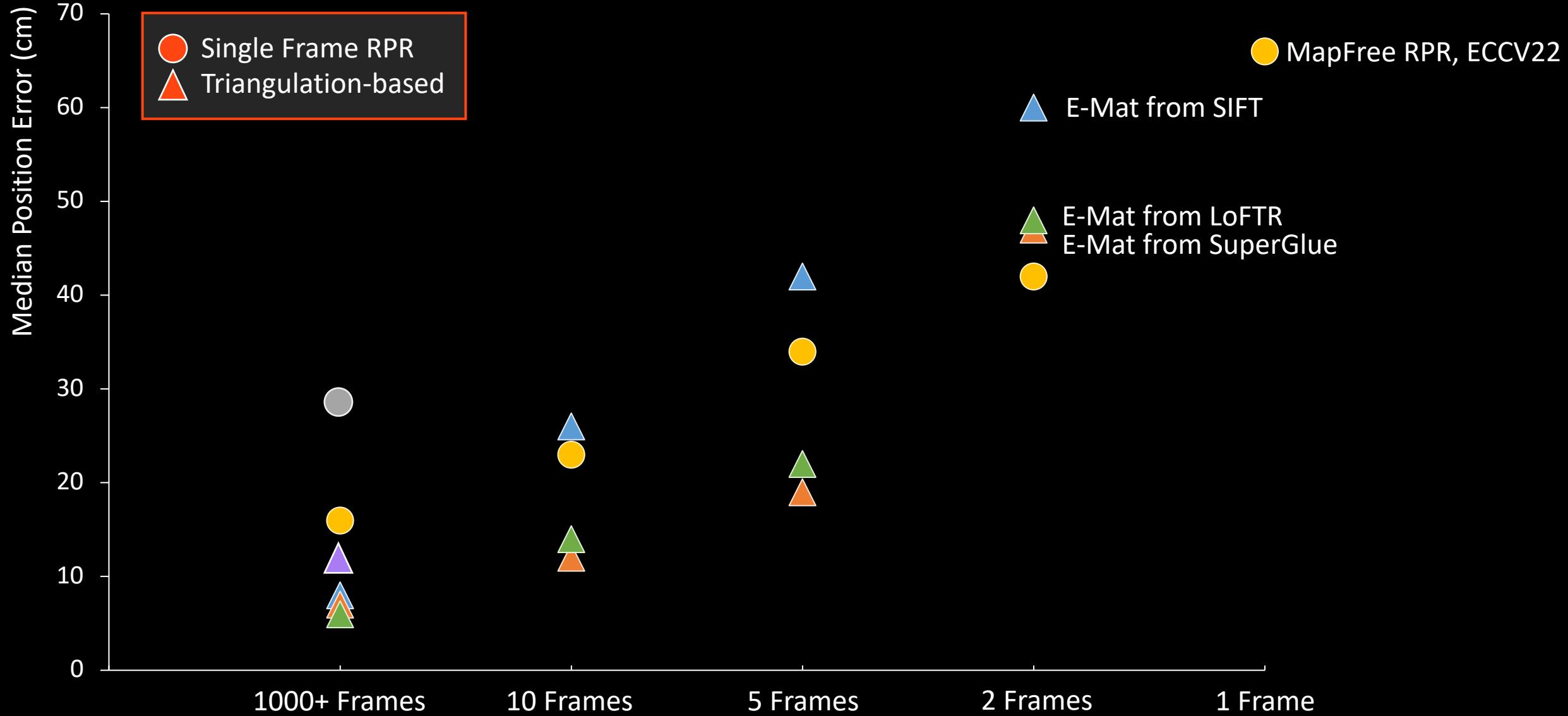
- Fast at query time
- Fast mapping time
- Low memory demand
- Moderate accuracy

- Fast at query time
- Moderate mapping time
- Low memory demand
- High accuracy

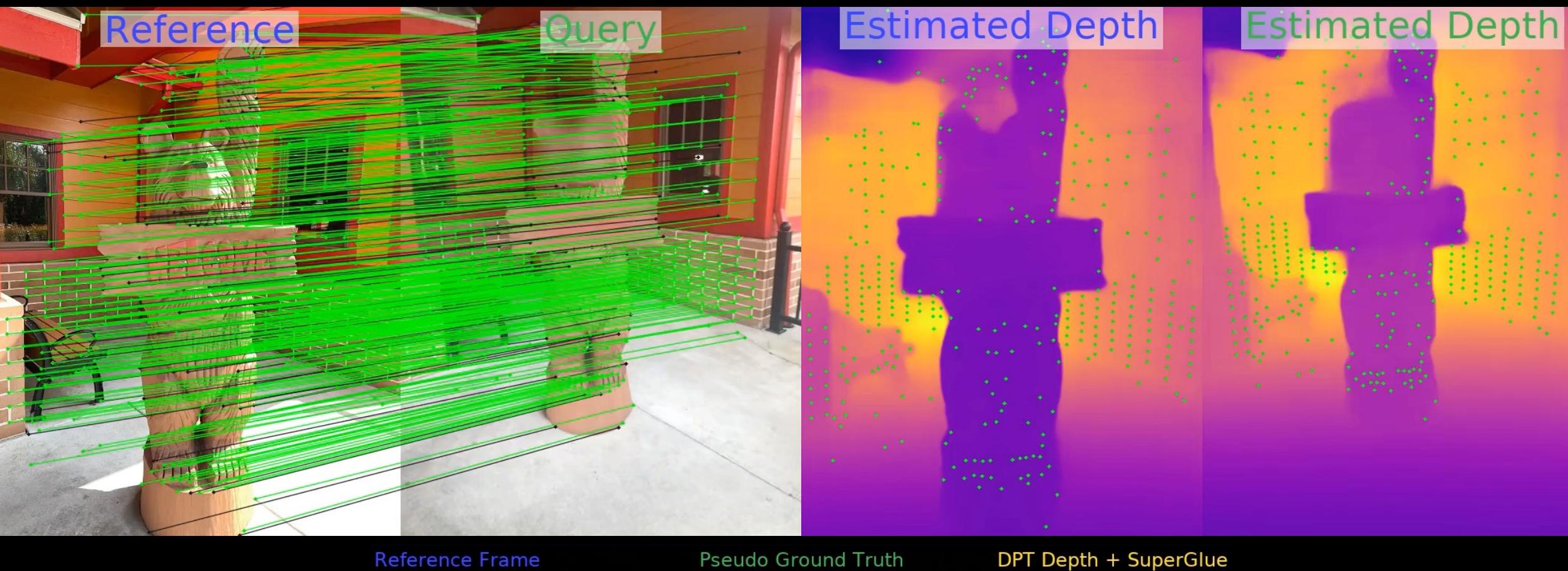
- Moderate query time
- Slow at mapping time
- High memory demand
- High accuracy

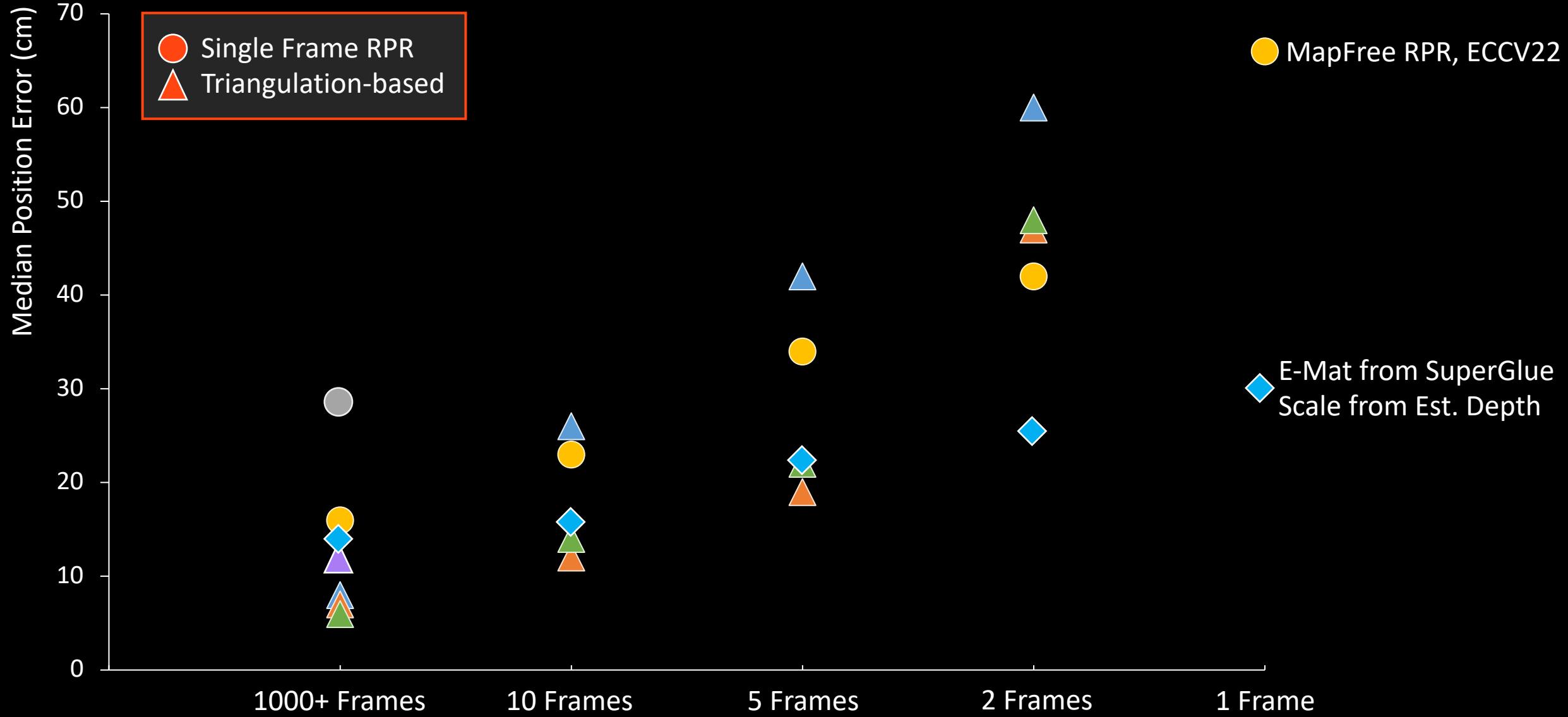


Data from Arnold et al., "Map-free Visual Relocalization: Metric Pose Relative to a Single Image", ECCV22



Data from Arnold et al., "Map-free Visual Relocalization: Metric Pose Relative to a Single Image", ECCV22





Data from Arnold et al., "Map-free Visual Relocalization: Metric Pose Relative to a Single Image", ECCV22

## Preparation

### Scene-Agnostic Training

Pre-Train Image Retrieval  
Pre-Train Relative Pose Regressor

## Mapping

### Scene-Specific Training

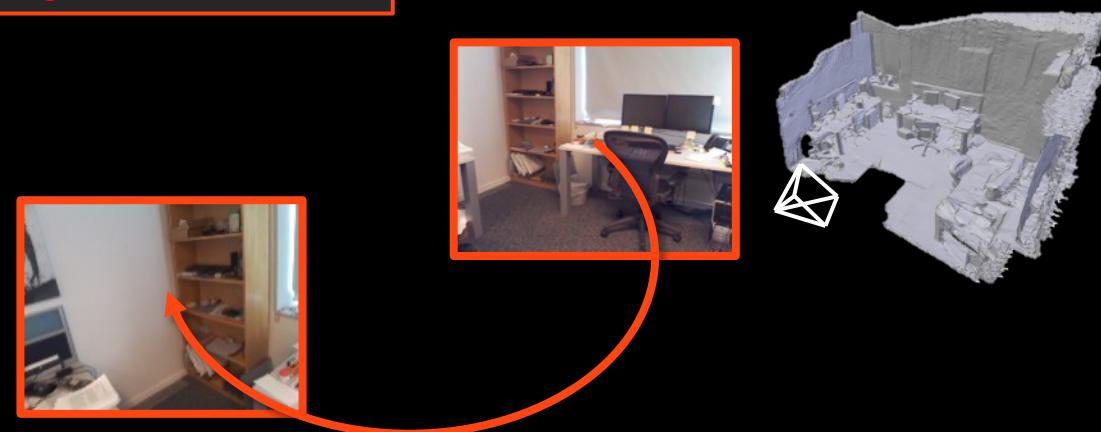
Obtain Posed Images  
Build Retrieval Index  
Shoot Reference Image

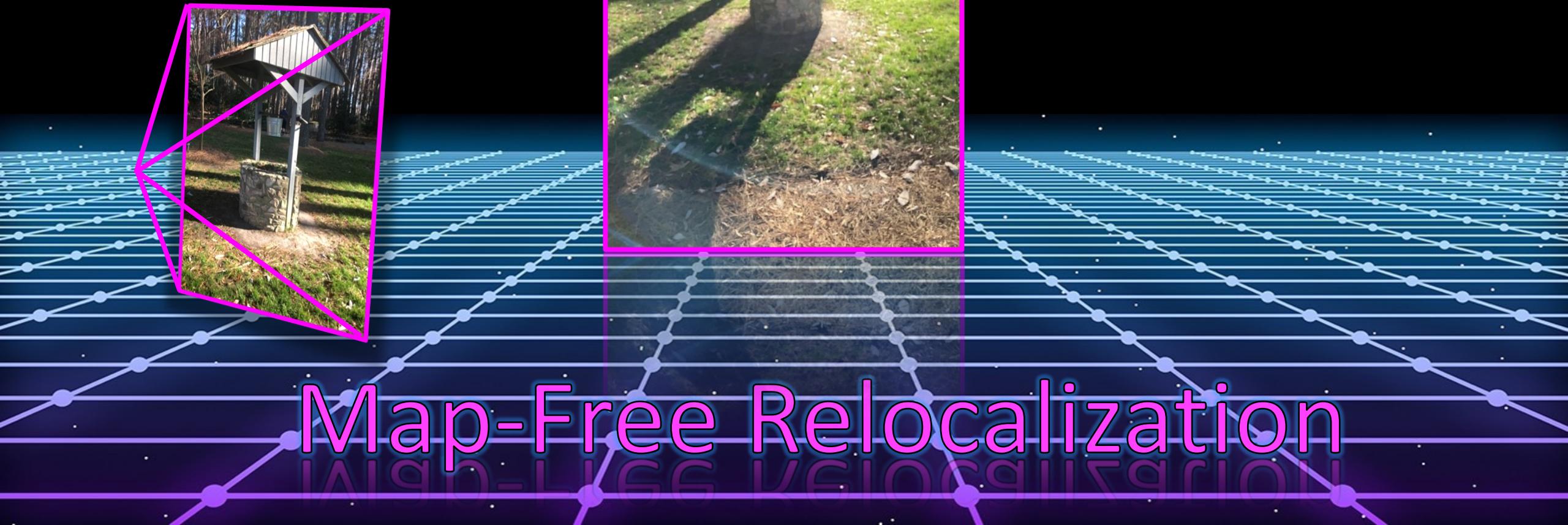
**Mapping Time: Zero**

## Re-Localisation

### Evaluation

Estimate Relative Pose





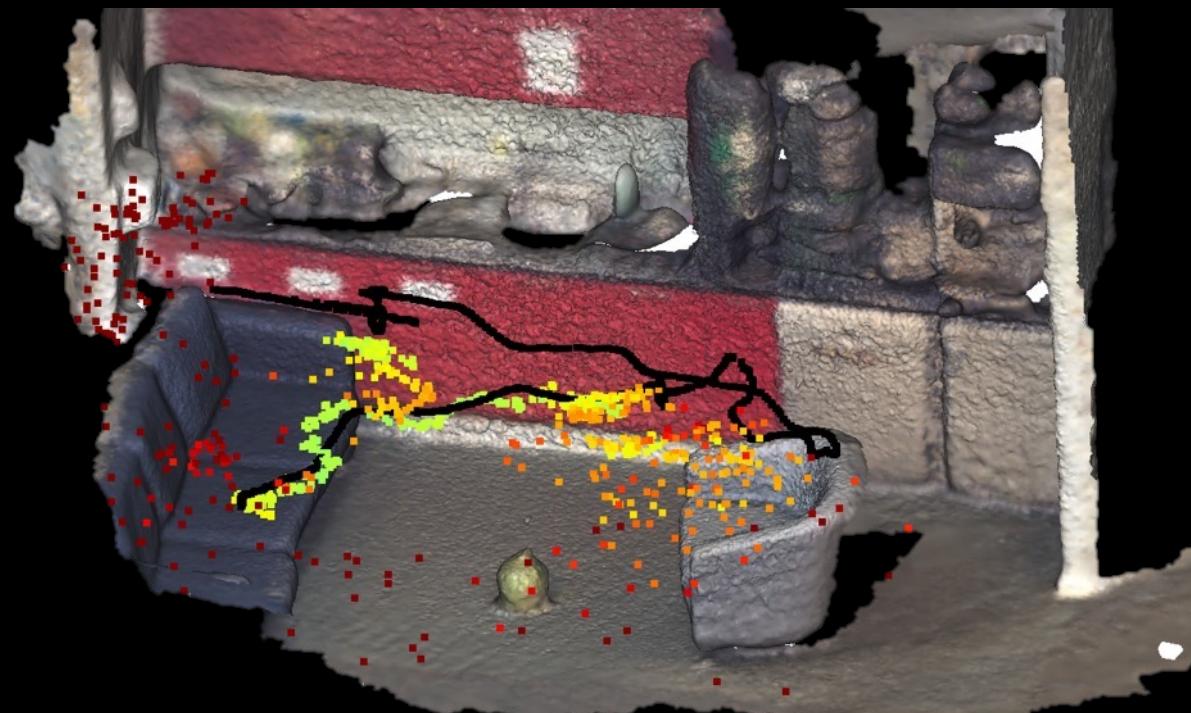
Reference Image



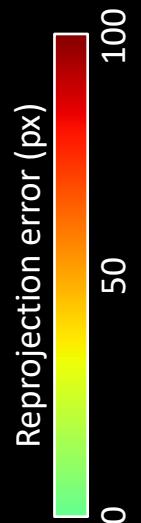
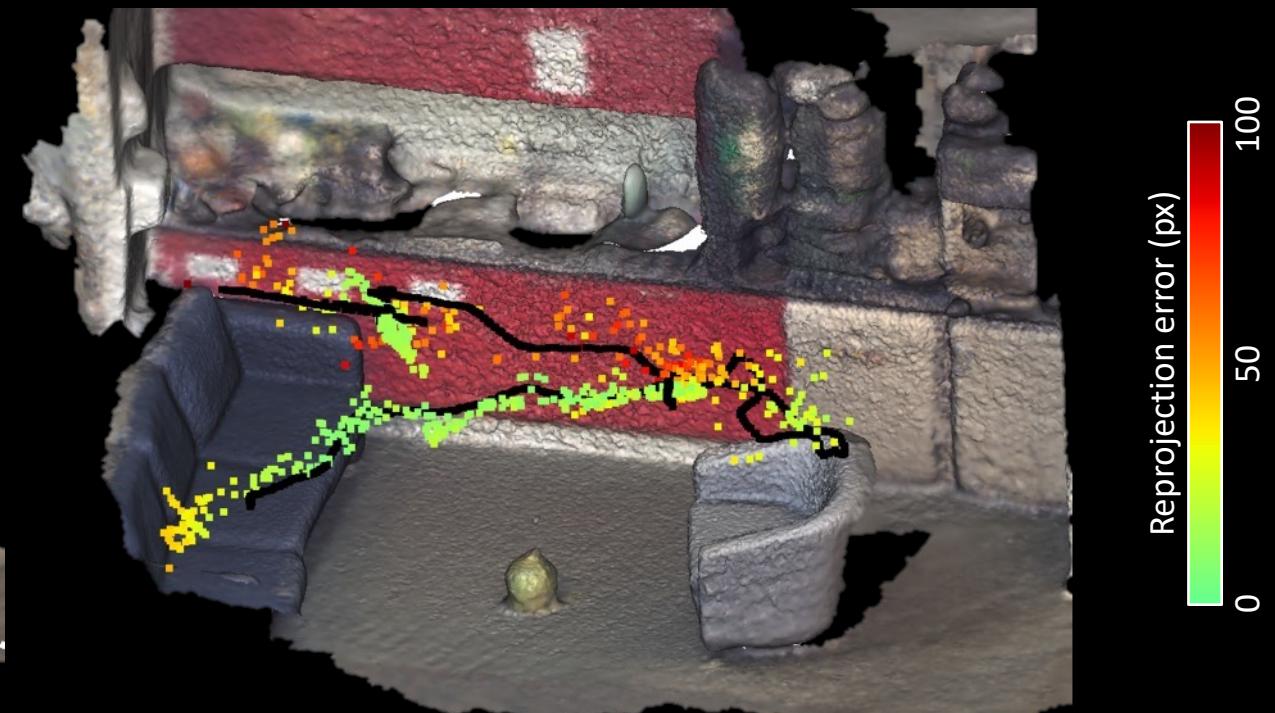
Query Images



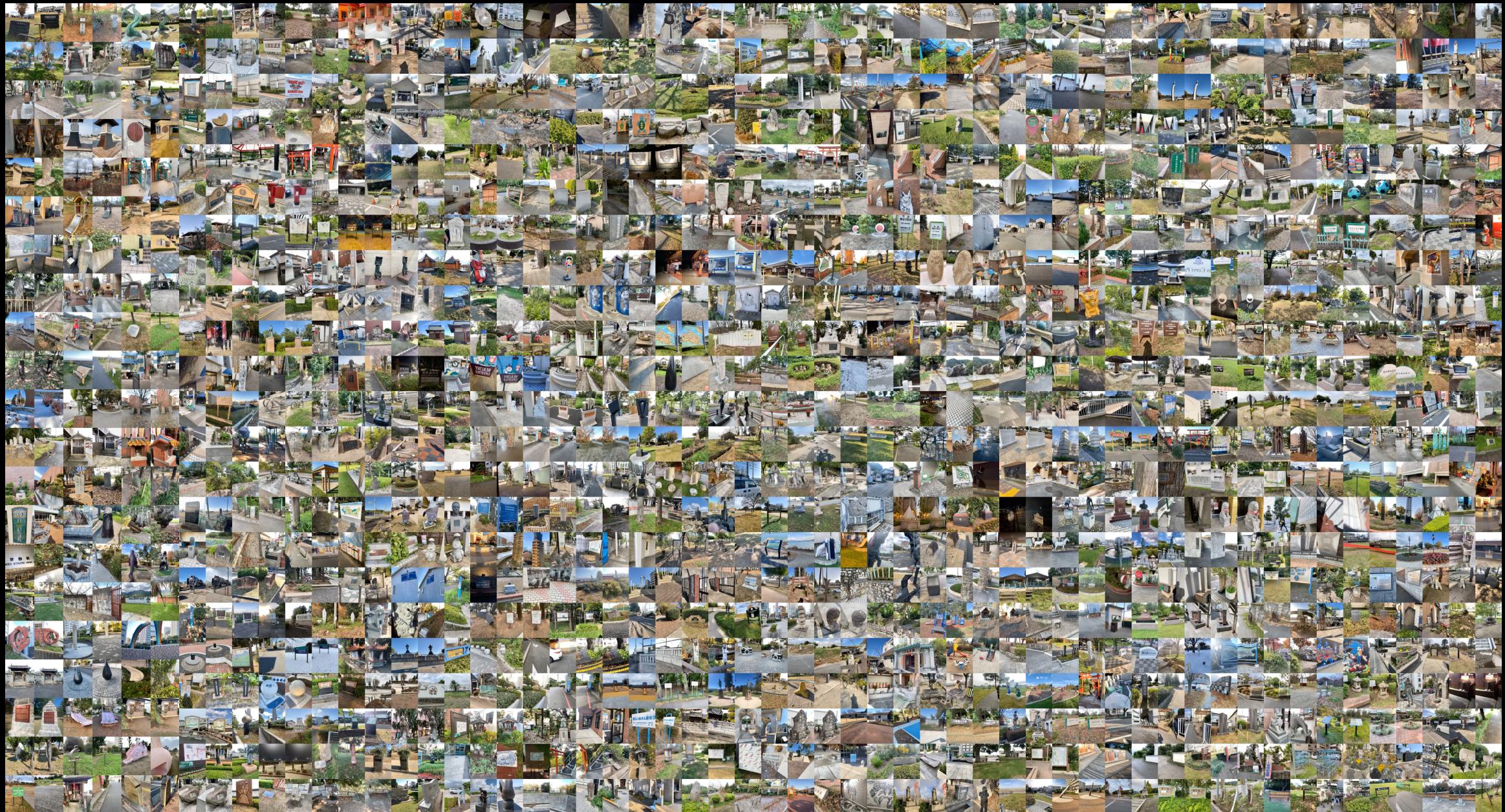
MapFree RPR, ECCV22

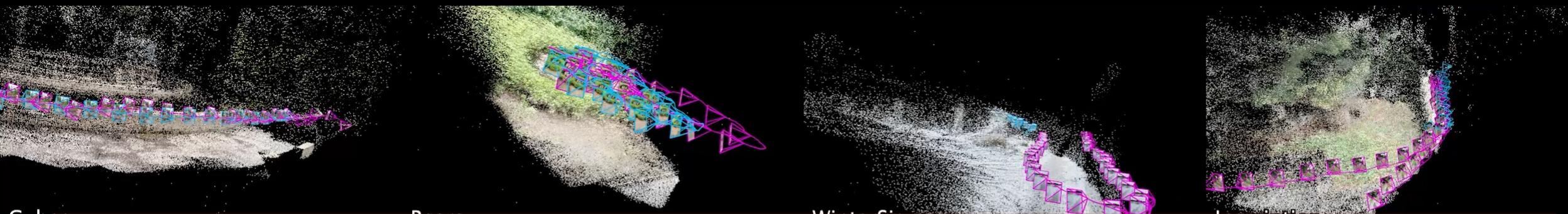


E-Mat from SuperGlue, Scale from Est. Depth





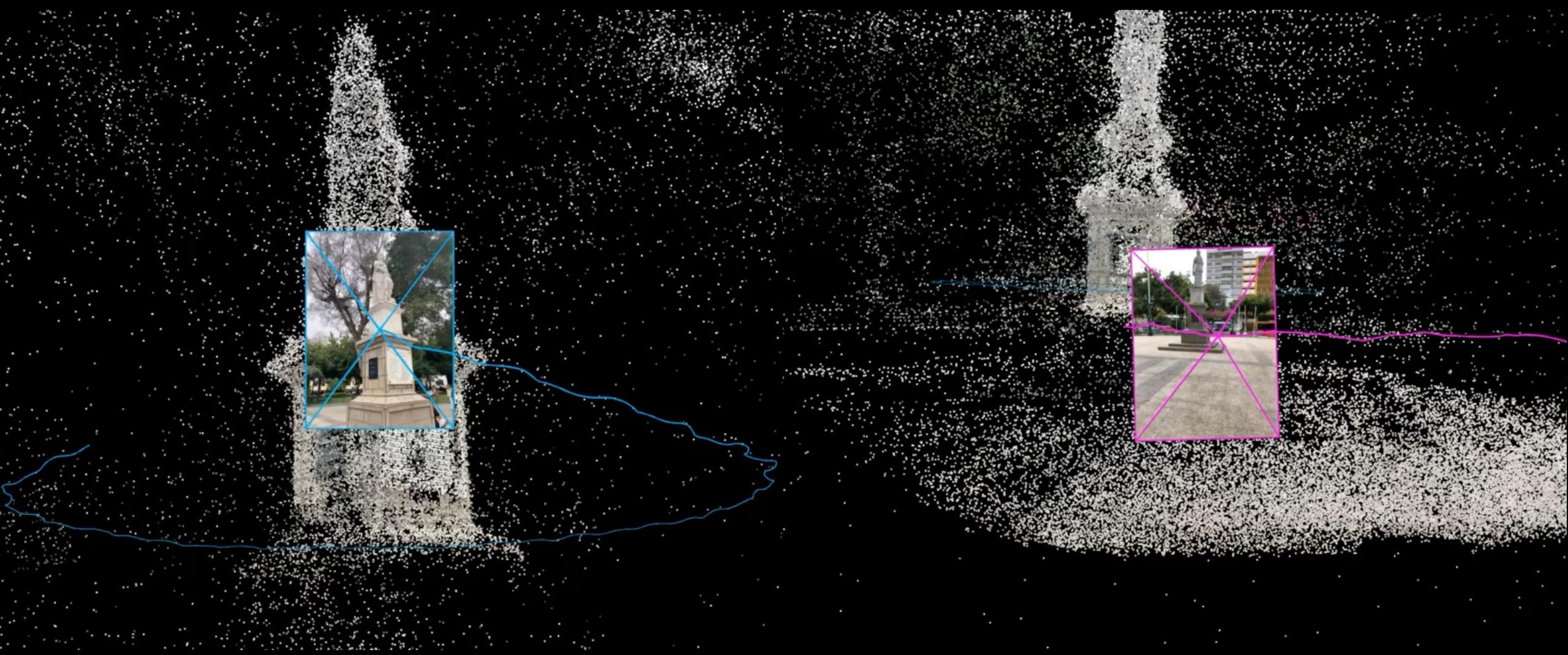


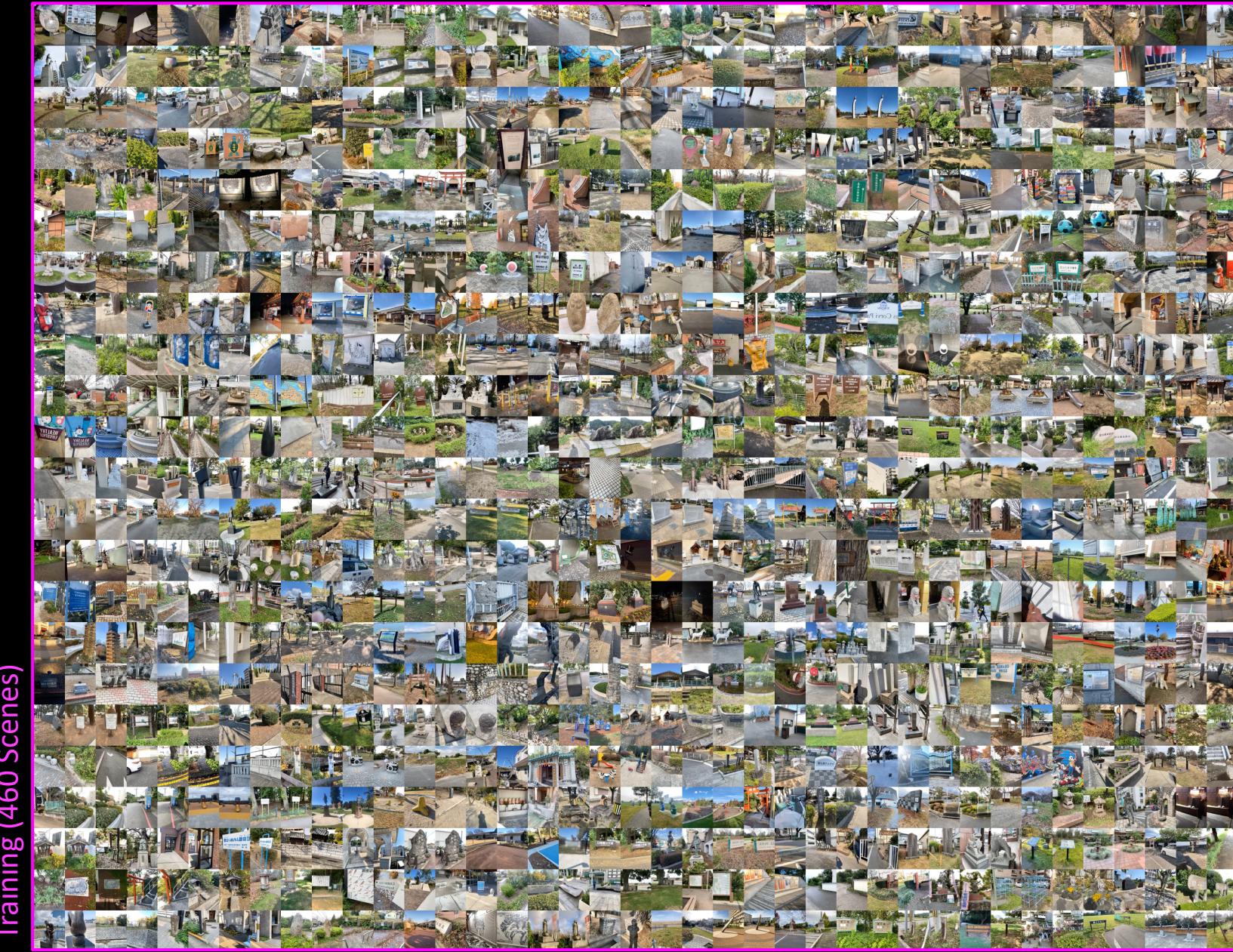
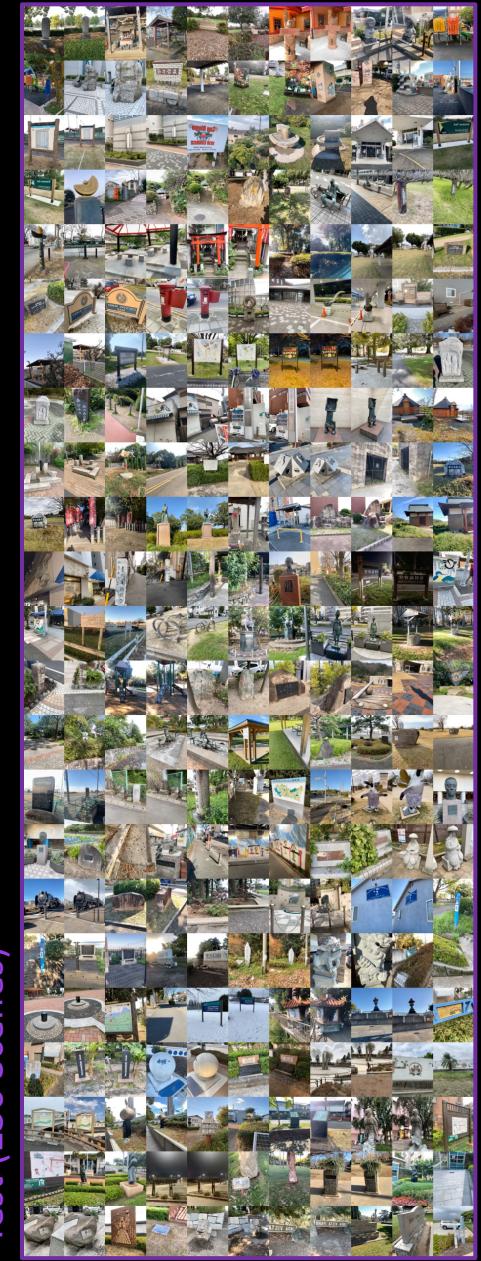


460 SfM models online for the training set

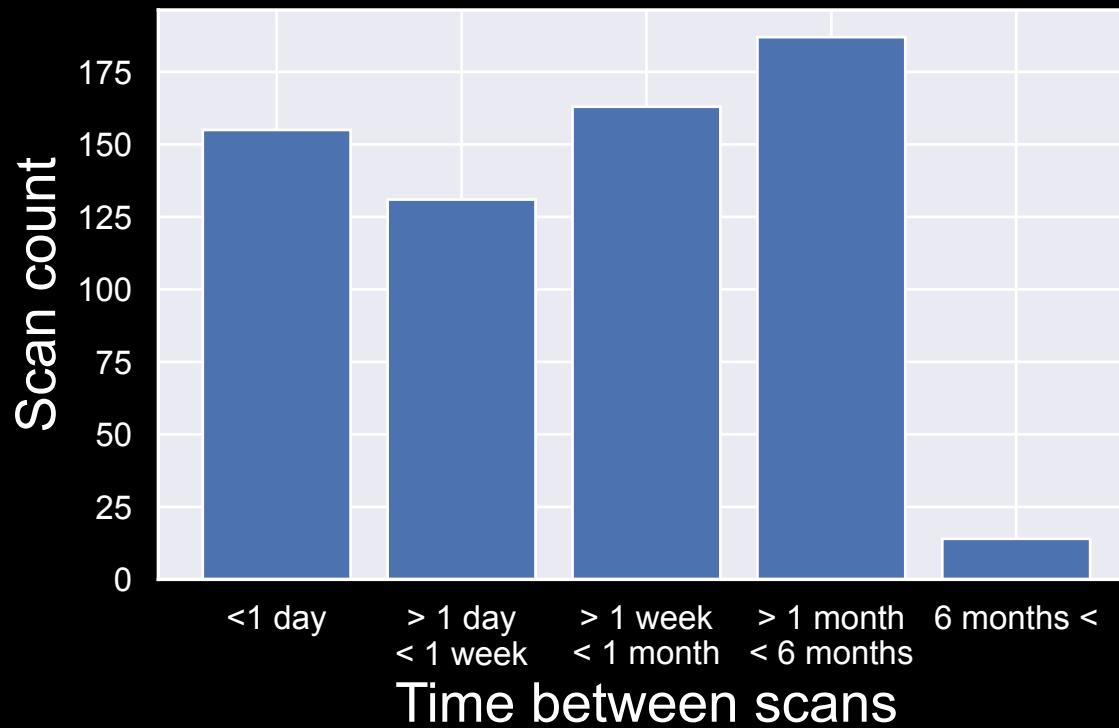
<https://research.nianticlabs.com/mapfree-reloc-benchmark>



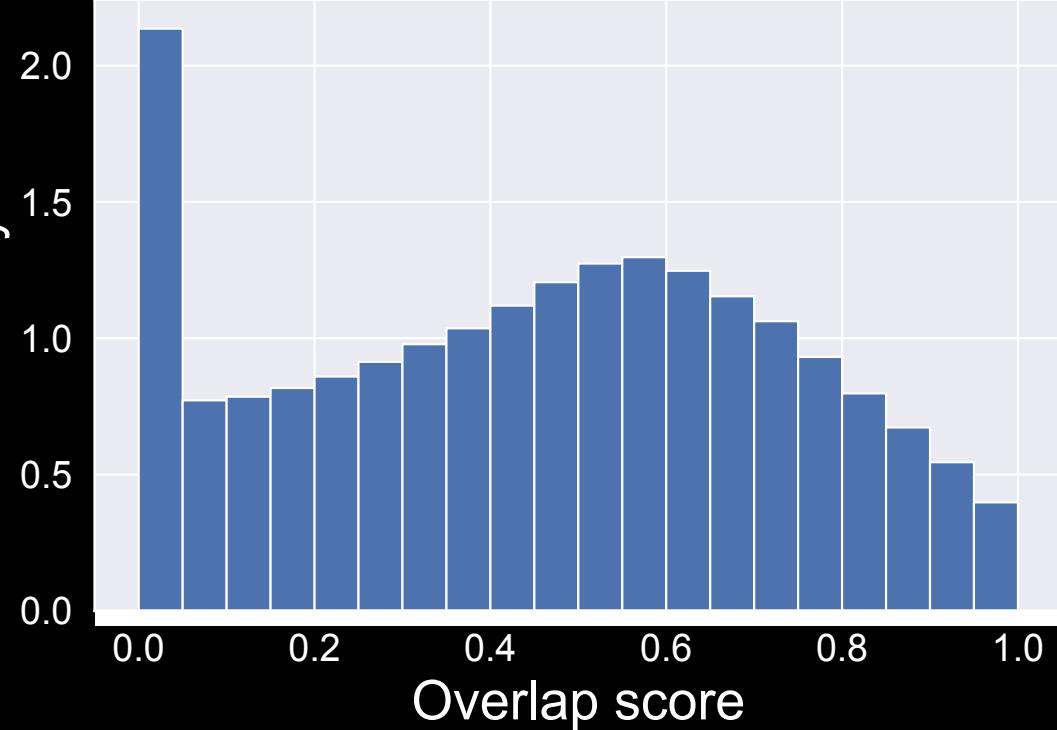


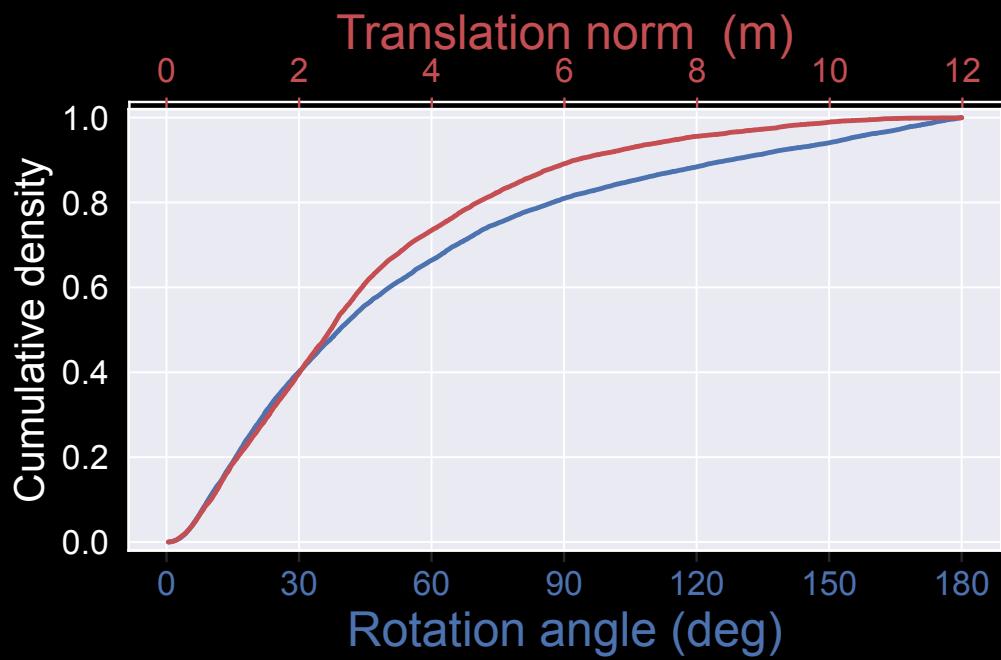


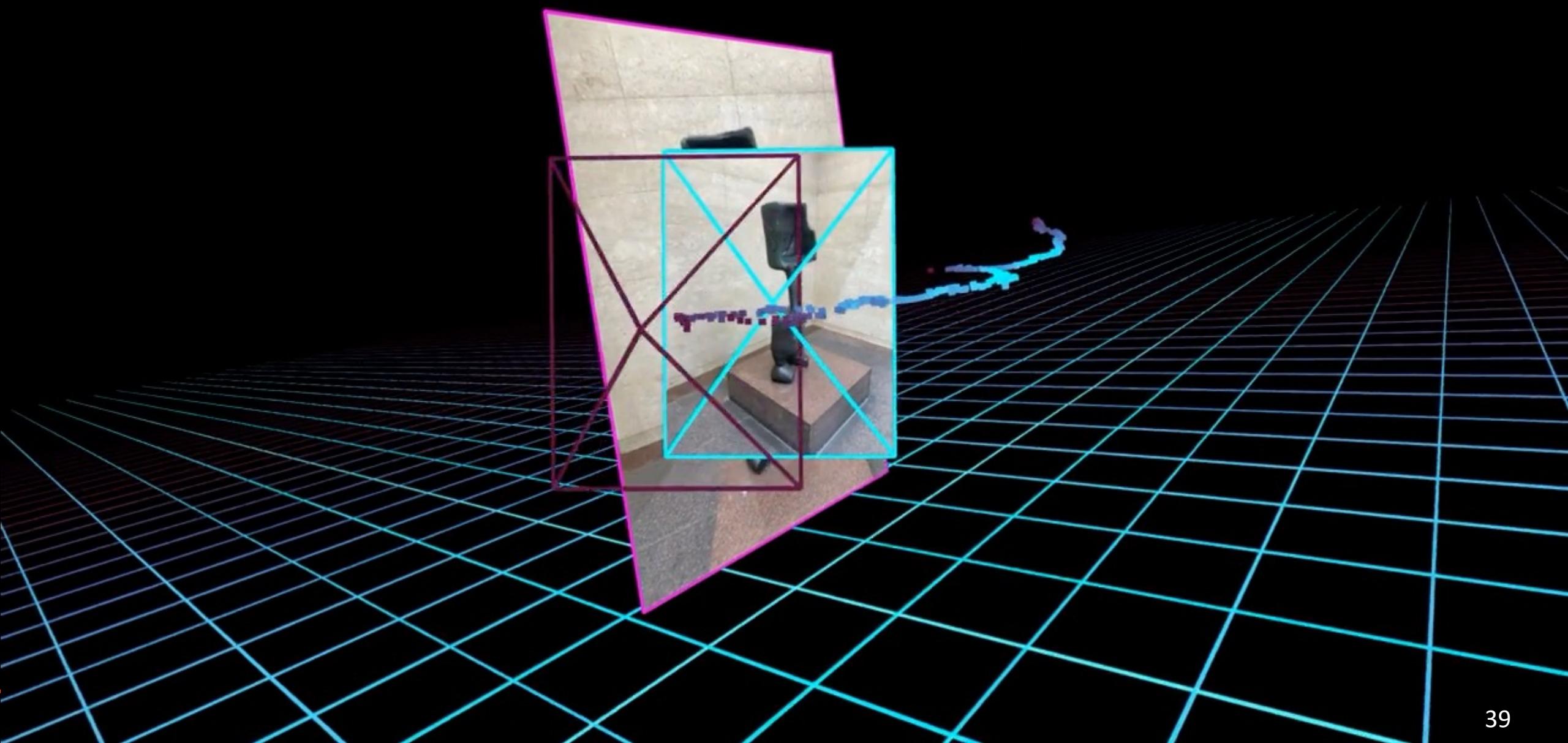




Density





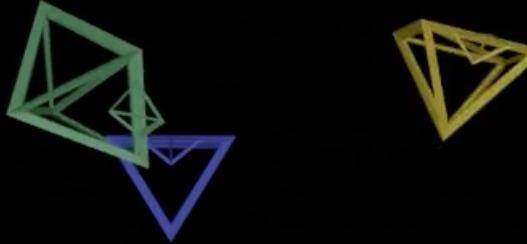


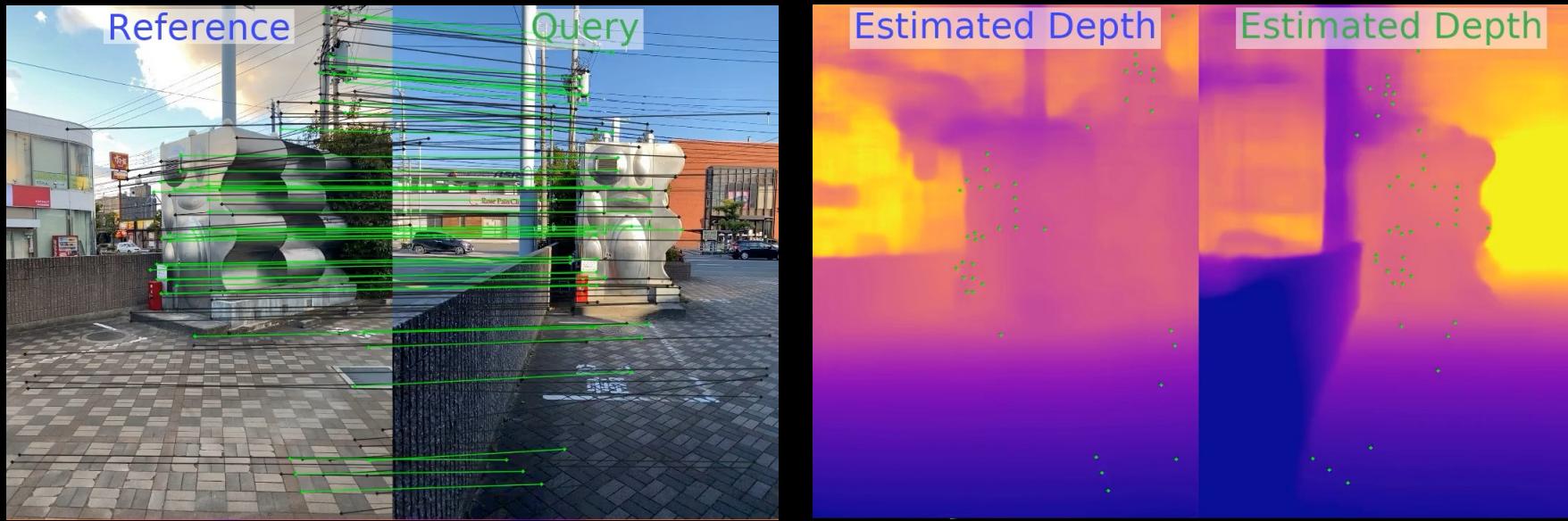


Reference Frame

Pseudo Ground Truth

DPT Depth + SuperGlue





Reference Frame

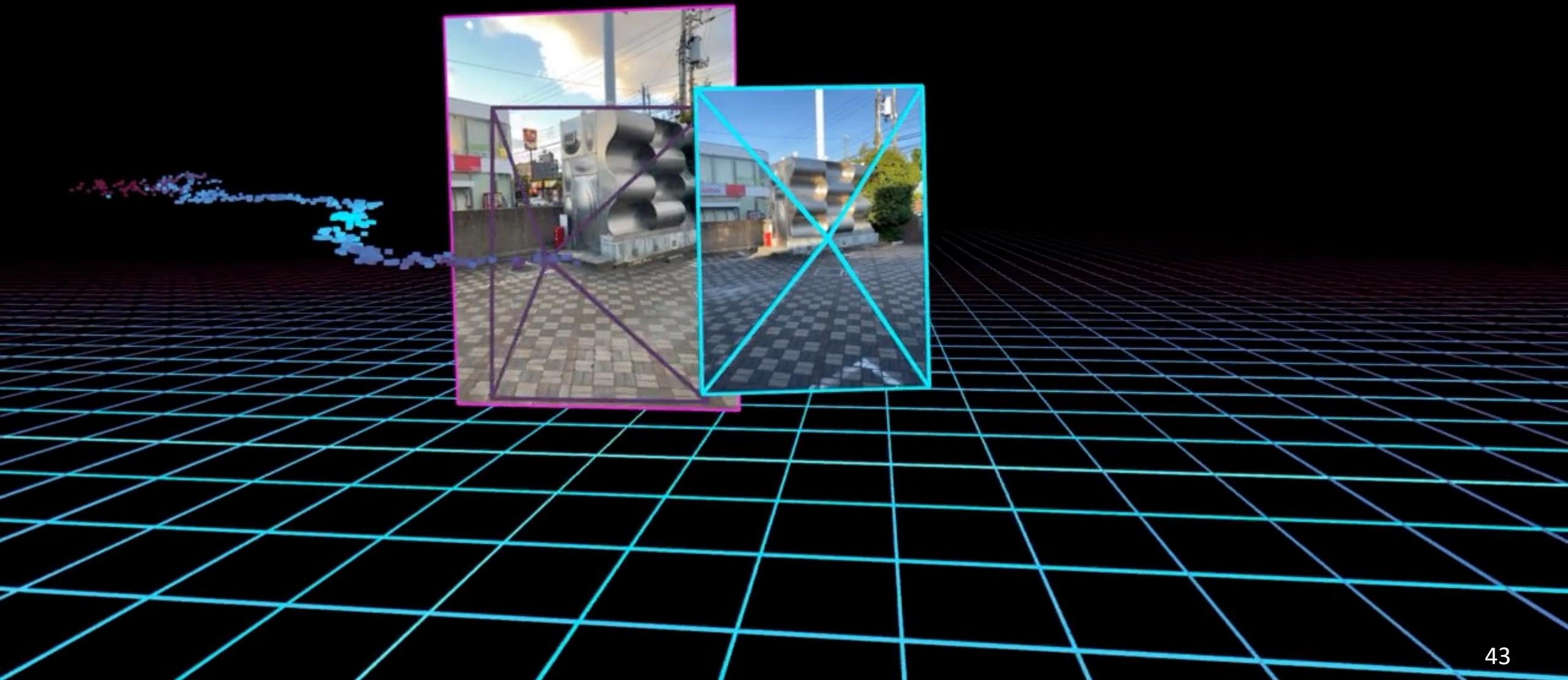


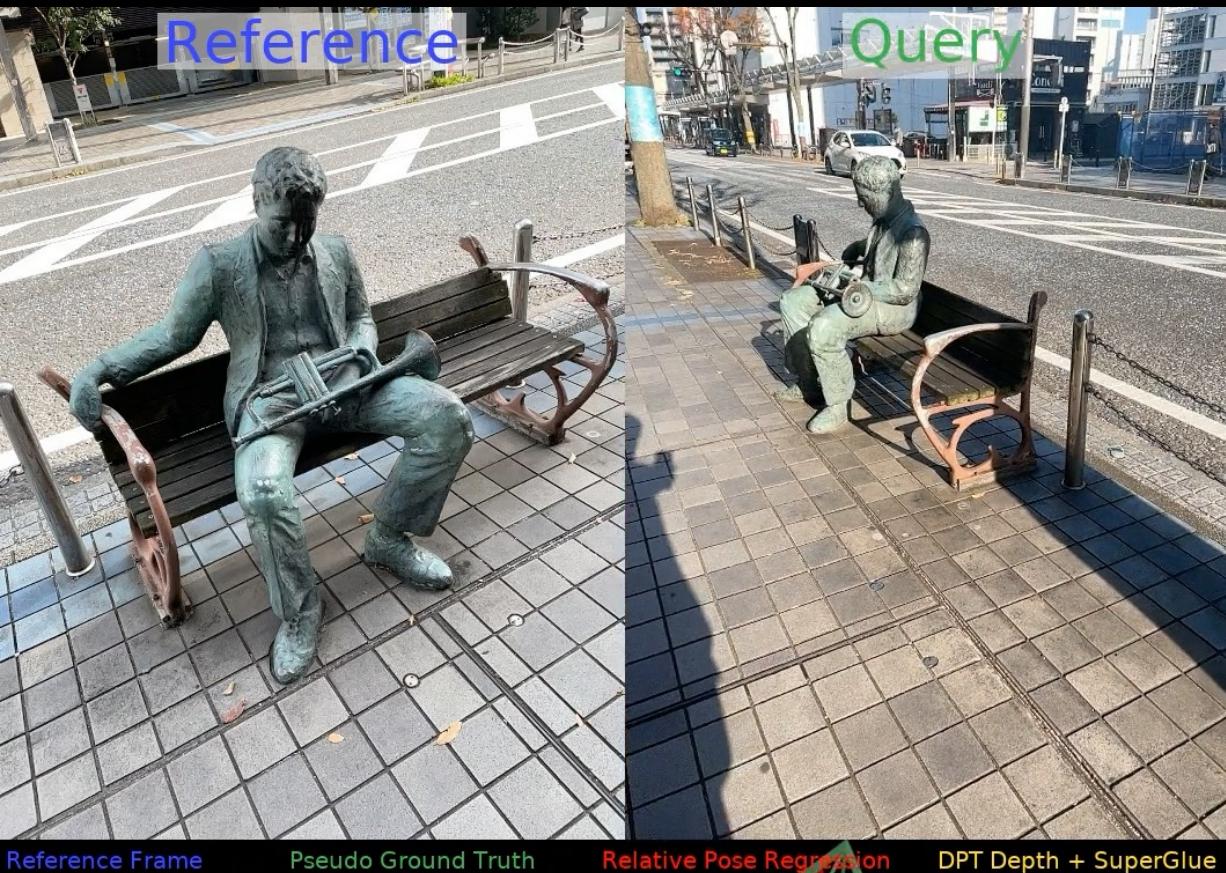
Pseudo Ground Truth



DPT Depth + SuperGlue







Reference Frame

Pseudo Ground Truth

Relative Pose Regression

DPT Depth + SuperGlue





Reference Frame

Pseudo Ground Truth



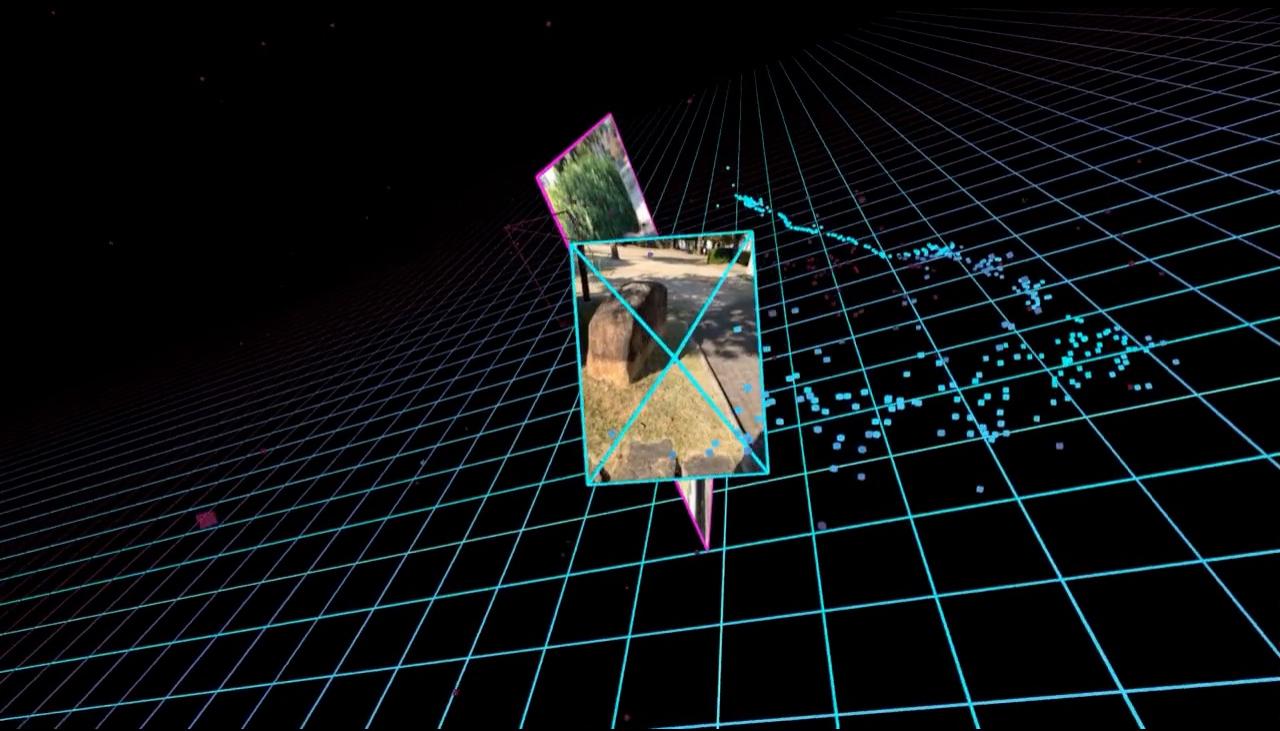
Relative Pose Regression



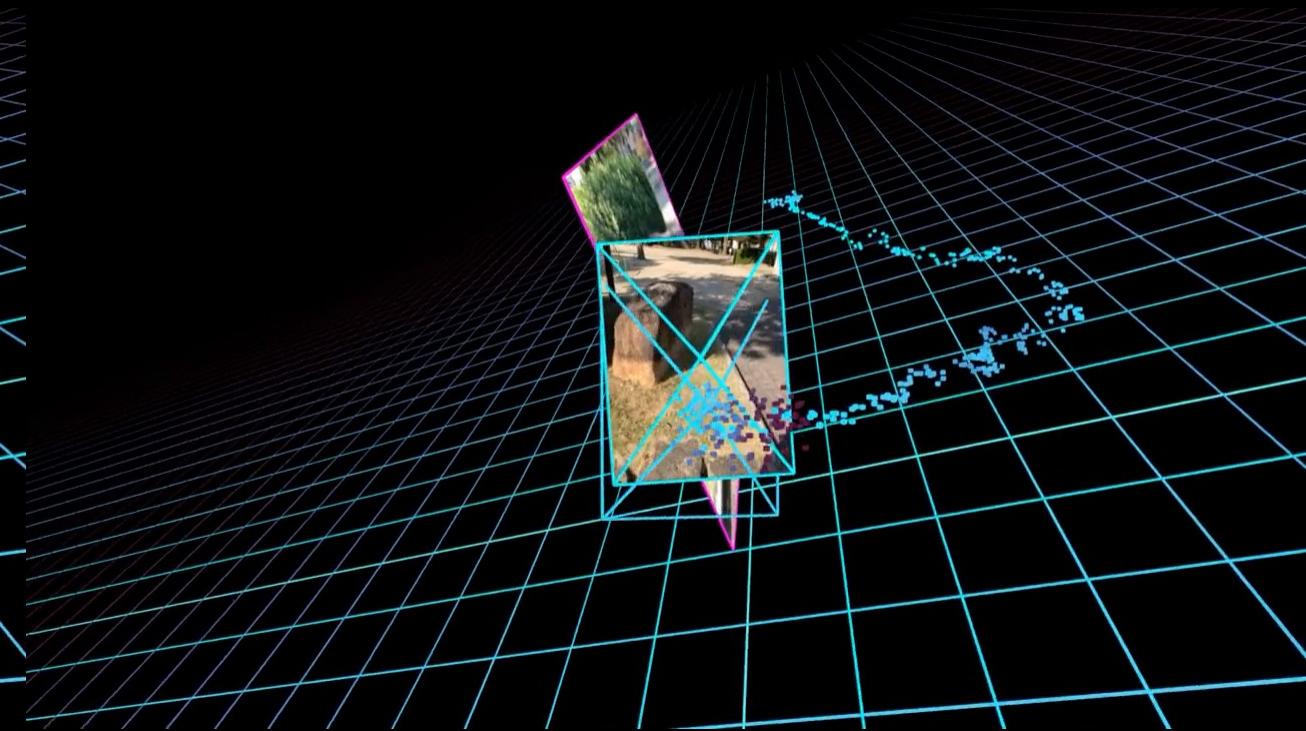
DPT Depth + SuperGlue



SuperGlue + Depth



Map-Free RPR





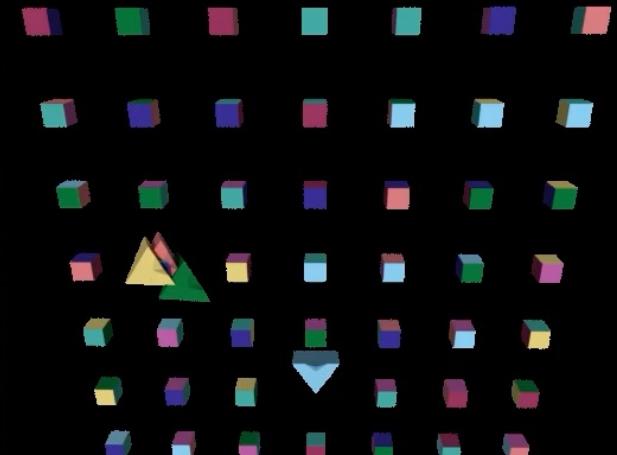
Reference Frame

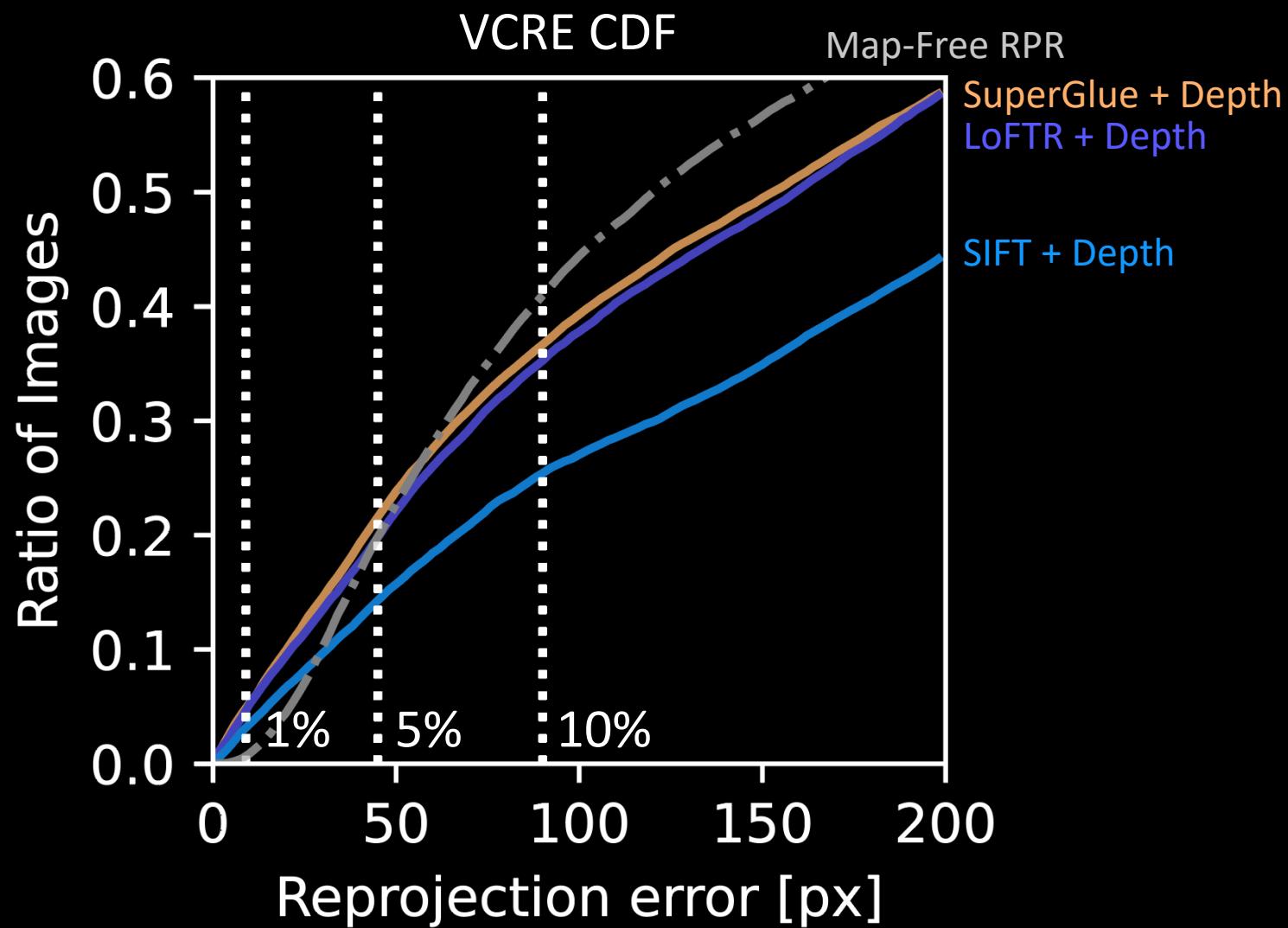
Pseudo Ground Truth

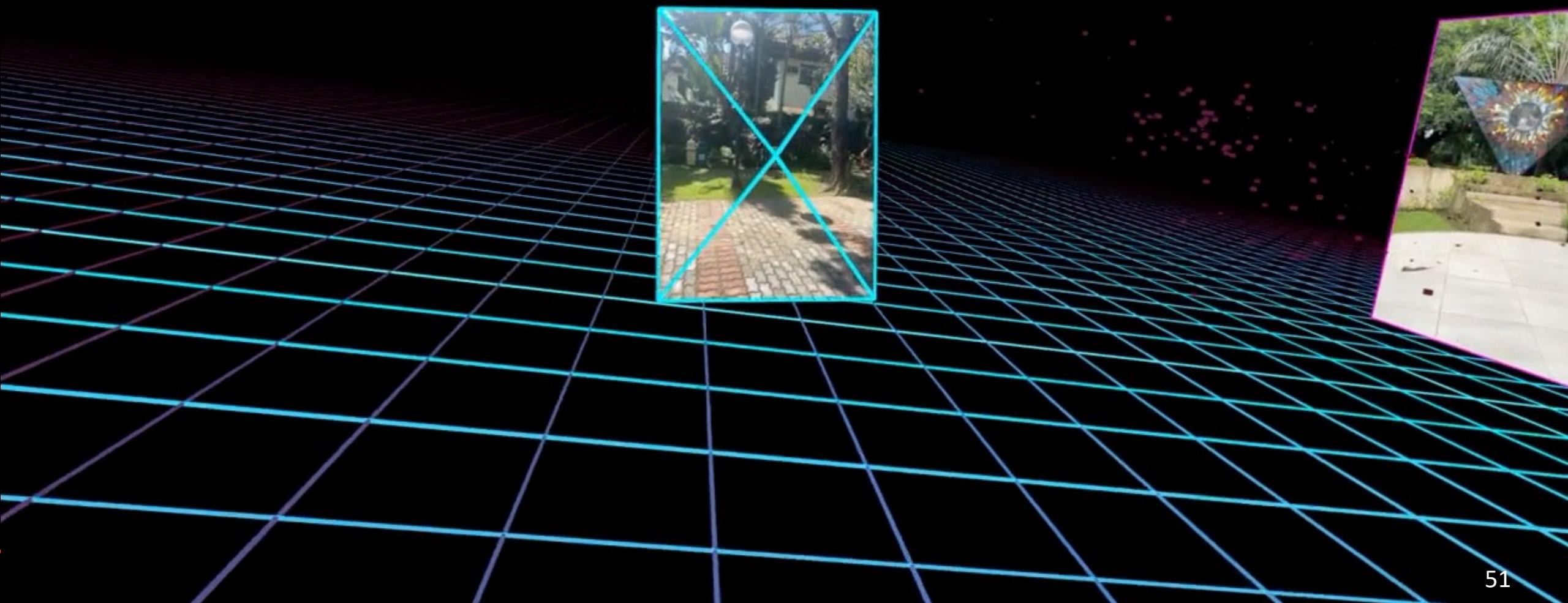
Relative Pose Regression

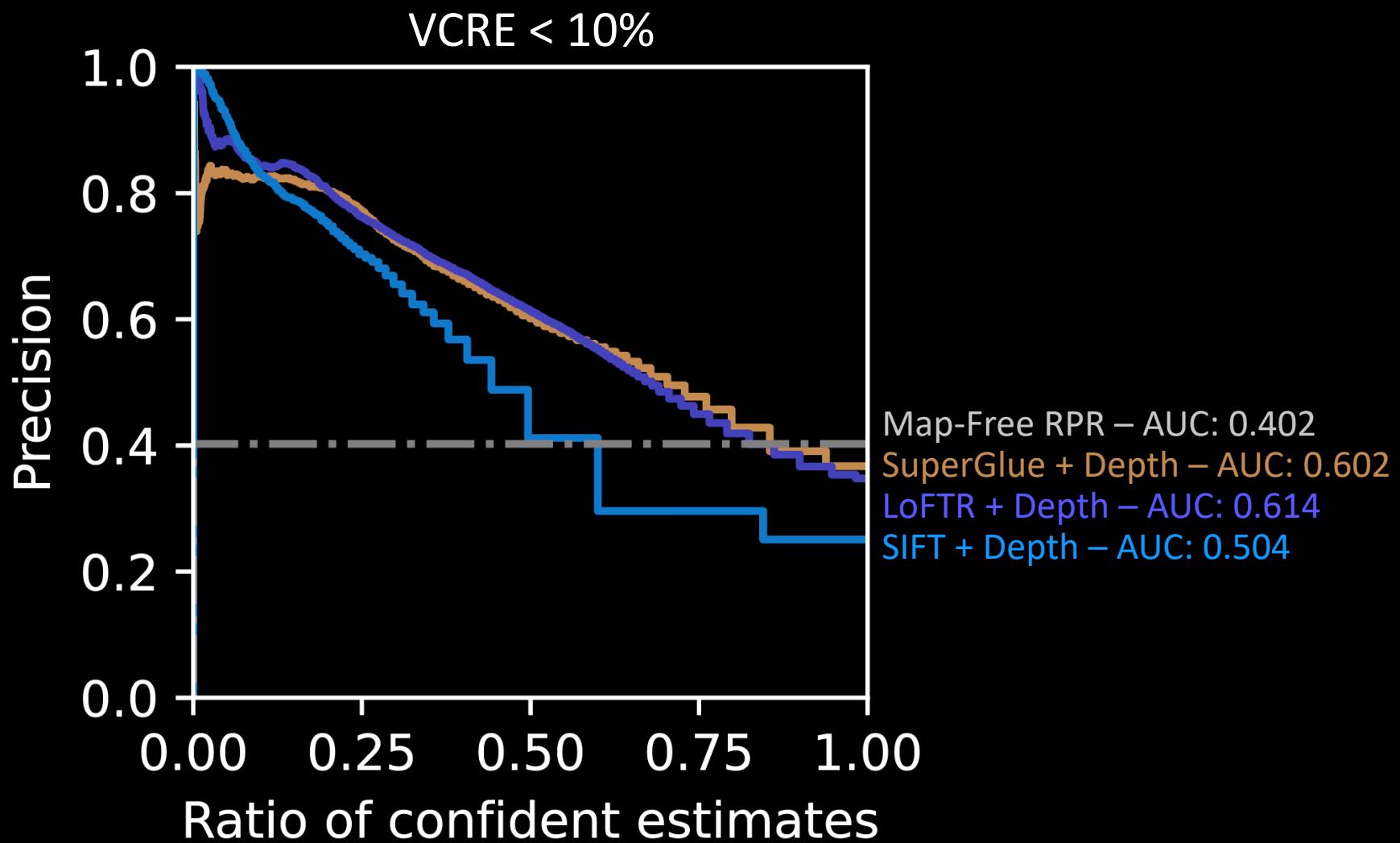
DPT Depth + SuperGlue



**Reference view****Ground Truth****Relocalized with  
Feature Matching + D.Scale****All Top**







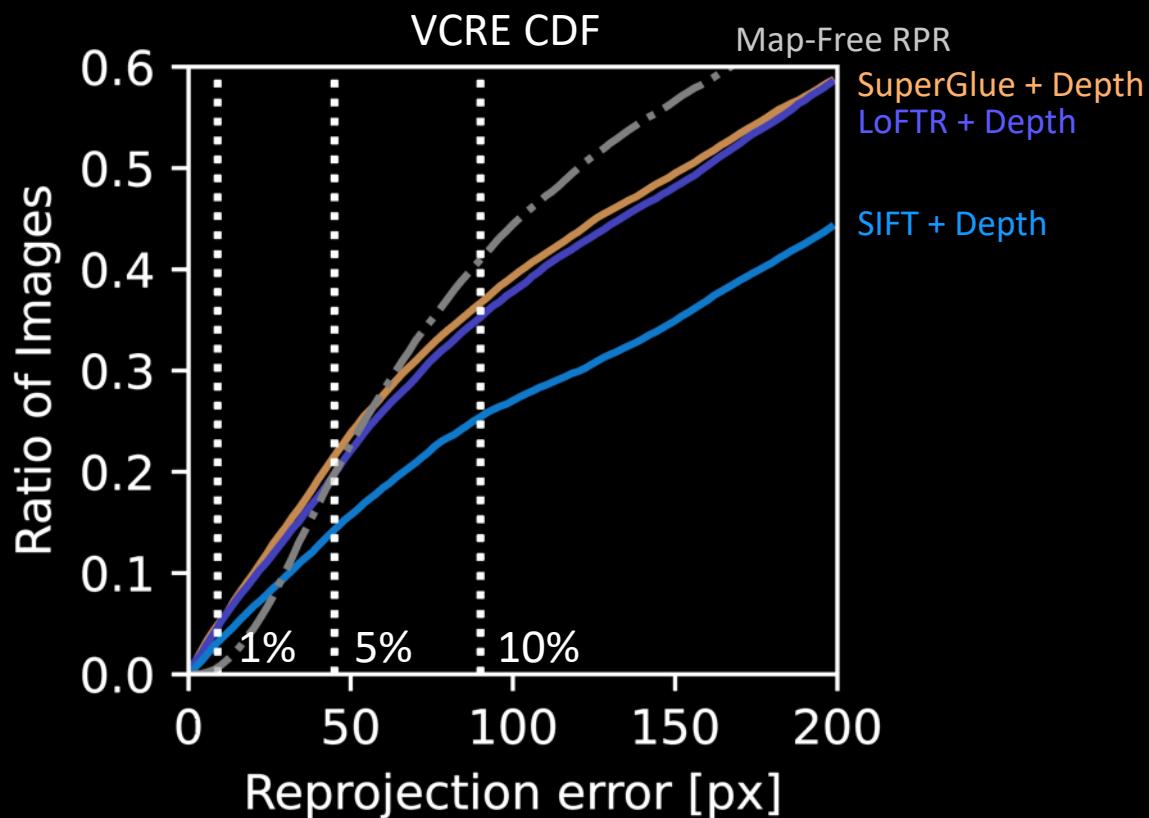
## Evaluation Leaderboard

Method	AUC (VCRE < 90px) ▾	Precision (VCRE < 90px)	Median Reproj. Error (px)	AUC (Err < 25cm, 5°)	Precision (Err < 25cm, 5°)	Median Trans. Error (m)	Median Rot. Error (°)	Estimates for % of frames
ⓘ DPT-KITTI & LoFTR (PnP)	0.618	33.4%	166.7	0.269	9.8%	2.31	39.4	100.0
ⓘ DPT-KITTI & LoFTR (Ess.Mat. + D.Scale)	0.614	34.7%	167.6	0.346	15.4%	1.98	30.5	100.0
ⓘ DPT-KITTI & SuperGlue (Ess.Mat. + D.Scale)	0.602	36.1%	160.3	0.346	16.8%	1.88	25.4	100.0
ⓘ DPT-KITTI & SuperGlue (PnP)	0.598	36.0%	156.9	0.252	10.7%	2.10	32.5	100.0
ⓘ DPT-KITTI & SIFT (Ess.Mat. + D.Scale)	0.504	25.0%	222.8	0.252	10.3%	2.93	61.4	100.0
ⓘ DPT-KITTI & SIFT (PnP)	0.468	25.1%	192.0	0.190	7.8%	3.35	63.7	100.0
ⓘ RPR [R(6D) +t]	0.402	40.2%	147.1	0.060	6.0%	1.68	22.5	100.0
ⓘ RPR [3D-3D]	0.387	38.7%	148.7	0.060	6.0%	1.69	22.9	100.0
ⓘ RPR [R( $\alpha$ , $\beta$ , $\gamma$ ) + s · t( $\theta$ , $\phi$ )]	0.354	35.4%	166.3	0.105	10.5%	1.83	23.2	100.0

Dataset, Baselines, Evaluation Code, Leaderboard  
<https://research.nianticlabs.com/mapfree-reloc-benchmark>



# Pose Estimation Beyond Feature Matching



<https://research.nianticlabs.com/mapfree-reloc-benchmark>