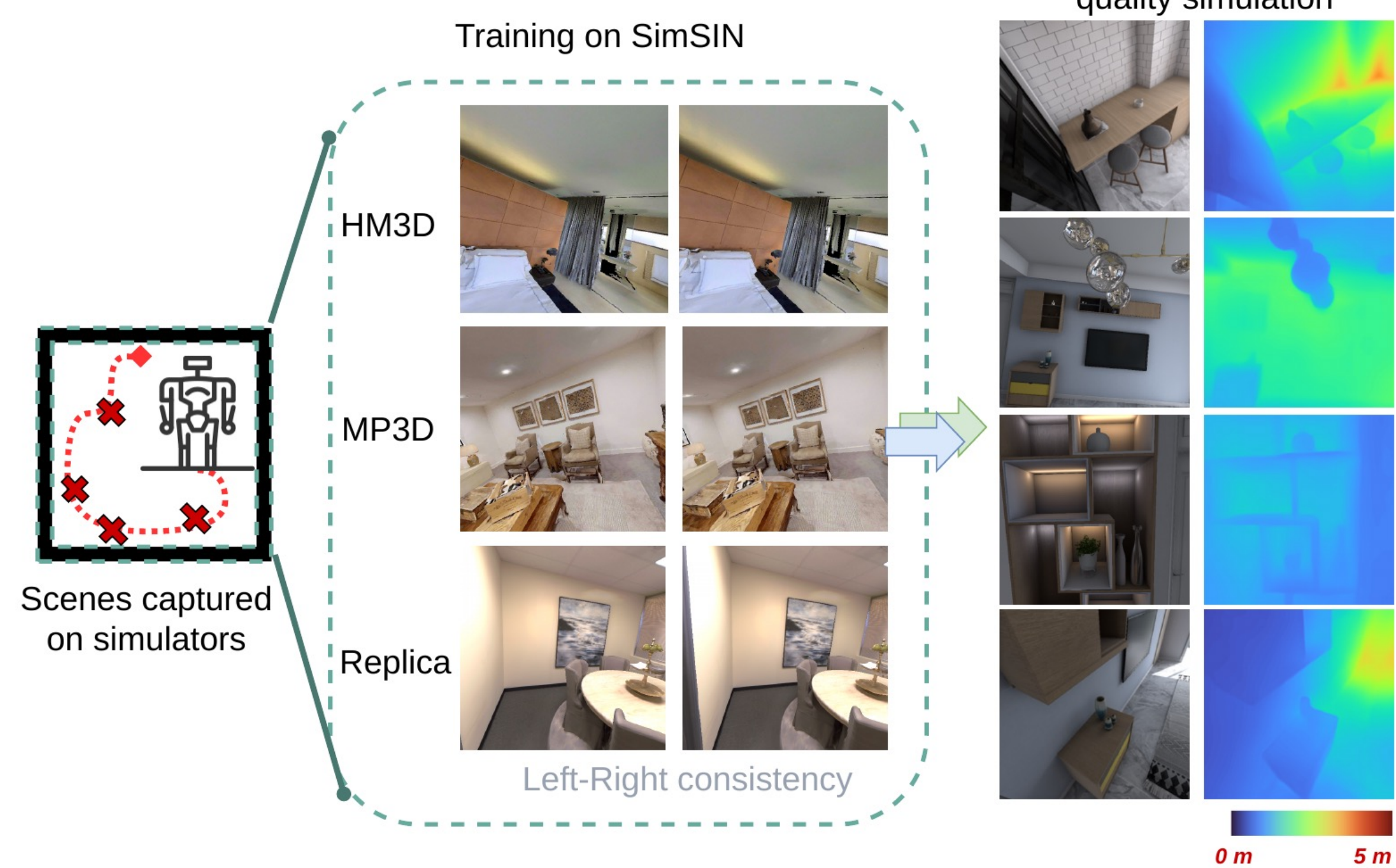
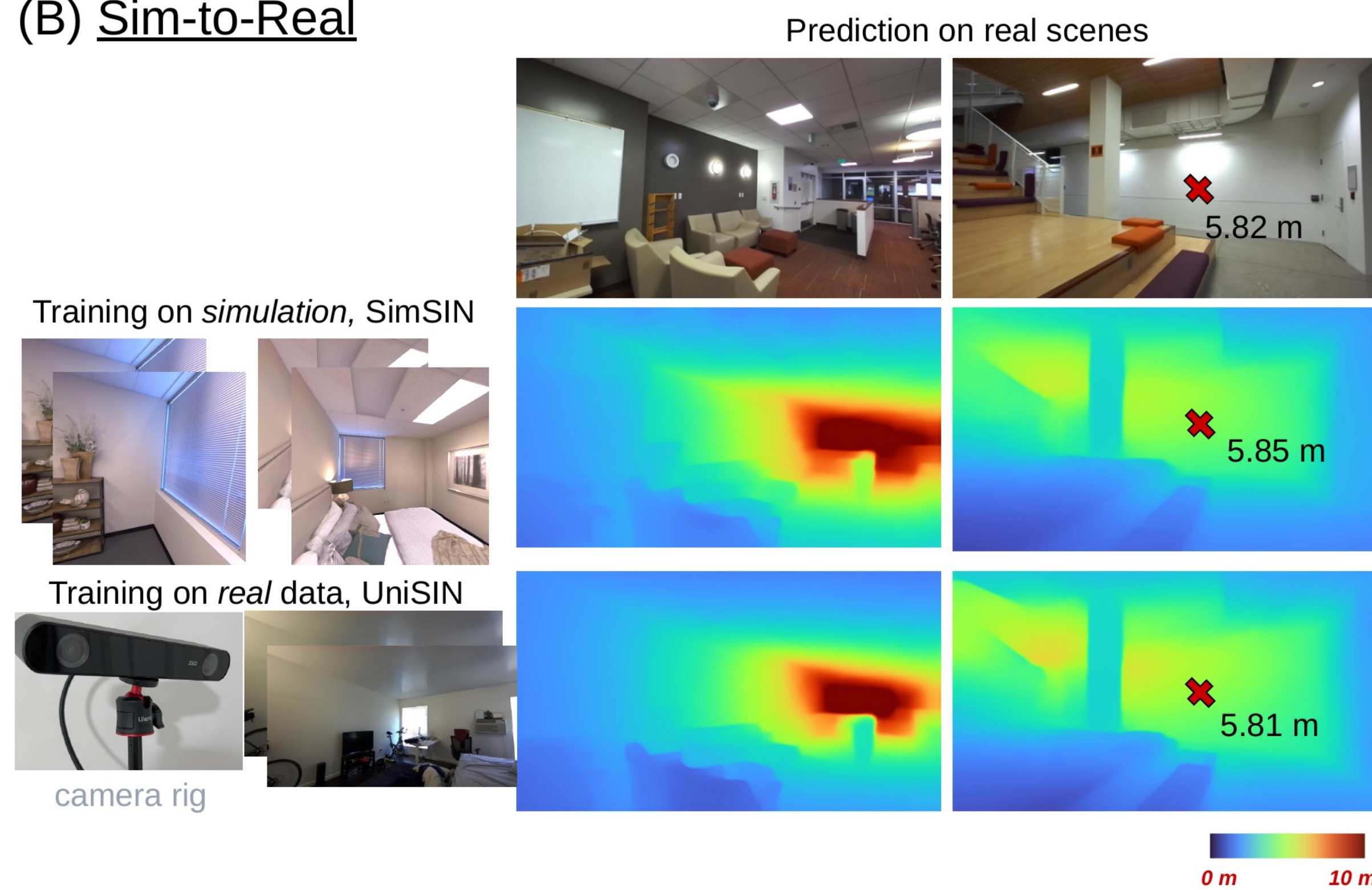


## Advantages

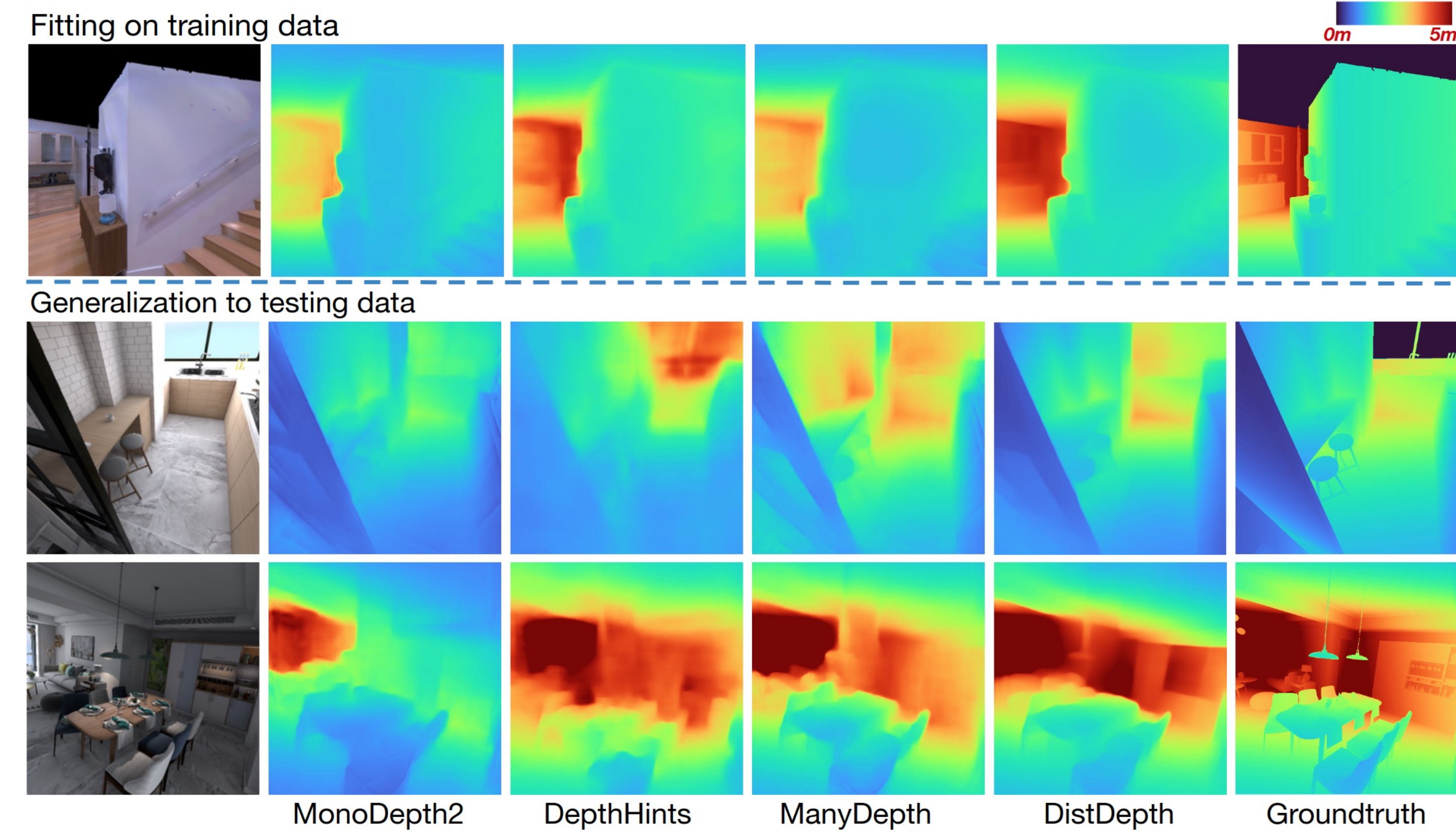
### (A) Zero-Shot Cross-Dataset Inference



### (B) Sim-to-Real

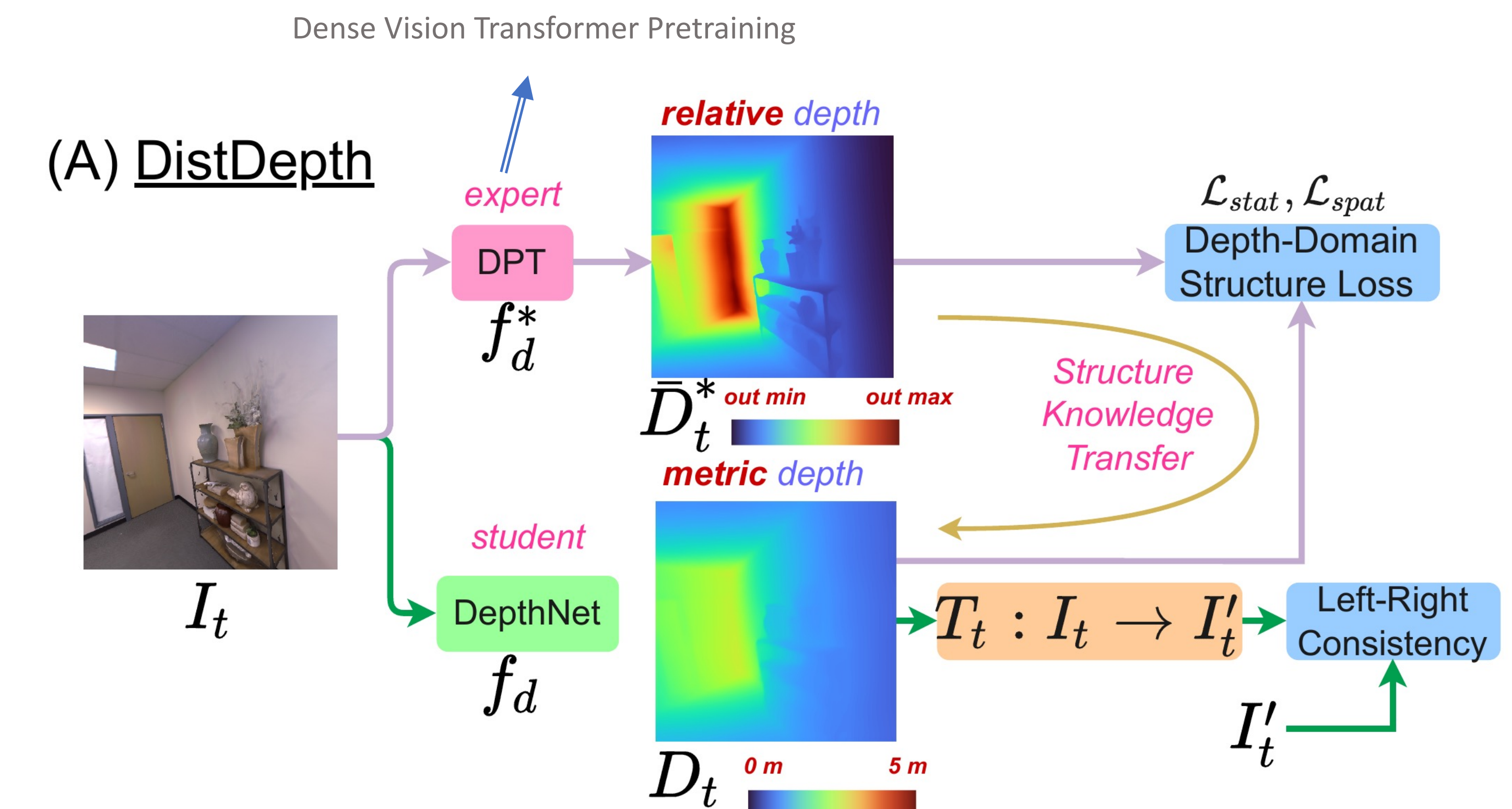


## Fitting on Training Scenes

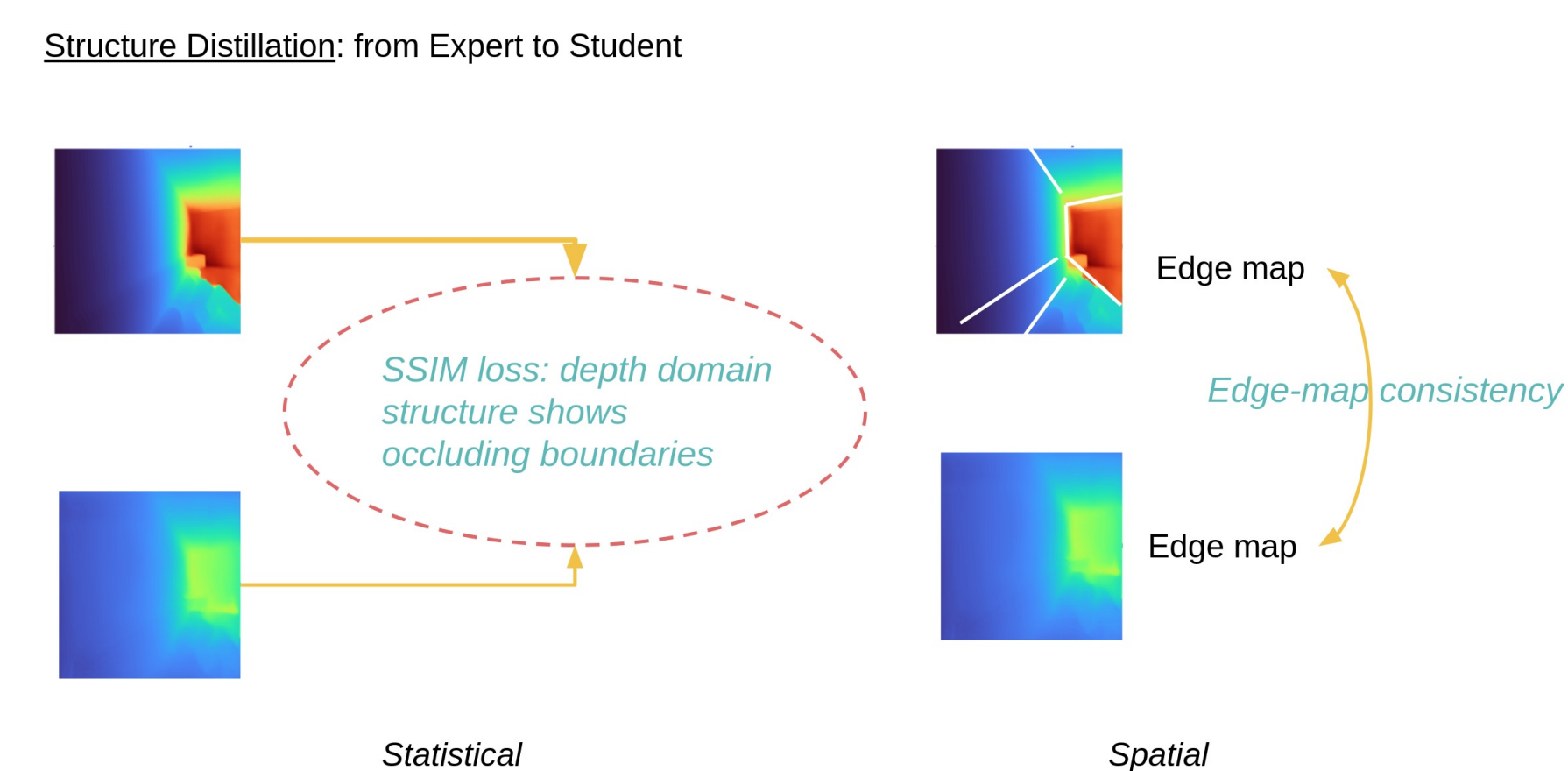


Prior left-right consistency methods can fit well on training scenes, but they do not generalize to testing scenes with complex structures.

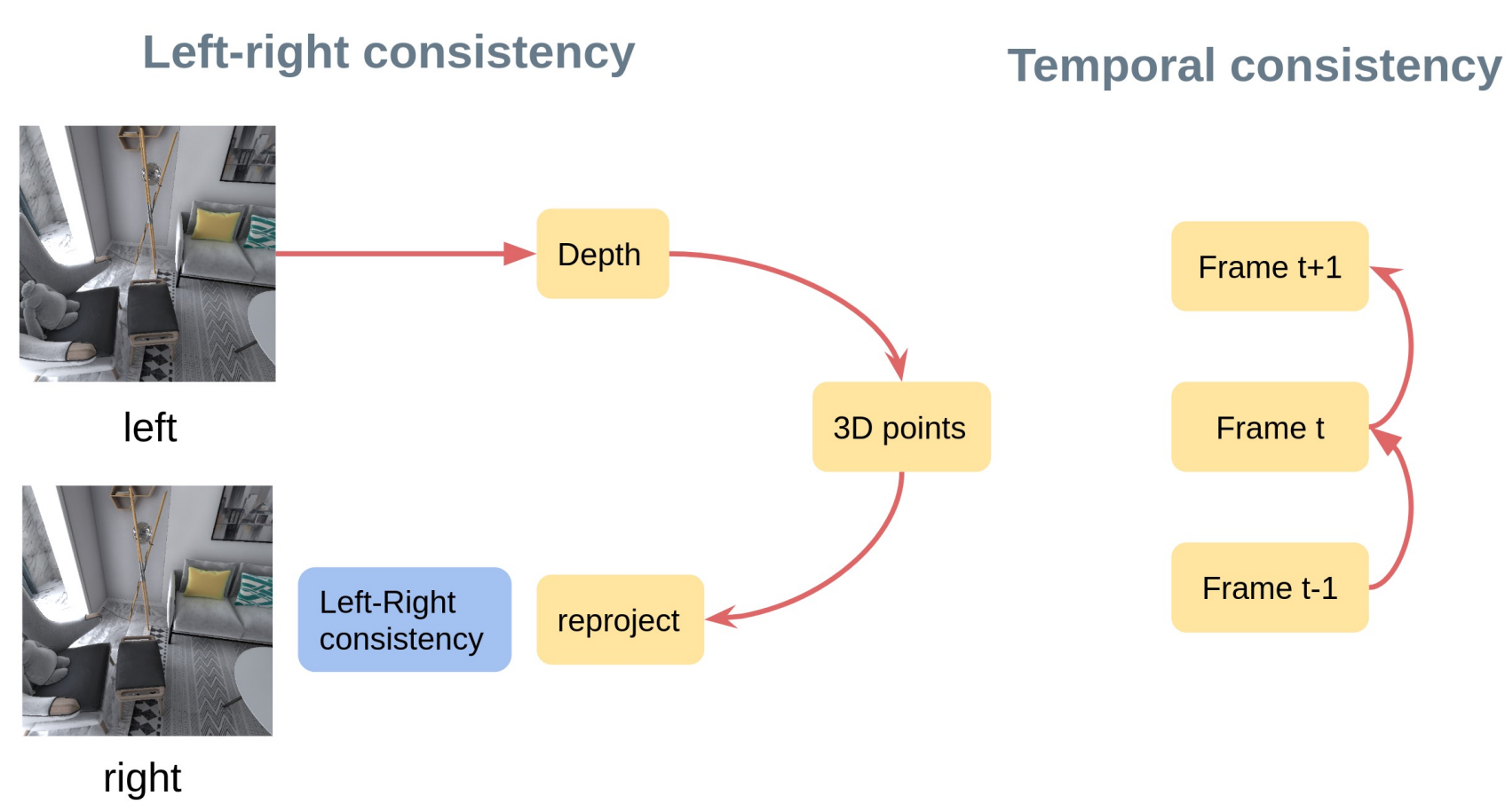
## Method



### (1) Depth-Domain Structure Loss:

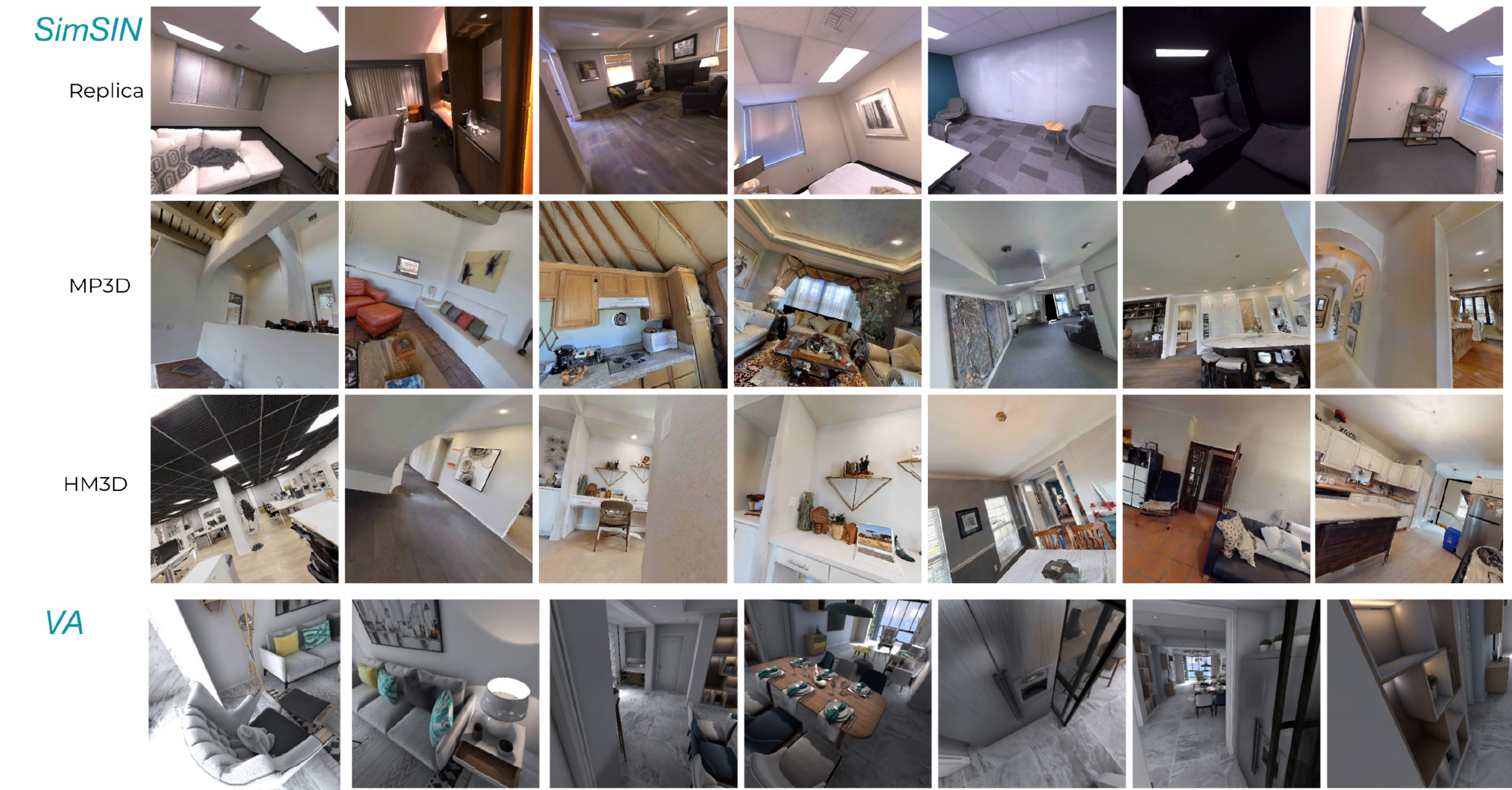


### (2) Left-Right Consistency:

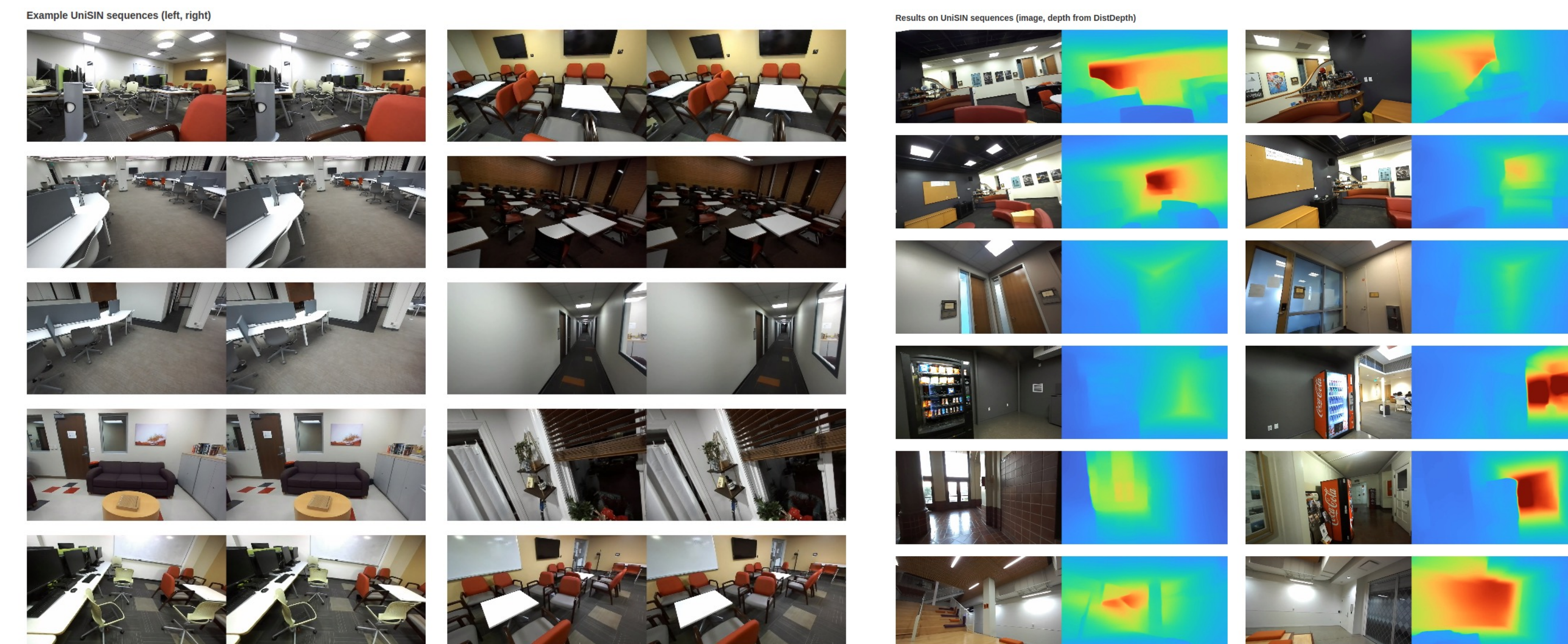


Combining metrics from left-right pairs and structures from relative depth pretraining

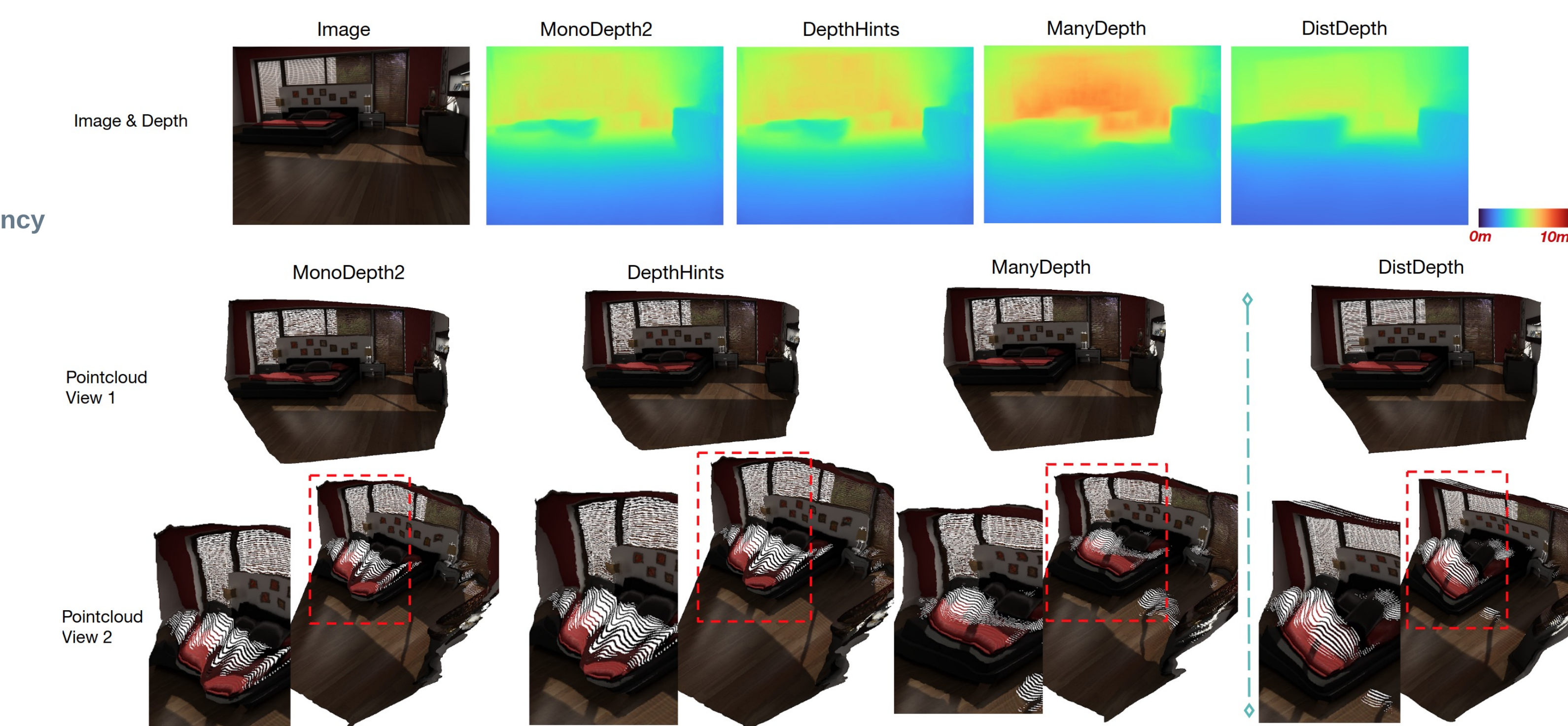
## Collected Datasets and Results



## UniSIN



## Results



## Practical Depth Estimation

- Learning without curated depth groundtruth
- Efficient and effective data collection by simulator
- High generalizability and accurate and real-time performance

## Motivation

- Most prior works for self-supervised depth focus on outdoor driving scenes. However, the applicability is yet to be investigated.
- Either cross-data evaluation and sim-to-real evaluation for indoor scenes has not explored much.

