



for Robotics



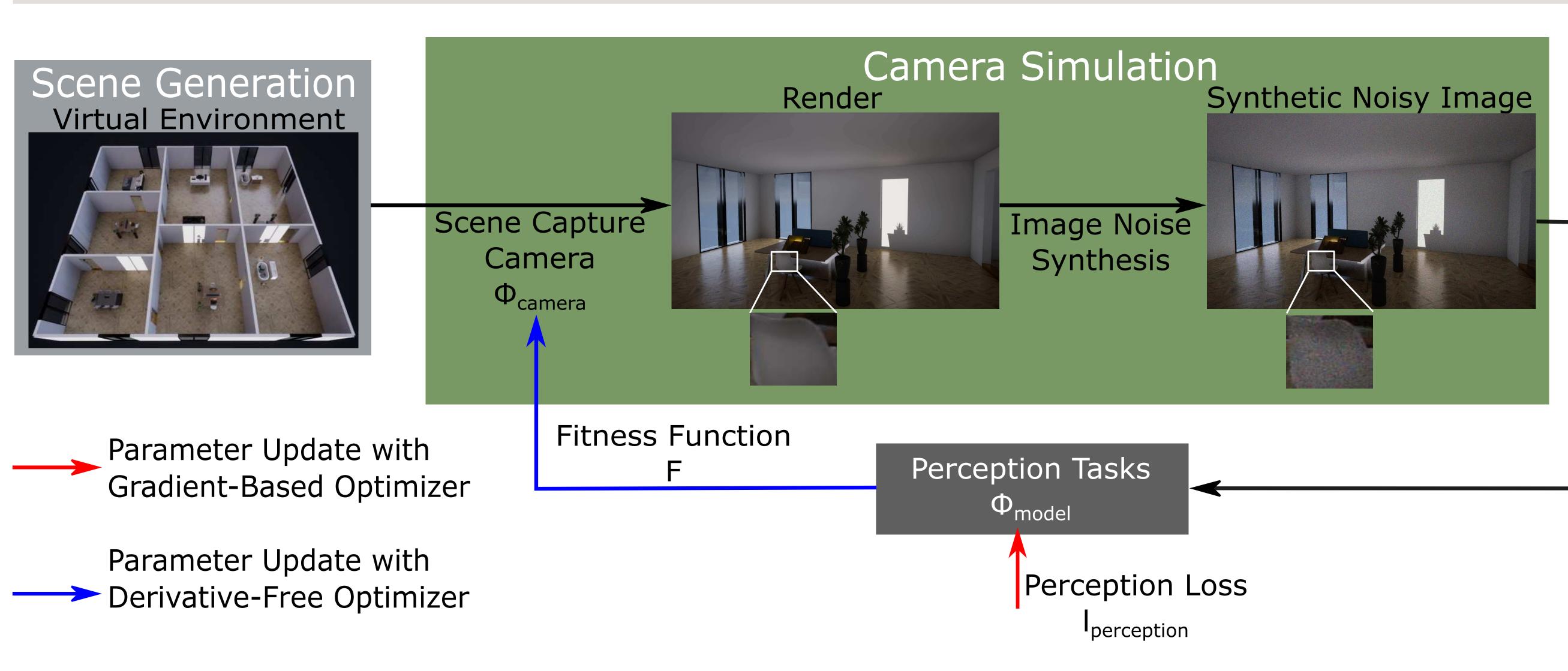
Camera Design For Perception Tasks

- The performance of perception tasks is influenced by cameras.
- Designing cameras with high performance is costly, requires human labour and hardware experiments.
- Most cameras and perception tasks are designed in isolation.

Previous Works

- Fully differentiable task-specific optics optimization. [1] Cannot optimize camera's resolution, placement, and unconventional cameras such as light field and stereo cameras.
- Task-specific camera optimization with Reinforcement Learning. [2] Long optimization time.
- In this work, we combine derivative-free and gradient-based optimizers to design cameras efficiently while supporting continuous, discrete, and categorical parameters.
- We additionally develop a camera simulation method including a scene capture component and a physics-based noise model, and provide a procedurally generated indoor virtual environment.

Methodology



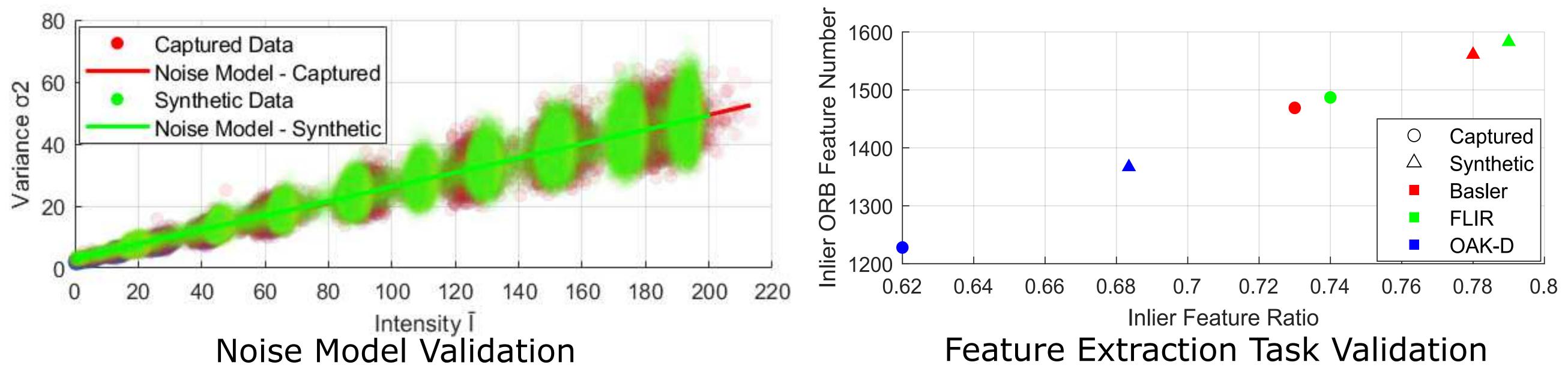
Camera Simulation

- We use the Unreal Engine camera to capture scene renders of simulation environments.
- We incorporate the affine noise model [3] to introduce noise.
- Validation shows our simulation has simular intensity variances and numbers of extracted ORB features with captured images.

TaCOS: Task-Specific Camera Optimization with Simulation

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Optimization

- Camera parameters are optimized with the genetic algorithm, perception tasks are jointly optimized with gradient-based optimizers.
- Discrete camera parameters with interdependency are optimized with the "quantized continuous" method proposed in [1].

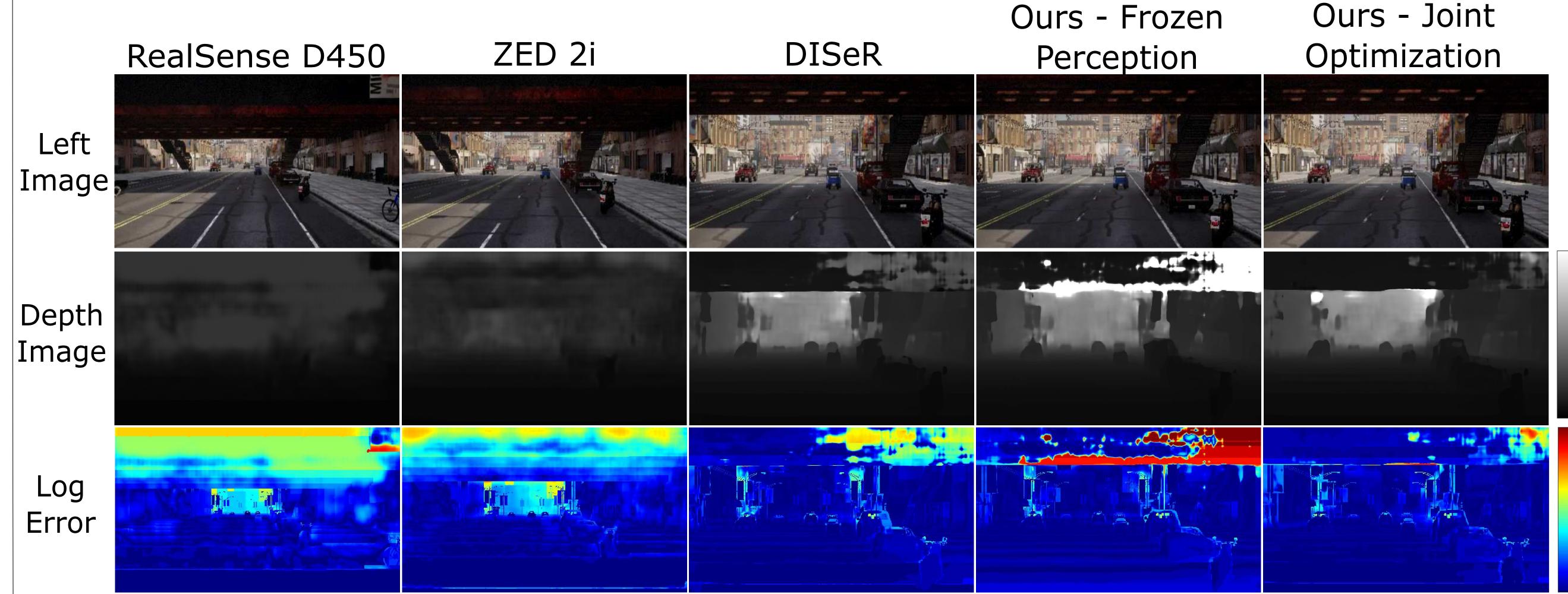
Experiment: Stereo Camera Design For Autonomous Vehicle

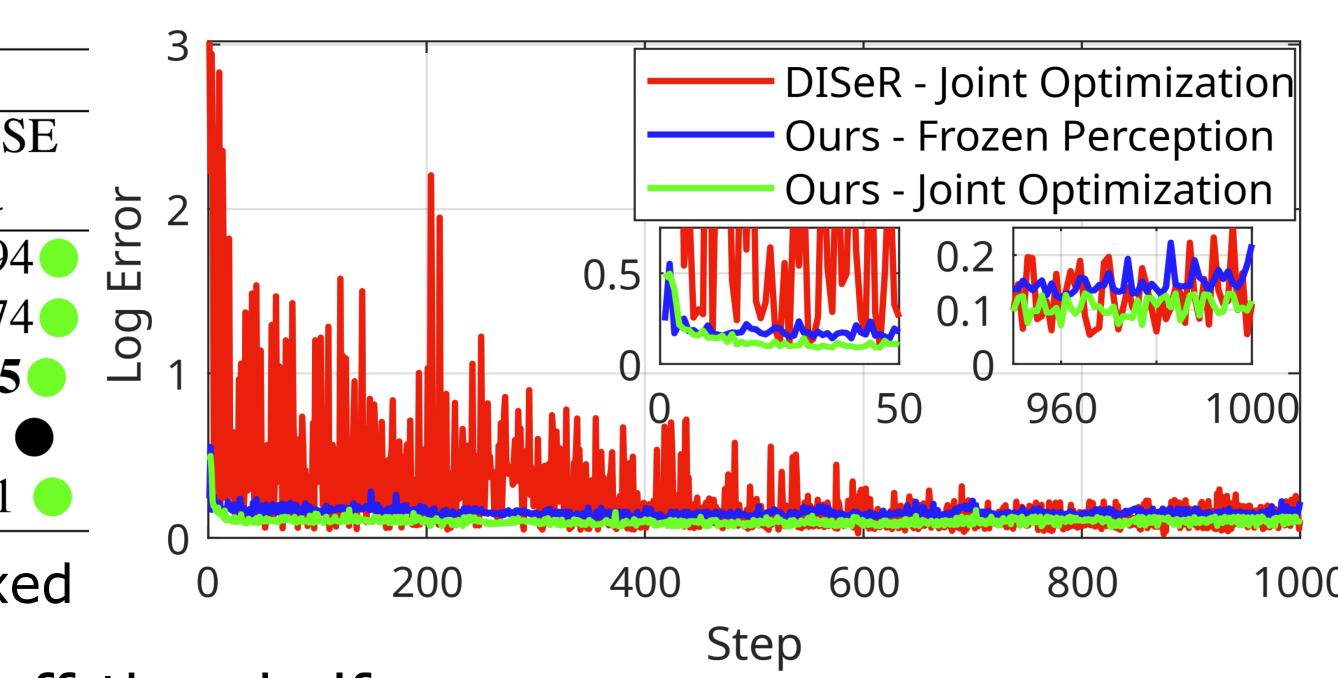
- We optimize the field of view and the baseline distance for a stereo camera pair using the CARLA simulator [4].
- We optimize the stereo camera for depth estimation.

	ra Parameters	Performance		
Baseline	Horizontal FOV	Log Error	RMS	
<i>b</i> (m)	fov (°)	\downarrow	\downarrow	
0.095	87 🔴	0.39	178.94	
0.12	72 🔴	0.29	134.74	
1.84	50	0.16 🛑	75.05	
1.41	50 🔴	0.19	111	
1.6	50 🔴	0.14	79.81	
	<i>b</i> (m) 0.095 • 0.12 • 1.84 • 1.41 •	$\begin{array}{c} b \ (m) & fov \ (^{\circ}) \\ \hline 0.095 \bullet & 87 \bullet \\ 0.12 \bullet & 72 \bullet \\ 1.84 \bullet & 50 \bullet \\ 1.41 \bullet & 50 \bullet \end{array}$	b (m) fov (°) \downarrow $0.095 \bullet$ $87 \bullet$ $0.39 \bullet$ $0.12 \bullet$ $72 \bullet$ $0.29 \bullet$ $1.84 \bullet$ $50 \bullet$ $0.16 \bullet$ $1.41 \bullet$ $50 \bullet$ $0.19 \bullet$	

Optimized • - Fixed

- TaCOS designs camera that outperforms off-the-shelf ones.
- TaCOS achieves similar performance with the SOTA method DISeR but with fewer optimization steps and less time.
- Jointly optimizing task and camera hardware outperforms isolated optimization.

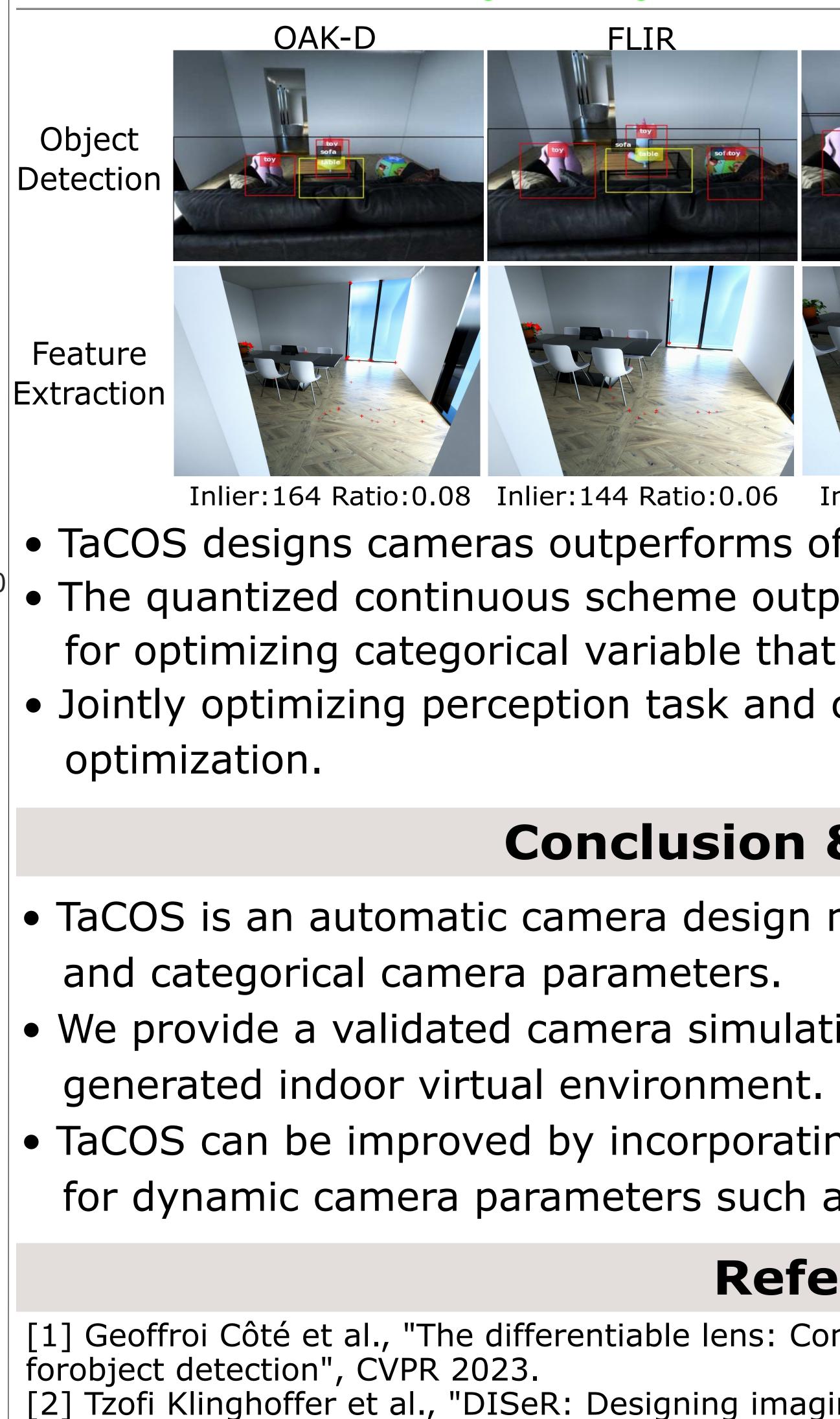




Experiment: Monocular Camera Design For MR Headset

- feature extraction

reature extraction	UII.					- (Optimized	Fixed
	Camera Parameters				Performance			
Camera	Pitch Angle	Focal Length	Sensor Size	Pixel Size	Obstacle Detect.	Object Detect.	Inlier Number	Inlier Ratio
	θ (°)	f (mm)	$w \times h \text{ (mm)}$	p (μ m)	Accuracy ↑	$AP\uparrow$	↑	↑
OAK-D	-20.04 🔴	2.75 •	6.29×4.71 ●	1.55 🔴	1	0.37 🔴	115	0.07
FLIR	-23.28 🔴	3.6	6.2×4.65 ●	1.55 🔴	1	0.37 🔴	77	0.05
Basler	-25.90 🔴	3.6	4.8×3.6 ●	3.75 🔴	1	0.23 😑	131	0.08
Ours - Fully Discrete	-27.87 🔴	4.01 🔴	8.45×6.76 😑	6.6 😑	1	0.43 •	191	0.13
	-24.69 😑	3.77 🔴	8.45×6.76 😑	6.6 😑	1	0.51 😑	189	0.12
Ours - Quantized Continuous	-21.86 🔵	2.99 😑	8.45×6.76 😑	6.6 🔴	1	0.46 •	230	0.13
	-26.34 🔴	2.88 😑	7.31×5.58 😑	4.5 😑	1	0.64 😑	224	0.13
							• •	



Processing 2009.

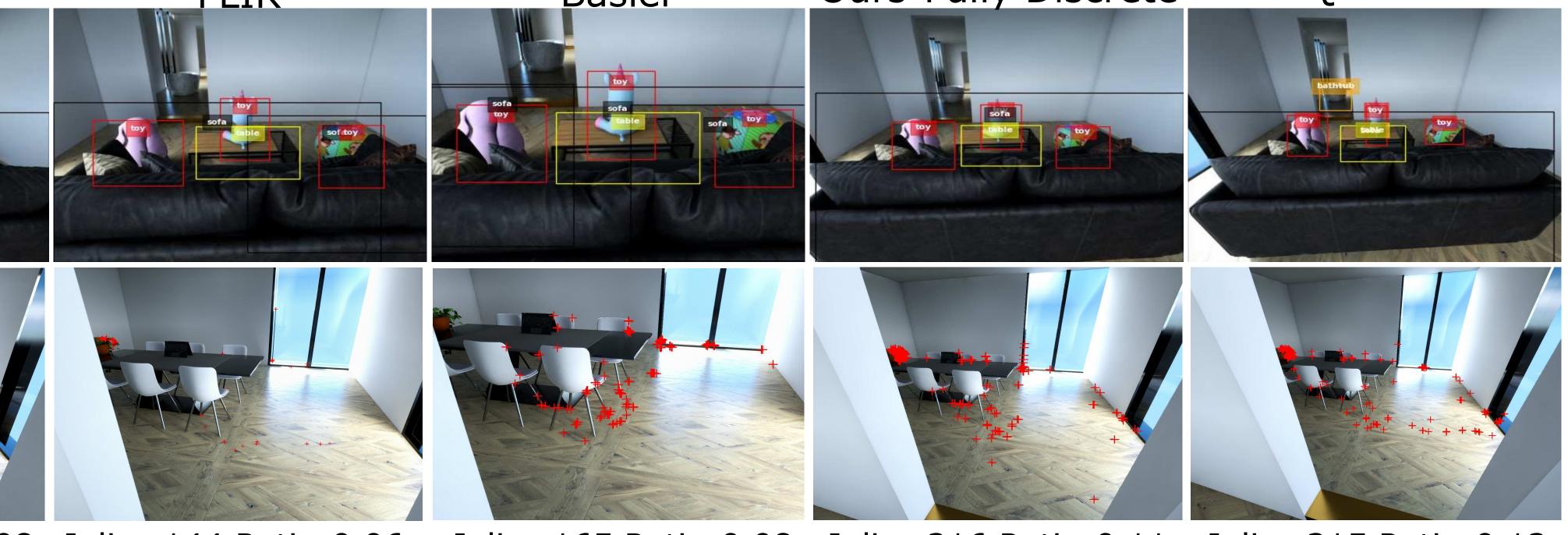




• We optimize the focal length, image sensor, which is a categorical variable including sensor size and pixel size, and camera mounting orientation on the MR headset using our provided procedurally generated indoor environment.

We optimize the monocular camera for object detection, obstacle avoidance, and

Ours-Fully Discrete Ours-Quant. Cont.



Inlier:164 Ratio:0.08 Inlier:144 Ratio:0.06 Inlier:165 Ratio:0.08 Inlier:216 Ratio:0.11 Inlier:215 Ratio:0.12 • TaCOS designs cameras outperforms off-the-shelf ones for all tasks. • The quantized continuous scheme outperforms traditional fully discrete scheme for optimizing categorical variable that includes interdependent variables. Jointly optimizing perception task and camera hardware outperforms isolated

Conclusion & Future Work

TaCOS is an automatic camera design method that supports continous, discrete,

We provide a validated camera simulation method and provide a procedurally

 TaCOS can be improved by incorporating an adaptive camera control algorithm for dynamic camera parameters such as auto-exposure.

References

[1] Geoffroi Côté et al., "The differentiable lens: Compound lens search over glass surfaces and materials

2] Tzofi Klinghoffer et al., "DISeR: Designing imaging systems with reinforcement learning", ICCV 2023. [3] Alessandro Foi, "Clipped noisy images: Heteroskedas-tic modeling and practical denoising", Signal

[4] Alexey Dosovitskiy et al., "CARLA: An open urban driving simulator", CoRL 2017.