



DistillNeRF: Perceiving 3D Scenes from Single-Glance Images by Distilling Neural Fields and Foundation Model Features



Letian Wang, Seung Wook Kim, Jiawei Yang, Cunjun Yu, Boris Ivanovic, Steven Waslander, Yue Wang, Sanja Fidler, Marco Pavone, Peter Karkus

Inference Training Per-Scene Optimized NeRrs Distillation Sparse Hierarchical Voxels Distillation Foundation Model Differentiable Rendering Downstream Tasks Downstream Tasks

- Problem: Perceiving 3D scenes from 2D observations
 - Classic Perception task: un-scalable due to expensive annotation
- NeRF
 Objectiv
 Inference Training
 Sparse Hierarchical Voxels
 Insights:

 Enhan
 Enrich

 Insights:

 Encoder & Lifting Lifting Liftisplates
 Enrich

 Per-Scene Optimized NeRF

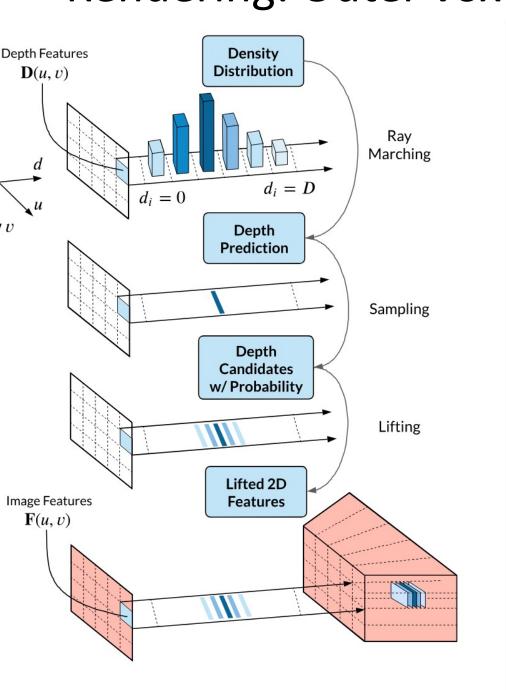
 Distillation Distillation
 Distillation Poundation Model

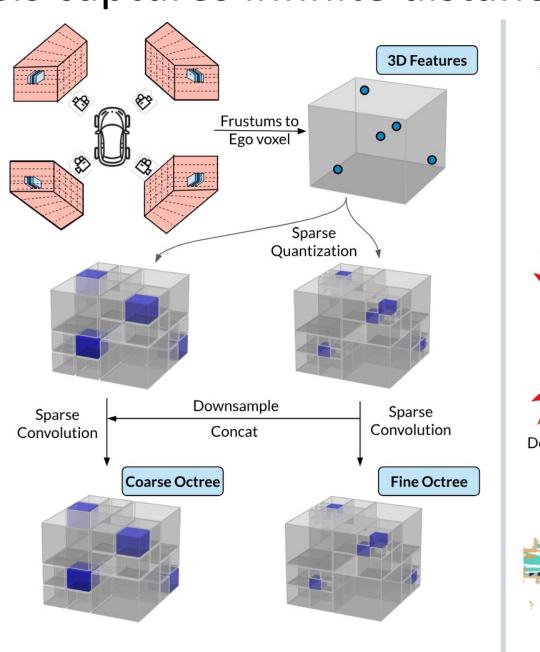
 Open-Vocabulary Query

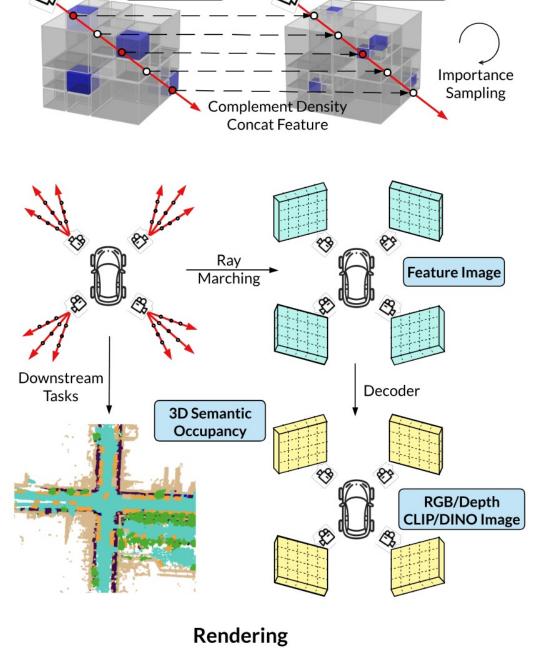
 Zero-Shot 3D Semantic Occupancy Prediction
 RFS
 es

Architecture Details

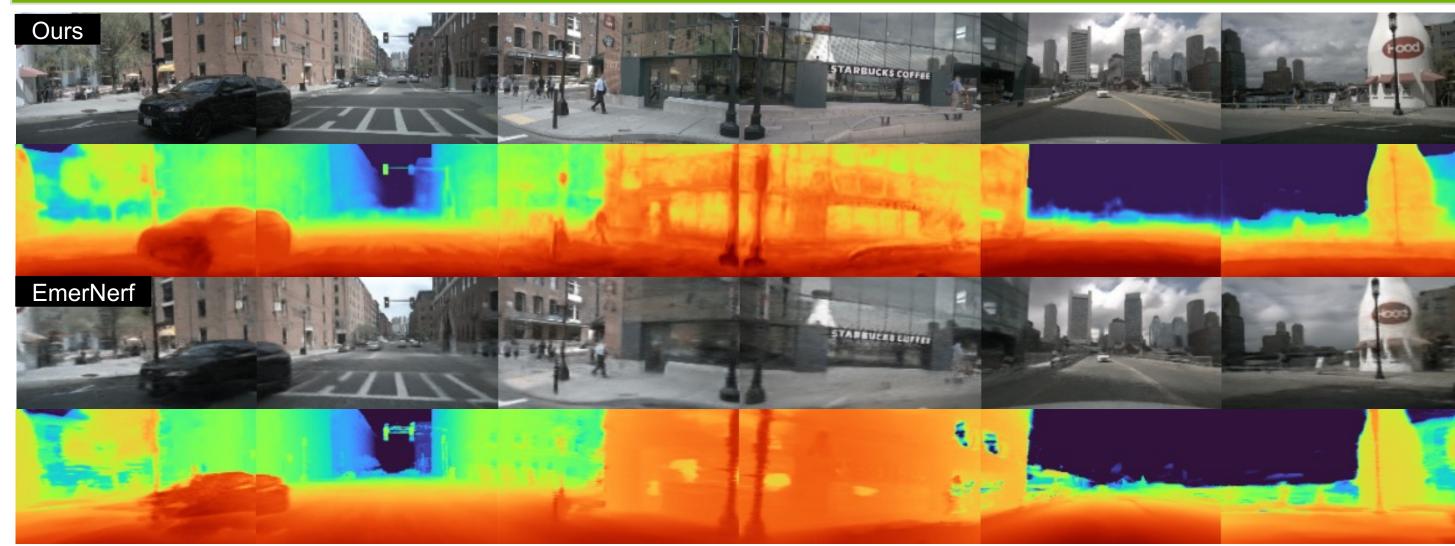
- Sparse Hierarchical Voxel
- Encoder and lift single-view image via two-stage Lift-Splat-Shoot
- Fuse multi-view features via sparse quantization and convolution
- Render from hierarchical sparse voxel and enable downstream tasks
- Parameterized Neural Field
 - Downstream: Keep inner voxels at the real scale and high resolution
 - Rendering: Outer voxels captures infinite distance at a lower resolution



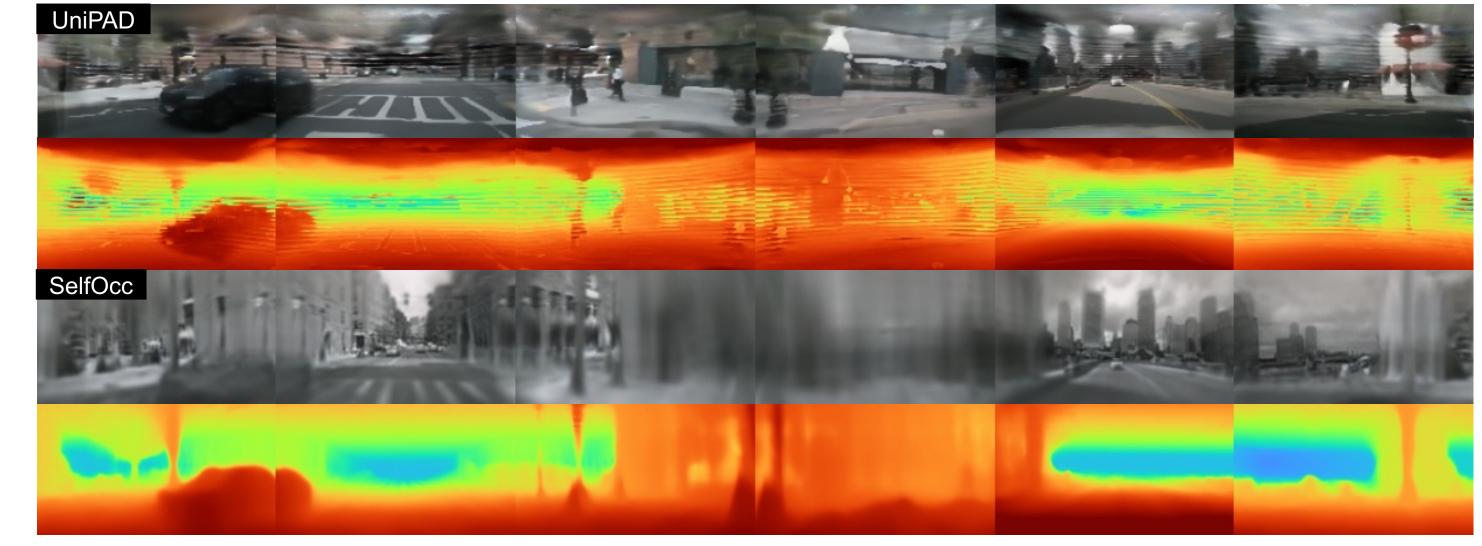




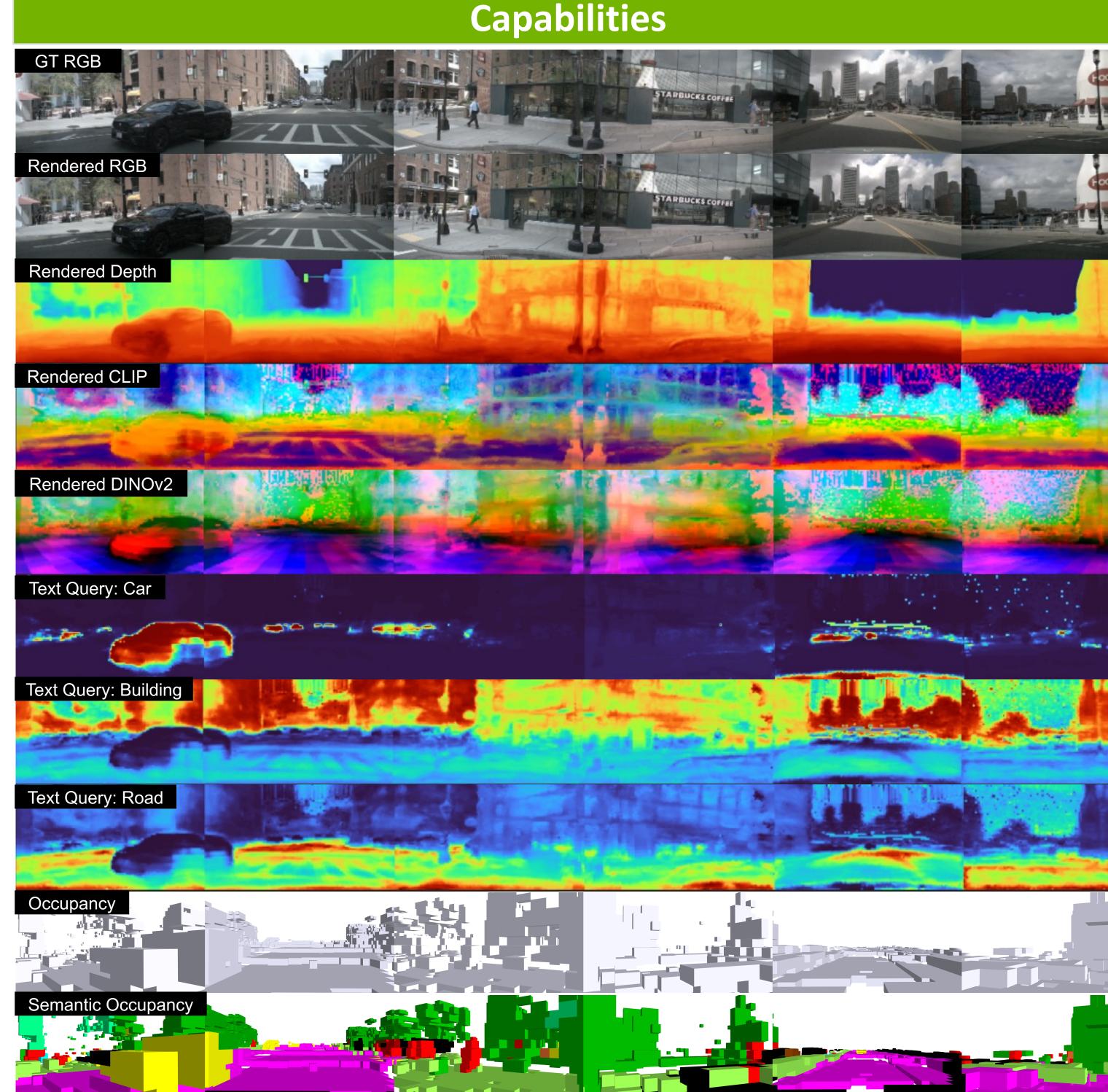
Results & Experiments



On Par with SOTA offline per-scene optimized NeRF



Significantly outperform SOTA online generalizable NeRFs



- Rendering without test-time per-scene optimization
 - Reconstruction & novel-view synthesis: RGB, Depth, Foundation Feat
- Downstream Tasks without annotation
- Open-vocabulary query
- 3D semantic occupancy prediction

....