

# Panoptic Neural Field

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**Supervisors:** Markus Herb, Artem Savkin

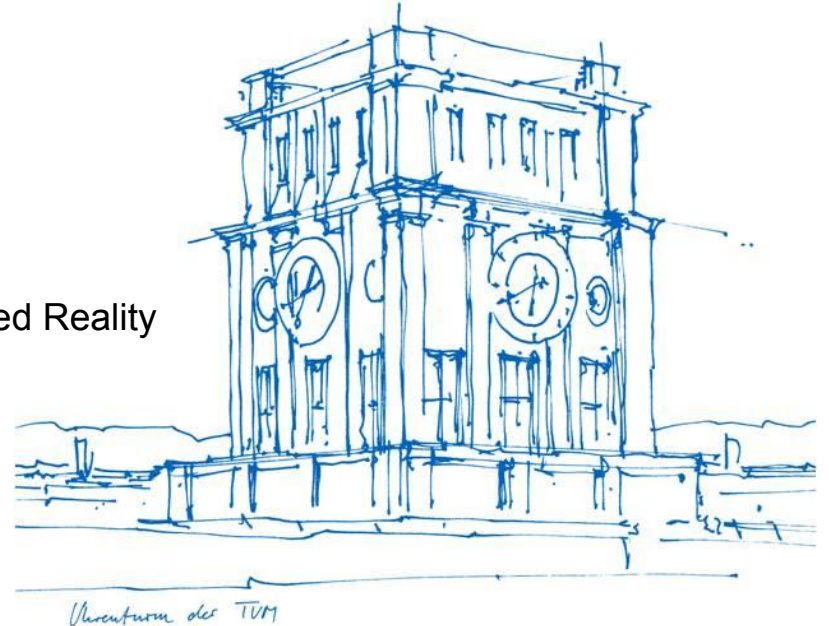
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Technische Universität München

TUM School of Computation, Information and Technology

Chair for Computer Aided Medical Procedures & Augmented Reality

Munich, 03. February 2023



# Agenda

**Motivation**

**Approach**

**Evaluation**

**Conclusion**

Data  
Preparation

NeRF

Semantic  
NeRF

Object  
Aware  
NeRF

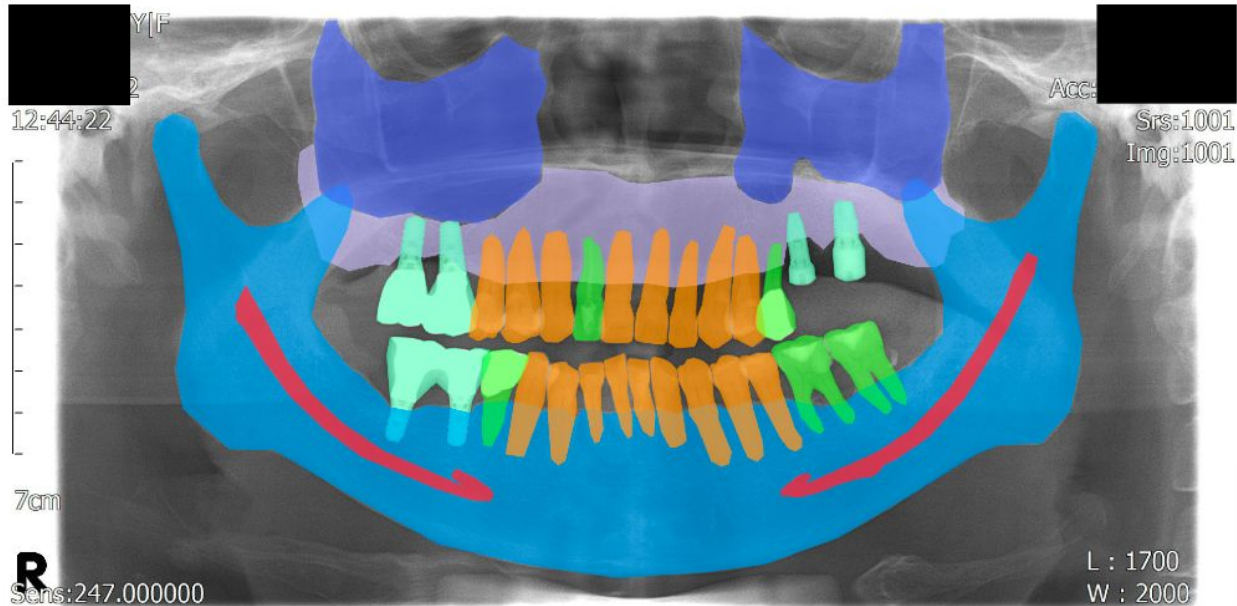




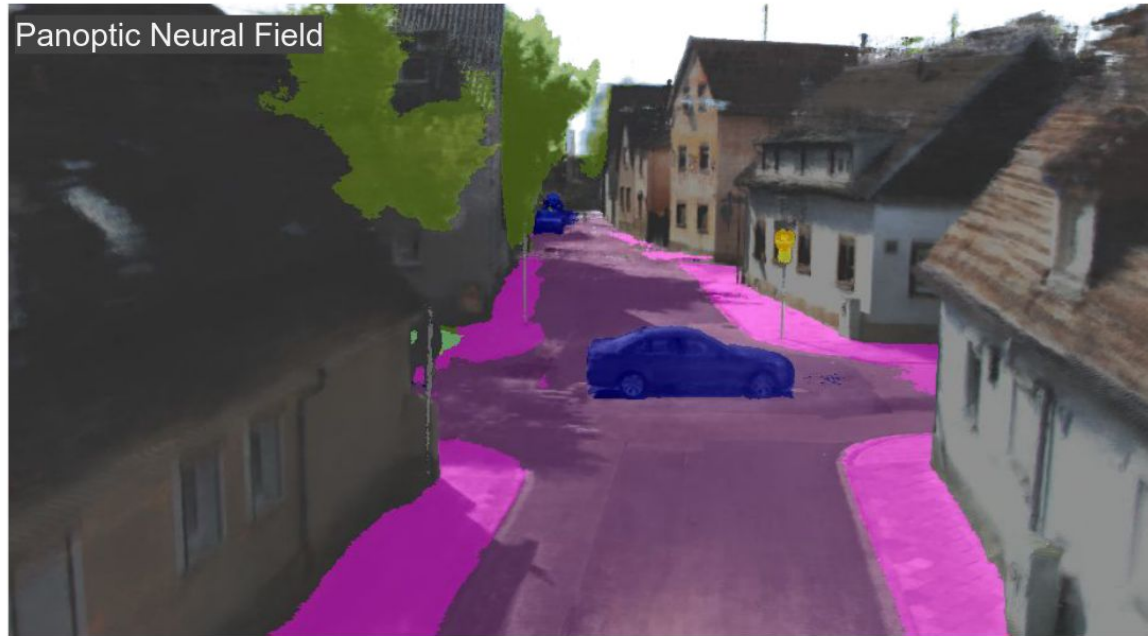
# Panoptic Reconstruction in Autonomous Driving



# Panoptic Reconstruction in Medical Imaging

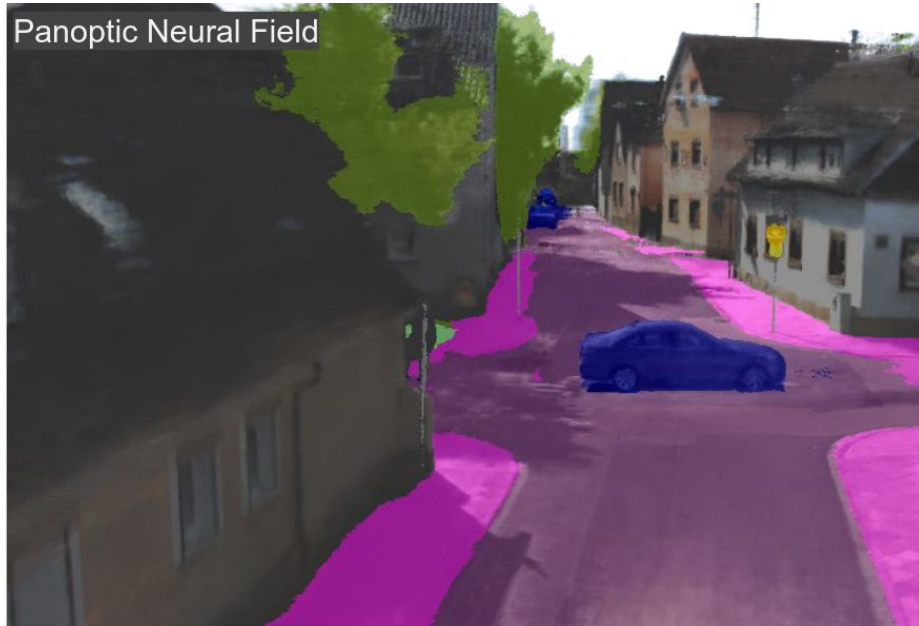


# Panoptic Neural Field



Kundu, Abhijit, et al. "Panoptic neural fields: A semantic object-aware neural scene representation." *CVPR 2022*.

# Panoptic Neural Fields



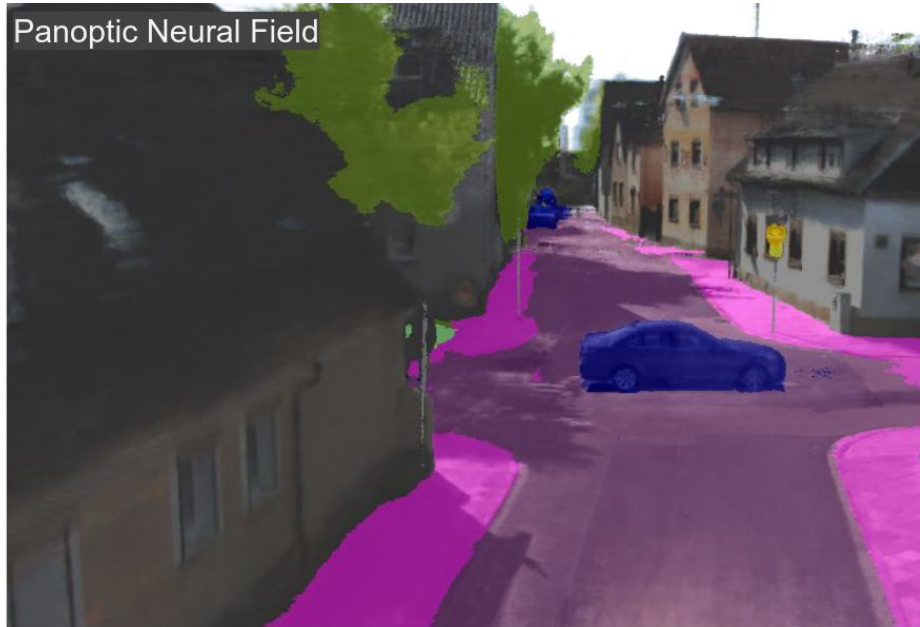
## Idea

- NeRF
- NeRF with Object Decomposition
- NeRF with Semantics
- NeRF with Dynamics

Kundu, Abhijit, et al. "Panoptic neural fields: A semantic object-aware neural scene representation." *CVPR 2022*.



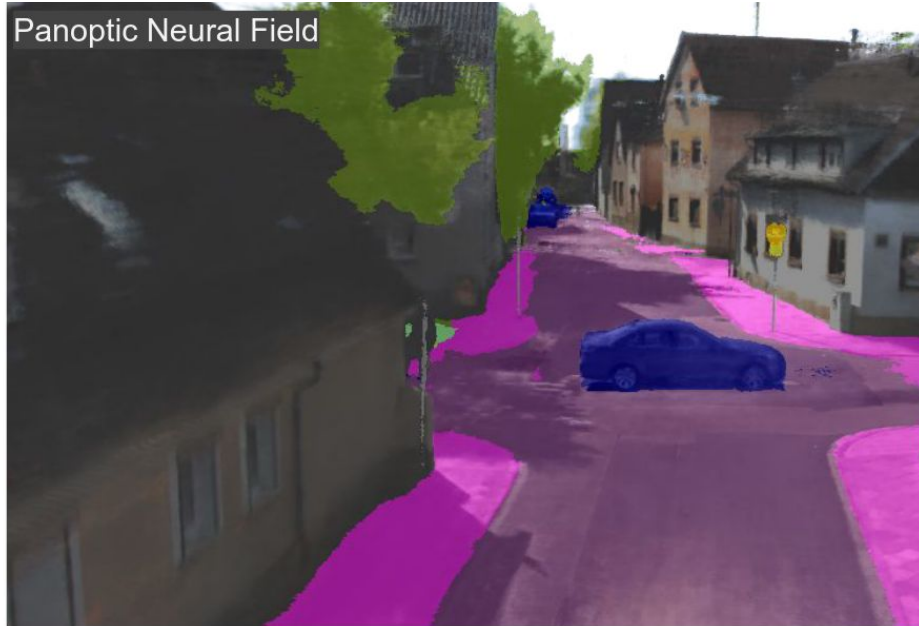
# Panoptic Neural Fields



## State of the Art

Paper	Sem	Obj	Pan	Dyn	Opt	Syn
MeshRCNN [16]		✓				
Total3D [39]	✓	✓				
Atlas [36]	✓					
SLAM++ [50]		✓.				
PanopticFusion [37]	✓		✓			
Kimera [48]	✓					
DynSceneGraphs [49]	✓	✓	✓	✓	✓	
SemanticNerF [66]	✓					✓
NSG [40]		✓		✓		✓
ObjectNeRF [61]		✓				✓
PNF (Ours)	✓	✓	✓	✓	✓	✓

# Panoptic Neural Fields



## Problem

*“Like most other NeRF-style methods, our model is compute-intensive and hence currently only suited for offline applications.”*

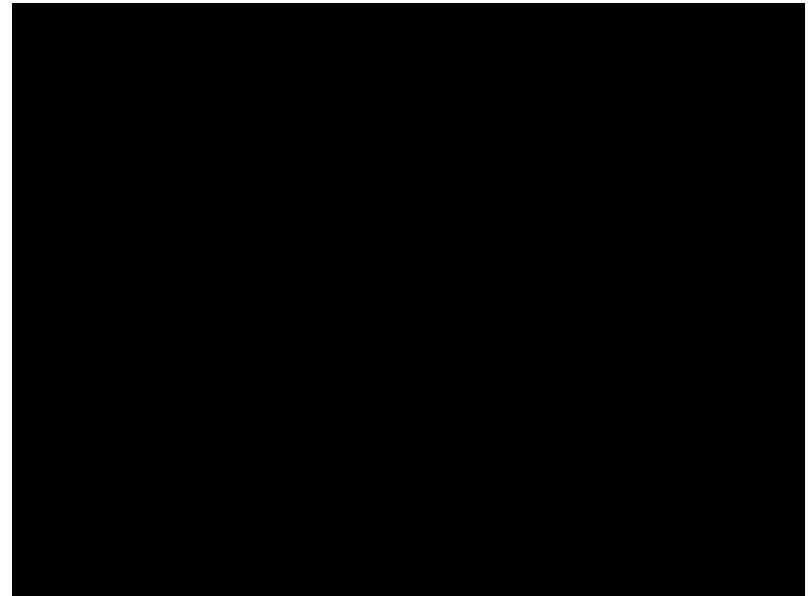
# Instant NGP for Fast NeRF Training



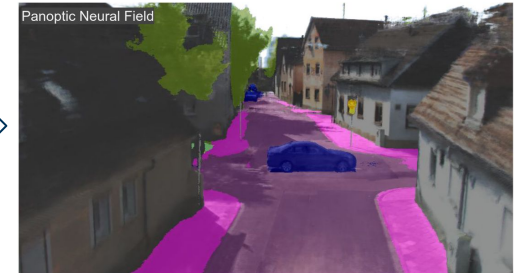
# Goal of the Project

## ★ Use **Instant NGP** to implement a **Neural Panoptic Field**

- ❑ **Given**
  - ❑ Groundtruth Images
  - ❑ Groundtruth Poses
  - ❑ Groundtruth Semantic Labels
  - ❑ Groundtruth Bounding Boxes
  
- ❑ **Tools**
  - ❑ Instant NGP
  - ❑ Kaolin WISP



# Approach



# Data Preparation

## ❑ Given

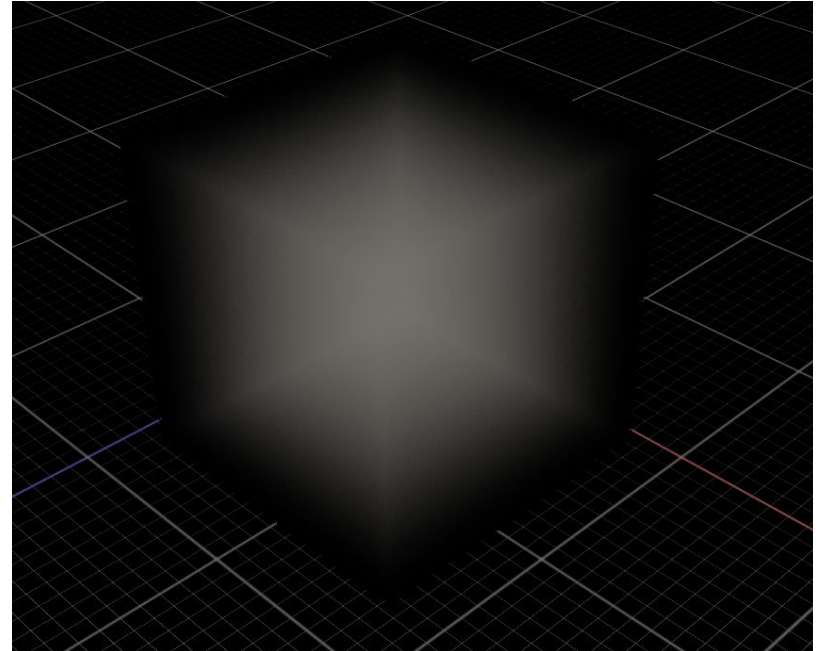
- ❑ Groundtruth Images
- ❑ Groundtruth Poses
- ❑ Groundtruth Semantic Labels
- ❑ Groundtruth Bounding Boxes

## ❑ Tool

### ❑ Instant NGP



Load them according to the assumptions of the framework!



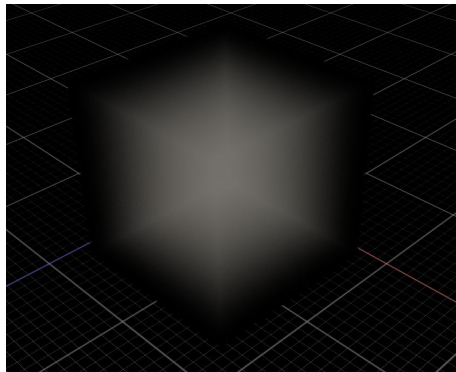
*Unit Cube where the scene should be centered*

# Data Preparation

## ❑ Problem

- ❑ We have to scale and offset the scene onto the unit cube

➔ Offset and scale **the poses**



Unit Cube

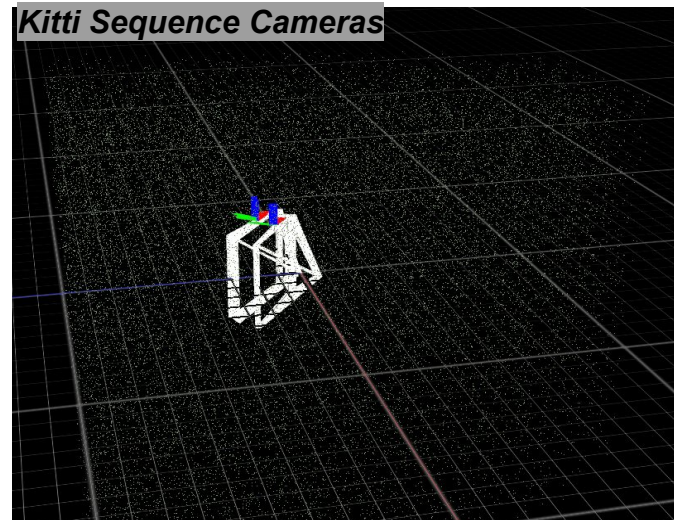
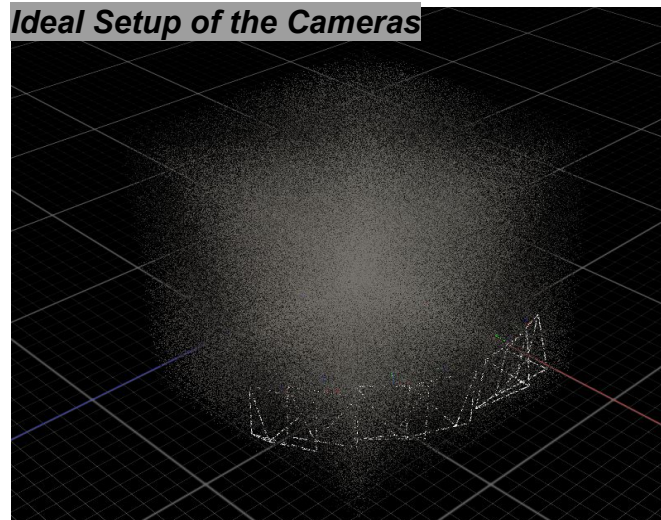


Our scene

# Data Preparation: Poses

## ❑ Challenges

- ❑ Cameras only moving in one direction

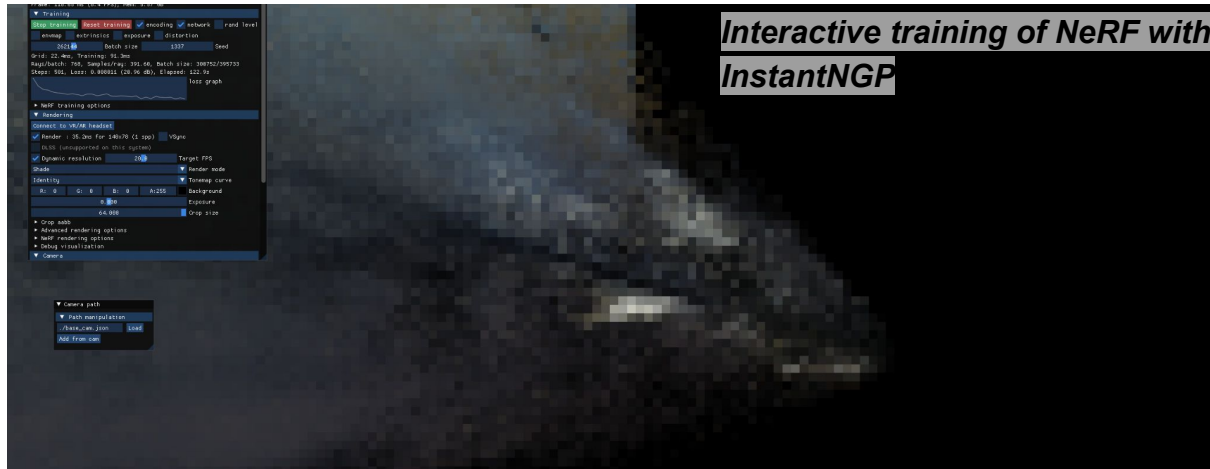




# Data Preparation: Poses

## Challenges

- Cameras only moving in one direction

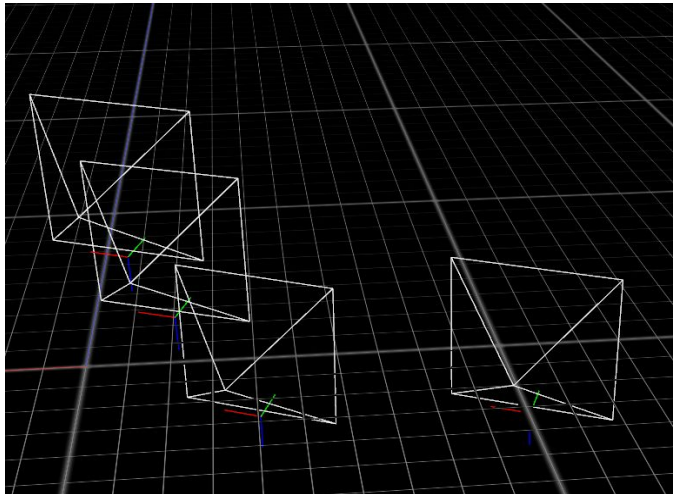


# Data Preparation: Poses

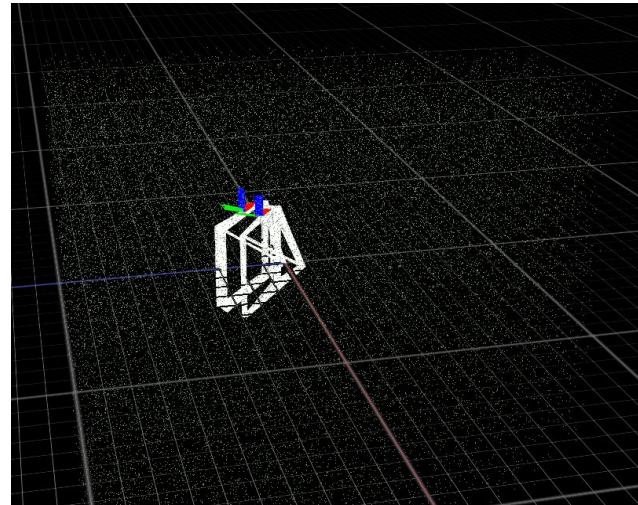
## ❑ Challenges

- ❑ How to test it?

*Loaded Groundtruth Poses*



*Colmap's Output*



# Data Preparation: Poses

## ❑ Challenges

❑ How to test it?

➔ Output of NeRF?

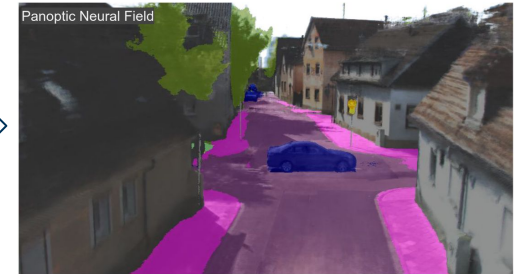


*Colmap's Output*

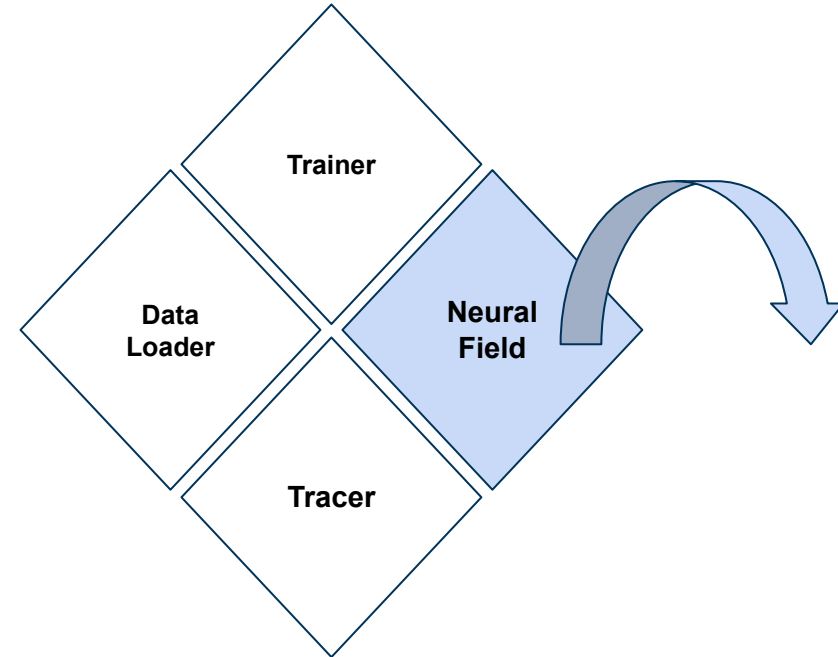
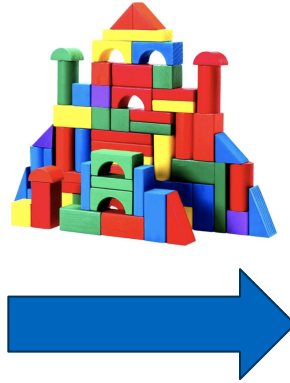
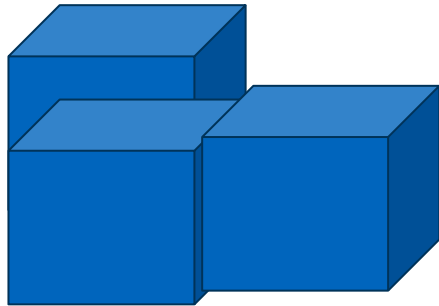


*Loaded Groundtruth Poses*

# Approach

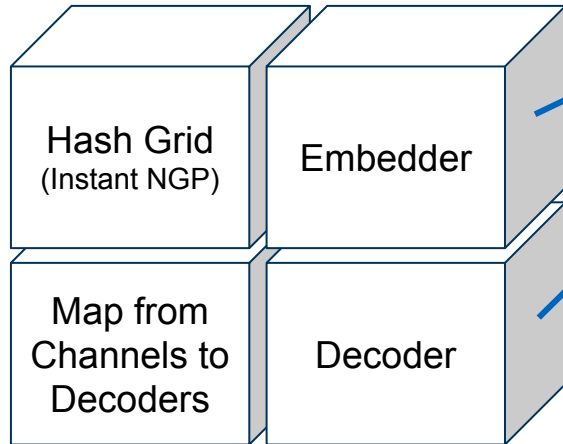


# NeRF with Kaolin WISP

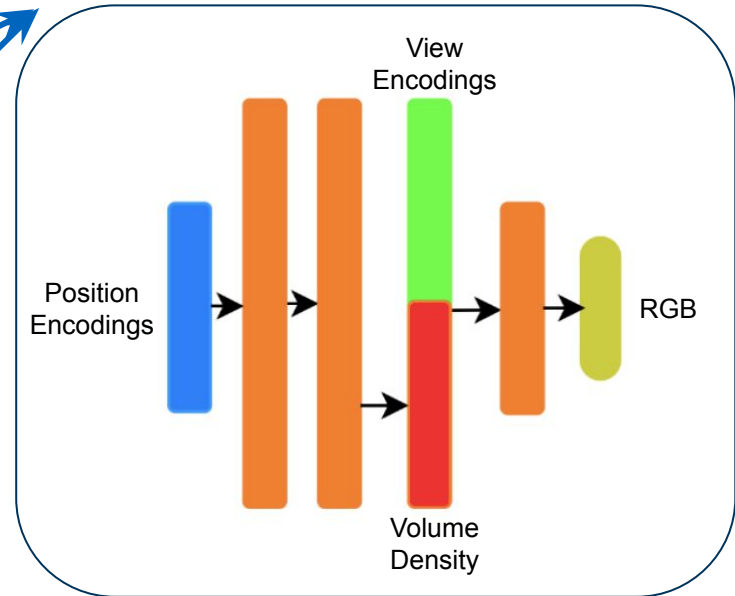


# NeRF with Instant NGP

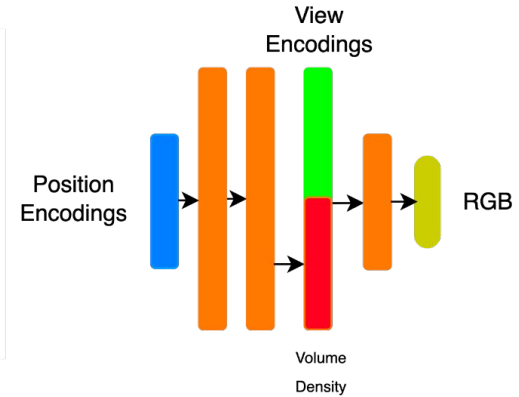
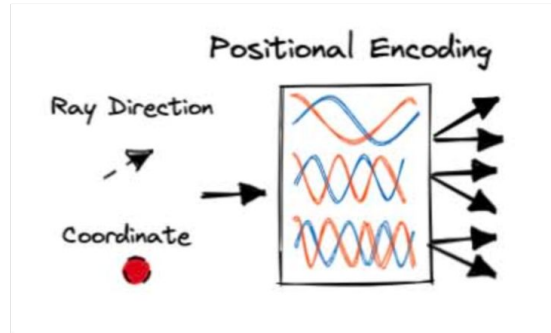
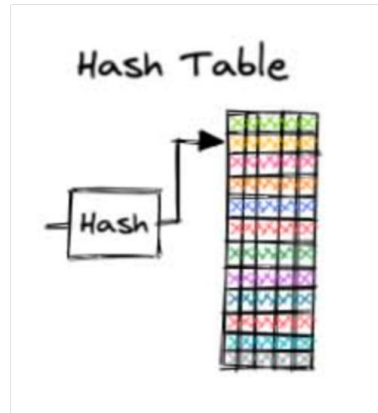
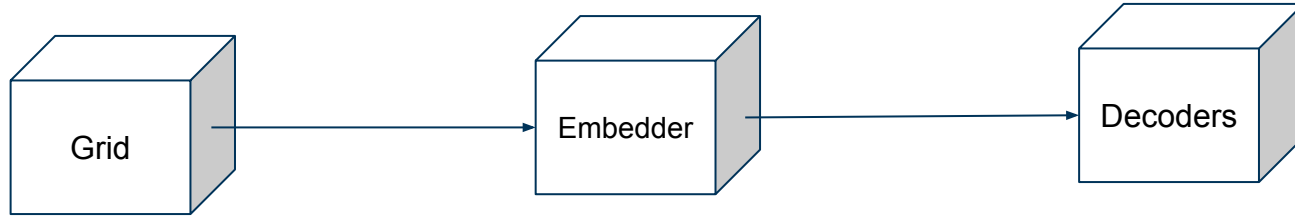
## Neural Field



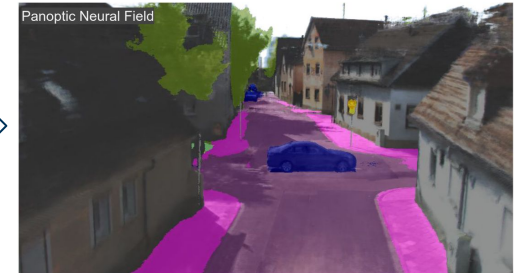
## Embedders + Decoders



# NeRF



# Approach

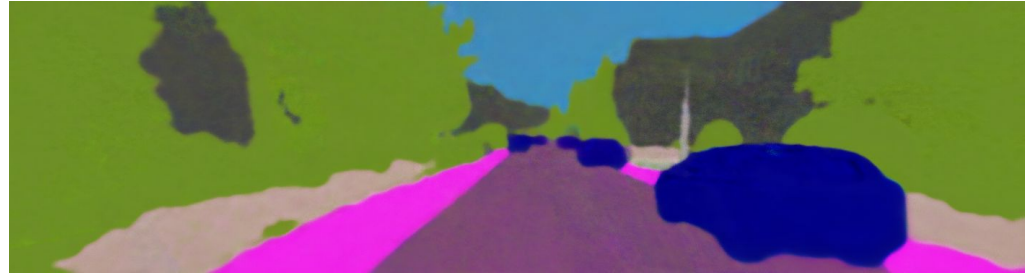




# Semantic NeRF

- Using semantic images as RGB images
- Using labels from KITTI Dataset
  - Using 46 semantic labels
  - Using 8 category labels

With RGB



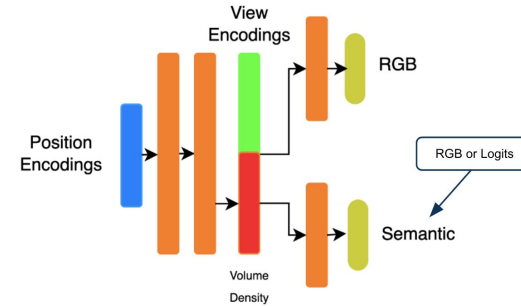
With logits (8 category labels)



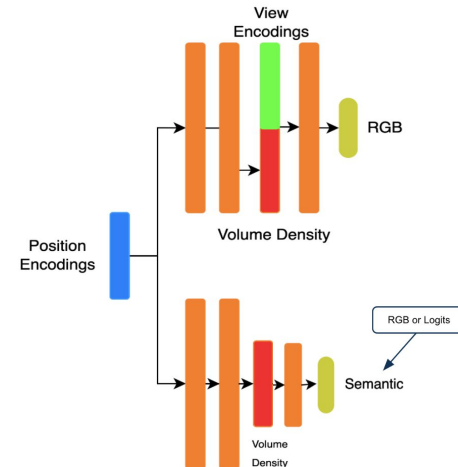
# Semantic NeRF

- Using semantic images as RGB images
- Using labels from KITTI Dataset
  - Using 46 semantic labels
  - Using 8 category labels
- Single density decoder
- Separate density decoders

## Single Density Decoder



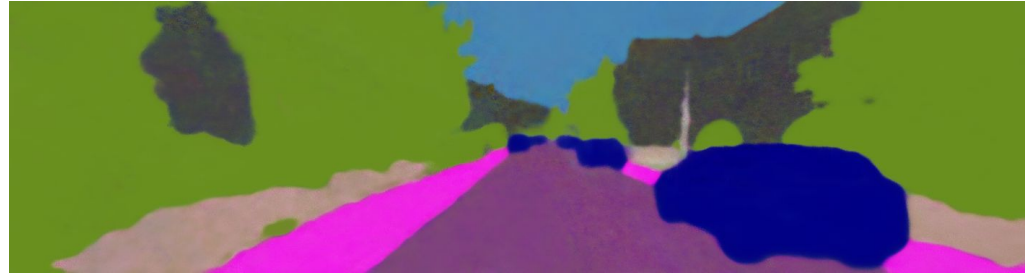
## Two Density Decoders



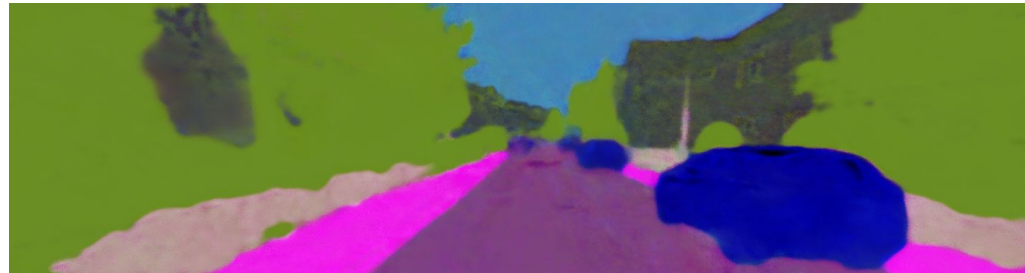
# Semantic NeRF

- Using semantic images as RGB images
- Using labels from KITTI Dataset
  - Using 46 semantic labels
  - Using 8 category labels
- Single density decoder
- Separate density decoders
- Single pass through density decoder
- Double pass through density decoder

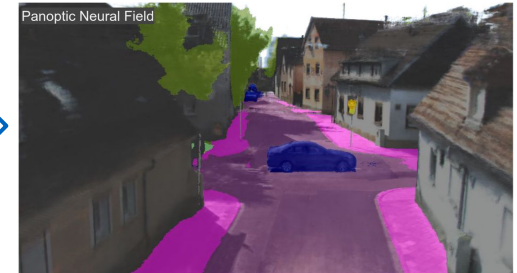
Single pass through density decoder



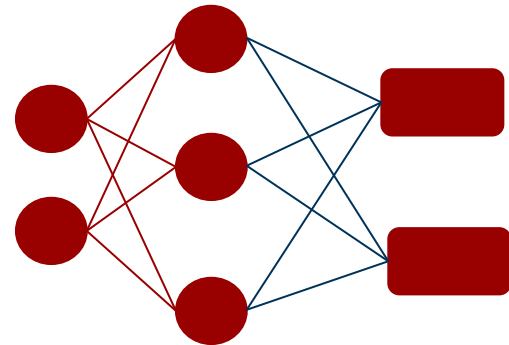
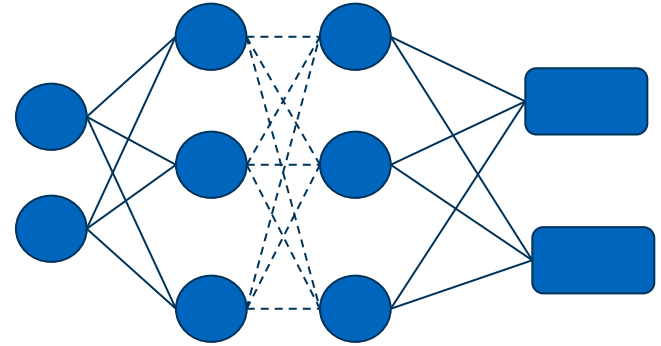
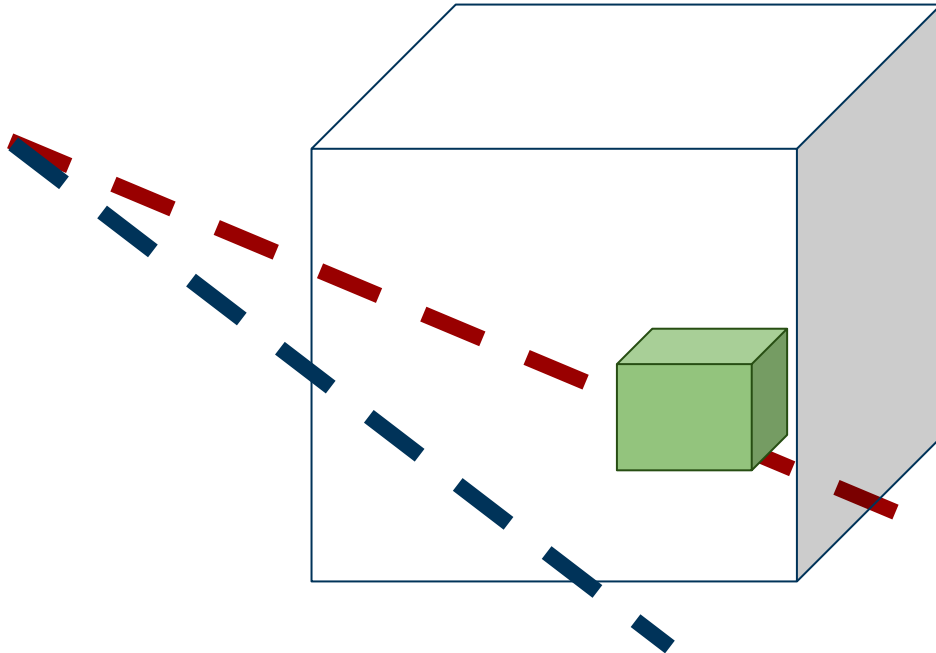
Double pass through density decoder



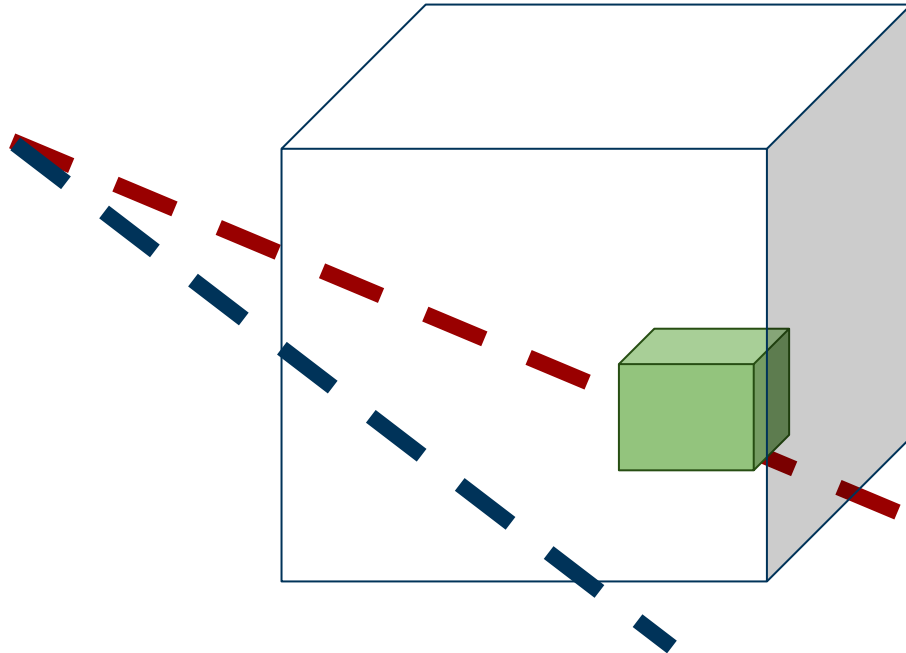
# Approach



# Object-Aware NeRF



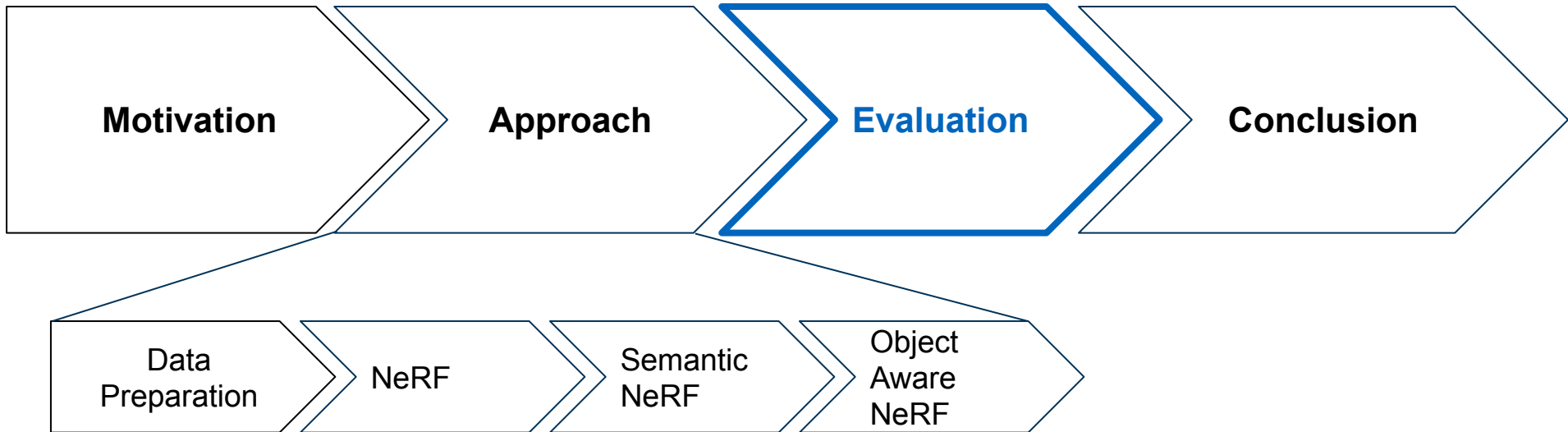
# Object-Aware NeRF



## Problems:

- Too slow!
- Not enough time for given technical knowledge

# Agenda



# Qualitative



Ground Truth



NeRF

- 2 Layers
- 256 Neurons
- 0.001 Learning Rate
- 1.0 Semantic Weight
- 19 Codebook Size

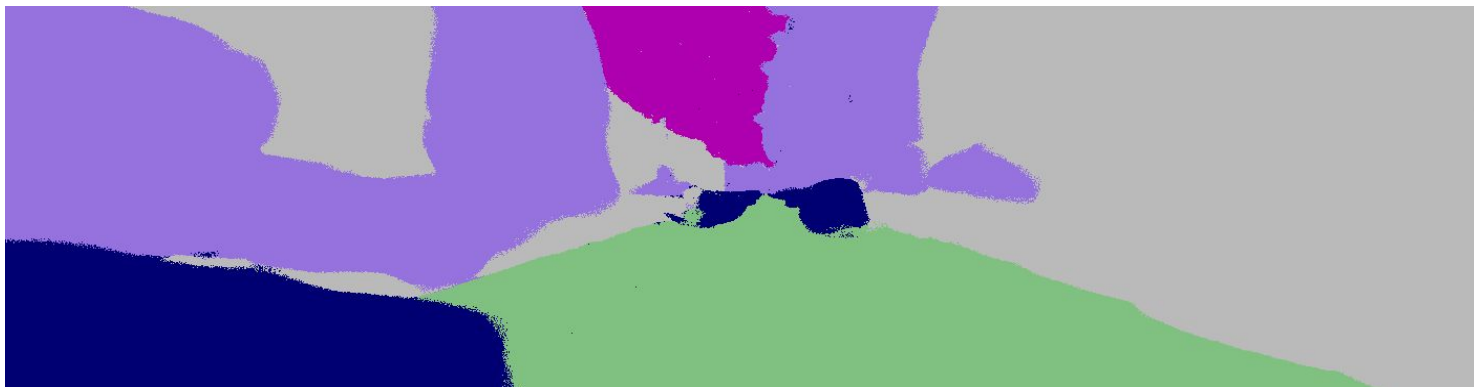


# Qualitative: Semantic NeRF



Ground Truth

- 2 Layers
- 32 Neurons
- 0.001 Learning Rate
- 1 Semantic Weight
- 8 Category Label
- 19 Codebook Size



Semantic Nerf

# Qualitative

- 2 Layers
- 32 Neurons
- 0.001 Learning Rate
- 19 Codebook Size



**Semantic Loss Weight 0.2**



**Semantic Loss Weight 1.0**

# Quantitative Evaluation

## Metrics:

- **mIoU**
  - Mean Intersection over Union

$$IoU = \frac{|A \cap B|}{|A \cup B|}$$

- **PSNR**
  - Peak Signal to Noise Ratio

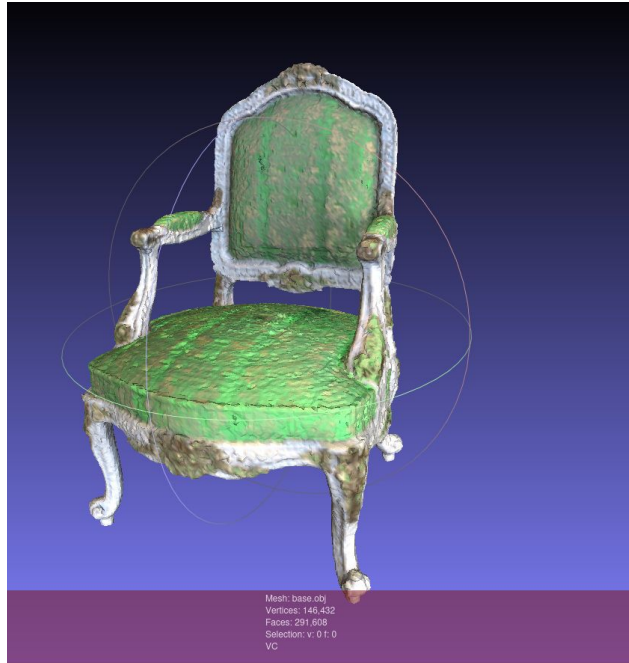
$$PSNR = 10 * \log_{10}\left(\frac{1}{MSE}\right)$$

	<b>Semantics mIoU</b>	<b>Appearance PSNR</b>
<b>PNF</b>	74.28	21.91
<b>ours</b>	53.29	16.82

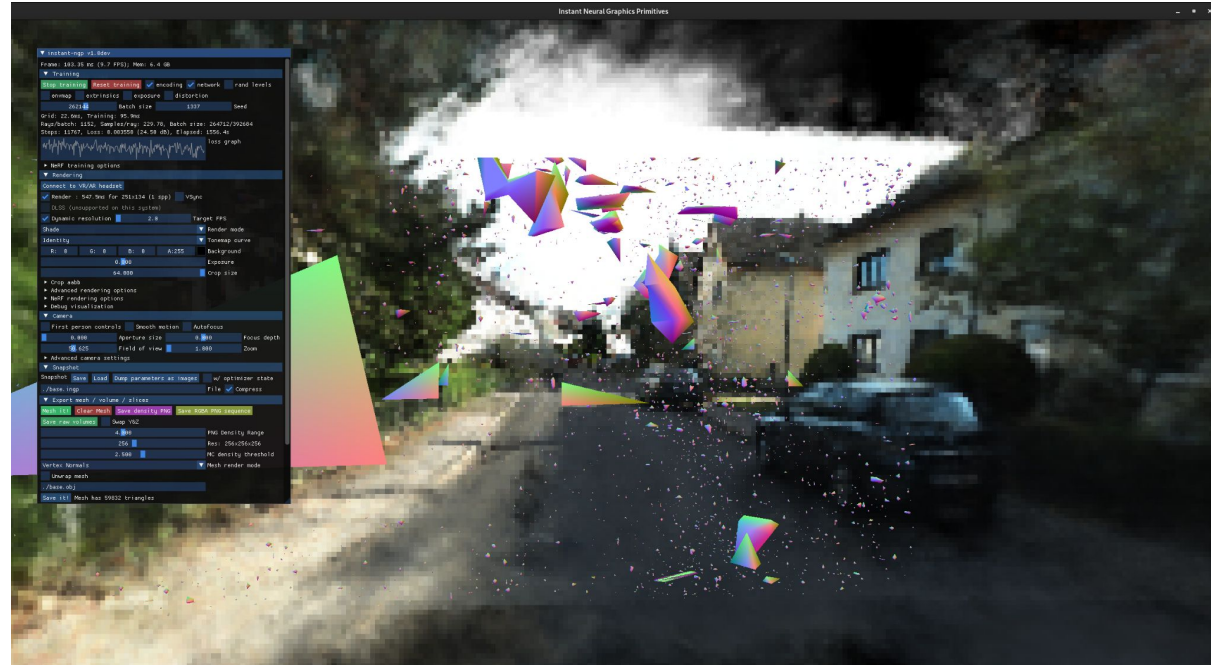
# Quantitative Evaluation

Hidden Dim	Pass Count	Decoder	Prediction	Pixel Accuracy
256	1	1	RGB	0.7526
256	1	2	RGB	0.7541
128	2	1	RGB	0.7455
128	1	1	RGB	0.7540
64	1	2	RGB	0.7476
32	1	1	Logits	0.8219

# 3D Reconstruction



Mesh of the chair



Instant NGP on KITTI



```
graph LR; Motivation --> Approach; Approach --> Evaluation; Evaluation --> Conclusion
```

**Motivation**

**Approach**

**Evaluation**

**Conclusion**

# Conclusion

- **The devil lies in the detail!**
- We managed to build a **Neural Radiance Field** with **Semantic Category Logits**
- Still so much to learn about NeRF:
  - Impact in quality of One Density Decoder vs Two Density Decoders
- Preparing the input to your system and serving the output of your NeRF are no trivial tasks
- Efficient processing of the points is very crucial! No for-loops allowed!

Thank you!



262183 Batch size 1337 Seed

Grid: 22.5ms, Training: 97.2ms

Rays/batch: 1280, Samples/ray: 288.29, Batch size: 266688/502187

Steps: 810, Loss: 0.002400 (26.14 dB), Elapsed: 182.4s

Loss graph

► NeRF training options

▼ Rendering

Connect to VR/AR headset

Render : 373.2ms for 225x126 (1 spp)  VSync

DLSS (unsupported on this system)

Dynamic resolution  2.0 Target FPS

Shade  Render mode

Identity  Tonemap curve

R: 0	G: 0	B: 0	A: 255	Background
				Exposure
				<input type="checkbox"/> Crop size

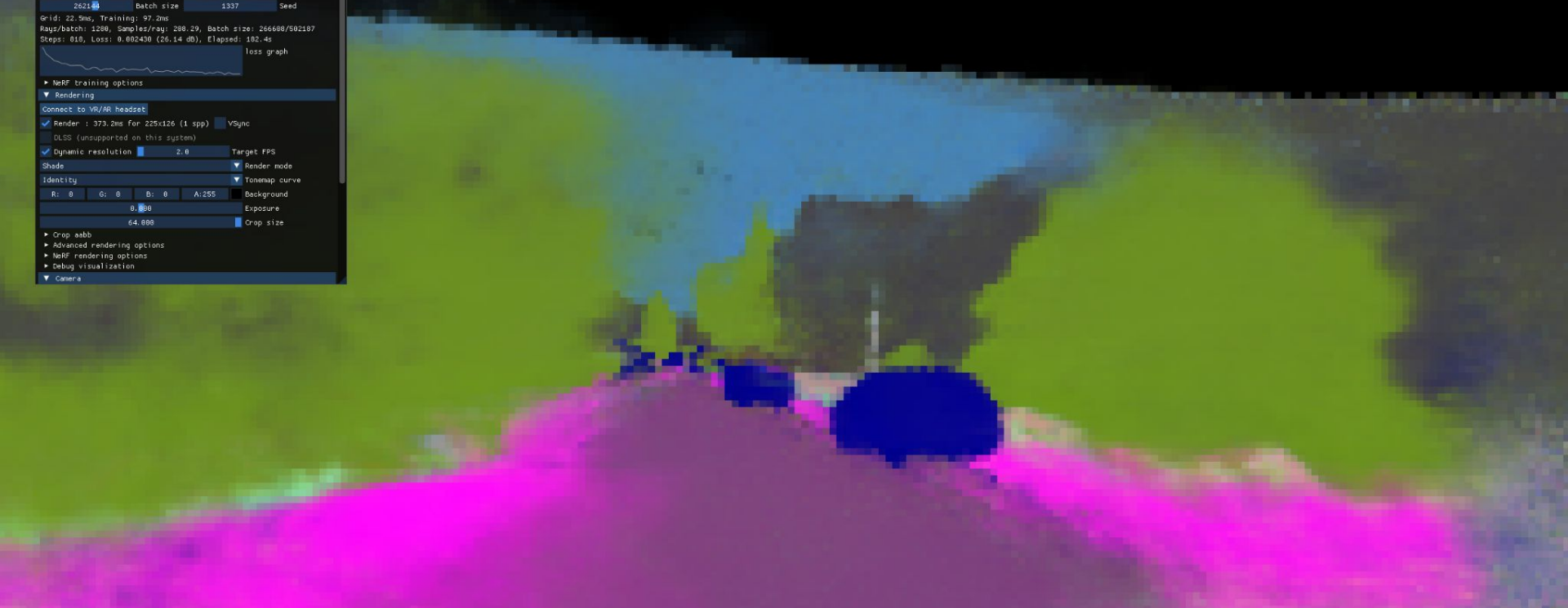
► Crop aabb

► Advanced rendering options

► NeRF rendering options

► Debug visualization

▼ Camera



262143 Batch size 1337 Seed

Training paused  
 Rays/batch: 1536, Samples/ray: 175.16, Batch size: 26900/523186  
 Steps: 1050, Loss: 0.002396 (26.21 dB), Elapsed: 215.0s

Loss graph

► NeRF training options

▼ Rendering

Connect to VR/AR headset

Render : 411.1ms for 286x156 (9 spp)  VSync

DLSS (unsupported on this system)

Dynamic resolution  2.0 Target FPS

Shade ▼ Render mode

Identity ▼ Tonemap curve

R: 0	G: 0	B: 0	A: 255	Background
				Exposure
				<input type="checkbox"/> Crop size

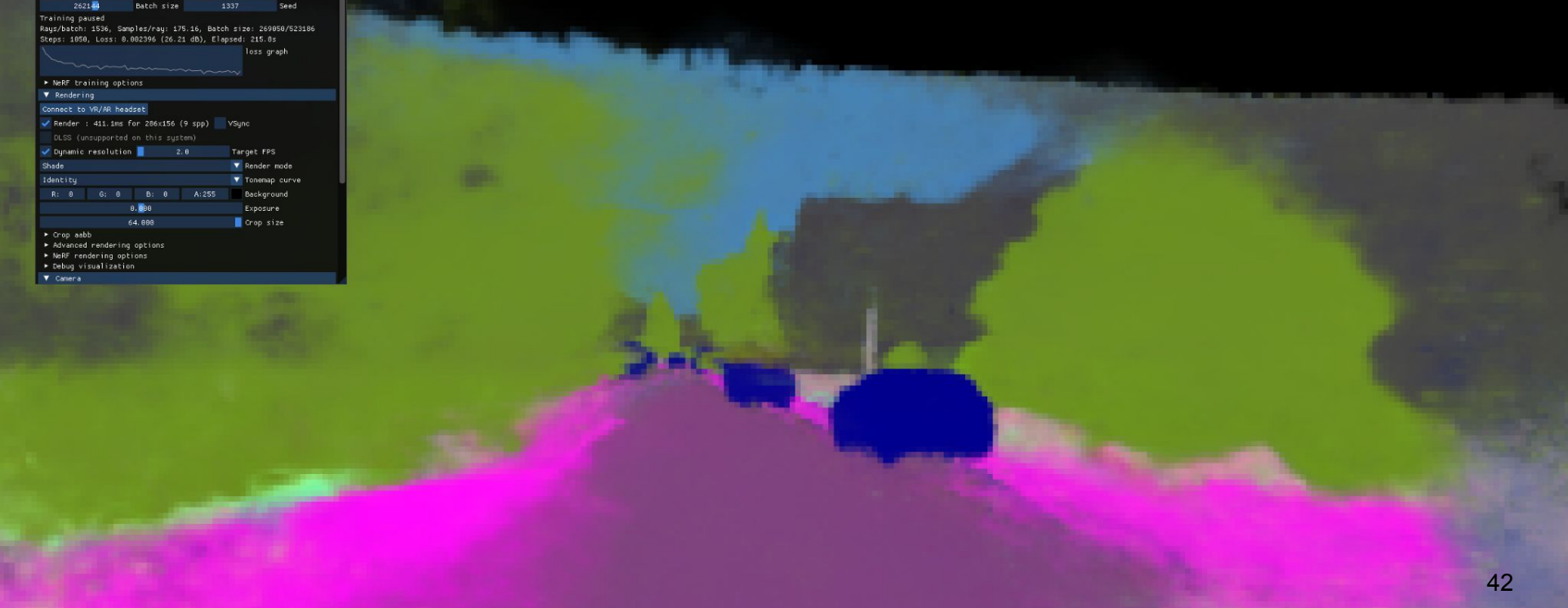
► Crop aabb

► Advanced rendering options

► NeRF rendering options

► Debug visualization

▼ Camera



Training paused

Rays/batch: 1824, Samples/ray: 291.75, Batch size: 298787/466262  
 Steps: 1337, Loss: 0.006191 (22.88 db), Elapsed: 371.6s

loss graph

Start training    Reset training    encoding    network    rand leve  
 enmap    extrinsics    exposure    distortion

262149   Batch size: 1337   Seed

NRf training options  
 Rendering  
 Connect to VR/AR headset  
 Render : 381.3ms for 199x109 (42 spp)    VSync  
 DLSS (unsupported on this system)  
 Dynamic resolution   2.0   Target FPS  
 Shade    Render mode  
 Identity    ToneMap curve  
 R: 0   G: 0   B: 0   Ai:255   Background  
 0.000   Exposure  
 64.000   Crop size  
 Crop sabb  
 Advanced rendering options  
 NRf rendering options  
 Debug visualization  
 Camera

Camera path  
 Path manipulation  
 ./base\_cam.json   Load  
 Add from cam



▼ Training

Start training Reset training  encoding  network  rand level

enemap extrinsics exposure distortion

2621 Batch size 1337 Seed

Training paused

Rays/batch: 1824, Samples/ray: 291.75, Batch size: 298787/466262

Steps: 1937, Loss: 0.006191 (22.88 db), Elapsed: 423.1s

loss graph

► Nerf training options

▼ Rendering

Connect to VR/AR headset

Render : 335.8ms for 199x109 (23 spp)  VSync

DLSS (unsupported on this system)

Dynamic resolution  2.0 Target FPS

Shade  Render mode

Identity  ToneMap curve

R: 0 G: 0 B: 0 Ai:255 Background

0.000 Exposure

64.000 Crop size  Crop size

► Crop sabb

► Advanced rendering options

► Nerf rendering options

► Debug visualization

▼ Camera

▼ Camera path

▼ Path manipulation

./base\_cam.json Load

Add from cam



Training

Stop training  Reset training  encoding  network  rand level

enmap extrinsics exposure distortion

262149 Batch size 1337 Seed

Grid: 22.4m, Training: 91.3ms  
Rays/batch: 768, Samples/ray: 391.48, Batch size: 388752/395733  
Steps: 591, Loss: 0.008011 (20.96 dB), Elapsed: 123.92

loss graph

NeRF training options

Rendering

Connect to VR/AR headset

Render : 35.2ms for 148x78 (1 spp)  VSync

DLSS (unsupported on this system)

Dynamic resolution 20.0 Target FPS

Shade  Render mode

Identity  Tone-map curve

R: 0 G: 0 B: 0 A:255 Background

0.000 Exposure

64,000 Crop size

Crop sabb

Advanced rendering options

NeRF rendering options

Debug visualization

Camera

Camera path

Path manipulation

../base\_cam.json Load

Add from cam

