

IEEE VR



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# A Hand-on Tutorial on 3DGS for XR Applications

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**T2: A PRACTICAL GUIDE TO RADIANCE FIELDS FOR XR  
RESEARCH AND APPLICATIONS**

# Agenda

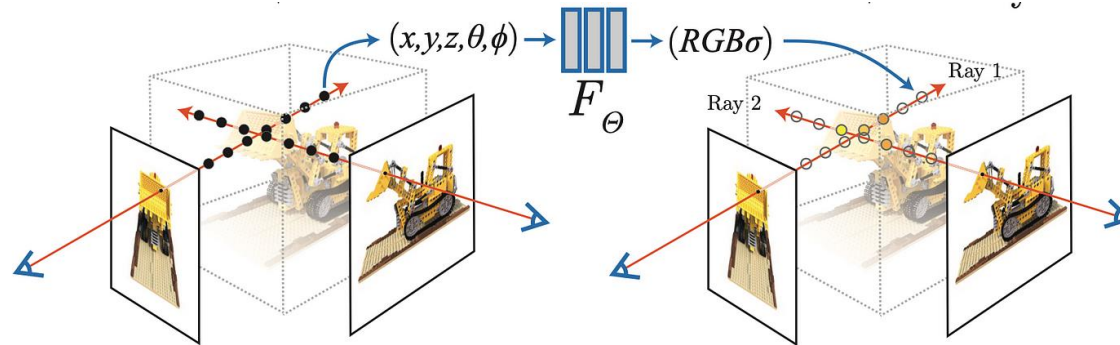
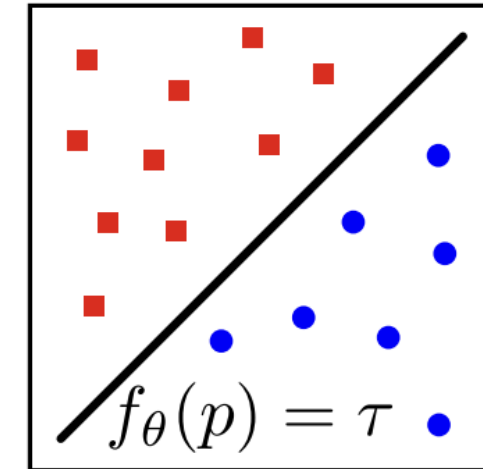
- Brief introduction to **3D Gaussian Splatting (3DGS)**
- Hands on of **UnityGaussianSplatting**
- Live demo on UnityGaussianSplatting toolkit
  - Scene rendering in VR
  - Interaction & editing



# 3D World World Representation

## Implicit Functions (Implicit & Continuous)

- Represent 3D world with neural networks
- Compress the 3D scene in to a function
- NeRF achieved the photorealistic representation

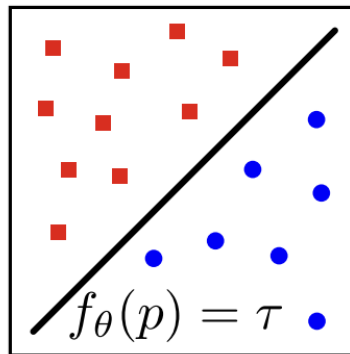


Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019



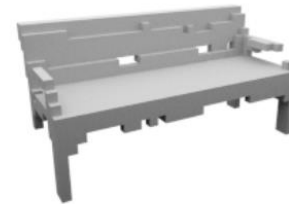
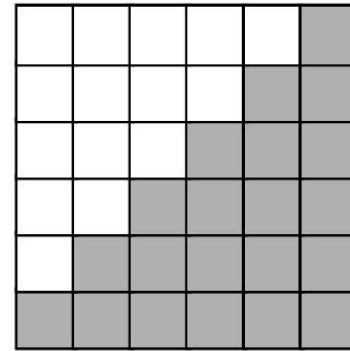
# 3D World Representation

## Implicit

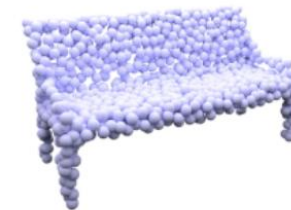
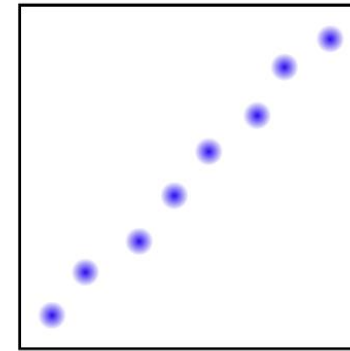


Scene Function

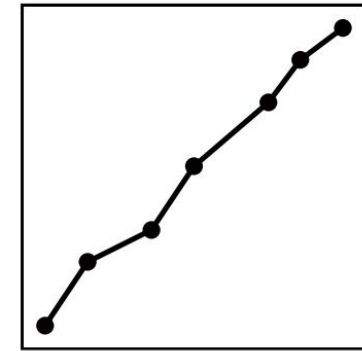
## Explicit



Voxel



Point Cloud



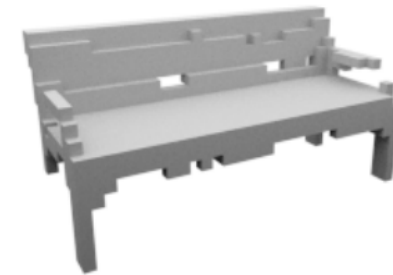
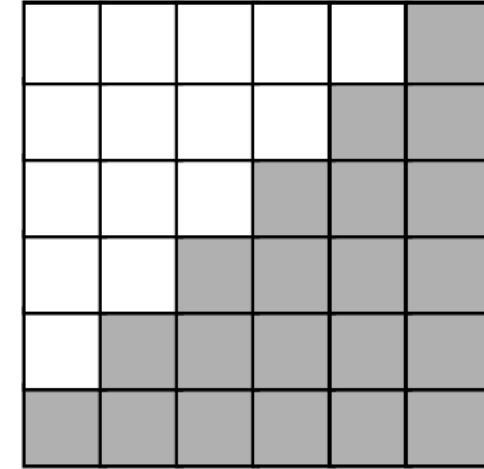
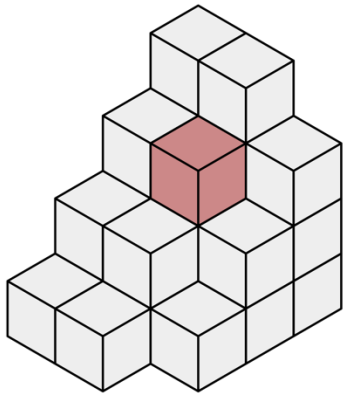
Mesh

Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019

# 3D World Representation

## Voxels (Explicit & Discrete)

- Represent 3D space with grids
- Easy to process
- Cubic memory  $O(n^3) \rightarrow$  limited resolution



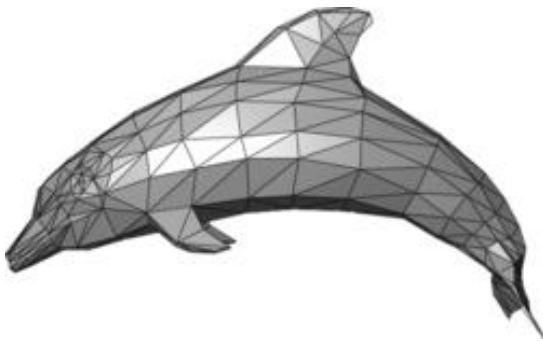
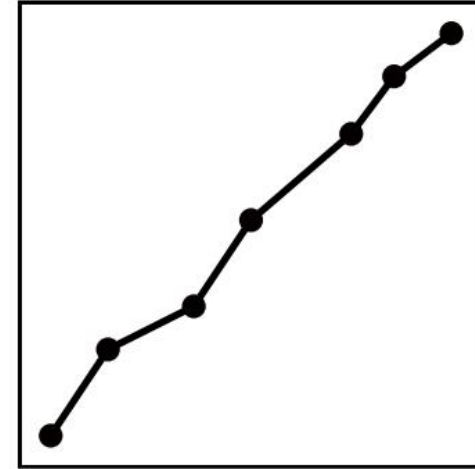
Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019

# 3D World Representation

## Mesh (Explicit & Discrete)

- Represent objects surface with vertices and faces
- Limited number of vertices

→ reconstruction accuracy is limited

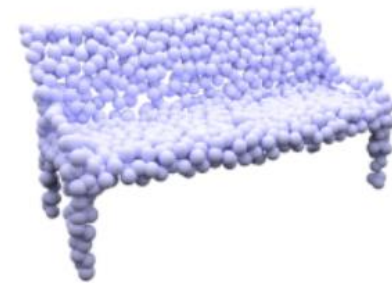
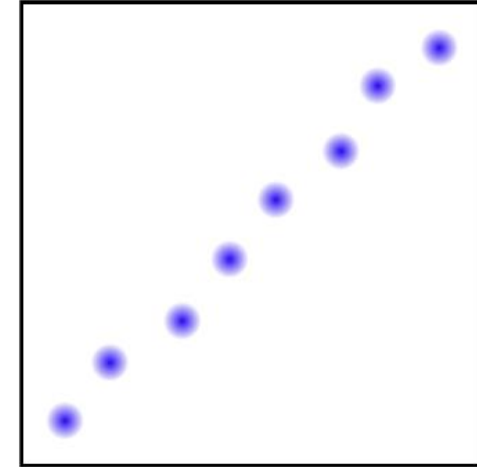


Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019

# 3D World Representation

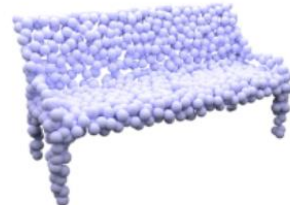
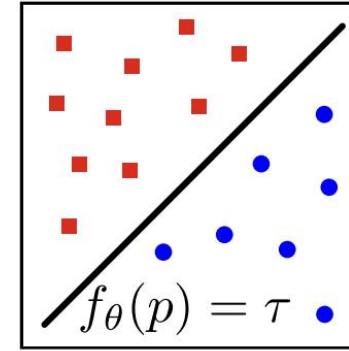
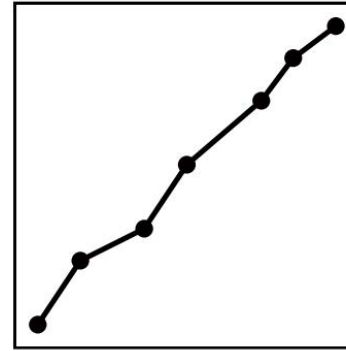
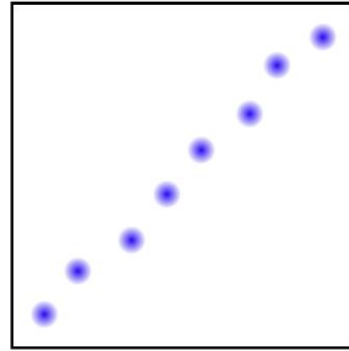
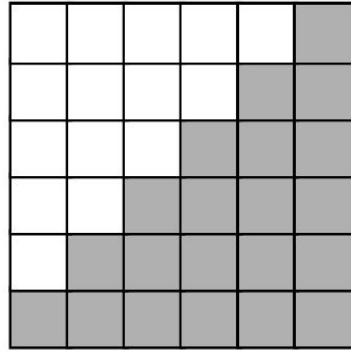
## Point Clouds (Explicit & Discrete)

- Represent surface with 3D points
- Does not have connected surface
- Limited number of points



Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019

# How to Represent the 3D World?



Voxel

Point Cloud

Mesh

Implicit Function

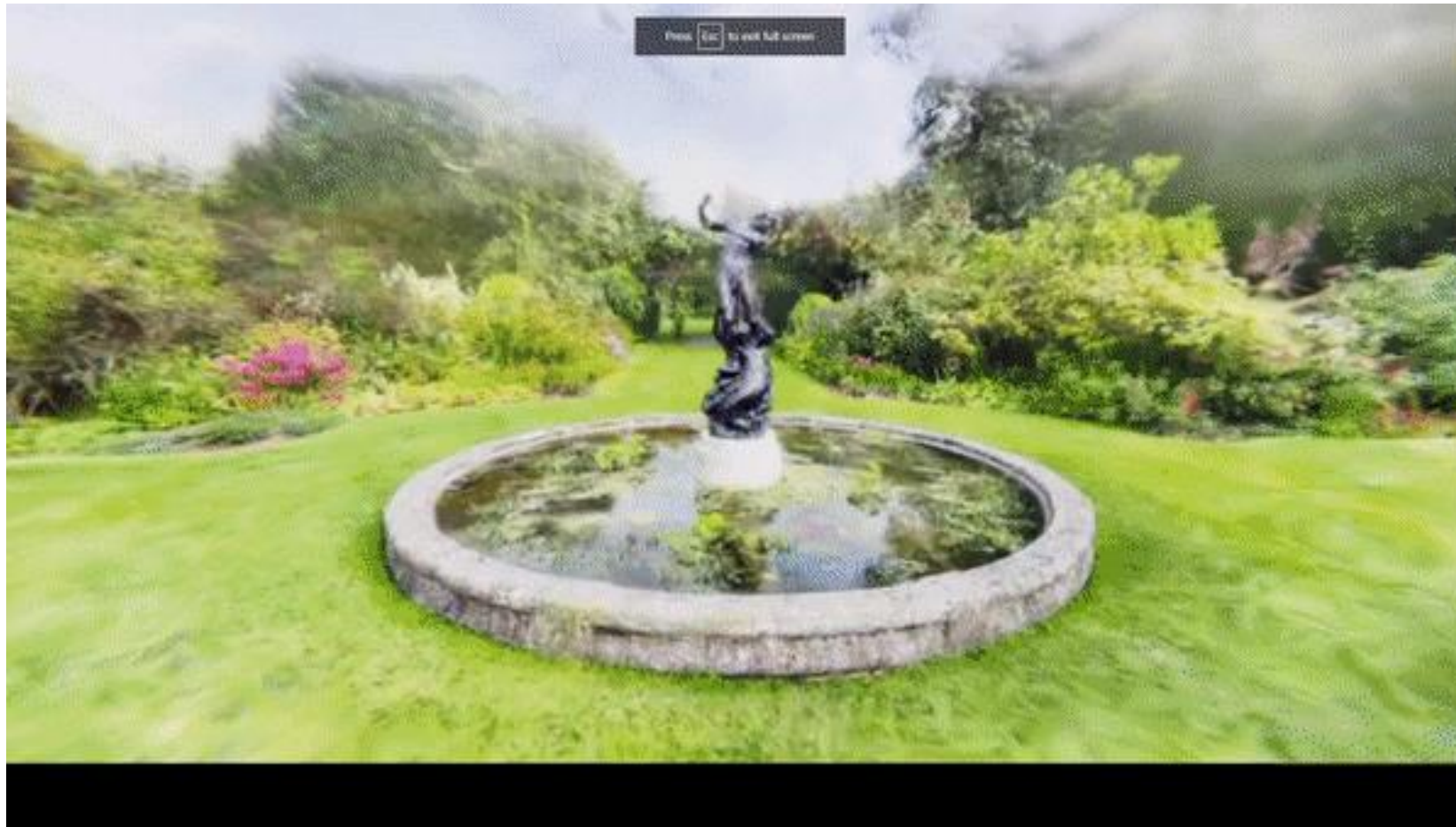
Which is the best representation of 3D world for **photo realistic** and **fast rendering speed**?

Mescheder, Oechsle, Niemeyer, Nowozin and Geiger: Occupancy Networks: Learning 3D Reconstruction in Function Space. CVPR, 2019



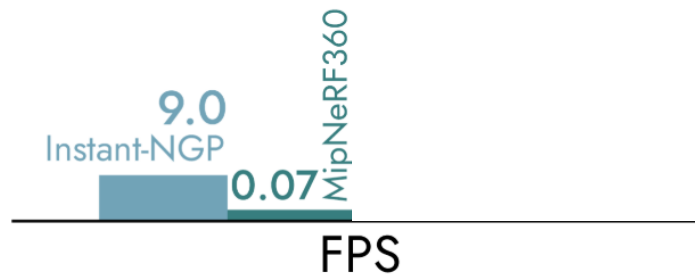
# 3D Gaussian Splatting

Represents radiance fields with points!

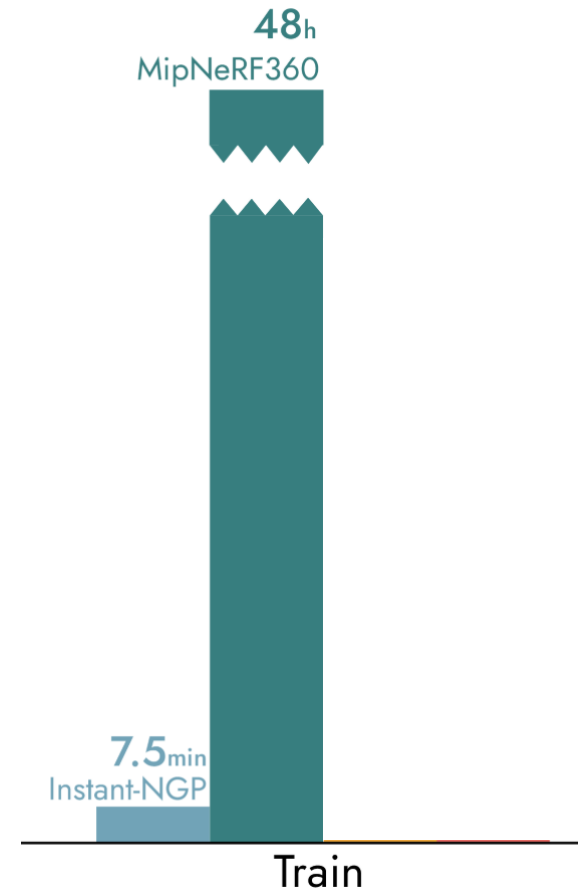


# Rendering/Training Speed

Rendering Speed



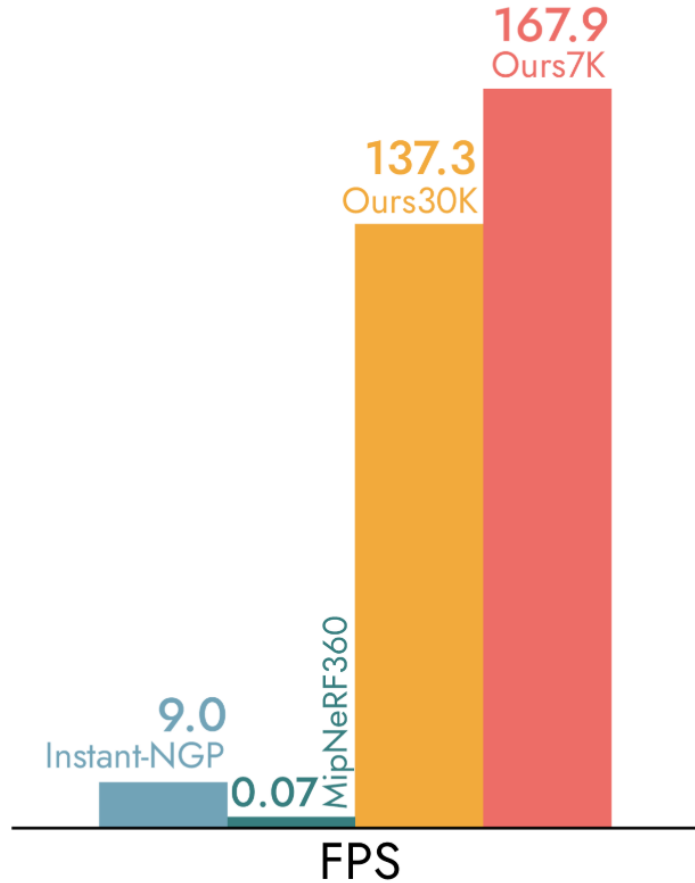
Training Speed



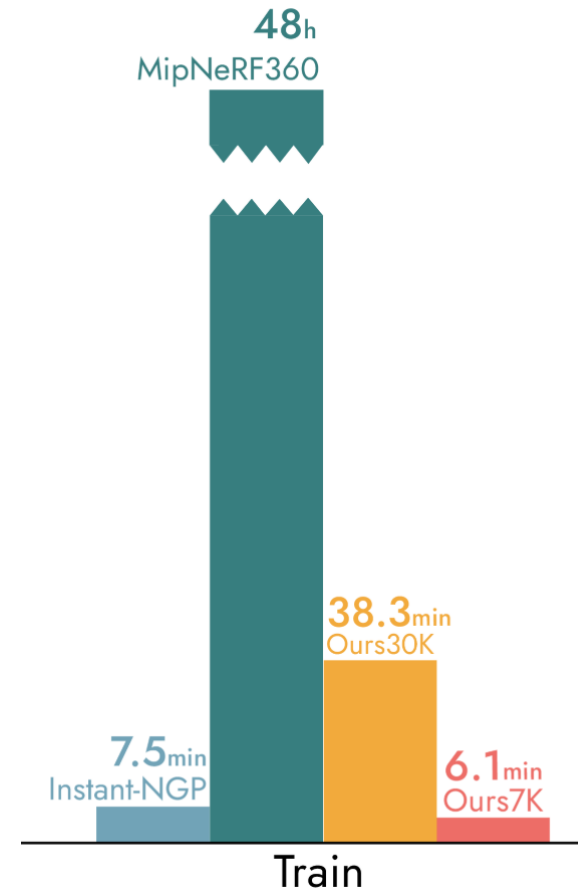
<https://repo-sam.inria.fr/fungraph/3d-gaussian-splatting/>

# Rendering/Training Speed

Rendering Speed



Training Speed



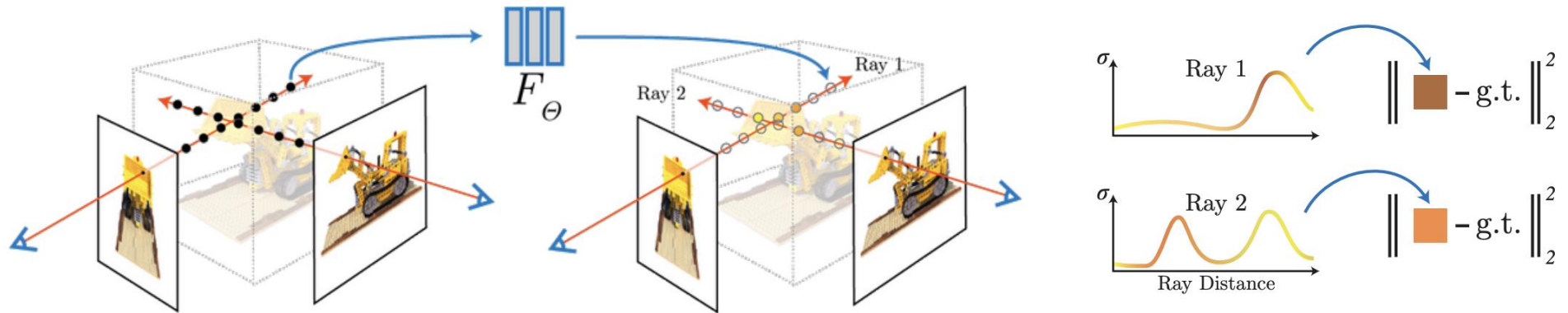
<https://repo-sam.inria.fr/fungraph/3d-gaussian-splatting/>



# Neural Radiance Field

Parameterize Radiance Field **densely**, at **every point** in space

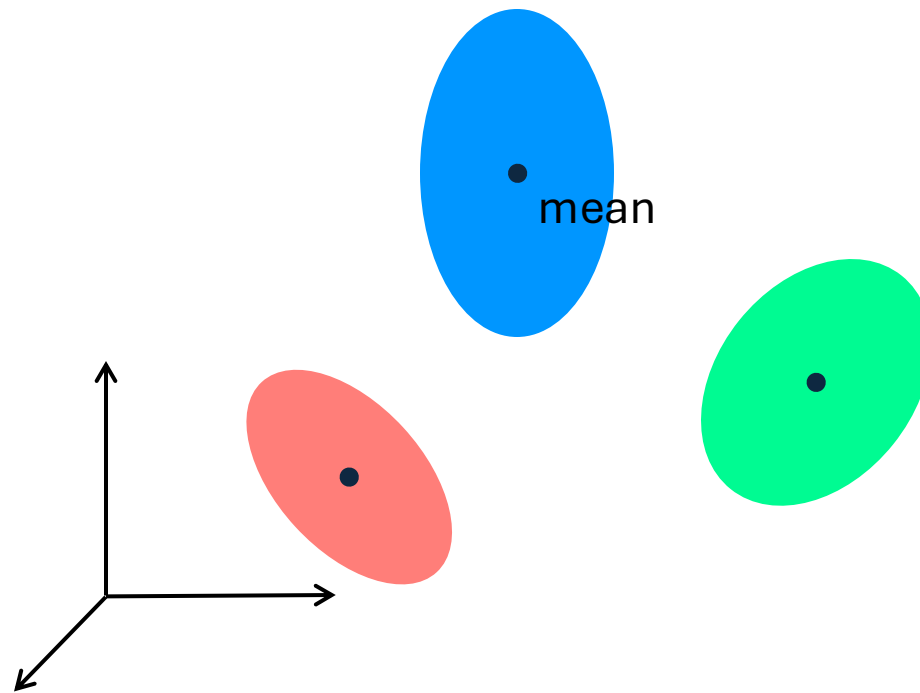
$$(x, y, z, \theta, \phi) \rightarrow \begin{matrix} \text{[Neural Network]} \\ F_{\Theta} \end{matrix} \rightarrow (RGB\sigma)$$





# 3D Gaussians

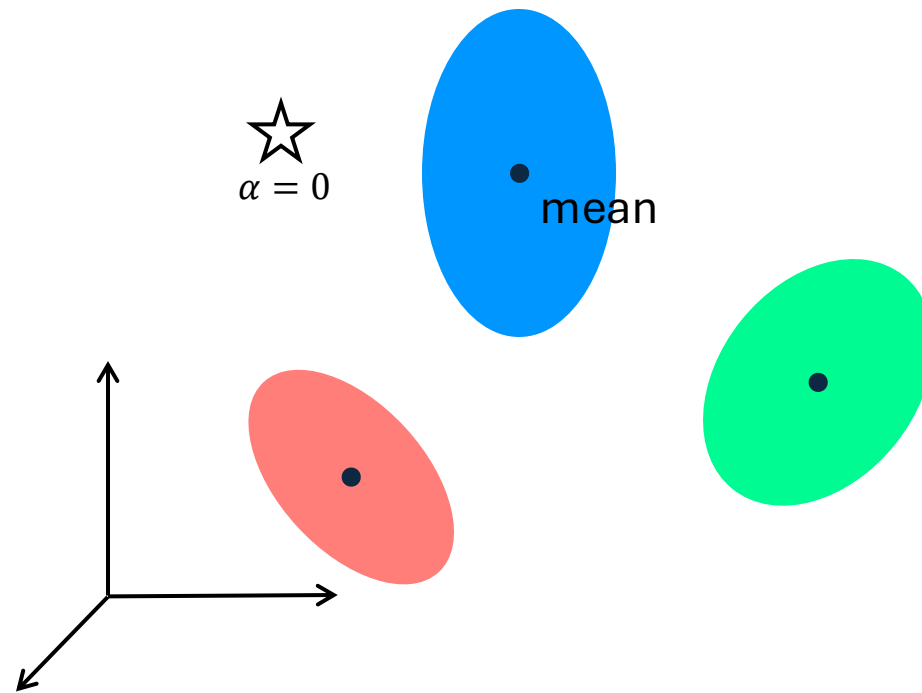
Key Idea: Parameterize Radiance Field **sparsely**, only where **density**  $\neq 0$



Reference: 6.S980 – ML for Inverse Graphics – Vincent Sitzmann

# 3D Gaussians

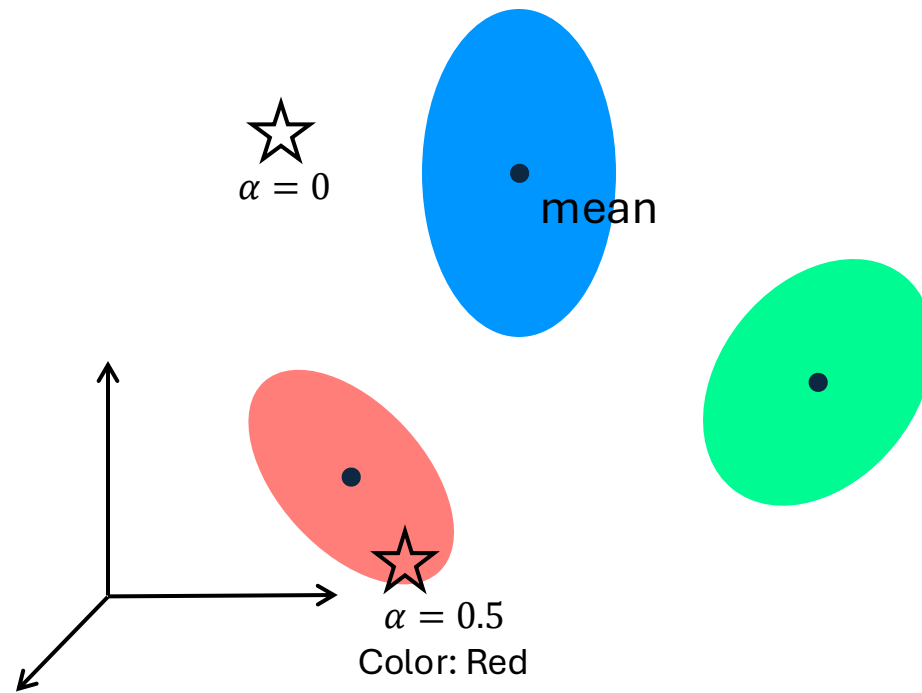
Key Idea: Parameterize Radiance Field **sparsely**, only where **density**  $\neq 0$



Reference: 6.S980 – ML for Inverse Graphics – Vincent Sitzmann

# 3D Gaussians

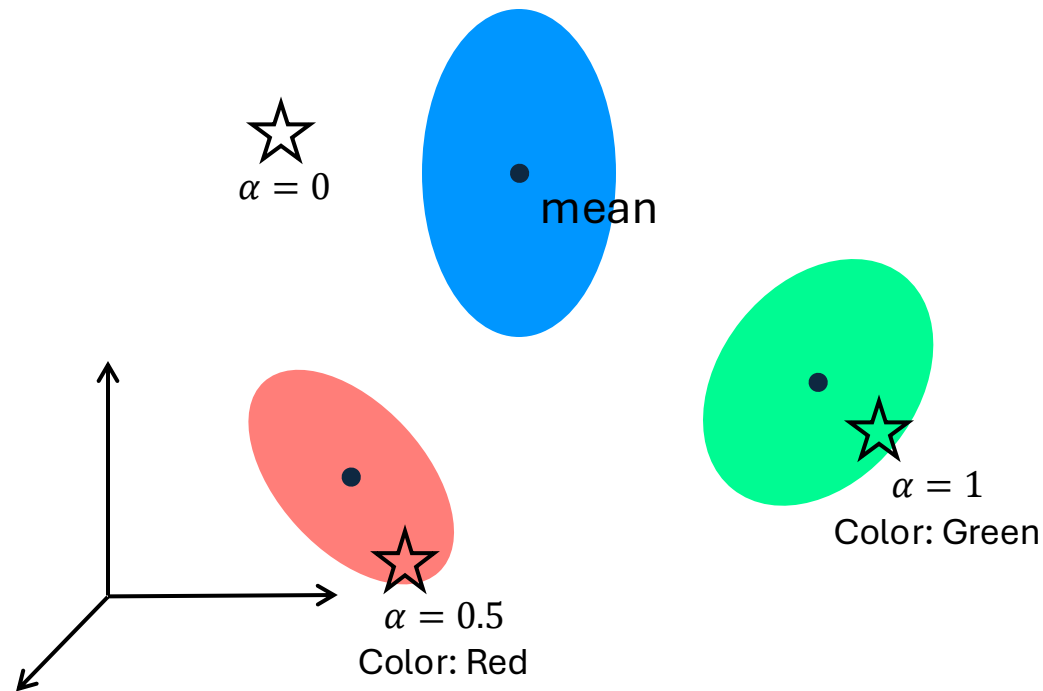
Key Idea: Parameterize Radiance Field **sparsely**, only where **density**  $\neq 0$



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# 3D Gaussians

Key Idea: Parameterize Radiance Field **sparsely**, only where **density**  $\neq 0$

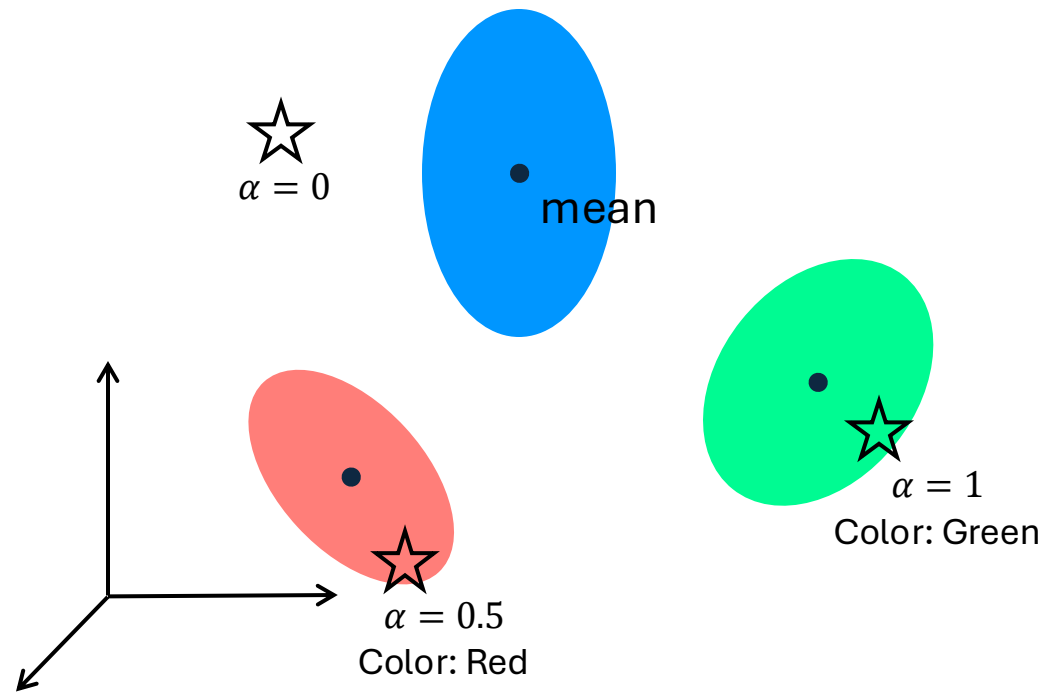


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# 3D Gaussians

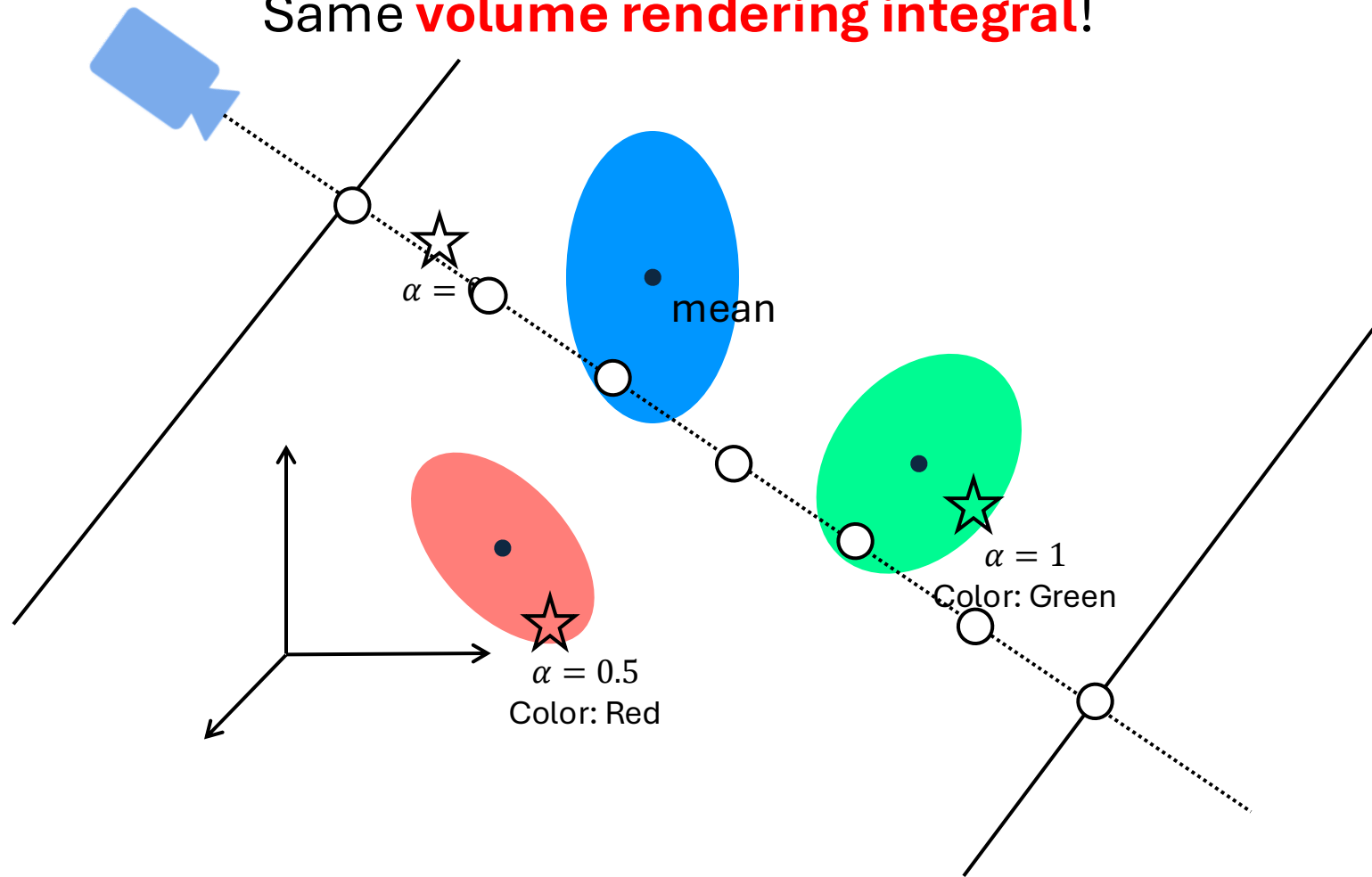
Key Idea: Parameterize Radiance Field **sparsely**, only where **density**  $\neq 0$



...How to render?

# 3D Gaussians -- Rendering

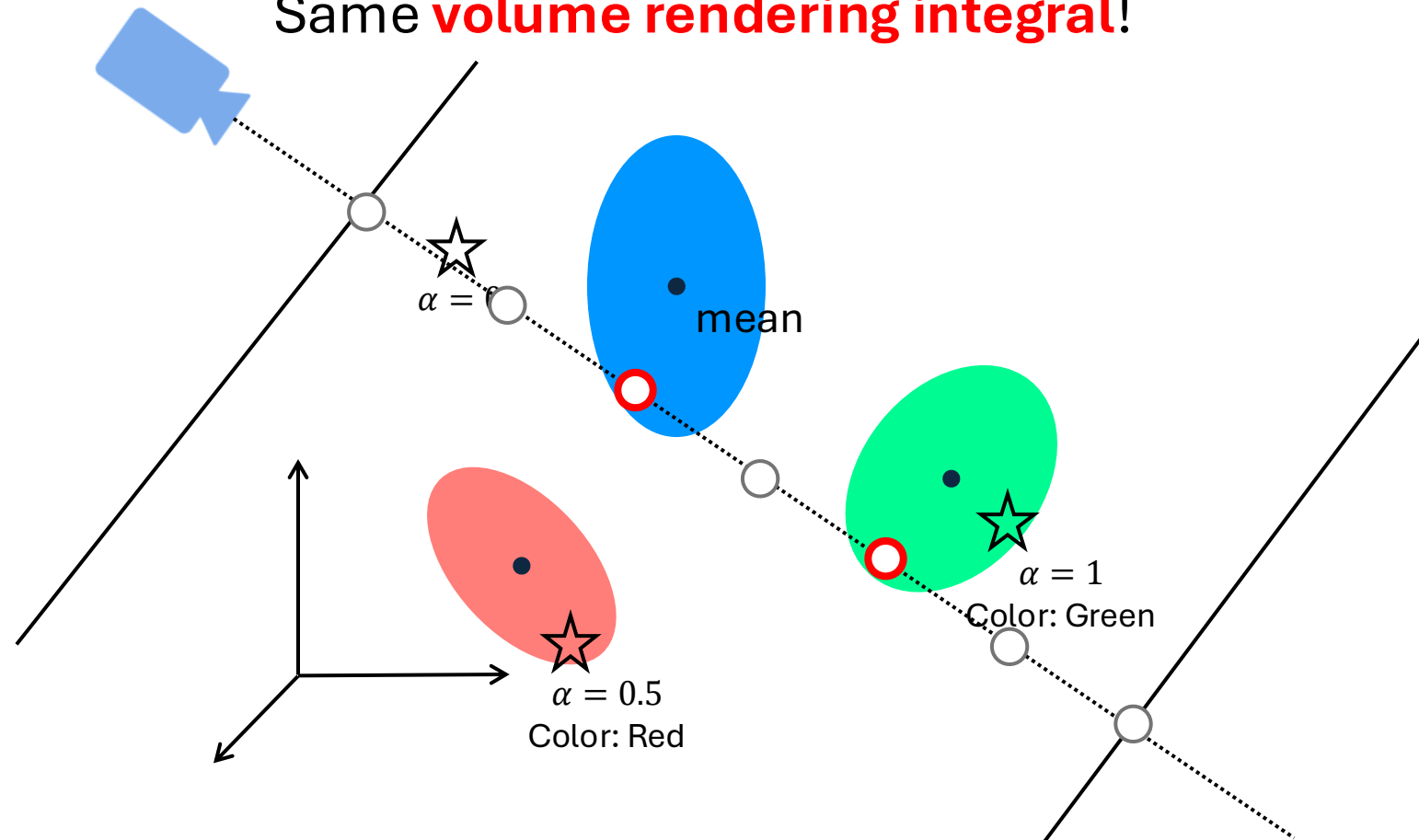
Same **volume rendering integral!**



Reference: 6.S980 – ML for Inverse Graphics – Vincent Sitzmann

# 3D Gaussians -- Rendering

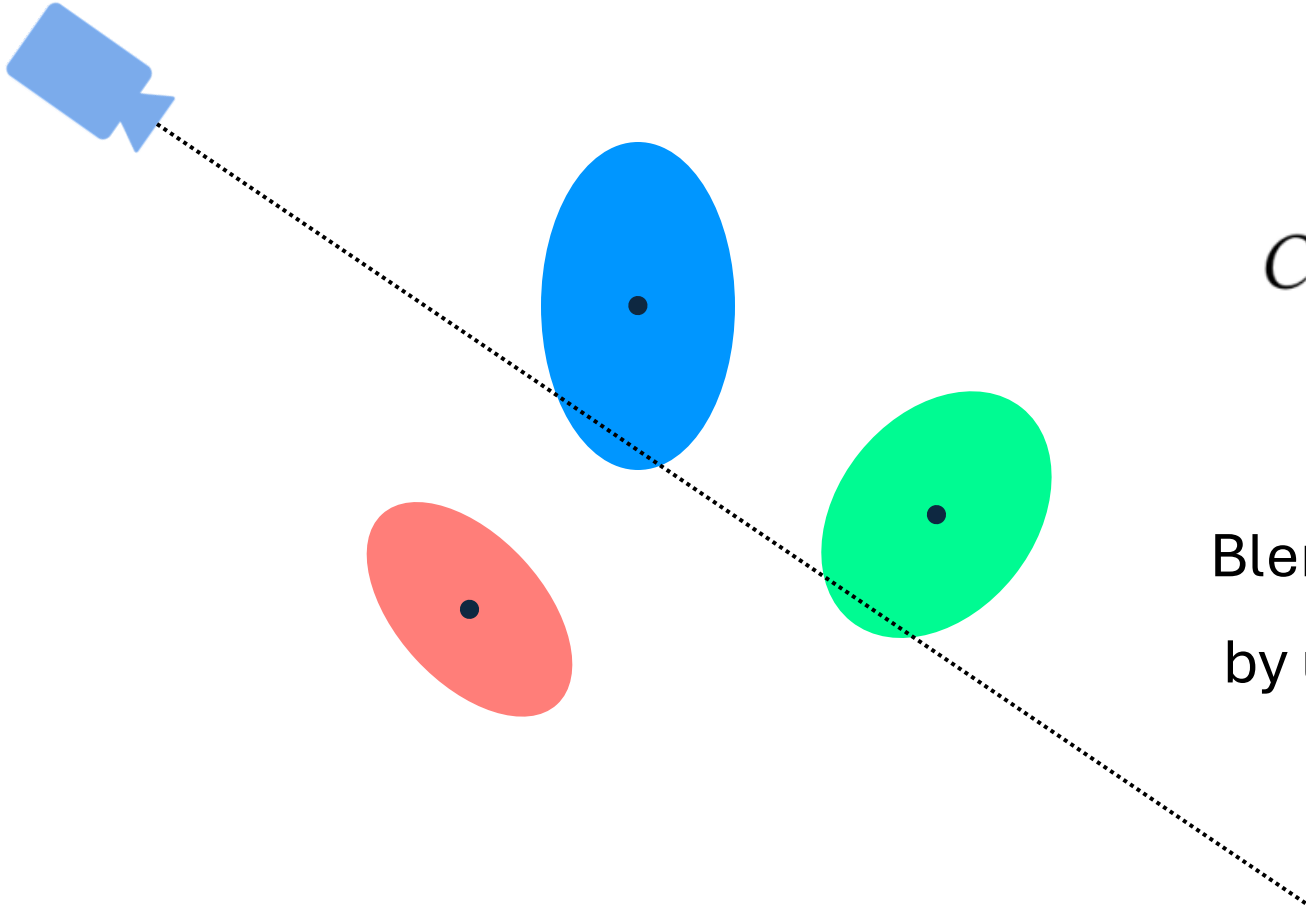
Same **volume rendering integral!**



But we can compute only on the points within the Gaussians!

Reference: 6.S980 – ML for Inverse Graphics – Vincent Sitzmann

# 3D Gaussians



$$C = \sum_{i \in \mathcal{N}} c_i \alpha_i \prod_{j=1}^{i-1} (1 - \alpha_j),$$

Blend together the 3D Gaussian points  
by using the density of each points



# Rendering Functions – NeRF vs 3DGS

- Discreet volumetric rendering equation for NeRF

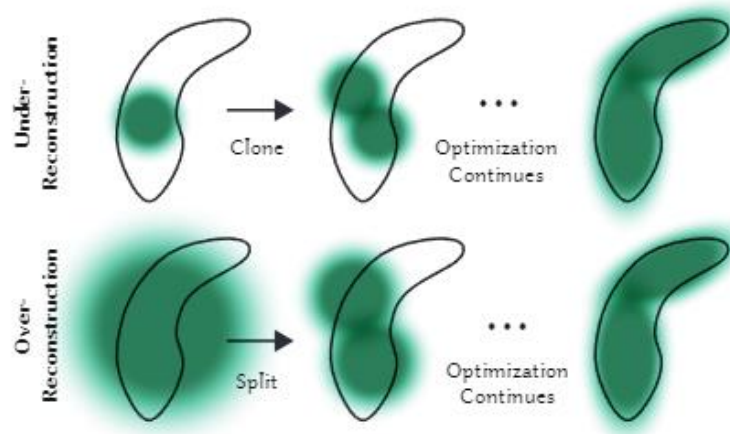
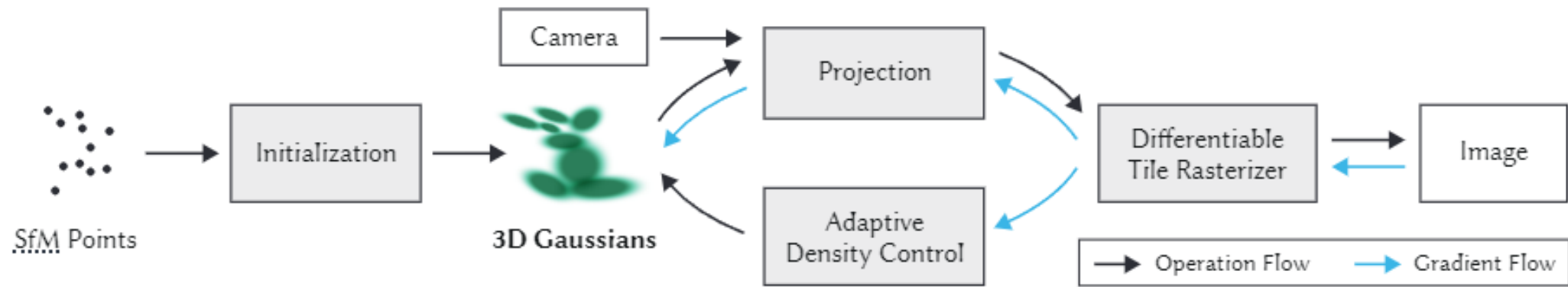
$$C = \sum_{i=1}^N T_i (1 - \exp(-\sigma_i \delta_i)) c_i \quad \text{with} \quad T_i = \exp\left(-\sum_{j=1}^{i-1} \sigma_j \delta_j\right)$$

$$C = \sum_{i=1}^N T_i \alpha_i c_i, \quad \alpha_i = (1 - \exp(-\sigma_i \delta_i)) \quad \text{and} \quad T_i = \prod_{j=1}^{i-1} (1 - \alpha_j).$$

- Neural point-based rendering for blending N order points overlapping the pixels

$$C = \sum_{i \in N} c_i \alpha_i \prod_{j=1}^{i-1} (1 - \alpha_j),$$

# 3DGS Training / Optimization



Split / Combine Gaussians until the Gaussian fits to the scene details!

# Summary: NeRF or 3DGS

	NeRF	3D Gaussian Splatting
Representation	Implicit Function	3D Gaussian Points
Computational Cost	Heavy volumetric rendering	<b>Fast rasterization rendering</b>
Compatibility with 3D tools	Limited	<b>Relatively Easy</b>
Interactbility	Complicated geometry-based editing	<b>Relatively easy to add objects</b>
When insufficient input views?	Hallucinate details	Could not render details - Missing Gaussians
Pose Sensitivity	Require precise camera calibration	Require precise camera calibration

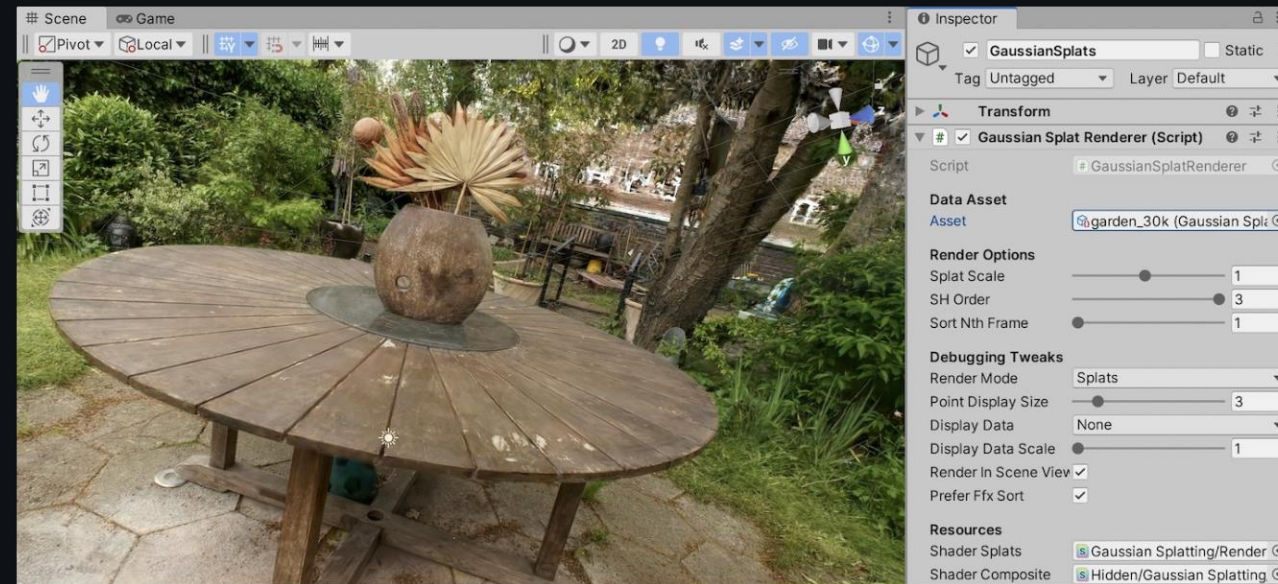


# UnityGaussianSplatting

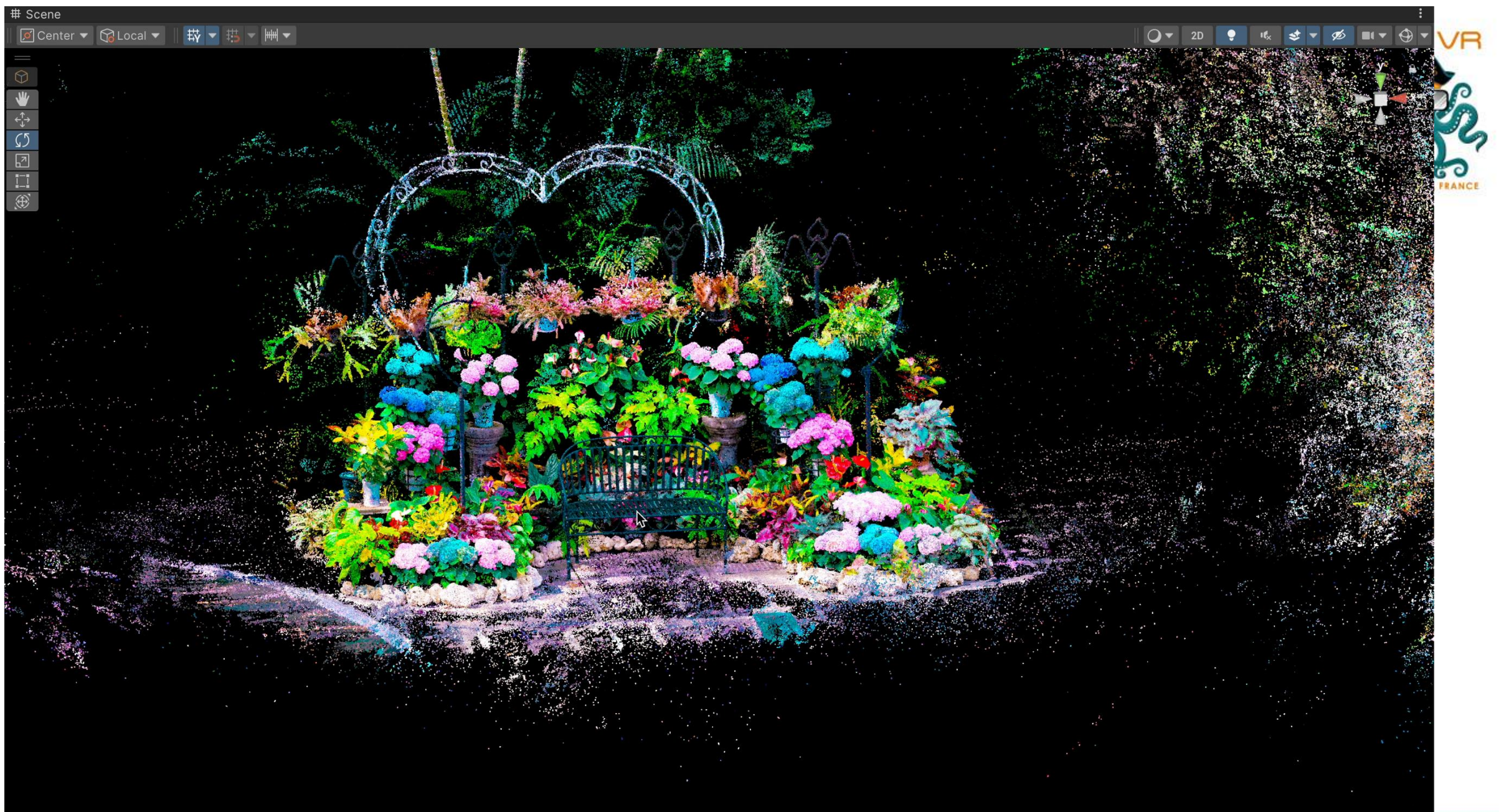
The open source tool to play 3D Gaussian splatting data in Unity!

## Gaussian Splatting playground in Unity

SIGGRAPH 2023 had a paper "[3D Gaussian Splatting for Real-Time Radiance Field Rendering](#)" by Kerbl, Kopanas, Leimkühler, Drettakis that is really cool! Check out their website, source code repository, data sets and so on. I've decided to try to implement the realtime visualization part (i.e. the one that takes already-produced gaussian splat "model" file) in Unity.





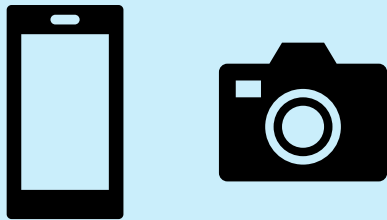




# Just 3 steps to play in Unity!

## Step1

Recording



## Step2

Train 3DGS



<https://aras-p.info/blog/2023/09/27/Making-Gaussian-Splats-more-smaller/>

## Step3

Import to Unity



+

UnityGaussianSplatting



# Create 3D Gaussians?

There are lot of ways to create 3DGS scenes!



... and also original 3DGS code!

# How to use?



# Live Demo!