

NeRF & Plenoxels

MEGVII 旷视

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2022年5月

保密信息

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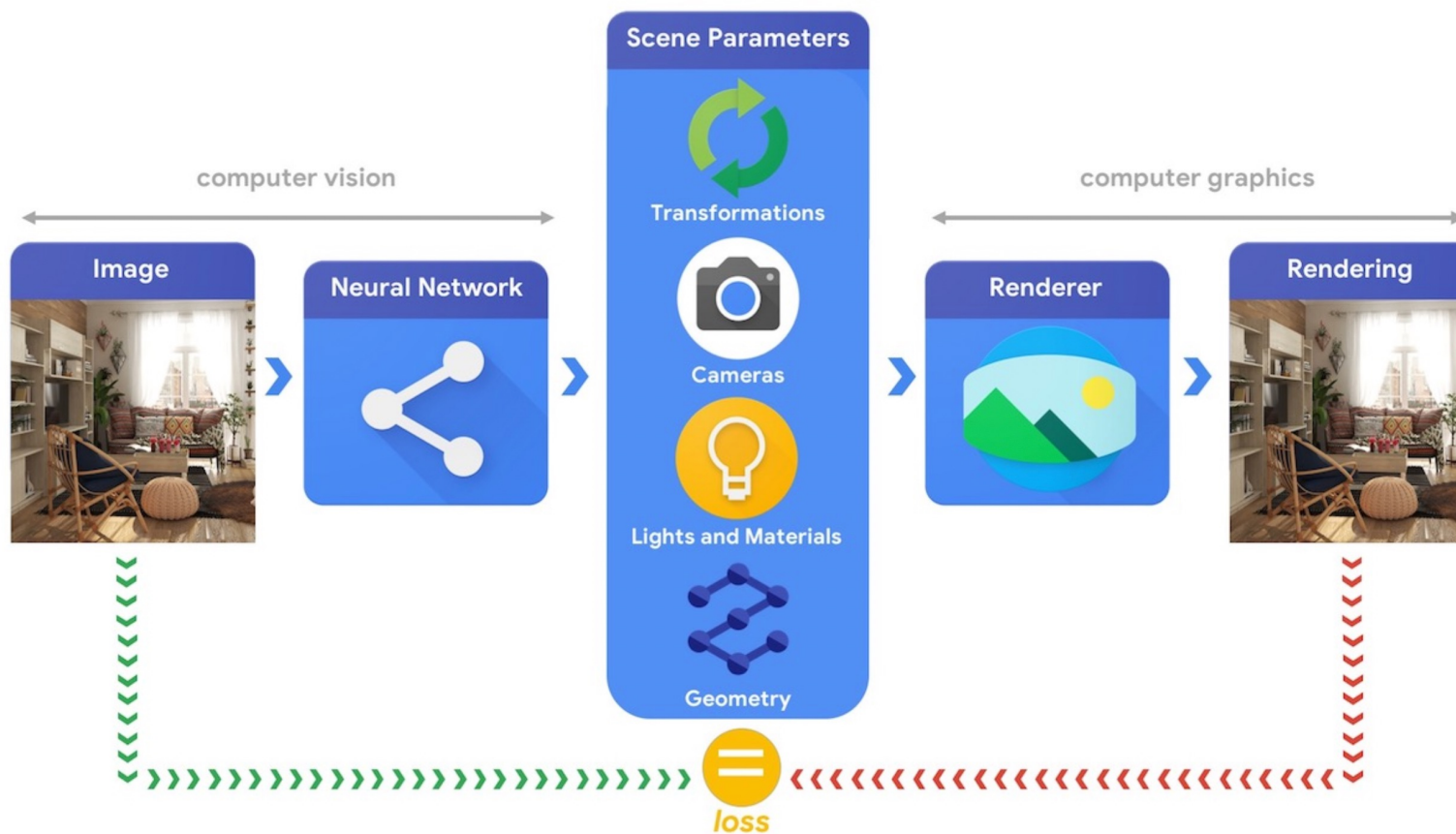
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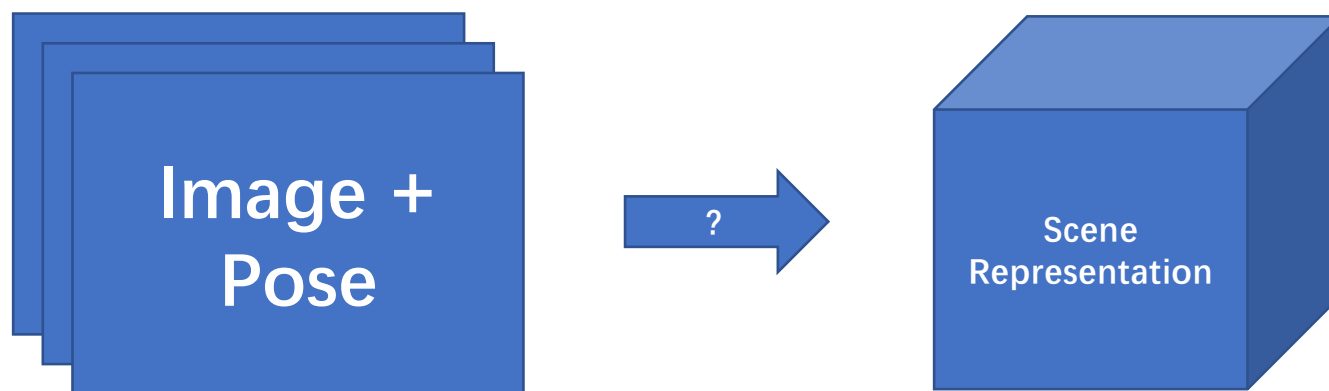
Overview

”

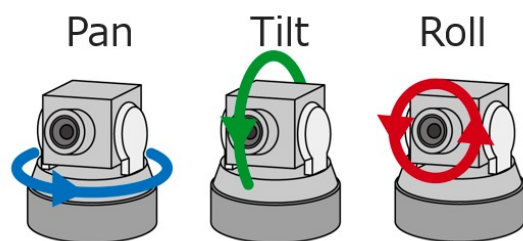
Computer Graphics meets Computer Vision



| Differentiable Rendering



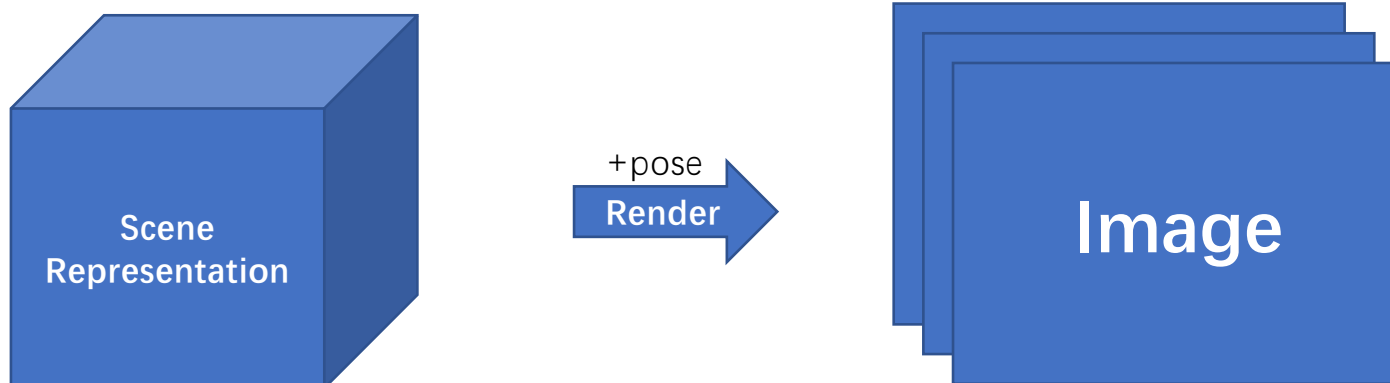
6 DoF Camera Pose



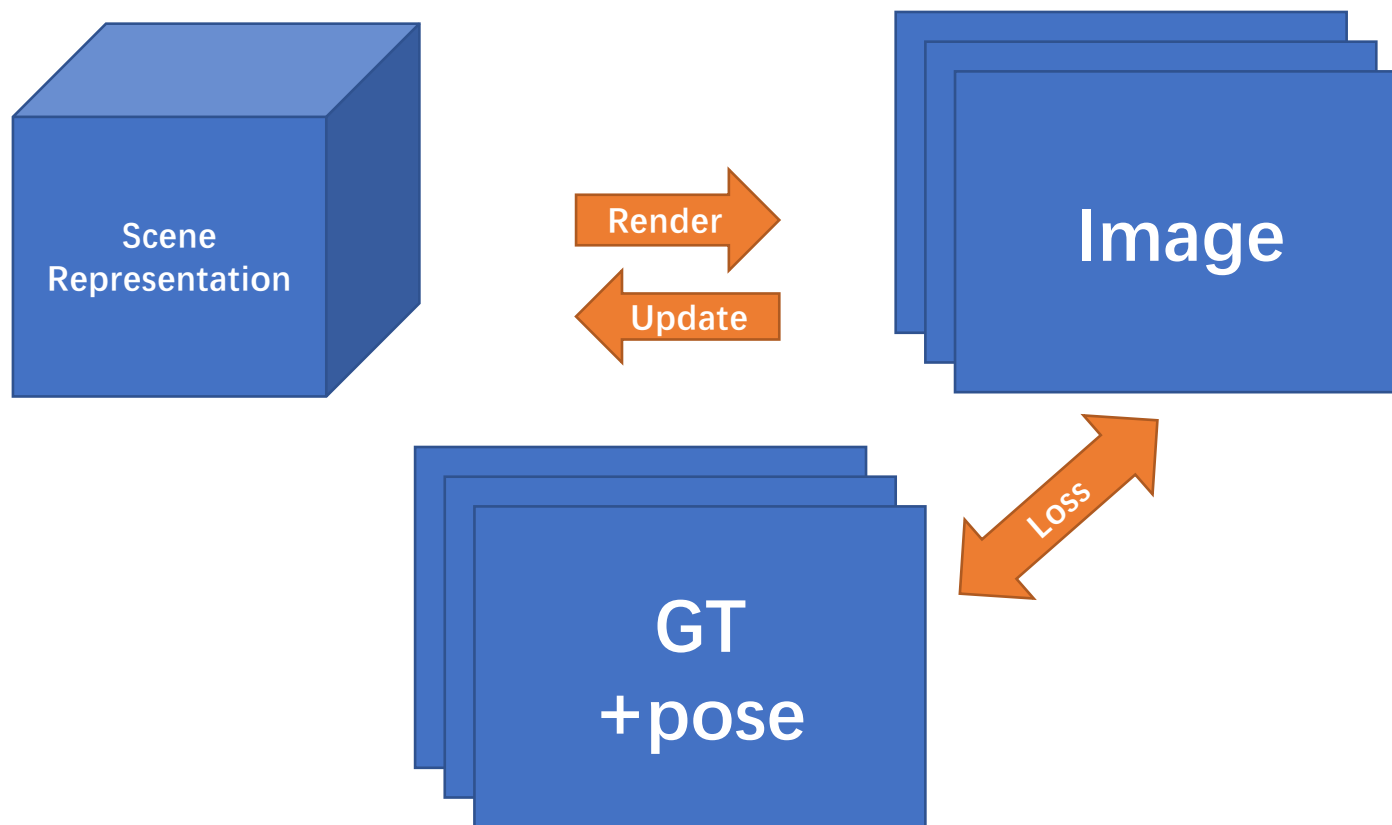
+ (x, y, z)



| Differentiable Rendering

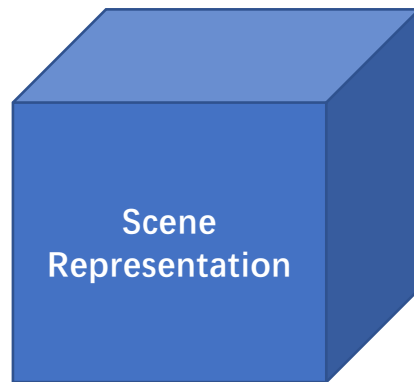


| Differentiable Rendering

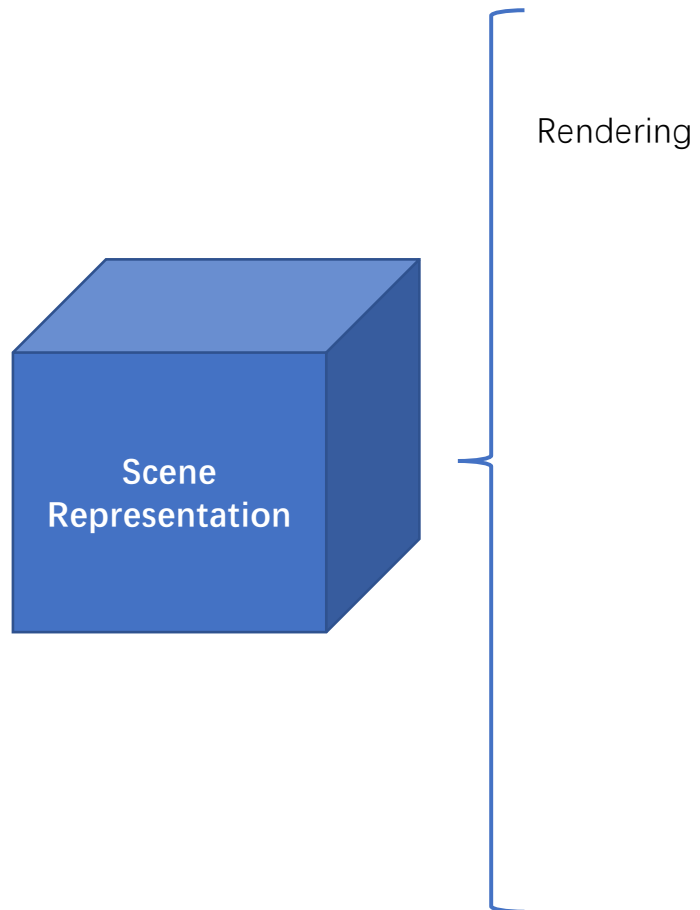


Applications

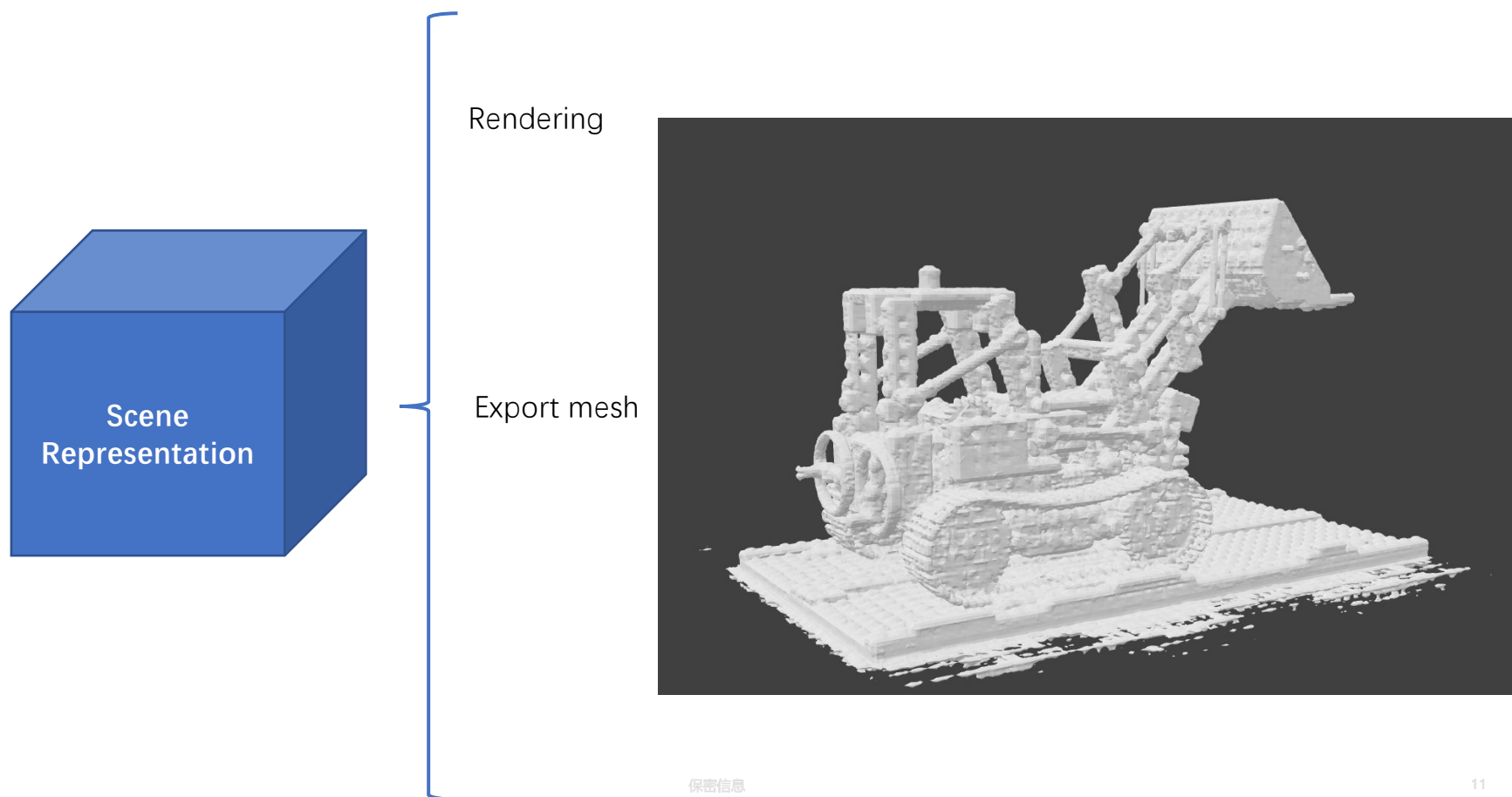
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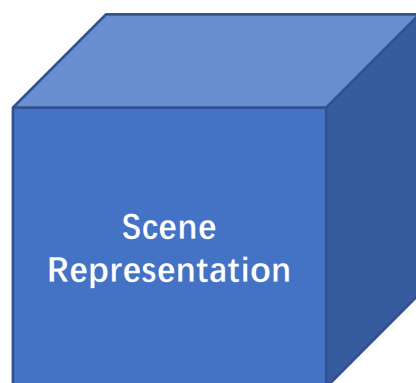
| Applications



Applications



Applications

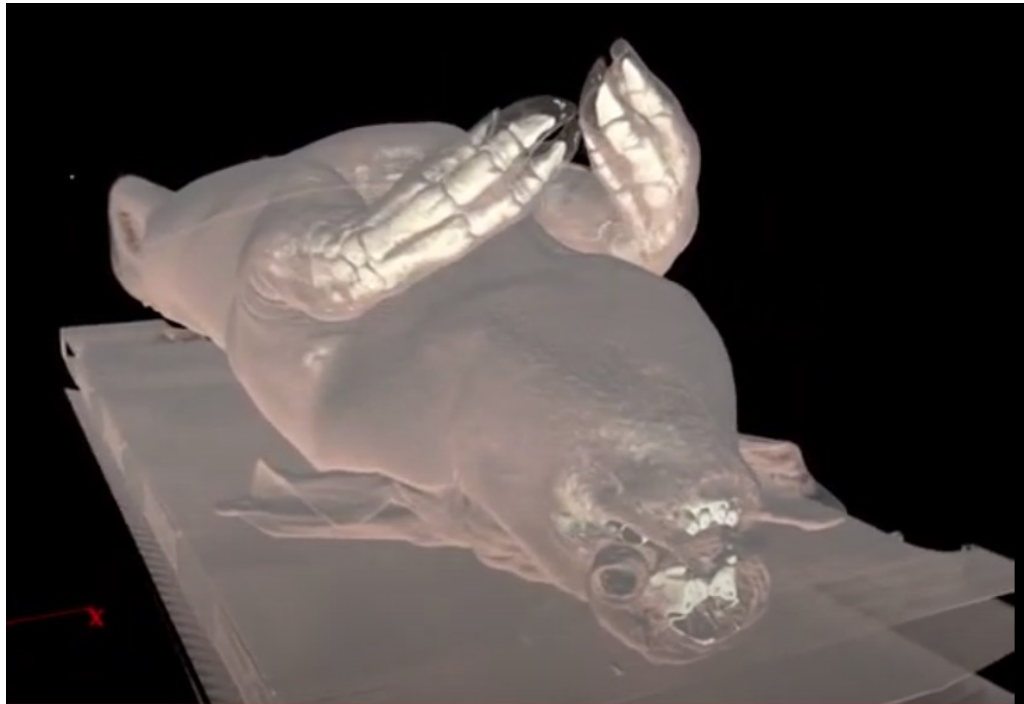


Computer Vision tasks (e.g. Depth Estimation, iNeRF)

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“ Volume Rendering ”

| Volume Rendering

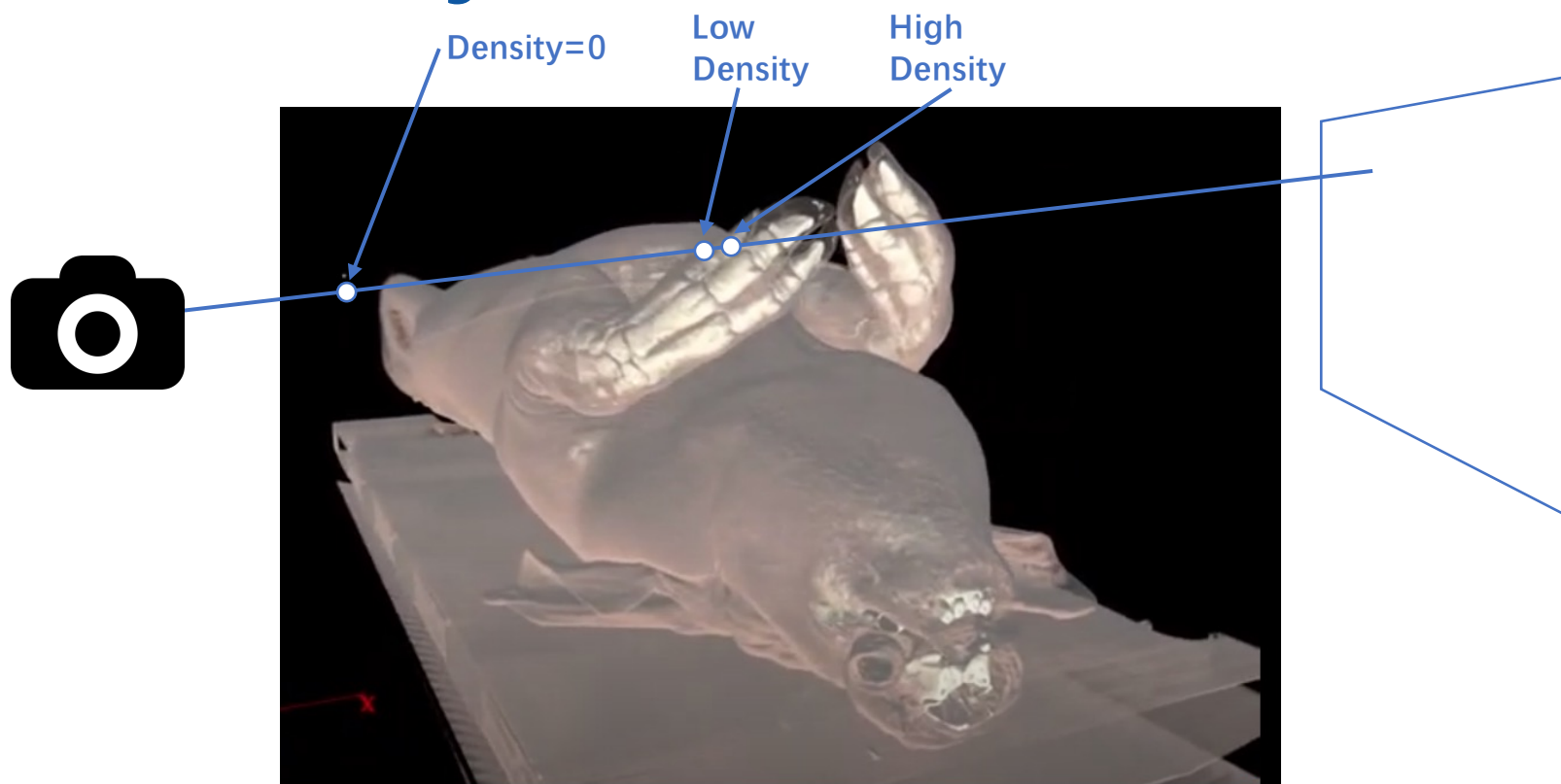


source: <https://www.youtube.com/watch?v=b0uH-hqUOrk>
CT Scan of a Pig, CC-BY 3.0

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| Volume Rendering

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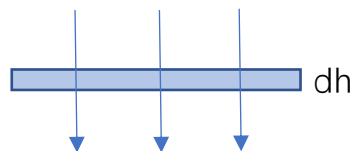
source: <https://www.youtube.com/watch?v=b0uH-hqUOrk>
CT Scan of a Pig, CC-BY 3.0

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| Volume Density



Opacity = 0.6

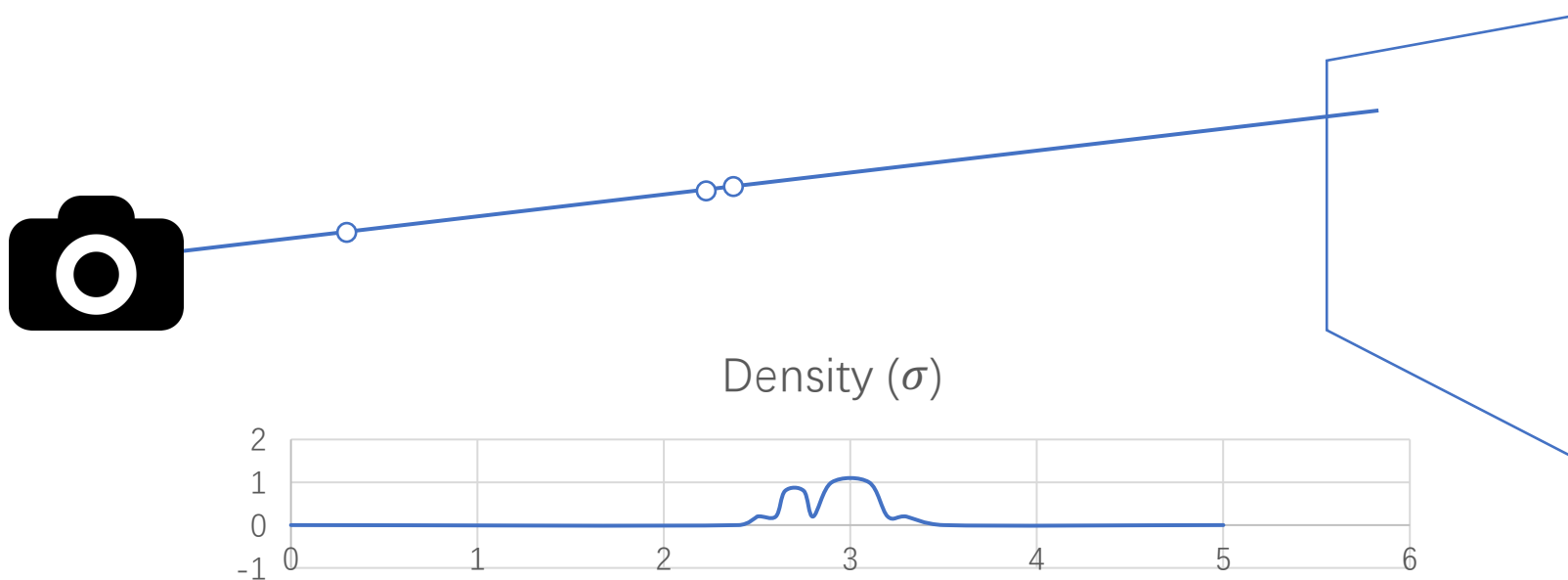


$$\sigma = \frac{d(\text{opacity})}{dh}$$

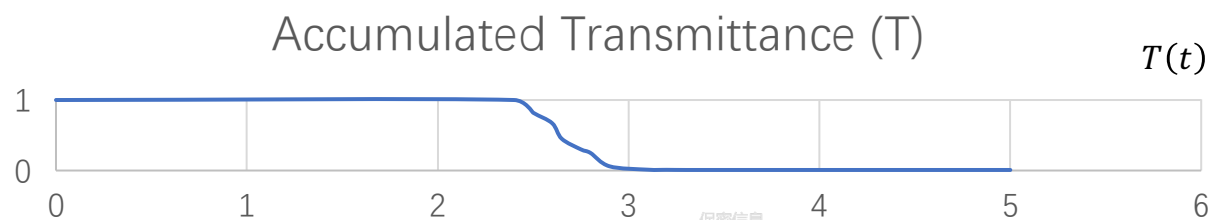
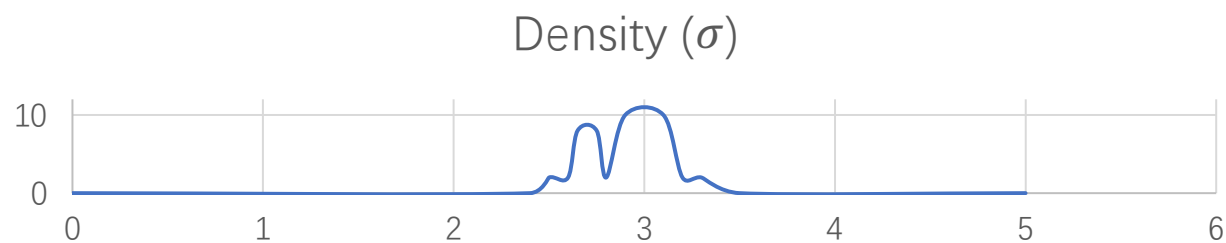
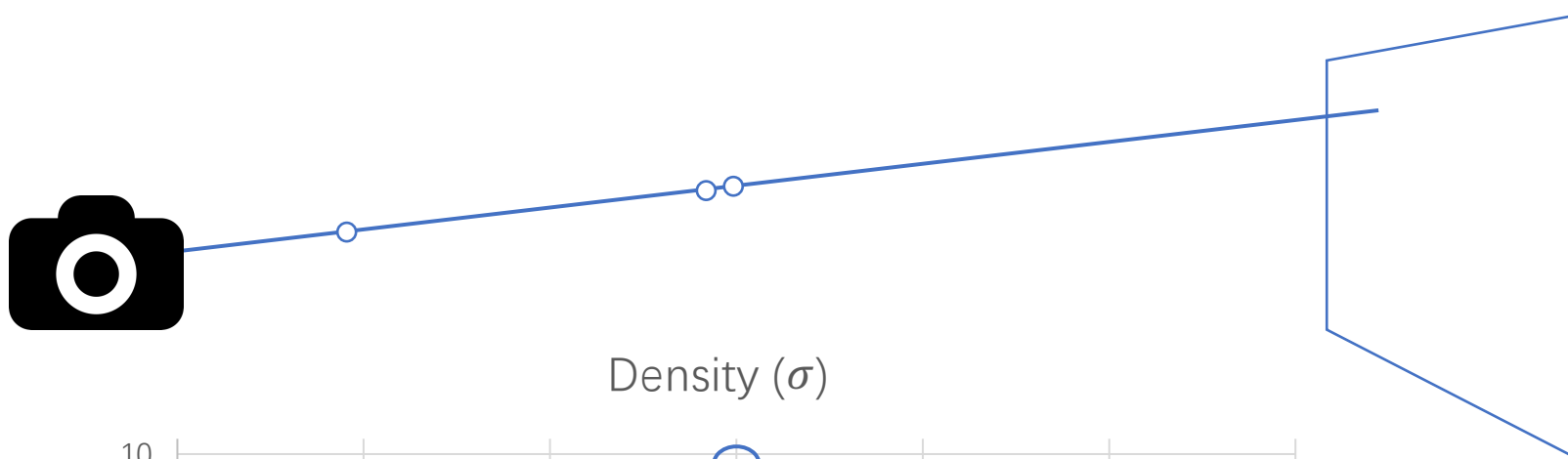
$$\log T = \int_0^{h_t} \log(1 - \text{opacity})$$

$$T = \exp\left(-\int_0^{h_t} \sigma(h)dh\right)$$

Volume Rendering

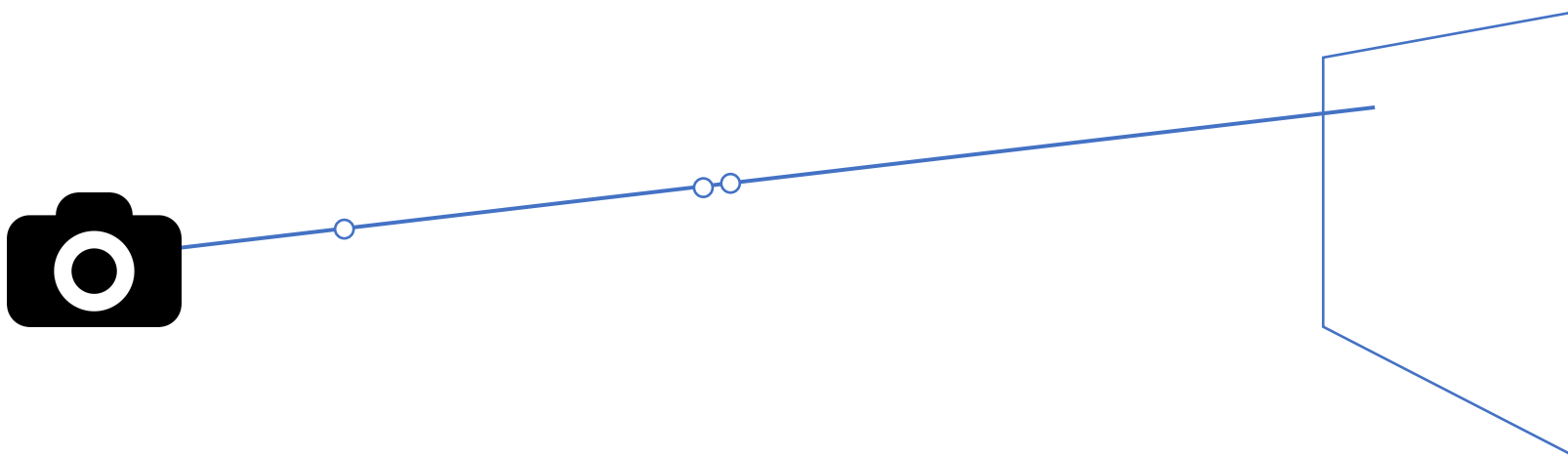


Volume Rendering



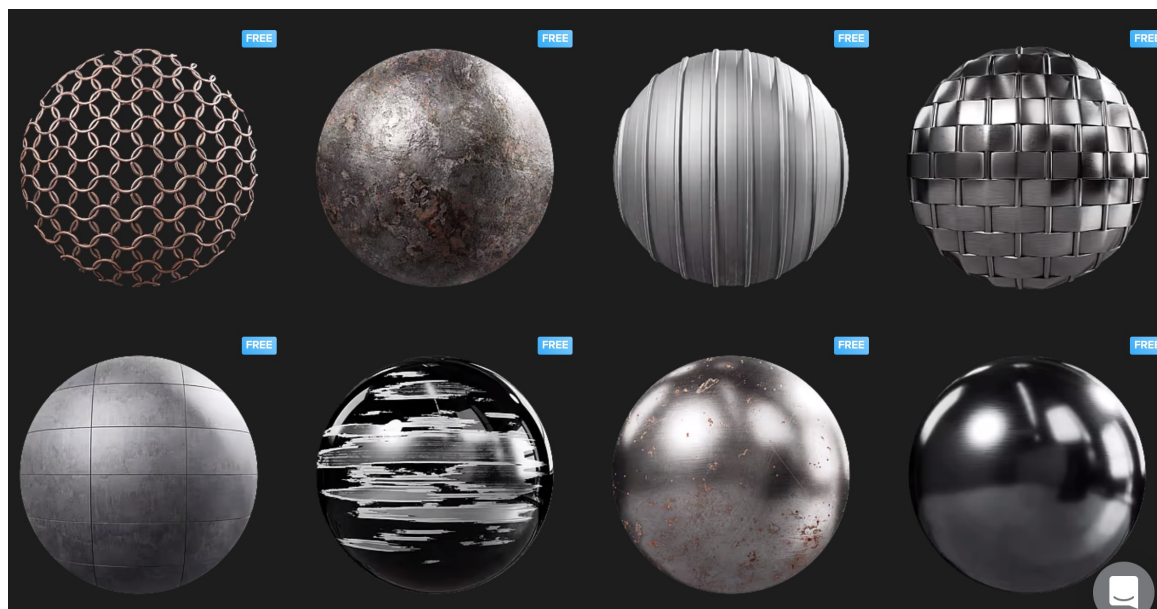
$$T(t) = \exp\left(-\int_{t_n}^t \sigma(\mathbf{r}(s))ds\right)$$

| Volume Rendering



$$C(\mathbf{r}) = \int_{t_n}^{t_f} T(t) \sigma(\mathbf{r}(t)) \mathbf{c}(\mathbf{r}(t)) dt, \text{ where } T(t) = \exp\left(-\int_{t_n}^t \sigma(\mathbf{r}(s)) ds\right)$$

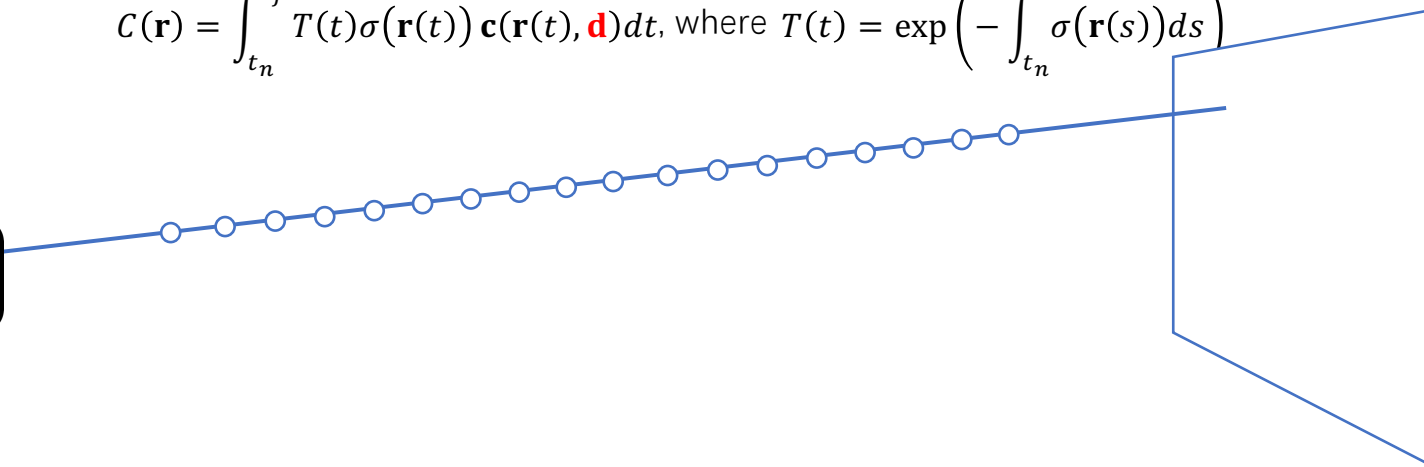
| Volume Rendering



$$C(\mathbf{r}) = \int_{t_n}^{t_f} T(t) \sigma(\mathbf{r}(t)) \mathbf{c}(\mathbf{r}(t), \mathbf{d}) dt, \text{ where } T(t) = \exp\left(-\int_{t_n}^t \sigma(\mathbf{r}(s)) ds\right)$$

Discretization

$$C(\mathbf{r}) = \int_{t_n}^{t_f} T(t) \sigma(\mathbf{r}(t)) \mathbf{c}(\mathbf{r}(t), \mathbf{d}) dt, \text{ where } T(t) = \exp\left(-\int_{t_n}^t \sigma(\mathbf{r}(s)) ds\right)$$

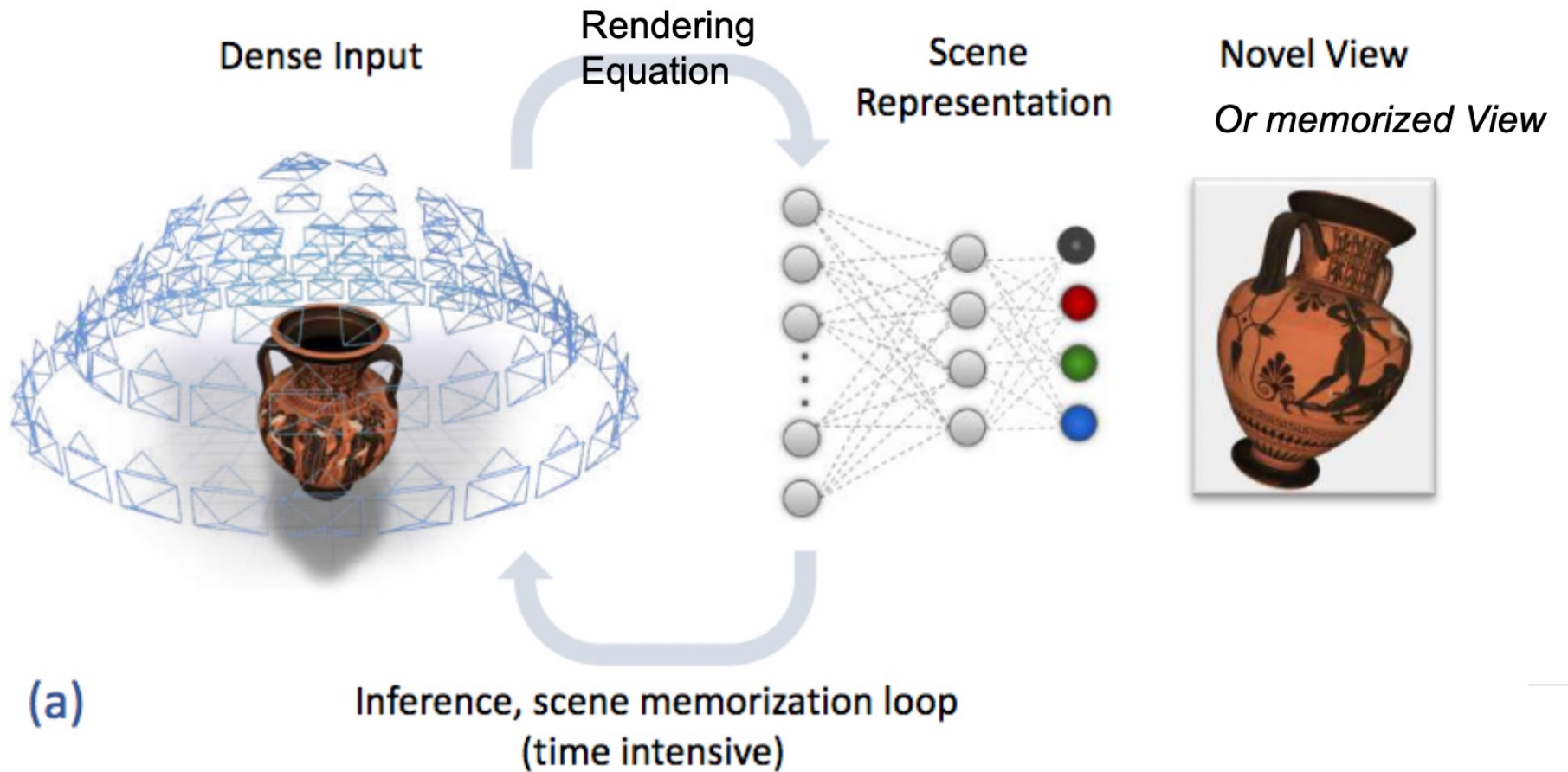


$$\hat{C}(\mathbf{r}) = \sum_{i=1}^N T_i (1 - \exp(-\sigma_i \delta_i)) \mathbf{c}_i, \text{ where } T_i = \exp\left(-\sum_{j=1}^{i-1} \sigma_j \delta_j\right)$$

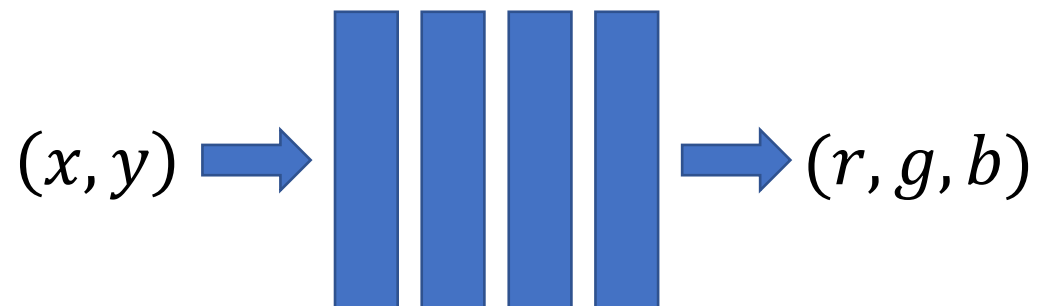
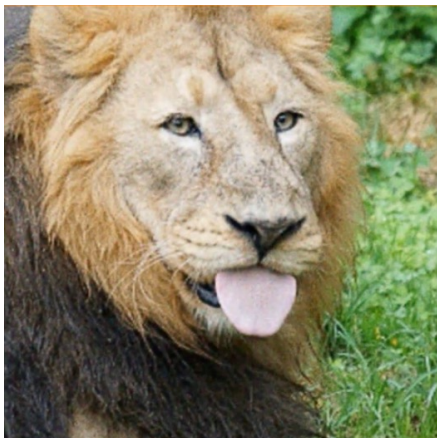
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“Neural Radiance Field”

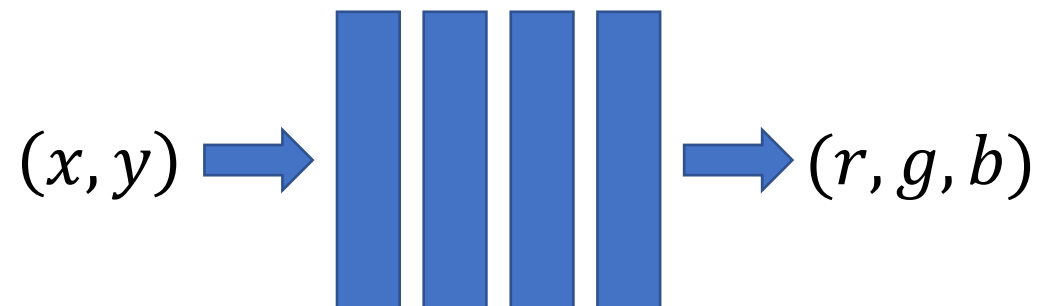
Neural Radiance Field



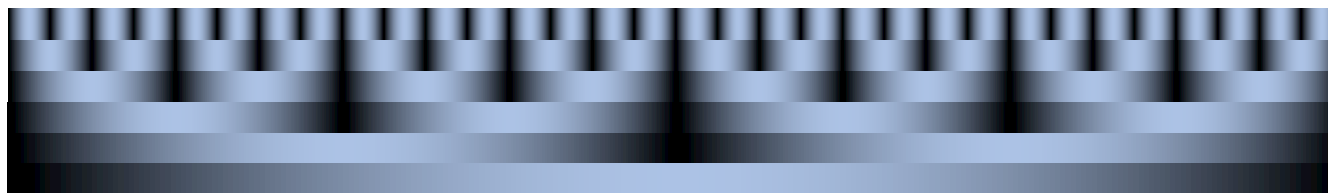
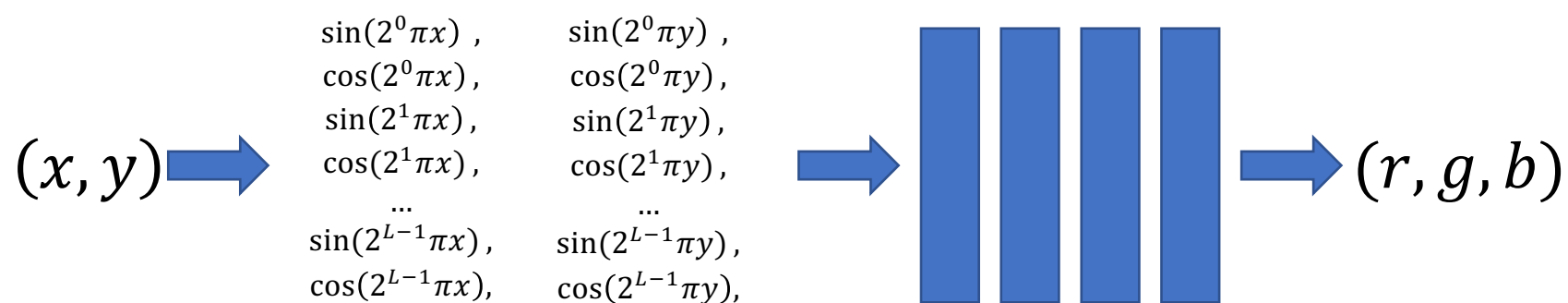
Coordinate MLP (2D)



Coordinate MLP (2D)



| Positional Encoding



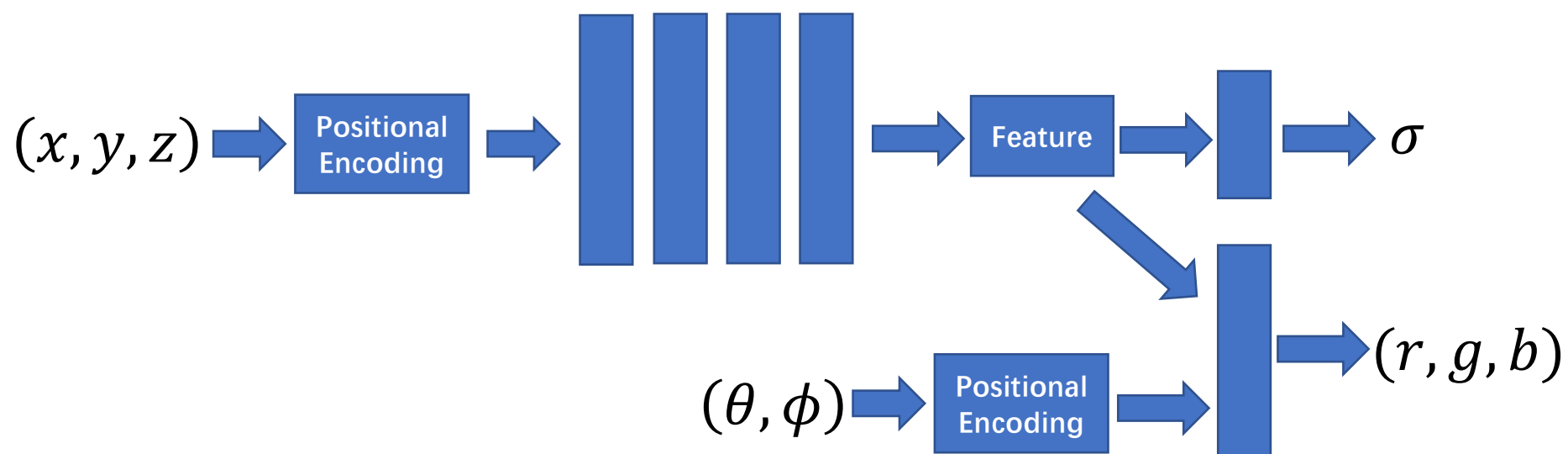
| 3D Coordinate MLP



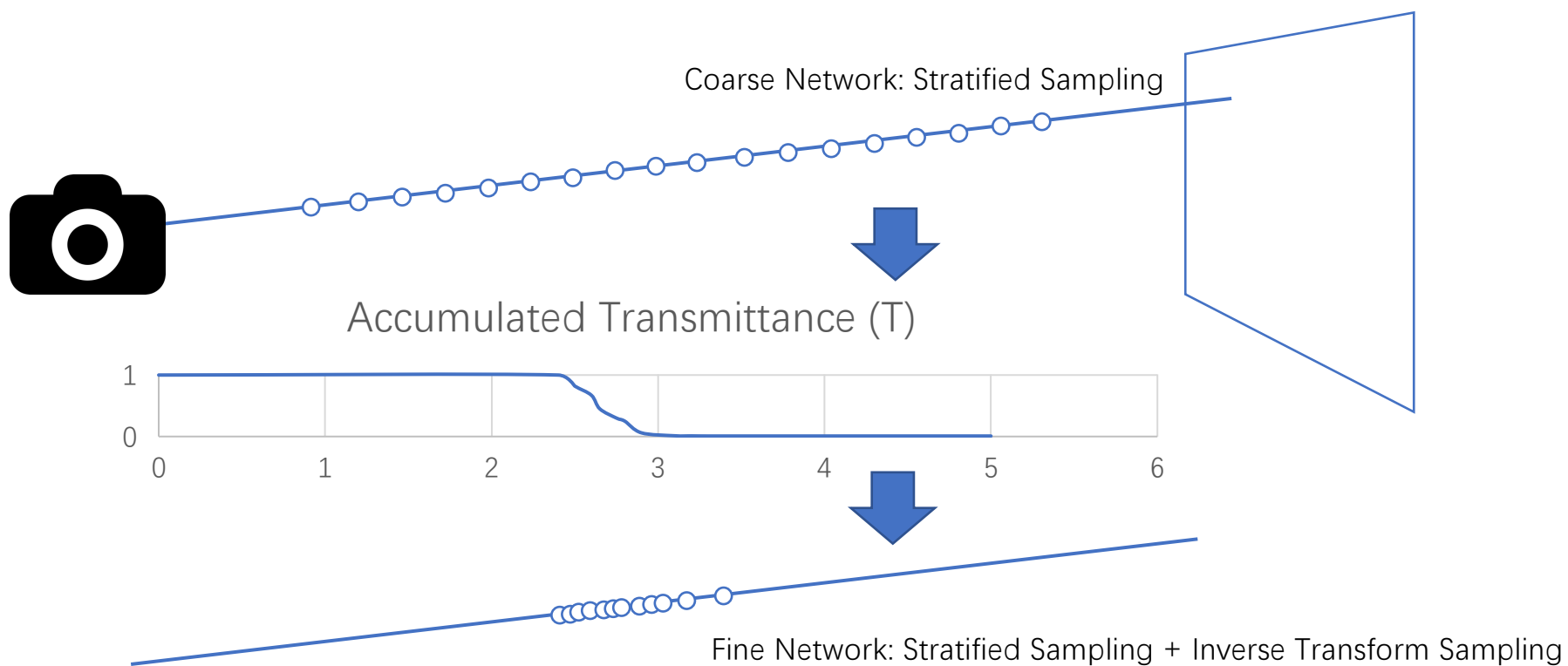
| 3D Coordinate MLP



| The Model



Coarse to Fine



| More Examples

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Plenoxels

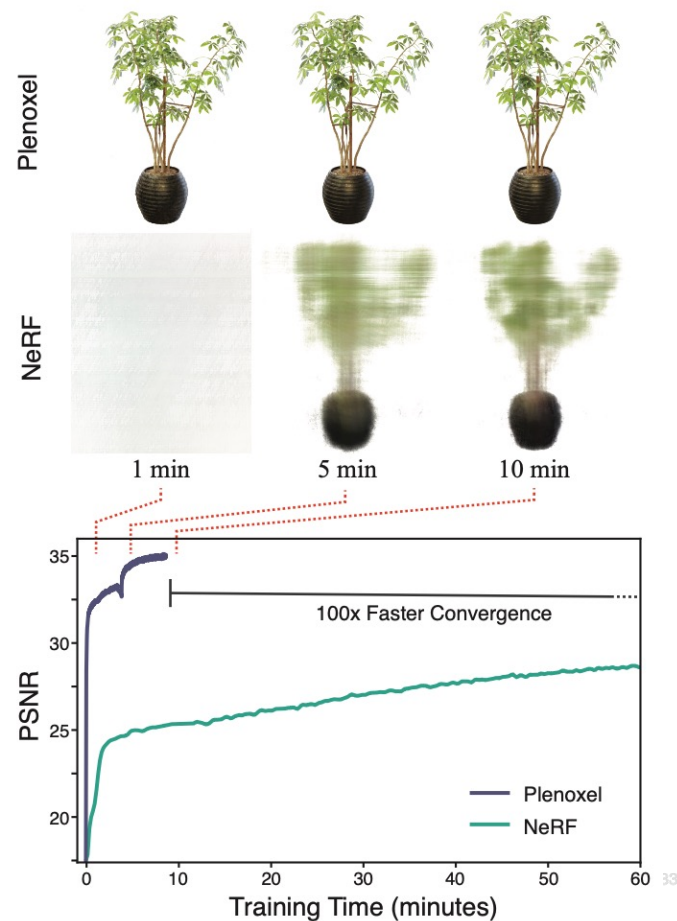
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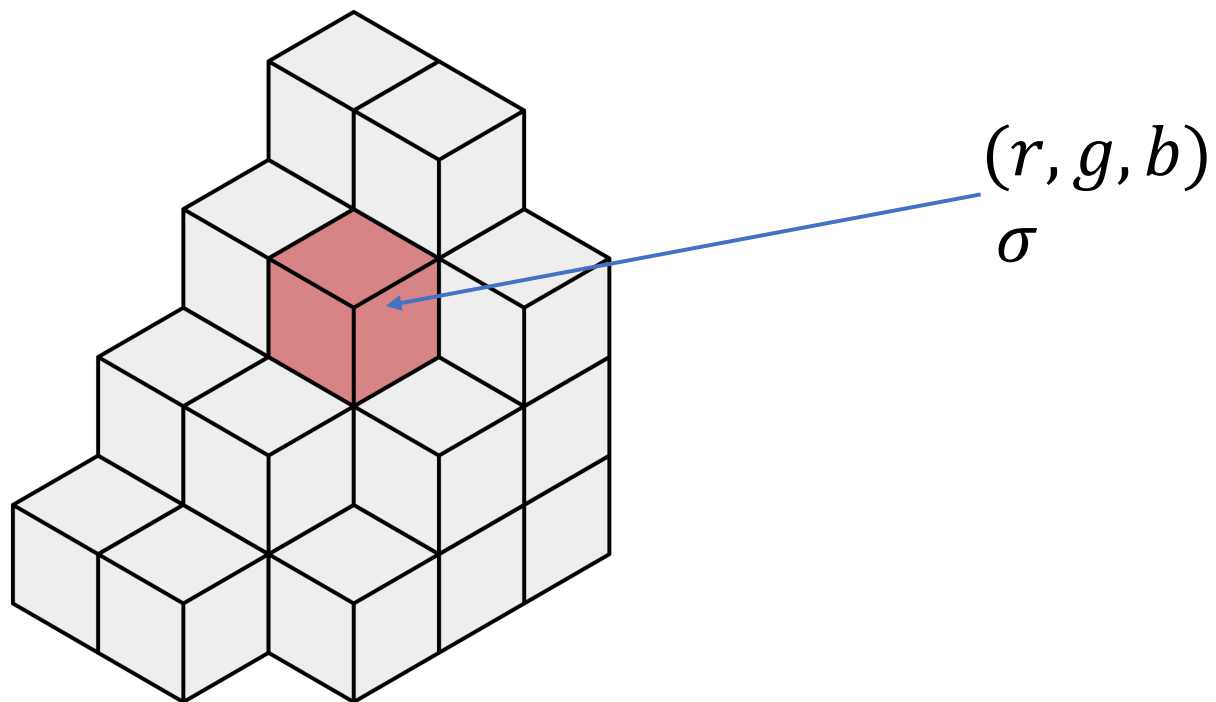
Plenoxels

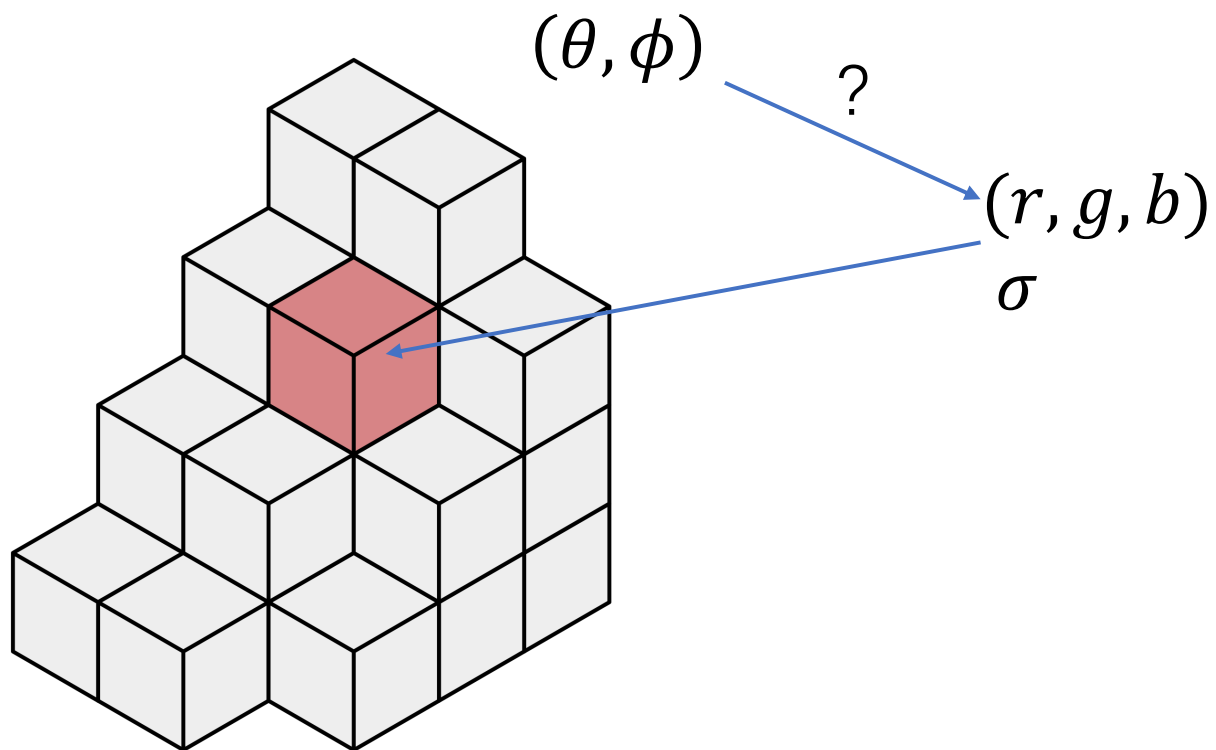
Faster training & inference

No neural networks

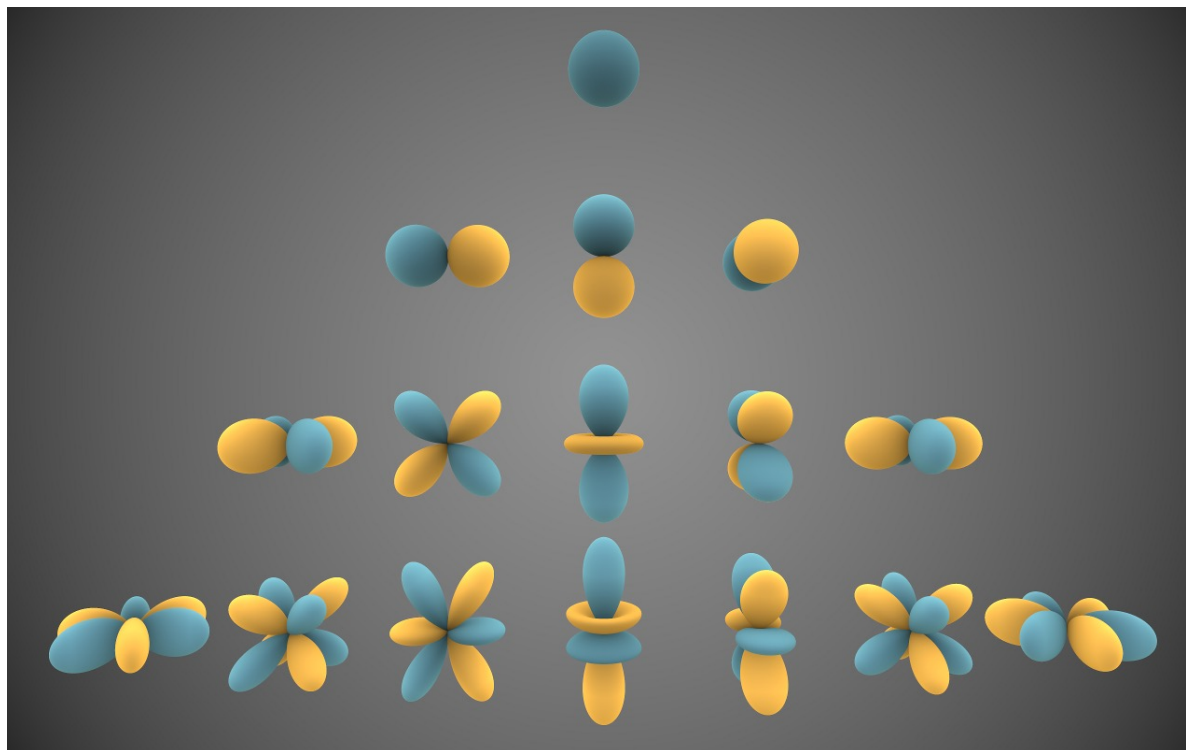
Custom CUDA implementation





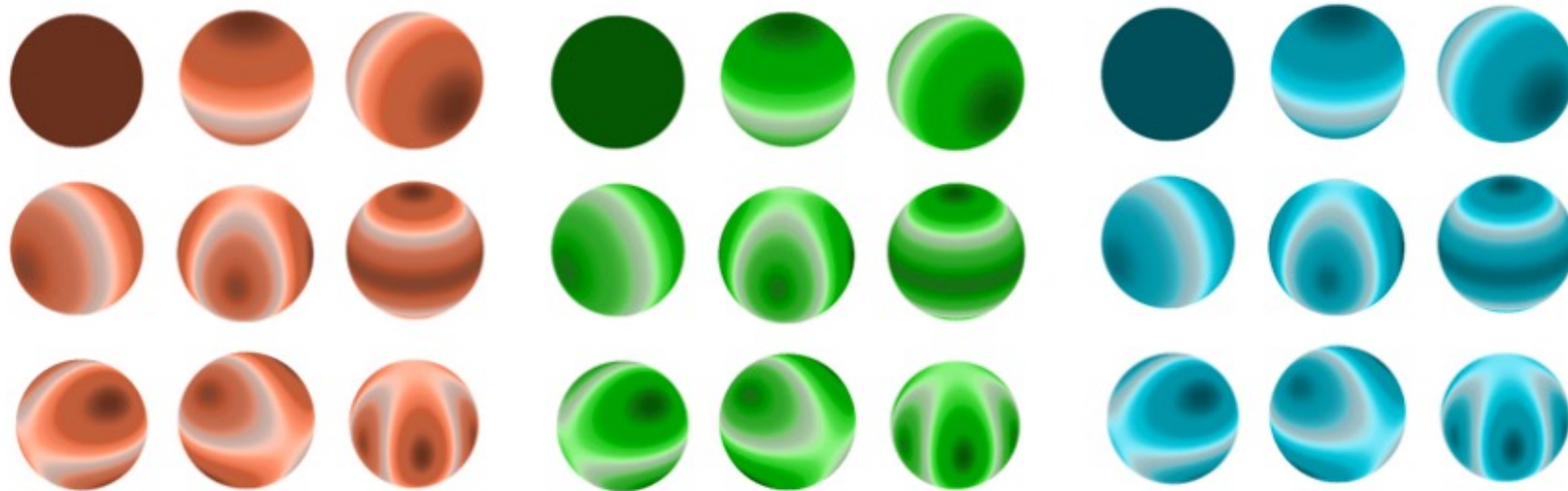


“Plenoptic” : Spherical Harmonics

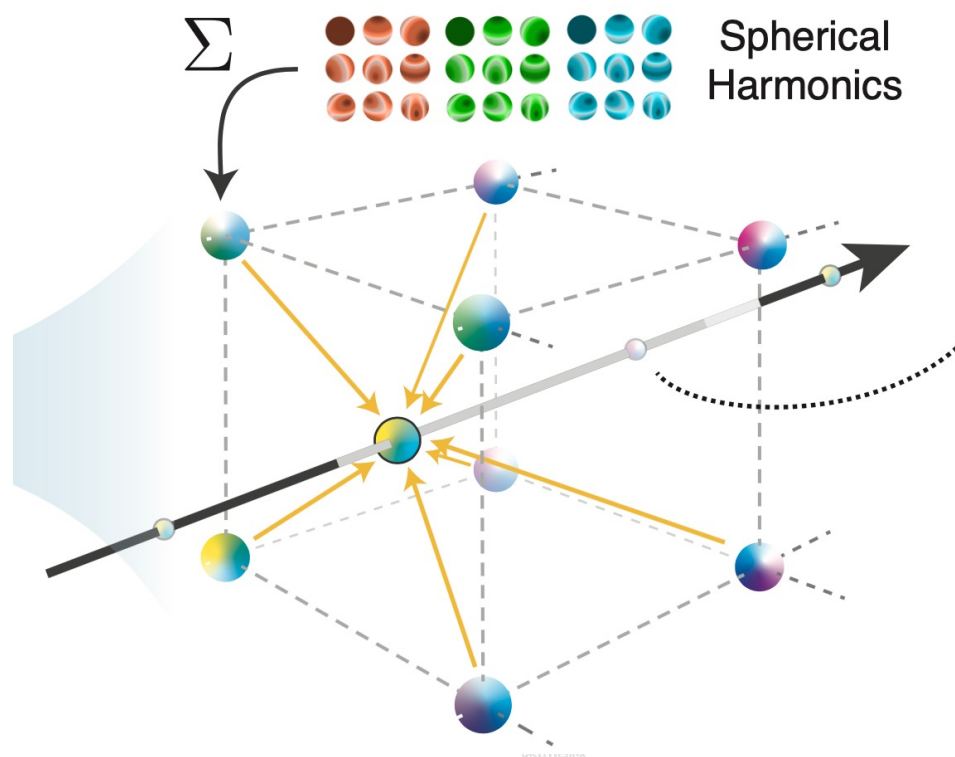


“Plenoptic” : Spherical Harmonics

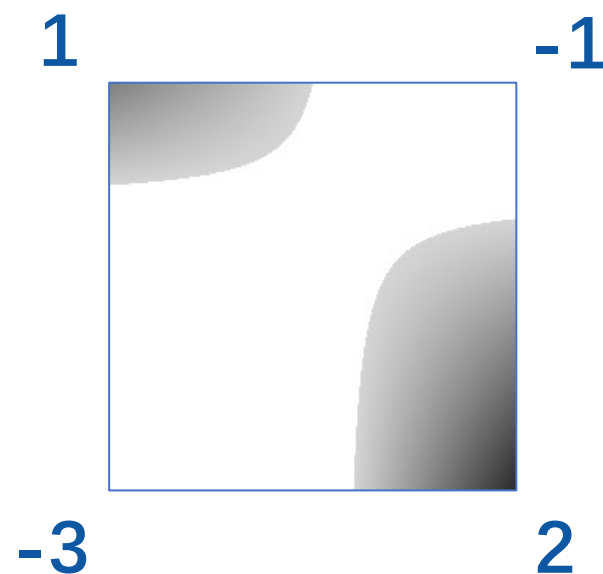
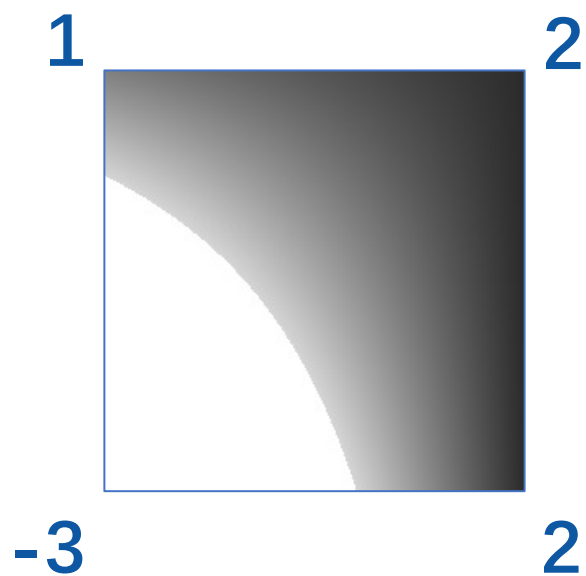
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Trilinear Interpolation



Trilinear Interpolation of Density

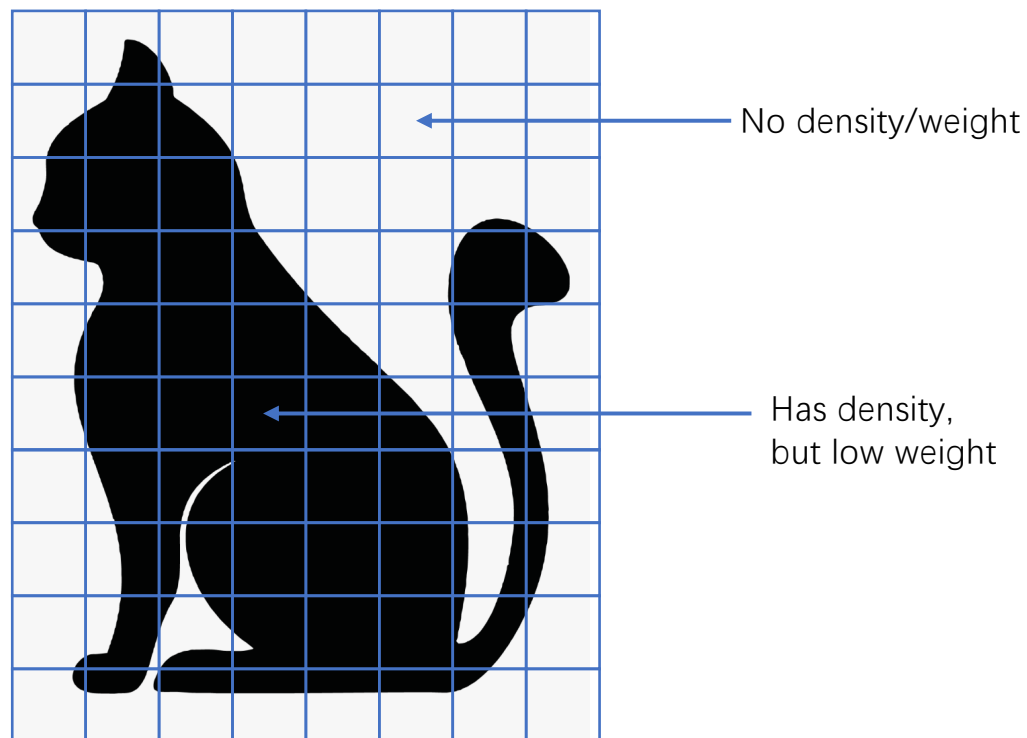


Sparsity: Voxel Pruning

Weight:

$$T_i(1 - \exp(-\sigma_i \delta_i))$$

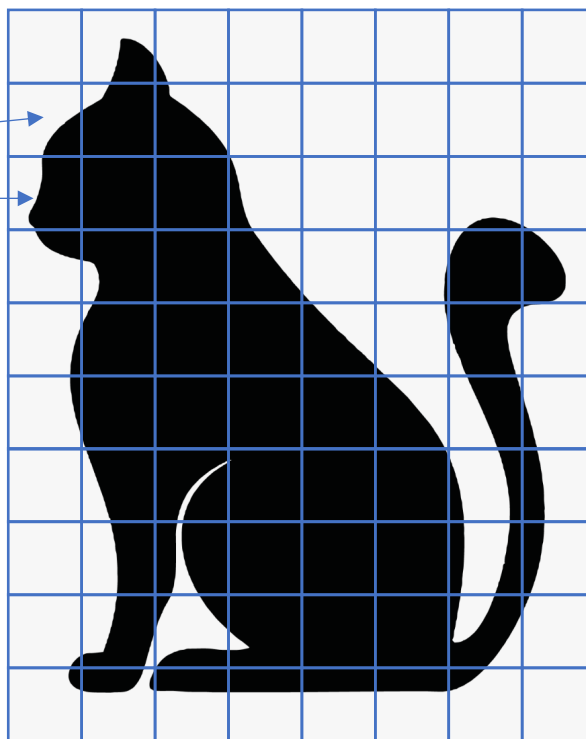
$$\text{where } T_i = \exp\left(-\sum_{j=1}^{i-1} \sigma_j \delta_j\right)$$



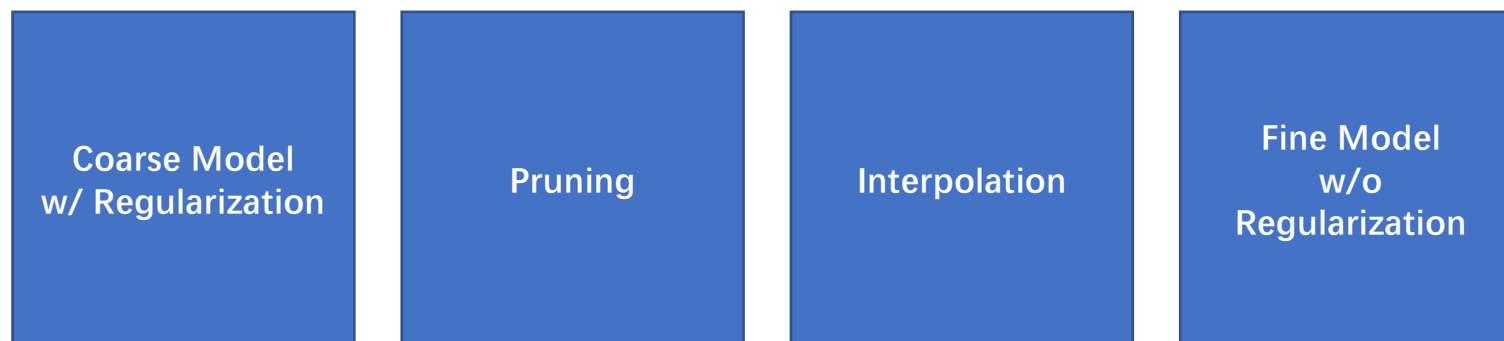
Regularization: Total Variation

Adjacent voxels
should be similar

$$\frac{1}{|\mathcal{V}|} \sum_{\substack{v \in \mathcal{V} \\ d \in [D]}} \sqrt{\Delta_x^2(\mathbf{v}, d) + \Delta_y^2(\mathbf{v}, d) + \Delta_z^2(\mathbf{v}, d)}$$



| Coarse to Fine



THANK YOU

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