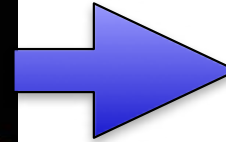


# Neural Radiance Fields and Surfaces



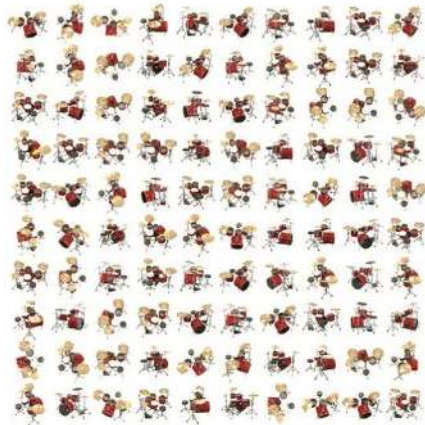
# Neural Radiance Fields



Input Images

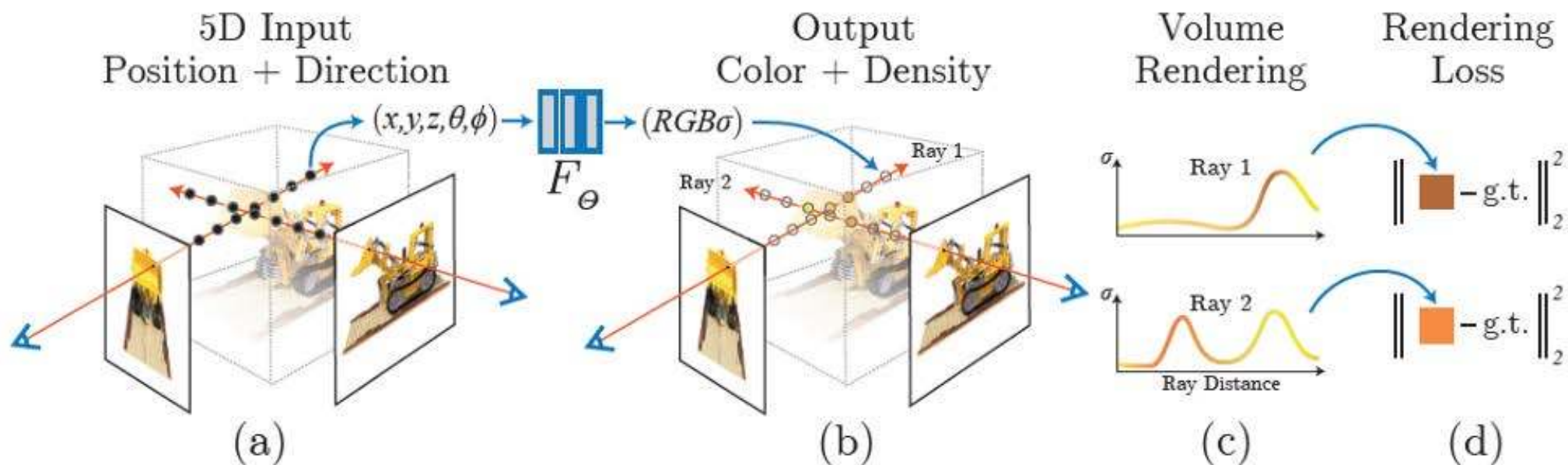
Optimize NeRF

Render new views



Multiple views of a complex scene

# Neural Radiance Fields




- A. Sampling 5D coordinates---location  $x, y, z$  and viewing direction  $\theta, \phi$ ---along camera rays.
- B. Feeding those locations into an MLP to produce a color and volume density.
- C. Using volume rendering techniques to composite these values into an image.
- D. Optimizing scene representation by minimizing the residual between synthesized and ground truth images.

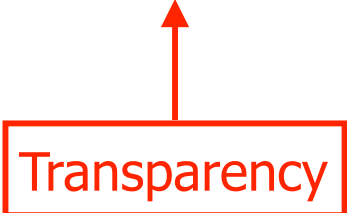
# Physically Inspired Volume Rendering

For a ray  $\mathbf{r}(t) = \mathbf{o} + t\mathbf{d}$ , the rendered color can be computed as

$$C(\mathbf{r}) = \int_{t_n}^{t_f} T(t) \sigma(\mathbf{r}(t)) \mathbf{c}(\mathbf{r}(t), \mathbf{d}) dt$$

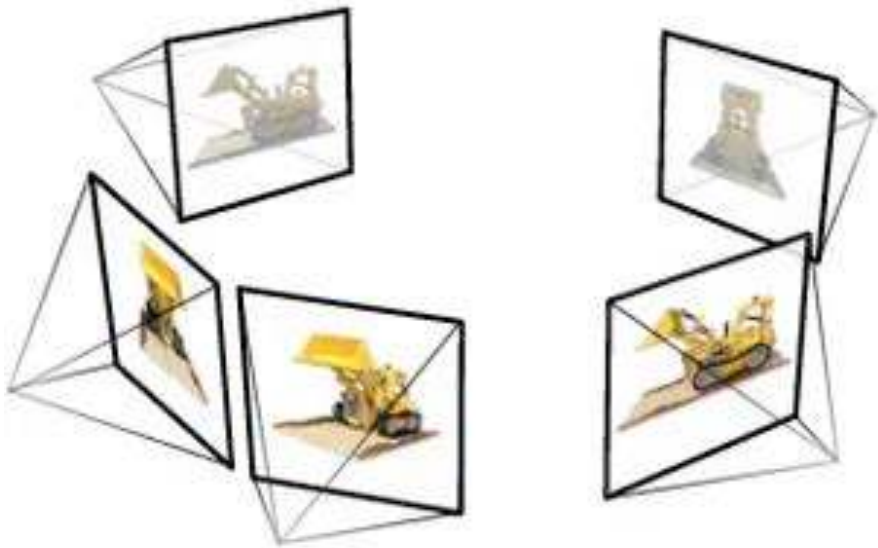


with  $T(t) = \exp\left(-\int_{t_n}^{t_f} \sigma(\mathbf{r}(s)) ds\right)$





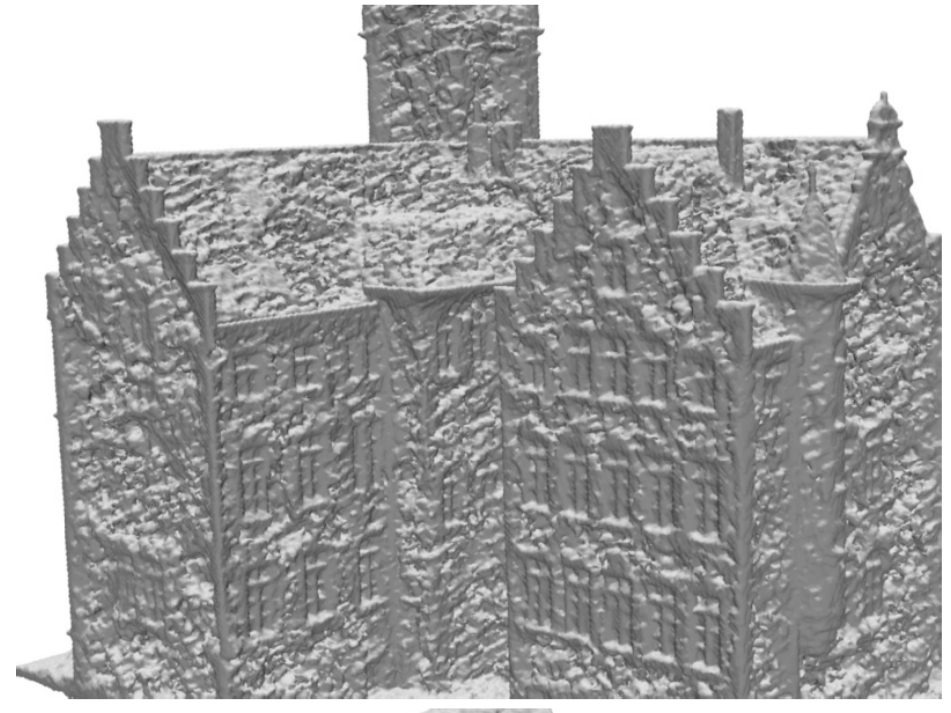
# Neural Rendering



Given a few images of a tractor



# Thresholding the Density

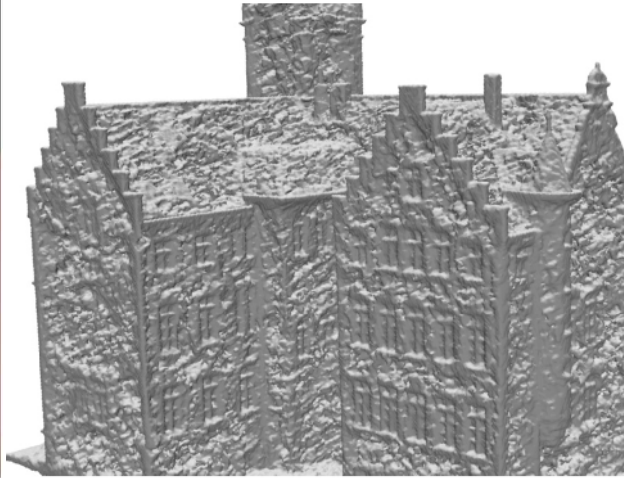


- Surfaces obtained by thresholding the density
- Choosing the threshold can be problematic

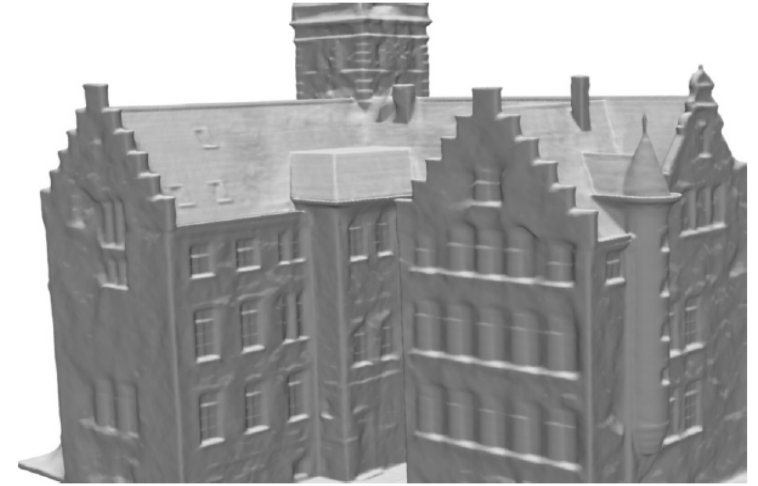
# From NerF to NeuS



Image



NeRF



NeuS

- Volume density is expressed a function of an SDF
- The reconstructed surfaces are smoother



# From Interpolation to Reconstruction



Images of a shiny statue



View Interpolation



3D Reconstruction