

# Neural Radiance Fields pt 3



made with



CS180/280A: Intro to Computer Vision and Computational Photography

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UC Berkeley Fall 2023

Lots of content from Noah Snavely and Ben Mildenhall, Pratul Srinivasan, and Matt Tancik from [ECCV 2022 Tutorial on Neural Volumetric Rendering for Computer Vision](#)

# Class Choice Award Project 3

Winner:

Chloe Zhong

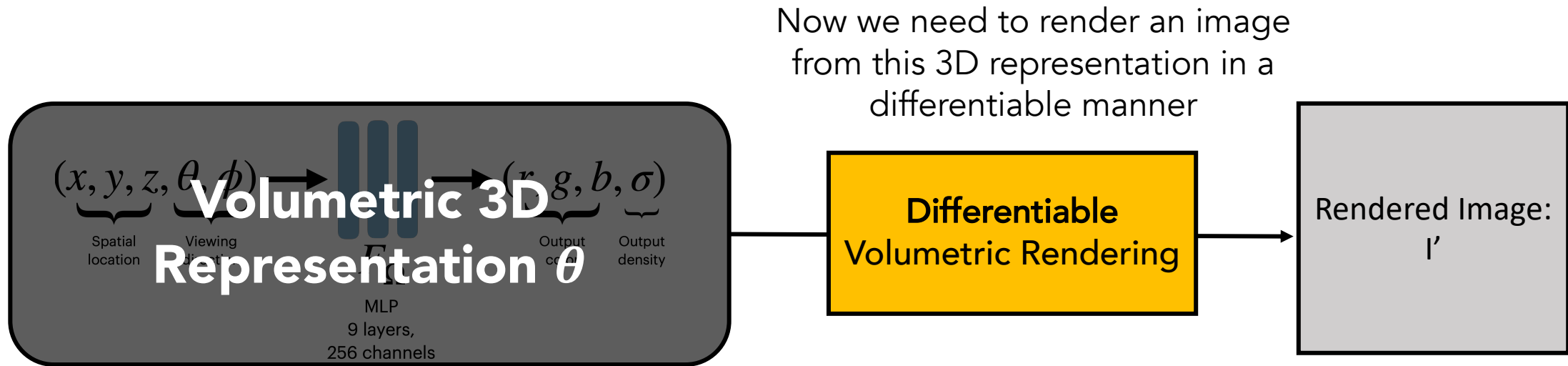
<https://inst.eecs.berkeley.edu/~cs180/fa23/upload/files/proj3/czhongx4>

Tied 2<sup>nd</sup> and 3<sup>rd</sup>: Irene Geng and Jai Singh

<https://inst.eecs.berkeley.edu/~cs180/fa23/upload/files/proj3/irenegeng2/>

<https://inst.eecs.berkeley.edu/~cs180/fa23/upload/files/proj3/jai.s>

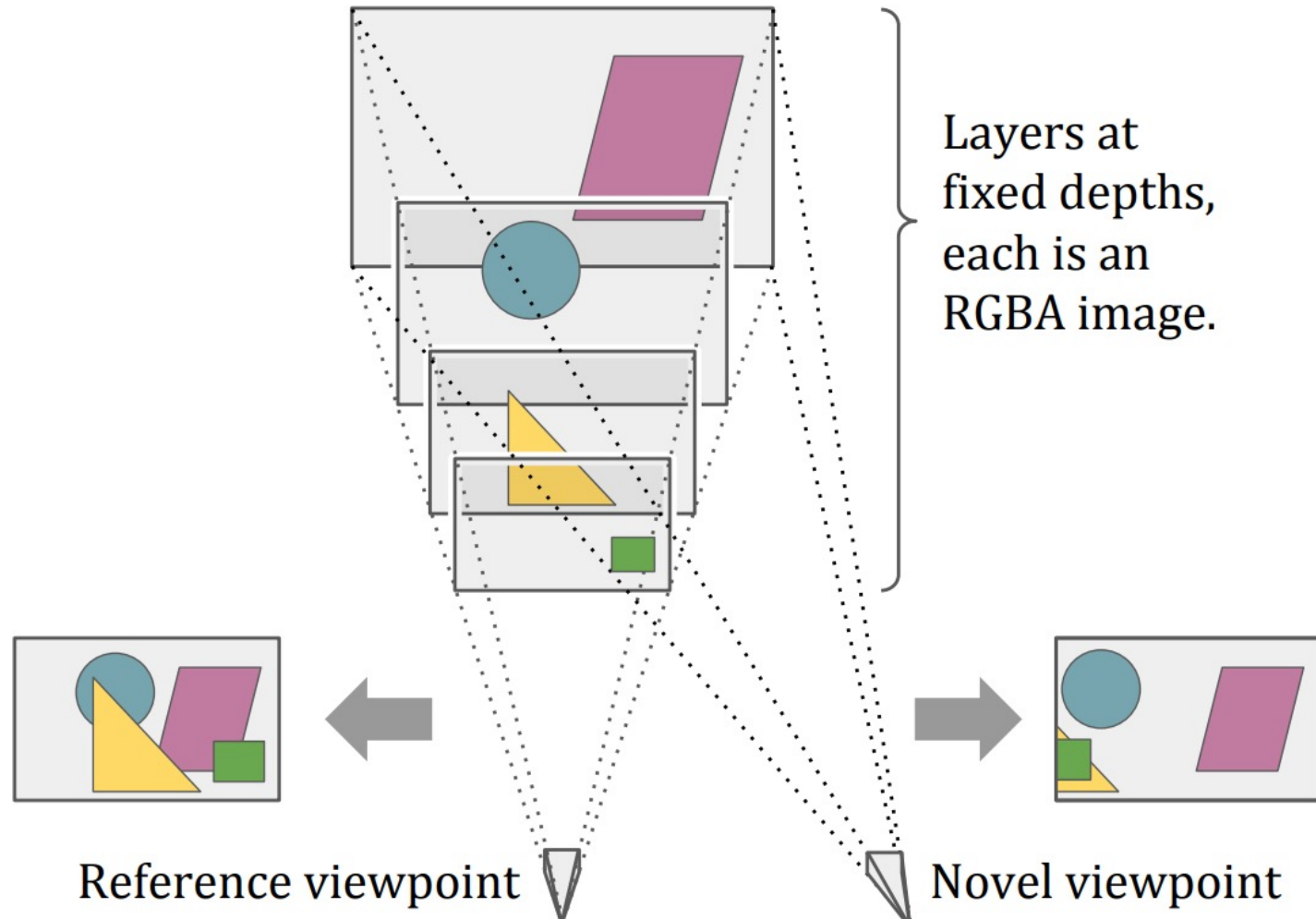
# Where we are



“Training” Objective (aka Analysis-by-Synthesis):

$$\min_{\theta} \left\| \begin{array}{c} \text{Rendered Image:} \\ I' \end{array} - \begin{array}{c} \text{Observed Image:} \\ I \end{array} \right\|_2$$

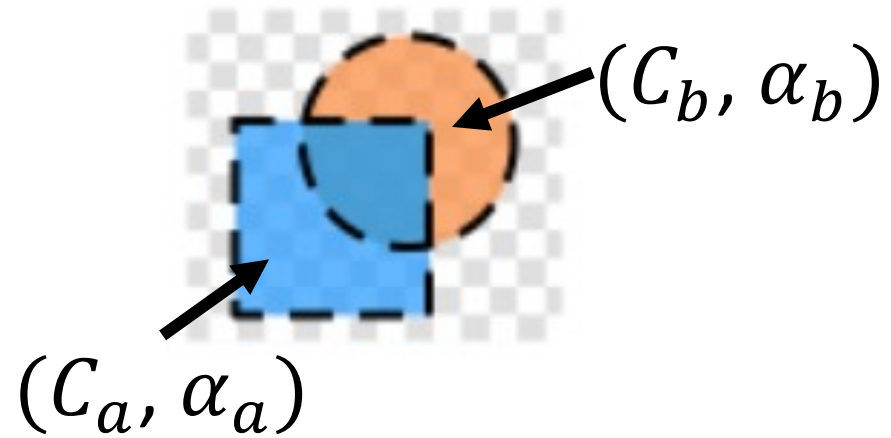
# A Precursor: Multi-plane Images



Also called front-to-back compositing or "over" operation

# Alpha Blending

for two image case, A and B,  
both partially transparent:

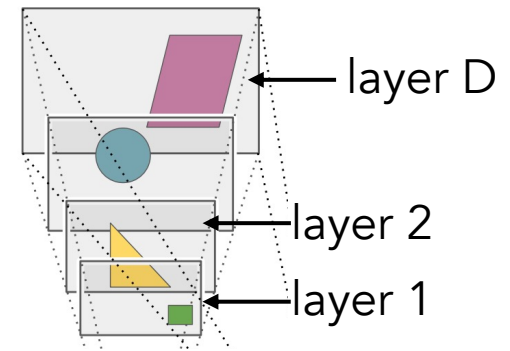


$$I = C_a \alpha_a + C_b \alpha_b \underbrace{(1 - \alpha_a)}$$

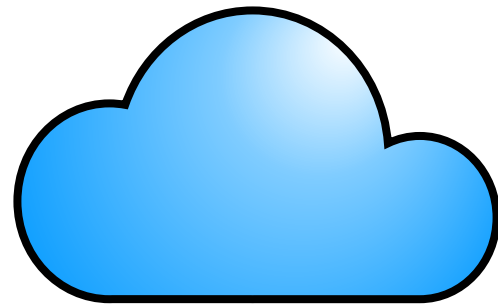
How much light is the previous layer letting through?

General D layer case:

$$I = \sum_{i=1}^D C_i \alpha_i \prod_{j=1}^{i-1} (1 - \alpha_j)$$

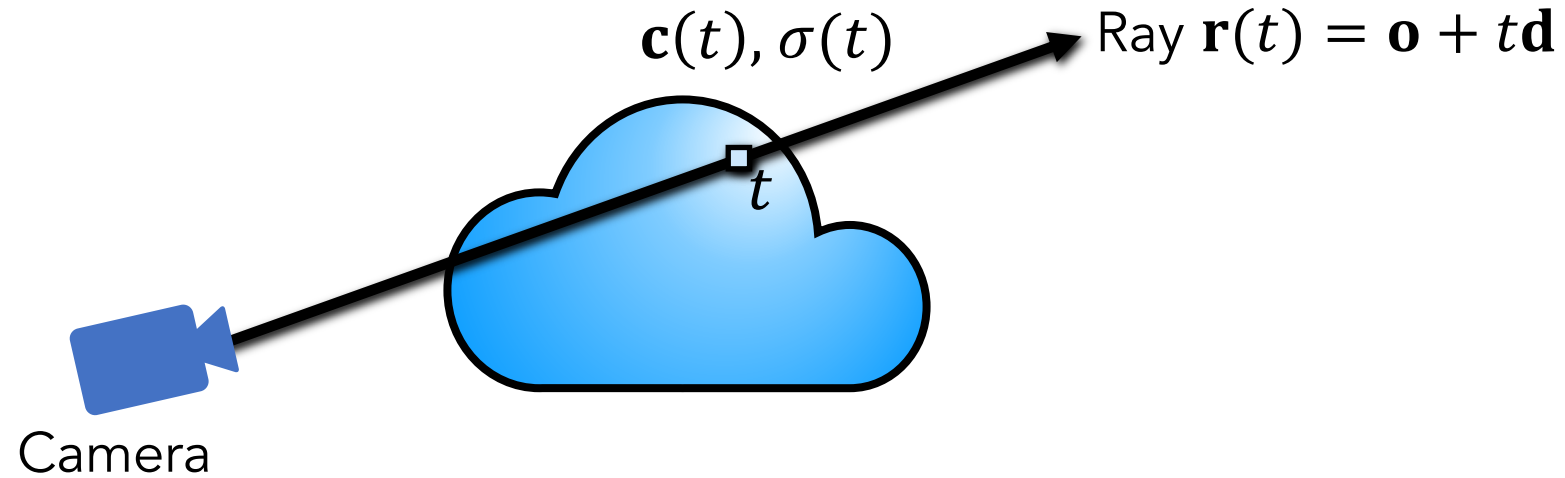


# Volumetric formulation for NeRF



Scene is a cloud of tiny colored particles

# Volumetric formulation for NeRF



at a point on the ray  $\mathbf{r}(t)$  , we can query color  $\mathbf{c}(t)$  and density  $\sigma(t)$

How to integrate all the info along the ray to get a color per ray?

# Idea: Expected Color

- Pose probabilistically.
- Each point on the ray has a probability to be the first "hit" :  $P[\textit{first hit at } t]$
- Color per ray = Expected value of color with this probability of first "hit"

for a ray  $\mathbf{r}(t) = \mathbf{o} + t\mathbf{d}$ :

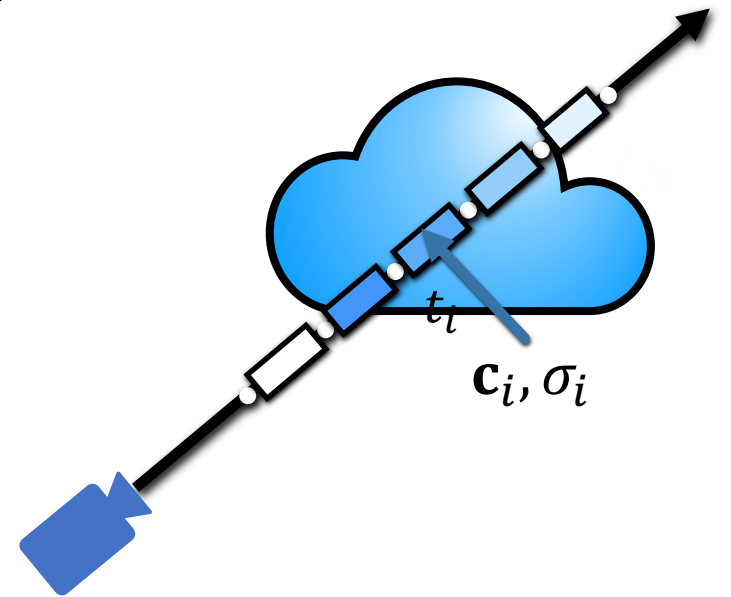
$$\mathbf{c}(\mathbf{r}) = \int_{t_0}^{t_1} P[\textit{first hit at } t] \mathbf{c}(t) dt$$

$$\approx \sum_{t=0}^T P[\textit{first hit at } t] \mathbf{c}(t)$$

$$\approx \sum_{t=0}^T w_t \mathbf{c}(t)$$

$$= \sum_{i=1}^n T_i \alpha_i \mathbf{c}_i$$

$$\text{where } T_i = \prod_{j=1}^{i-1} (1 - \alpha_j) \quad \alpha_i = 1 - \exp(-\sigma_i \delta_i)$$





# Differentiable Volumetric Rendering Formula

for a ray  $\mathbf{r}(t) = \mathbf{o} + t\mathbf{d}$ :

$$\mathbf{c} \approx \sum_{i=1}^n w_i \mathbf{c}_i$$

differentiable w.r.t.  $\mathbf{c}, \sigma$

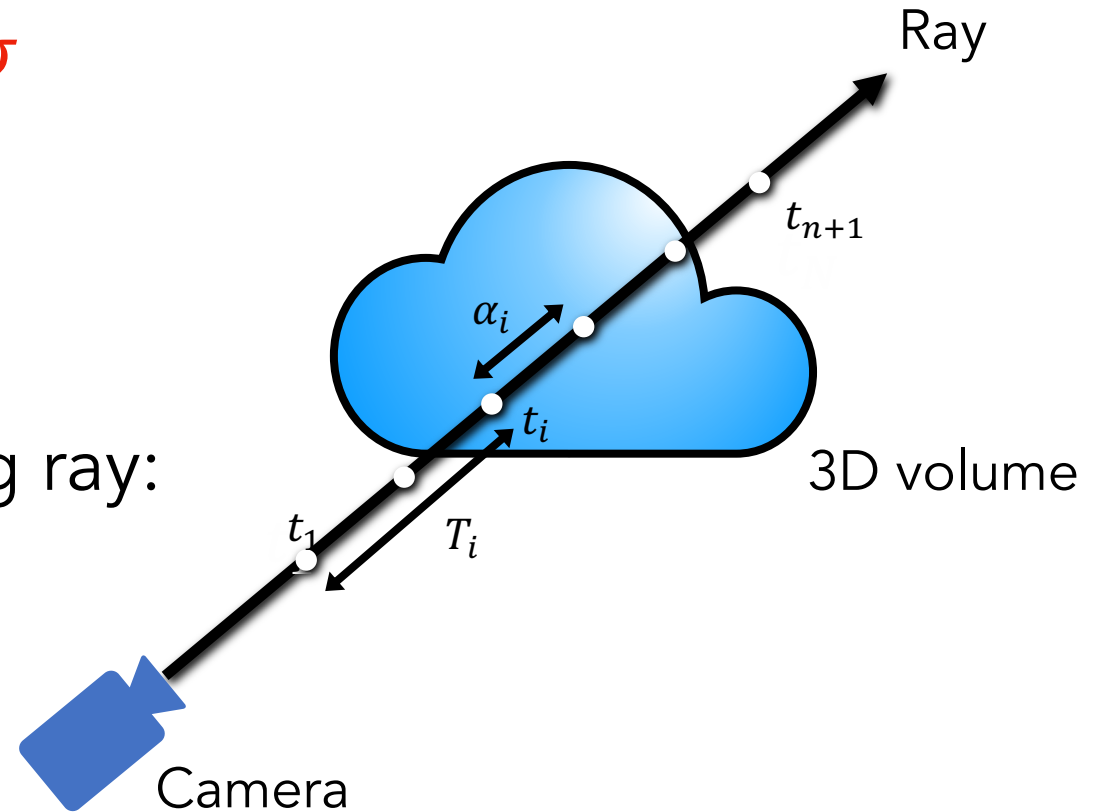
weights → colors

How much light is blocked earlier along ray:

$$T_i = \prod_{j=1}^{i-1} (1 - \alpha_j)$$

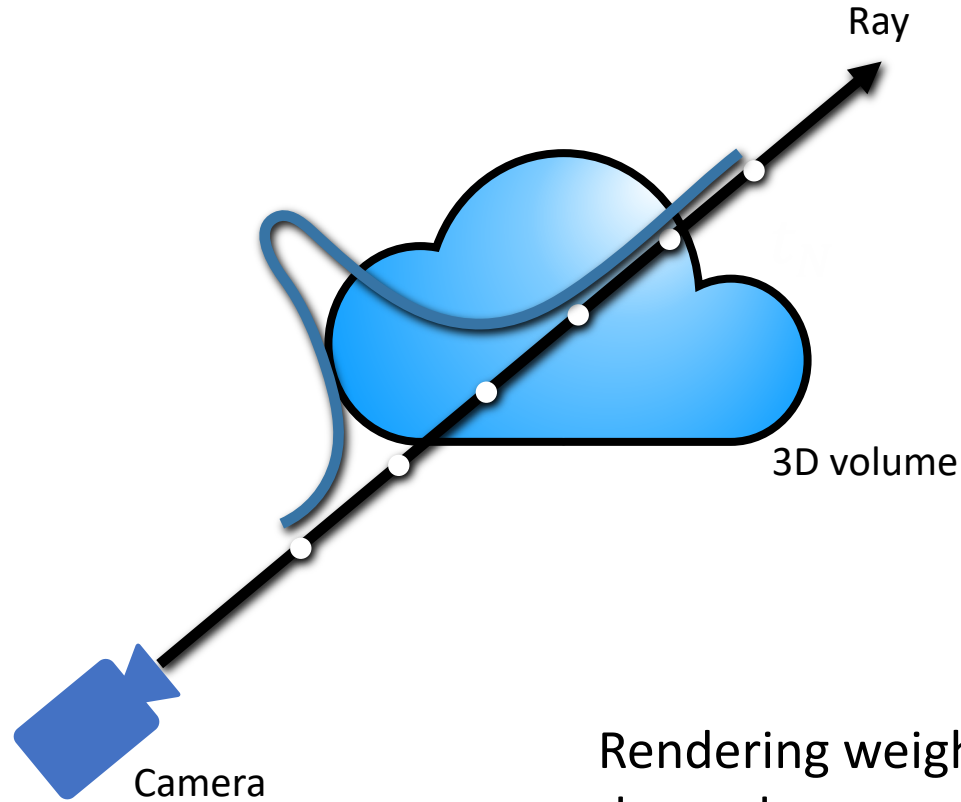
How much light is contributed by ray segment  $i$ :

$$\alpha_i = 1 - \exp(-\sigma_i \delta_i)$$



# Visual intuition: rendering weights is specific to a ray

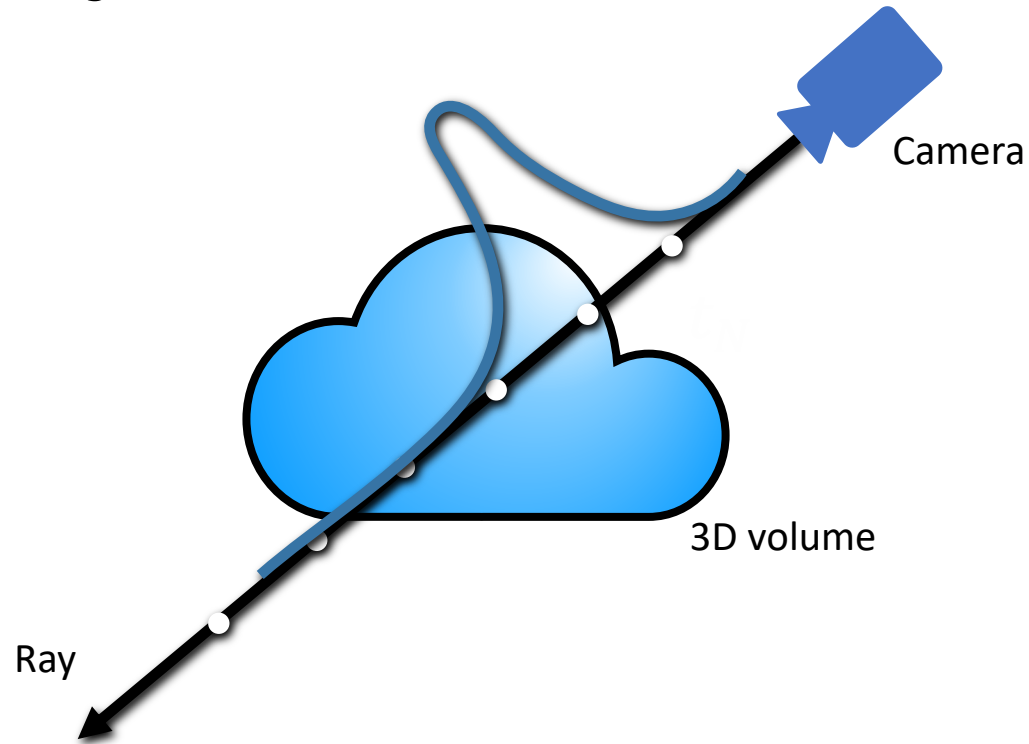
$$C \approx \sum_{i=1}^N T_i \alpha_i c_i$$



Rendering weights are not a 3D function — depends on ray, because of transmittance!

# Visual intuition: rendering weights is specific to a ray

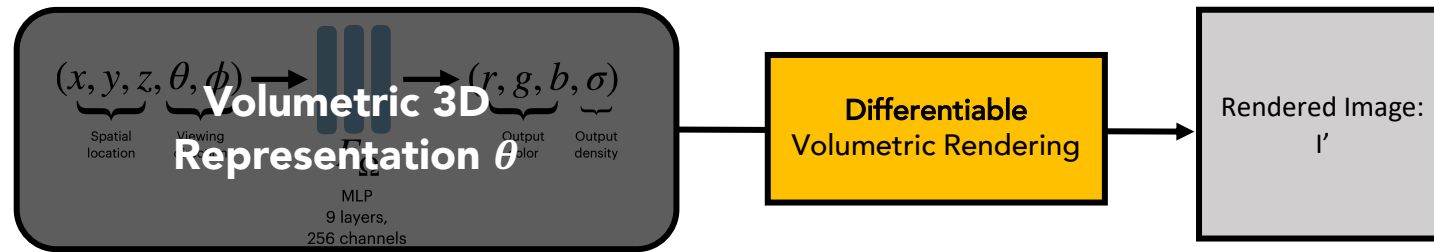
$$C \approx \sum_{i=1}^N T_i \alpha_i c_i$$



Rendering weights are not a 3D function — depends on ray, because of tranmistance!

# What's the point

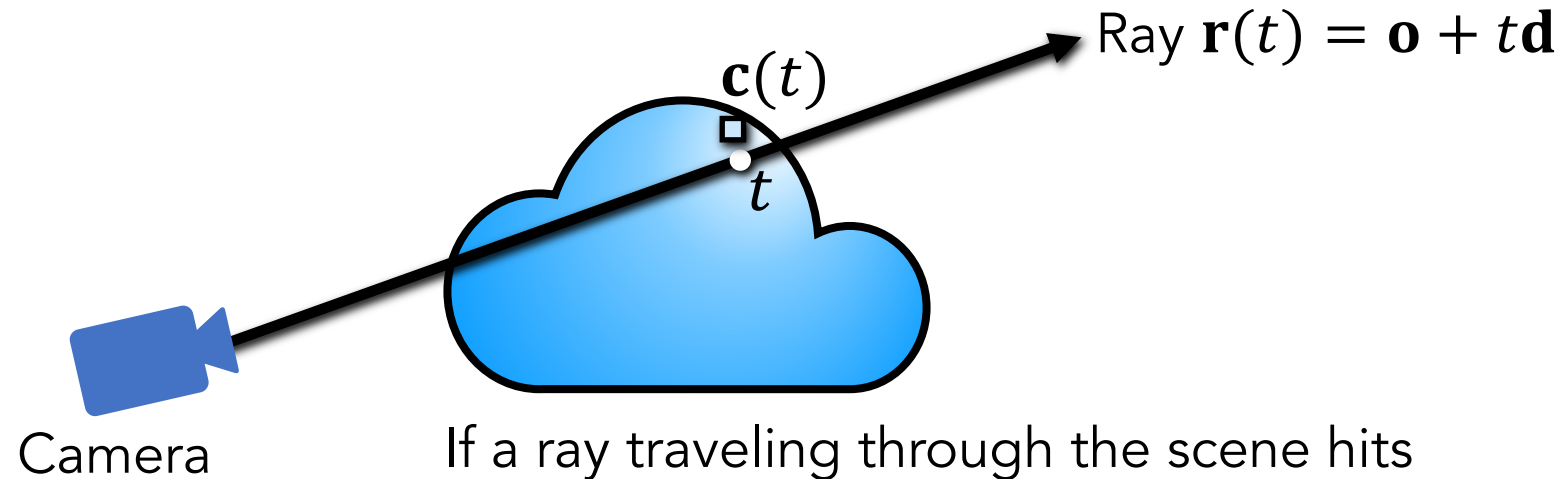
- Remember, for each pixel or a ray we render a color with this formula based on the Volumetric 3D Representation
- We use this to supervise the 3D Representation (sigma, RGB volume)



"Training" Objective (aka Analysis-by-Synthesis):

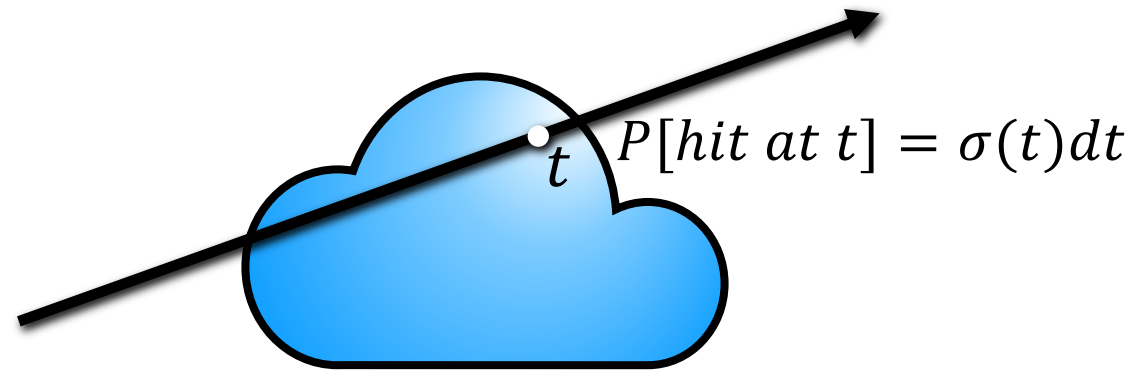
$$\min_{\theta} \left\| \begin{array}{|c|} \hline \text{Rendered Image:} \\ \hline I' \\ \hline \end{array} - \begin{array}{|c|} \hline \text{Observed Image:} \\ \hline I \\ \hline \end{array} \right\|_2$$

Let's derive this:



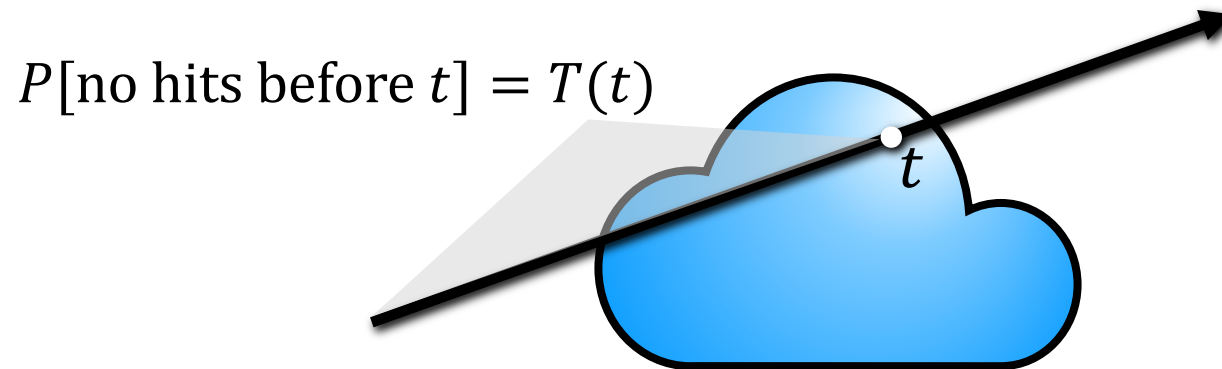
If a ray traveling through the scene hits a particle at distance  $t$  along the ray, we return its color  $\mathbf{c}(t)$

What does it mean for a ray to “hit” the volume?



This notion is *probabilistic*: chance that ray hits a particle in a small interval around  $t$  is  $\sigma(t)dt$ .  
 $\sigma$  is called the “volume density”

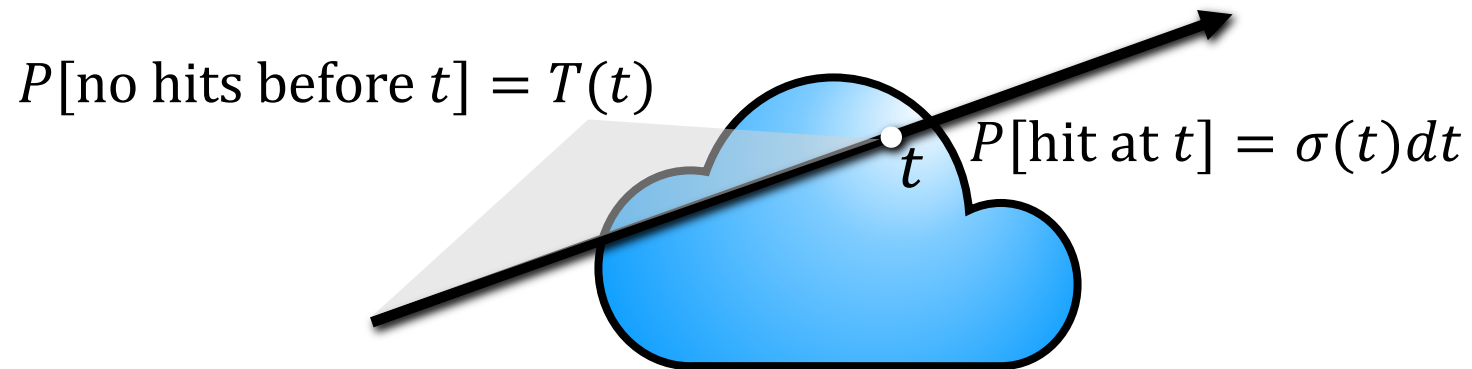
# Is it the **first** hit?



To determine if  $t$  is the *first* hit along the ray, need to know  $T(t)$ : **the probability that the ray makes it through the volume up to  $t$ .**

$T(t)$  is called "transmittance"

# Define First Fit



The product of these probabilities tells us how much you see the particles at  $t$ :

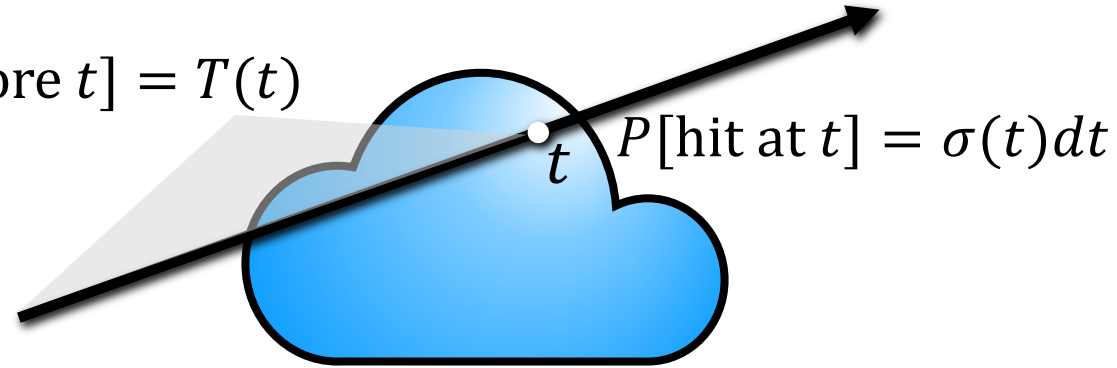
$$P[\text{first hit at } t] = P[\text{no hit before } t] \times P[\text{hit at } t] = T(t)\sigma(t)dt$$

Also called Ray Termination  
Let's write  $T$  as a function of  $\sigma$  ! How?



# Calculating $T$ given $\sigma$

$$P[\text{no hits before } t] = T(t)$$



$$\text{We got: } P[\text{first hit at } t] = P[\text{no hit before } t] \times P[\text{hit at } t] = T(t)\sigma(t)dt$$

Now use a slightly different equation to relate  $\sigma$  and  $T$ :

$$\underbrace{P[\text{no hit before } t + dt]}_{T(t + dt)} = \underbrace{P[\text{no hit before } t]}_{T(t)} \times \underbrace{P[\text{no hit at } t]}_{(1 - \sigma(t)dt)}$$

Now we can solve for  $T$  as a function of  $\sigma$

# Solve for $T$ as a function of $\sigma$

$$\underbrace{P[\text{no hit before } t + dt]}_{T(t + dt)} = \underbrace{P[\text{no hit before } t]}_{T(t)} \times \underbrace{P[\text{no hit at } t]}_{(1 - \sigma(t)dt)}$$

Solve the differential equation

$$T(t + dt) = T(t)(1 - \sigma(t)dt)$$

Expanded Righthand side

$$\text{Taylor expansion} \Rightarrow \cancel{T(t)} + T'(t)dt = \cancel{T(t)} - T(t)\sigma(t)dt$$

$$\text{Rearrange} \Rightarrow \frac{T'(t)}{T(t)} dt = -\sigma(t)dt$$

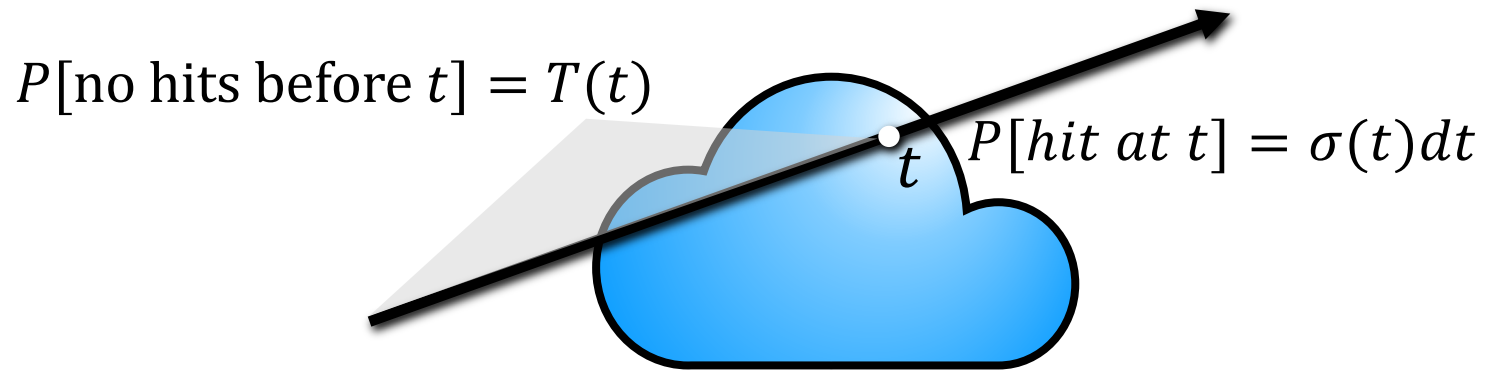
$$\text{Integrate} \Rightarrow \log T(t) = -\int_{t_0}^t \sigma(s)ds$$

$$\text{Exponentiate} \Rightarrow T(t) = \exp\left(-\int_{t_0}^t \sigma(s)ds\right)$$

Integral of:

$$\int \frac{f'(x)}{f(x)} dx = \log f(x)$$

# Finally, we can write the ray termination PDF



Finally, we can write the probability that a ray terminates at  $t$  as a function of only sigma

$$P[\text{first hit at } t] = P[\text{no hit before } t] \times P[\text{hit at } t]$$

$$= T(t)\sigma(t)dt$$

$$= \exp\left(-\int_{t_0}^t \sigma(s)ds\right) \sigma(t)dt$$

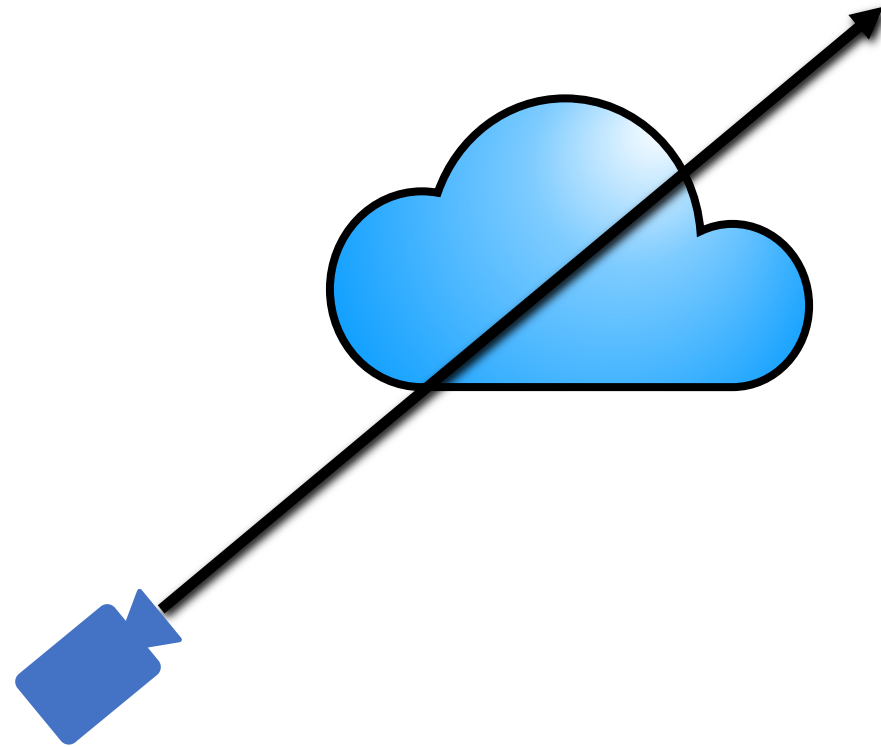
# Finally, Expected Color along the ray

Then, the expected color returned by the ray will be

$$\begin{aligned}\mathbf{c}(\mathbf{r}) &= \int_{t_0}^{t_1} P[\textit{first hit at } t] \mathbf{c}(t) dt \\ &= \int_{t_0}^{t_1} T(t) \sigma(t) \mathbf{c}(t) dt \\ &= \int_{t_0}^{t_1} \exp\left(-\int_{t_0}^t \sigma(s) ds\right) \sigma(t) \mathbf{c}(t) dt\end{aligned}$$

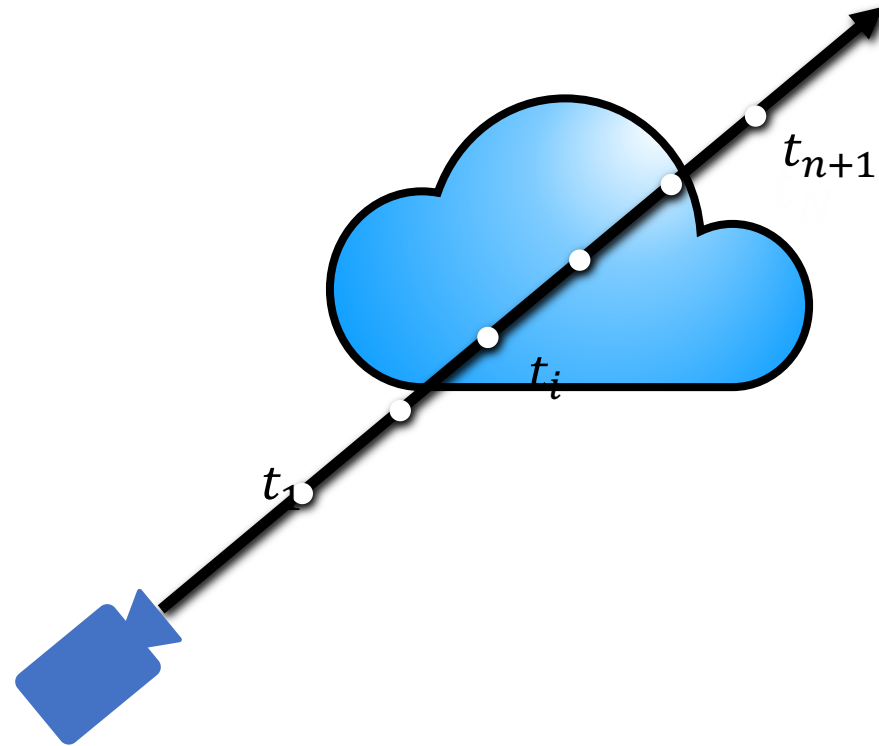
Note the nested integral!

# Approximating the nested integral



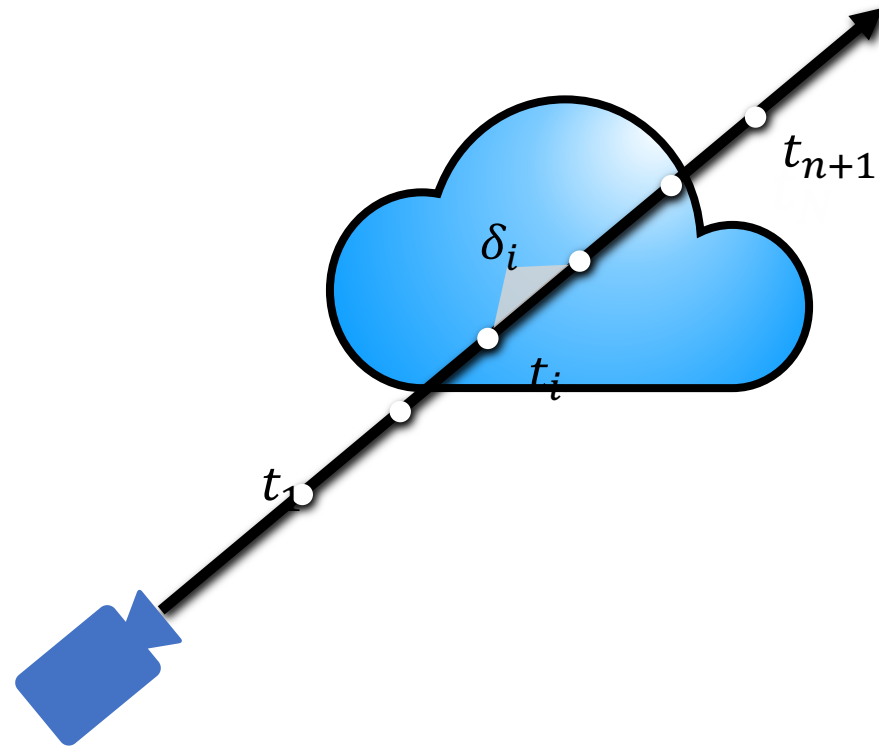
We use quadrature to approximate the nested integral,

# Approximating the nested integral



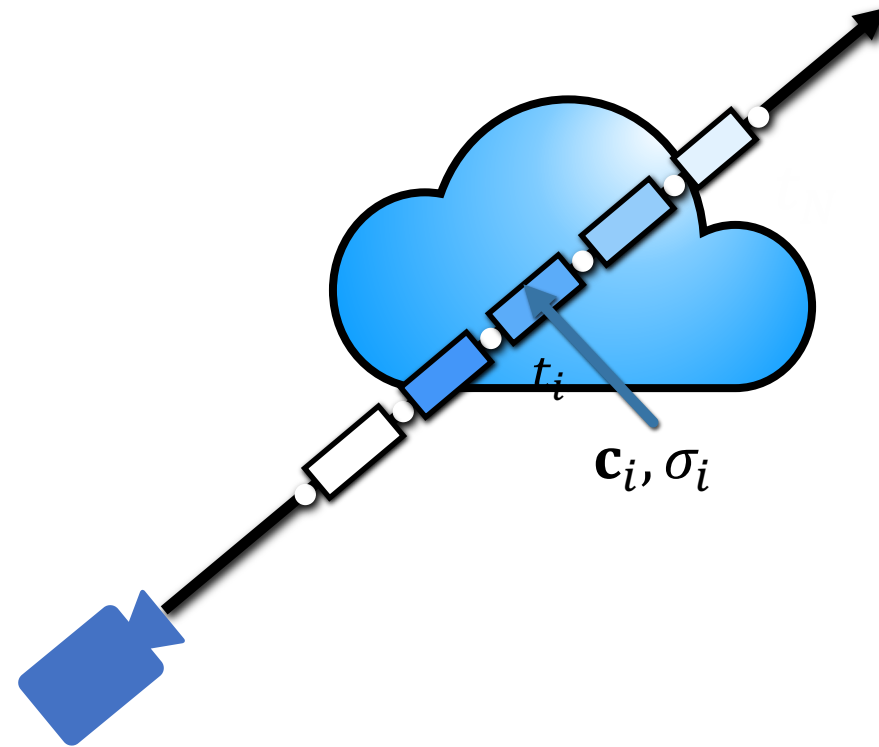
We use quadrature to approximate the nested integral, splitting the ray up into  $n$  segments with endpoints  $\{t_1, t_2, \dots, t_{n+1}\}$

# Approximating the nested integral



We use quadrature to approximate the nested integral, splitting the ray up into  $n$  segments with endpoints  $\{t_1, t_2, \dots, t_{n+1}\}$  with lengths  $\delta_i = t_{i+1} - t_i$

# Approximating the nested integral



We assume volume density and color are roughly constant within each interval



# Deriving quadrature estimate

Expected color:  $\int T(t)\sigma(t)\mathbf{c}(t)dt \approx$

This allows us to break the outer integral

# Deriving quadrature estimate

Expected color:  $\int T(t)\sigma(t)\mathbf{c}(t)dt \approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt$

This allows us to break the outer integral into a sum of analytically tractable integrals

# Deriving quadrature estimate

$$\int T(t)\sigma(t)\mathbf{c}(t)dt \approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt$$

Caveat: piecewise constant density and color  
**do not** imply constant transmittance!

# Deriving quadrature estimate

$$\int T(t)\sigma(t)\mathbf{c}(t)dt \approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt$$

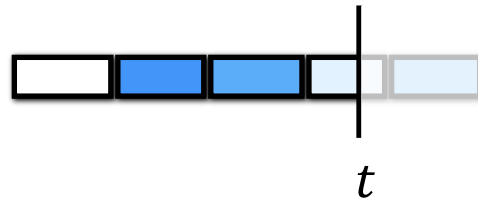
Caveat: piecewise constant density and color **do not** imply constant transmittance!

Important to account for how early part of a segment blocks later part when  $\sigma_i$  is high

# Evaluating $T$ for piecewise constant density

$$\text{For } t \in [t_i, t_{i+1}], T(t) = \exp\left(-\int_{t_1}^{t_i} \sigma_i ds\right) \exp\left(-\int_{t_i}^t \sigma_i ds\right)$$

We need to evaluate at continuous  $t$  values that can lie *partway through* an interval

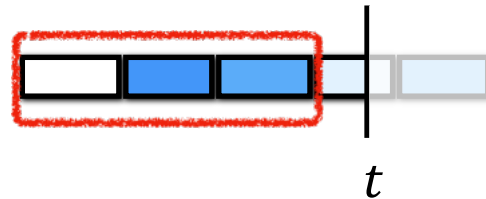


# Evaluating $T$ for piecewise constant density

$$\text{For } t \in [t_i, t_{i+1}], T(t) = \exp\left(-\int_{t_1}^{t_i} \sigma_i ds\right) \exp\left(-\int_{t_i}^t \sigma_i ds\right)$$




$$\exp\left(-\sum_{j=1}^{i-1} \sigma_j \delta_j\right) = T_i \text{ "How much light is blocked by all previous segments?"}$$

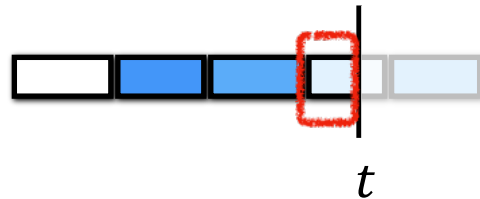


# Evaluating $T$ for piecewise constant density

$$\text{For } t \in [t_i, t_{i+1}], T(t) = \exp\left(-\int_{t_1}^{t_i} \sigma_i ds\right) \exp\left(-\int_{t_i}^t \sigma_i ds\right)$$

“How much light is blocked partway through the current segment?”


$$\exp(-\sigma_i(t - t_i))$$



# Deriving quadrature estimate

$$\int T(t)\sigma(t)\mathbf{c}(t)dt \approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt$$



# Deriving quadrature estimate

$$\int T(t)\sigma(t)\mathbf{c}(t)dt \approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt$$

Substitute=  $\sum_{i=1}^n T_i\sigma_i\mathbf{c}_i \int_{t_i}^{t_{i+1}} \exp(-\sigma_i(t - t_i))dt$

# Deriving quadrature estimate

$$\begin{aligned}\int T(t)\sigma(t)\mathbf{c}(t)dt &\approx \sum_{i=1}^n \int_{t_i}^{t_{i+1}} T(t)\sigma_i\mathbf{c}_i dt \\ &= \sum_{i=1}^n T_i\sigma_i\mathbf{c}_i \int_{t_i}^{t_{i+1}} \exp(-\sigma_i(t-t_i))dt\end{aligned}$$

Integral of Exponential:

$$\int \exp(-ax) dx = -\frac{1}{a} \exp(-ax)$$

$$\int_{t_i}^{t_{i+1}} \exp(-\sigma(t-t_i)) dt = -\frac{1}{\sigma} \exp(-\sigma(t-t_i)) \Big|_{t_i}^{t_{i+1}}$$

$$\frac{\exp(-\sigma_i(t_{i+1}-t_i)) - \exp(-\sigma_i(t_i-t_i))}{-\sigma_i} = \frac{\exp(-\sigma_i(t_{i+1}-t_i)) - 1}{-\sigma_i}$$

$$\text{Integrate} = \sum_{i=1}^n T_i\sigma_i\mathbf{c}_i \frac{\exp(-\sigma_i(t_{i+1}-t_i)) - 1}{-\sigma_i}$$

$$\text{Cancel } \sigma_i = \sum_{i=1}^n T_i\mathbf{c}_i(1 - \exp(-\sigma_i\delta_i))$$

$$\text{Expected Color} = \sum_{i=1}^n T_i\mathbf{c}_i(1 - \exp(-\sigma_i\delta_i))$$

# Putting it all together

$$\text{Expected Color} = \sum_{i=1}^n T_i \mathbf{c}_i (1 - \exp(-\sigma_i \delta_i))$$

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

# Connection to alpha compositing

$$\text{Expected Color} = \sum_{i=1}^n T_i \mathbf{c}_i (1 - \exp(-\sigma_i \delta_i))$$

segment  
opacity  $\alpha_i$

$$\text{Expected Color} = \sum_{i=1}^n T_i \mathbf{c}_i \alpha_i$$

where

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

$$\prod_i \exp(x_i) = \exp\left(\sum_i x_i\right)$$
$$\alpha_i = 1 - \exp(-\sigma_i \delta_i)$$
$$1 - \alpha_i = \exp(-\sigma_i \delta_i)$$

# Summary

for a ray  $\mathbf{r}(t) = \mathbf{o} + t\mathbf{d}$ :

$$\mathbf{c} \approx \sum_{i=1}^n w_i \mathbf{c}_i = \sum_{i=1}^n T_i \alpha_i \mathbf{c}_i$$

weights      colors

differentiable w.r.t.  $\mathbf{c}, \sigma$

How much light is blocked earlier along ray:

$$T_i = \prod_{j=1}^{i-1} (1 - \alpha_j)$$

How much light is contributed by ray segment  $i$ :

$$\alpha_i = 1 - \exp(-\sigma_i \delta_i)$$

