# Volume Rendering Digest (for NeRF) 

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Neural Radiance Fields [3] employ simple volume rendering as a way to overcome the challenges of differentiating through ray-triangle intersections by leveraging a probabilistic notion of visibility. This is achieved by assuming the scene is composed by a cloud of light-emitting particles whose density changes in space (in the terminology of physically-based rendering, this would be described as a volume with absorption and emission but no scattering [4, Sec 11.1]. In what follows, for the sake of exposition simplicity, and without loss of generality, we assume the emitted light does not change as a function of view-direction. This technical report is a condensed version of previous reports [1, 2], but rewritten in the context of NeRF, and adopting its commonly used notation 1 .

Transmittance. Let the density field $\sigma(\mathbf{x})$, with $\mathbf{x} \in \mathbb{R}^{3}$ indicate the differential likelihood of a ray hitting a particle (i.e. the probability of hitting a particle while travelling an infinitesimal distance). We reparameterize the density along a given ray $\mathbf{r}=(\mathbf{o}, \mathbf{d})$ as a scalar function $\sigma(t)$, since any point $\mathbf{x}$ along the ray can be written as $\mathbf{r}(t)=\mathbf{o}+t \mathbf{d}$. Density is closely tied to the transmittance function $\mathcal{T}(t)$, which indicates the probability of a ray traveling over the interval $[0, t)$ without hitting any particles. Then the probability $\mathcal{T}(t+d t)$ of not hitting a particle when taking a differential step $d t$ is equal to $\mathcal{T}(t)$, the likelihood of the ray reaching $t$, times $(1-d t \cdot \sigma(t))$, the probability of not hitting anything during the step:

$$
\begin{align*}
\mathcal{T}(t+d t) & =\mathcal{T}(t) \cdot(1-d t \cdot \sigma(t))  \tag{1}\\
\frac{\mathcal{T}(t+d t)-\mathcal{T}(t)}{d t} & \equiv \mathcal{T}^{\prime}(t)=-\mathcal{T}(t) \cdot \sigma(t) \tag{2}
\end{align*}
$$

This is a classical differential equation that can be solved as follows:

$$
\begin{align*}
\mathcal{T}^{\prime}(t) & =-\mathcal{T}(t) \cdot \sigma(t)  \tag{3}\\
\frac{\mathcal{T}^{\prime}(t)}{\mathcal{T}(t)} & =-\sigma(t)  \tag{4}\\
\int_{a}^{b} \frac{\mathcal{T}^{\prime}(t)}{\mathcal{T}(t)} d t & =-\int_{a}^{b} \sigma(t) d t  \tag{5}\\
\left.\log \mathcal{T}(t)\right|_{a} ^{b} & =-\int_{a}^{b} \sigma(t) d t  \tag{6}\\
\mathcal{T}(a \rightarrow b) \equiv \frac{\mathcal{T}(b)}{\mathcal{T}(a)} & =\exp \left(-\int_{a}^{b} \sigma(t) d t\right) \tag{7}
\end{align*}
$$

where we define $\mathcal{T}(a \rightarrow b)$ as the probability that the ray travels from distance $a$ to $b$ along the ray without hitting a particle, which is related to the previous notation by $\mathcal{T}(t)=\mathcal{T}(0 \rightarrow t)$.

[^0]Probabilistic interpretation. Note that we can also interpret the function $1-\mathcal{T}(t)$ (often called opacity) as a cumulative distribution function (CDF) indicating the probability that the ray does hit a particle sometime before reaching distance $t$. Then $\mathcal{T}(t) \cdot \sigma(t)$ is the corresponding probability density function (PDF), giving the likelihood that the ray stops precisely at distance $t$.

Volume rendering. We can now calculate the expected value of the light emitted by the particles in the volume as the ray travels from $t=0$ to $D$, composited on top of a background color. Since the probability density for stopping at $t$ is $\mathcal{T}(t) \cdot \sigma(t)$, the expected color is

$$
\begin{equation*}
C=\int_{0}^{D} \mathcal{T}(t) \cdot \sigma(t) \cdot \mathbf{c}(t) d t+\mathcal{T}(D) \cdot \mathbf{c}_{\mathrm{bg}} \tag{8}
\end{equation*}
$$

where $\mathbf{c}_{\mathrm{bg}}$ is a background color that is composited with the foreground scene according to the residual transmittance $\mathcal{T}(D)$. Without loss of generality, we omit the background term in what follows.

Homogeneous media. We can calculate the color of some homogeneous volumetric media with constant color $\mathbf{c}_{a}$ and density $\sigma_{a}$ over a ray segment $[a, b]$ by integration:

$$
\begin{array}{rlr}
\boldsymbol{C}(a \rightarrow b) & =\int_{a}^{b} \mathcal{T}(a \rightarrow t) \cdot \sigma(t) \cdot \mathbf{c}(t) d t & \\
& =\sigma_{a} \cdot \mathbf{c}_{a} \int_{a}^{b} \mathcal{T}(a \rightarrow t) d t & \text { constant density/radiance } \\
& =\sigma_{a} \cdot \mathbf{c}_{a} \int_{a}^{b} \exp \left(-\int_{a}^{t} \sigma(u) d u\right) d t & \\
& =\sigma_{a} \cdot \mathbf{c}_{a} \int_{a}^{b} \exp \left(-\left.\sigma_{a} u\right|_{a} ^{t}\right) d t & \text { substituting (7) } \\
& =\sigma_{a} \cdot \mathbf{c}_{a} \int_{a}^{b} \exp \left(-\sigma_{a}(t-a)\right) d t & \\
& =\left.\sigma_{a} \cdot \mathbf{c}_{a} \cdot \frac{\exp \left(-\sigma_{a}(t-a)\right)}{-\sigma_{a}}\right|_{a} ^{b} & \\
& =\mathbf{c}_{a} \cdot\left(1-\exp \left(-\sigma_{a}(b-a)\right)\right) \tag{15}
\end{array}
$$

Transmittance is multiplicative. Note that transmittance factorizes as follows:

$$
\begin{align*}
\mathcal{T}(a \rightarrow c)= & =\exp \left(-\left[\int_{a}^{b} \sigma(t) d t+\int_{b}^{c} \sigma(t) d t\right]\right)  \tag{16}\\
& =\exp \left(-\int_{a}^{b} \sigma(t) d t\right) \exp \left(-\int_{b}^{c} \sigma(t) d t\right)  \tag{17}\\
& =\mathcal{T}(a \rightarrow b) \cdot \mathcal{T}(b \rightarrow c) \tag{18}
\end{align*}
$$

This also follows from the probabilistic interpretation of $\mathcal{T}$, since the probability that the ray does not hit any particles within $[a, c]$ is the product of the probabilities of the two independent events that it does not hit any particles within $[a, b]$ or within $[b, c]$.

Transmittance for piecewise constant data. Given a set of intervals $\left\{\left[t_{n}, t_{n+1}\right]\right\}_{n=1}^{N}$ with constant density $\sigma_{n}$ within the $n$-th segment, and with $t_{1}=0$ and $\delta_{n}=t_{n+1}-t_{n}$, transmittance is equal to:

$$
\begin{equation*}
\mathcal{T}_{n}=\mathcal{T}\left(t_{n}\right)=\mathcal{T}\left(0 \rightarrow t_{n}\right)=\exp \left(-\int_{0}^{t_{n}} \sigma(t) d t\right)=\exp \left(\sum_{k=1}^{n-1}-\sigma_{k} \delta_{k}\right) \tag{19}
\end{equation*}
$$

Volume rendering of piecewise constant data. Combining the above, we can evaluate the volume rendering integral through a medium with piecewise constant color and density:

$$
\begin{array}{rlr}
\boldsymbol{C}\left(t_{N+1}\right) & =\sum_{n=1}^{N} \int_{t_{n}}^{t_{n+1}} \mathcal{T}(t) \cdot \sigma_{n} \cdot \mathbf{c}_{n} d t & \text { piecewise constant } \\
& =\sum_{n=1}^{N} \int_{t_{n}}^{t_{n+1}} \mathcal{T}\left(0 \rightarrow t_{n}\right) \cdot \mathcal{T}\left(t_{n} \rightarrow t\right) \cdot \sigma_{n} \cdot \mathbf{c}_{n} d t & \\
& =\sum_{n=1}^{N} \mathcal{T}\left(0 \rightarrow t_{n}\right) \int_{t_{n}}^{t_{n+1}} \mathcal{T}\left(t_{n} \rightarrow t\right) \cdot \sigma_{n} \cdot \mathbf{c}_{n} d t & \text { from (18) } \\
& =\sum_{n=1}^{N} \mathcal{T}\left(0 \rightarrow t_{n}\right) \cdot\left(1-\exp \left(-\sigma_{n}\left(t_{n+1}-t_{n}\right)\right)\right) \cdot \mathbf{c}_{n} & \text { constant } \tag{23}
\end{array}
$$

This leads to the volume rendering equations from NeRF [3, Eq.3]:

$$
\begin{equation*}
\boldsymbol{C}\left(t_{N+1}\right)=\sum_{n=1}^{N} \mathcal{T}_{n} \cdot\left(1-\exp \left(-\sigma_{n} \delta_{n}\right)\right) \cdot \mathbf{c}_{n}, \quad \text { where } \quad \mathcal{T}_{n}=\exp \left(\sum_{k=1}^{n-1}-\sigma_{k} \delta_{k}\right) \tag{24}
\end{equation*}
$$

Finally, rather than writing these expressions in terms of volumetric density, we can re-express them in terms of alpha-compositing weights $\alpha_{n} \equiv 1-\exp \left(-\sigma_{n} \delta_{n}\right)$, and by noting that $\prod_{i} \exp x_{i}=\exp \left(\sum_{i} x_{i}\right)$ in (19):

$$
\begin{equation*}
\boldsymbol{C}\left(t_{N+1}\right)=\sum_{n=1}^{N} \mathcal{T}_{n} \cdot \alpha_{n} \cdot \mathbf{c}_{n}, \quad \text { where } \quad \mathcal{T}_{n}=\prod_{n=1}^{N-1}\left(1-\alpha_{n}\right) \tag{25}
\end{equation*}
$$

Alternate derivation. By making use of the earlier connection between CDF and PDF that $(1-\mathcal{T})^{\prime}=\mathcal{T} \sigma$, and by assuming constant color $\mathbf{c}_{a}$ along an interval $[a, b]$ :

$$
\begin{align*}
\int_{a}^{b} \mathcal{T}(t) \cdot \sigma(t) \cdot \mathbf{c}(t) d t & =\mathbf{c}_{a} \int_{a}^{b}(1-\mathcal{T})^{\prime}(t) d t  \tag{26}\\
& =\left.\mathbf{c}_{a} \cdot(1-\mathcal{T}(t))\right|_{a} ^{b}  \tag{27}\\
& =\mathbf{c}_{a} \cdot(\mathcal{T}(a)-\mathcal{T}(b))  \tag{28}\\
& =\mathbf{c}_{a} \cdot \mathcal{T}(a) \cdot(1-\mathcal{T}(a \rightarrow b)) \tag{29}
\end{align*}
$$

Combined with constant per-interval density, this identity yields the same expression for color as (24).

## References

[1] Nelson Max and Min Chen. Local and global illumination in the volume rendering integral. Technical report, Lawrence Livermore National Lab (LLNL), Livermore, CA (United States), 2005.
[2] Nelson Max and Min Chen. Local and global illumination in the volume rendering integral. Technical report, Schloss Dagstuhl, Leibniz Center for Informatics (Germany), 2010. https://drops.dagstuhl.de/opus/volltexte/2010/2709/pdf/18.pdf.
[3] Ben Mildenhall, Pratul P. Srinivasan, Matthew Tancik, Jonathan T. Barron, Ravi Ramamoorthi, and Ren Ng. NeRF: Representing scenes as neural radiance fields for view synthesis. In ECCV, 2020.
[4] Matt Pharr, Wenzel Jakob, and Greg Humphreys. Physically based rendering: From theory to implementation. Morgan Kaufmann, 2016. https://pbr-book.org/3ed-2018/Volume_Scattering/Volume_Scattering_Processes.

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[^0]:    ${ }^{1}$ If you are interested in borrowing the LaTeX notation, please refer to: https://www.overleaf.com/read/fkhpkzxhnyws

