Importance Sampling Techniques for Path Tracing in Participating Media

Christopher Kulla Marcos Fajardo











Requirements

- Area Lights
- Global illumination
- ▶ Motion Blur of camera, lights, surfaces and volumes
- Procedural Shaders
- Minimal memory usage
- Progressive rendering
- Artist friendly quality controls

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}\left(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}\right) + \sigma_{s}\left(\mathbf{x}_{t}\right) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

Nested integral for incoming light

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

- Nested integral for incoming light
- Nested integral for heterogeneous media

$$\underline{L}(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) \underline{L}(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

- Nested integral for incoming light
- Nested integral for heterogeneous media
- Recursive integral for multiple scattering

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}\left(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}\right) + \sigma_{s}\left(\mathbf{x}_{t}\right) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

- Nested integral for incoming light
- Nested integral for heterogeneous media
- Recursive integral for multiple scattering

Our approach

New importance sampling techniques for each problem

Importance Sampling Incoming Light

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$

Homogenous media ▼

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{-\sigma}^{\sigma} e^{-\sigma_t t} \left(\sigma_s \int_{S^2} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_t, \overrightarrow{\mathbf{v}}) \, d\overrightarrow{\mathbf{v}} \right) \, dt$

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_{0}^{\infty} e^{-\sigma_t t} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{x}_t c}) L(\mathbf{x}_t, \overrightarrow{\mathbf{x}_t c}) dt$

Integrate point light contribution along the ray

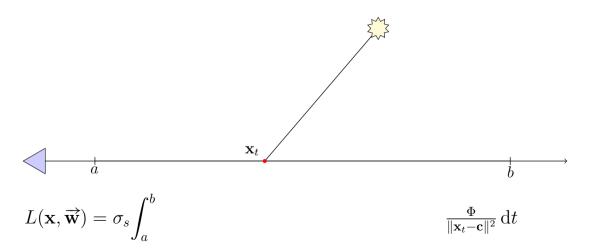




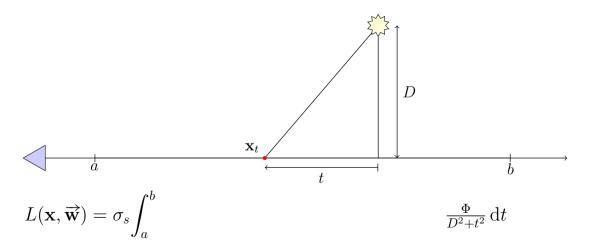
$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b$$

 $\mathrm{d}t$

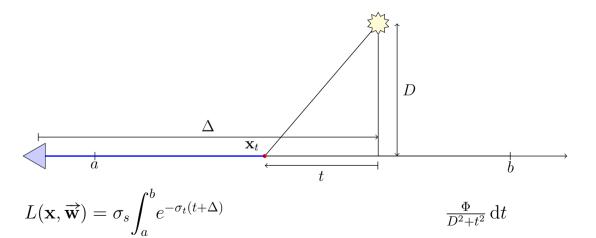
Radiance reaching the ray varies as $1/r^2$



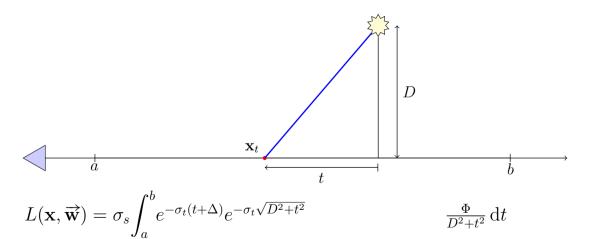
Express in terms of t



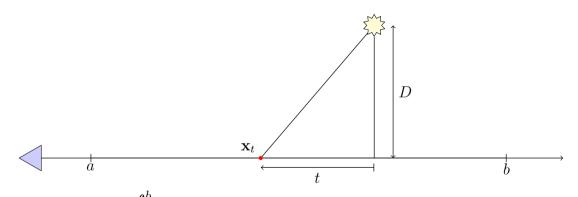
Account for extinction up to sample point



Add extinction towards the light

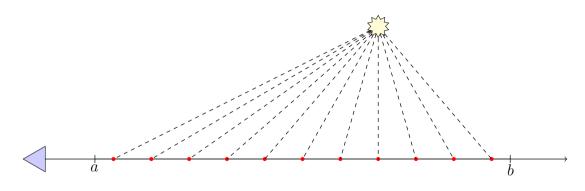


Finally add phase function and visibility



$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_{a}^{b} e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{x}_t \mathbf{c}}) V(\mathbf{x}_t, \mathbf{c}) \frac{\Phi}{D^2+t^2} dt$$

What is the best sample distribution for Monte Carlo integration?



$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_{a}^{b} e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{x}_t \mathbf{c}}) V(\mathbf{x}_t, \mathbf{c}) \frac{\Phi}{D^2+t^2} dt$$

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

Only two terms can be easily integrated and inverted

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

Only two terms can be easily integrated and inverted

Transmission $e^{-\sigma_t t}$



$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

Only two terms can be easily integrated and inverted

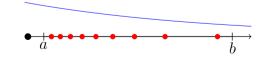
Transmission $e^{-\sigma_t t}$



$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

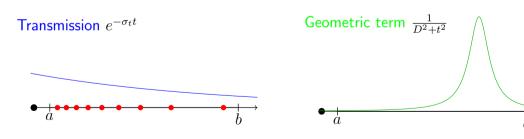
Only two terms can be easily integrated and inverted

Transmission $e^{-\sigma_t t}$

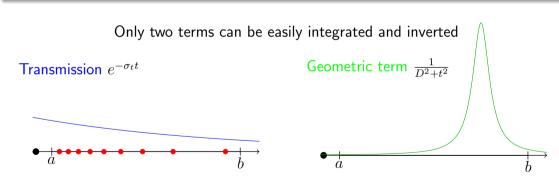


$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

Only two terms can be easily integrated and inverted



$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c}, \mathbf{x}_t\right) \frac{\Phi}{D^2 + t^2} dt$$

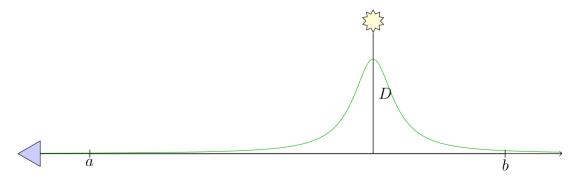


$$L(\mathbf{x},\overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}},\overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c},\mathbf{x}_t\right) \frac{\Phi}{D^2+t^2} \, \mathrm{d}t$$
 Only two terms can be easily integrated and inverted
Transmission $e^{-\sigma_t t}$ Geometric term $\frac{1}{D^2+t^2}$

$$L(\mathbf{x},\overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t(t+\Delta)} e^{-\sigma_t\sqrt{D^2+t^2}} \rho\left(\overrightarrow{\mathbf{w}},\overrightarrow{\mathbf{v}_t}\right) V\left(\mathbf{c},\mathbf{x}_t\right) \frac{\Phi}{D^2+t^2} \, \mathrm{d}t$$
 Only two terms can be easily integrated and inverted
Transmission $e^{-\sigma_t t}$ Geometric term $\frac{1}{D^2+t^2}$

Goal is to get a pdf proportional to geometric term:

$$\mathsf{pdf}(t) \propto rac{1}{D^2 + t^2}$$

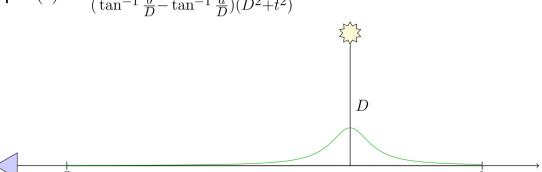


Integrate pdf to obtain cdf:

$$\mathsf{cdf}(t) = \int \frac{1}{D^2 + t^2} \, dt = \frac{1}{D} \tan^{-1} \frac{t}{D}$$

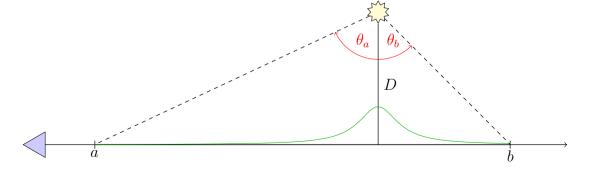
Use cdf to normalize over [a, b]:

$$pdf(t) = \frac{D}{(\tan^{-1}\frac{b}{D} - \tan^{-1}\frac{a}{D})(D^2 + t^2)}$$



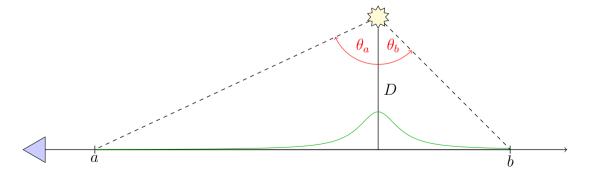
Use cdf to normalize over [a, b]:

$$\mathsf{pdf}(t) = \frac{D}{(\theta_b - \theta_a)(D^2 + t^2)}$$



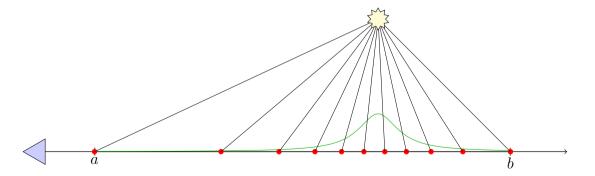
Invert cdf to obtain distribution for $\xi_i \in [0, 1)$:

$$t_i = D \tan ((1 - \xi_i) \theta_a + \xi_i \theta_b)$$

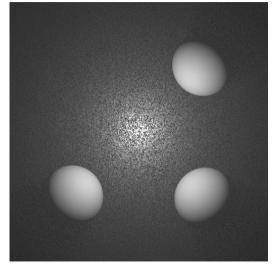


Sample distribution is equi-angular

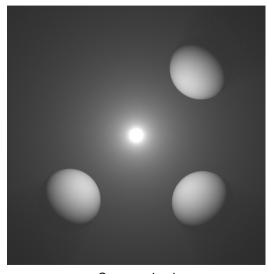
$$t_i = D \tan ((1 - \xi_i) \theta_a + \xi_i \theta_b)$$



Results with 16 samples/pixel

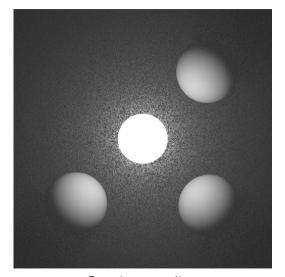


Density sampling



Our method

Sphere lights can use same equations! $(\Omega \propto 1/r^2)$



Density sampling

Our method

Importance sampling for arbitrary area lights

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_0^b e^{-\sigma_t t} e^{-\sigma_t \|\mathbf{x}_t - \mathbf{c}\|} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}_t}) V(\mathbf{c}, \mathbf{x}_t) \frac{\Phi}{\|\mathbf{x}_t - \mathbf{c}\|^2} dt$$

Area light ▼

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \sigma_s \int_a^b e^{-\sigma_t t} \int_A \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{x}_t \mathbf{y}}) e^{-\sigma_t \|\mathbf{x}_t - \mathbf{y}\|} \frac{\cos \theta_{\mathbf{y}}}{\|\mathbf{x}_t - \mathbf{y}\|^2} V(\mathbf{x}_t, \mathbf{y}) L_e(\mathbf{y}, \overrightarrow{\mathbf{y}} \overrightarrow{\mathbf{x}_t}) \, dA(\mathbf{y}) \, dt$$

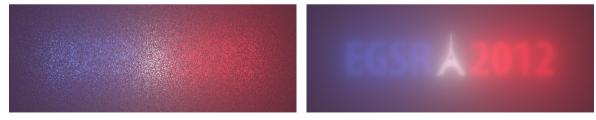
► Same two choices for importance sampling

- ► Same two choices for importance sampling
 - ▶ Transmission $e^{-\sigma_t t}$



Density sampling

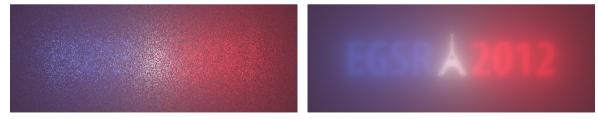
- ► Same two choices for importance sampling
 - ▶ Transmission $e^{-\sigma_t t}$
 - Geometric term $1/\|\mathbf{x}_t \mathbf{y}\|^2$



Density sampling

Equi-angular sampling

- Same two choices for importance sampling
 - ▶ Transmission $e^{-\sigma_t t}$
 - Geometric term $1/\|\mathbf{x}_t \mathbf{y}\|^2$
- lacktriangle To sample geometric term, choose f y before $f x_t$



Density sampling

Equi-angular sampling

- Same two choices for importance sampling
 - ▶ Transmission $e^{-\sigma_t t}$
 - Geometric term $1/\|\mathbf{x}_t \mathbf{y}\|^2$

Density sampling

- \triangleright To sample geometric term, choose y before \mathbf{x}_t
- ► Can combine both line sampling strategies by MIS
- ▶ More details in paper about using the phase function



Equi-angular sampling

Importance Sampling for Heterogeneous Media

Importance sampling in heterogeneous media

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

► Transmission term contains an integral

Importance sampling in heterogeneous media

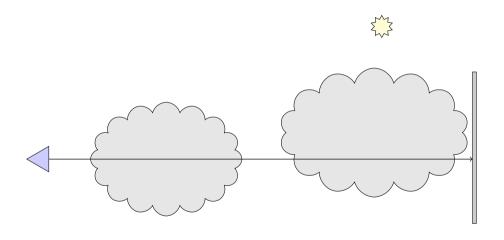
$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}\left(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}\right) + \sigma_{s}\left(\mathbf{x}_{t}\right) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

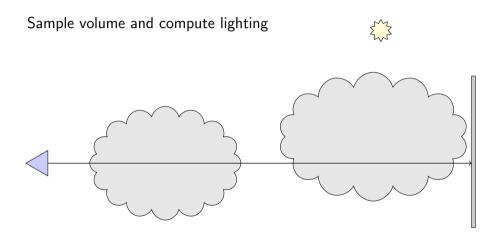
- Transmission term contains an integral
- Integral limit is also integration variable

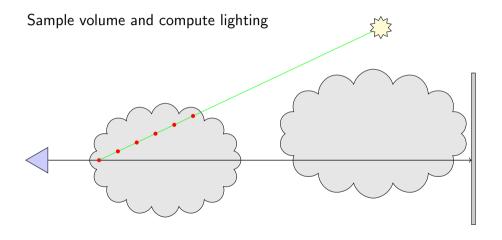
Importance sampling in heterogeneous media

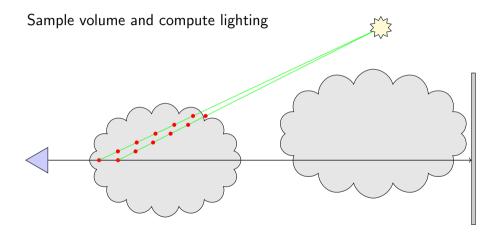
$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{S^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

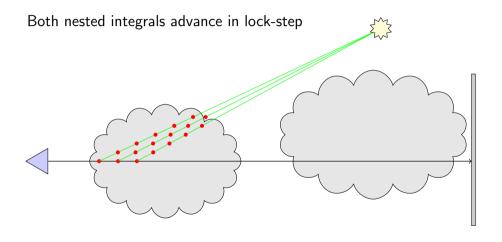
- ▶ Transmission term contains an integral
- Integral limit is also integration variable
- Previous method assumed we could take samples at any t

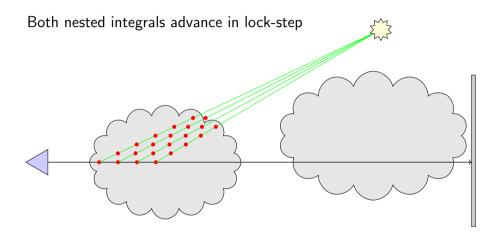


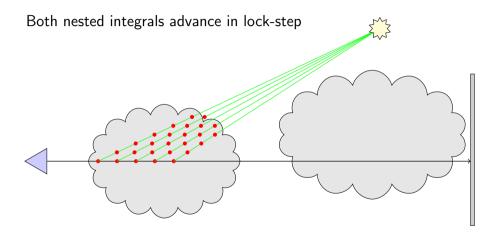


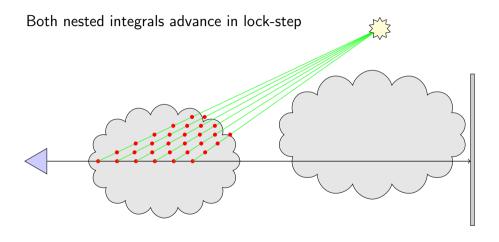


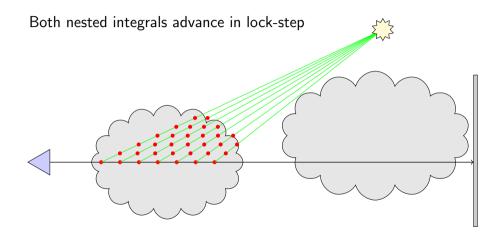


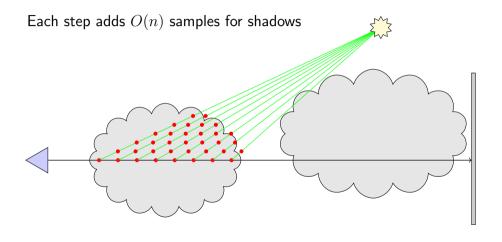


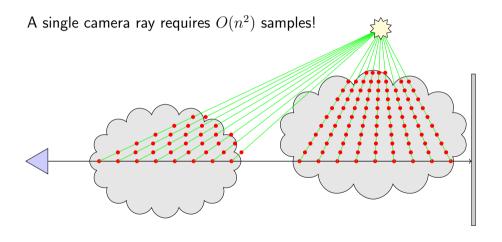


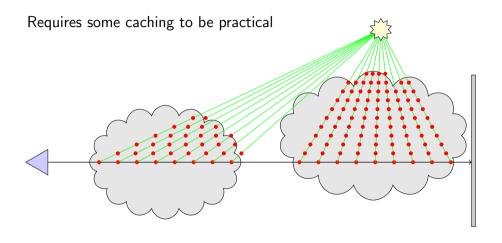


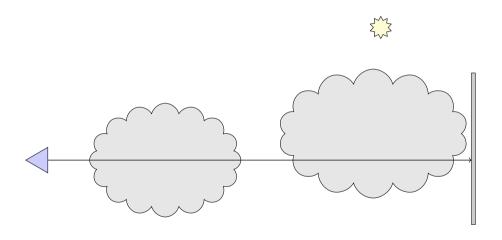


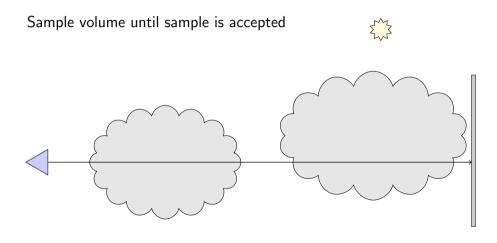


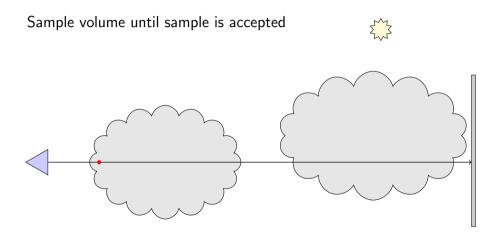


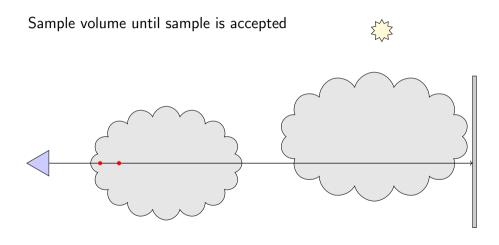


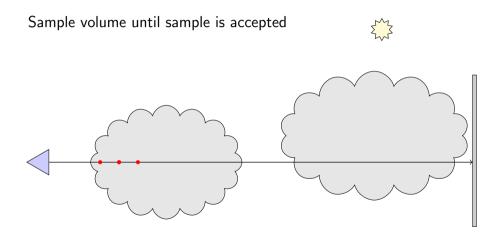


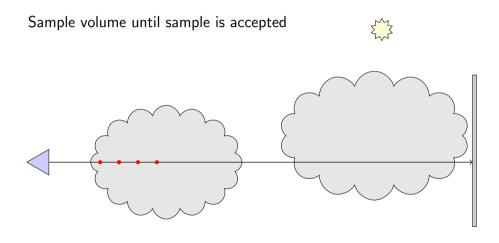


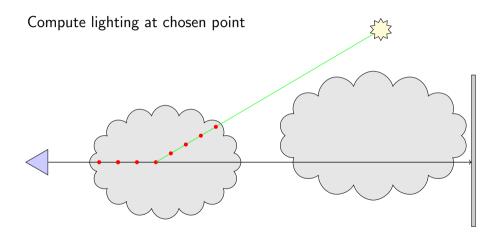


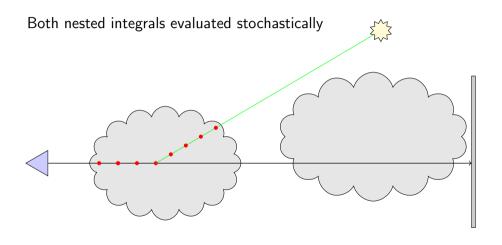


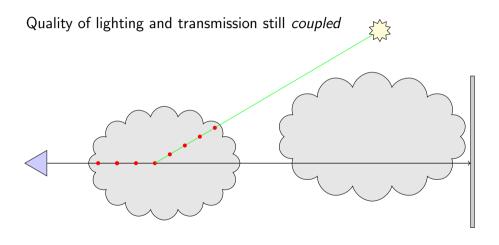


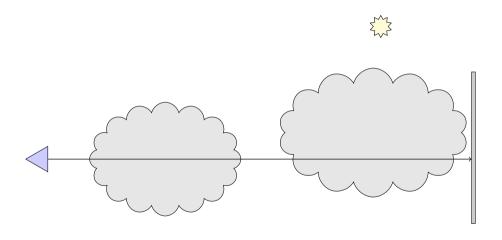


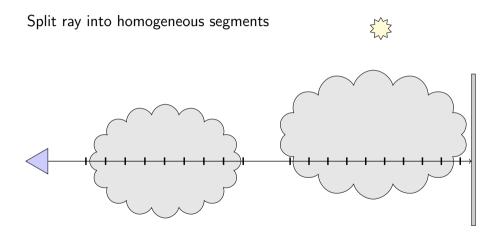








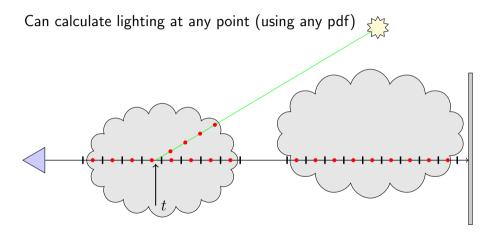




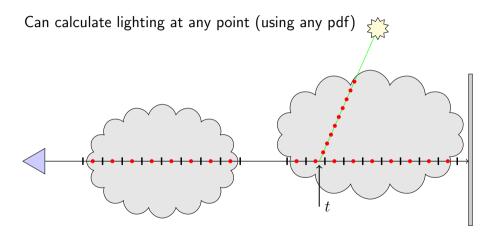
Run shader once per segment (front to back)

Store
$$\sigma_{s_i}, \sigma_{t_i}, T_i = T_{i-1}e^{-\sigma_{t_{i-1}}\Delta_{i-1}}$$

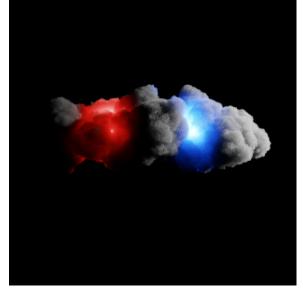
Given any t, locate segment by binary search

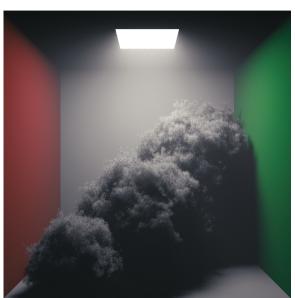


Our approach - Decoupled Ray Marching



Equi-angular sampling in heterogeneous media

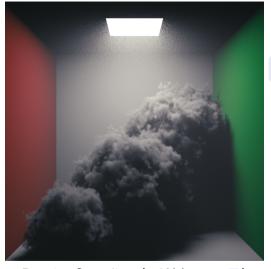






Equi-angular Sampling $(pdf(t) \propto 1/r^2)$

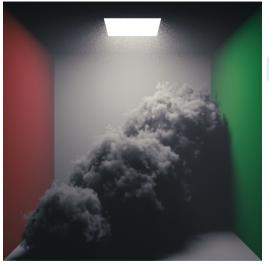
 Equi-angular sampling not optimal everywhere



Density Sampling $(pdf(t) \propto \sigma_{s_i}T_i)$

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_a^b \sigma_s(\mathbf{x}_t) e^{-\int_0^t \sigma_t(\mathbf{x}_s) \, \mathrm{d}s} \left(\int_{S^2} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_t, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

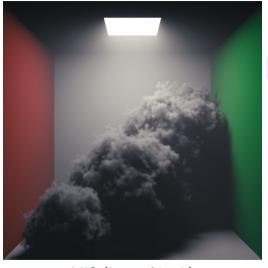
- Equi-angular sampling not optimal everywhere
- Build discrete pdf from ray marched samples



Density Sampling $(pdf(t) \propto \sigma_{s_i}T_i)$

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_a^b \sigma_s(\mathbf{x}_t) e^{-\int_0^t \sigma_t(\mathbf{x}_s) \, \mathrm{d}s} \left(\int_{S^2} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_t, \overrightarrow{\mathbf{v}}) \, \mathrm{d}\overrightarrow{\mathbf{v}} \right) \, \mathrm{d}t$$

- Equi-angular sampling not optimal everywhere
- Build discrete pdf from ray marched samples
- ▶ Samples are focused where $\sigma_s > 0$



MIS (best of both)

$$L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{a}^{b} \sigma_{s}(\mathbf{x}_{t}) e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) ds} \left(\int_{S^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) d\overrightarrow{\mathbf{v}} \right) dt$$

- Equi-angular sampling not optimal everywhere
- Build discrete pdf from ray marched samples
- Samples are focused where $\sigma_s > 0$
- Can combine multiple pdfs by MIS

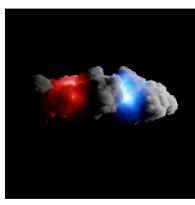
Results - Decoupled Ray Marching



Equi-angular Sampling $(\mathrm{pdf}(t) \propto 1/r^2)$



Density Sampling $(pdf(t) \propto \sigma_{s_i}T_i)$



MIS (best of both)

Multiple Scattering



Emissive term ▼

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{a}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, \mathrm{d}s} L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) \, \mathrm{d}t$

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) \, ds} \left(L_{e}(\mathbf{x}_{t}, \overrightarrow{\mathbf{w}}) + \sigma_{s}(\mathbf{x}_{t}) \int_{\mathbb{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) \, d\overrightarrow{\mathbf{v}} \right) \, dt$

Scattering term ▼

 $L(\mathbf{x}, \overrightarrow{\mathbf{w}}) = \int_{0}^{b} \sigma_{s}(\mathbf{x}_{t}) e^{-\int_{0}^{t} \sigma_{t}(\mathbf{x}_{s}) ds} \int_{\mathbf{S}^{2}} \rho(\overrightarrow{\mathbf{w}}, \overrightarrow{\mathbf{v}}) L(\mathbf{x}_{t}, \overrightarrow{\mathbf{v}}) d\overrightarrow{\mathbf{v}} dt$





Multiple Scattering

- ▶ Use discrete pdf to generate sampling locations for path tracing
- ▶ Works well for 1-2 bounces



No bounce (2m15s) 1 Bounce (9m03s)



2 Bounces (18m14s)

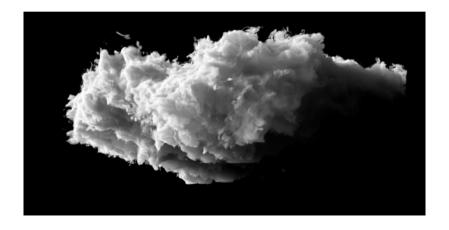
Multiple Scattering Approximation

▶ High albedo media require many bounces



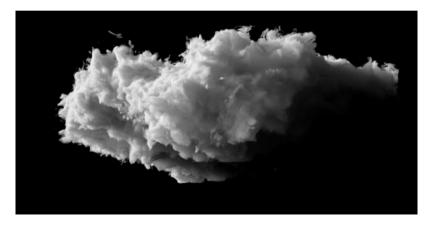
Multiple Scattering Approximation

- ▶ High albedo media require many bounces
- ► Approximate higher order bounces by changing density along shadow rays



Multiple Scattering Approximation

- ▶ High albedo media require many bounces
- ► Approximate higher order bounces by changing density along shadow rays
- ► Combine with one real bounce to achieve more natural diffusion



Implementation Details

- Ray tracer reports interval hits with volume primitives
- Motion blurred media, lights, camera handled automatically
- Only two parameters exposed: step size and light samples (very intuitive for artists)
- Can unify handling of transparent objects into the discrete pdf (hair in volumes)

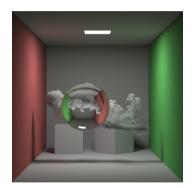
Results

Videos

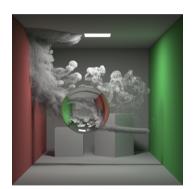
Future Work

- ► Generalize to bi-directional methods
- ▶ Improve primitive bounds for motion blur
- ► GPU implementation

Thanks for listening!







Questions?