



**SIGGRAPH** THINK  
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# PATH-SPACE DIFFERENTIABLE RENDERING

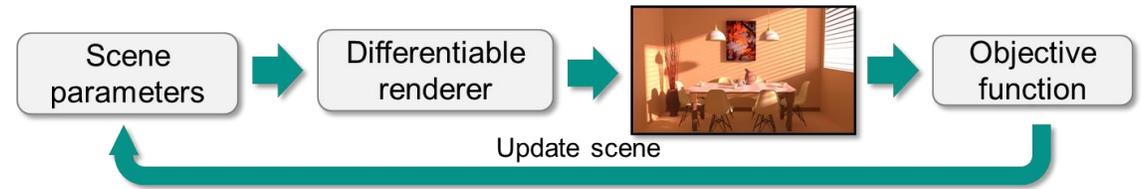
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Ioannis Gkioulekas<sup>2</sup>, Shuang Zhao<sup>1</sup>

<sup>1</sup>University of California, Irvine

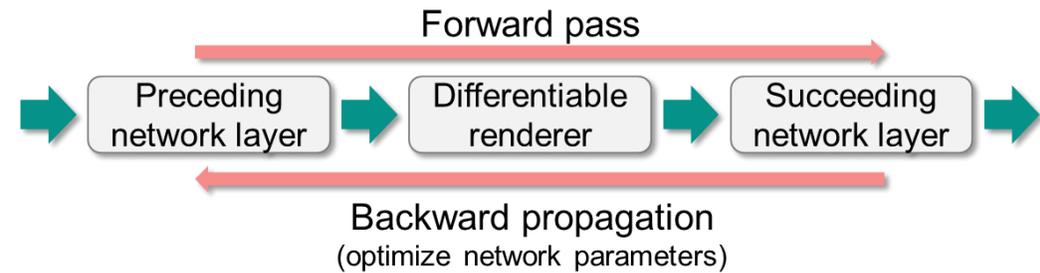
<sup>2</sup>Carnegie Mellon University

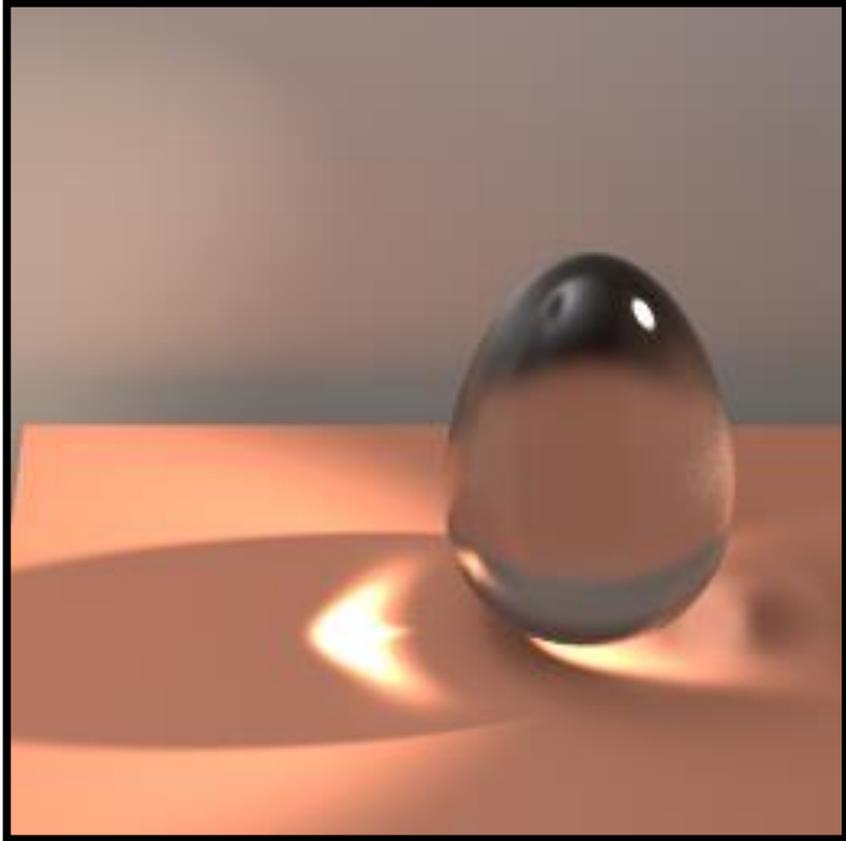
# WHY WE NEED DIFF. RENDERING?

- Inverse rendering
  - Enabling gradient-based optimization

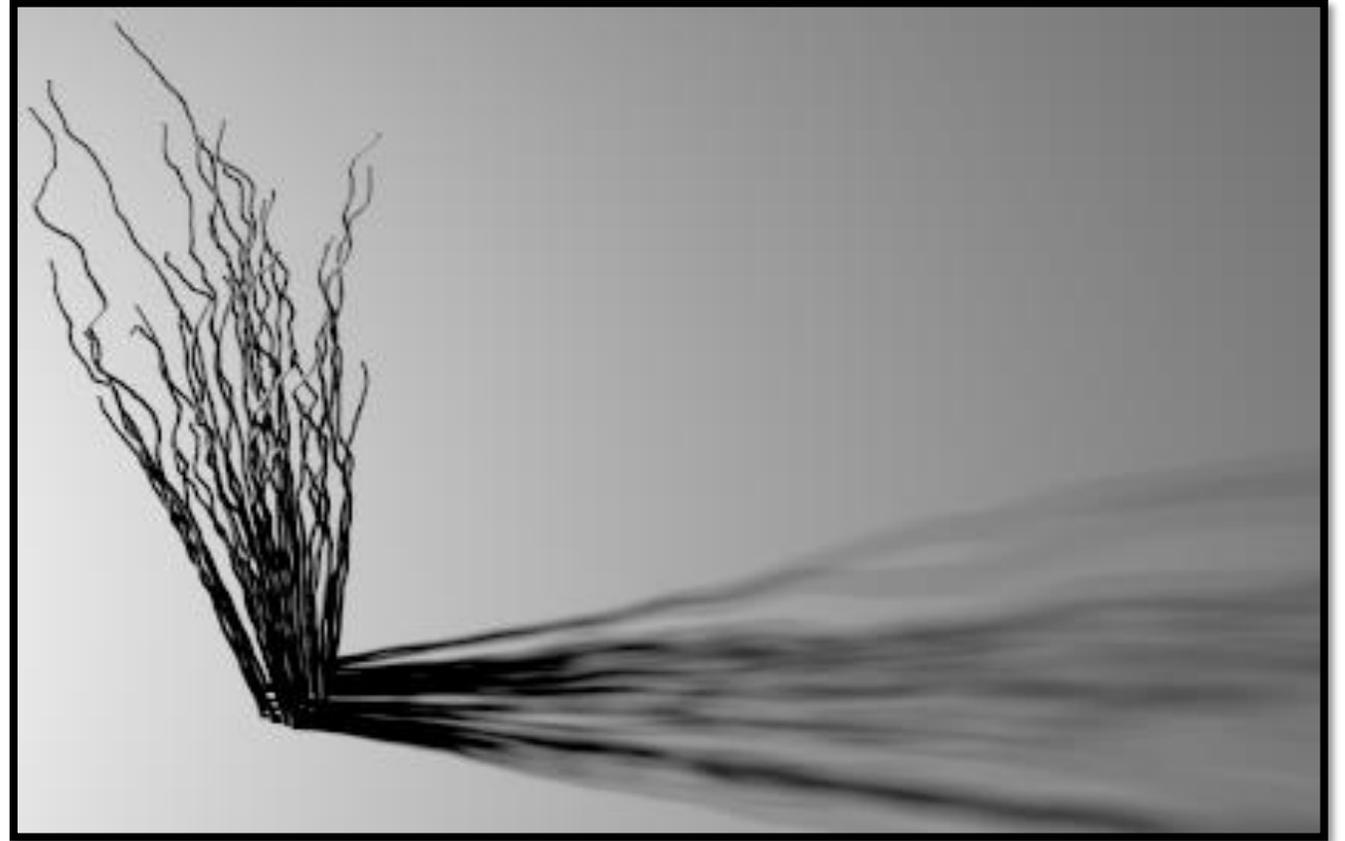


- Machine learning
  - Backpropagation through rendering





Complex light transport effects



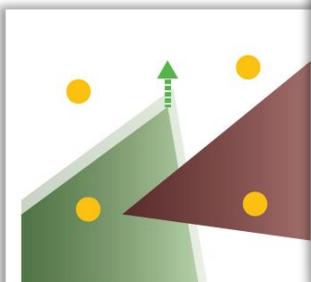
Complex geometry

Li et al. 2018

Zhang et al. 2019

Loubet et al. 2019

## Our method addresses both challenges efficiently!



(a) area sampling

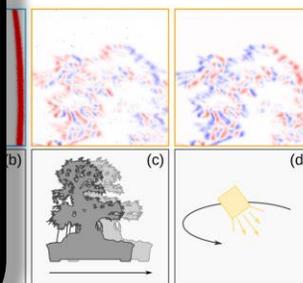
(b) edge sampling

(a) Initial

(b) Est. density

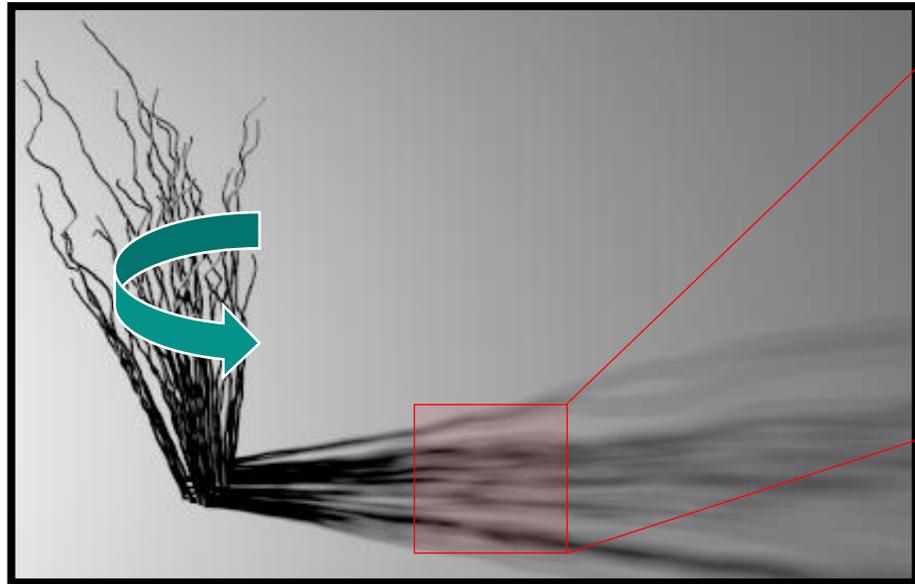
A scene with complex geometry and visibility (T&M triangles)

Gradients with respect to scene parameters that affect visibility

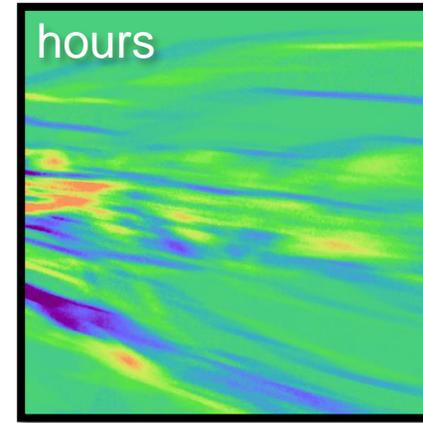
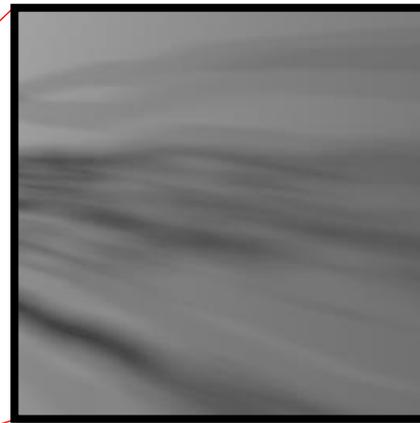


- Handling **complex geometry**
  - [Li et al. 2018, Zhang et al. 2019] Expensive silhouette detection
  - [Loubet et al. 2019] Biased approximation
- Handling **complex light transport effects**
  - All previous methods Unidirectional path tracing only

# PREVIEW OF OUR RESULTS



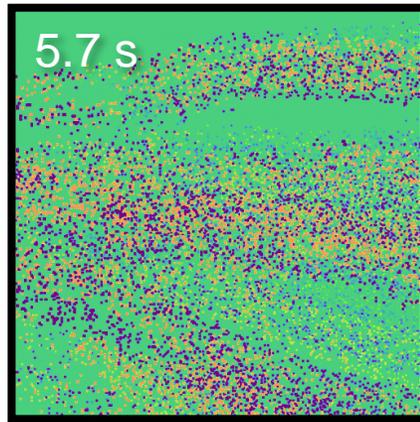
Complex geometry



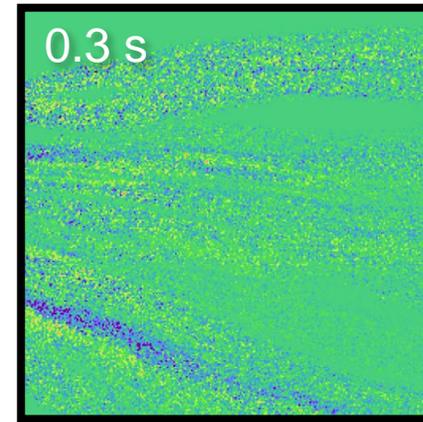
Reference



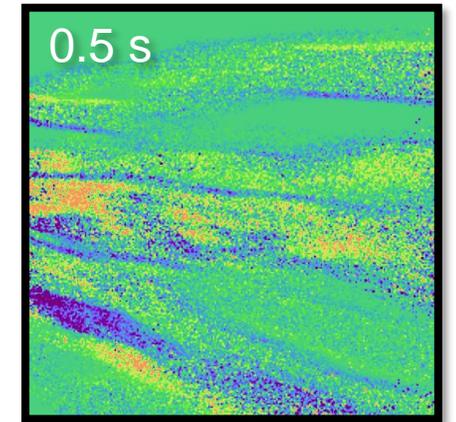
**Equal-sample comparison**



[Zhang et al. 2019]

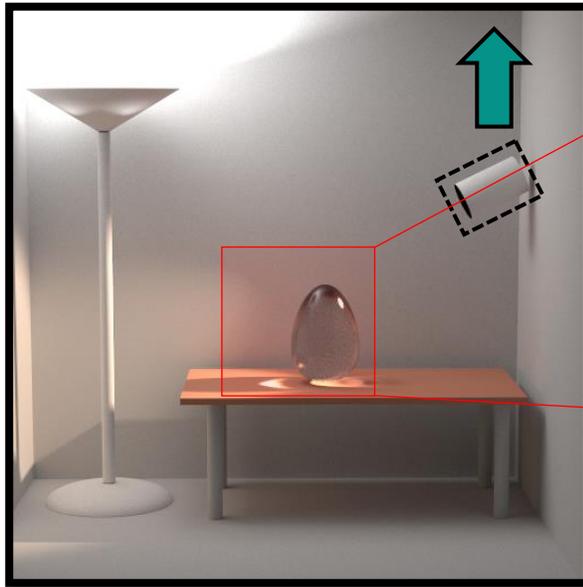


[Loubet et al. 2019]

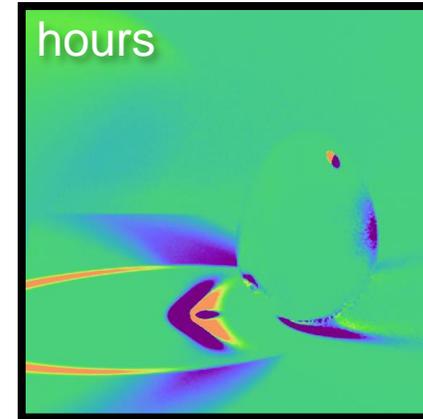
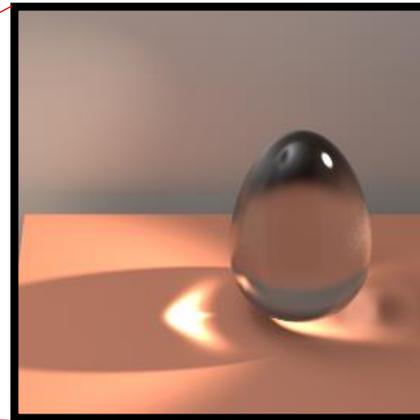


**Ours**

# PREVIEW OF OUR RESULTS



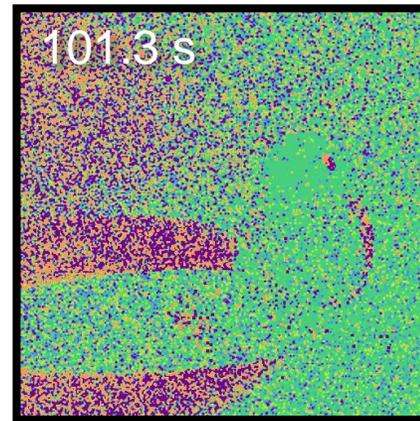
Complex light transport effects



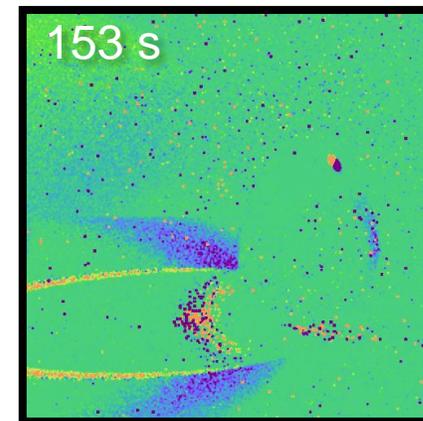
Reference



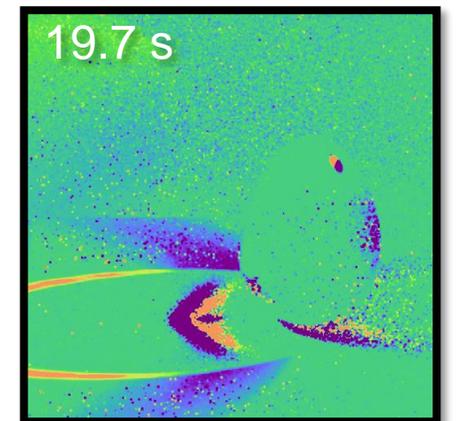
**Equal-sample  
comparison**



[Zhang et al. 2019]



[Loubet et al. 2019]



**Ours**

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# PRELIMINARIES

# REYNOLDS TRANSPORT THEOREM

$$\frac{d}{d\pi} \int_{\Omega} f dA \stackrel{?}{=} \int_{\Omega} \frac{df}{d\pi} dA + \int_{\partial\Omega} g dl$$

**Reynolds transport theorem**  
Generalization of Leibniz's rule

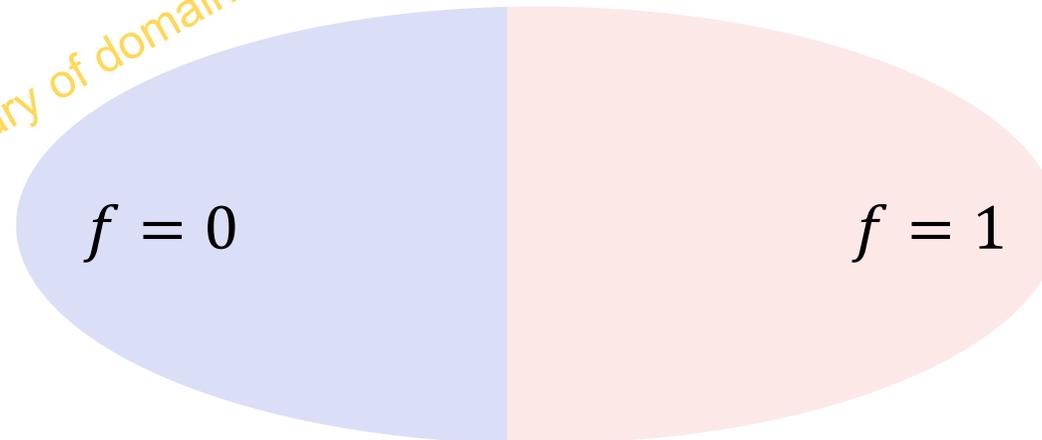
Interior integral

Boundary domain

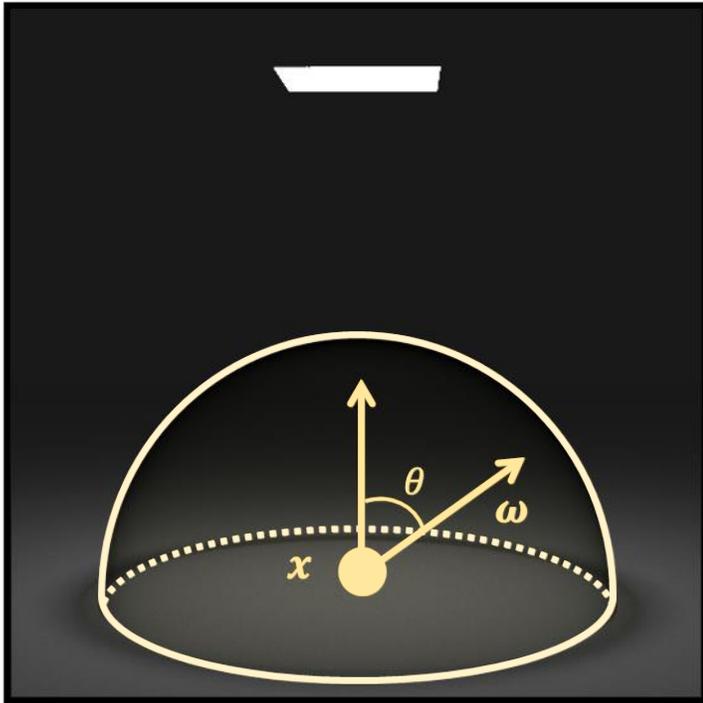
boundary integral

discontinuity points  $\cup$  boundary of domain  $\Omega$

boundary of domain  $\Omega$



$\pi$ : size of the emitter



Irradiance at  $x$

$$E = \int_{\mathbb{H}^2} L_i(\boldsymbol{\omega}) \cos\theta \, d\sigma(\boldsymbol{\omega})$$

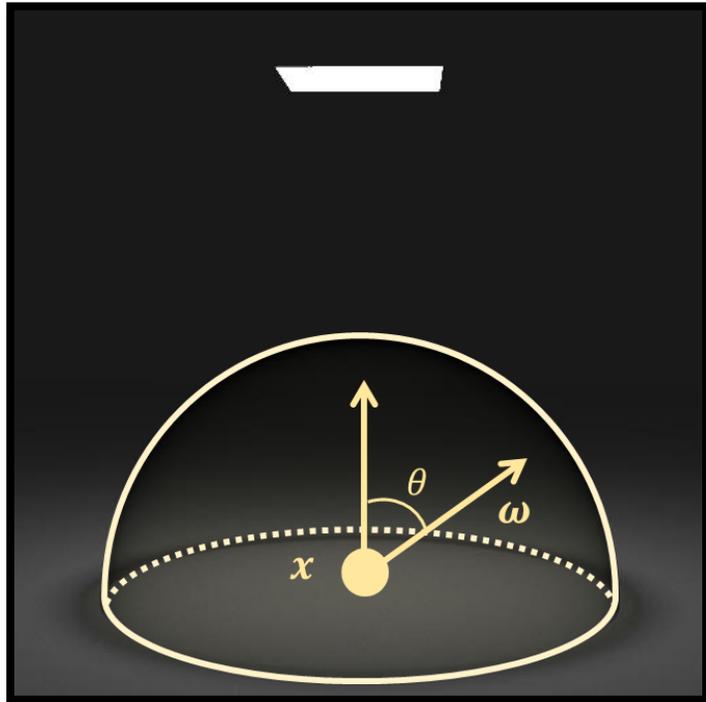
Unit hemisphere

Differential irradiance at  $x$

$$\frac{dE}{d\pi} = \frac{d}{d\pi} \int_{\mathbb{H}^2} L_i(\boldsymbol{\omega}) \cos\theta \, d\sigma(\boldsymbol{\omega})$$

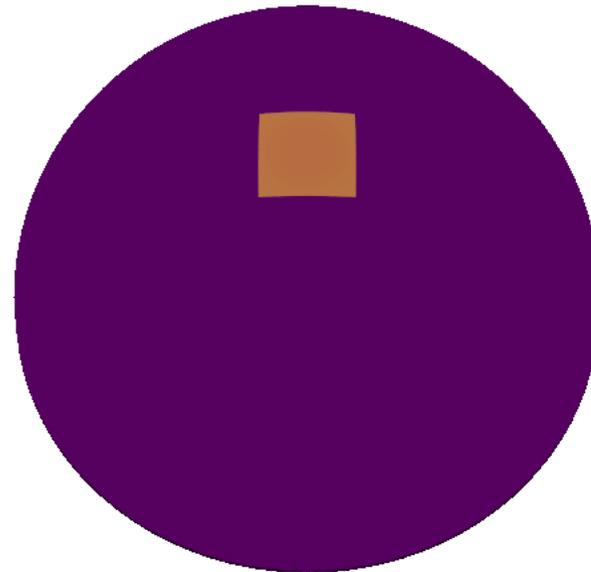
# REYNOLDS TRANSPORT THEOREM

$\pi$ : size of the emitter

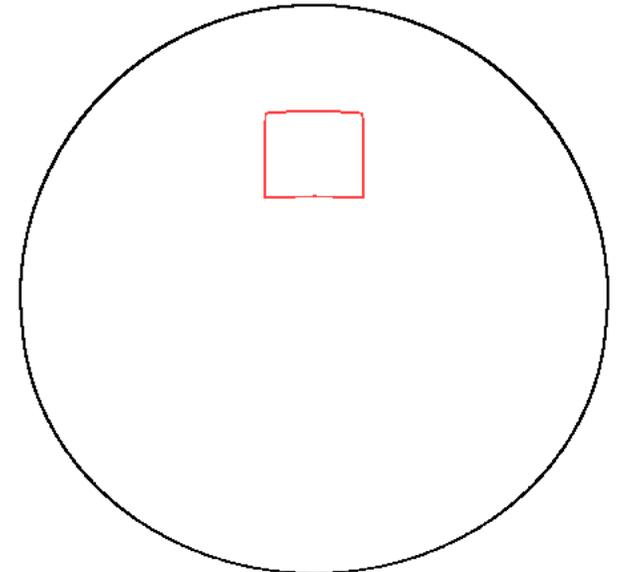


$$E = \int_{\mathbb{H}^2} L_i(\omega) \cos\theta \, d\sigma(\omega)$$

Low  High



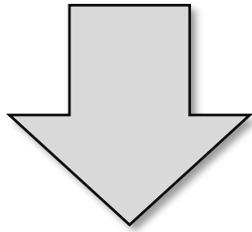
The integrand



Discontinuous points  
( $\pi$ -dependent)

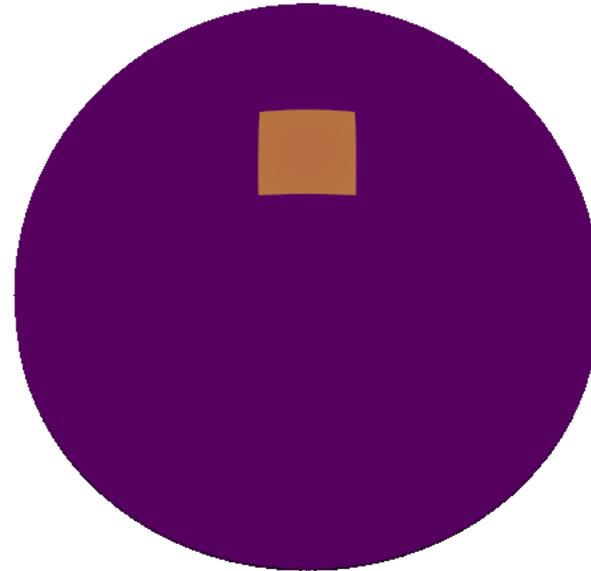
# REYNOLDS TRANSPORT THEOREM

$$E = \int_{\mathbb{H}^2} \overbrace{L_i(\omega) \cos\theta}^f d\sigma(\omega)$$

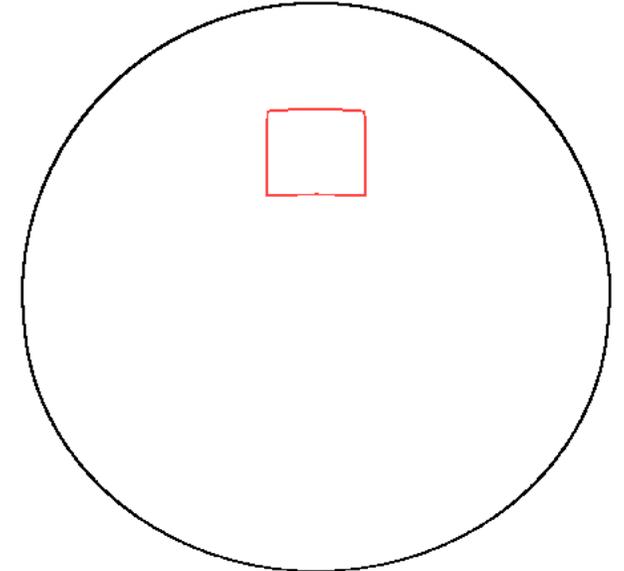


$$\frac{dE}{d\pi} = \int_{\mathbb{H}^2} \underbrace{\frac{df}{d\pi}}_{=0} d\sigma + \underbrace{\int_{\partial\mathbb{H}^2} g dl}_{\neq 0}$$

Low  High



The integrand



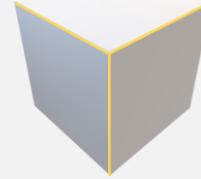
Discontinuous points  
( $\pi$ -dependent)

# SOURCE OF DISCONTINUITIES

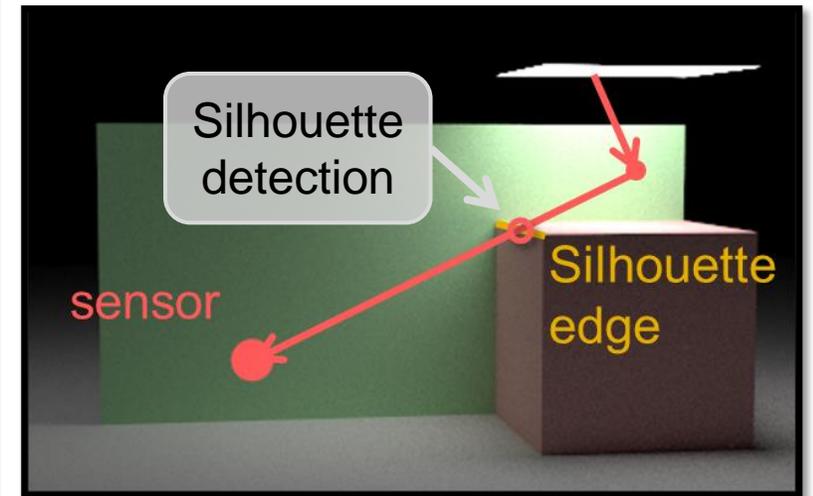
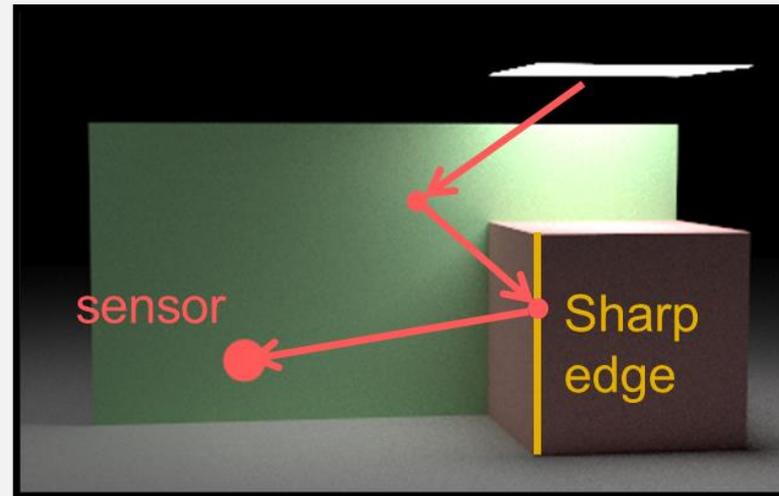
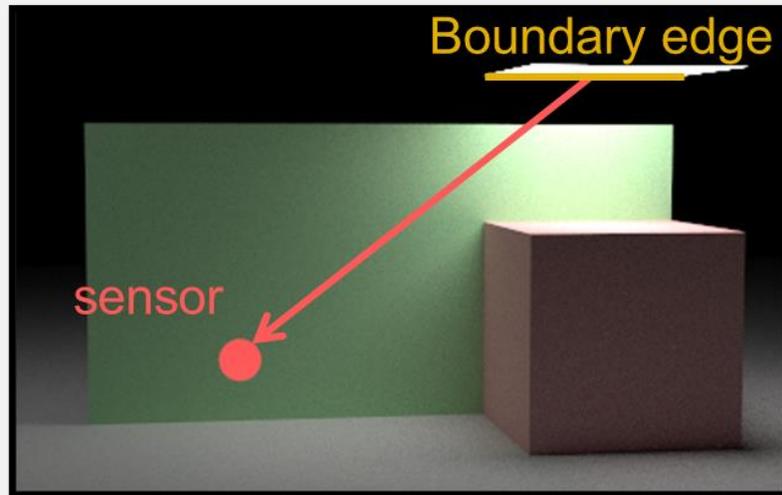
Boundary edge



Sharp edge



Silhouette edge



Topology-driven

Visibility-driven

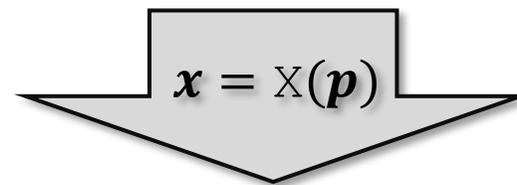
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# OUR TECHNIQUE

$$\frac{d}{d\pi} \int \text{[Diagram of a path integral with points } x_0, x_1, \dots, x_{N-1}, x_N \text{ and a sun icon]} d\bar{x}$$

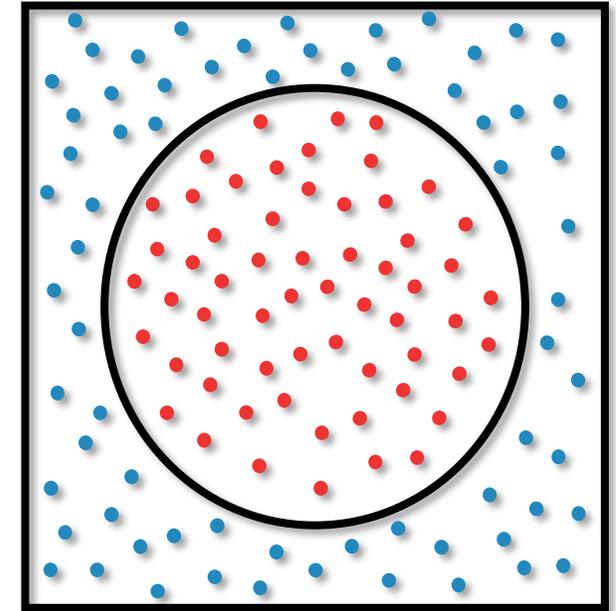
Differential path integral

$$f(\mathbf{x})$$



$$f(X(\mathbf{p}))$$

Reparameterization

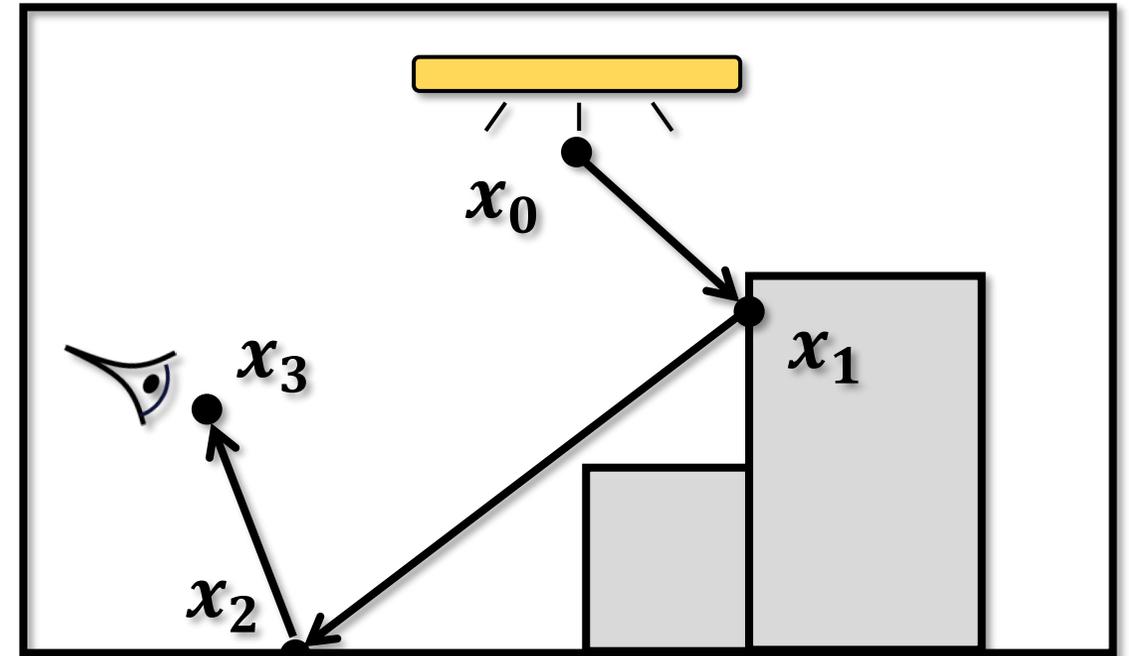


Monte Carlo Estimator

$$I = \int_{\Omega} \overset{\text{Measurement contribution function}}{f(\bar{x})} d \underset{\text{Area-product measure}}{\mu(\bar{x})}$$

Path space

- Introduced by Veach [1997]
- Foundation of sophisticated Monte Carlo algorithms (e.g., BDPT, MCMC rendering)



Light path  $\bar{x} = (x_0, x_1, x_2, x_3)$

Path Integral

A generalization of Reynolds theorem

$$I = \int_{\Omega} f(\bar{\mathbf{x}}) d\mu(\bar{\mathbf{x}}) \quad \longrightarrow \quad \frac{dI}{d\pi} = ?$$

We now derive  $\partial I_N / \partial \pi$  in Eq. (25) using the recursive relations provided by Eqs. (21) and (24). Let

$$h_n^{(0)} := [\prod_{n'=n+1}^N g(\mathbf{x}_{n'}; \mathbf{x}_{n'-2}, \mathbf{x}_{n'-1})] W_e(\mathbf{x}_N \rightarrow \mathbf{x}_{N-1}), \quad (52)$$

$$h_n^{(1)} := \sum_{n'=n+1}^N \kappa(\mathbf{x}_{n'}) V(\mathbf{x}_{n'}), \quad (53)$$

$$\Delta h_{n,n'}^{(0)} := h_n^{(0)} \Delta g(\mathbf{x}_{n'}; \mathbf{x}_{n'-2}, \mathbf{x}_{n'-1}) / g(\mathbf{x}_{n'}; \mathbf{x}_{n'-2}, \mathbf{x}_{n'-1}), \quad (54)$$

for  $0 \leq n < n' \leq N$ . We omit the dependencies of  $h_n^{(0)}$ ,  $h_n^{(1)}$ , and  $\Delta h_{n,n'}^{(0)}$  on  $\mathbf{x}_{n+1}, \dots, \mathbf{x}_N$  for notational convenience.

We now show that, for all  $0 \leq n < N$ , it holds that

$$h_n(\mathbf{x}_n; \mathbf{x}_{n-1}) = \int_{\mathcal{M}^{N-n}} h_n^{(0)} \prod_{n'=n+1}^N dA(\mathbf{x}_{n'}), \quad (55)$$

and

$$\begin{aligned} \dot{h}_n(\mathbf{x}_n; \mathbf{x}_{n-1}) &= \int_{\mathcal{M}^{N-n}} \left[ \left( h_n^{(0)} \right)' - h_n^{(0)} h_n^{(1)} \right] \prod_{\substack{n'=n+1 \\ i \neq n'}}^N dA(\mathbf{x}_{n'}) \\ &+ \sum_{n'=n+1}^N \int \Delta h_{n,n'}^{(0)} V_{\partial \mathcal{M}_{n'}}(\mathbf{x}_{n'}) d\ell(\mathbf{x}_{n'}) \prod_{\substack{n < i \leq N \\ i \neq n'}} dA(\mathbf{x}_i), \end{aligned} \quad (56)$$

where the integral domain of the second term on the right-hand side, which is omitted for notational clarity, is  $\mathcal{M}(\pi)$  for each  $\mathbf{x}_i$  with  $i \neq n'$  and  $\partial \mathcal{M}_{n'}(\pi)$ , which depends on  $\mathbf{x}_{n-1}$ , for  $\mathbf{x}_{n'}$ .

It is easy to verify that Eqs. (55) and (56) hold for  $n = N - 1$ . We now show that, if they hold for some  $0 < n < N$ , then it is also the case for  $n - 1$ . Let  $g_{n-1} := g(\mathbf{x}_n; \mathbf{x}_{n-2}, \mathbf{x}_{n-1})$  for all  $0 < n \leq N$ . Then,

$$\begin{aligned} h_{n-1}(\mathbf{x}_{n-1}; \mathbf{x}_{n-2}) &= \int_{\mathcal{M}} g_{n-1} \int_{\mathcal{M}^{N-n}} h_n^{(0)} \prod_{n'=n+1}^N dA(\mathbf{x}_{n'}) dA(\mathbf{x}_n) \\ &= \int_{\mathcal{M}^{N-n+1}} h_{n-1}^{(0)} \prod_{n'=n}^N dA(\mathbf{x}_{n'}), \end{aligned} \quad (57)$$

and

$$\begin{aligned} \dot{h}_{n-1}(\mathbf{x}_{n-1}; \mathbf{x}_{n-2}) &= \int_{\mathcal{M}} \left[ \dot{g}_{n-1} h_n + g_{n-1} (\dot{h}_n - h_n \kappa(\mathbf{x}_n) V(\mathbf{x}_n)) \right] dA(\mathbf{x}_n) \\ &+ \int_{\partial \mathcal{M}_n} \Delta g_{n-1} h_n V_{\partial \mathcal{M}_n} d\ell(\mathbf{x}_n) \\ &= \int_{\mathcal{M}^{N-n+1}} \left\{ \dot{g}_{n-1} h_n^{(0)} + g_{n-1} \left[ \left( h_n^{(0)} \right)' - h_n^{(0)} h_n^{(1)} \right] \right\} \prod_{n'=k}^N dA(\mathbf{x}_{n'}) \\ &+ \sum_{n'=n+1}^N \int g_{n-1} \Delta h_{n,n'}^{(0)} V_{\partial \mathcal{M}_{n'}}(\mathbf{x}_{n'}) d\ell(\mathbf{x}_{n'}) \prod_{\substack{n \leq i \leq N \\ i \neq n'}} dA(\mathbf{x}_i) \\ &+ \int \Delta g_{n-1} h_n^{(0)} V_{\partial \mathcal{M}_n} d\ell(\mathbf{x}_n) \prod_{n'=n+1}^N dA(\mathbf{x}_{n'}) \\ &= \int_{\mathcal{M}^{N-n+1}} \left[ \left( h_{n-1}^{(0)} \right)' - h_{n-1}^{(0)} h_{n-1}^{(1)} \right] \prod_{n'=n}^N dA(\mathbf{x}_{n'}) \\ &+ \sum_{n'=n}^N \int \Delta h_{n-1,n'}^{(0)} V_{\partial \mathcal{M}_{n'}}(\mathbf{x}_{n'}) d\ell(\mathbf{x}_{n'}) \prod_{\substack{n \leq i \leq N \\ i \neq n'}} dA(\mathbf{x}_i). \end{aligned} \quad (58)$$

Thus, using mathematical induction, we know that Eqs. (55) and (56) hold for all  $0 \leq n < N$ .

Notice that  $h_0^{(0)} = f$  and  $\Delta h_{0,n'}^{(0)} = \Delta f_{n'}$ , where  $\Delta f_{n'}$  follows the definition in Eq. (28). Letting  $n = 0$  in Eq. (56) yields

$$\begin{aligned} \dot{h}_0(\mathbf{x}_0) &= \int_{\mathcal{M}^N} \left[ \dot{f}(\bar{\mathbf{x}}) - f(\bar{\mathbf{x}}) \sum_{n'=1}^N \kappa(\mathbf{x}_{n'}) V(\mathbf{x}_{n'}) \right] \prod_{n'=1}^N dA(\mathbf{x}_{n'}) \\ &+ \sum_{n'=1}^N \int \Delta f_{n'}(\bar{\mathbf{x}}) V_{\partial \mathcal{M}_{n'}} d\ell(\mathbf{x}_{n'}) \prod_{\substack{0 < i \leq N \\ i \neq n'}} dA(\mathbf{x}_i). \end{aligned} \quad (59)$$

Lastly, based on the assumption that  $h_0$  is continuous in  $\mathbf{x}_0$ , Eq. (25) can be obtained by differentiating Eq. (23):

$$\begin{aligned} \frac{\partial I_N}{\partial \pi} &= \frac{\partial}{\partial \pi} \int_{\mathcal{M}} h_0(\mathbf{x}_0) dA(\mathbf{x}_0) \\ &= \int_{\mathcal{M}} \left[ \dot{h}_0(\mathbf{x}_0) - h_0(\mathbf{x}_0) \kappa(\mathbf{x}_0) V(\mathbf{x}_0) \right] dA(\mathbf{x}_0) \\ &+ \int_{\partial \mathcal{M}_0} h_0(\mathbf{x}_0) V_{\partial \mathcal{M}_0}(\mathbf{x}_0) d\ell(\mathbf{x}_0) \\ &= \int_{\Omega_N} \left[ \dot{f}(\bar{\mathbf{x}}) - f(\bar{\mathbf{x}}) \sum_{K=0}^N \kappa(\mathbf{x}_K) V(\mathbf{x}_K) \right] d\mu(\bar{\mathbf{x}}) \\ &+ \sum_{K=0}^N \int_{\Omega_{N,K}} \Delta f_K(\bar{\mathbf{x}}) V_{\partial \mathcal{M}_K} d\mu'_{N,K}(\bar{\mathbf{x}}). \end{aligned} \quad (60)$$

Full derivation in the paper

# DIFFERENTIAL PATH INTEGRAL

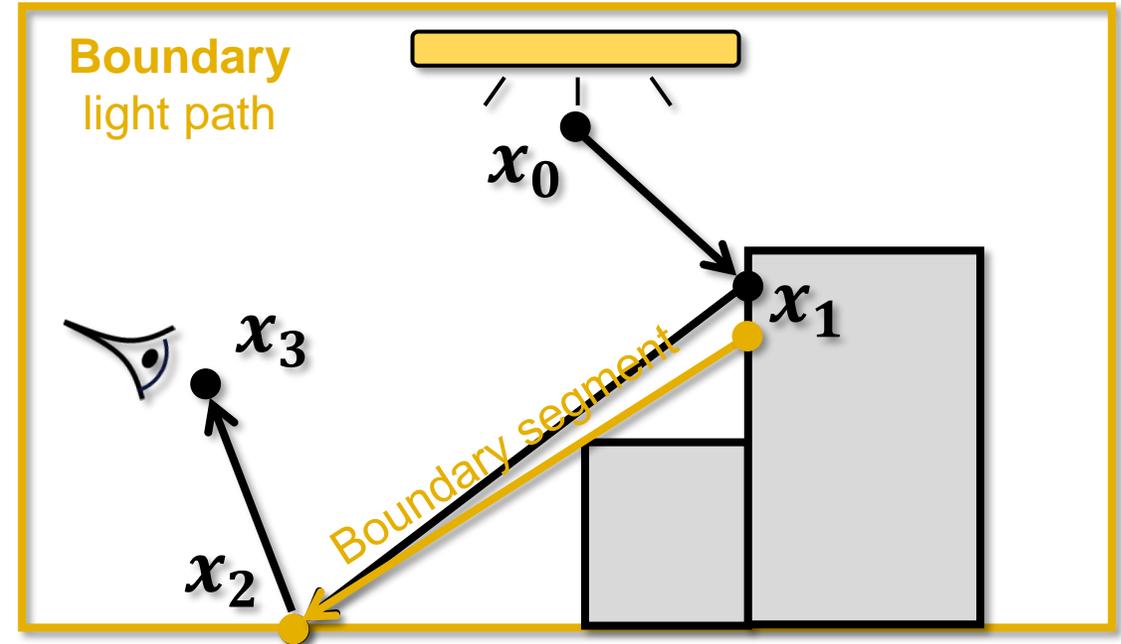
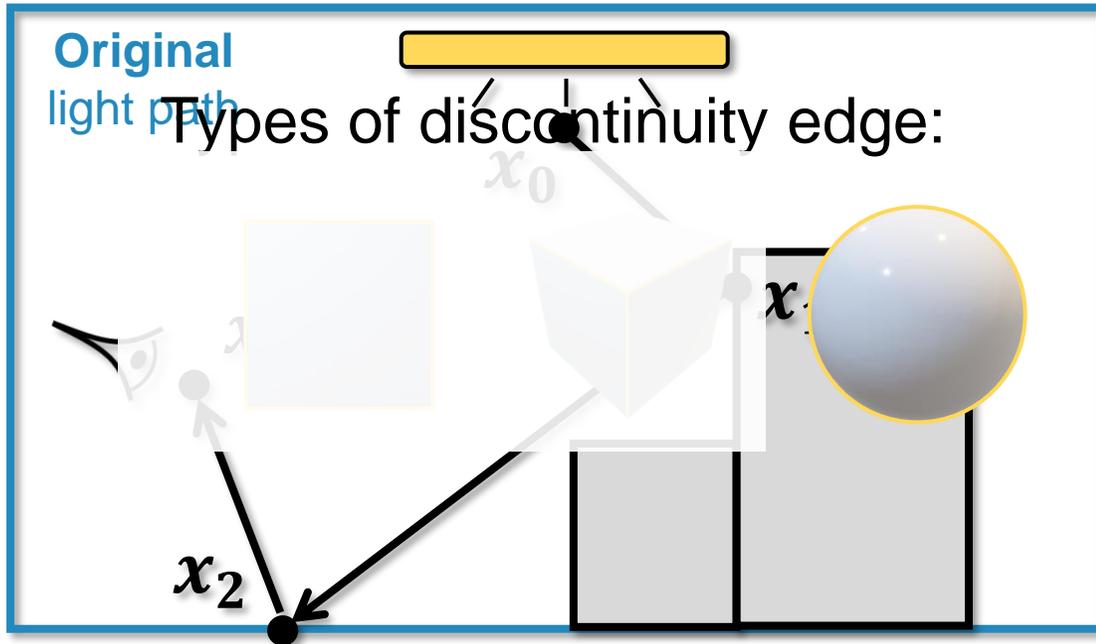
Path Integral

A generalization of Reynolds theorem

Differential Path Integral

$$I = \int_{\Omega} f(\bar{x}) d\mu(\bar{x}) \quad \longrightarrow \quad \frac{dI}{d\pi} = \int_{\Omega} \frac{d}{d\pi} f(\bar{x}) d\mu(\bar{x}) + \int_{\partial\Omega} g(\bar{x}) d\mu'(\bar{x})$$

path space Interior integral path space Boundary integral

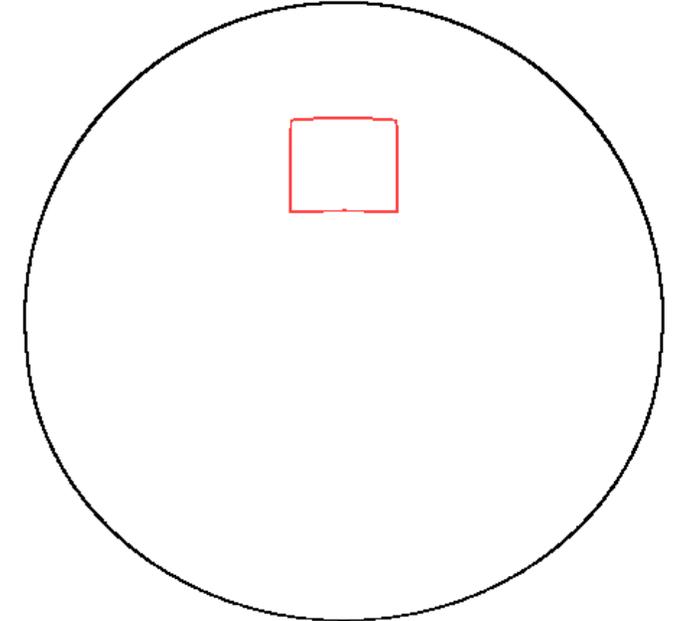
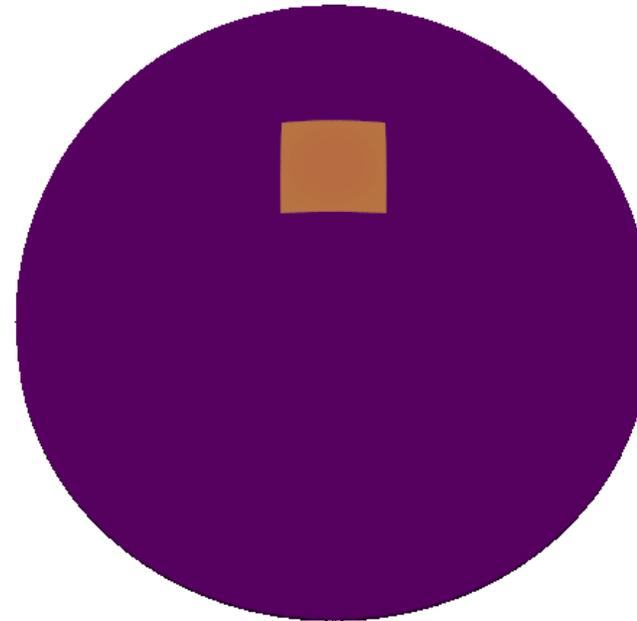
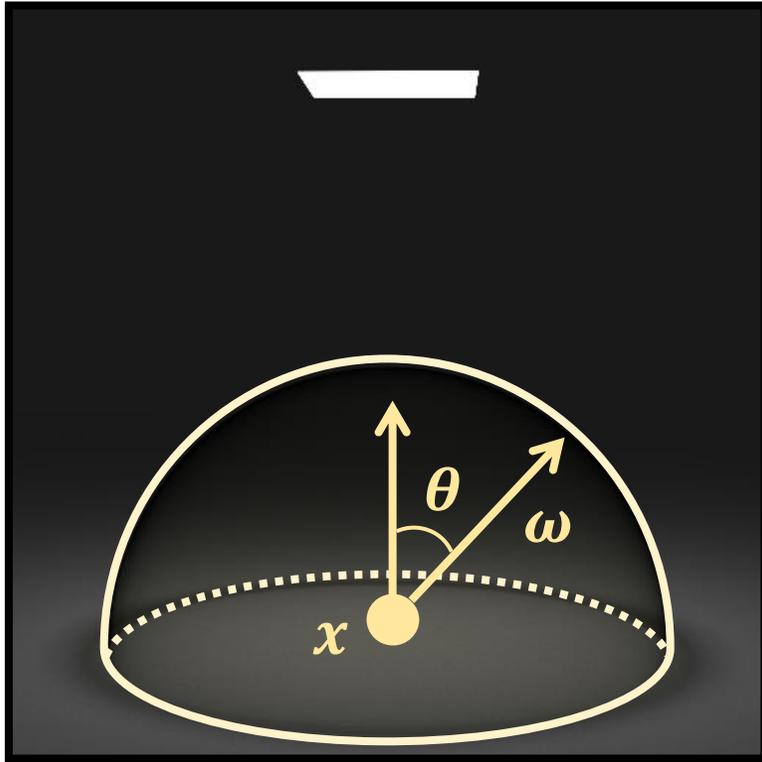


# REVISIT - DIFFERENTIAL IRRADIANCE

$\pi$ : size of the emitter

Low  High

Discontinuities of  $f$



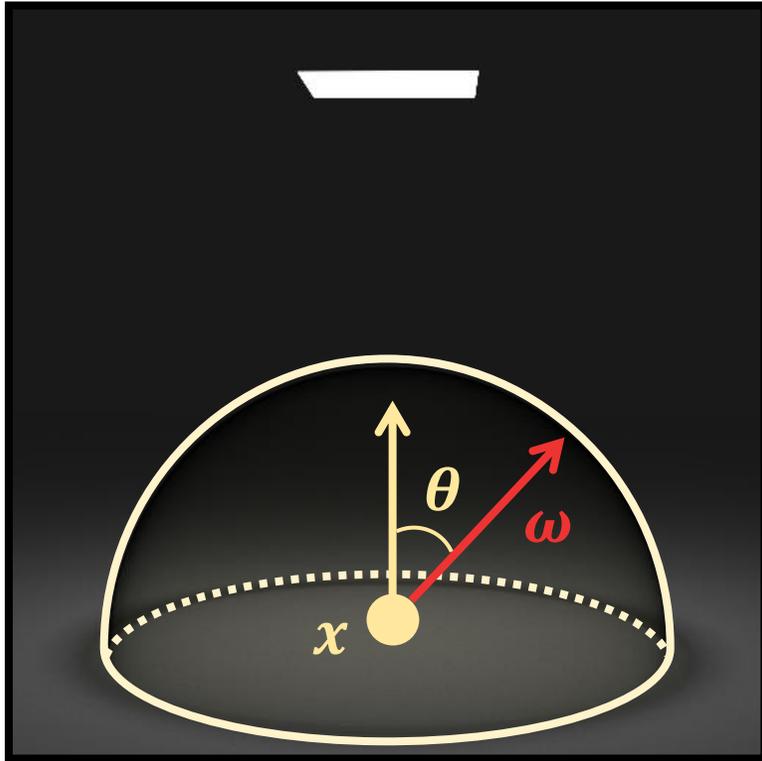
$$E = \int_{\mathbb{H}^2} \underbrace{L_i(\omega)}_f \cos\theta \, d\sigma(\omega)$$

Differentiation 

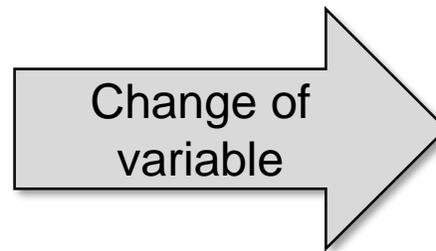
$$\frac{dE}{d\pi} = \int_{\mathbb{H}^2} \frac{df}{d\pi} \, d\sigma + \int_{\partial\mathbb{H}^2} g \, dl$$

$= 0$ 
 $\neq 0$

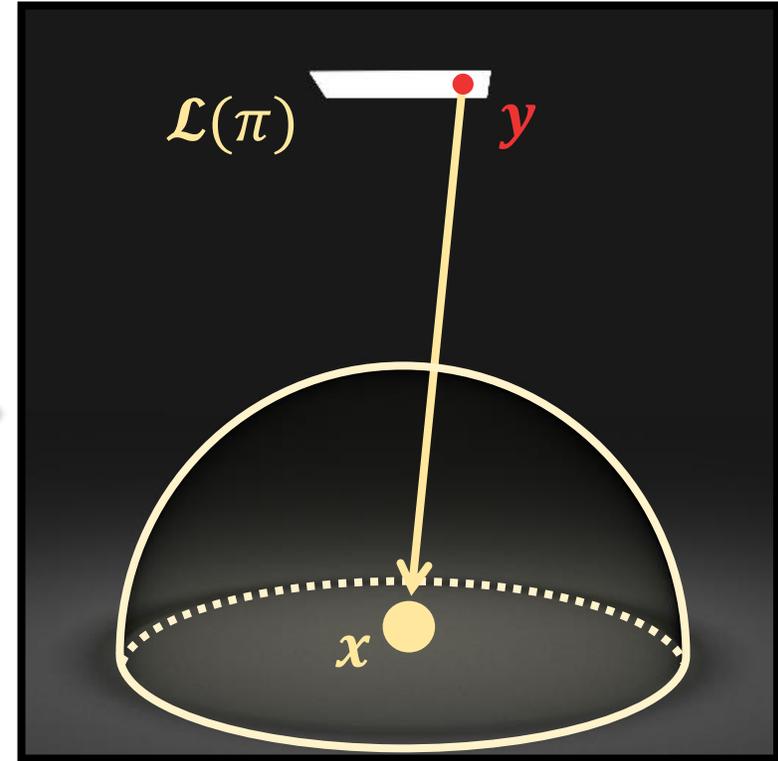
## Spherical integral



$$E = \int_{\mathbb{H}^2} L_i(\omega) \cos\theta \, d\sigma(\omega)$$

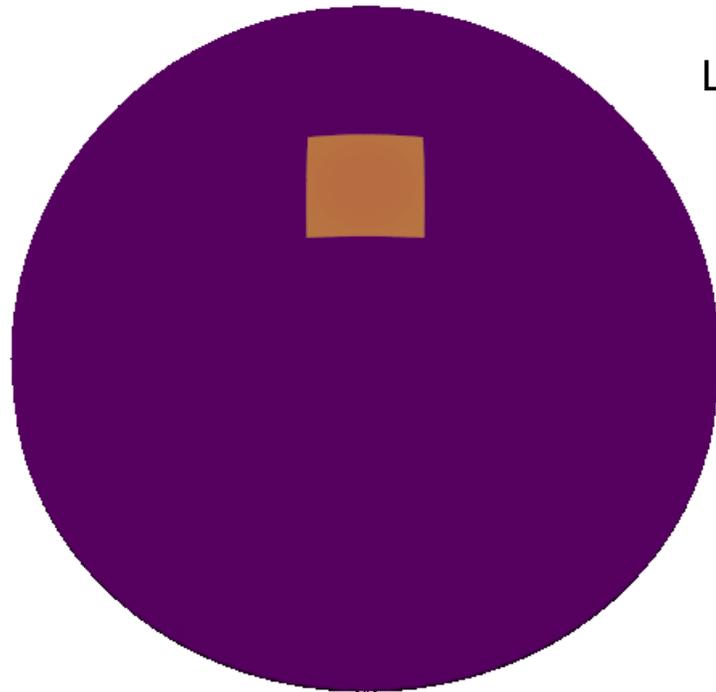


## Surface integral

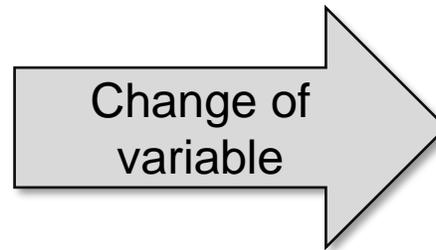


$$E = \int_{\mathcal{L}(\pi)} L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y}) \, dA(\mathbf{y})$$

## Spherical integral



Low  High



## Surface integral



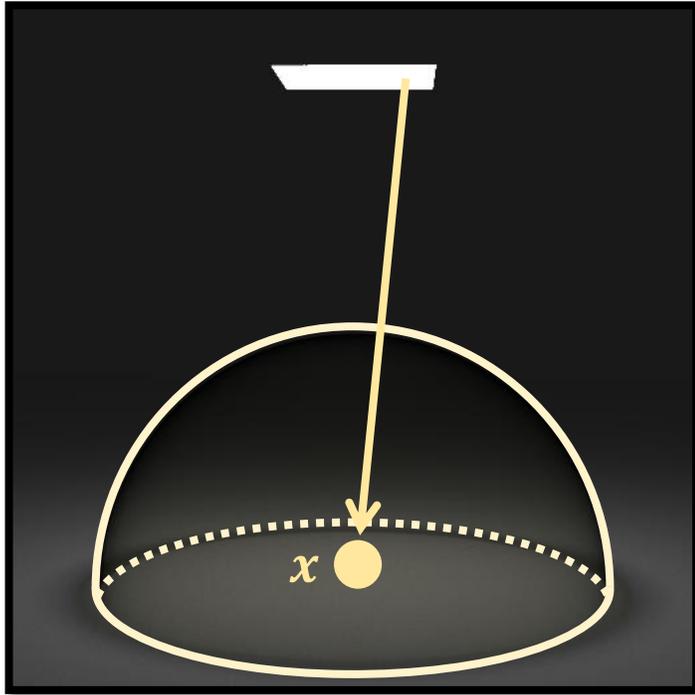
$$E = \int_{\mathbb{H}^2} \overset{\text{discontinuous}}{L_i(\boldsymbol{\omega}) \cos\theta} d\sigma(\boldsymbol{\omega})$$

constant domain

$$E = \int_{\mathcal{L}(\pi)} \overset{\text{continuous}}{L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y})} dA(\mathbf{y})$$

evolving domain

# DIFFERENTIAL IRRADIANCE



Low  High

Boundary of  $\mathcal{L}(\pi)$



$$E = \int_{\mathcal{L}(\pi)} \overbrace{L_e(\mathbf{y} \rightarrow \mathbf{x})G(\mathbf{x}, \mathbf{y})}^f dA(\mathbf{y})$$

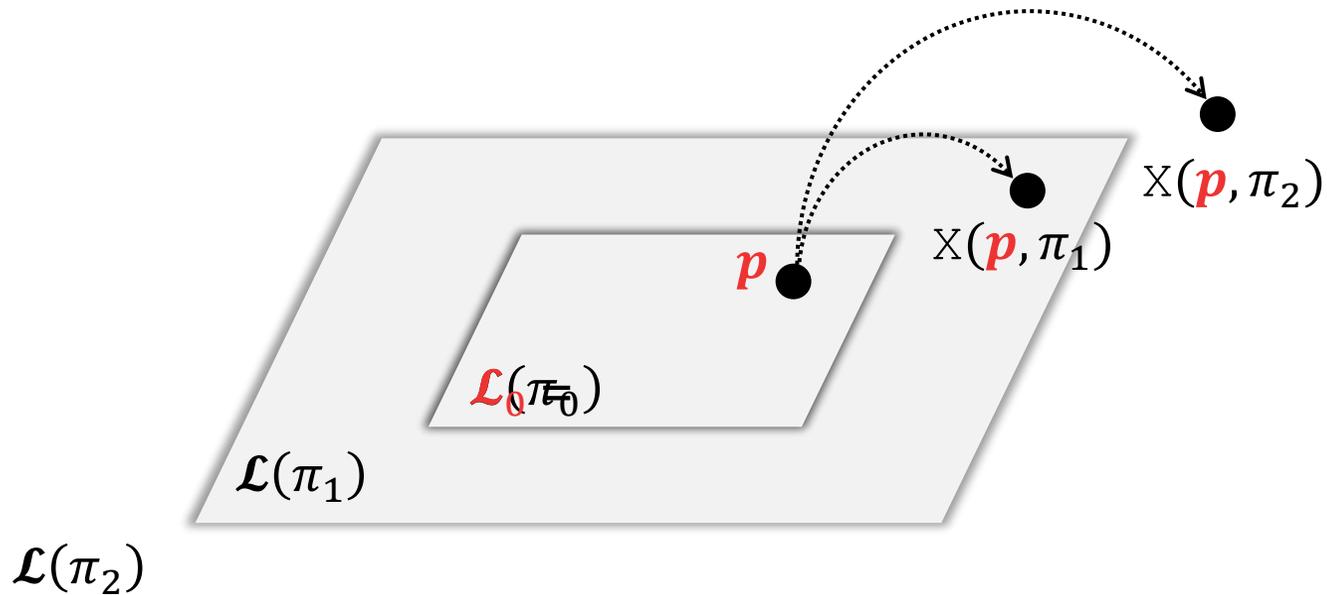


A generalization of Reynolds theorem

$$\frac{dE}{d\pi} = \int_{\mathcal{L}(\pi)} \frac{df}{d\pi} dA + \int_{\partial\mathcal{L}(\pi)} g dl$$

$\neq 0$

$$E = \int_{\mathcal{L}(\pi)} L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$



Parameterize  $\mathcal{L}(\pi)$  using some fixed  $\mathcal{L}_0$ :

$$\mathbf{y} = X(\mathbf{p}, \pi)$$

where  $X(\cdot, \pi)$  is one-to-one and continuous

Reparameterization  
with  $\mathbf{y} = X(\mathbf{p}, \pi)$ :

$$E = \int_{\mathcal{L}_0} L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y}) \left| \frac{dA(\mathbf{y})}{dA(\mathbf{p})} \right| dA(\mathbf{p})$$

# REPARAMETERIZATION



↓  $\mathbf{y} = \mathbf{x}(\mathbf{p}, \pi)$



$$E = \int_{\mathcal{L}(\pi)} \overbrace{L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y})}^f dA(\mathbf{y})$$

$$\frac{dE}{d\pi} = \underbrace{\int_{\mathcal{L}(\pi)} \frac{df}{d\pi} dA}_{= 0} + \underbrace{\int_{\partial\mathcal{L}(\pi)} g dl}_{\neq 0}$$

$$E = \int_{\mathcal{L}_0} \overbrace{L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y})}^{f_0} \left| \frac{dA(\mathbf{y})}{dA(\mathbf{p})} \right| dA(\mathbf{p})$$

$$\frac{dE}{d\pi} = \underbrace{\int_{\mathcal{L}_0} \frac{df_0}{d\pi} dA}_{\neq 0} + \underbrace{\int_{\partial\mathcal{L}_0} g_0 dl}_{= 0}$$

## Reparameterization for irradiance

$$E = \int_{\mathcal{L}(\pi)} L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y}) dA(\mathbf{y})$$

$$\mathbf{y} = \mathbb{X}(\mathbf{p}, \pi)$$



$$E = \int_{\mathcal{L}_0} L_e(\mathbf{y} \rightarrow \mathbf{x}) G(\mathbf{x}, \mathbf{y}) \left| \frac{dA(\mathbf{y})}{dA(\mathbf{p})} \right| dA(\mathbf{p})$$

↑  
Fixed surface

## Reparameterization for path integral

$$I = \int_{\Omega(\pi)} f(\bar{\mathbf{x}}) d\mu(\bar{\mathbf{x}})$$

$$\bar{\mathbf{x}} = \mathbb{X}(\bar{\mathbf{p}}, \pi)$$



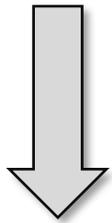
$$I = \int_{\Omega_0} f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| d\mu(\bar{\mathbf{p}})$$

↑  
Fixed path space

$$\parallel \prod_i \left| \frac{dA(\mathbf{x}_i)}{dA(\mathbf{p}_i)} \right|$$

Original

$$I = \int_{\Omega(\pi)} f(\bar{\mathbf{x}}) d\mu(\bar{\mathbf{x}})$$



$$\bar{\mathbf{x}} = \mathbb{X}(\bar{\mathbf{p}}, \pi)$$

Reparameterized

$$I = \int_{\Omega_0} f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| d\mu(\bar{\mathbf{p}})$$

Original

$$\frac{dI}{d\pi} = \int_{\Omega(\pi)} \frac{df(\bar{\mathbf{x}})}{d\pi} d\mu(\bar{\mathbf{x}}) + \int_{\partial\Omega(\pi)} g(\bar{\mathbf{x}}) d\mu'(\bar{\mathbf{x}})$$

- Pro:** No global parametrization required  
**Con:** More types of discontinuities

Reparameterized

$$\frac{dI}{d\pi} = \int_{\Omega_0} \frac{d}{d\pi} \left( f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| \right) d\mu(\bar{\mathbf{p}}) + \int_{\partial\Omega_0} g(\bar{\mathbf{p}}) d\mu'(\bar{\mathbf{p}})$$

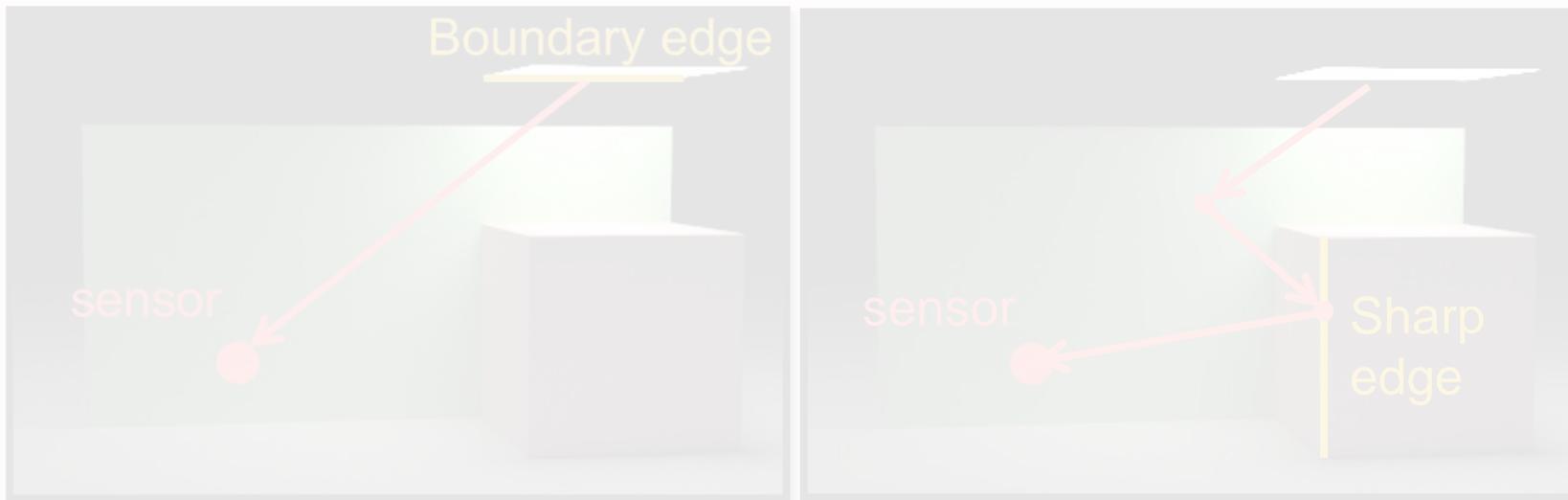
- Pro:** Fewer types of discontinuities  
**Con:** Requires global parametrization  $\mathbb{X}$

## Differential path integral

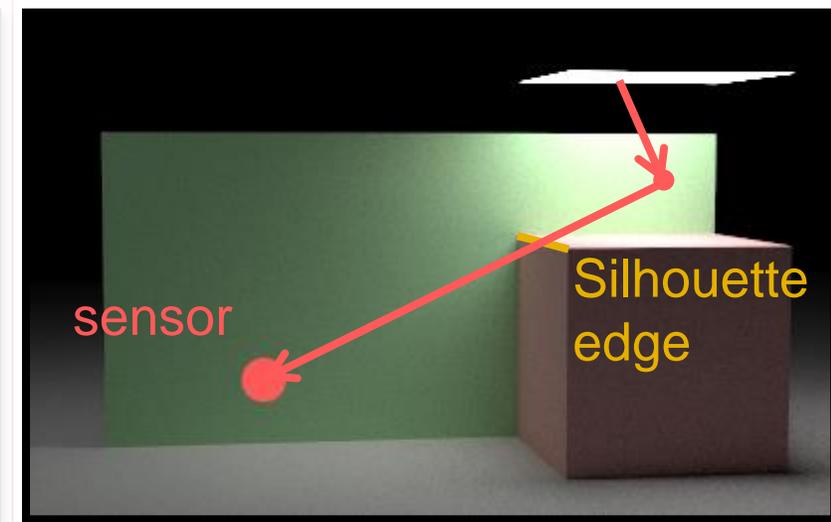
$$\frac{dI}{d\pi} = \int_{\Omega(\pi)} \frac{df(\bar{x})}{d\pi} d\mu(\bar{x}) + \int_{\partial\Omega(\pi)} g(\bar{x}) d\mu'(\bar{x})$$

$$\frac{dI}{d\pi} = \int_{\Omega_0} \frac{d}{d\pi} \left( f(\bar{x}) \left| \frac{d\mu(\bar{x})}{d\mu(\bar{p})} \right| \right) d\mu(\bar{p}) + \int_{\partial\Omega_0} g(\bar{p}) d\mu'(\bar{p})$$

### Topology-driven



### Visibility-driven



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# MONTE CARLO ESTIMATORS

# ESTIMATING INTERIOR INTEGRAL

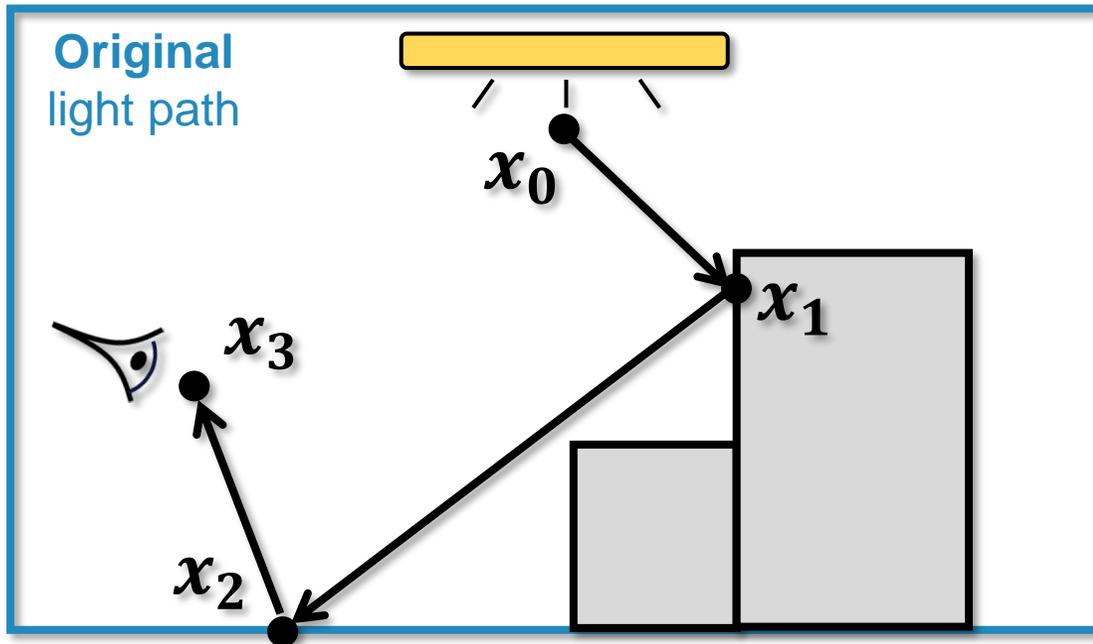
(Reparameterized)  
Differential path Integral

$$\frac{\partial I}{\partial \pi} = \int_{\Omega_0} \frac{\partial}{\partial \pi} \left( f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| \right) d\mu(\bar{\mathbf{p}}) + \int_{\partial\Omega_0} g(\bar{\mathbf{p}}) d\mu'(\bar{\mathbf{p}})$$

Interior integral

Boundary integral

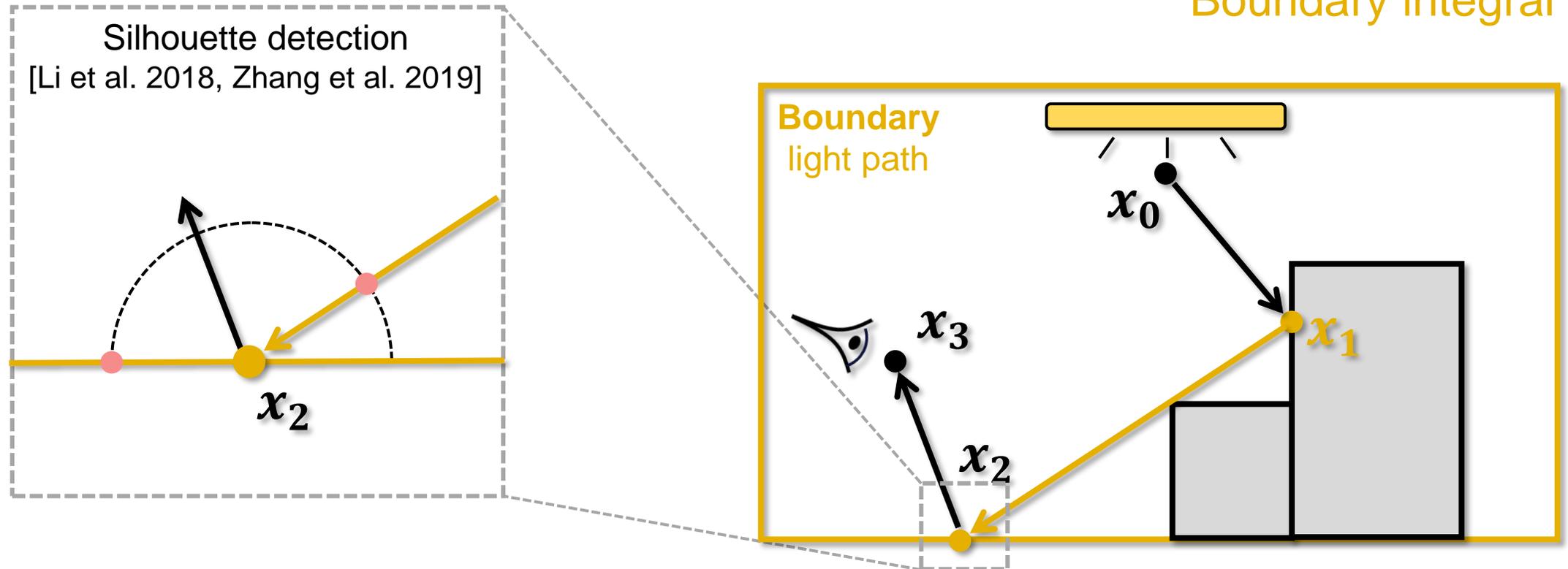
- Can be estimated using identical path sampling strategies as forward rendering
  - Unidirectional path tracing
  - Bidirectional path tracing
  - ...



(Reparameterized)  
Differential path Integral

$$\frac{\partial I}{\partial \pi} = \int_{\Omega_0} \frac{\partial}{\partial \pi} \left( f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| \right) d\mu(\bar{\mathbf{p}}) + \int_{\partial\Omega_0} g(\bar{\mathbf{p}}) d\mu'(\bar{\mathbf{p}})$$

Boundary integral



(Reparameterized)  
Differential path Integral

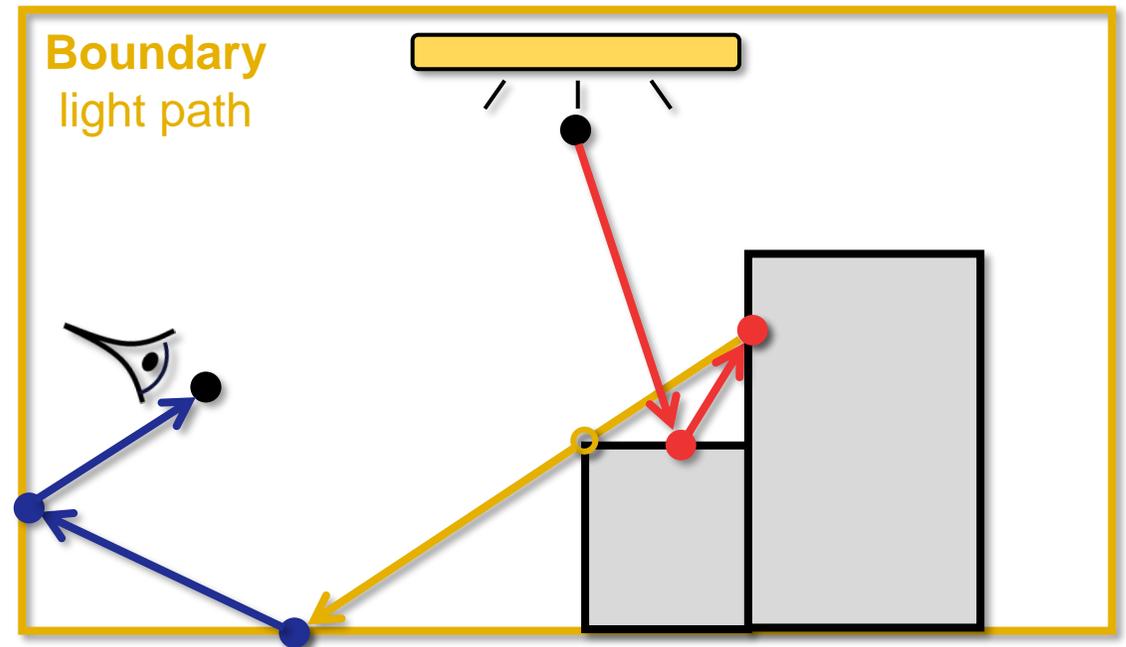
$$\frac{\partial I}{\partial \pi} = \int_{\Omega_0} \frac{\partial}{\partial \pi} \left( f(\bar{\mathbf{x}}) \left| \frac{d\mu(\bar{\mathbf{x}})}{d\mu(\bar{\mathbf{p}})} \right| \right) d\mu(\bar{\mathbf{p}}) + \int_{\partial\Omega_0} g(\bar{\mathbf{p}}) d\mu'(\bar{\mathbf{p}})$$

where  $\bar{\mathbf{x}} = \mathbb{X}(\bar{\mathbf{p}}, \pi)$

Boundary integral

- Construct **boundary segment**
- Construct **source** and **sensor** subpaths  

- To improve efficiency
  - Next-event estimation
  - Importance sampling of boundary segments

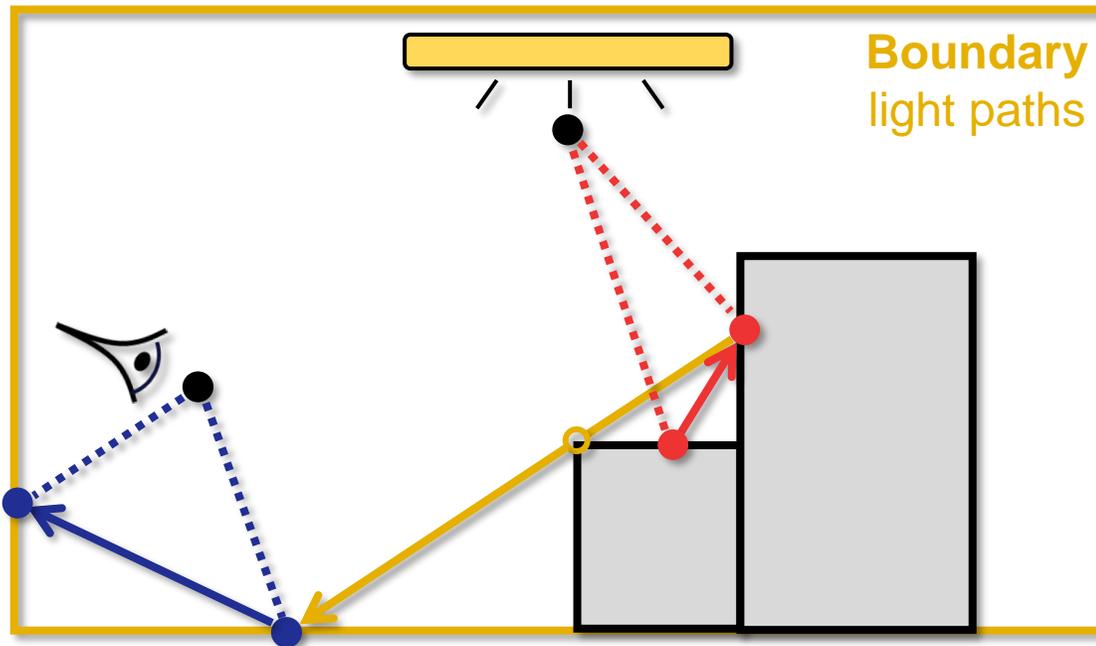


# OUR ESTIMATORS

## Unidirectional estimator

**Interior:** unidirectional path tracing

**Boundary:** unidirectional sampling of subpaths

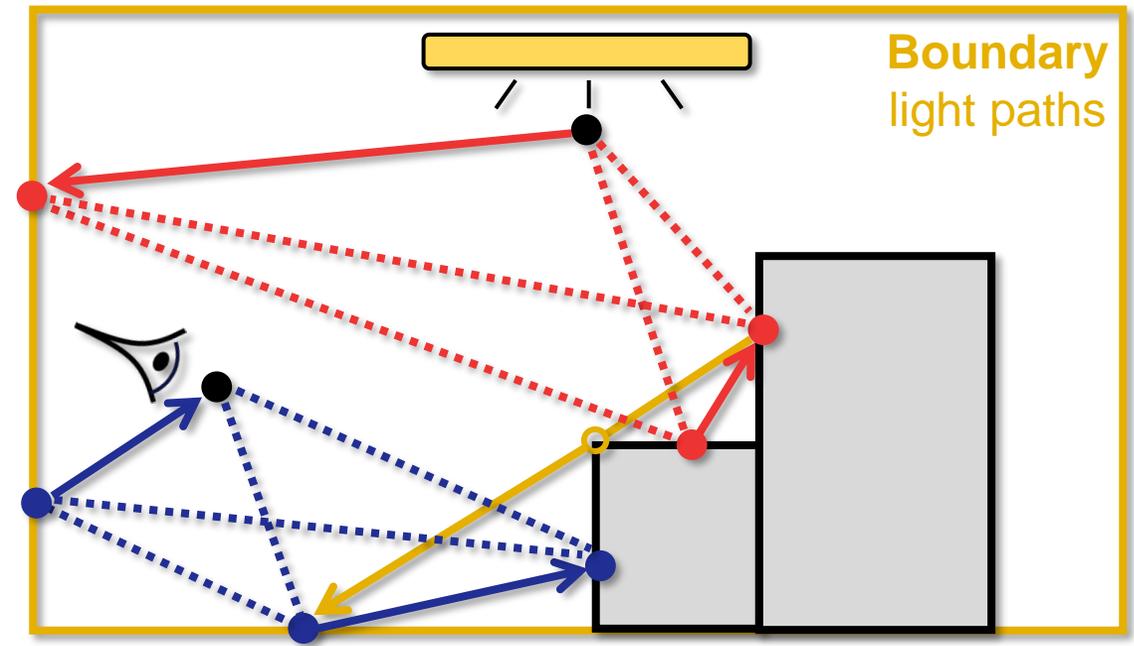


**Unidirectional path tracing + NEE**

## Bidirectional estimator

**Interior:** bidirectional path tracing

**Boundary:** bidirectional sampling of subpaths



**Bidirectional path tracing**

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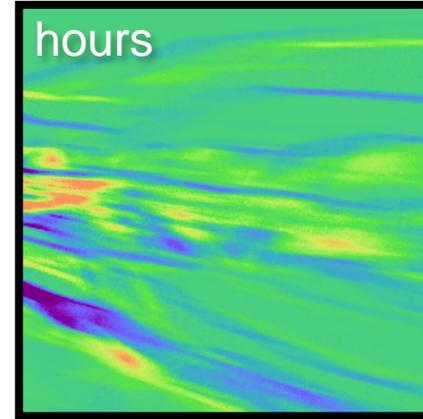
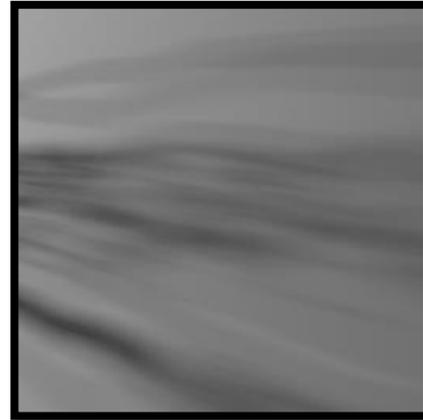
# RESULTS

# RESULTS

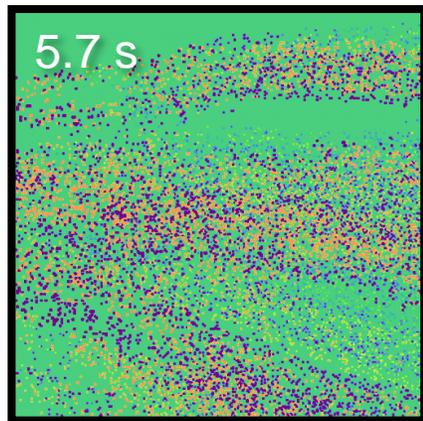
## COMPLEX GEOMETRY



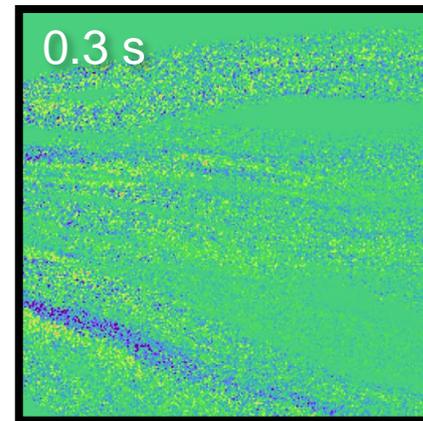
**Equal-sample comparison**



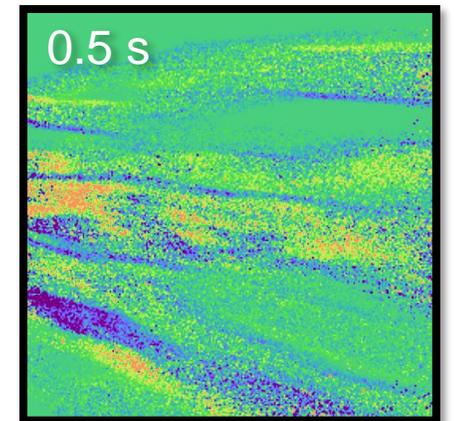
Reference



[Zhang et al. 2019]



[Loubet et al. 2019]

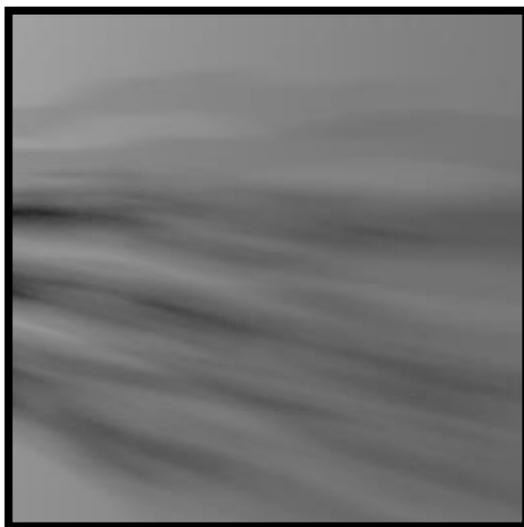


**Ours**

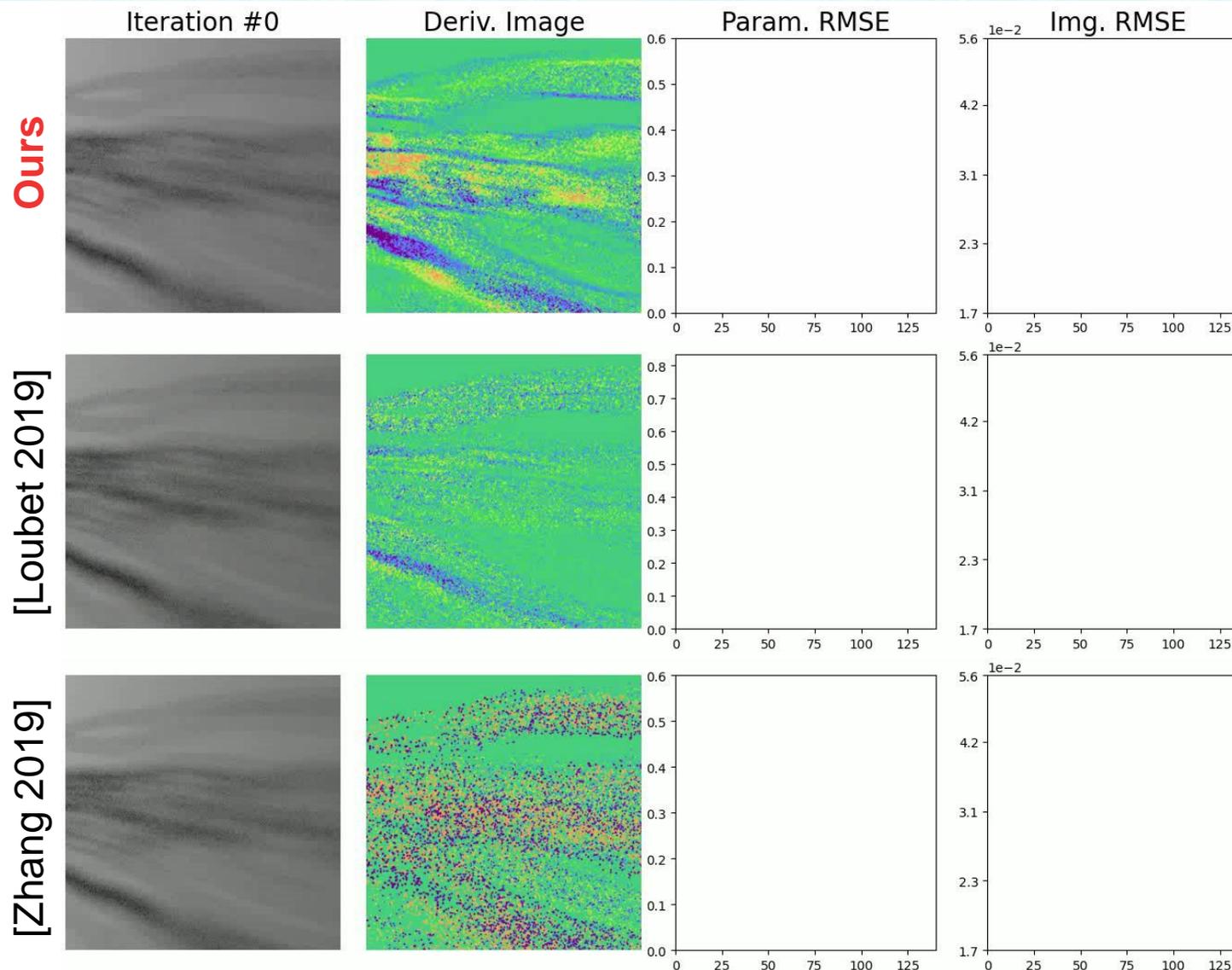
# RESULTS

## COMPLEX GEOMETRY

Target image

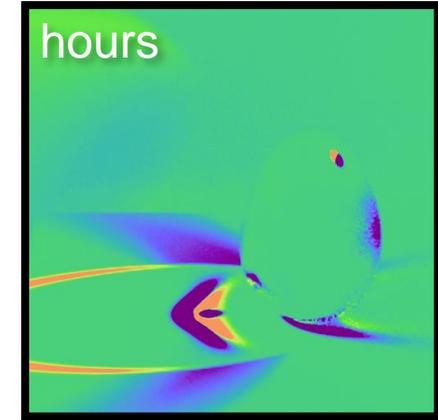
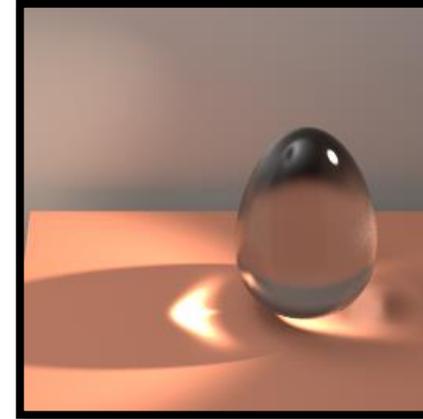


- Optimizing *rotation angle*
- **Equal-sample** per iteration
- **Identical** optimization setting
  - Learning rate (Adam)
  - Initializations



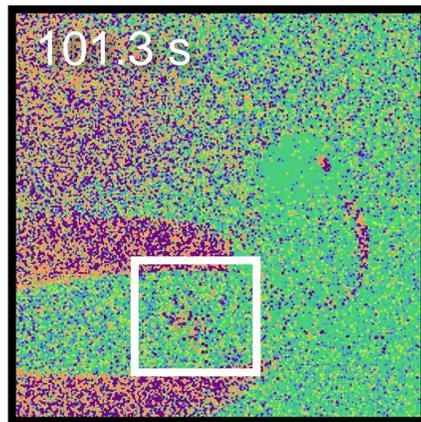
# RESULTS

## COMPLEX LIGHT TRANSPORT EFFECTS

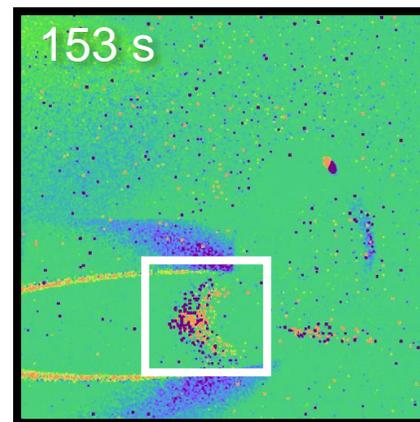


Reference

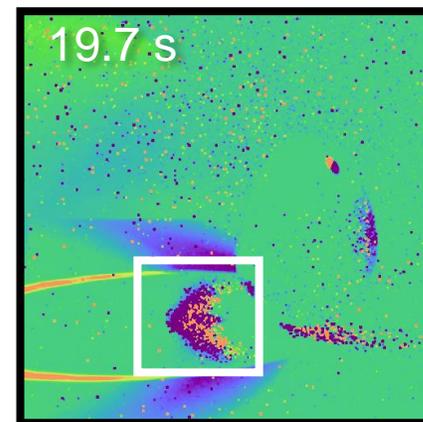
### Equal-sample comparison



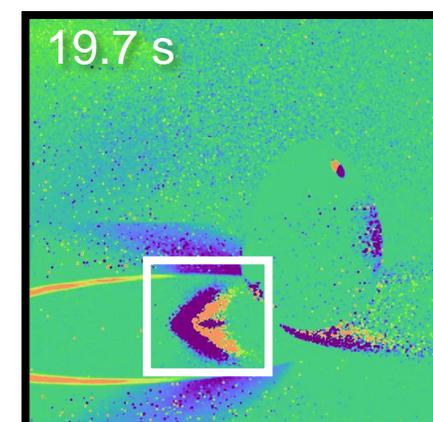
[Zhang et al. 2019]



[Loubet et al. 2019]



**Ours (unidirectional)**

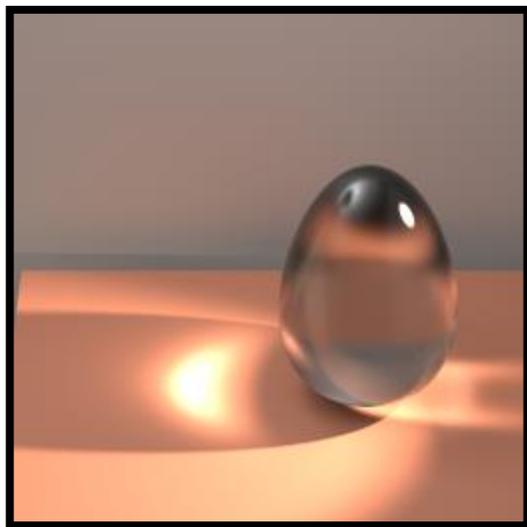


**Ours (bidirectional)**

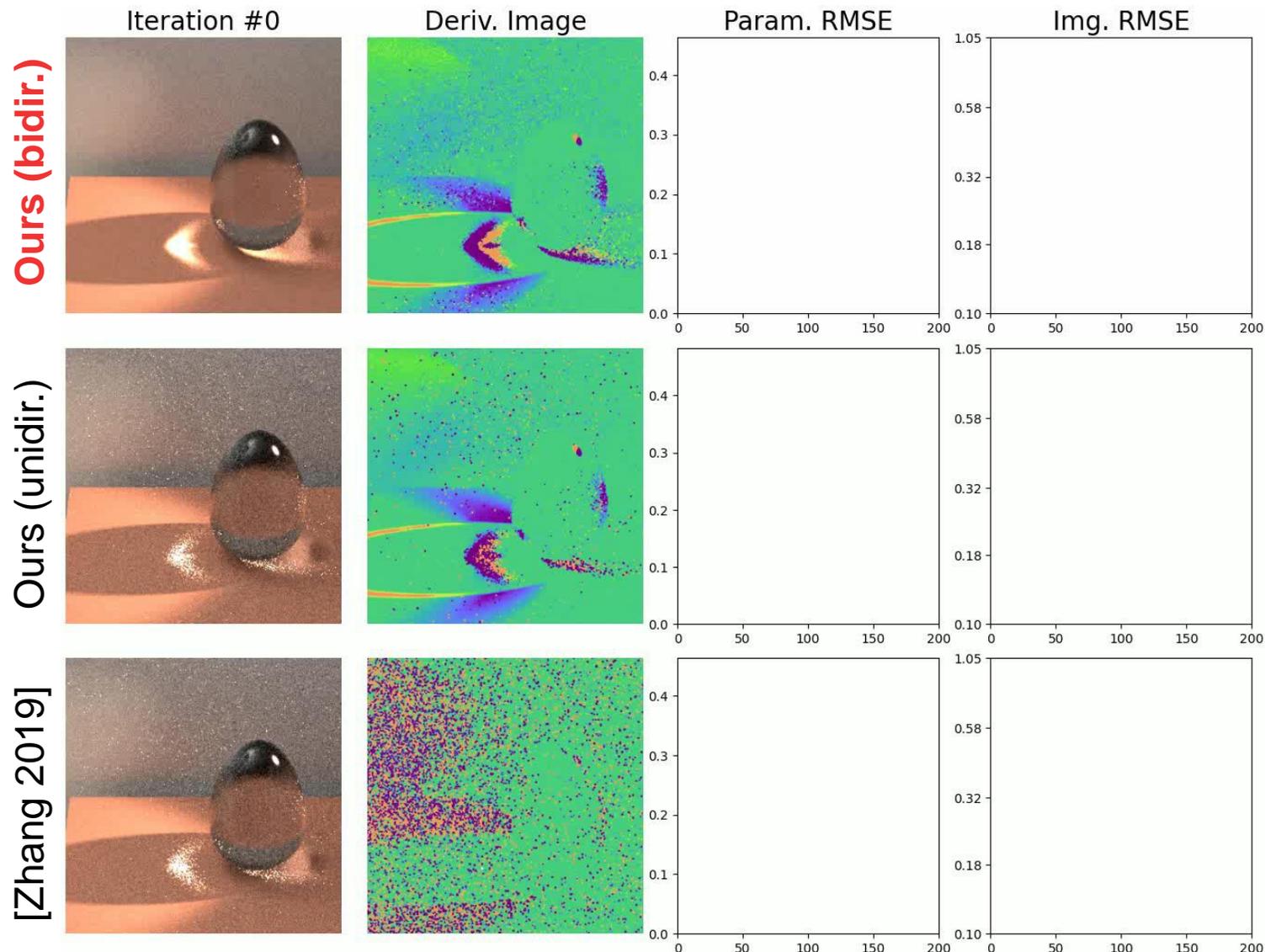
# RESULTS

## COMPLEX LIGHT TRANSPORT EFFECTS

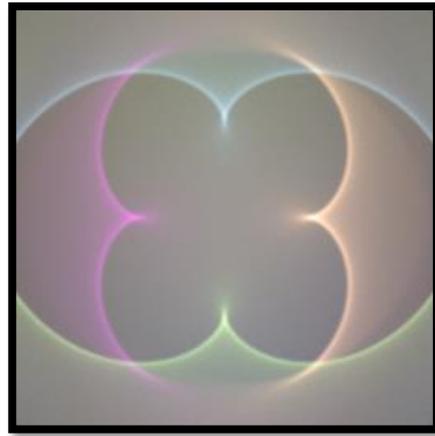
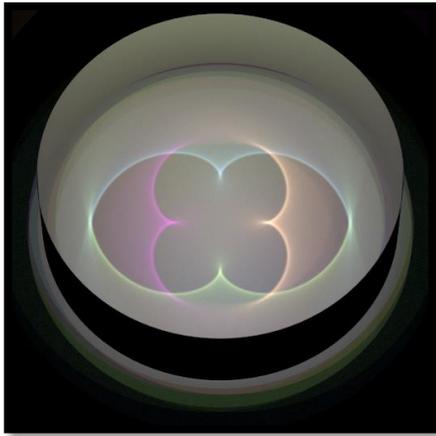
Target image



- Optimizing
  - Glass IOR
  - Spotlight position
- **Equal-time** per iteration
- **Identical** optimization setting

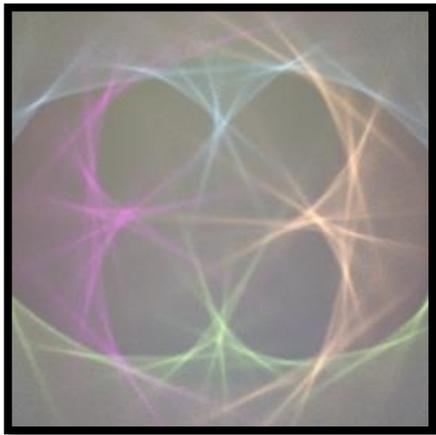


Initial

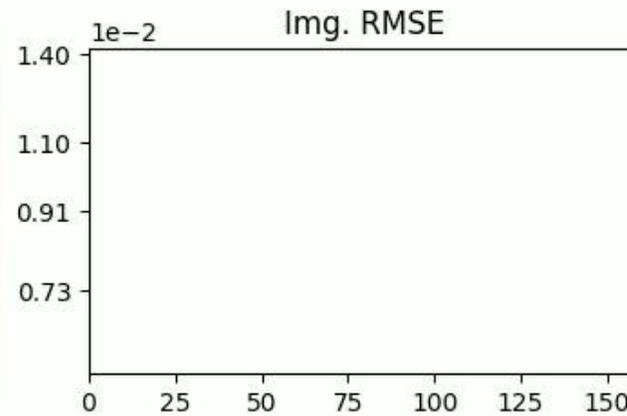
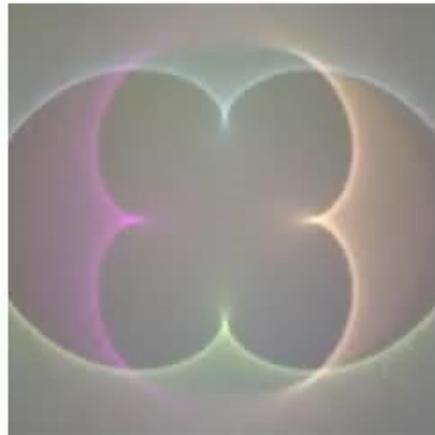


Optimizing **cross-sectional** shape (100 variables)

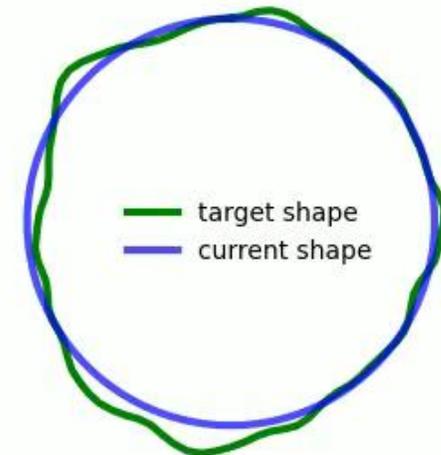
Target image



Iter #0

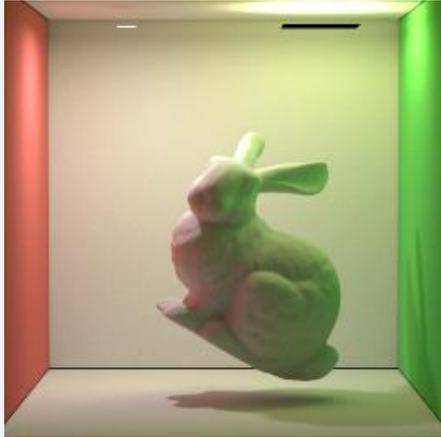


Cross-sectional shape  
(displacement x 20)

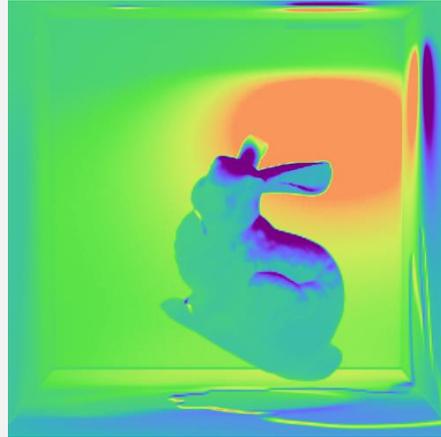


# RESULTS

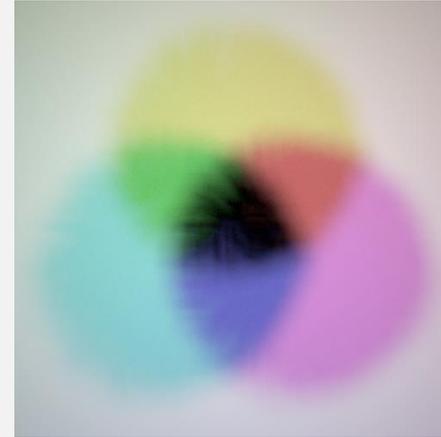
Original image



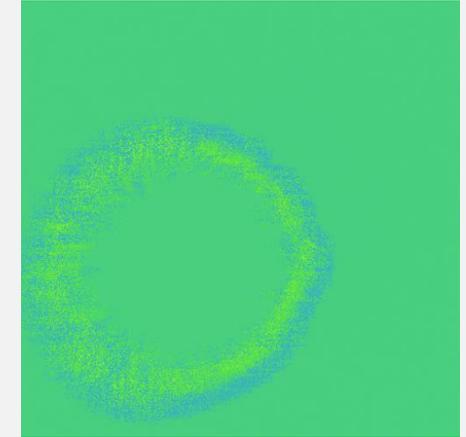
Derivative image



Original image



Derivative image



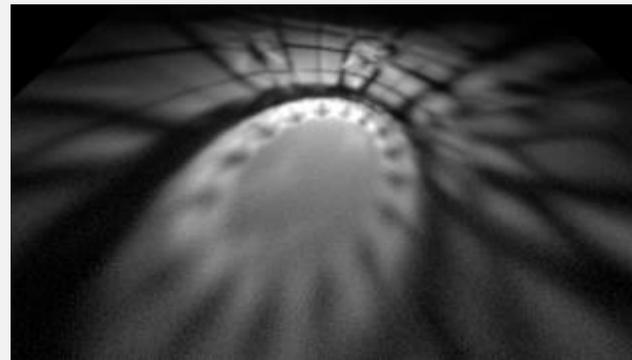
Config.



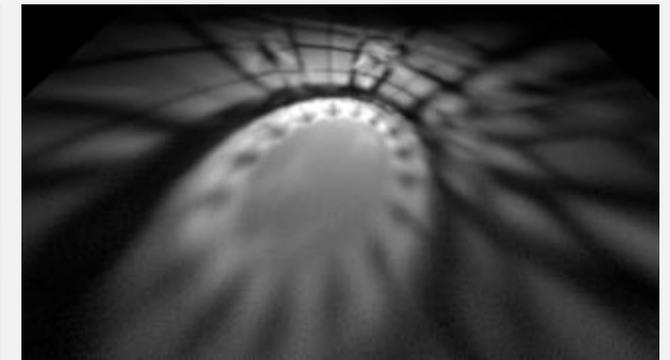
Optimize (initial)



Optimize (final)



Target



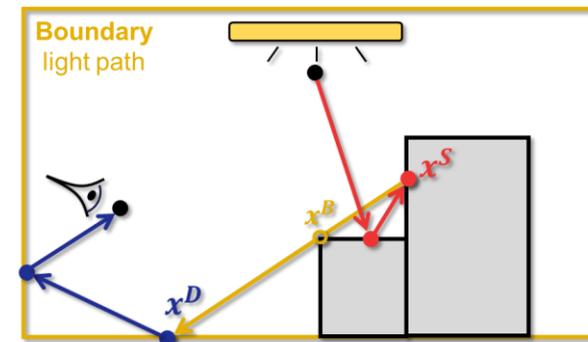
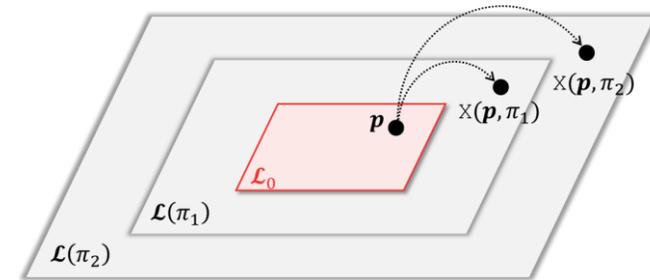
- Surface-based light transport
- More sophisticated Monte Carlo estimators
  - *Markov-chain Monte Carlo* (MCMC) methods
- Better importance sampling of paths

# CONCLUSION

- Differential path integral
  - Separated *interior* and *boundary*
- Reparameterization
  - Only need to consider silhouette edges
- Unbiased Monte Carlo methods
  - Unidirectional and bidirectional estimators
  - No silhouette detection is needed

$$\int_{\Omega} \frac{d}{d\pi} f(\bar{x}) d\mu(\bar{x}) + \int_{\partial\Omega} g(\bar{x}) d\mu'(\bar{x})$$

Interior integral                      Boundary integral



# ACKNOWLEDGMENTS

- Anonymous reviewers
- Funding
  - NSF grants 1730147, 1900849 and 1900927
  - AWS Cloud Credits for Research program



Project webpage  
<https://rb.gy/lb2z3s>

## Thank you!

