



Pre-recorded sessions:
From 4 December 2020

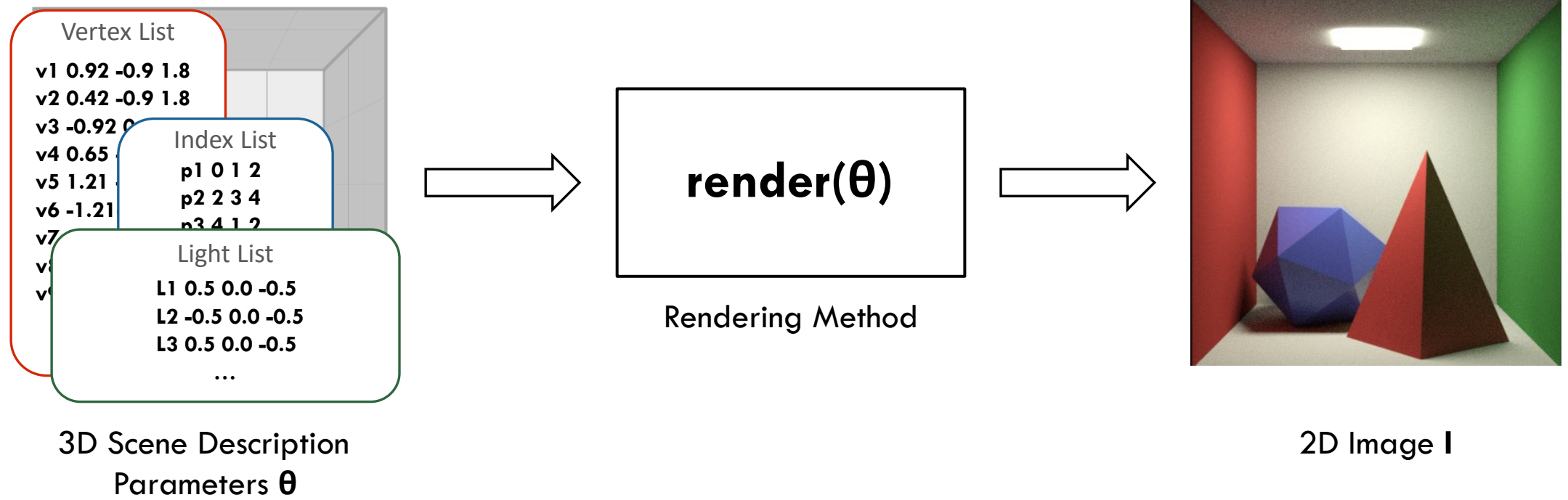
Live sessions:
10 – 13 December 2020

[SA2020.SIGGRAPH.ORG](https://sa2020.siggraph.org)
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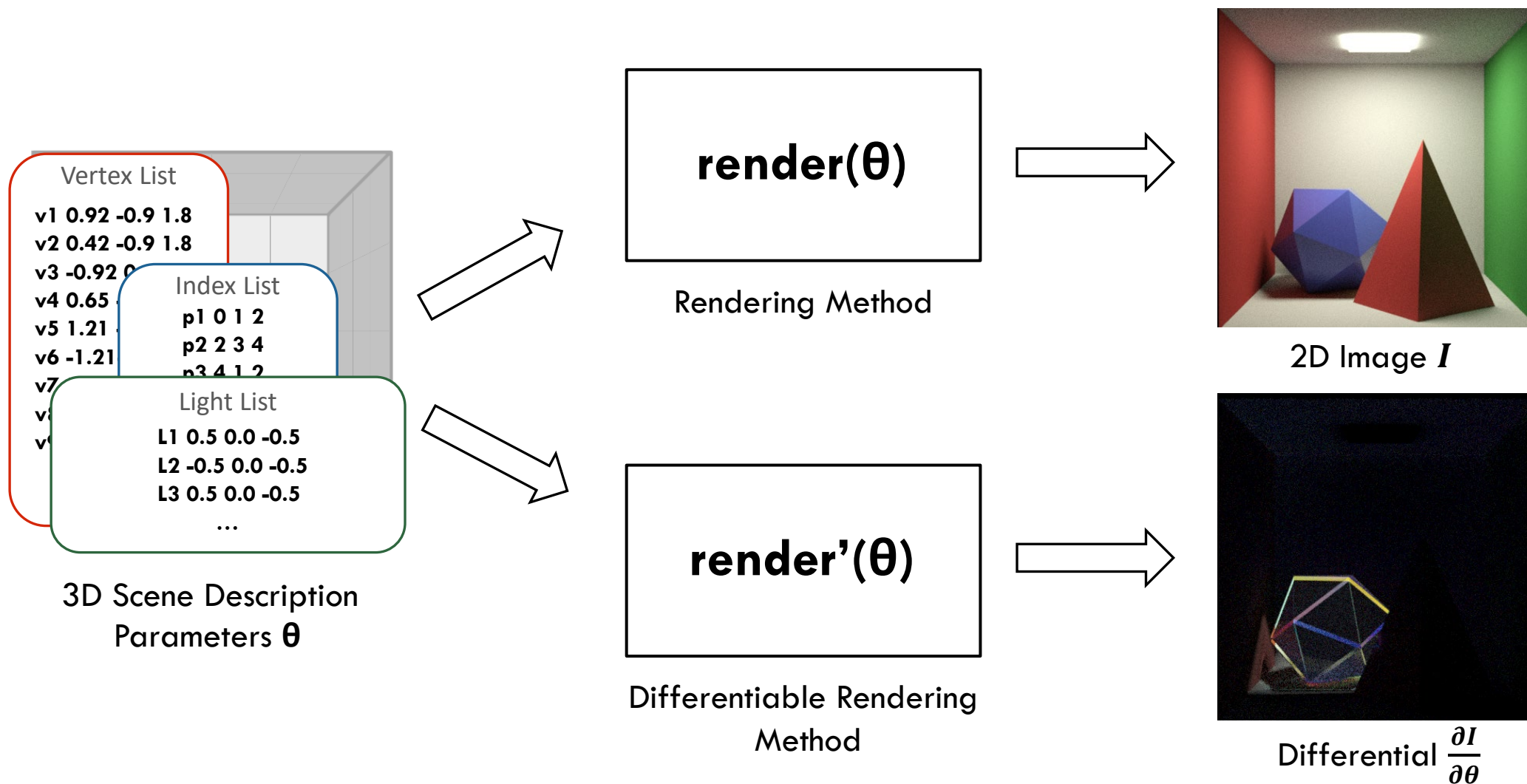
Unbiased Warped-Area Sampling for Differentiable Rendering

Sai Praveen Bangaru
Tzu-Mao Li
Frédo Durand

RENDERING



DIFFERENTIABLE RENDERING



WHY DIFFERENTIABLE RENDERING?

Motivation 1

Vertex List			
v1	0.92	-0.9	1.8
v2	0.42	-0.9	1.8
v3	-0.92	0	
v4	0.65		
v5	1.21		
v6	-1.21		
v7			
v8			
v9			

Index List			
p1	0	1	2
p2	2	3	4
p3	4	1	2

Light List			
L1	0.5	0.0	-0.5
L2	-0.5	0.0	-0.5
L3	0.5	0.0	-0.5
...			

3D Scene Description
Parameters θ

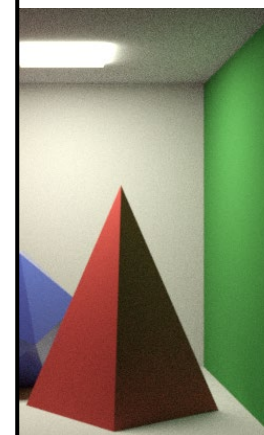
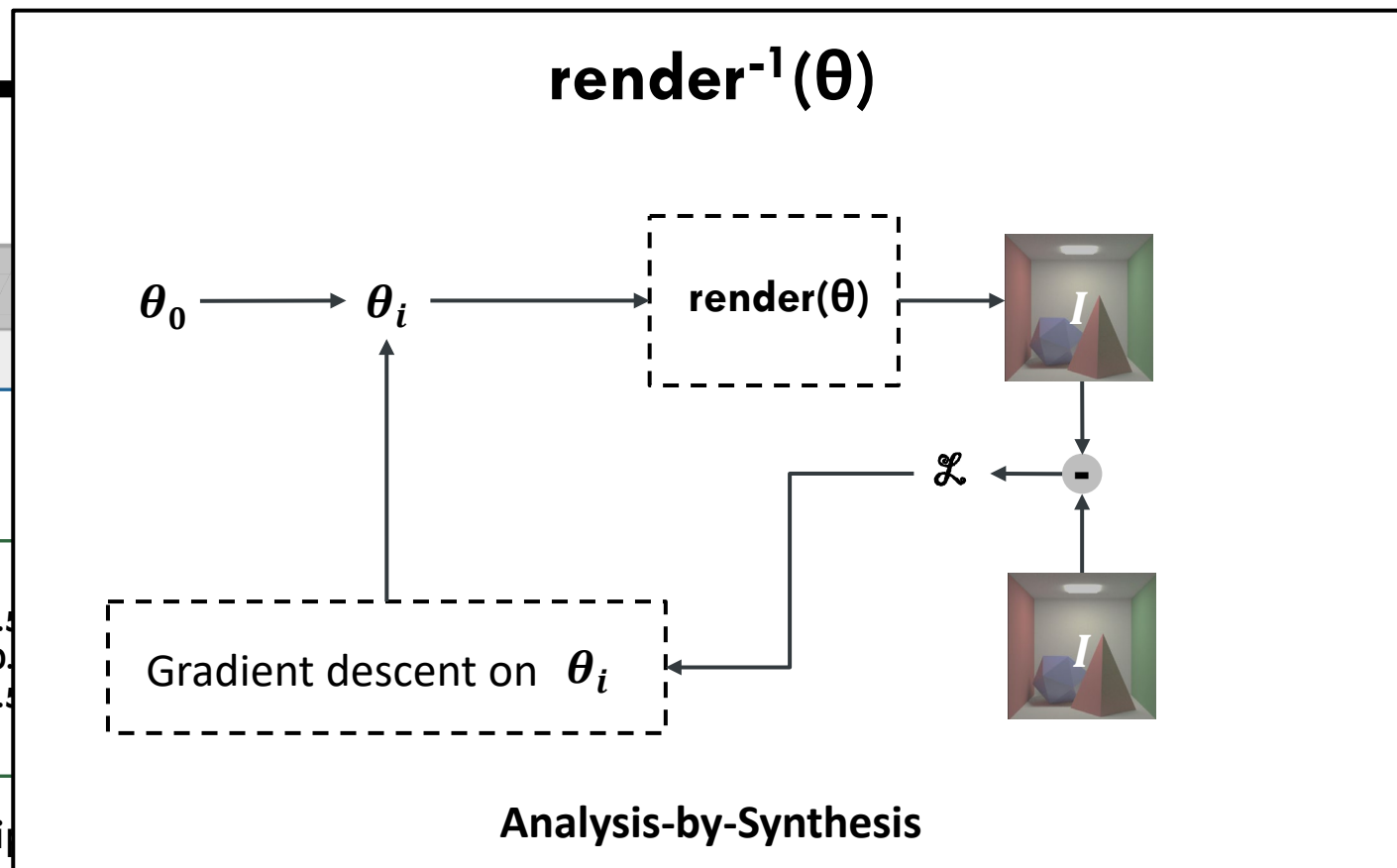
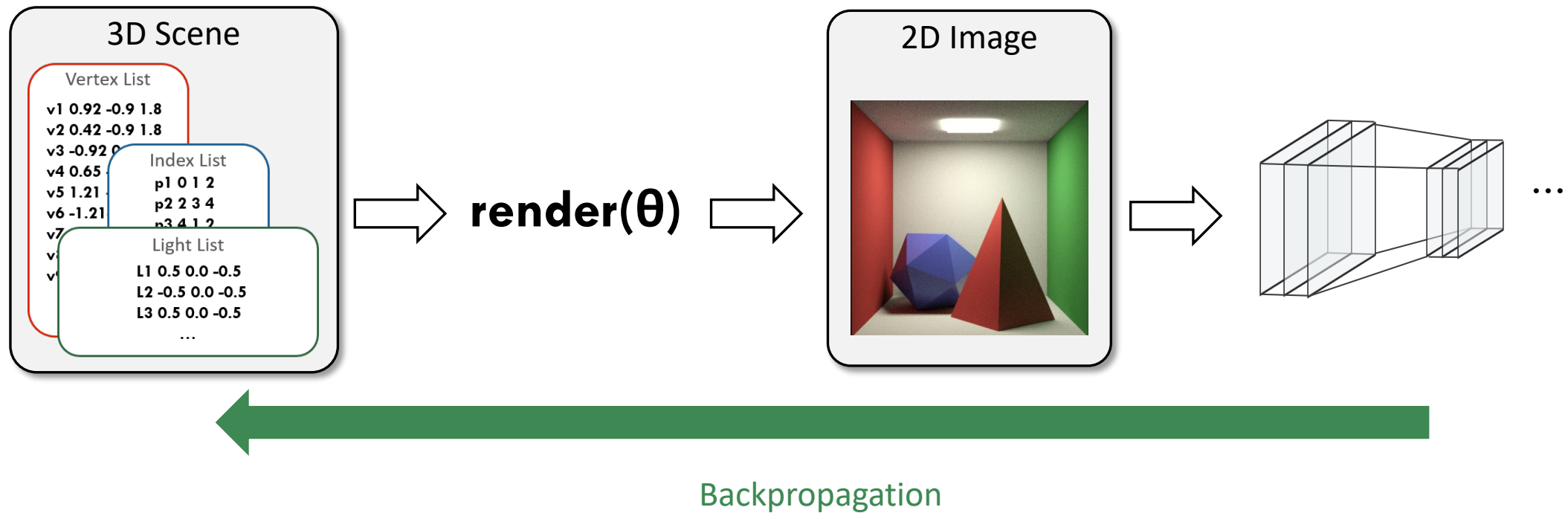


Image I

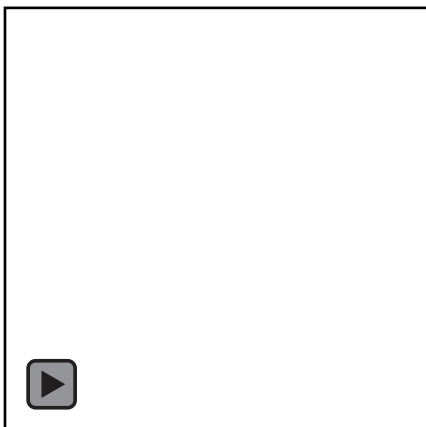
WHY DIFFERENTIABLE RENDERING?

Motivation 2 \longrightarrow Deep Learning (adversarial robustness, etc..)

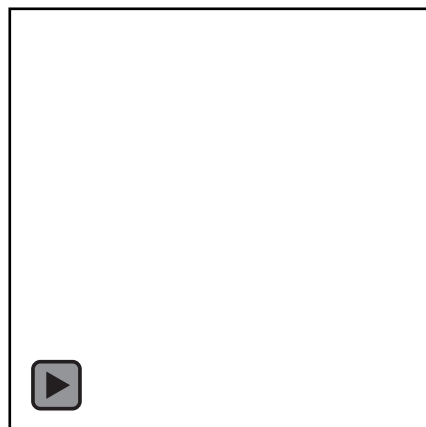


SCENE PARAMETER DERIVATIVES

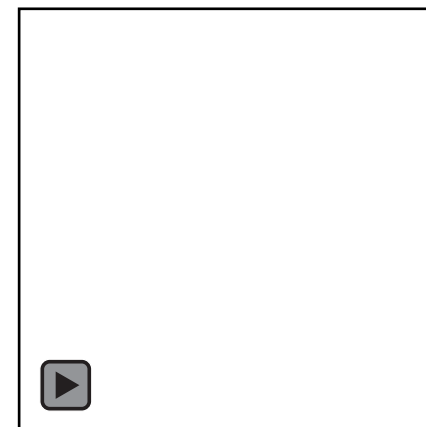
Image I



θ : Object Translation

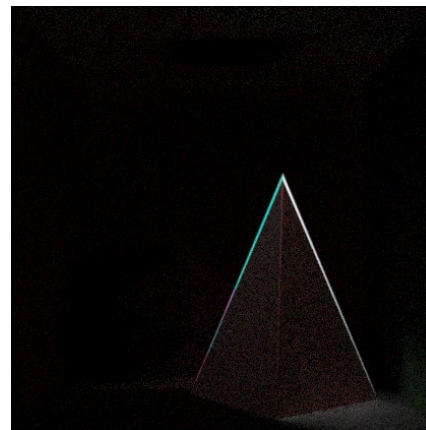
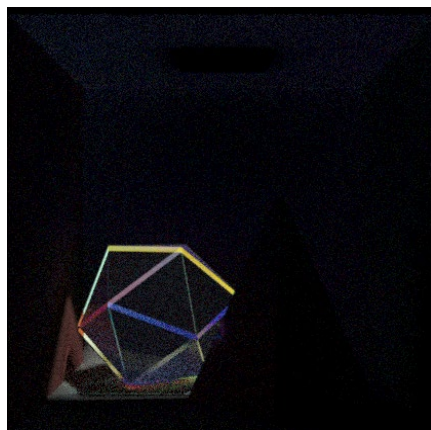


θ : Vertex position

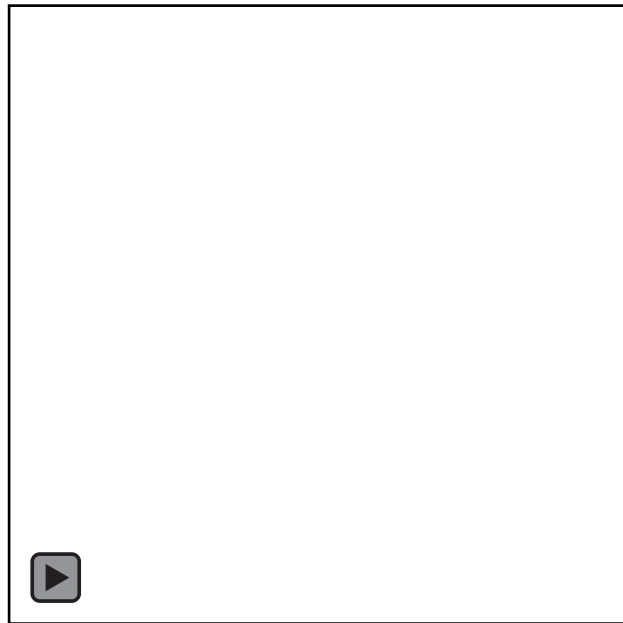


θ : Camera Rotation

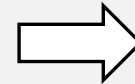
Differential $\frac{\partial I}{\partial \theta}$



AUTO-DIFF HAS A VISIBILITY PROBLEM

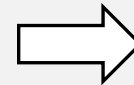


Sudden discontinuity



Auto-diff fails due to edges

Smooth function

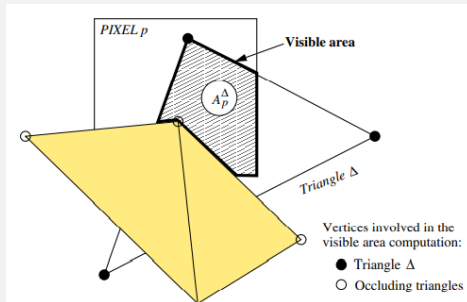


Auto-diff computes correct derivative

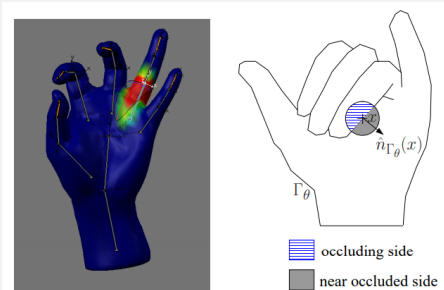
RASTERIZATION APPROACHES ARE LIMITED

Key Idea: *Analytical* occupancy

[Jalobeanu 2004]

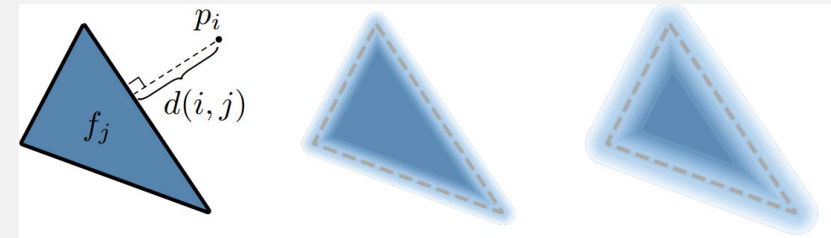


[de La Gorce 2008]

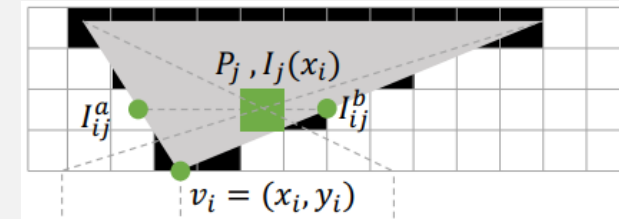


Key Idea: *Approximating* visibility

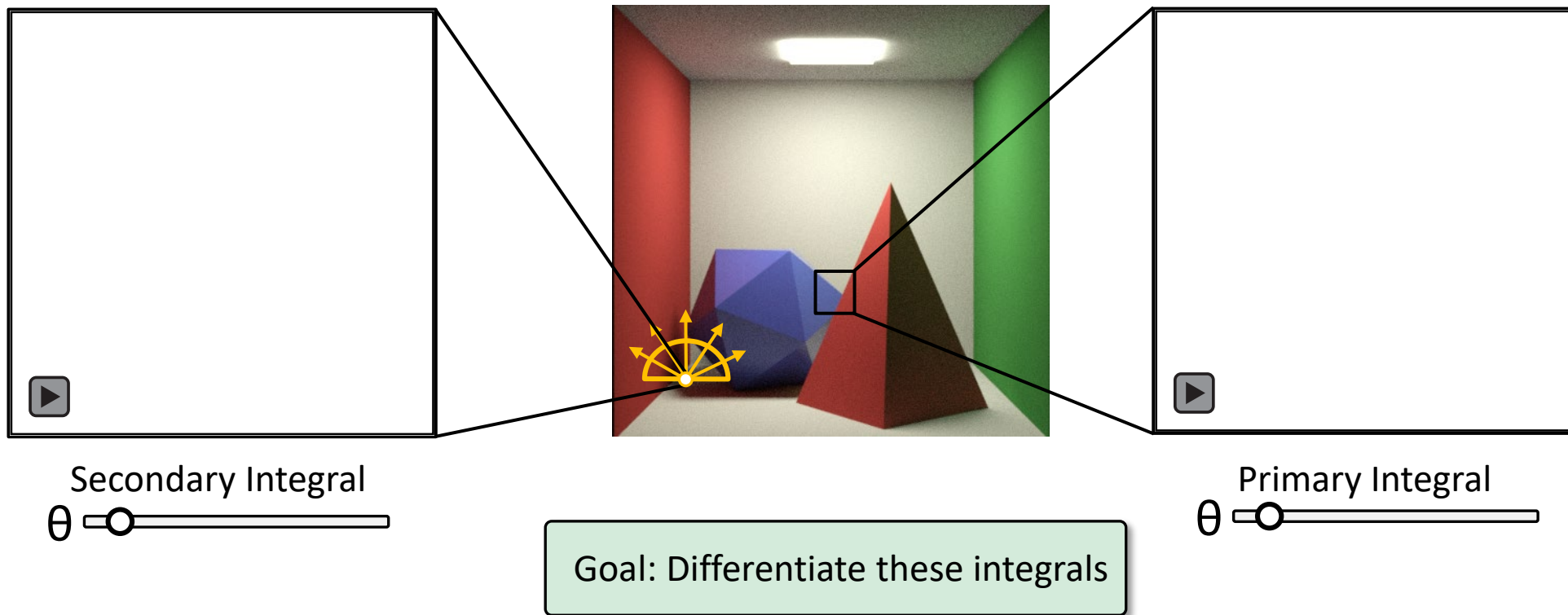
Soft Rasterizer [Liu 2019]



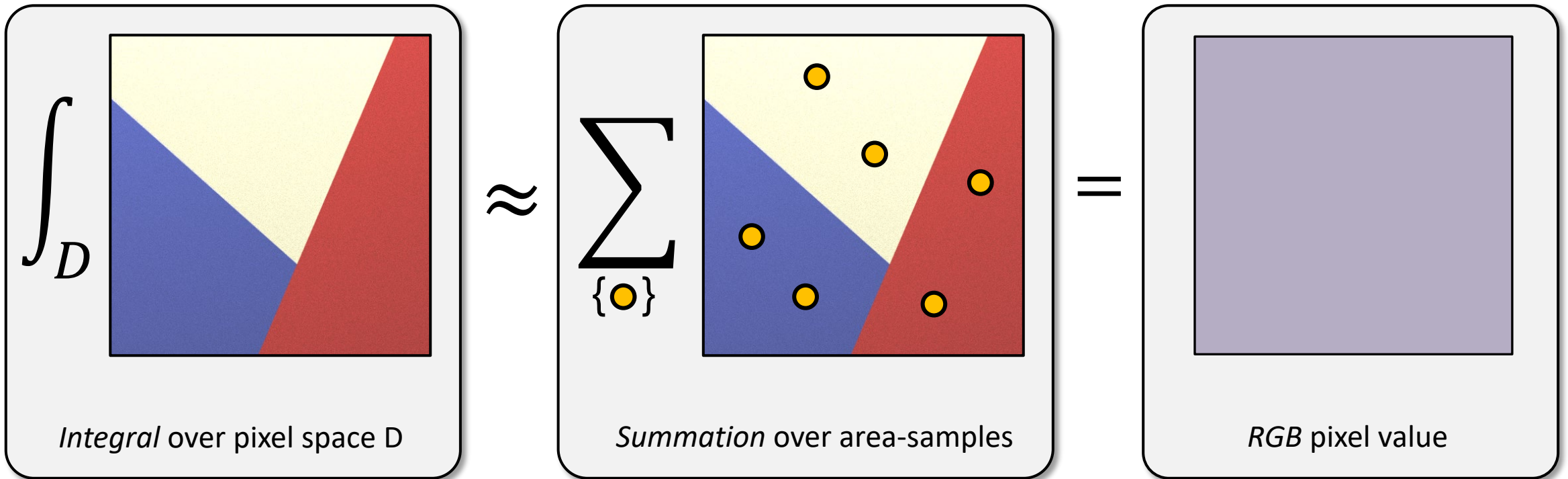
Neural 3D Mesh Renderer [Kato 2017]



RENDERING AS AN INTEGRAL



Monte Carlo Estimation



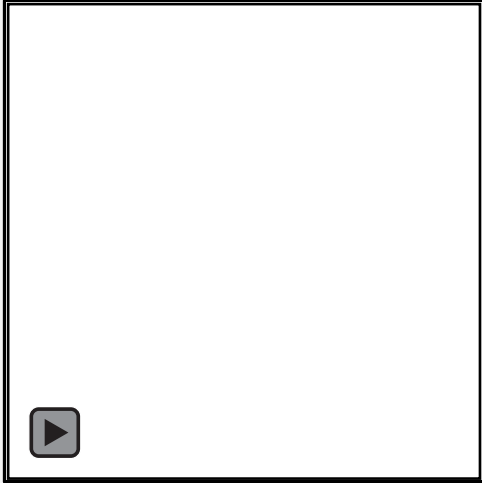
DISCONTINUOUS INTEGRANDS

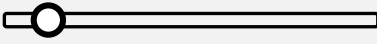
(Incorrect)

Attempt 1



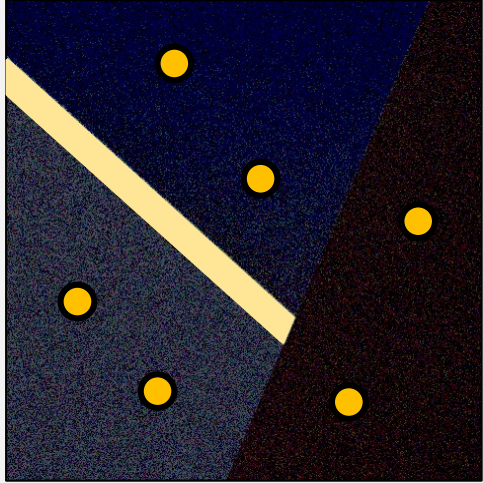
Apply *auto-diff* to summation

$$\partial_{\theta} \int_D$$


θ 

Integral over pixel space D

\neq

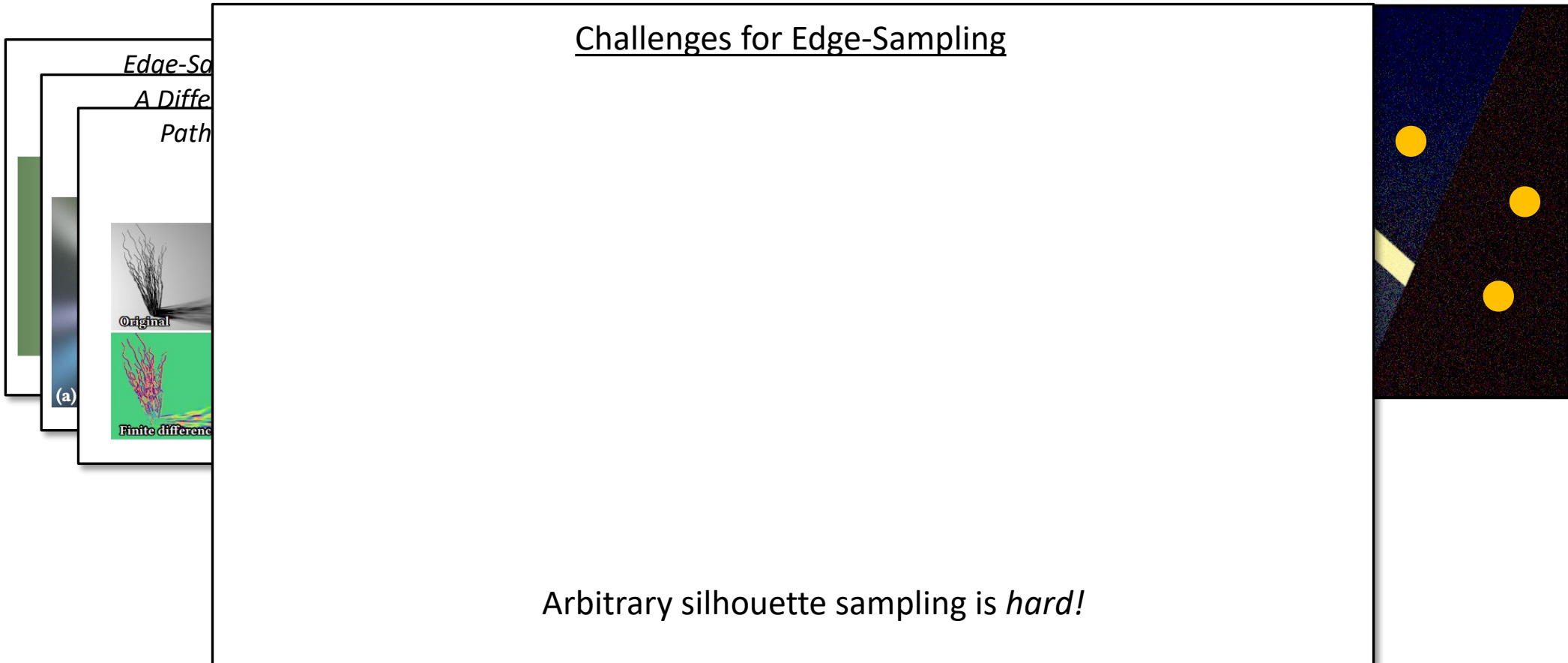
$$\sum_{\{\bullet\}} \partial_{\theta}$$


Summation over area-samples

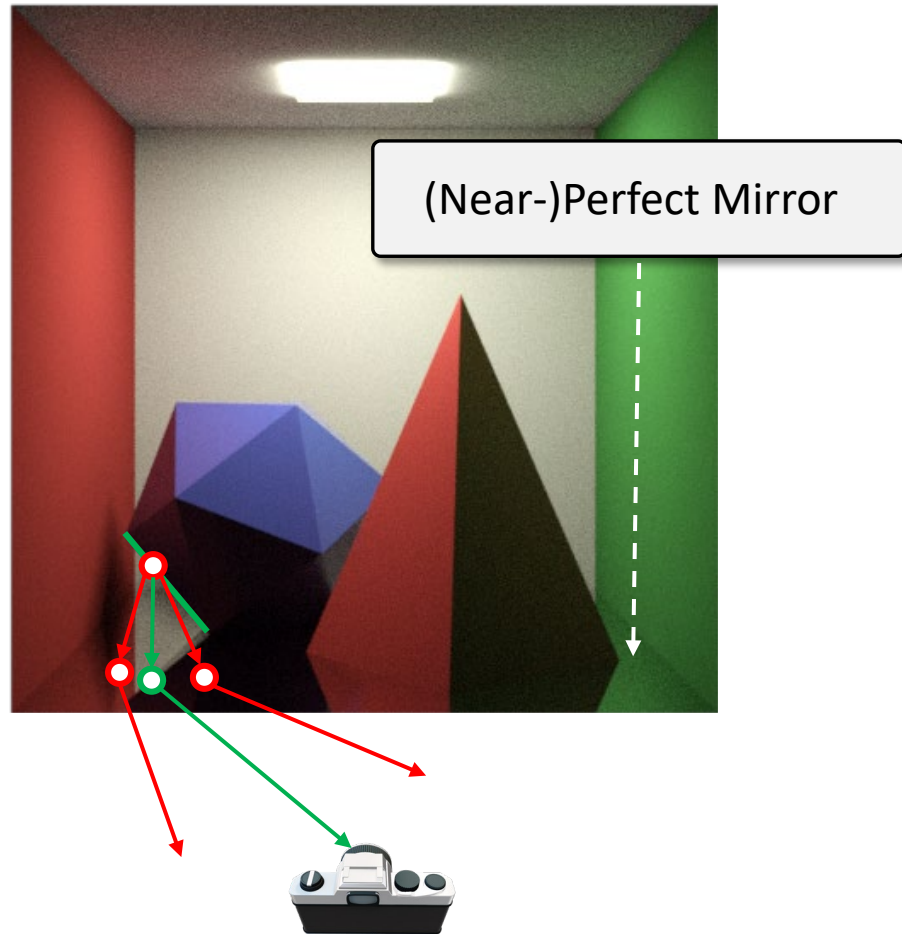
EDGE-SAMPLING

Challenges for Edge-Sampling

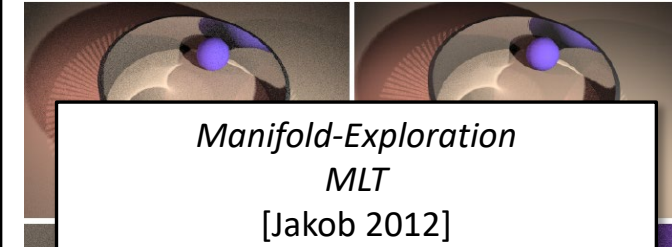
Arbitrary silhouette sampling is *hard!*



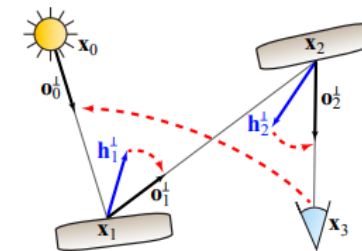
EDGE-SAMPLING HAS TROUBLE WITH SPECULAR REFLECTIONS



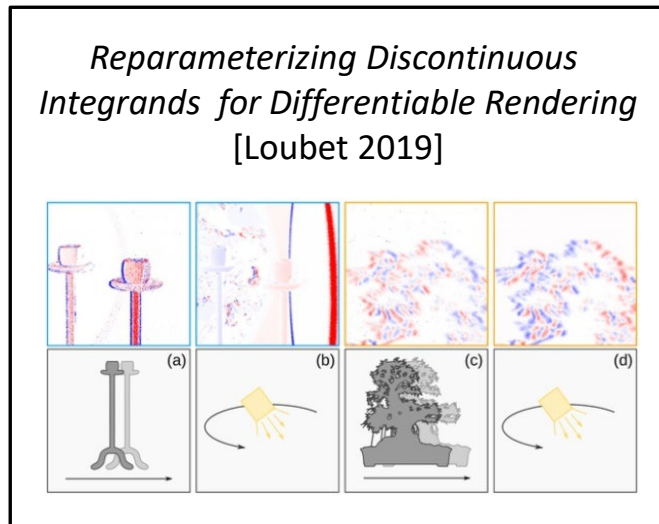
Rendering Caustics



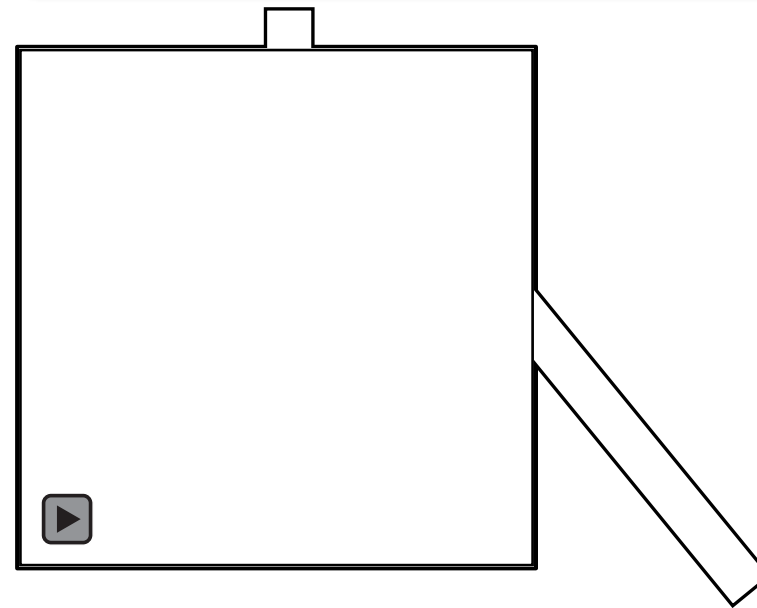
*Natural Constraint Representation
for MLT*
[Kaplanyan 2014]



AREA-SAMPLING



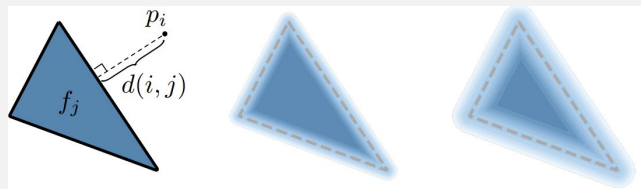
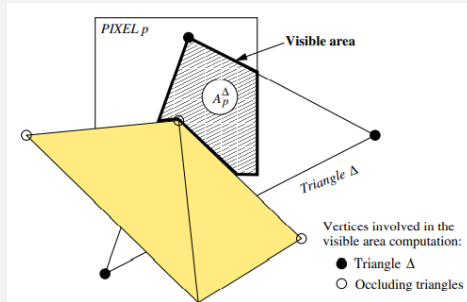
Transform samples with θ . Avoids discontinuities.



Heuristic Approximation!
May not work for all samples.

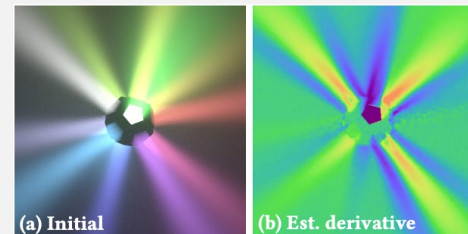
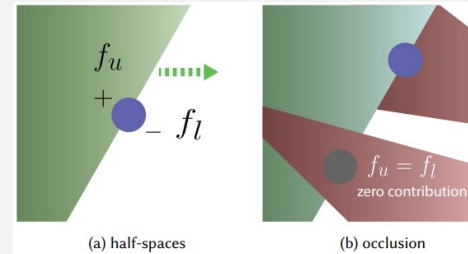
SUMMARY OF METHODS

Rasterization



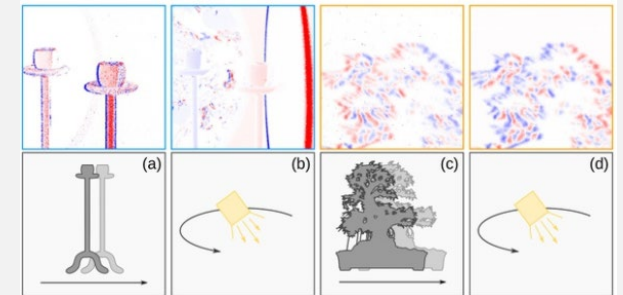
- + Fast
- Approximate visibility
- No secondary effects

Edge-sampling



- + Exact derivative
- Depth complexity
- No perfect specularities
- Complex data structures

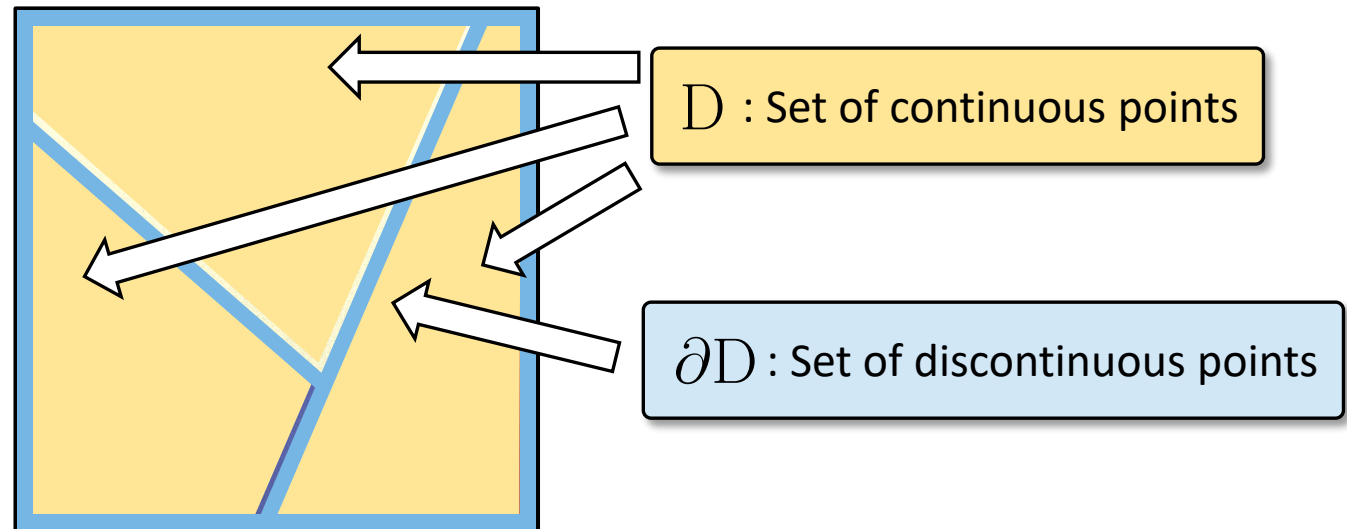
Area-sampling



- + Fast (No complex sampling)
- Approximate derivative

OUR APPROACH

THE REYNOLDS TRANSPORT THEOREM



$$\partial_\theta \int_D f = \int_D \partial_\theta f + \int_{\partial D} f \vec{v} \cdot \vec{n}$$

Interior term Edge term

CONVERTING EDGE-SAMPLES TO AREA-SAMPLES

$$\int_{\partial D} f \vec{v} \cdot \vec{n}$$

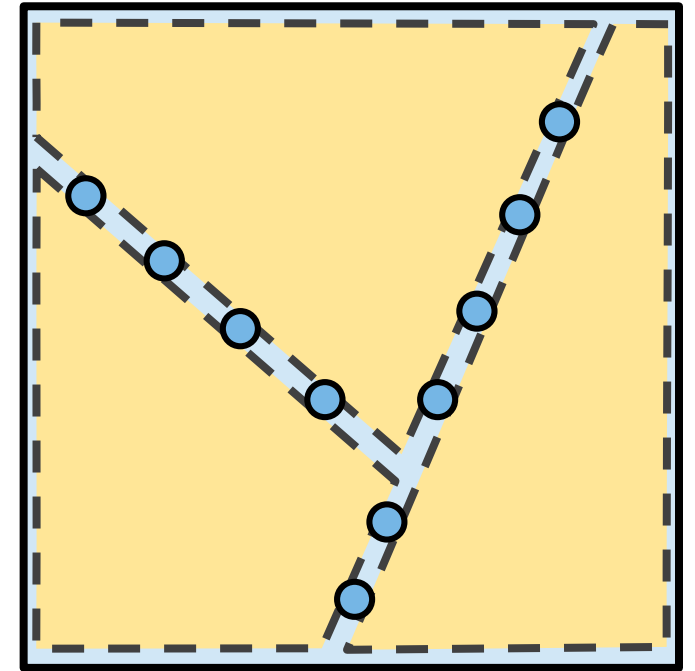
is estimated through edge-samples ●

Goal: Rewrite

$$\int_{\partial D} f \vec{v} \cdot \vec{n}$$

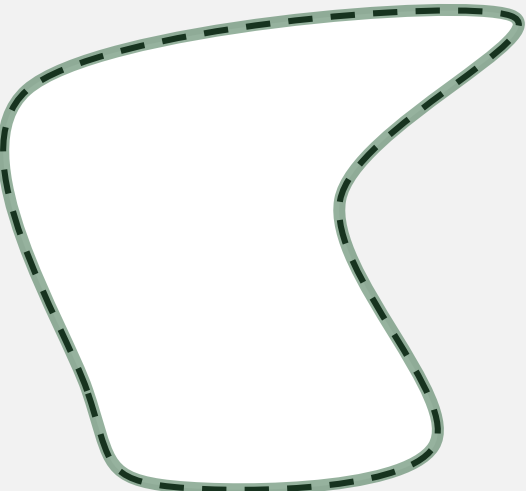
into area integral

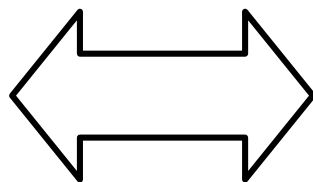
$$\int_D g$$

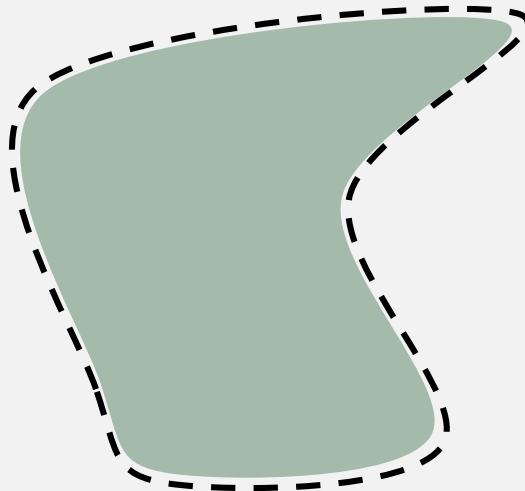


THE DIVERGENCE THEOREM

[Gauss 1813]


$$\int_{\partial D} \vec{f} \cdot \vec{n}$$




$$\int_D \nabla \cdot \vec{f}$$

APPLYING THE DIVERGENCE THEOREM TO THE EDGE INTEGRAL

Goal: Rewrite

$$\int_{\partial D} f \vec{v} \cdot \vec{n}$$

into area integral

$$\int_D g$$

Solution: Rewrite

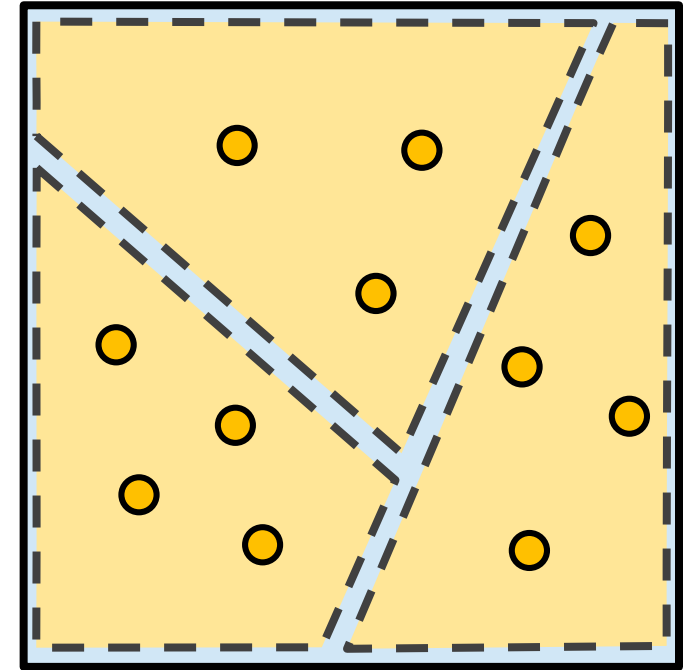
$$\int_{\partial D} f \vec{v} \cdot \vec{n}$$

into

$$\int_D \nabla \cdot (\vec{v}_\theta f)$$

$$\int_D \nabla \cdot (\vec{v}_\theta f)$$

can be estimated through area-samples ●



QUICK RECAP

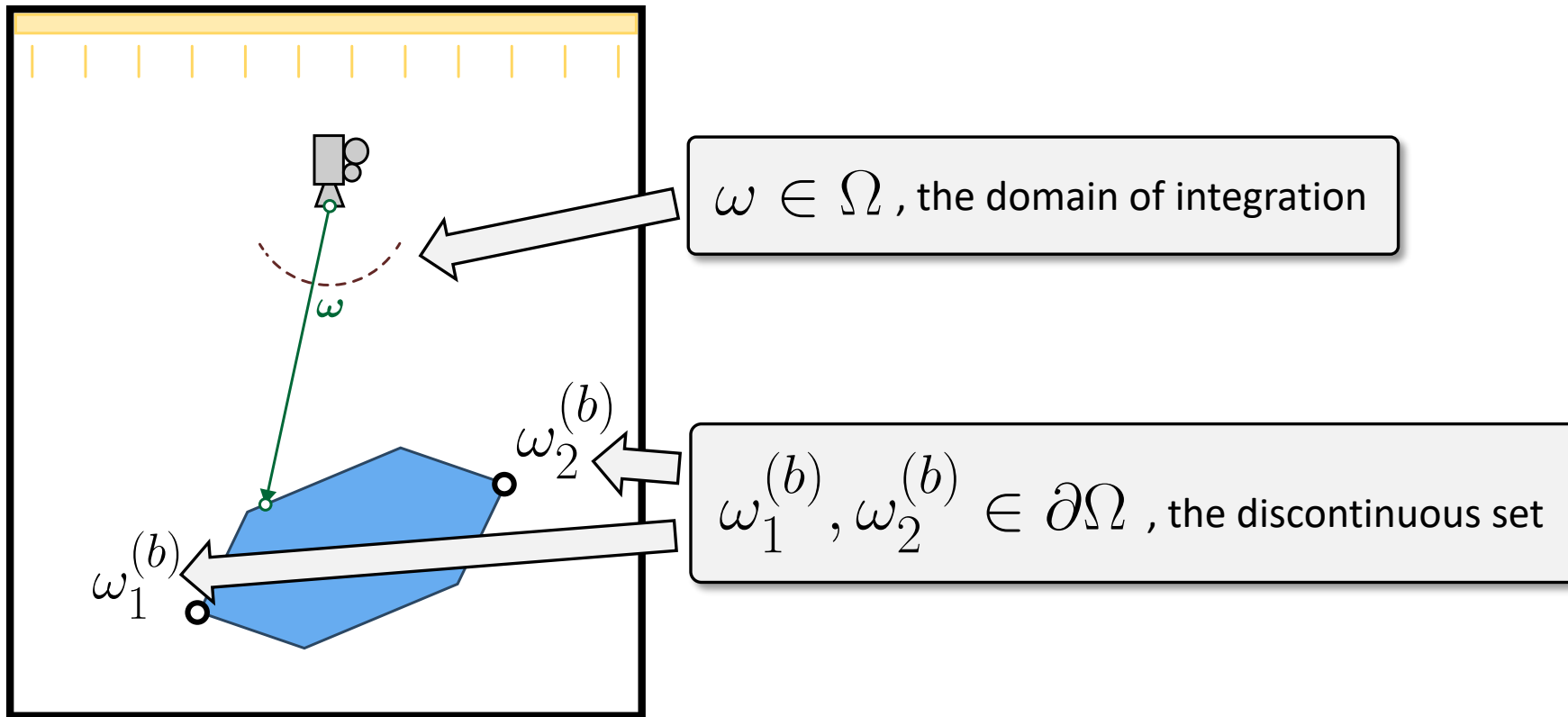
- Used *Reynolds transport theorem* to find the boundary integral

$$\int_{\partial D} f \vec{v} \cdot \vec{n}$$

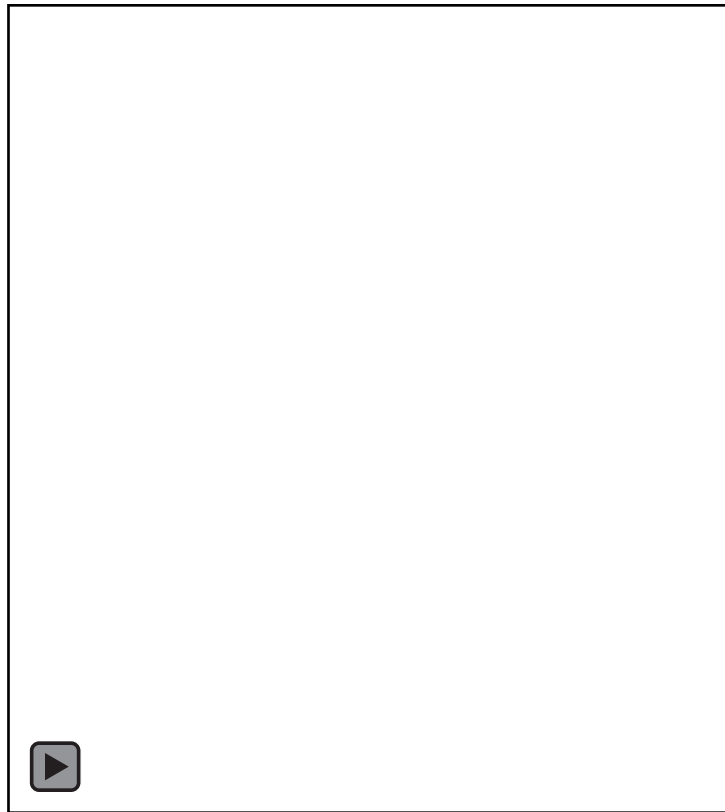
- Rewrote $\int_{\partial D} f \vec{v} \cdot \vec{n}$ to $\int_D \nabla \cdot (\vec{v}_\theta f)$ using the *divergence theorem*.

- Have to define the *vector field* \vec{v}_θ over domain D

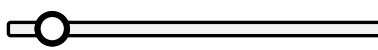
A 2D EXAMPLE SCENE



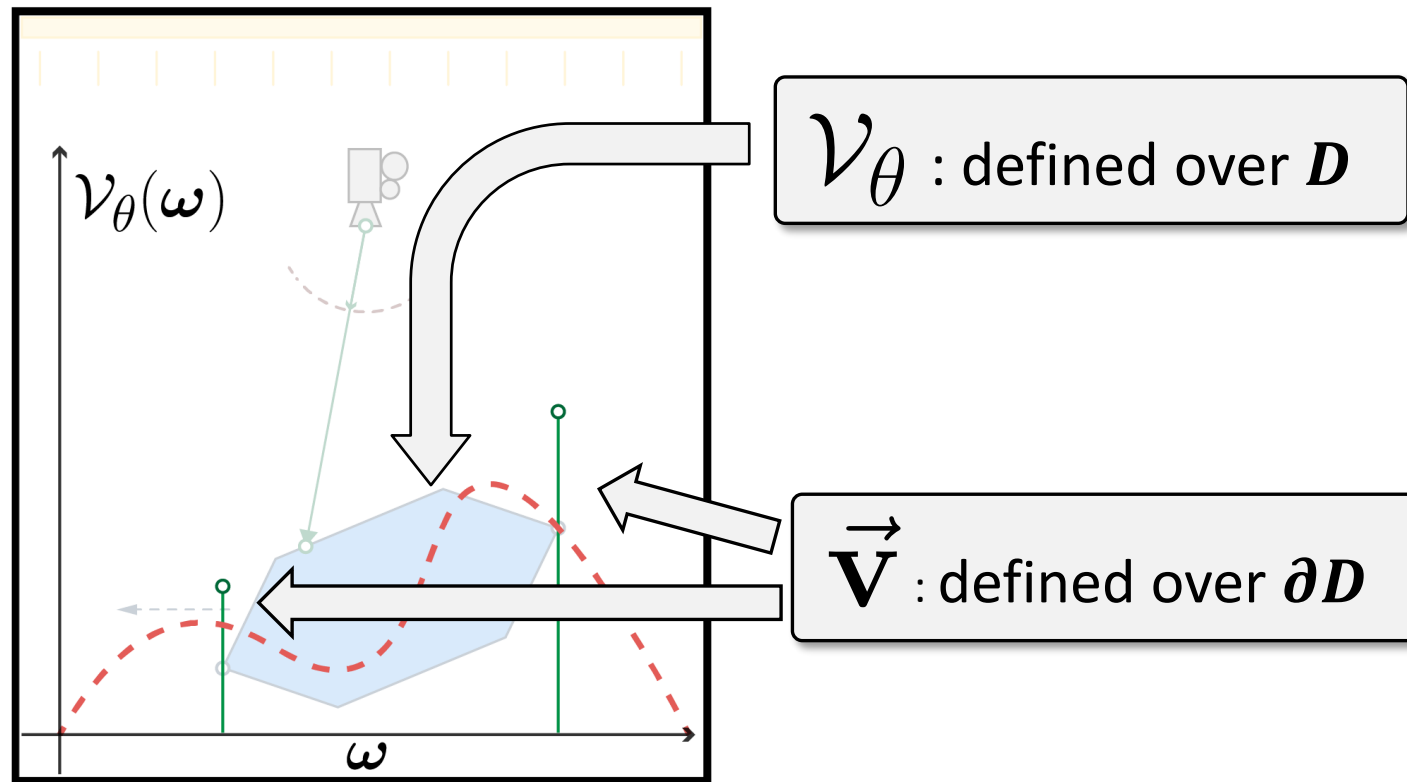
VELOCITY \vec{V} : THE BOUNDARY DERIVATIVE



$\partial_{\theta} \omega_i^{(b)}$: Derivative of boundary position w.r.t θ

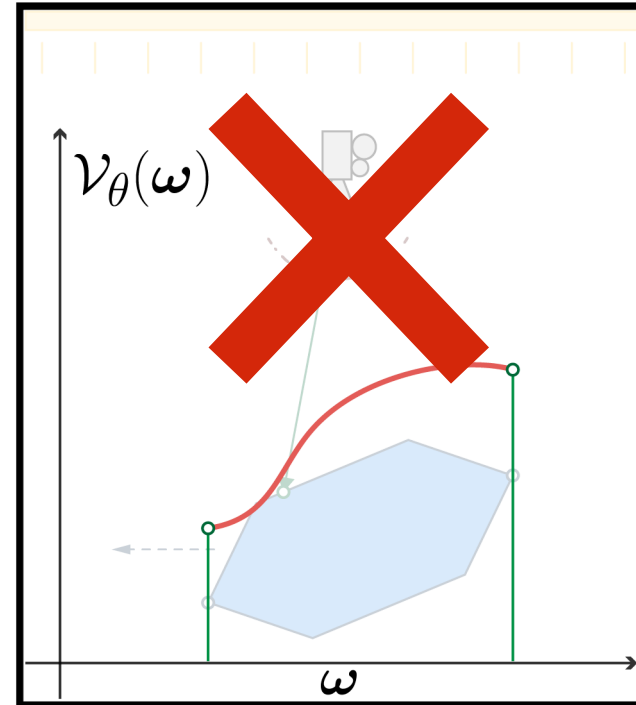
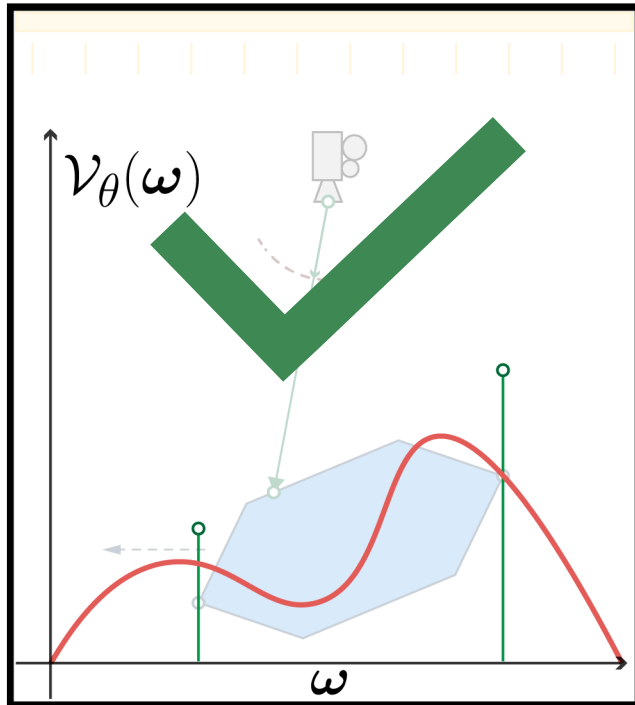
θ 

WARP FIELD \mathcal{V}_θ : EXTENSION OF $\vec{\mathbf{V}}$ TO ALL POINTS



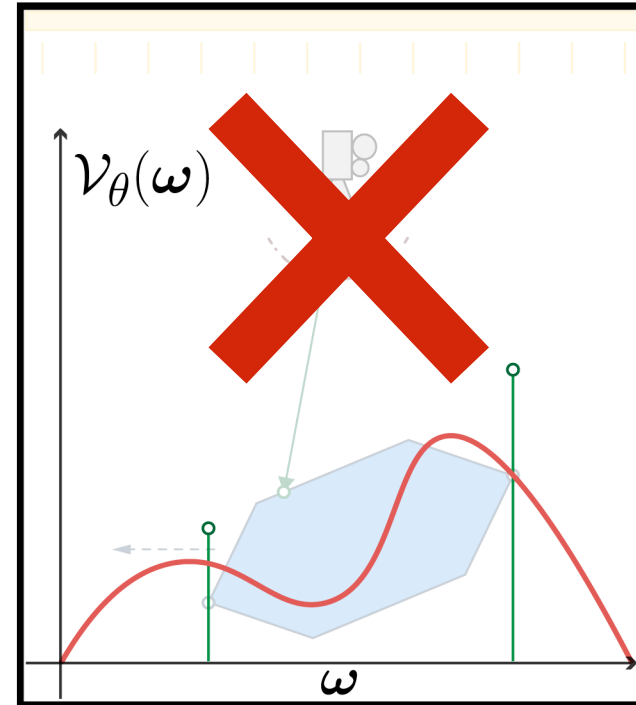
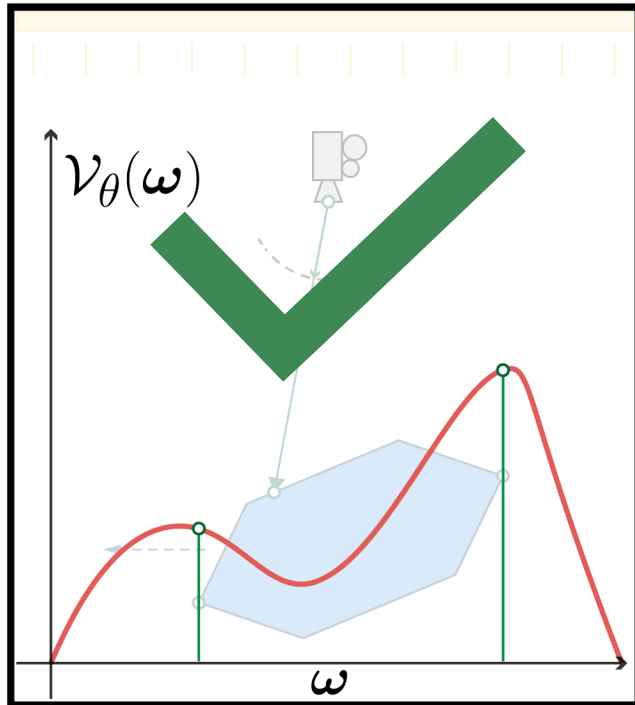
VALIDITY OF $\vec{\mathcal{V}}_\theta$

Rule 1: Continuous

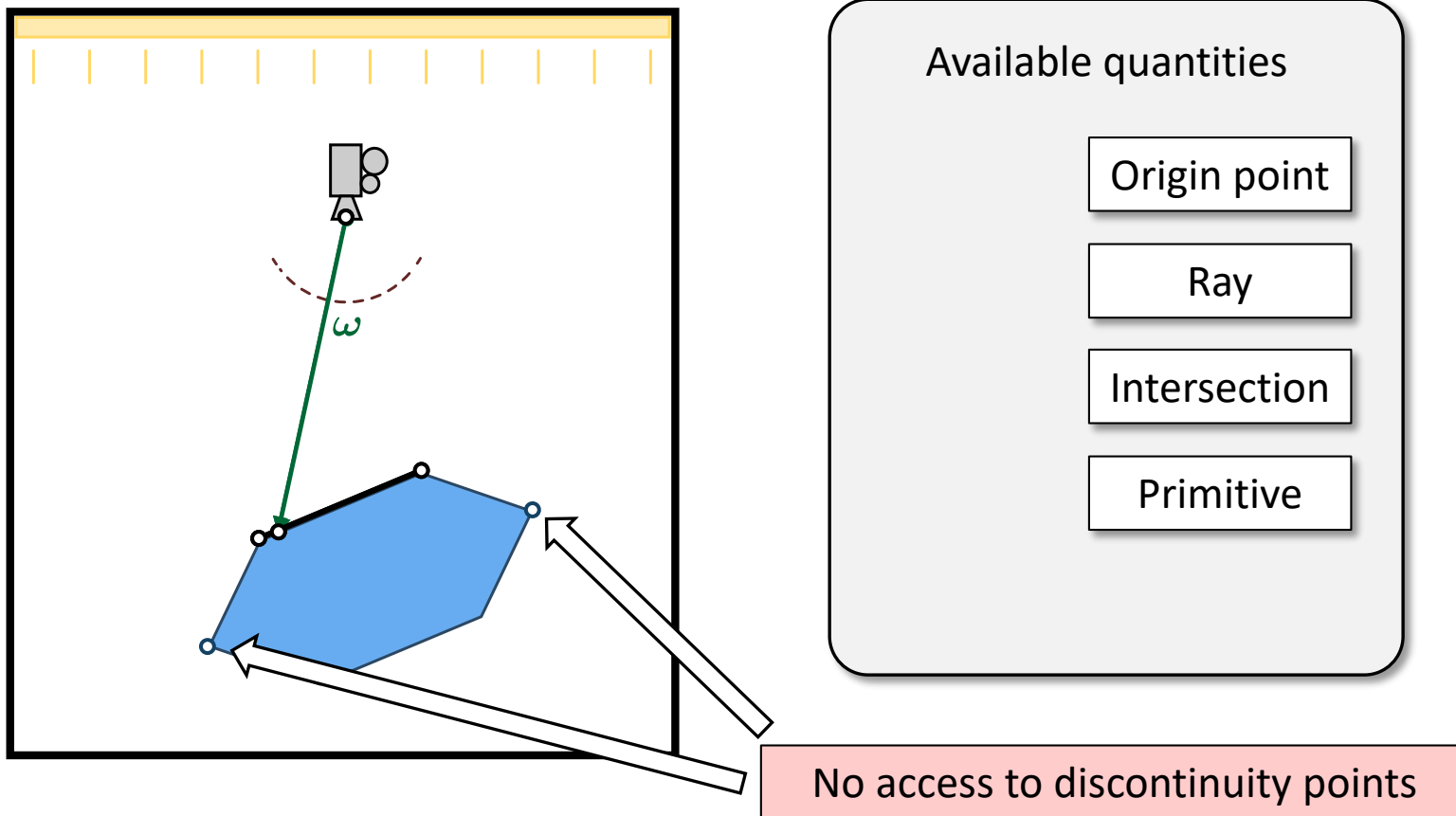


VALIDITY OF $\vec{\mathcal{V}}_\theta$

Rule 2: Boundary Consistent



INTERPOLATION WITHOUT KNOWLEDGE OF BOUNDARIES



CONSTRUCTING \vec{V}_θ

Attempt 1 \longrightarrow Find $\partial_\theta \omega$ through *implicit derivative*

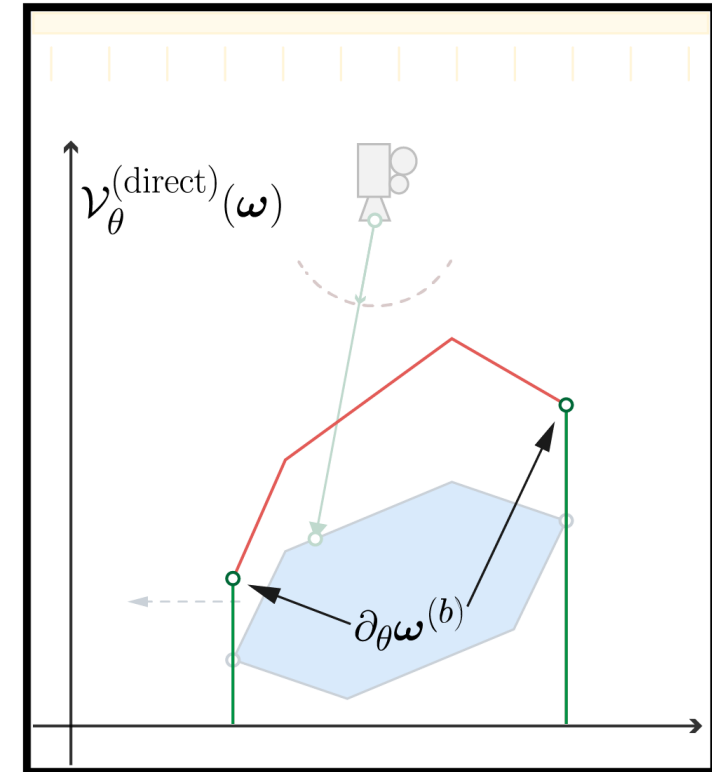
(Incorrect)

$$\mathbf{y} = \text{INTERSECT}(\omega, \theta) \implies \partial_\theta \omega = \frac{\partial_\omega \mathbf{y}}{\partial_\theta \mathbf{y}}$$

At all points (not just boundaries)

+ Boundary consistent

- Not continuous



CONSTRUCTING $\vec{\mathcal{V}}_{\theta}$

Attempt 2 \longrightarrow Filter *Attempt 1* with a Gaussian filter

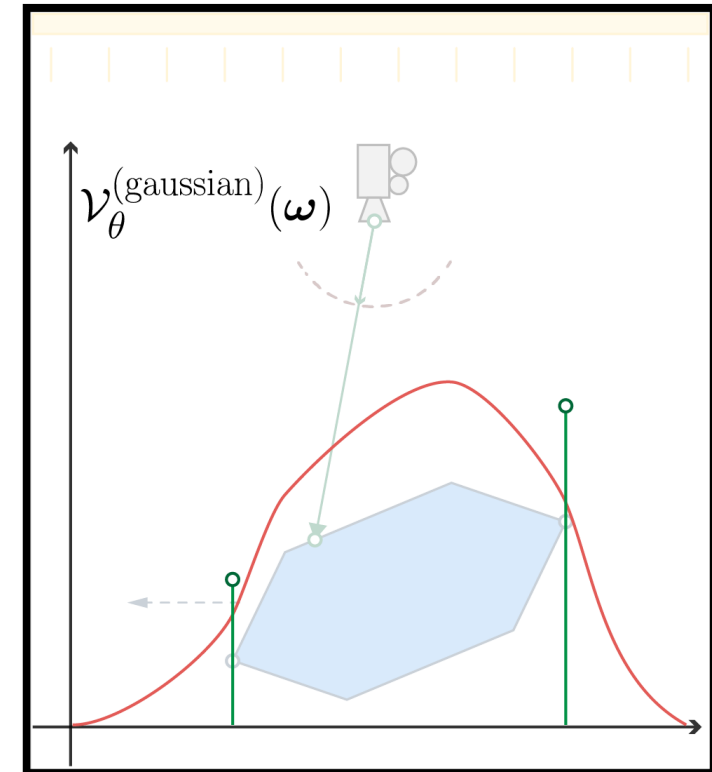
(Incorrect)

$$\int_{\Omega'} k(\omega, \omega') \frac{\partial_{\omega} \mathbf{y}}{\partial_{\theta} \mathbf{y}}$$

$k(.,.) = \text{Gaussian filter}$

+ Continuous

- Not boundary consistent



BOUNDARY-AWARE WEIGHTING

Goal: Find weights $k(\omega, \omega')$ s.t. $\vec{V}_\theta = \frac{\partial_\omega y}{\partial_\theta y}$ at boundaries.

Ideal weighting function



ω

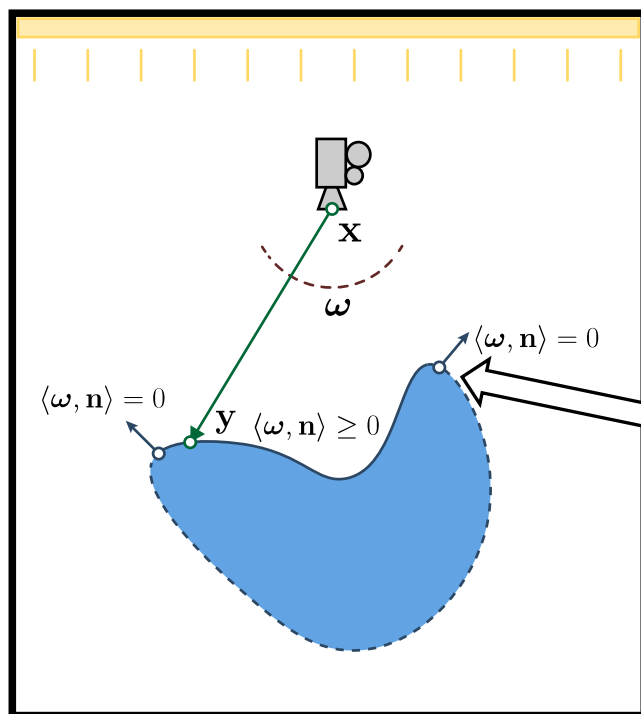


Approach Dirac delta near boundaries



BOUNDARY-AWARE WEIGHTING

Implicit Boundary through *geometric normals*



$\langle \omega, \mathbf{n} \rangle = 0$
at boundaries

ry sampling)

boundary
g)

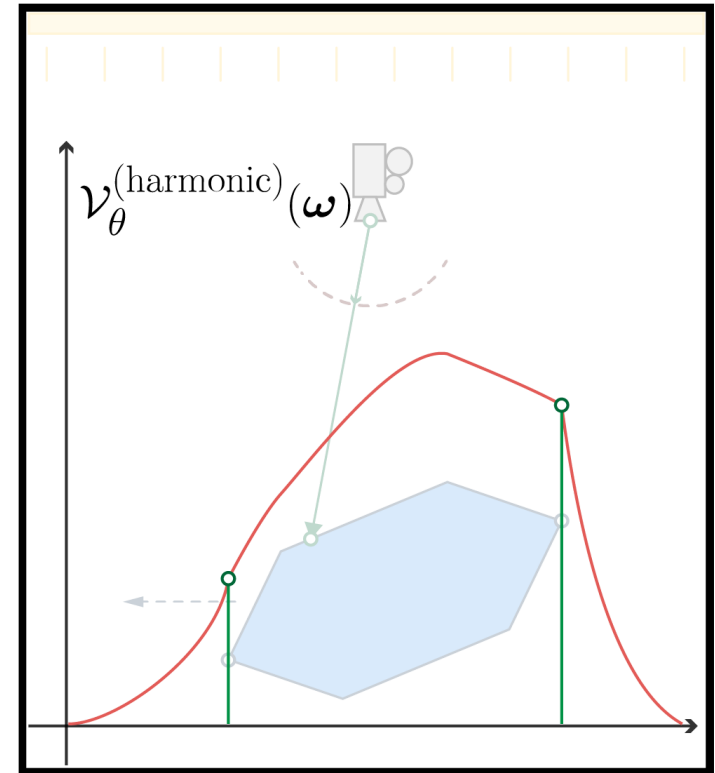
CONSTRUCTING $\vec{\mathcal{V}}_{\theta}$

Our Approach \longrightarrow Filter *Attempt 1* with harmonic weights

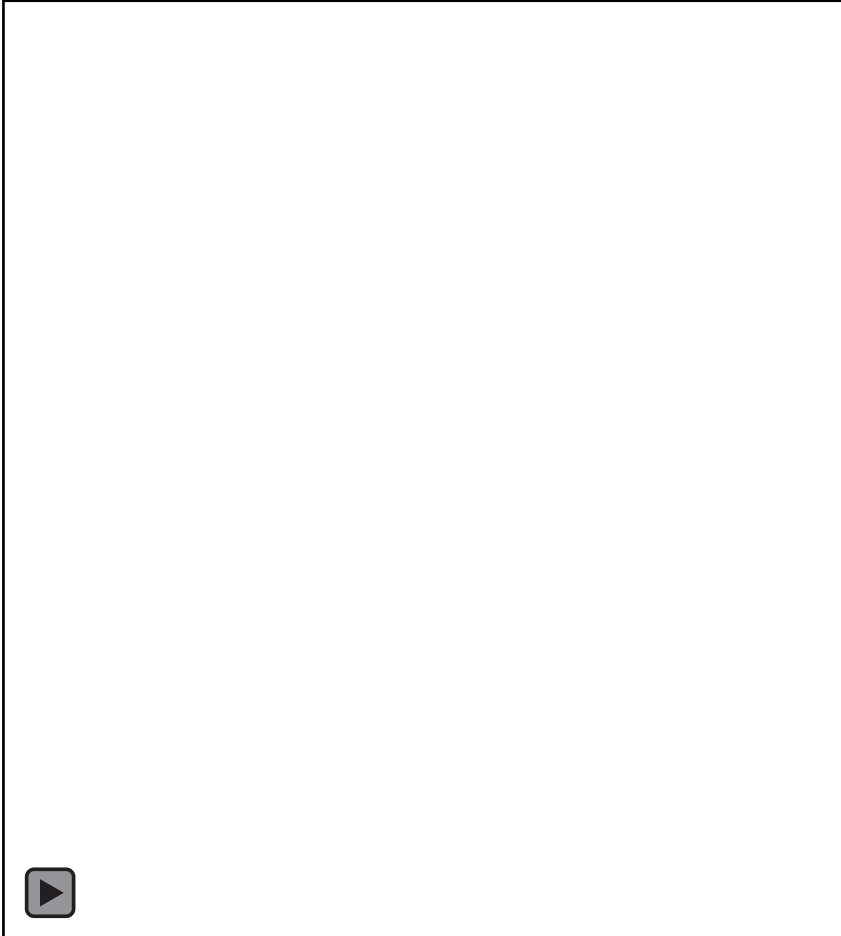
$$k(\omega, \omega') = \frac{1}{\boxed{\mathcal{D}(\omega, \omega')} + \boxed{\mathcal{B}(\omega')}}}$$

Distance function Boundary test

+ Boundary consistent
+ Continuous



COMPUTING \vec{V}_θ



1. Sample **path** using path tracer *(N paths)*

For each bounce:

→ 2. Sample **auxiliary** rays *(N' rays)*

3. Compute boundary term ***B()*** locally

4. Compute weight ***k(.,.)*** and $\partial_\theta \omega$

5. Find weighted mean

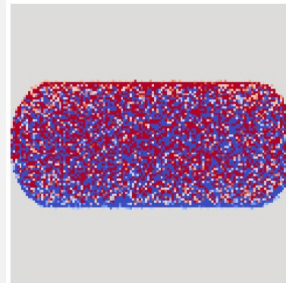
ADDITIONAL DETAILS

Russian Roulette

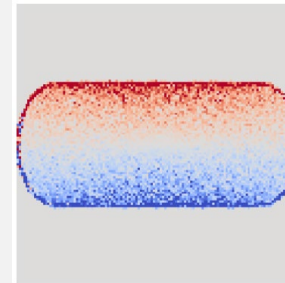
$$\frac{\sum k(\omega, \omega'_i) g(\omega'_i)}{\sum k(\omega, \omega'_i)}$$

$$N' \sim \text{GEOM}(p)$$

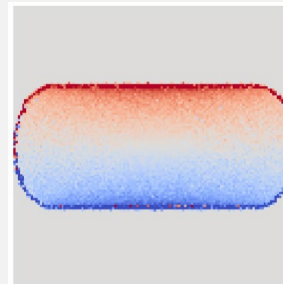
Variance Reduction



No Variance Reduction



Antithetic Variates



Antithetic Variates +
Control Variates

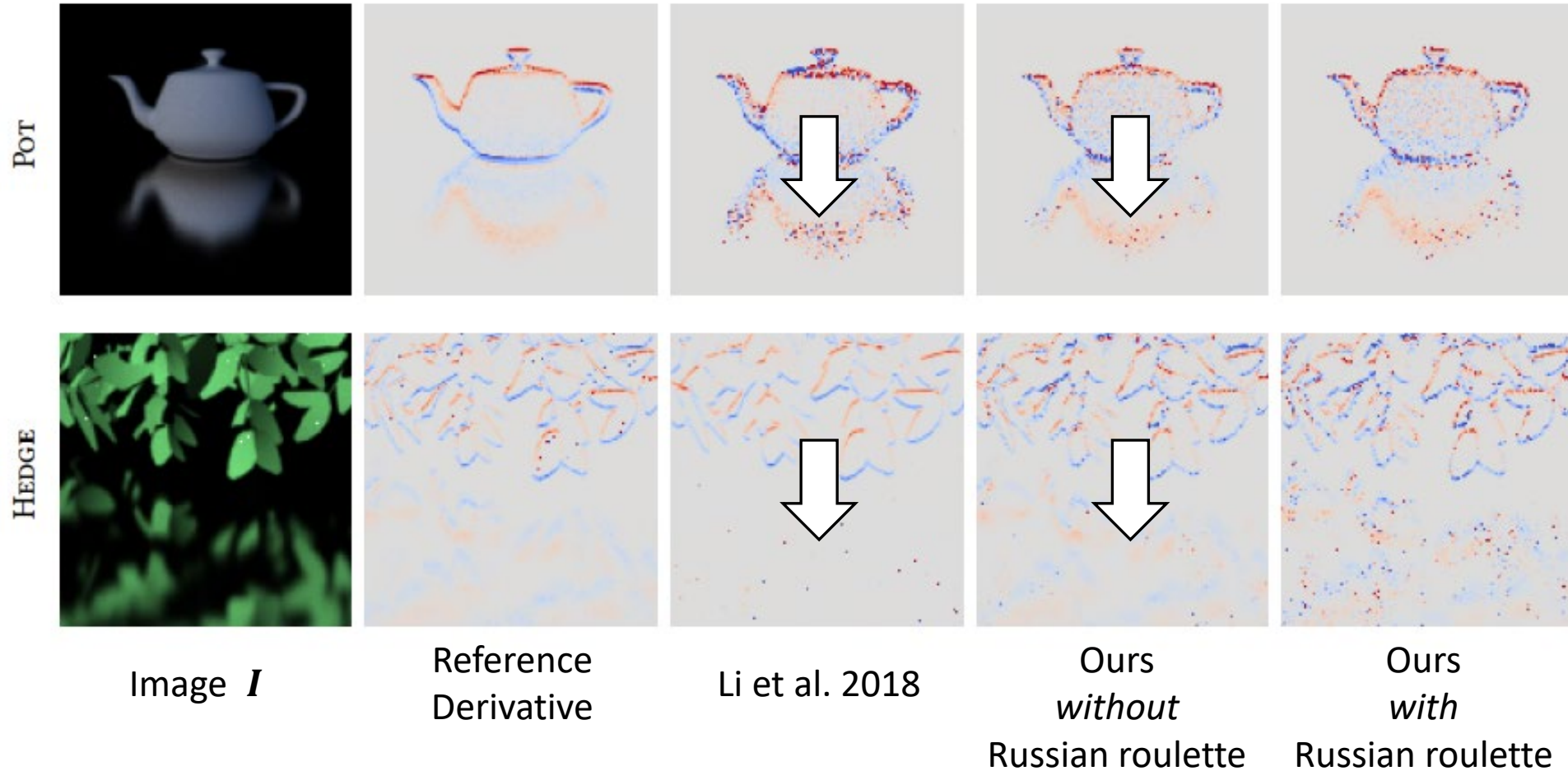
Relationship with Reparameterization

$$\mathcal{V}_\theta(\omega) \longleftrightarrow \mathcal{T}(\omega; \theta)$$

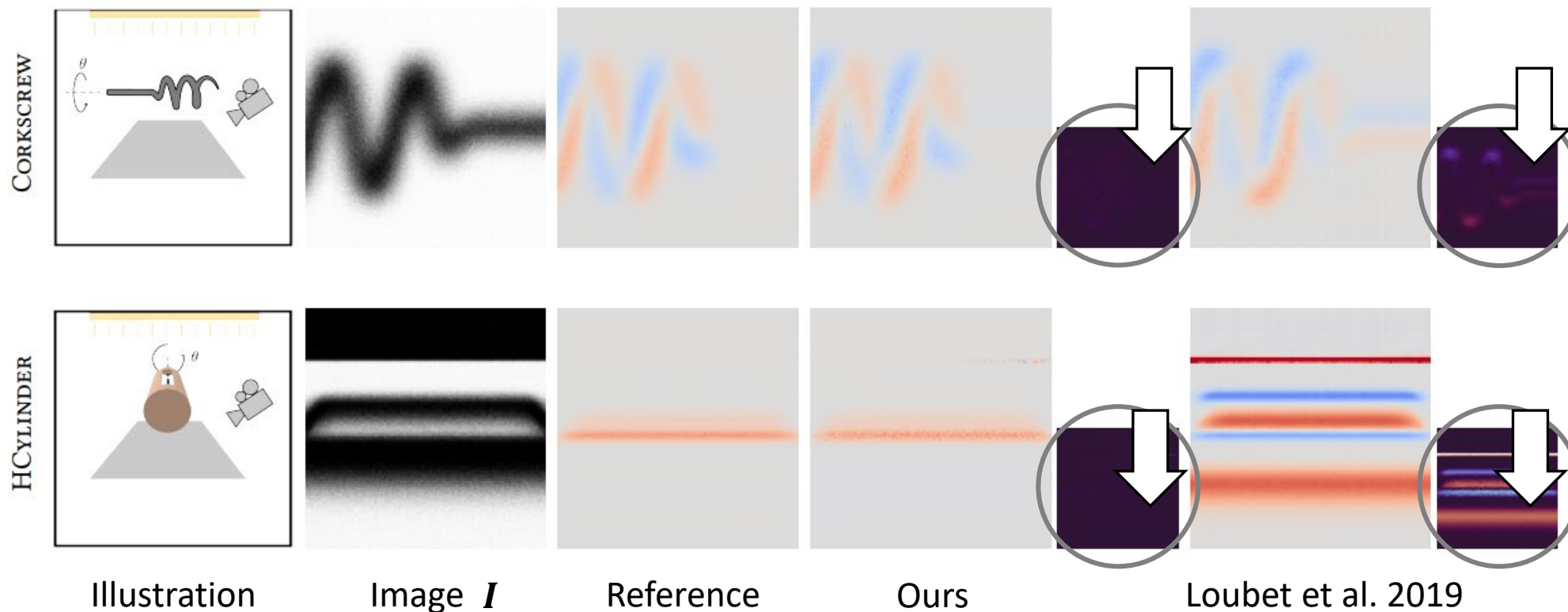
$$\mathcal{V}_\theta(\omega) = [\partial_\theta \mathcal{T}(\omega; \theta)]_{\theta=\theta_0}$$

RESULTS

VARIANCE COMPARISON WITH EDGE-SAMPLING

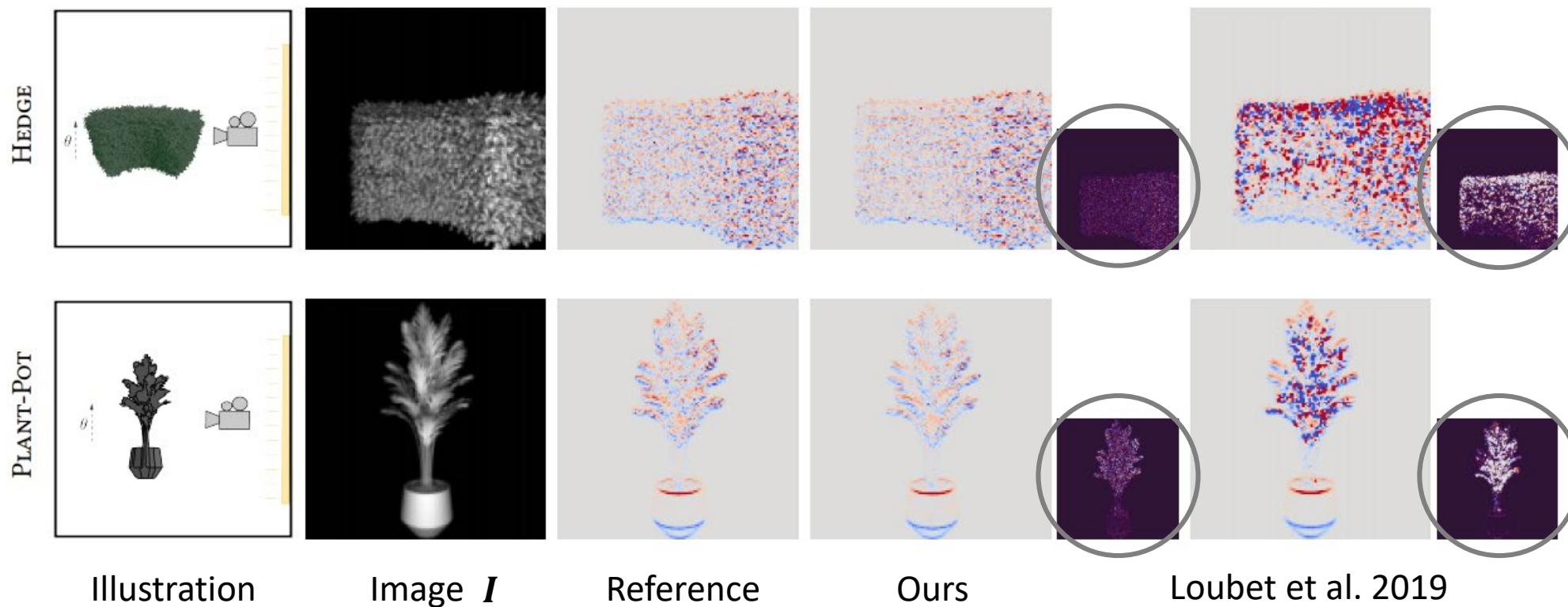


BIAS COMPARISON WITH REPARAMETERIZATION



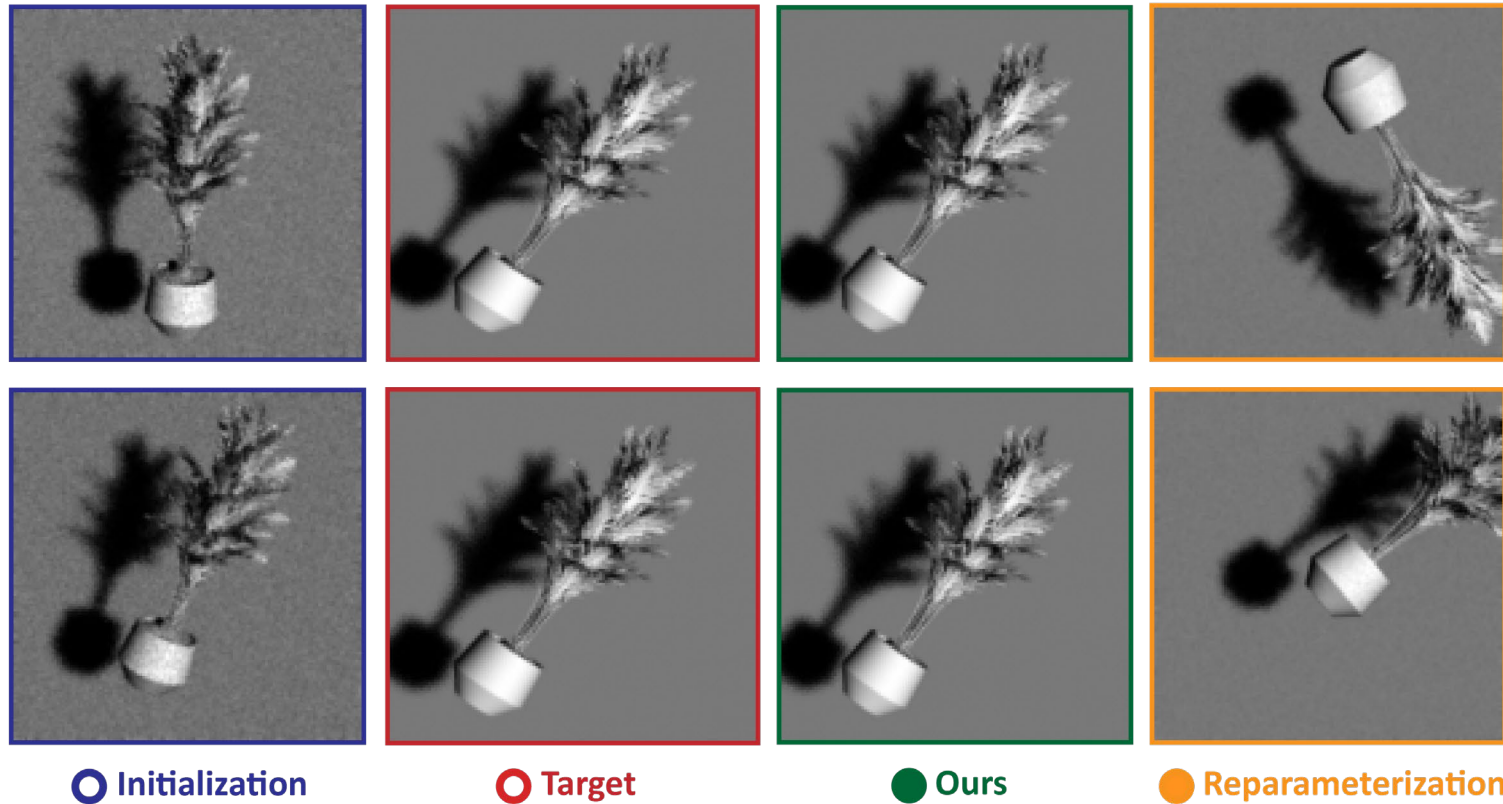
Rotating cylindrical objects present a complicated scenario for area-sampling

BIAS COMPARISON WITH REPARAMETERIZATION

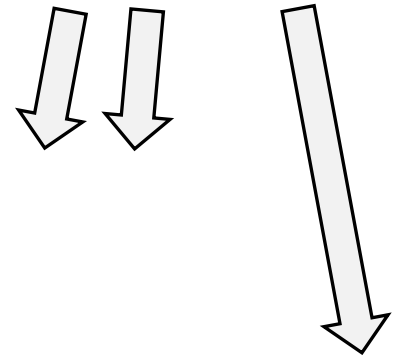


Extremely complex geometry like foliage can cause heuristic to fail

POSE ESTIMATION CAN FAIL WITH BIASED GRADIENTS

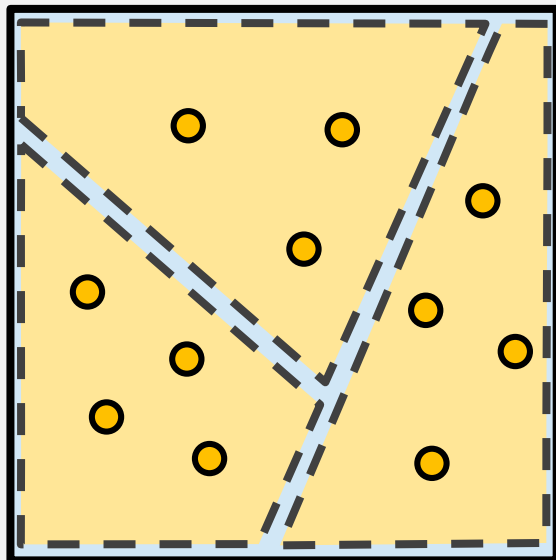


Multiple Initializations

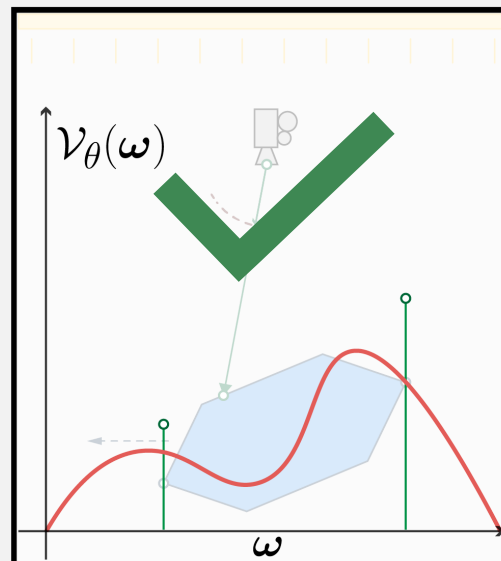


CONTRIBUTIONS

*Edge-integral to
Area-integral*

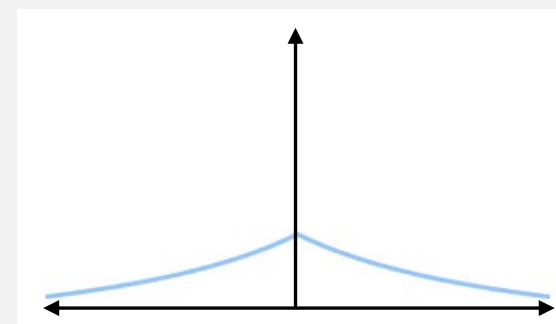


Warp field conditions



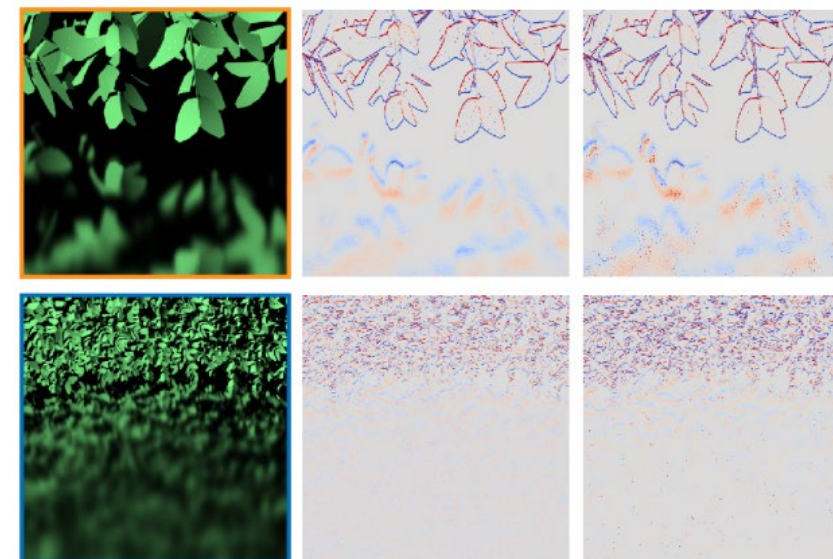
*Harmonic
interpolation*

$$k(\omega, \omega') = \frac{1}{\mathcal{D}(\omega, \omega') + \mathcal{B}(\omega')}$$



ACKNOWLEDGEMENTS

This project was funded by the **Toyota Research Institute**.

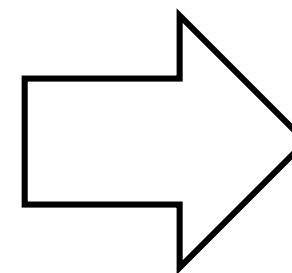


Highlighted Section

Ground Truth (FD)

Our Method

Project
Page



Code available on **redner**
Code *coming soon* on **mitsuba-2**