







Overview

Instant Radiosity

Virtual Point Lights

Virtual Ray Lights

LightCuts

Stochastic LightCuts

Lighting Grid Hierarchy

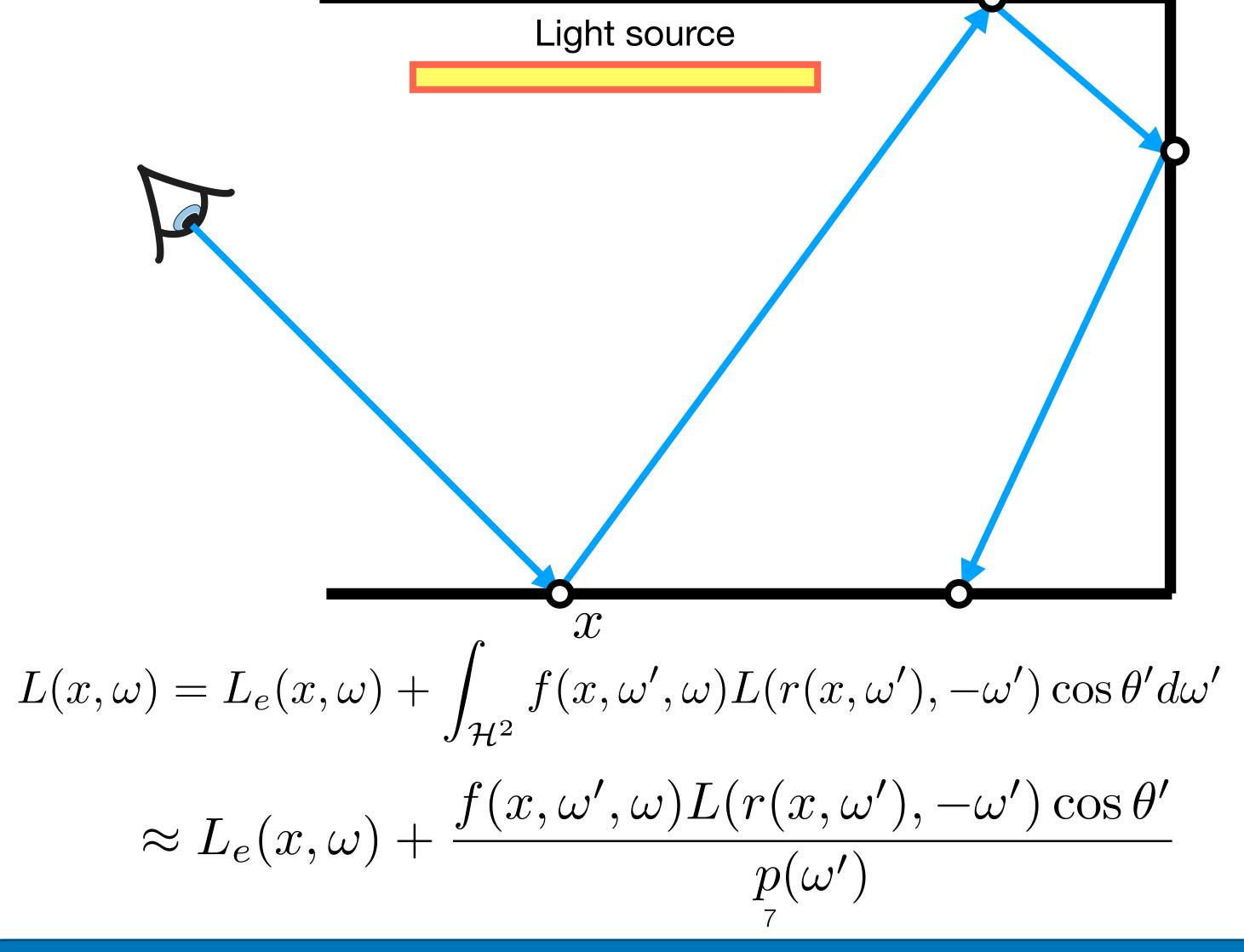
Light source



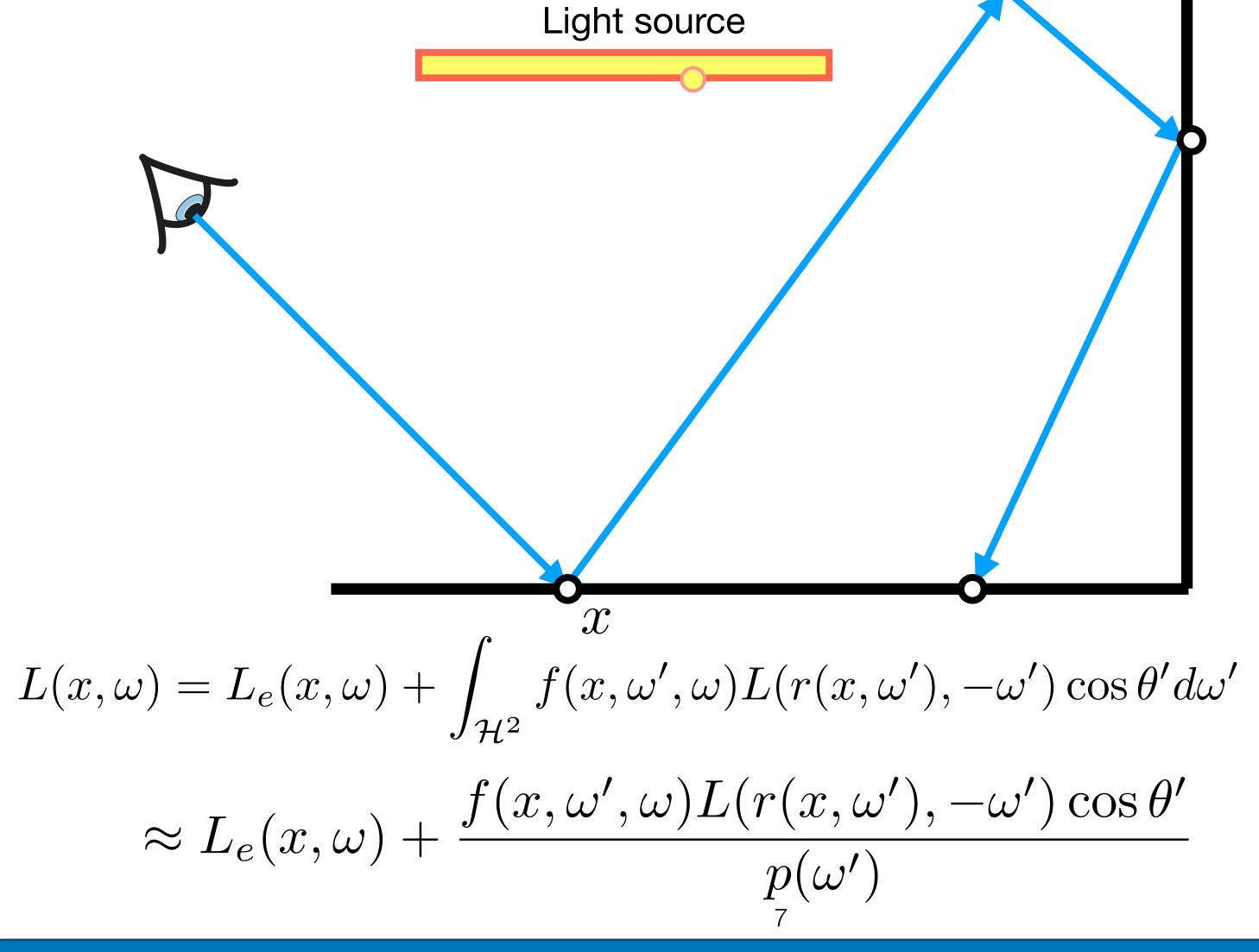
$$L(x,\omega) = L_e(x,\omega) + \int_{\mathcal{H}^2} \frac{x}{f(x,\omega',\omega)L(r(x,\omega'), -\omega')\cos\theta'd\omega'}$$

$$\approx L_e(x,\omega) + \frac{f(x,\omega',\omega)L(r(x,\omega'), -\omega')\cos\theta'}{p(\omega')}$$

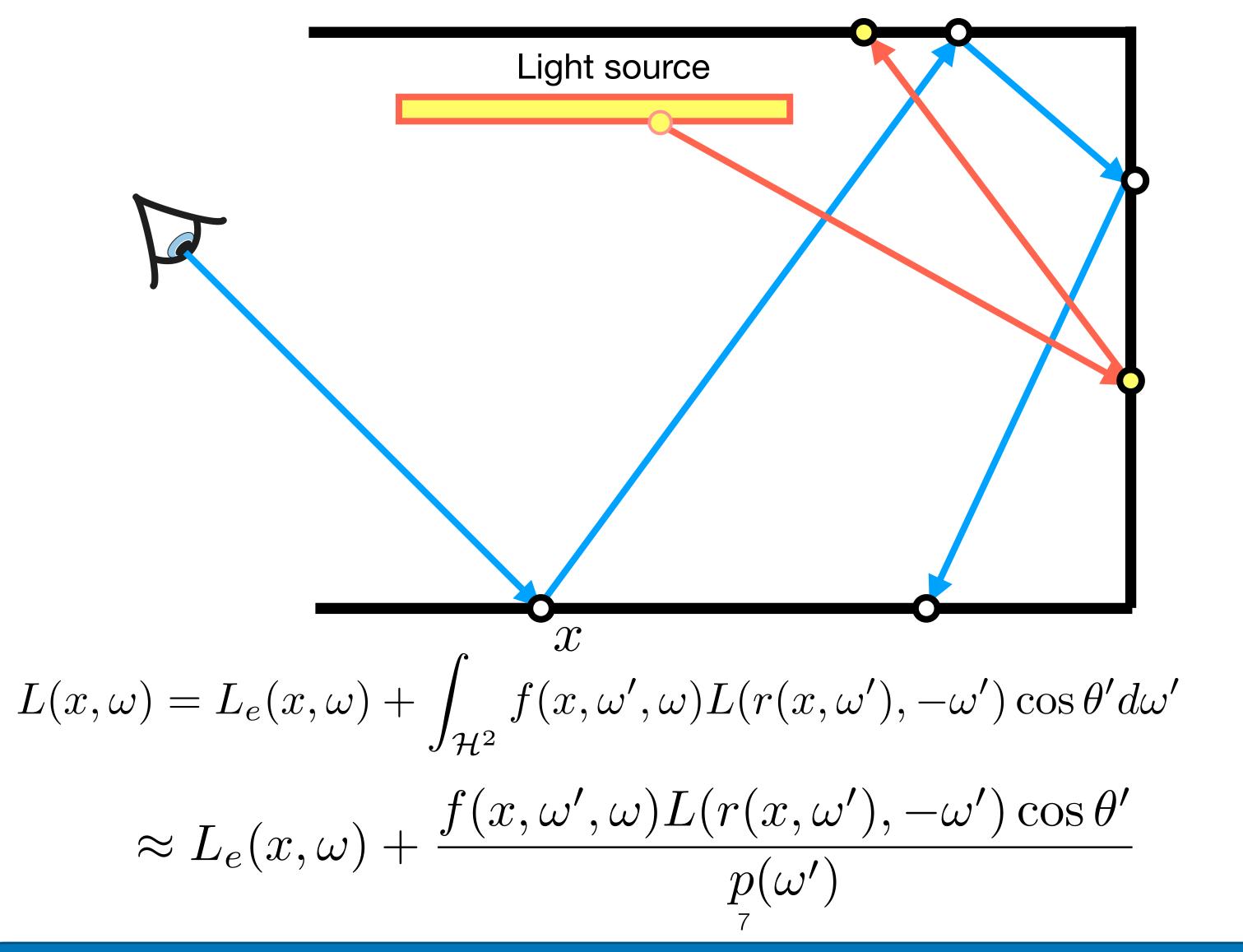




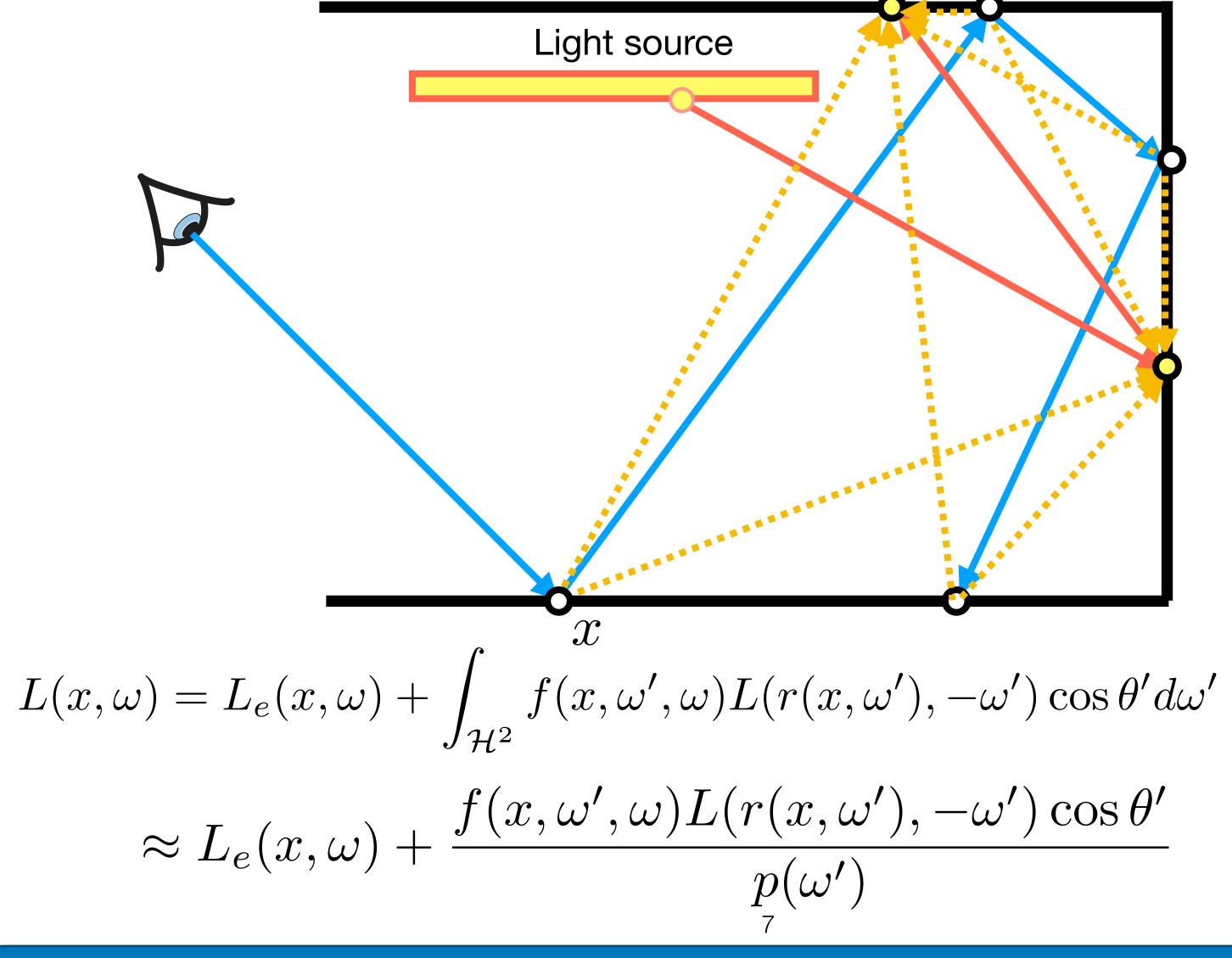














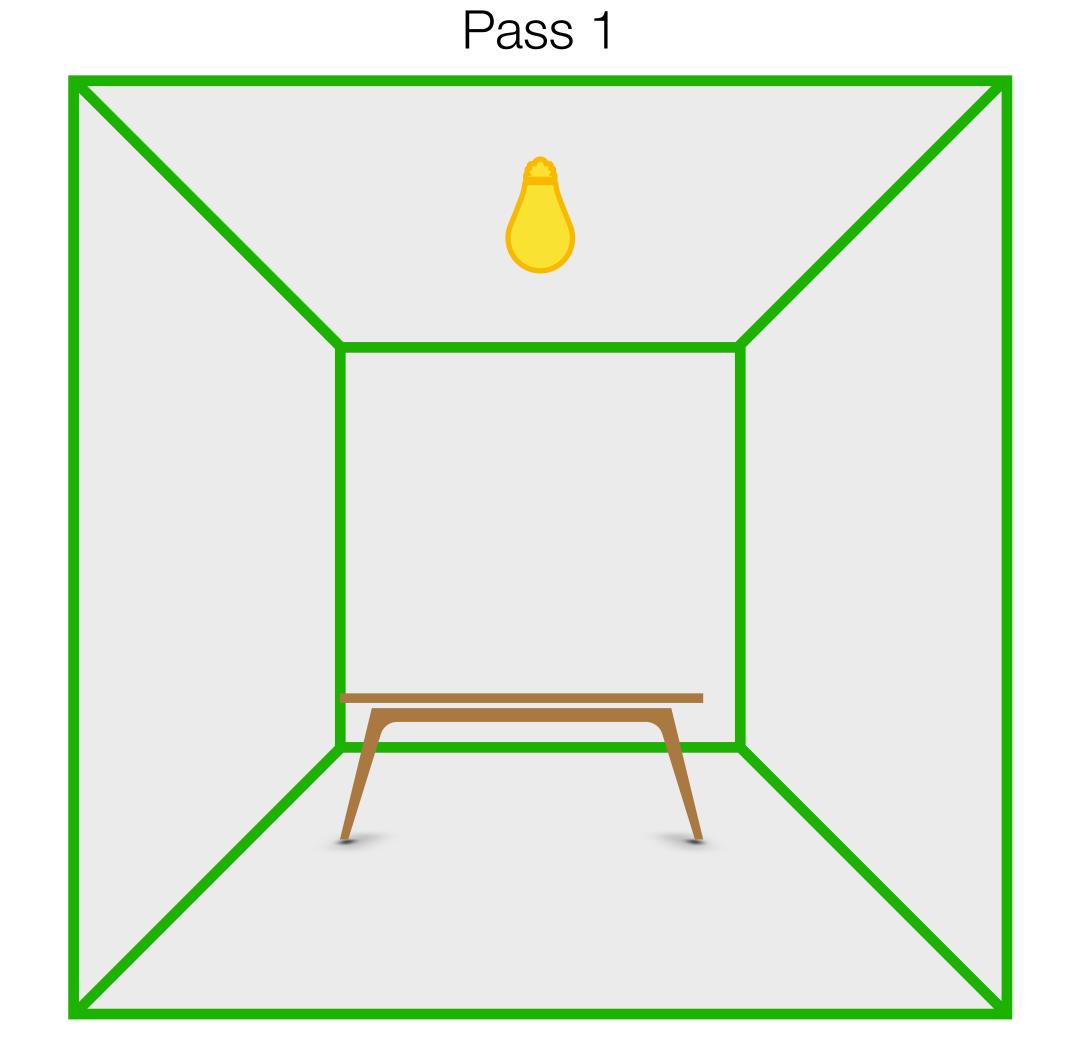
Instant Radiosity



Virtual Point Light: Generation

Step 1:

- Trace paths from light source(s)
- Treat path vertices as Vritual Point Lights (VPLs)

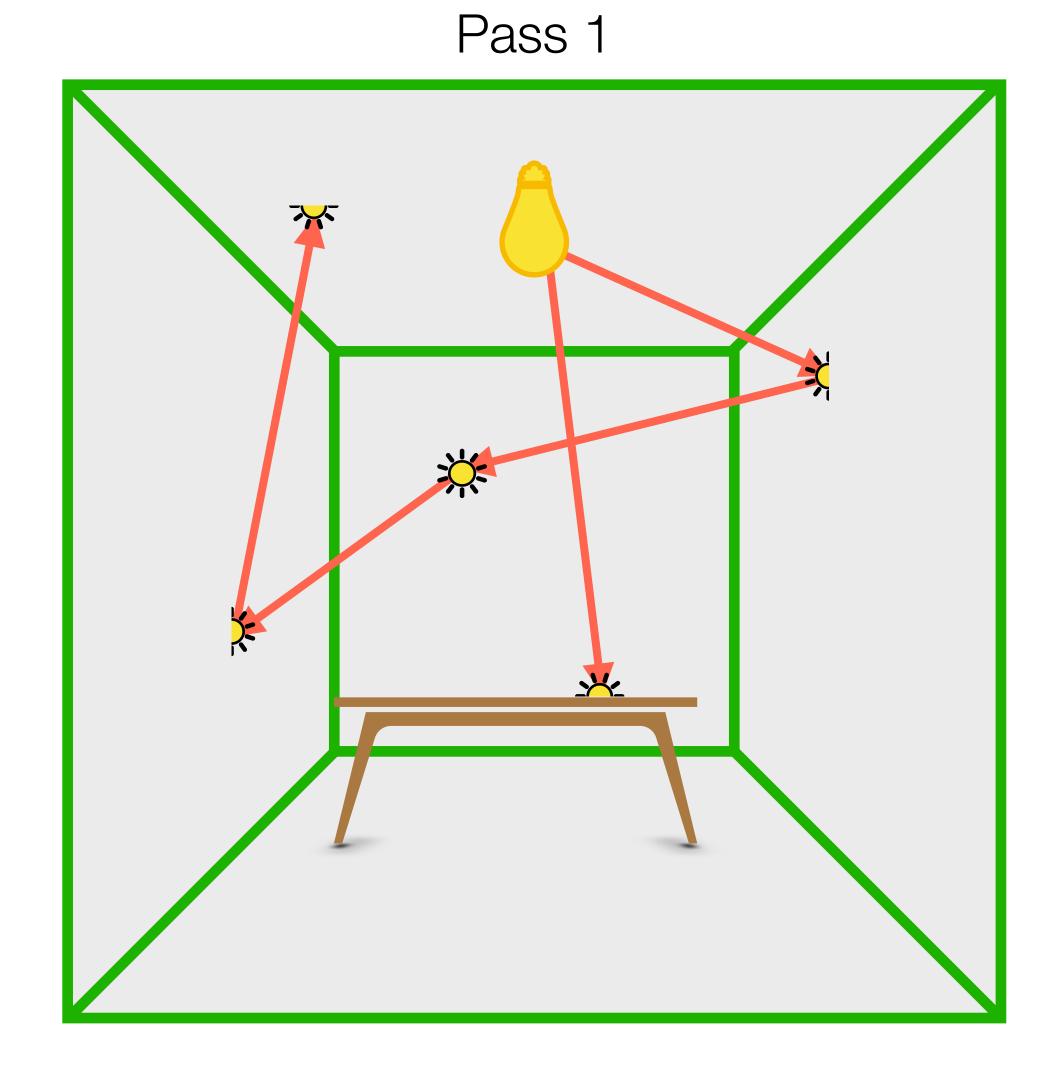




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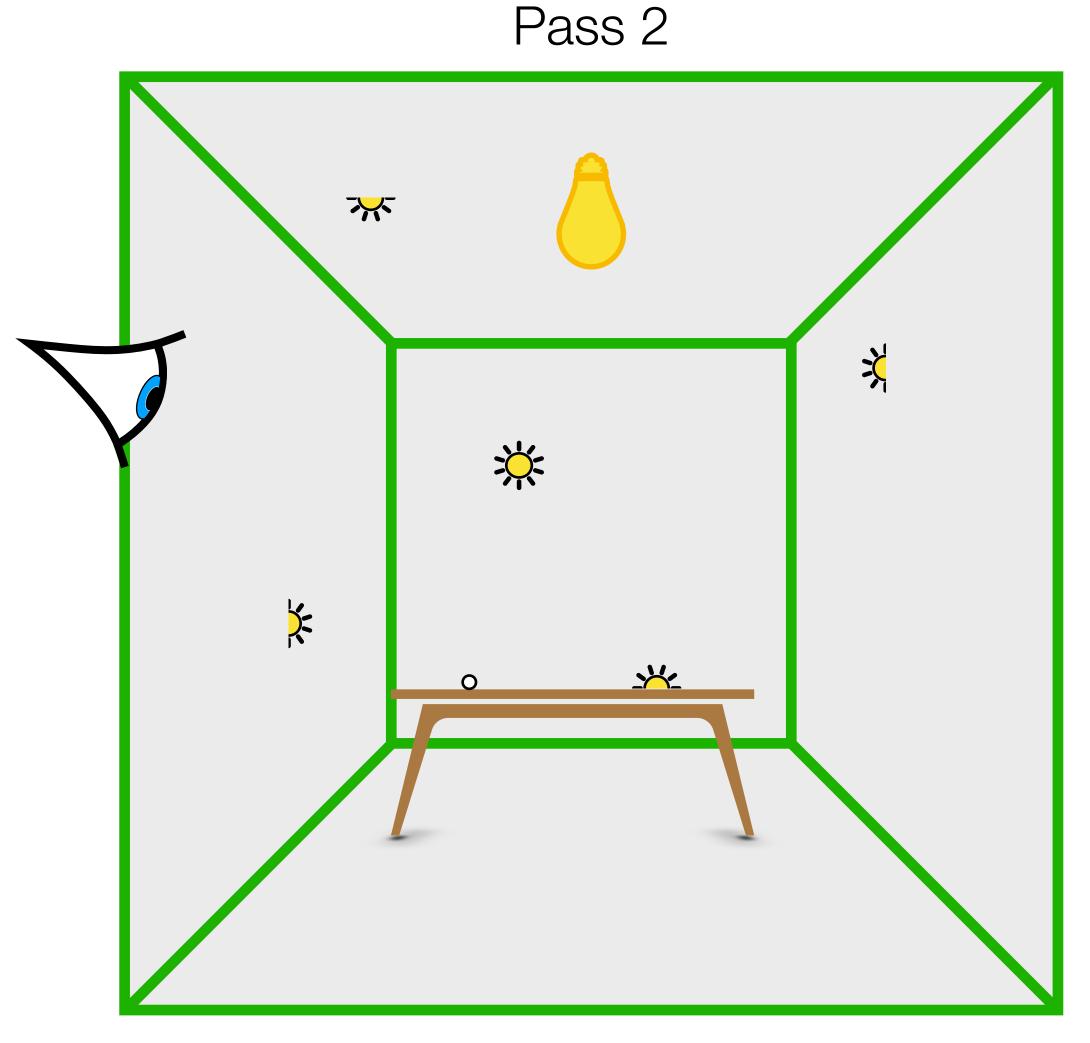
Virtual Point Light: Lighting

Step 1:

- Trace paths from light source(s)
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Step 2:

- Render scene with VPLs





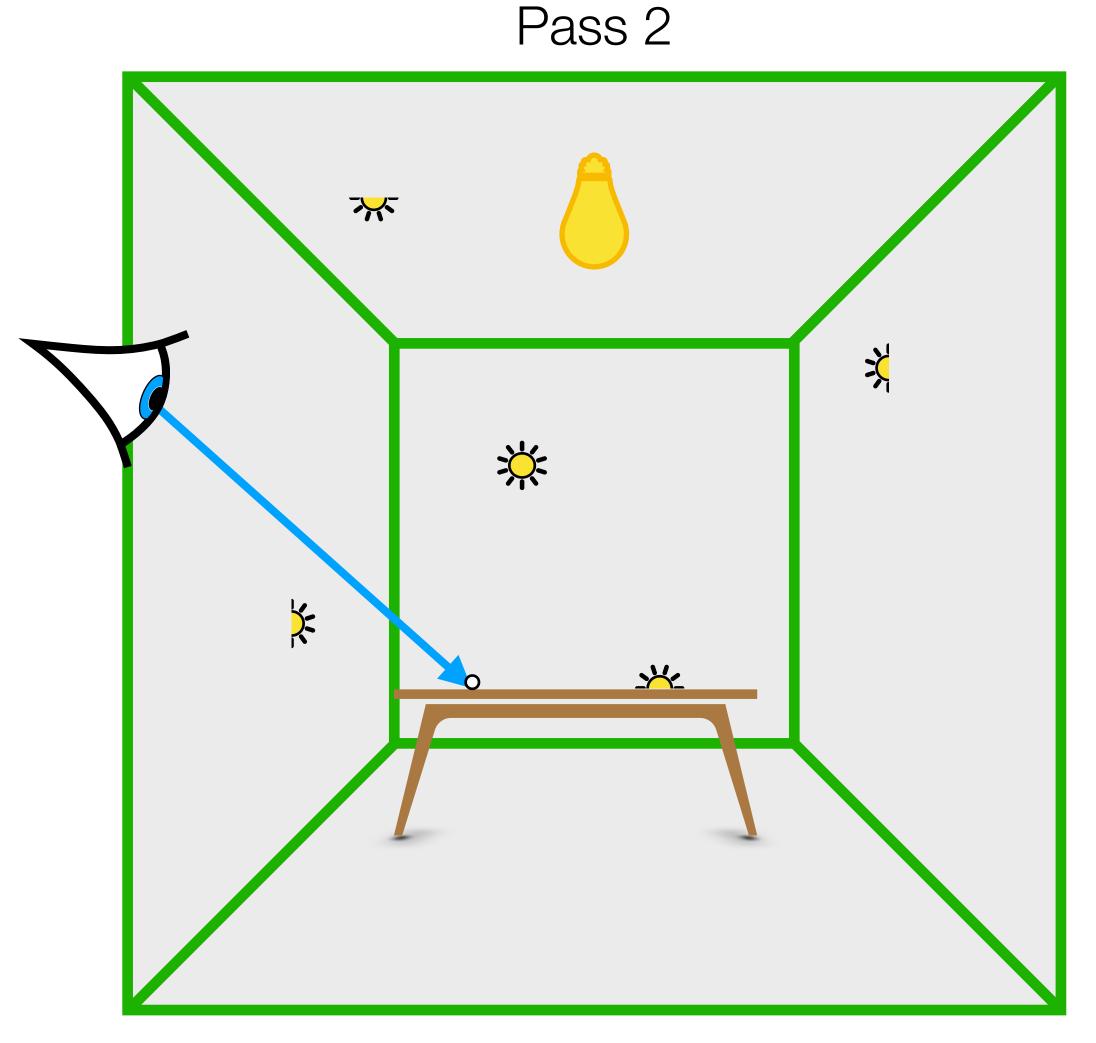
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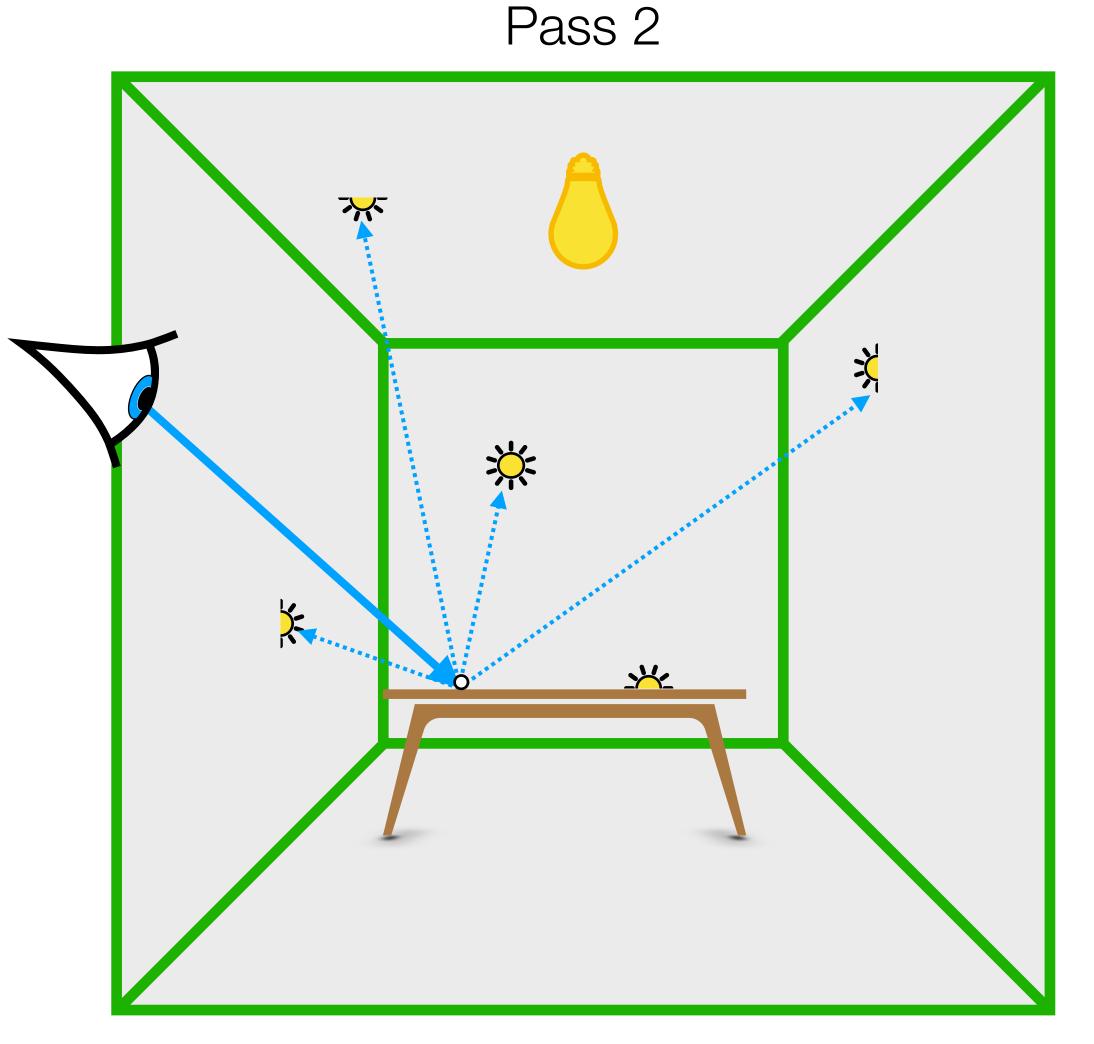
Virtual Point Light: Lighting

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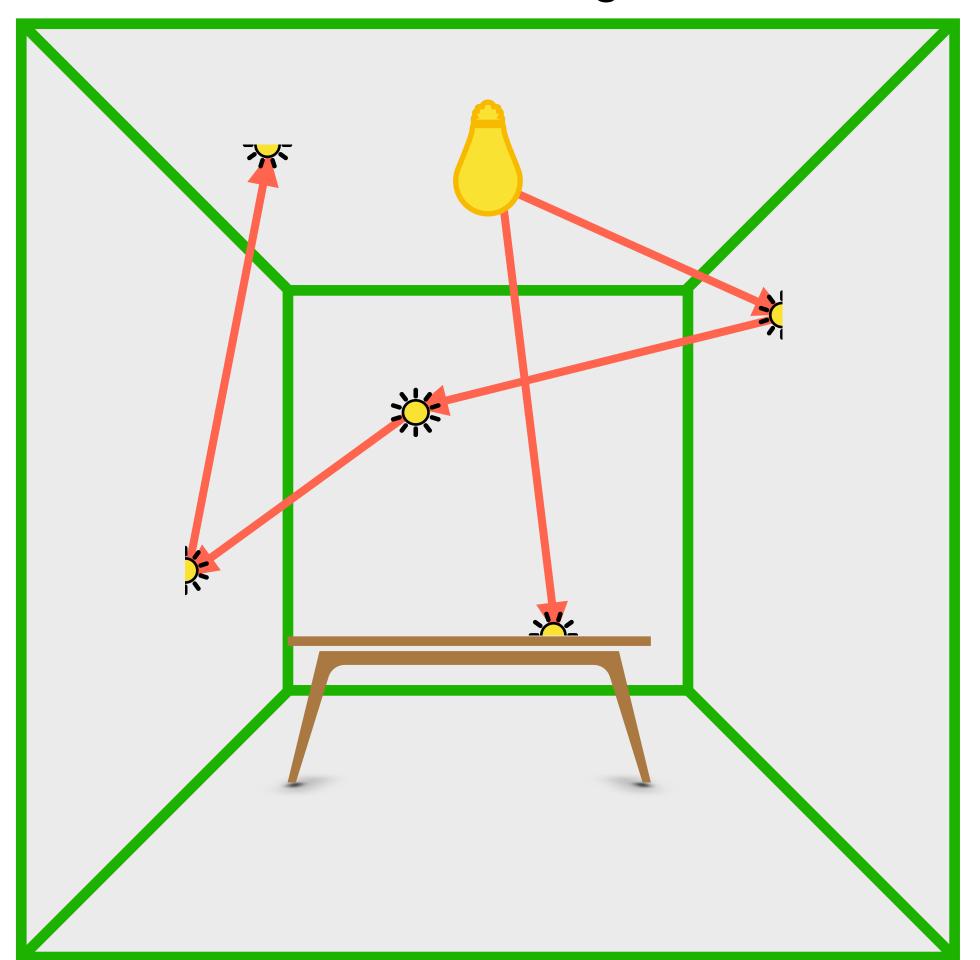




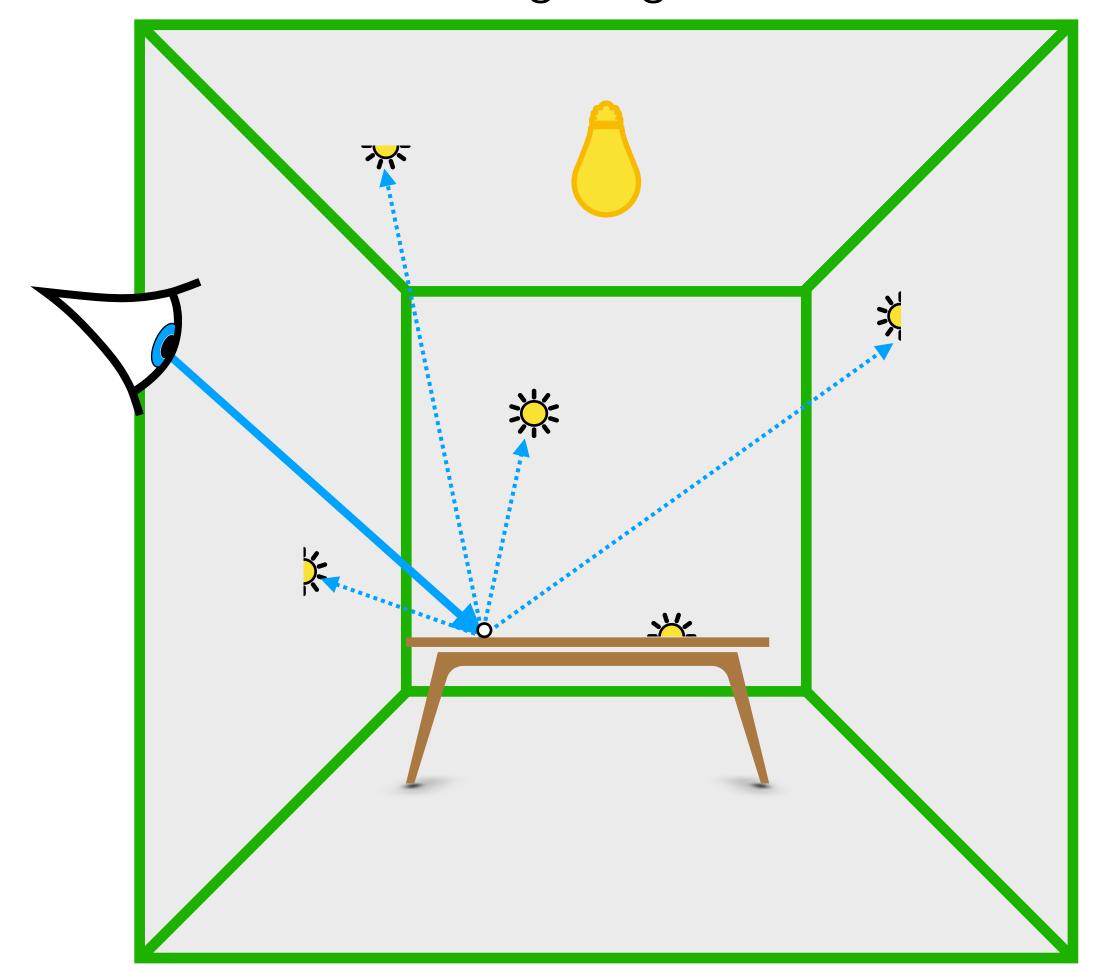


Virtual Point Light: Two-Pass

Pass 1: Generating VPLs



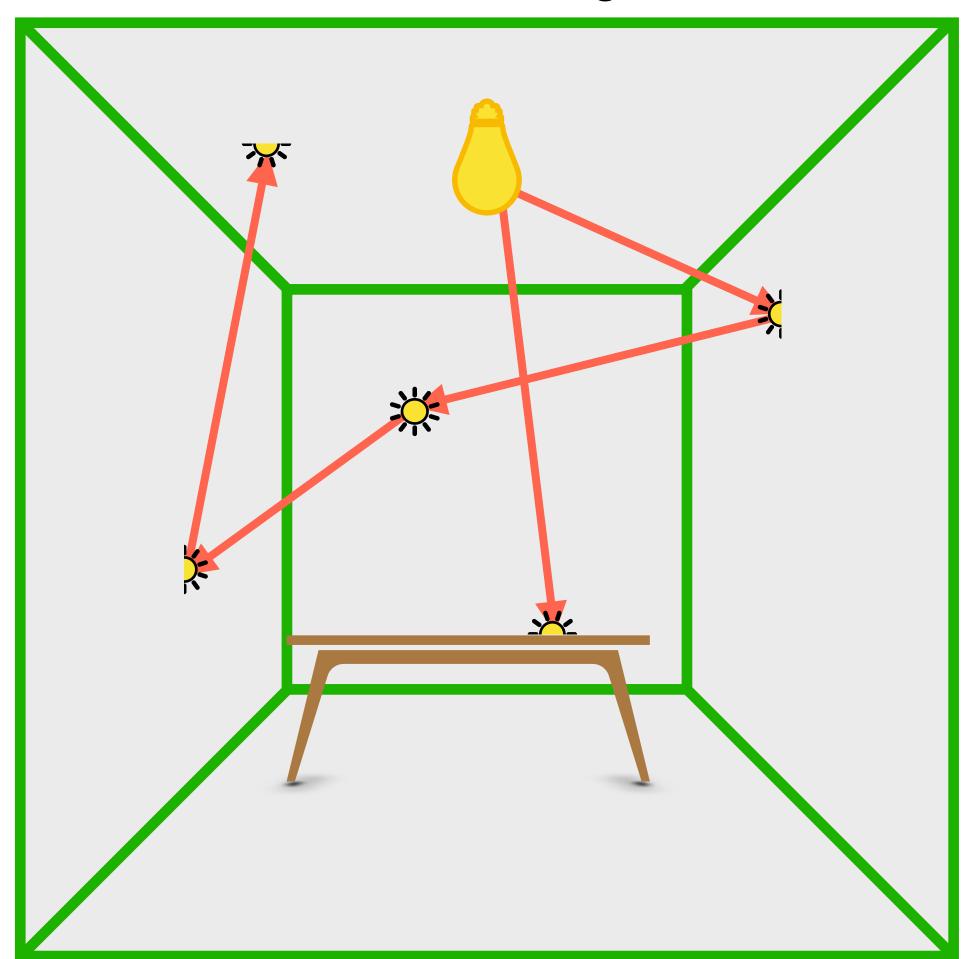
Pass 2: Lighting with VPLs



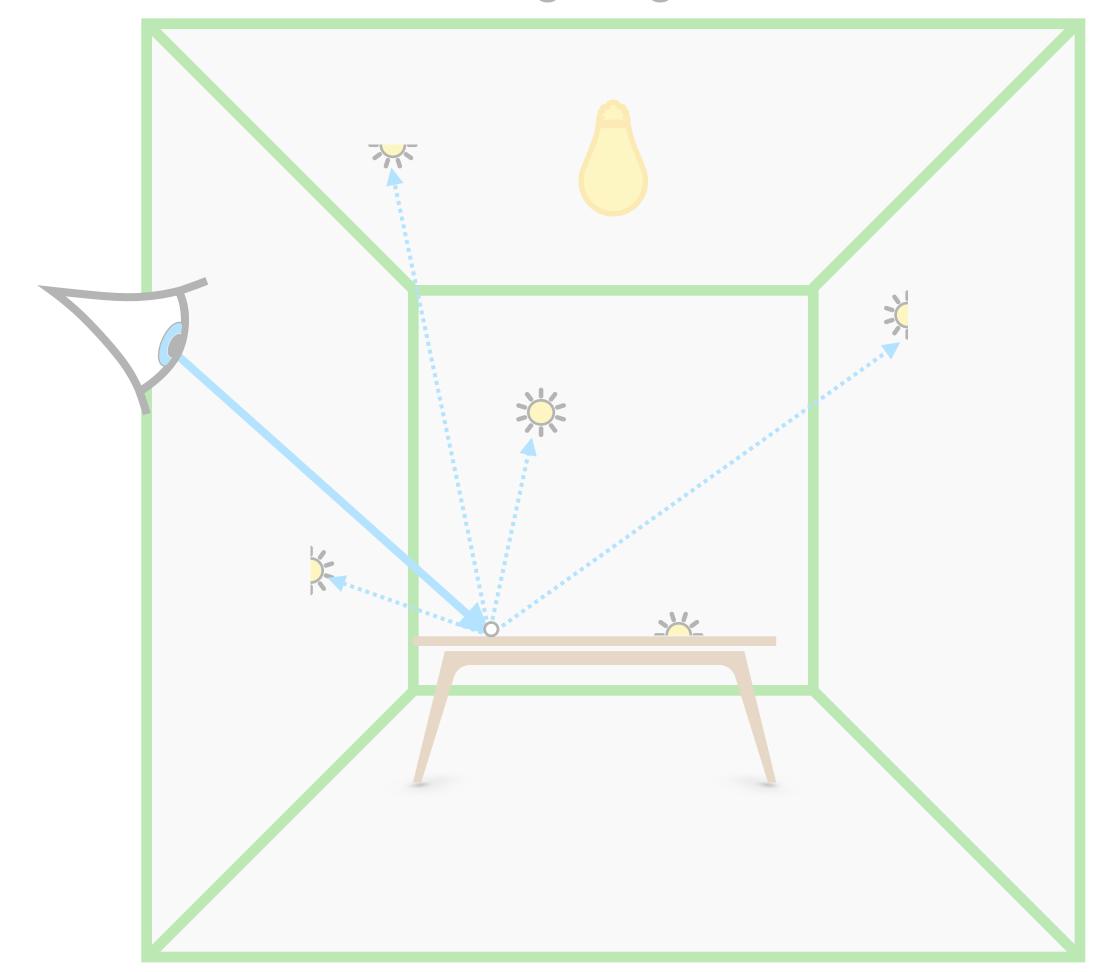


Virtual Point Light: Two-Pass

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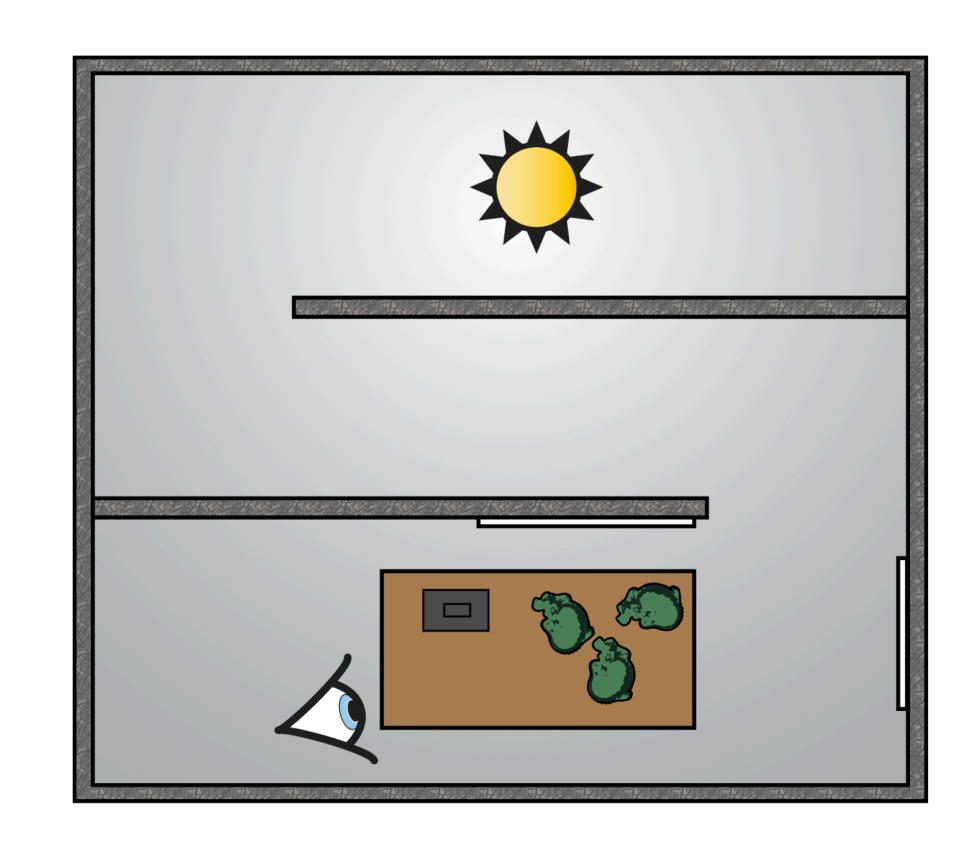
Pass 2: Lighting with VPLs





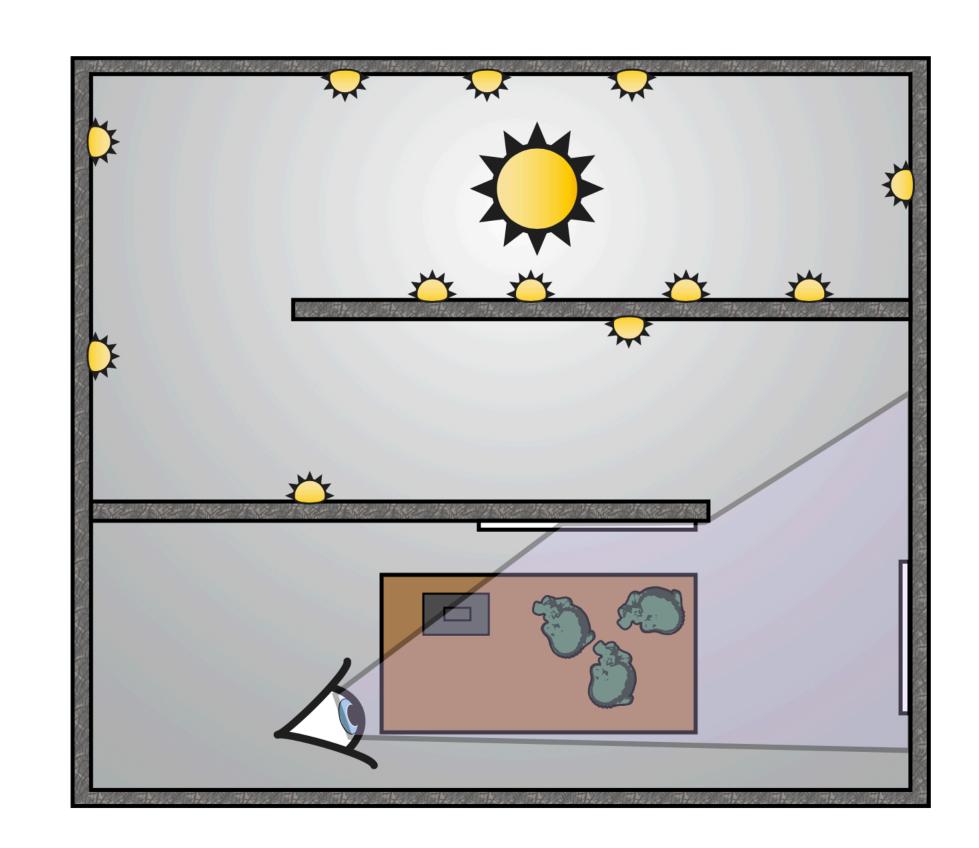
Generating Virtual Point Lights (VPLs)





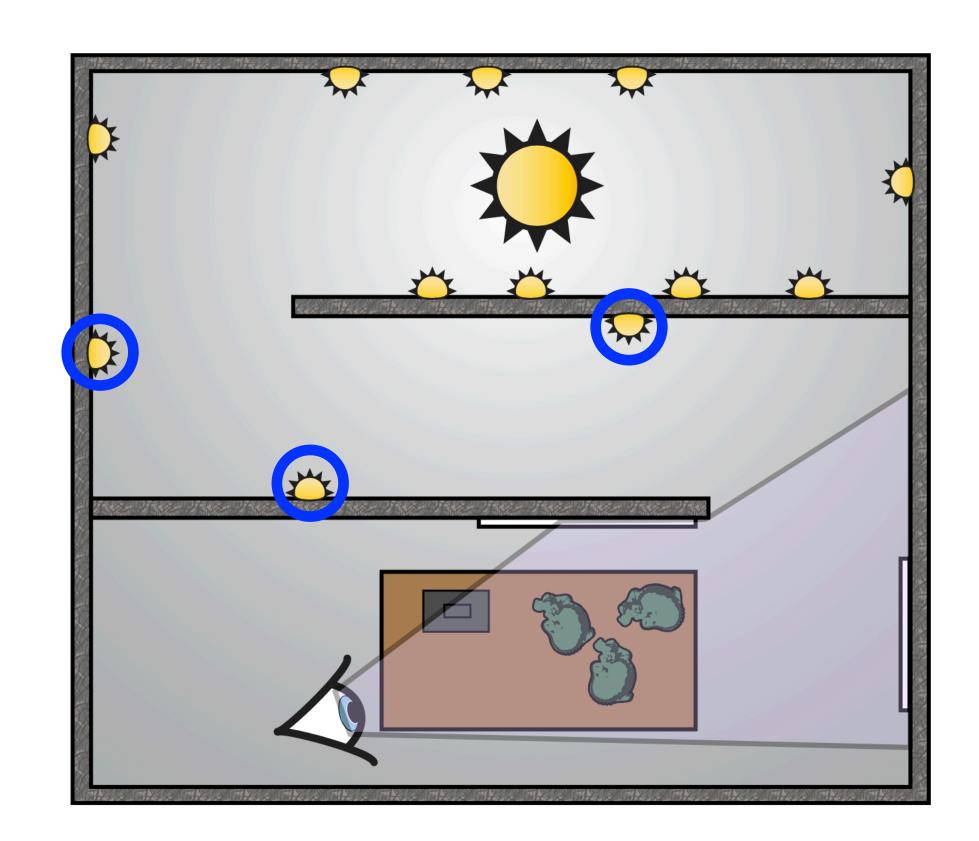






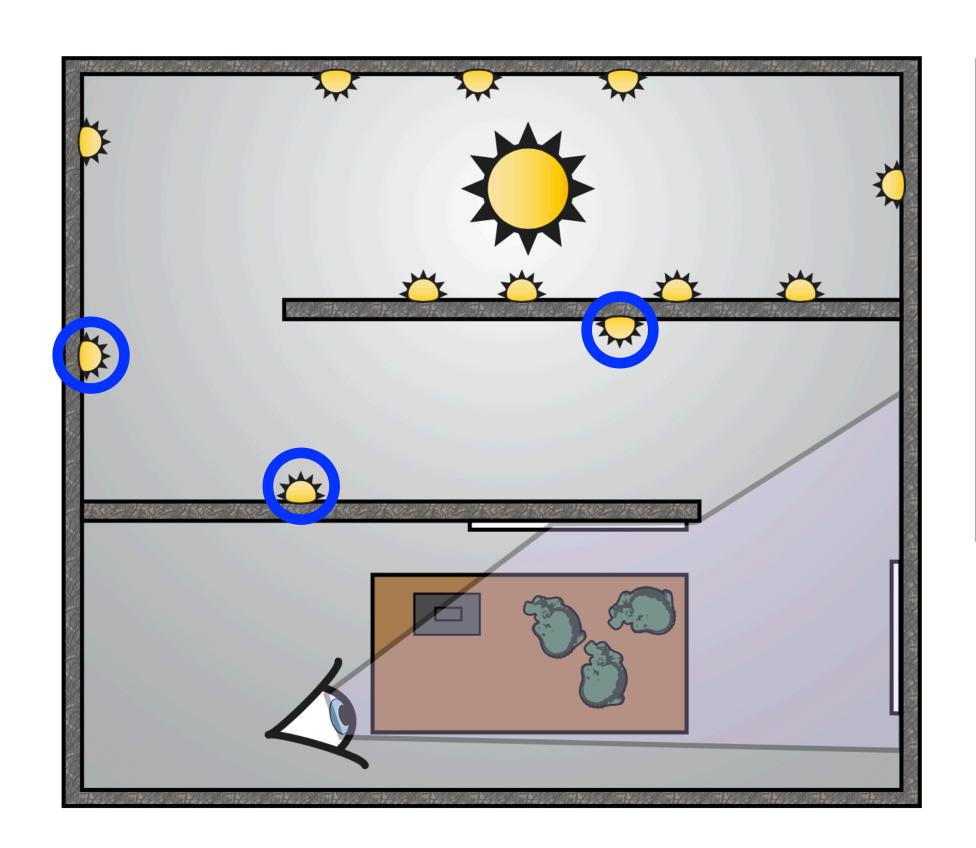




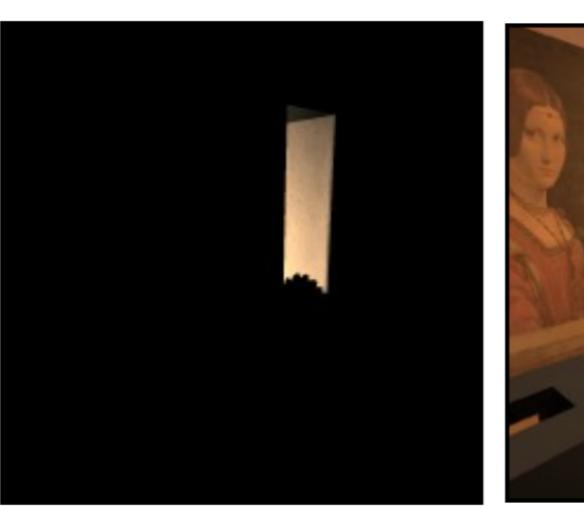








Instant Radiosity



Reference



Image courtesy Segovia et al.





Instant Radiosity





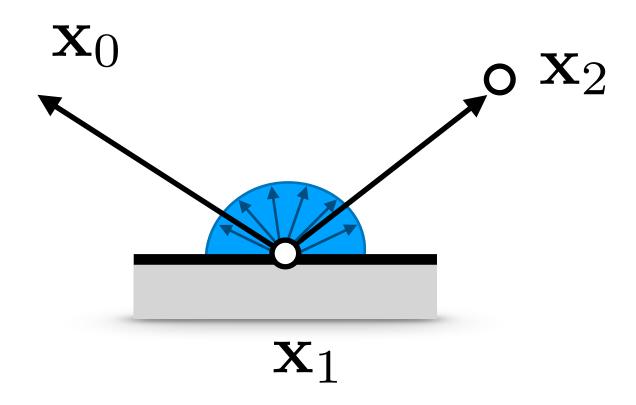


Problem 2 (in glossy scenes): - Glossy inter-reflections suffer from splotches

Instant Radiosity



Instant radiosity assumes all surfaces are diffuse





Instant Radiosity







Instant Radiosity



Clamped







Instant Radiosity



Clamped



Reference







Problem 2 (in glossy scenes):

- Glossy inter-reflections suffer from splotches
- Insufficient number of VPLs in some regions

Instant Radiosity



Clamped



Reference

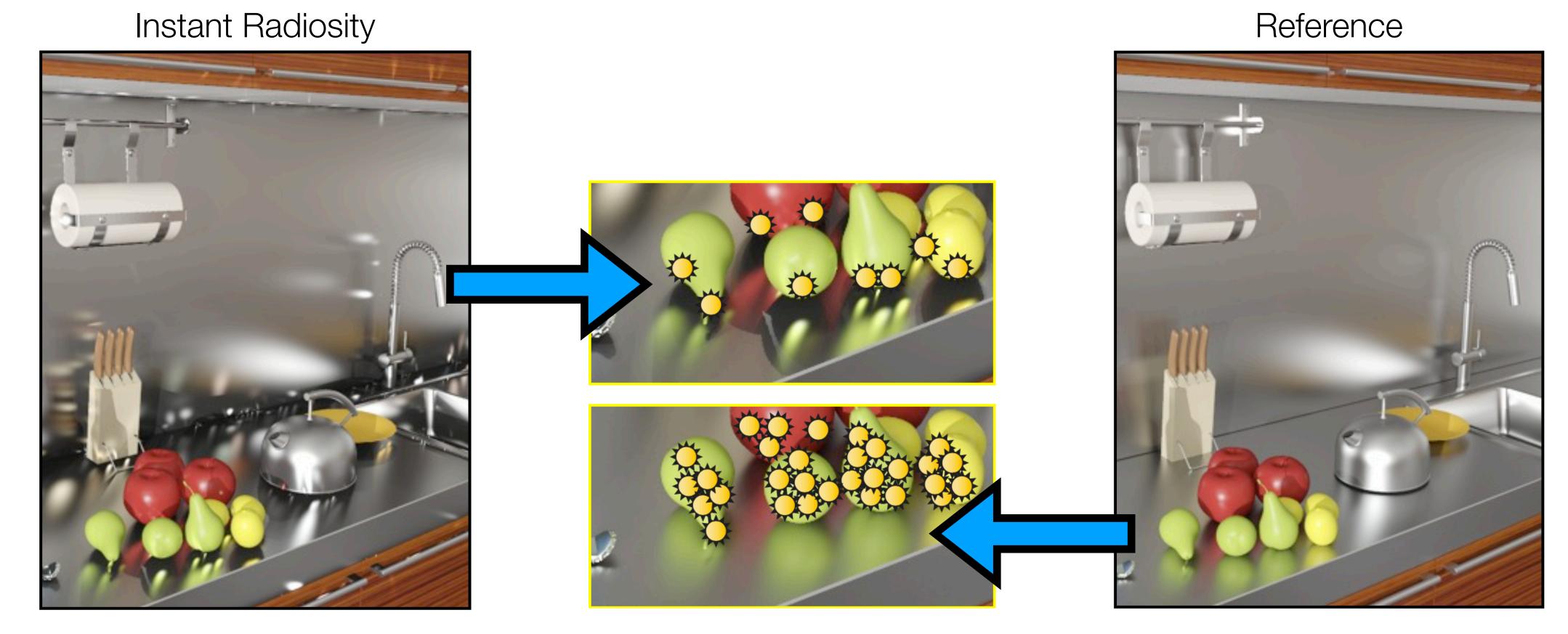






Problem 2 (in glossy scenes):

- Glossy inter-reflections suffer from splotches
- Insufficient number of VPLs in some regions







Goal:

- place VPLs only where needed

Approaches:

Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

Bidirectional Instant Radiosity [Segovia et al. 2006]

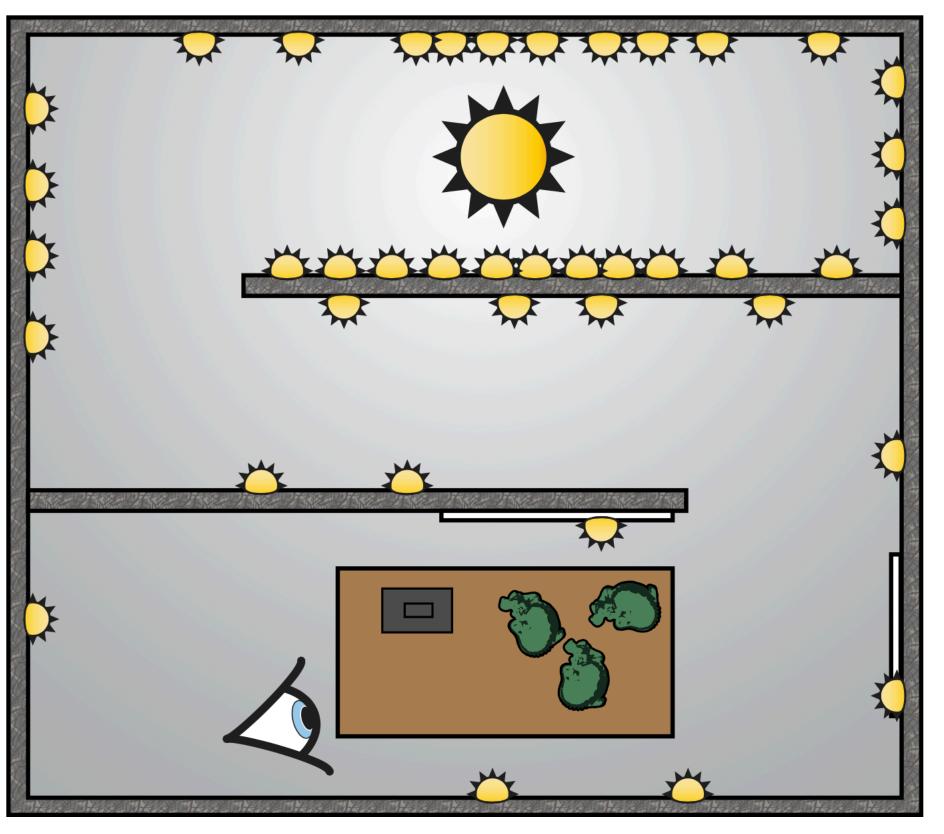
Metropolis sampling for VPL distributions [Segovia et al. 2007]





Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

- probabilistically reject VPLs with low expected contribution

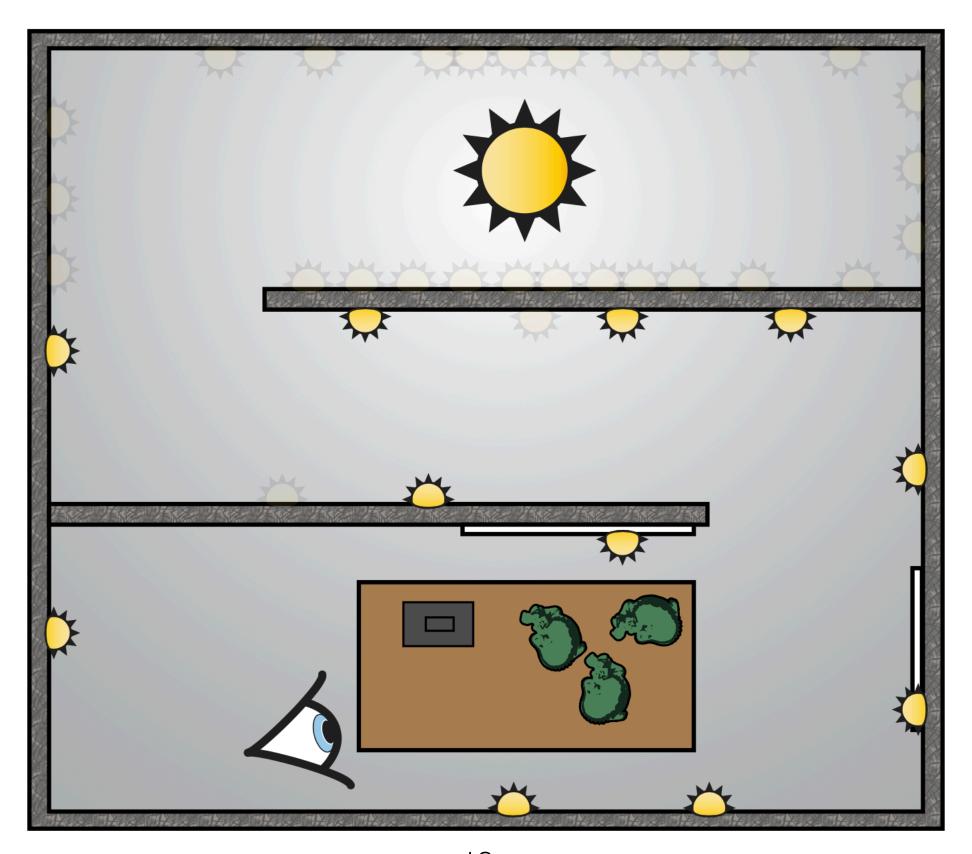






Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

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Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

Approach:





Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

- 1) estimate average contribution of a VPL Φ_v
 - few pilot VPLs illuminate few surface points seen by the camera



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 - for each VPL
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 - If accepted, divide its energy by p_i





Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

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 - for each VPL
 - estimate its contribution Φ_i to points seen by the camera

- accept with probability
$$\ p_i = \min\left(\frac{\Phi_i}{\Phi_v} + \epsilon, 1\right)$$

- If accepted, divide its energy by p_i

Russian roulette





Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

Advantages:

Cheap and simple to implement!

VPLs have roughly equal contribution

Works well most of the time



Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

Advantages:

Cheap and simple to implement!

VPLs have roughly equal contribution

Works well most of the time

Disadvantages:

Increase the cost of VPL distribution

"one-pixel image" assumption

Does not help with local inter-reflections



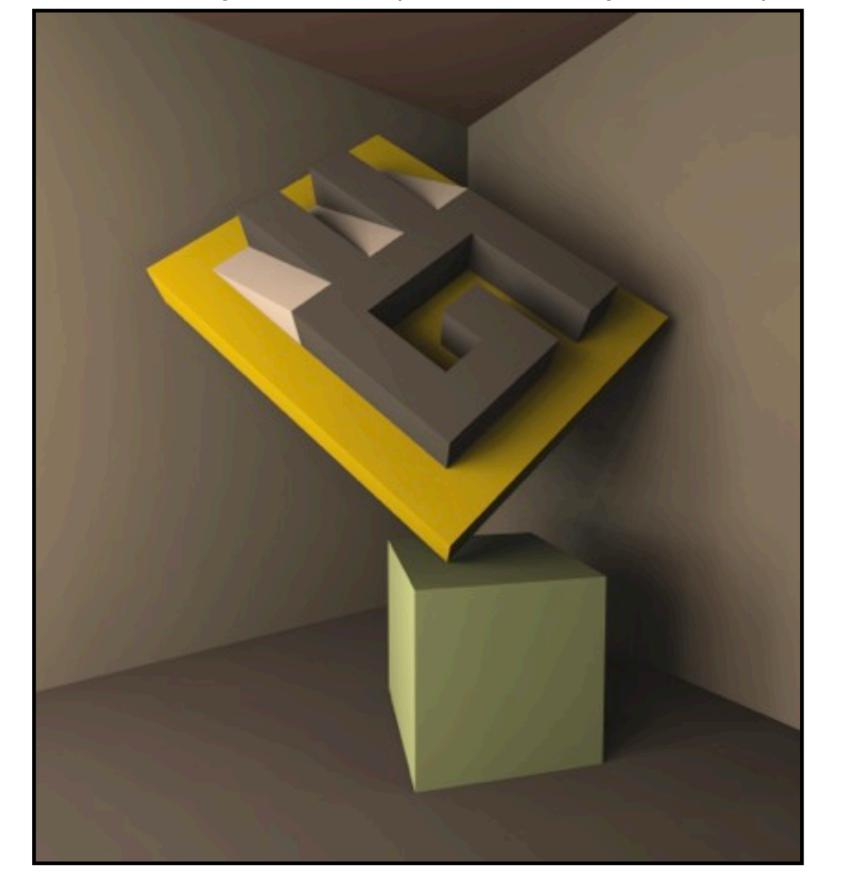


Rejection of unimportant VPLs [Gerogiev and Slusallek 2010]

Without rejection



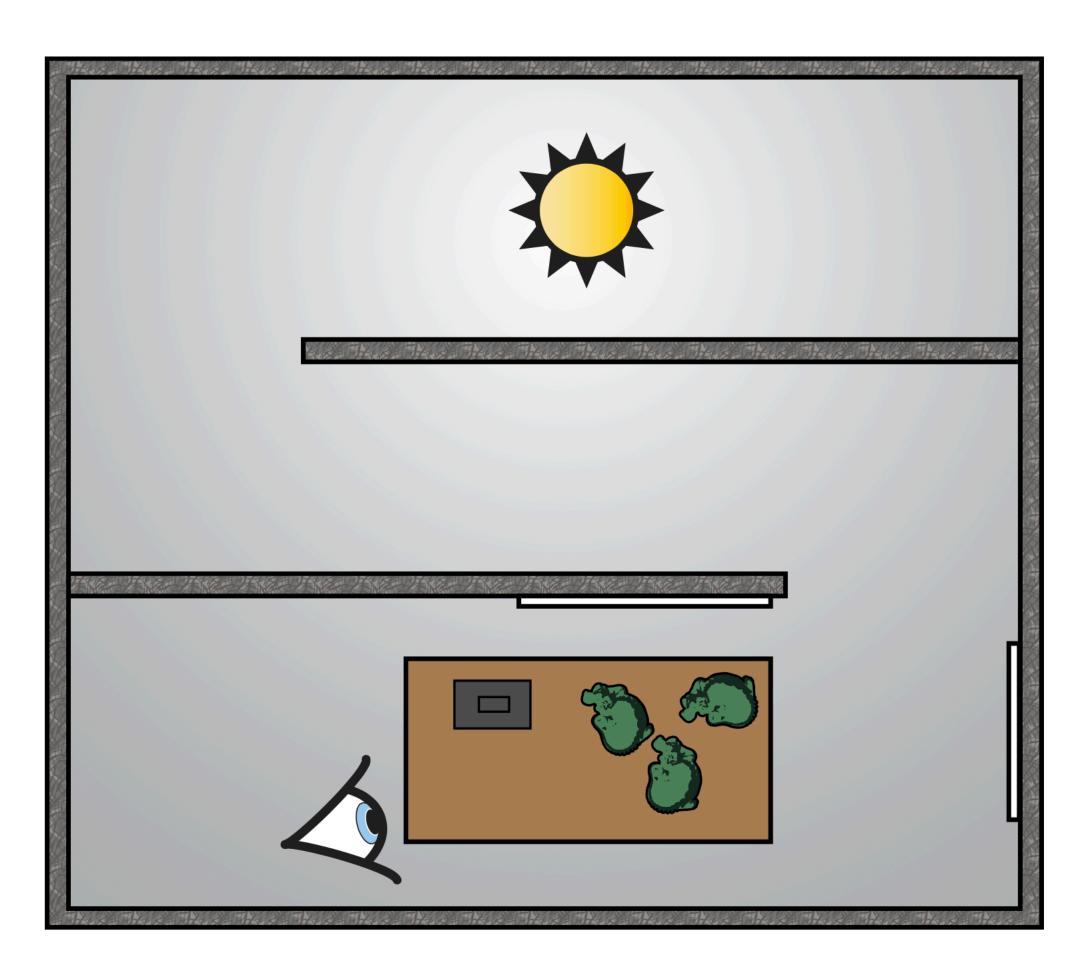
With rejection (7% acceptance)







Bidirectional Instant Radiosity [Segovia et al. 2006]

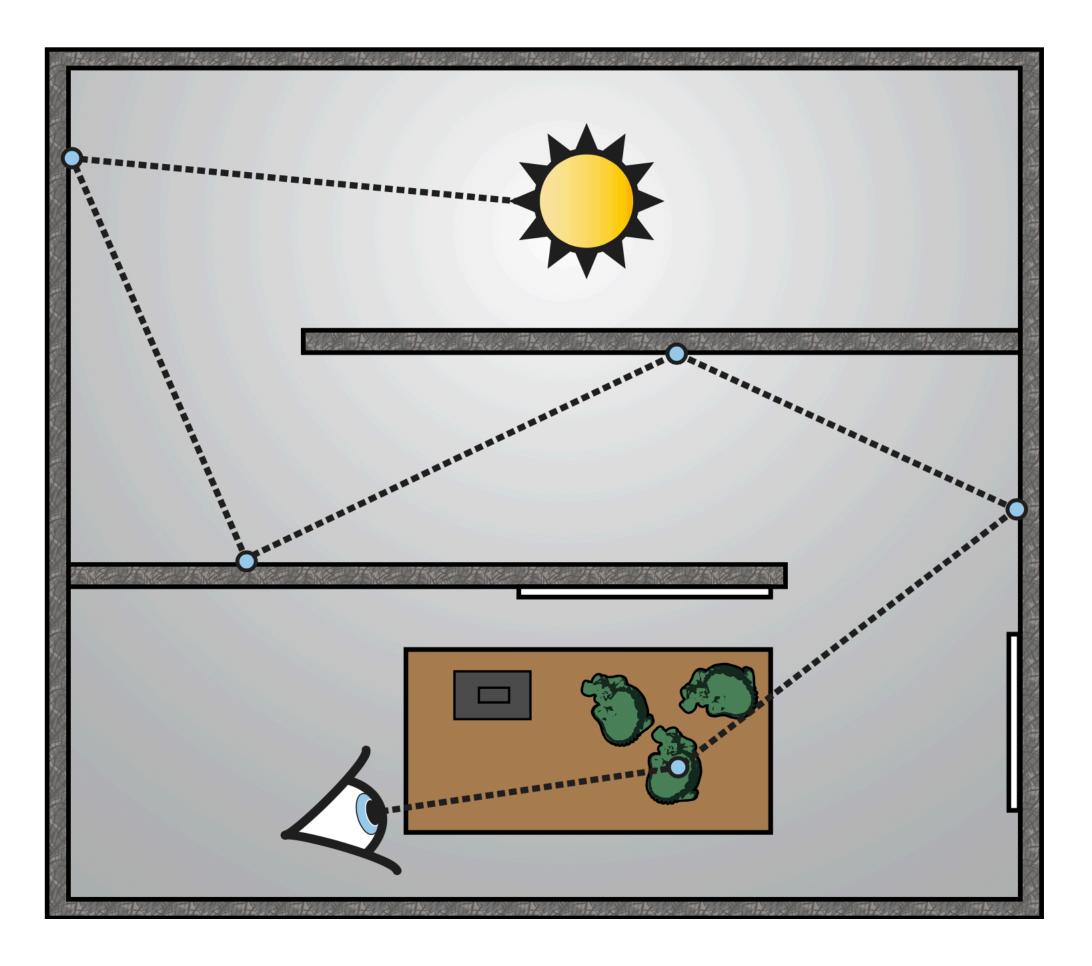


Create a VPL at the second bounce from the camera





Bidirectional Instant Radiosity [Segovia et al. 2006]

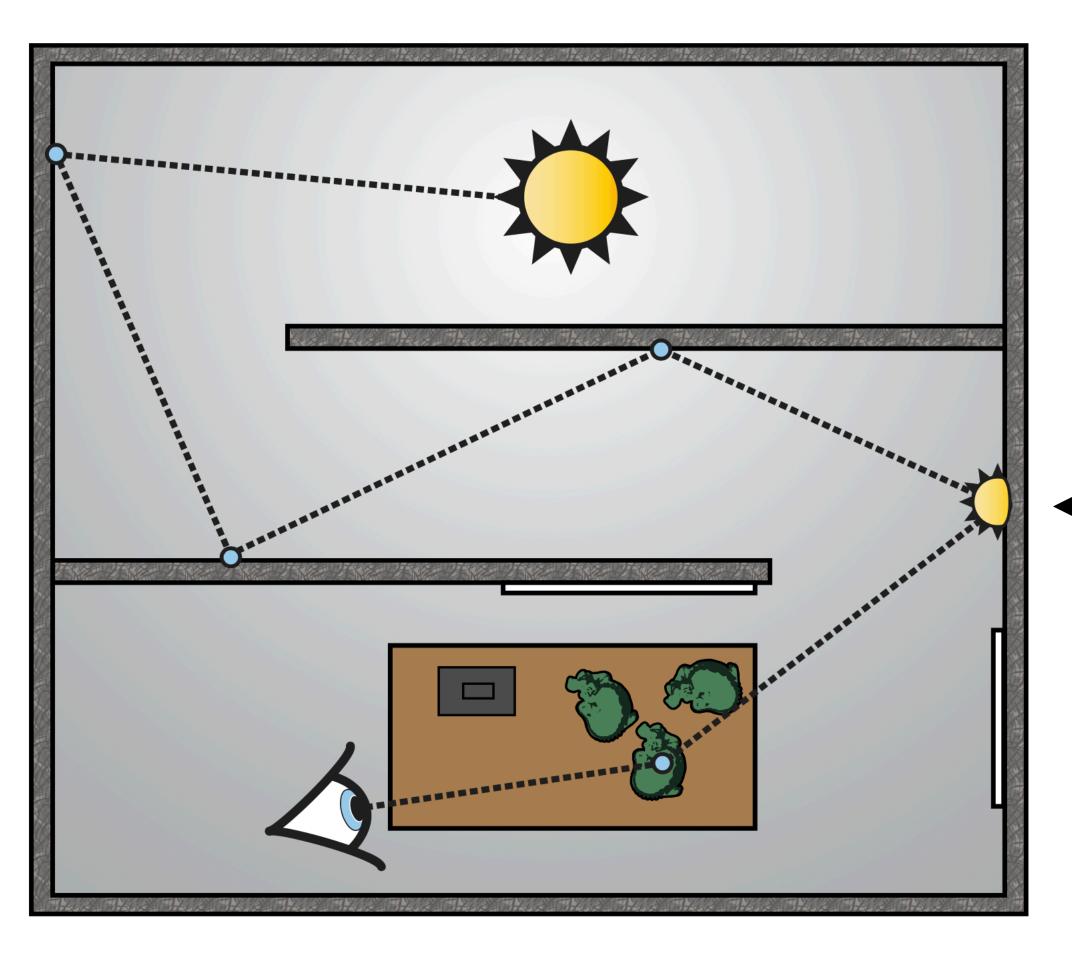


Create a VPL at the second bounce from the camera





Bidirectional Instant Radiosity [Segovia et al. 2006]



Create a VPL at the second bounce from the camera

2nd vertex from camera



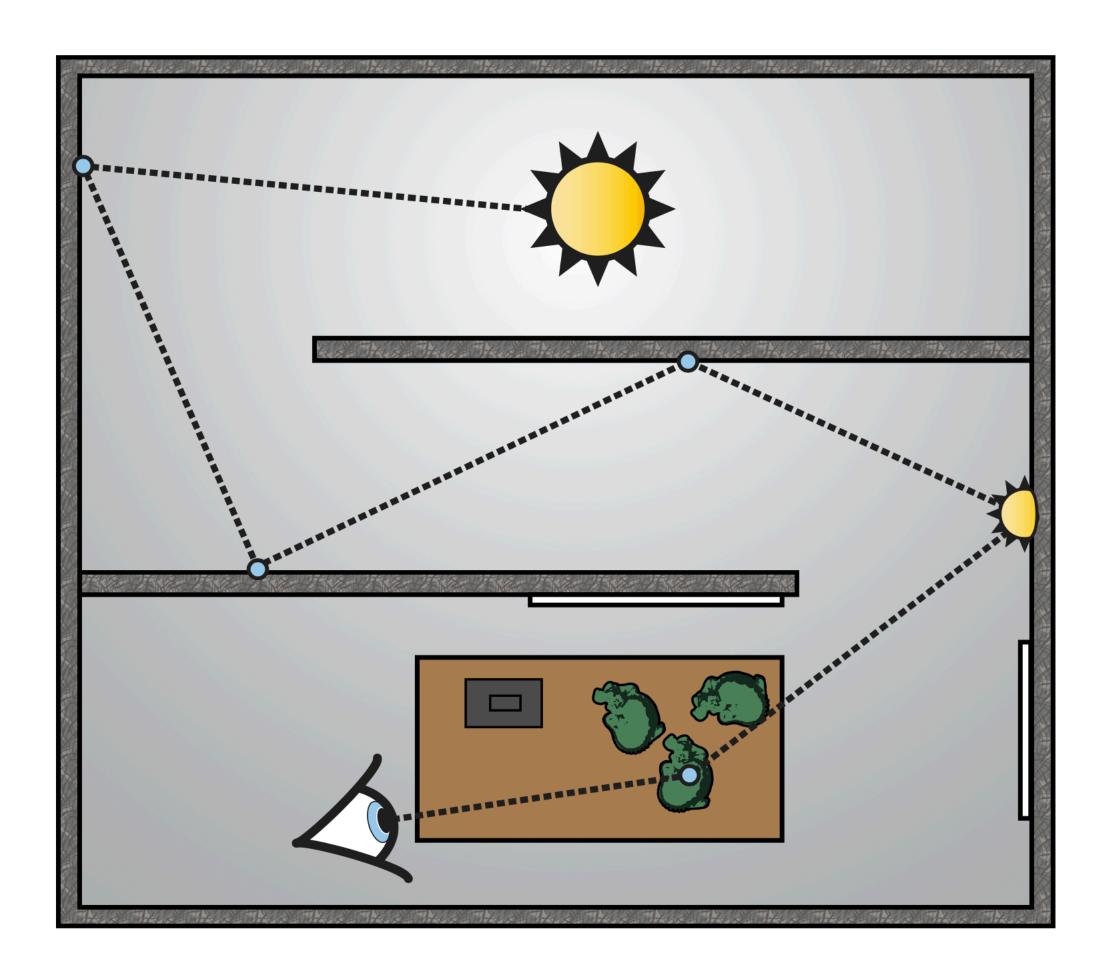


Metropolis Instant Radiosity [Segovia et al. 2007]





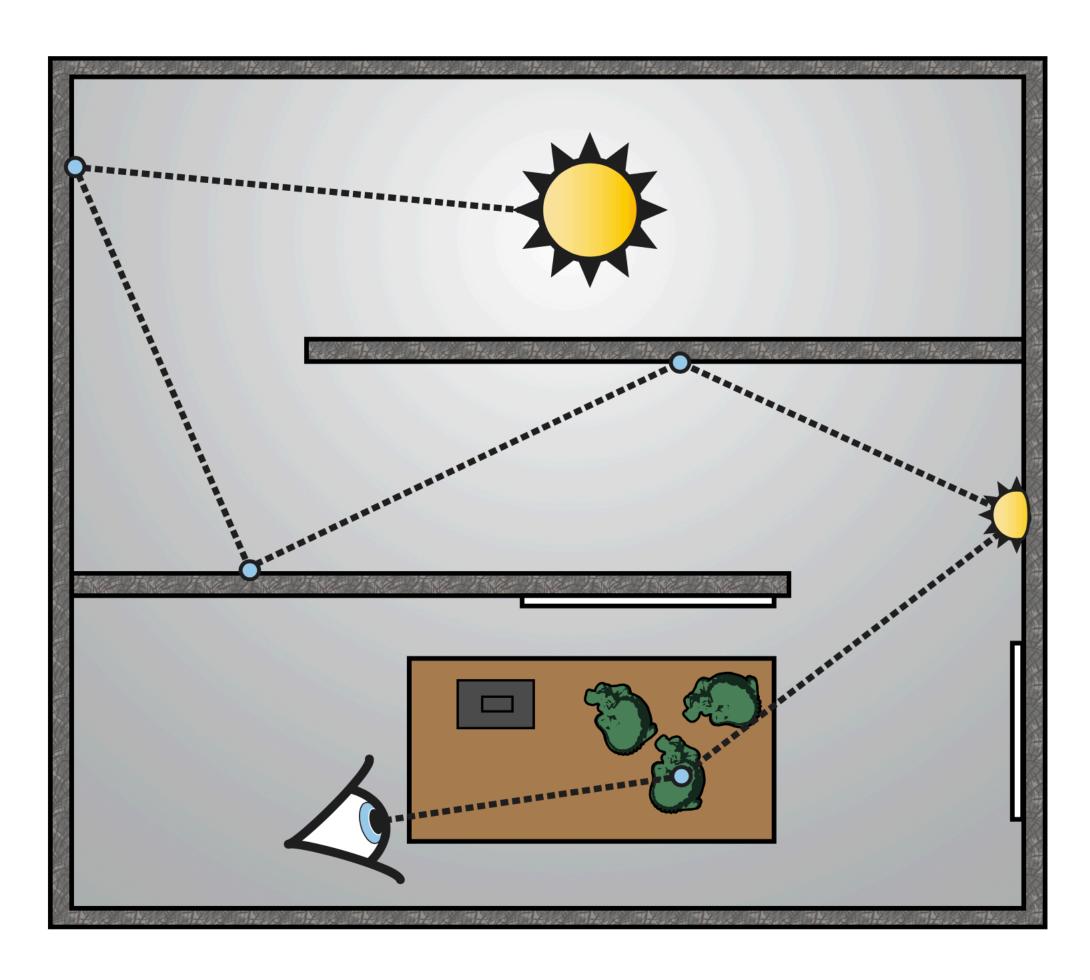
Metropolis Instant Radiosity [Segovia et al. 2007]







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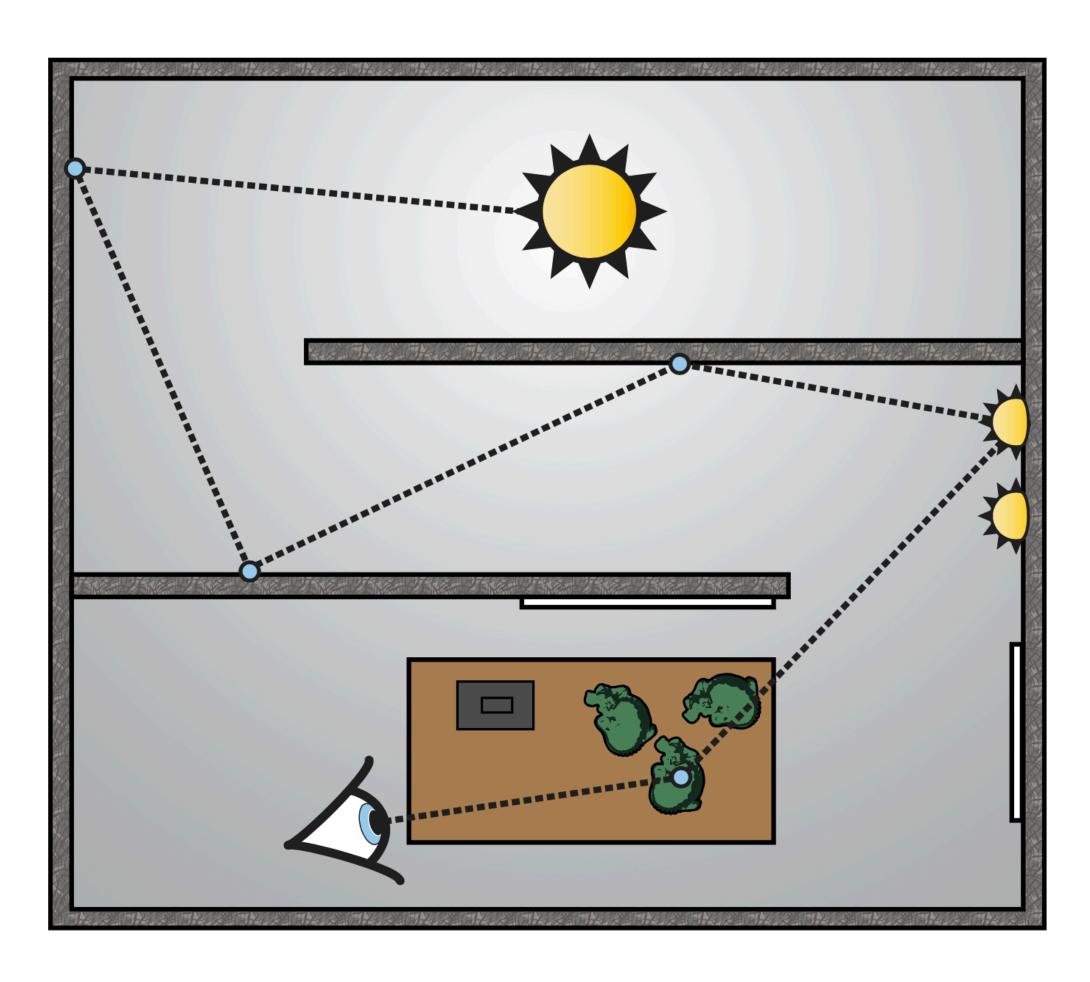


Generate VPLs by mutating paths





Metropolis Instant Radiosity [Segovia et al. 2007]

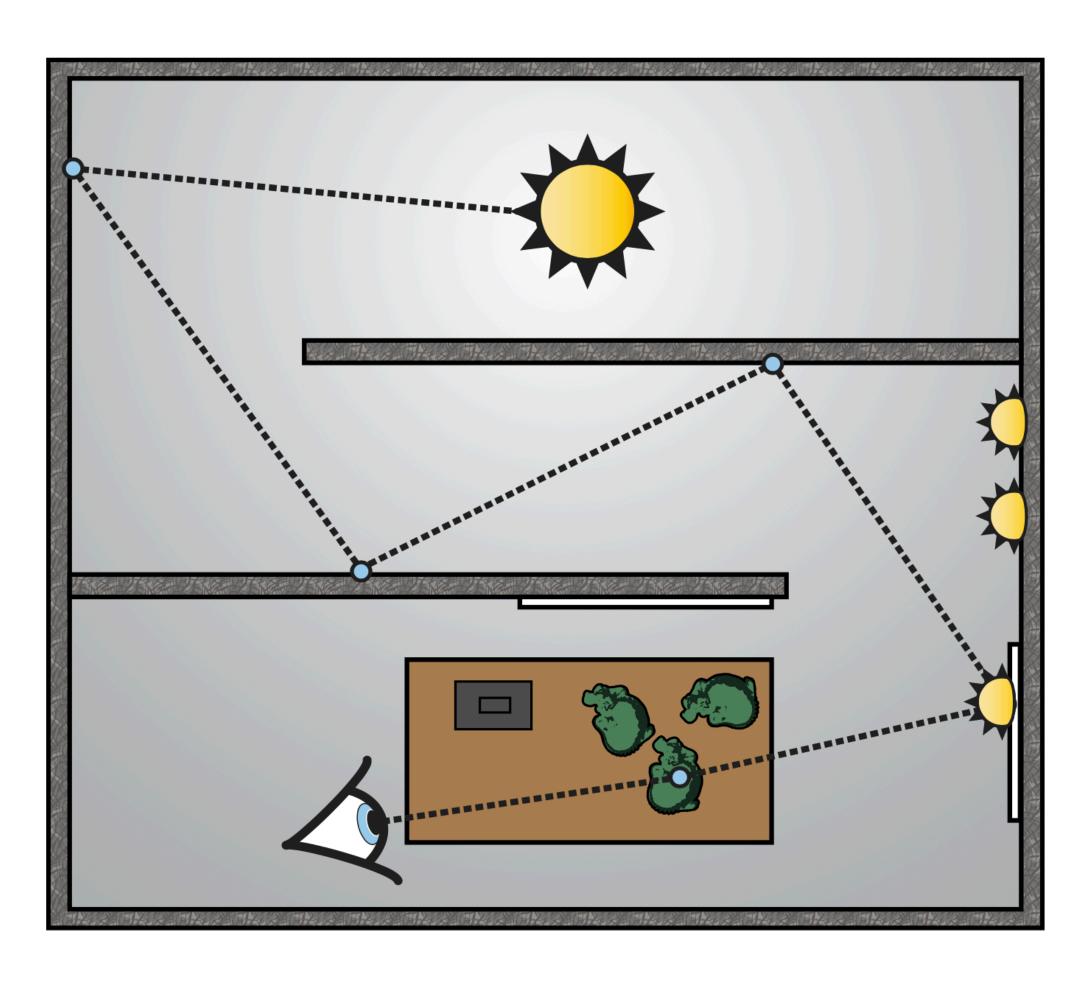


Generate VPLs by mutating paths





Metropolis Instant Radiosity [Segovia et al. 2007]



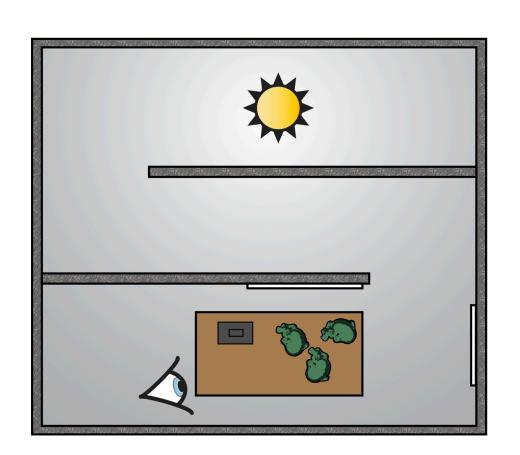
Generate VPLs by mutating paths





Comparisons

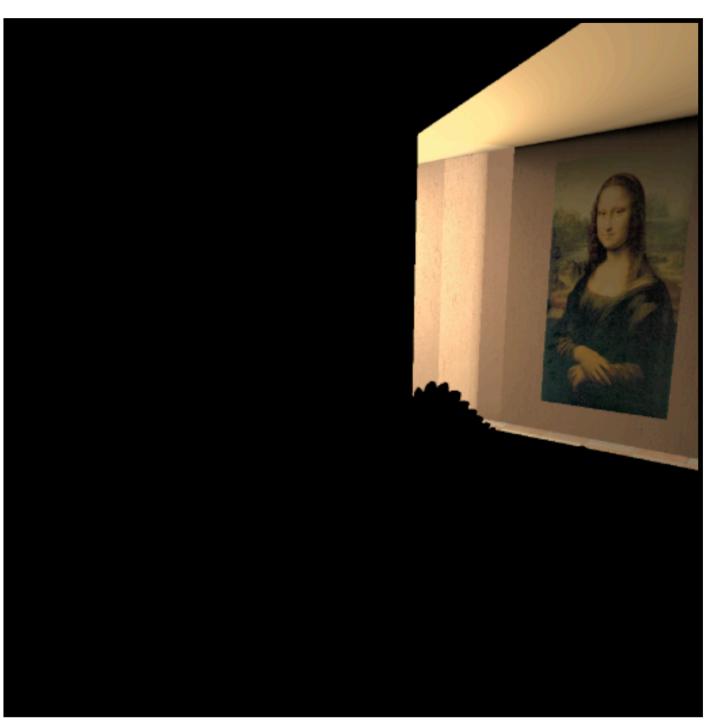
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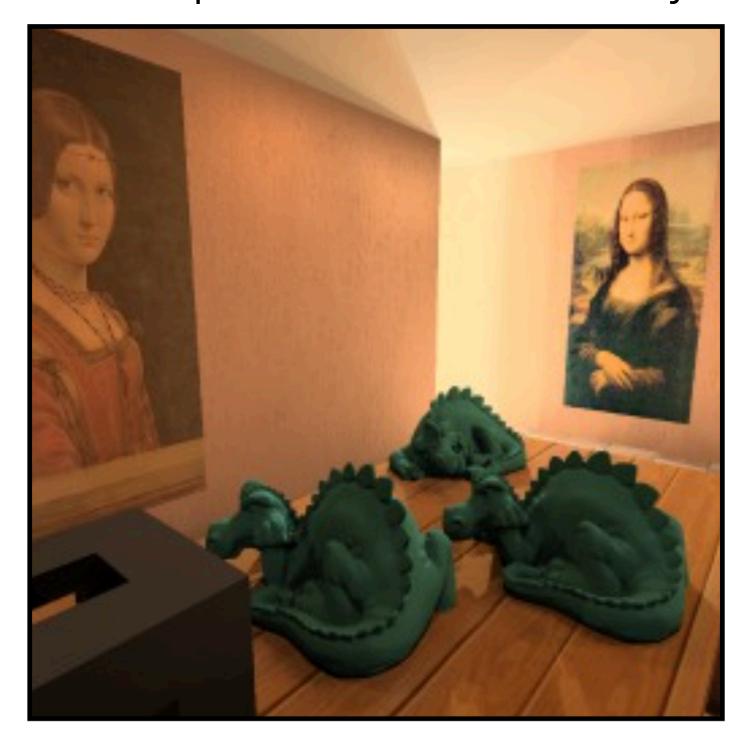
Instant Radiosity



Bidirectional Instant Radiosity



Metropolis Instant Radiosity







Comparisons

Metropolis Instant Radiosity [Segovia et al. 2006]

Advantages:

Handles large and difficult scenes

VPLs have equal contribution

Disadvantages:

Complicated implementation

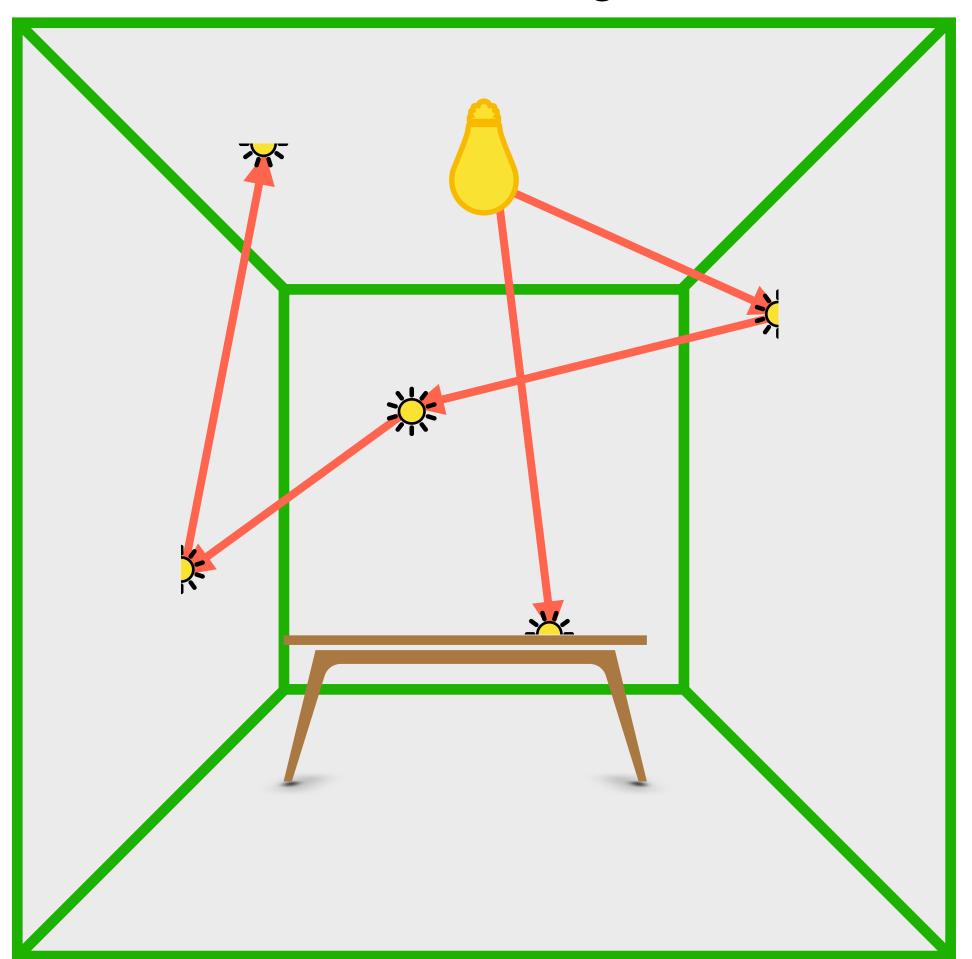
Does not help with local inter-reflections



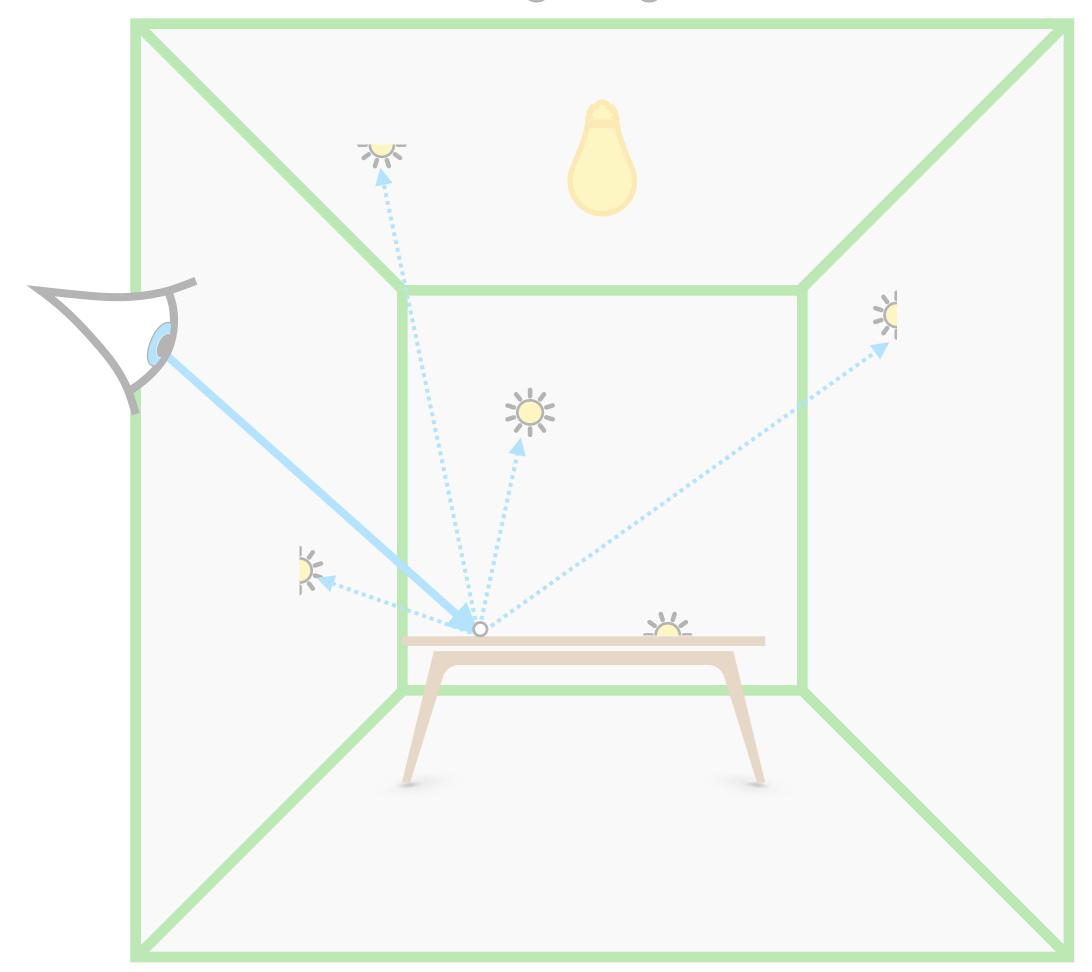


Virtual Point Light: Two-Pass

Pass 1: Generating VPLs



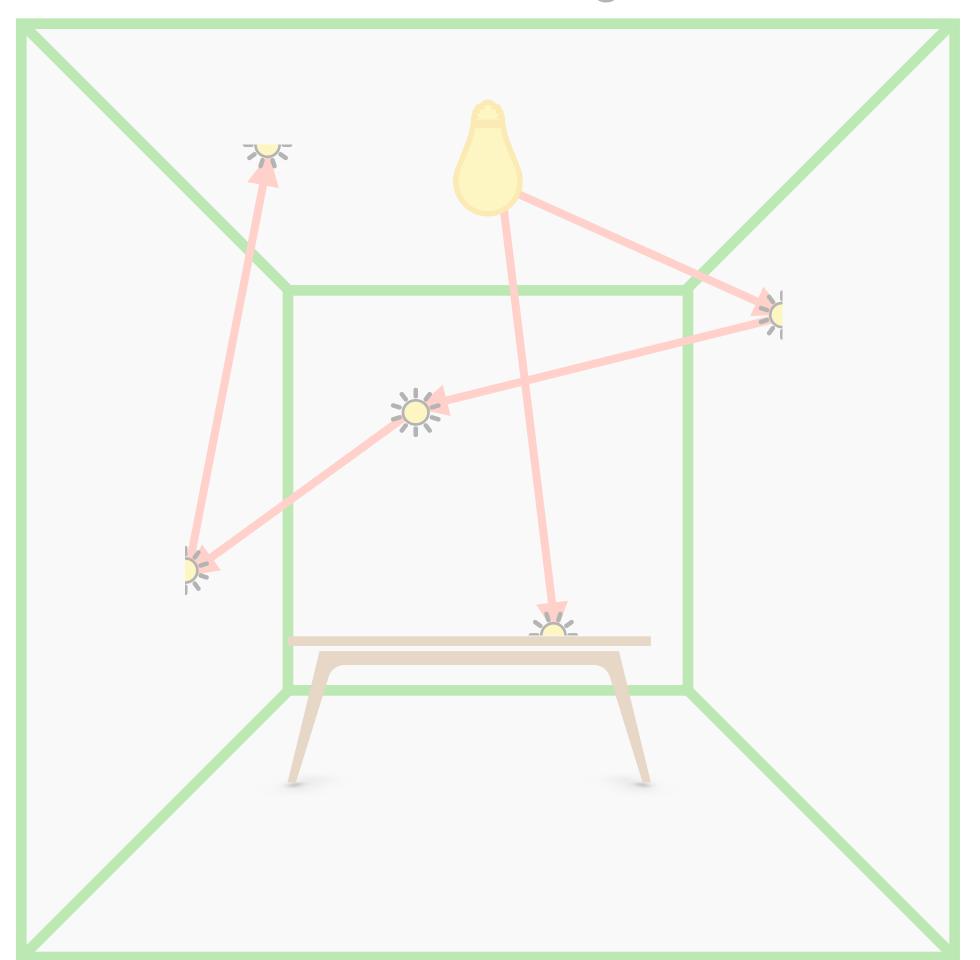
Pass 2: Lighting with VPLs



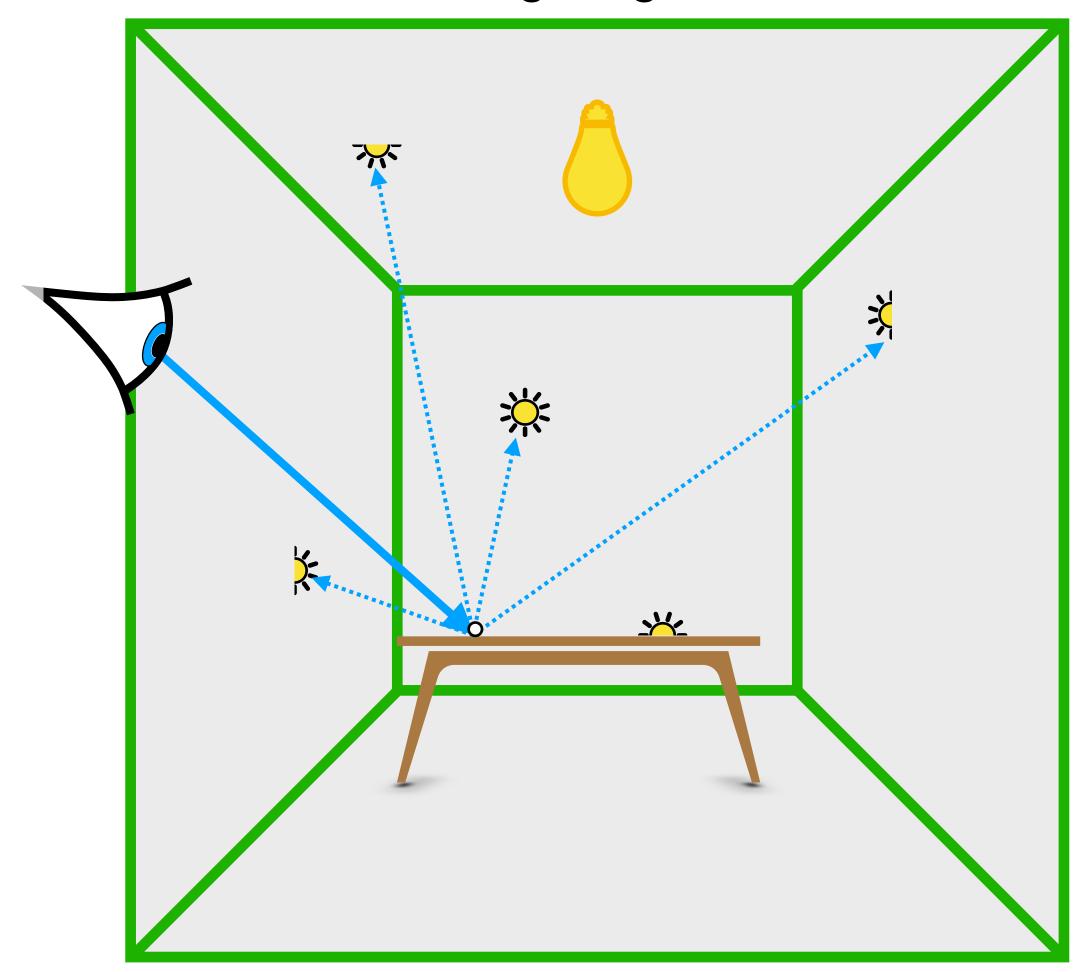


Virtual Point Light: Two-Pass

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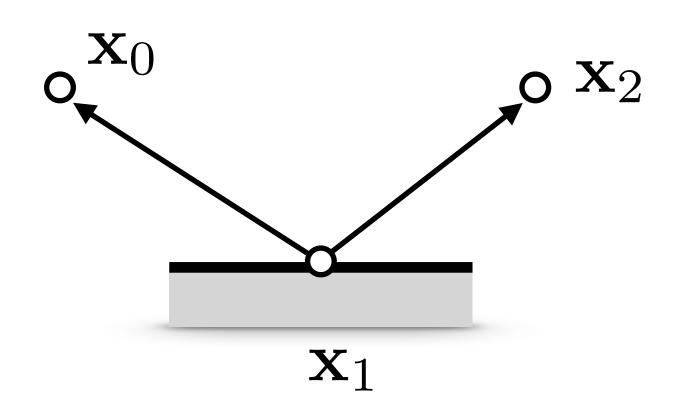
Pass 2: Lighting with VPLs





Lighting with Virtual Point Lights (VPLs)

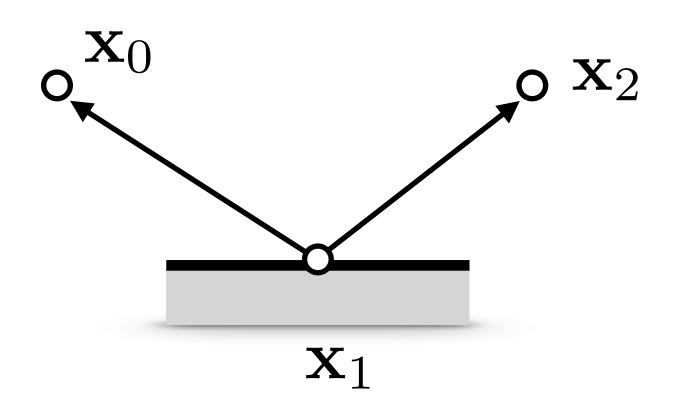








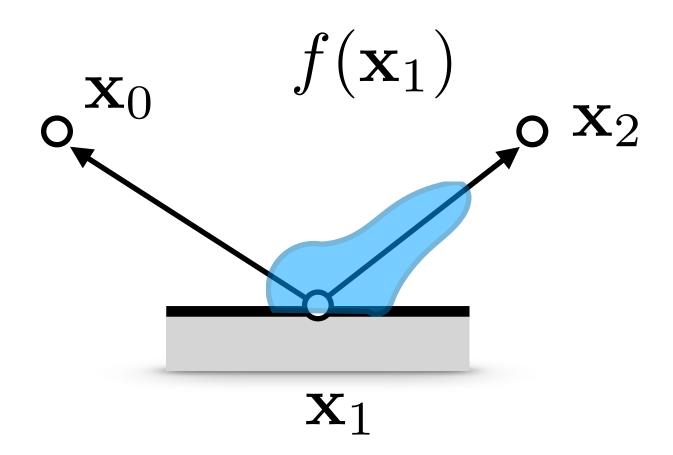
$$L(\mathbf{x}_1 \to \mathbf{x}_0) = L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \int_A f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$







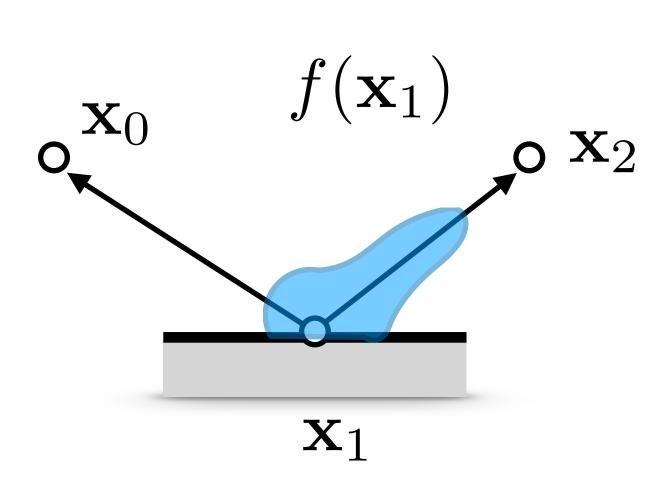
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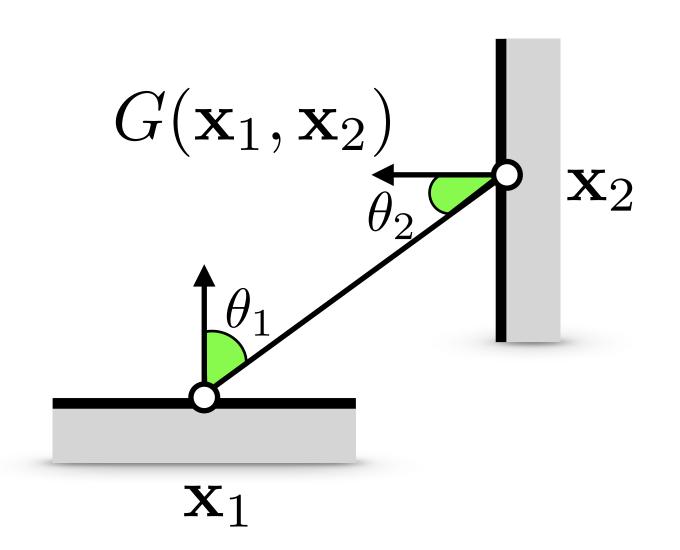






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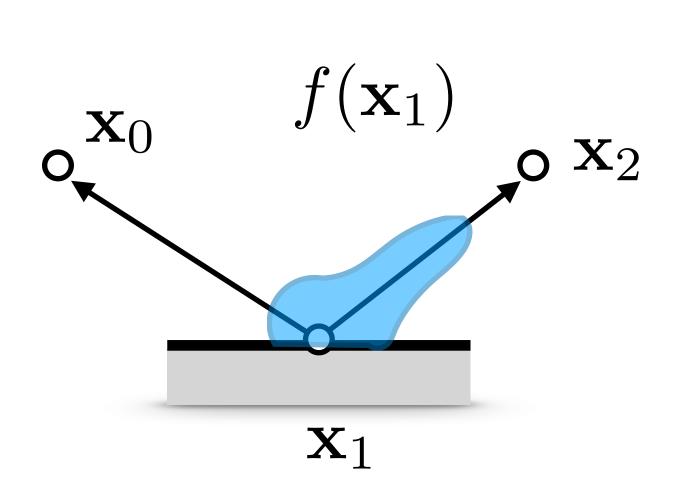


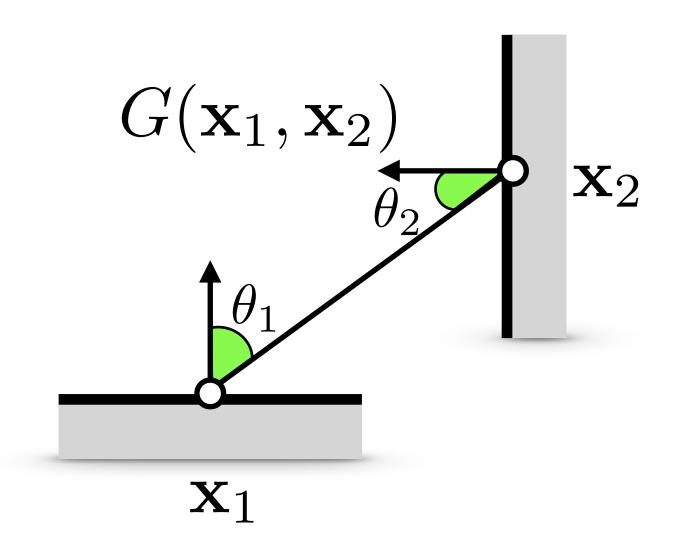


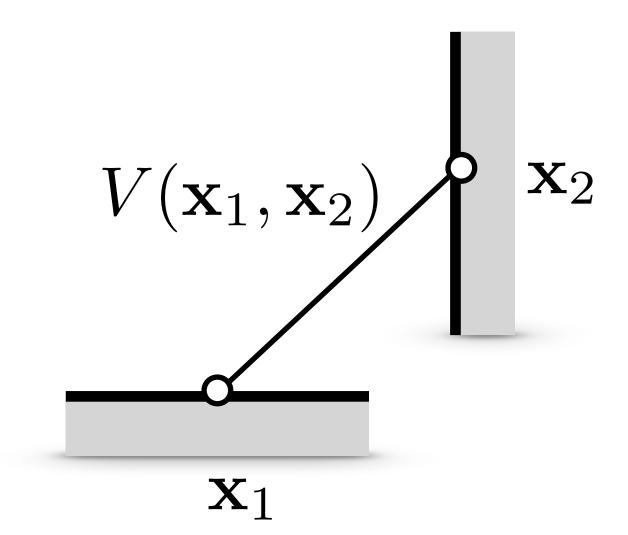




$$L(\mathbf{x}_1 \to \mathbf{x}_0) = L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \int_A f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$



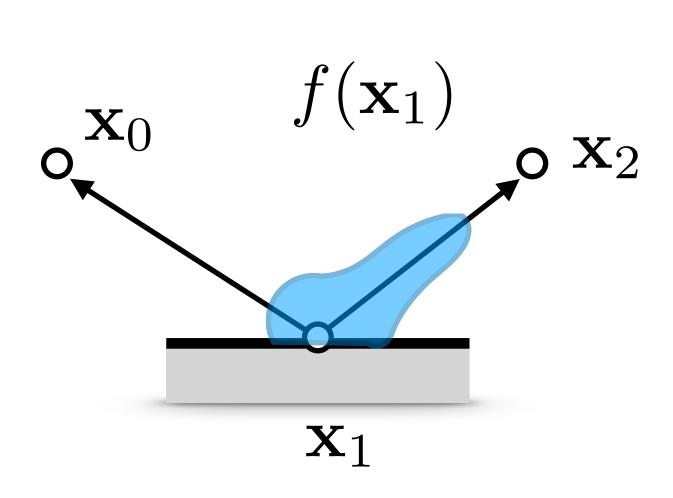


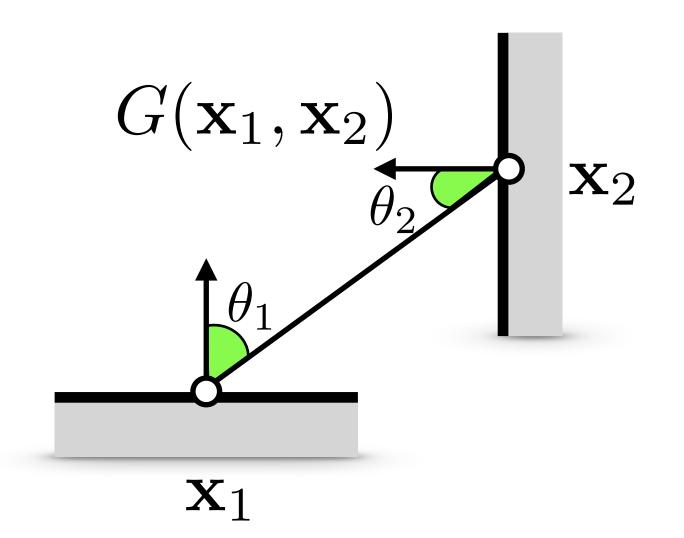


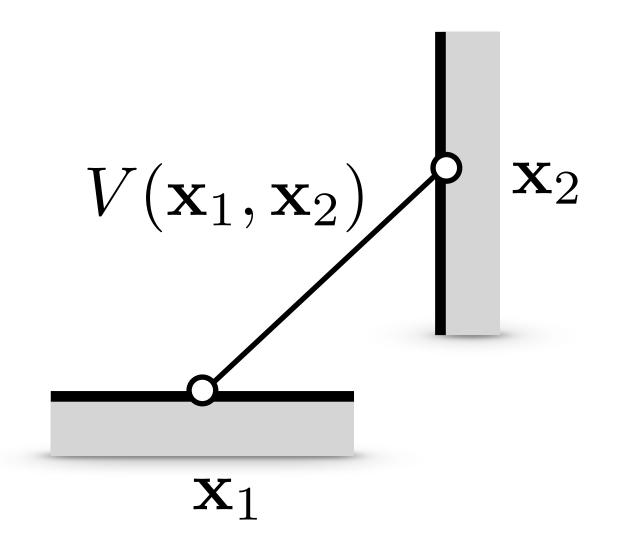




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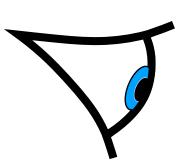


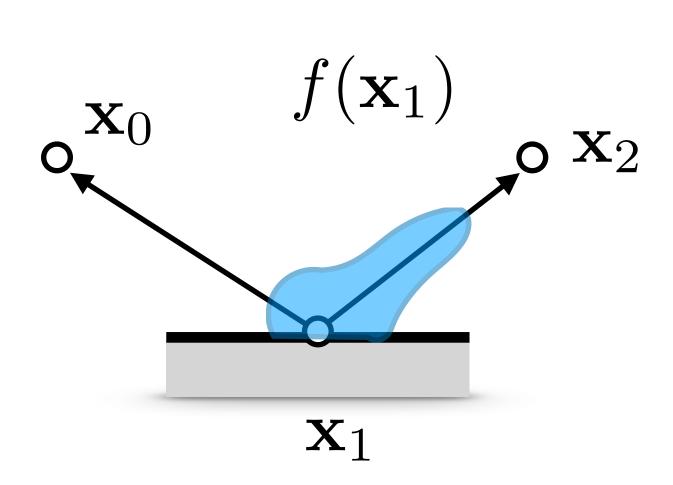


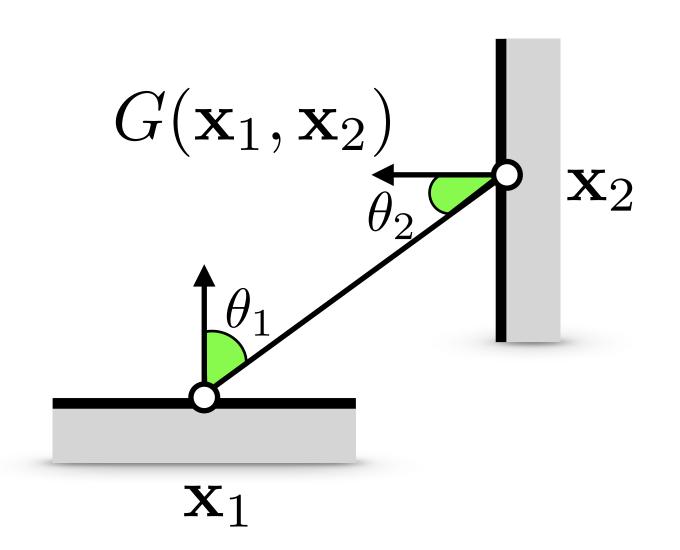


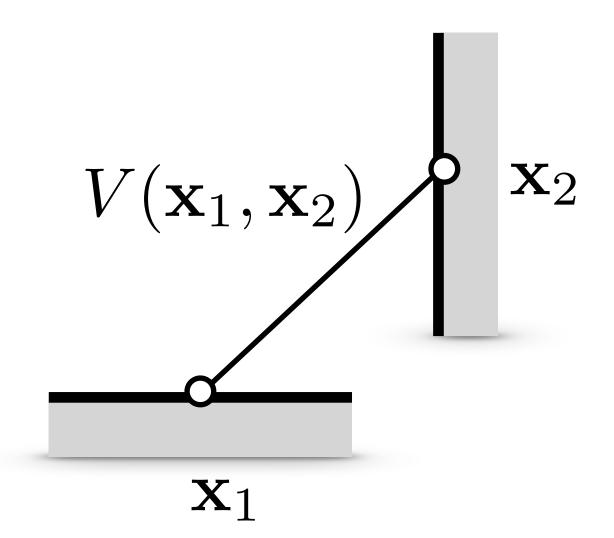






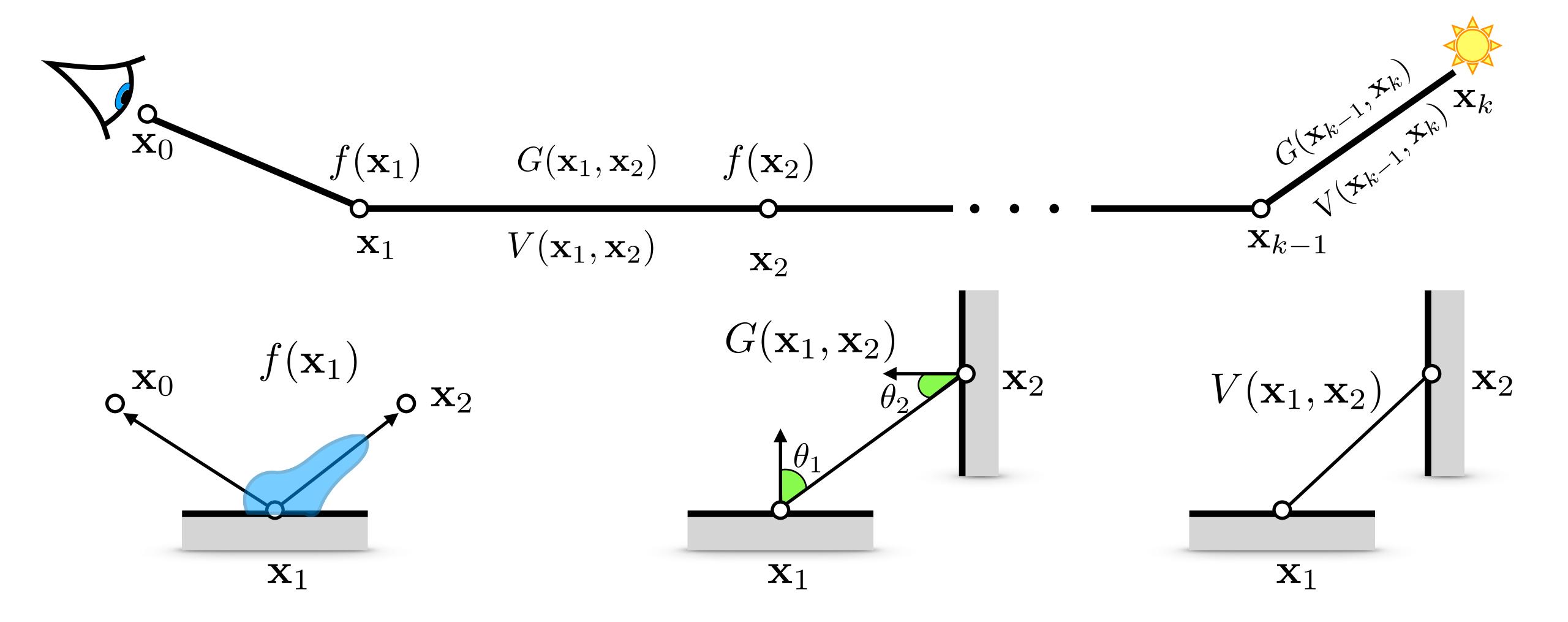






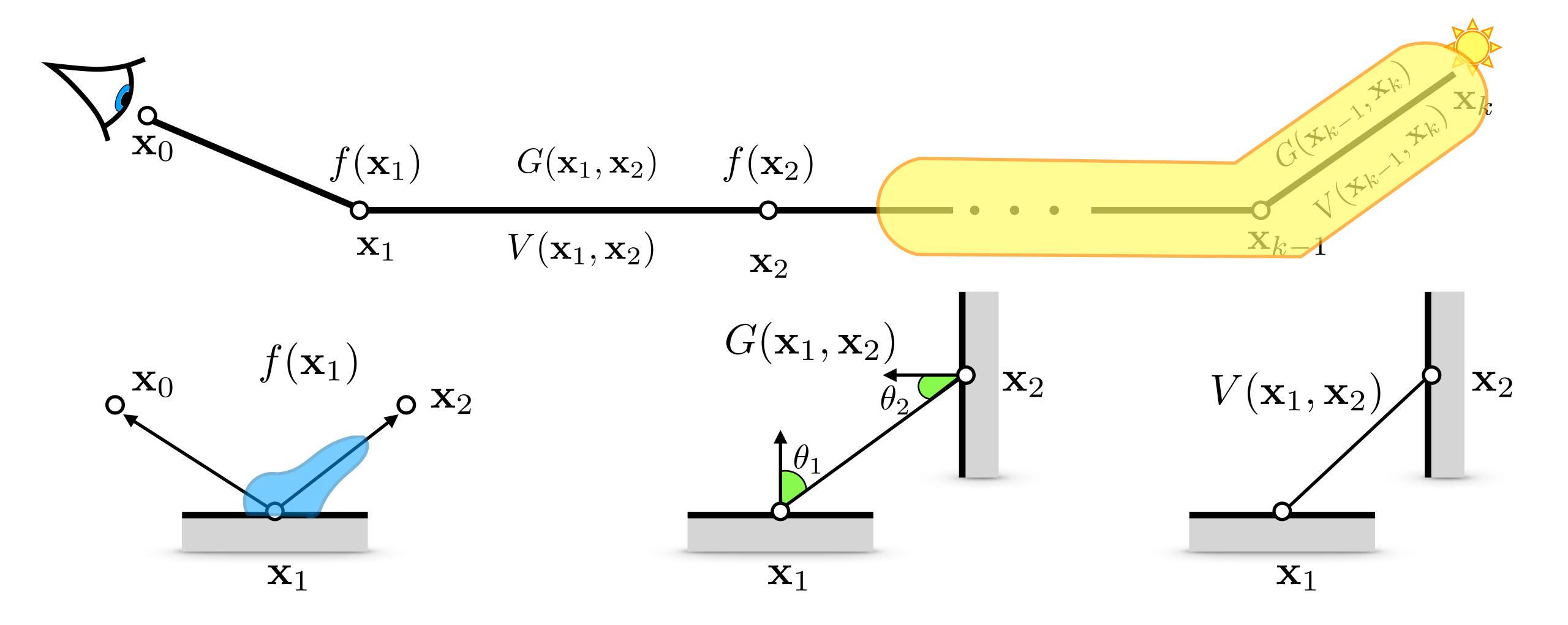






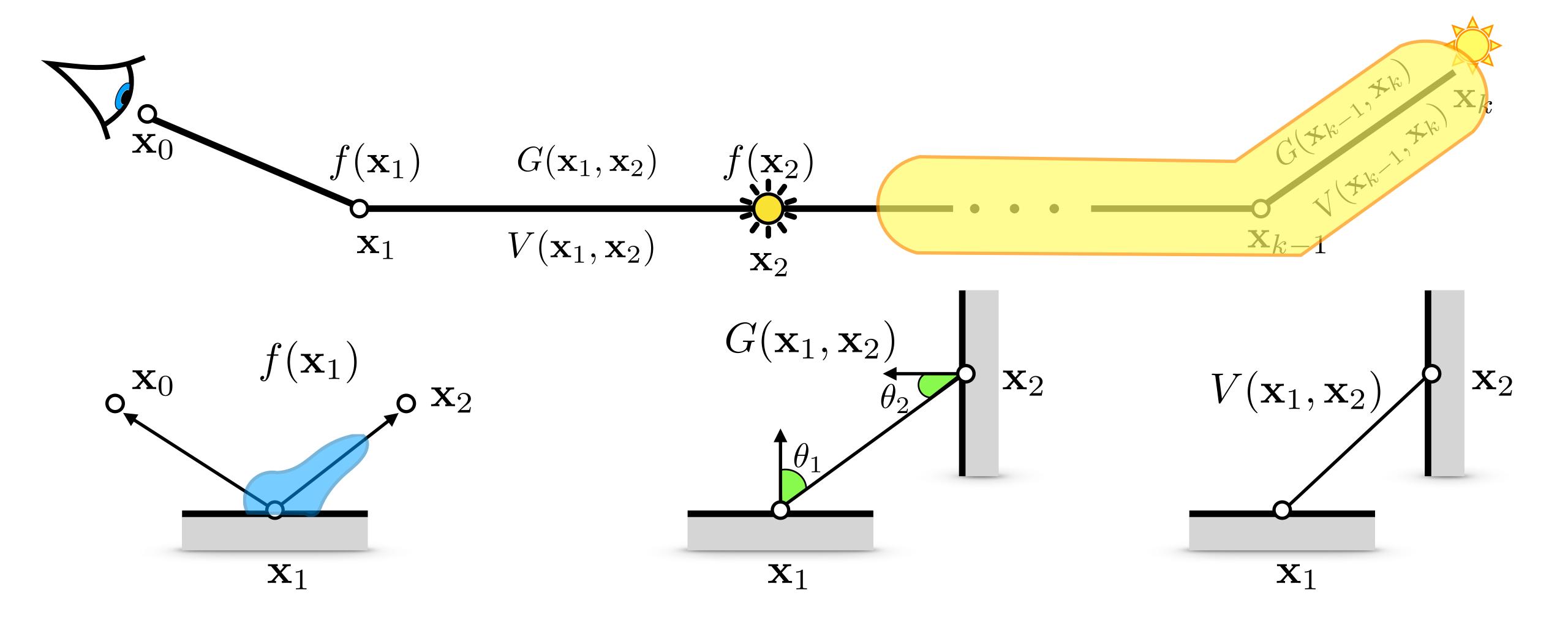
















$$L(\mathbf{x}_1 \to \mathbf{x}_0) = L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \int_A f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$



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$$L(\mathbf{x}_1 \to \mathbf{x}_0) \approx L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \sum_{k=1}^{N} f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2^i) V(\mathbf{x}_1, \mathbf{x}_2^i) f(\mathbf{x}_2^i) \Phi_i$$





$$L(\mathbf{x}_1 \to \mathbf{x}_0) = L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \int_A f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$
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$$L(\mathbf{x}_1 \to \mathbf{x}_0) = L_e(\mathbf{x}_1 \to \mathbf{x}_0) + \int_{\Delta} f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$

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Approximation using VPLs

 \mathbf{x}_2^i : position of the i-th VPL

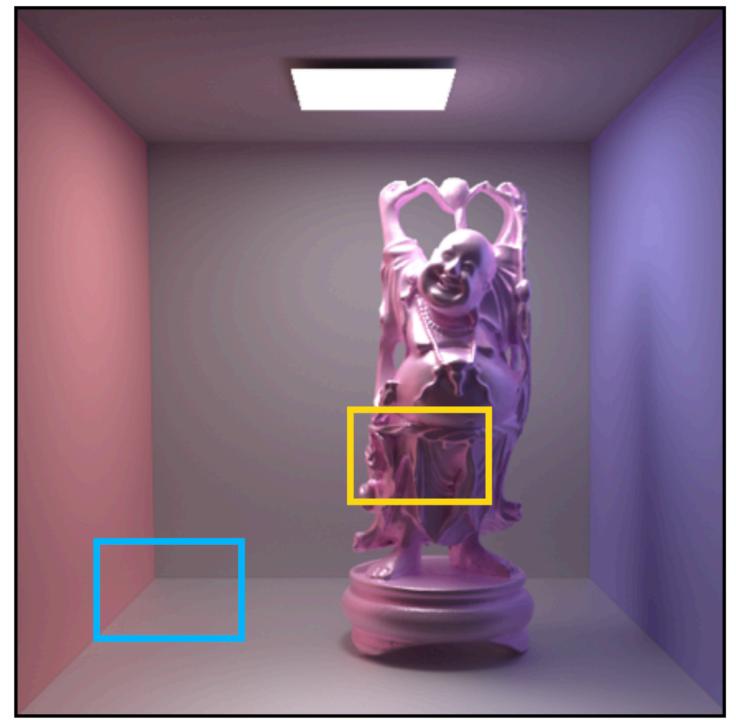
 Φ^i : flux of the i-th VPL

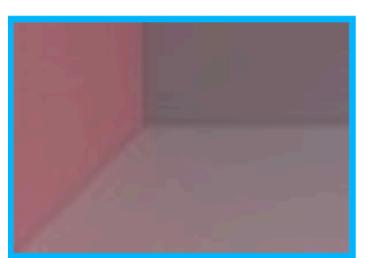
recursion is hidden in the generation of VPLs





Reference



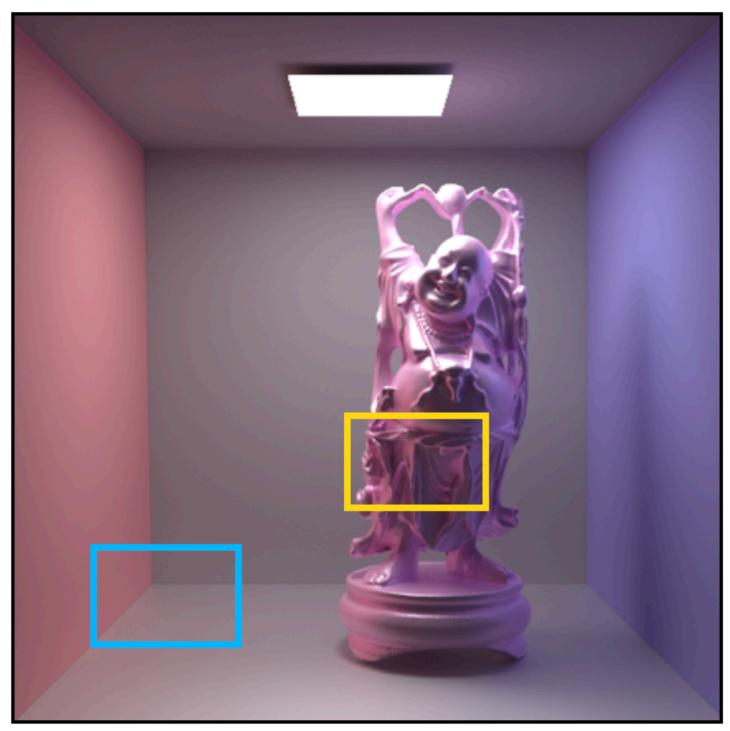


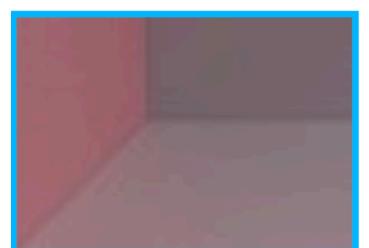






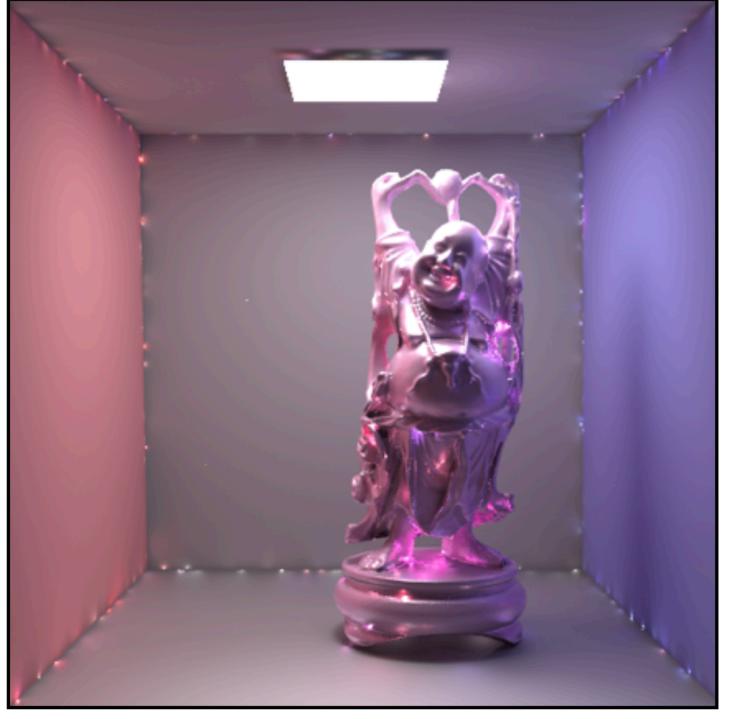
Reference







Approximation with VPLs





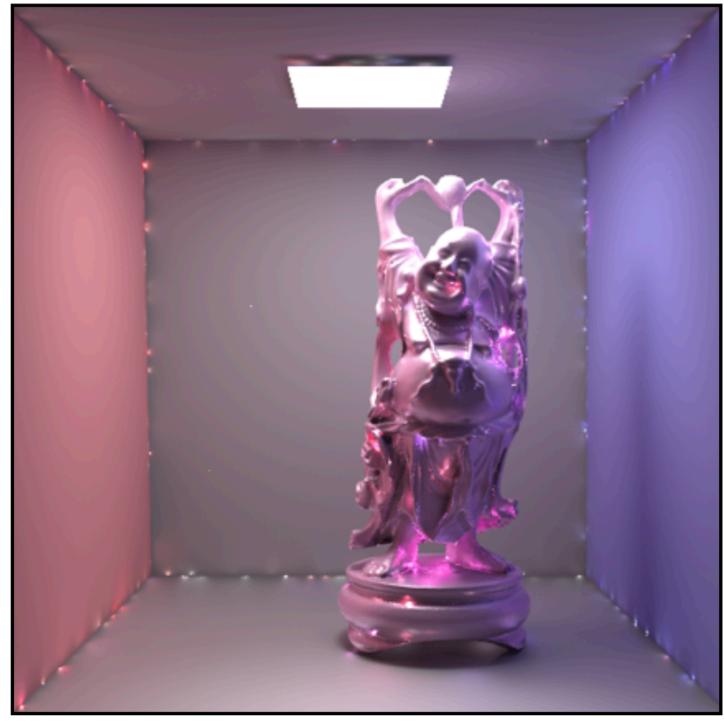


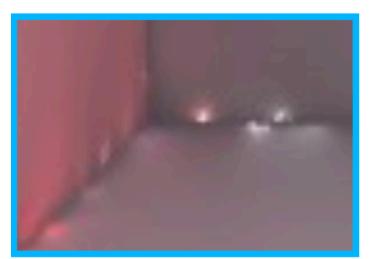




Why there are Splotches?

Approximation with VPLs







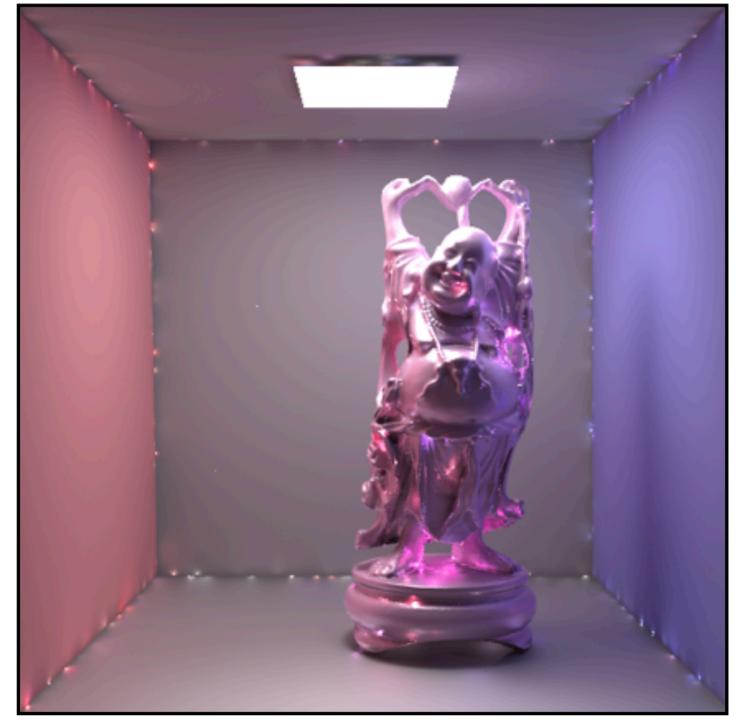


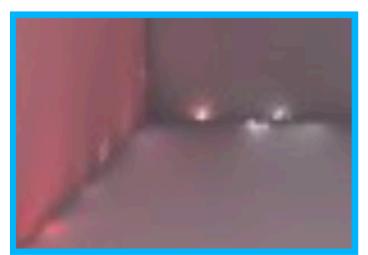


Why there are Splotches?

1. Have a look at the geometric term:

Approximation with VPLs







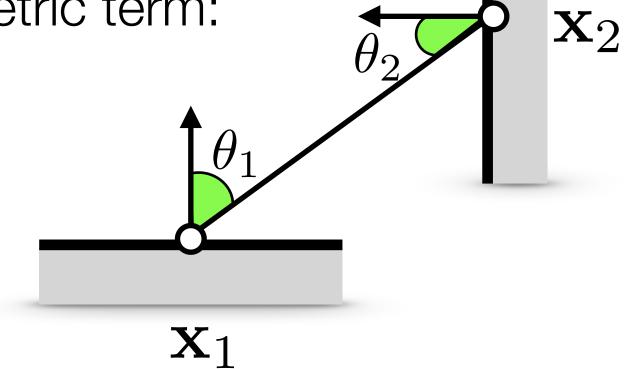


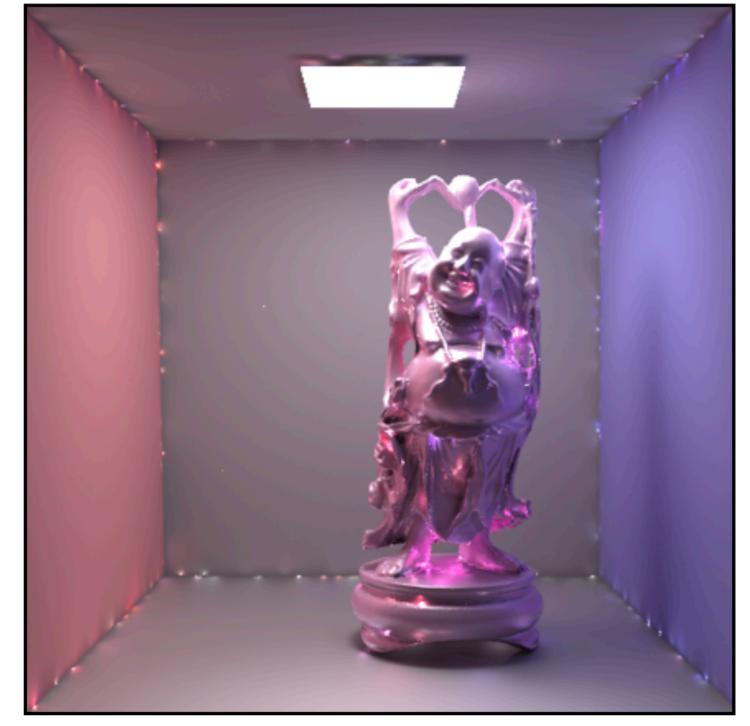


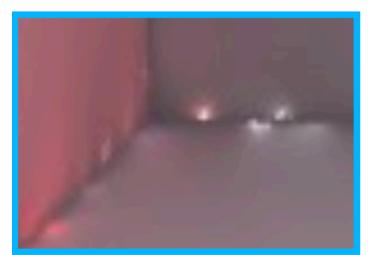
Why there are Splotches?

1. Have a look at the geometric term:

$$G(\mathbf{x}_1, \mathbf{x}_2) = \frac{\cos \theta_1 \cos \theta_2}{\|\mathbf{x}_1 - \mathbf{x}_2\|^2}$$









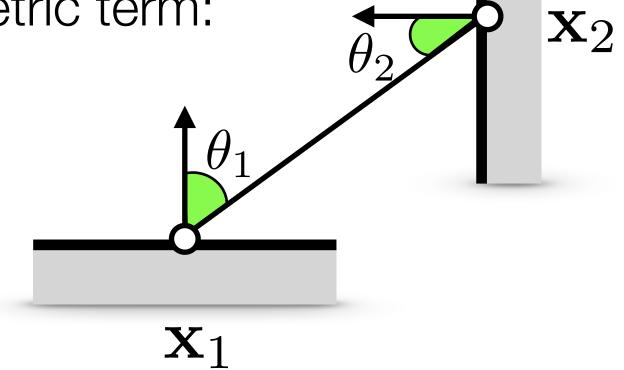


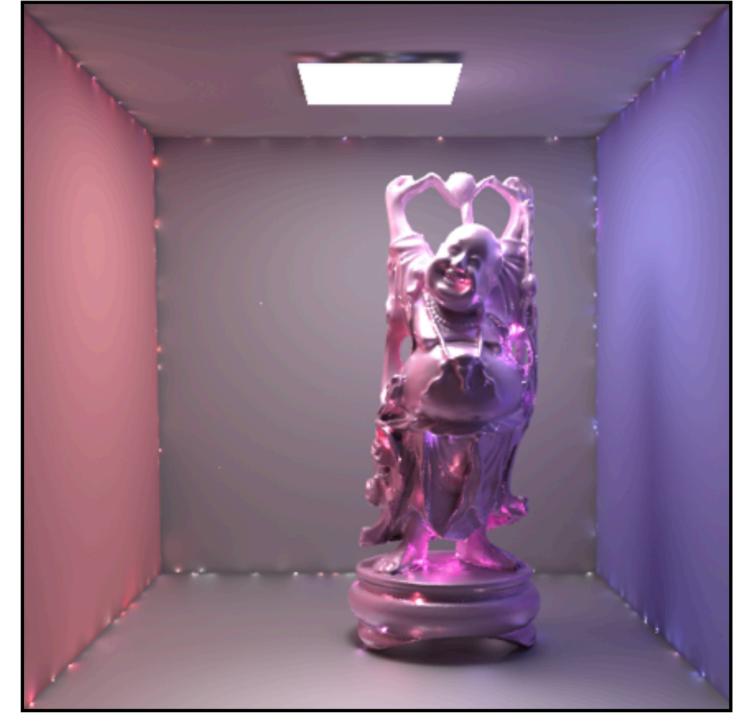


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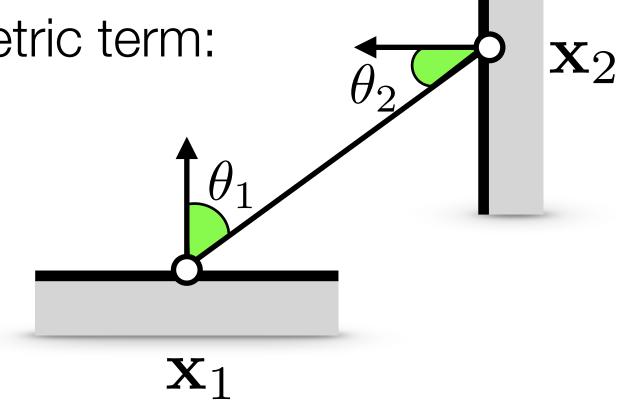


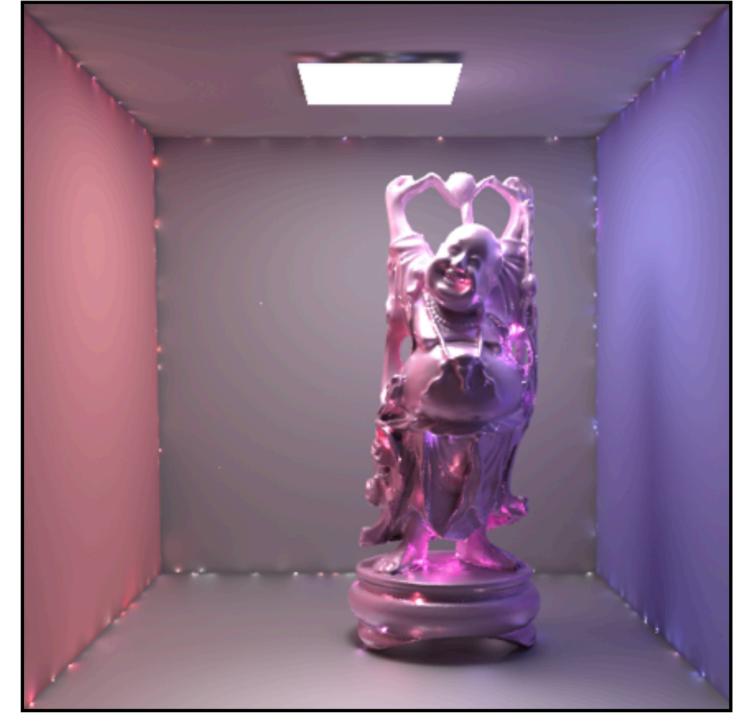


Why there are Splotches?

1. Have a look at the geometric term:

$$G(\mathbf{x}_1,\mathbf{x}_2) = \frac{\cos heta_1 \cos heta_2}{||\mathbf{x}_1 - \mathbf{x}_2||^2}$$
 Singularity







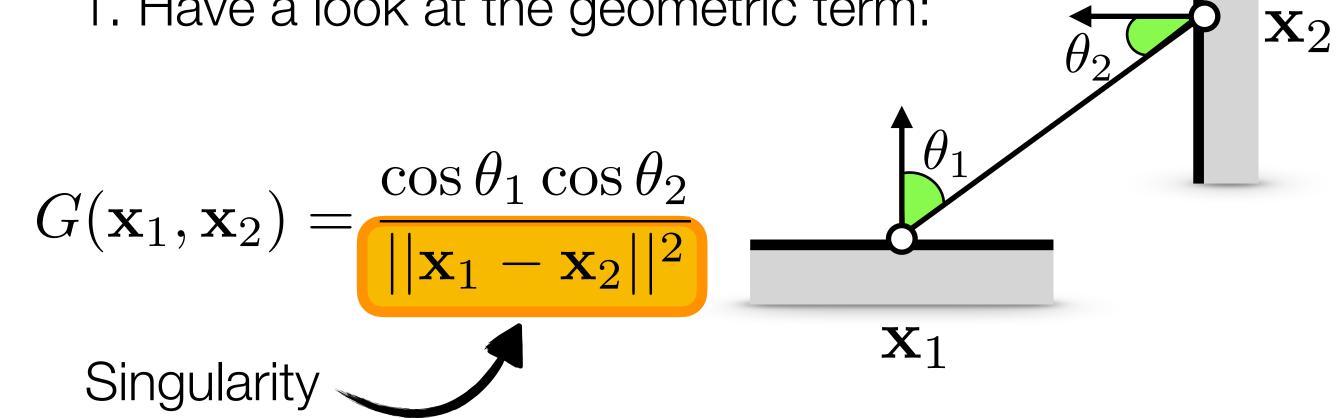




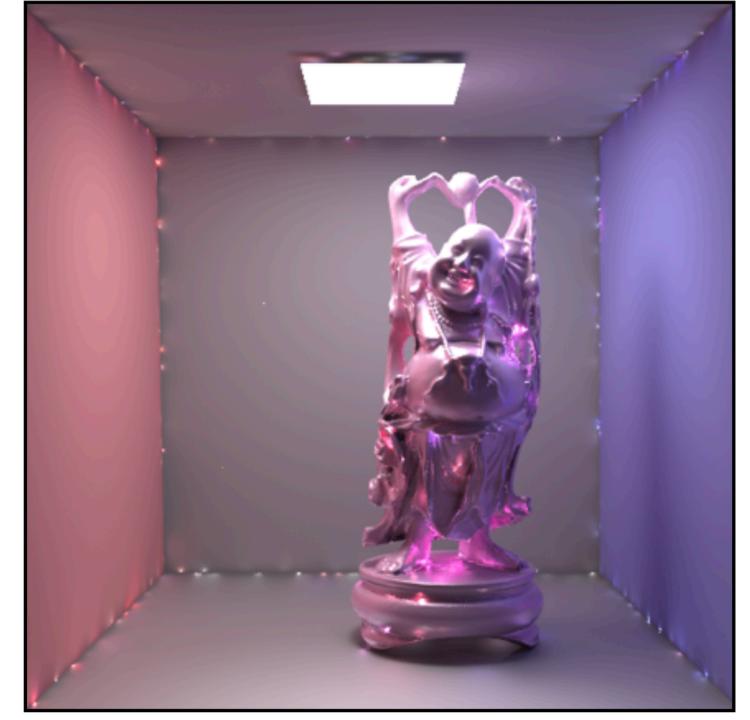


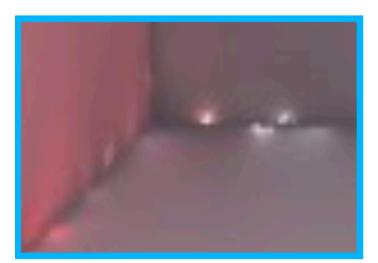
Why there are Splotches?

1. Have a look at the geometric term:



- 2. All points are lit by the same set of VPLs
 - introduce bad correlations











How to avoid Splotches?

Solutions:

- 1) Bound the geometry
 - remove energy, darkens the image
 - to get unbiased results, we need to compensate for the bounding
- 2) Distribute the flux of a VPL over area (volume)
 - redistributes energy, blurs the illumination
 - to get consistent results, progressively reduce the blurring





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- 1) Bound the geometry
 - prevent G from being very high
 - user-defined maximum value

$$G_b(\mathbf{x}_1, \mathbf{x}_2) = \min(G(\mathbf{x}_1, \mathbf{x}_2), b)$$





- 1) Bound the geometry
 - prevent G from being very high
 - user-defined maximum value

$$G_b(\mathbf{x}_1, \mathbf{x}_2) = \min(G(\mathbf{x}_1, \mathbf{x}_2), b)$$

Advantage:

- extremely simple and fast

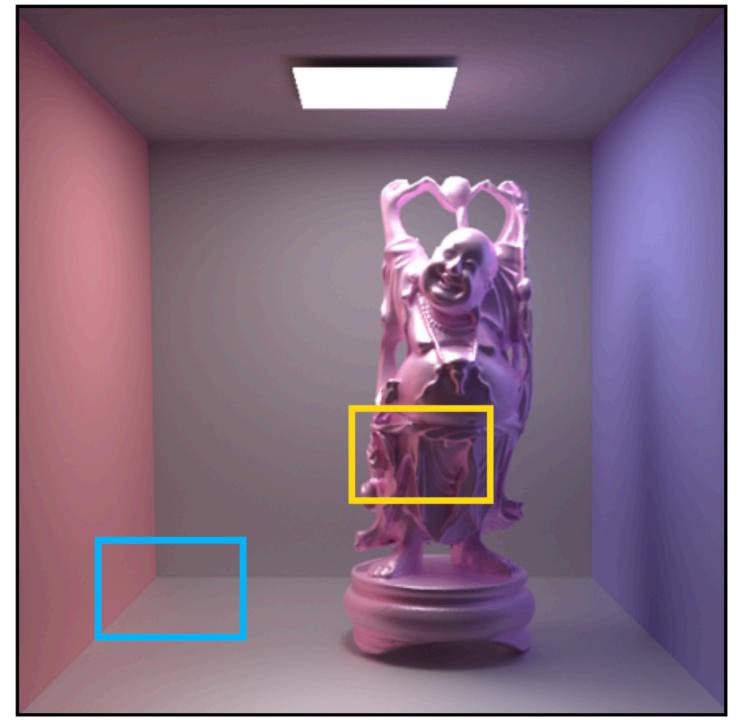
Disadvantages:

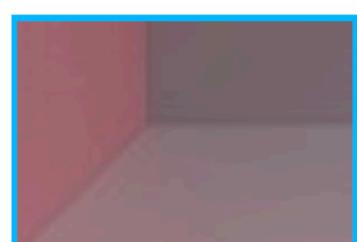
- removes energy, darkens the image





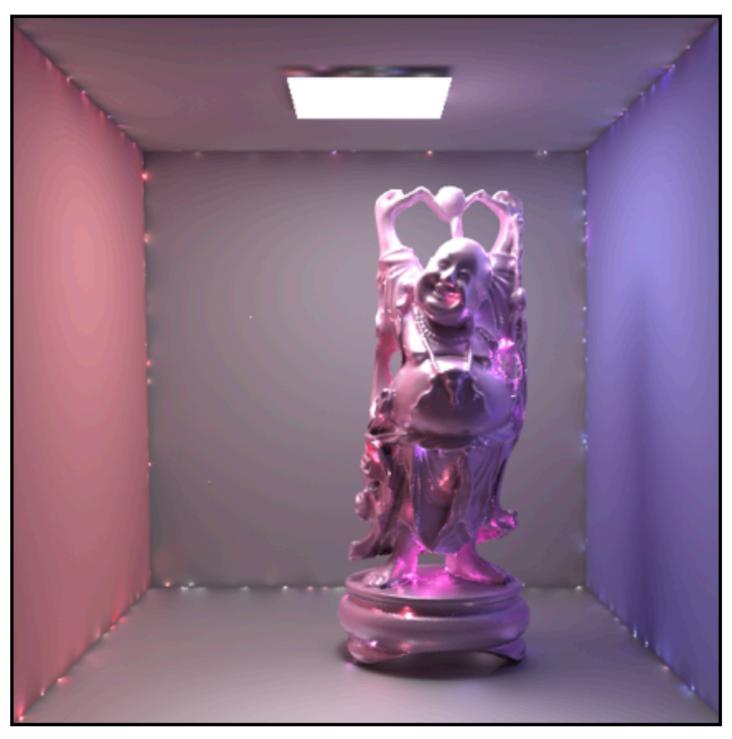
Reference



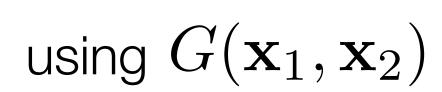




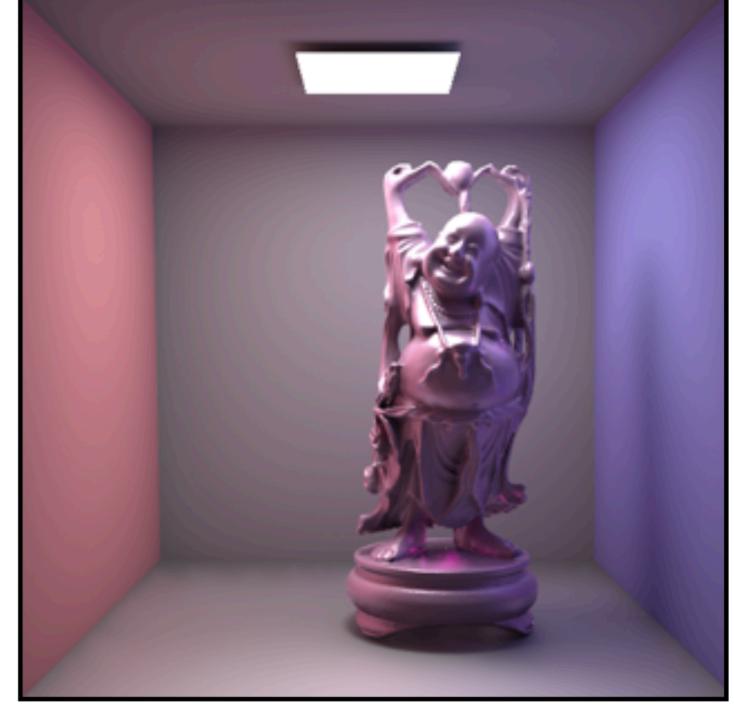
VPLs

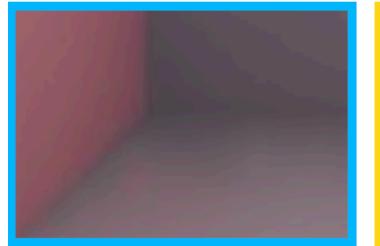






VPLs with bounded G





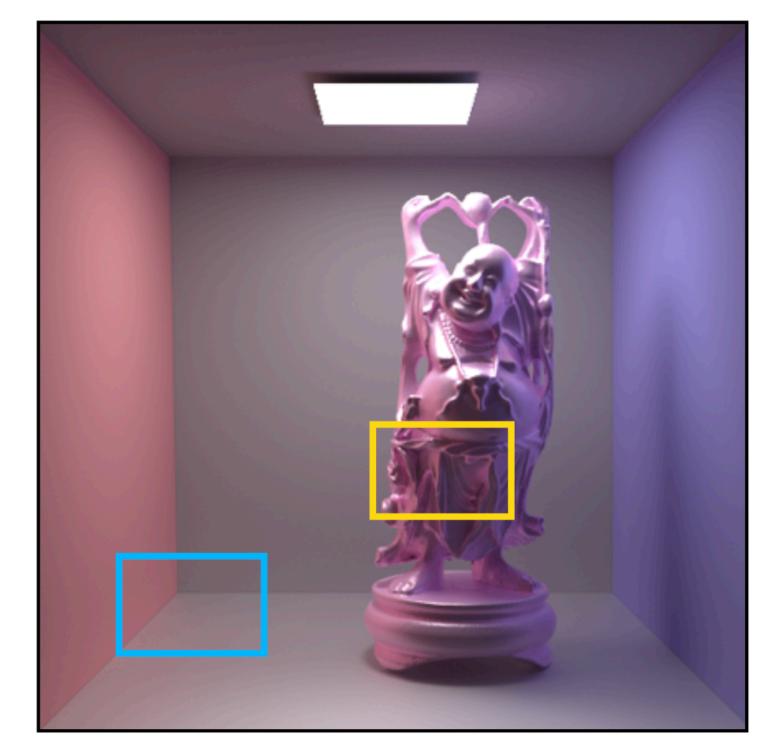


using $G_b(\mathbf{x}_1,\mathbf{x}_2)$





Reference







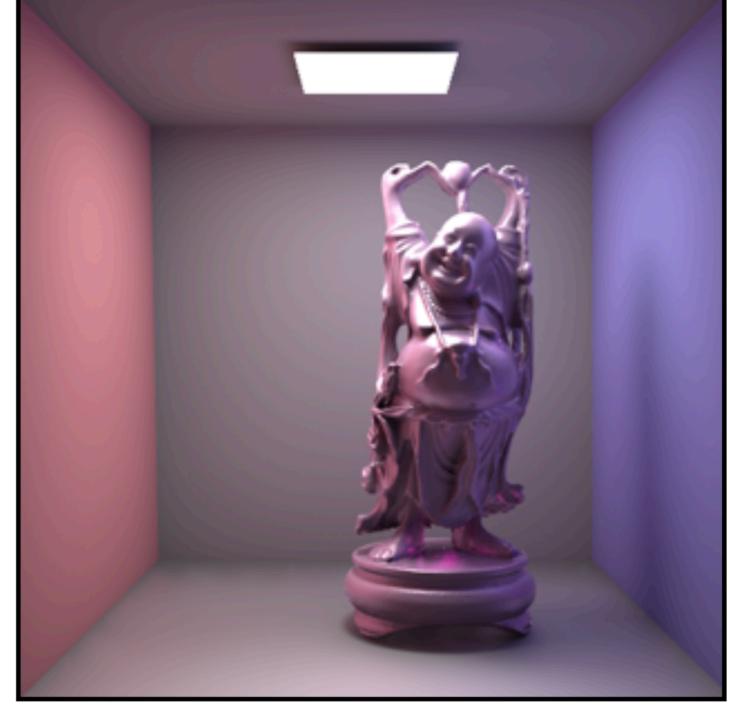
Difference



$$G(\mathbf{x}_1,\mathbf{x}_2)-G_b(\mathbf{x}_1,\mathbf{x}_2)$$

We need to compensate for the energy loss!

VPLs with bounded G







using $G_b(\mathbf{x}_1,\mathbf{x}_2)$





$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1) G(\mathbf{x}_1, \mathbf{x}_2) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$



Light transport operator ${f T}$

$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1)G(\mathbf{x}_1, \mathbf{x}_2)V(\mathbf{x}_1, \mathbf{x}_2)L(\mathbf{x}_2 \to \mathbf{x}_1)dA(\mathbf{x}_2)$$

Bounded light transport operator \mathbf{T}_b

$$(\mathbf{T}_b L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1) \min(G(\mathbf{x}_1, \mathbf{x}_2), b) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$





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Residual light transport operator \mathbf{T}_r

$$(\mathbf{T}_r L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1) \max(G(\mathbf{x}_1, \mathbf{x}_2) - b, 0) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$





Light transport operator ${f T}$

$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1)G(\mathbf{x}_1, \mathbf{x}_2)V(\mathbf{x}_1, \mathbf{x}_2)L(\mathbf{x}_2 \to \mathbf{x}_1)dA(\mathbf{x}_2)$$

Bounded light transport operator \mathbf{T}_b

$$(\mathbf{T}_b L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1) \min(G(\mathbf{x}_1, \mathbf{x}_2), b) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$

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Light transport operator ${f T}$

$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = (\mathbf{T}_b L)(\mathbf{x}_1 \to \mathbf{x}_0) + (\mathbf{T}_r L)(\mathbf{x}_1 \to \mathbf{x}_0)$$

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Residual light transport operator \mathbf{T}_r

$$(\mathbf{T}_r L)(\mathbf{x}_1 \to \mathbf{x}_0) = \int_A f(\mathbf{x}_1) \max(G(\mathbf{x}_1, \mathbf{x}_2) - b, 0) V(\mathbf{x}_1, \mathbf{x}_2) L(\mathbf{x}_2 \to \mathbf{x}_1) dA(\mathbf{x}_2)$$

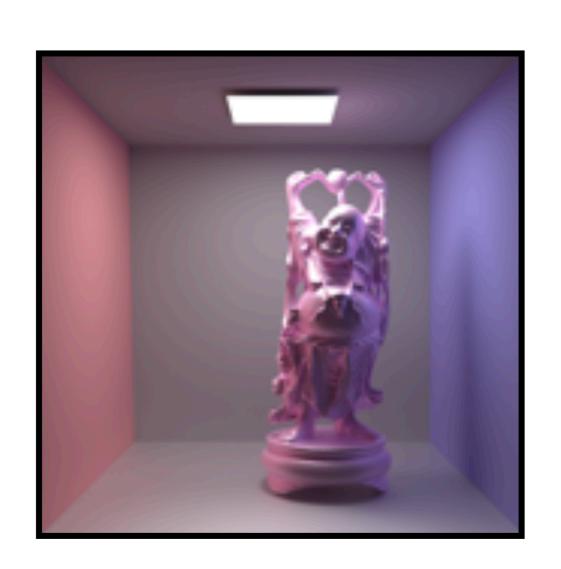




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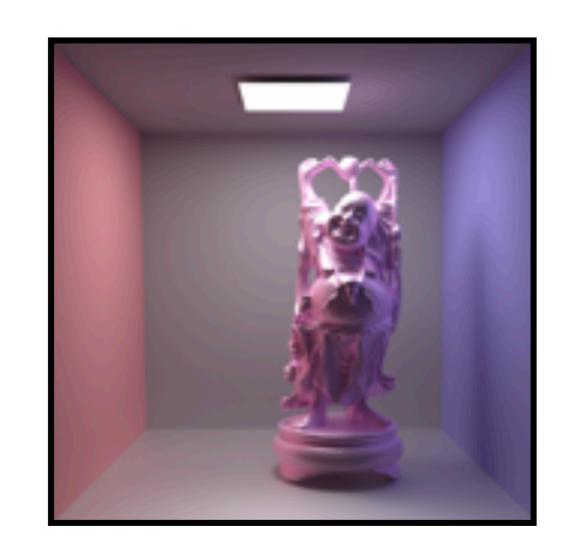
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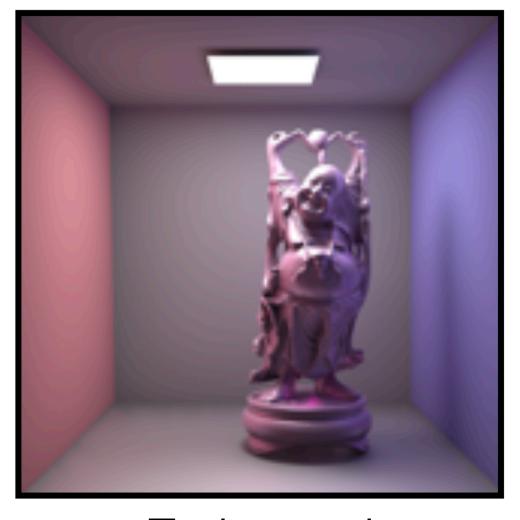
Estimated using VPLs





$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = (\mathbf{T}_b L)(\mathbf{x}_1 \to \mathbf{x}_0) + (\mathbf{T}_r L)(\mathbf{x}_1 \to \mathbf{x}_0)$$





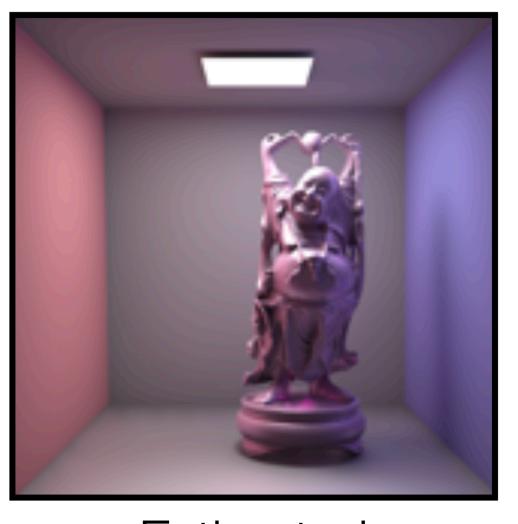
Estimated using VPLs





$$(\mathbf{T}L)(\mathbf{x}_1 \to \mathbf{x}_0) = (\mathbf{T}_b L)(\mathbf{x}_1 \to \mathbf{x}_0) + (\mathbf{T}_r L)(\mathbf{x}_1 \to \mathbf{x}_0)$$





Estimated using VPLs



Estimated differently



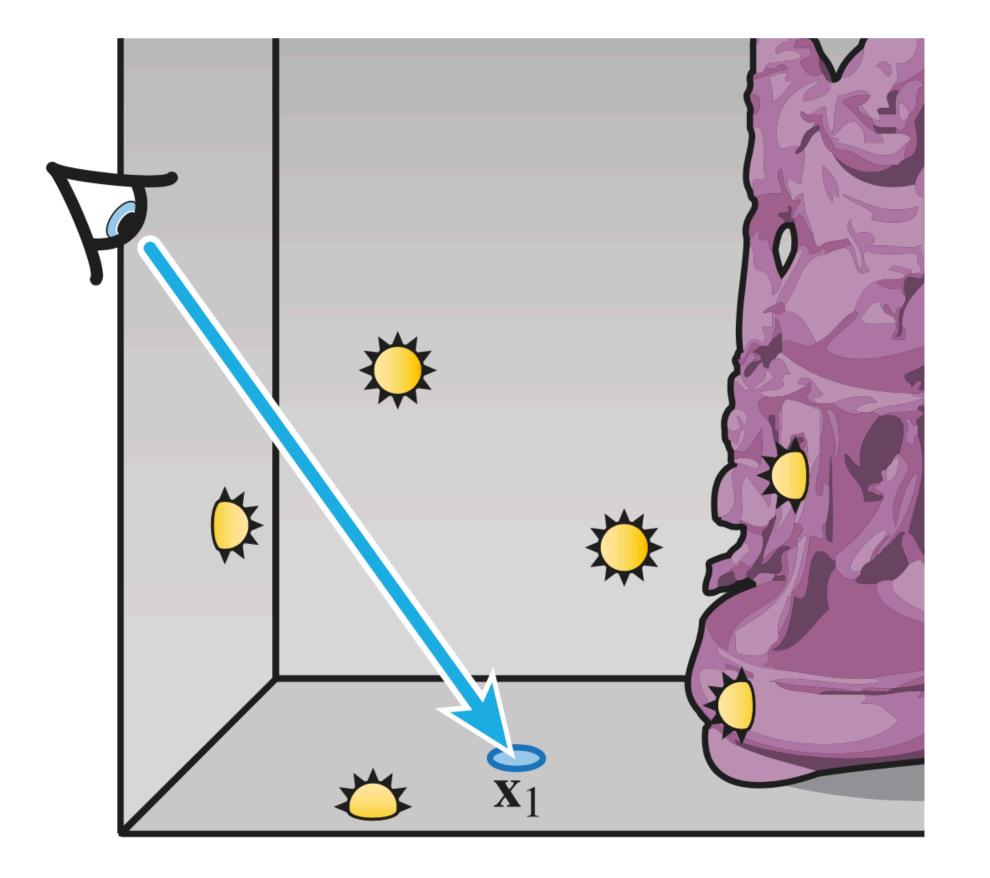


How to compute residual transport?





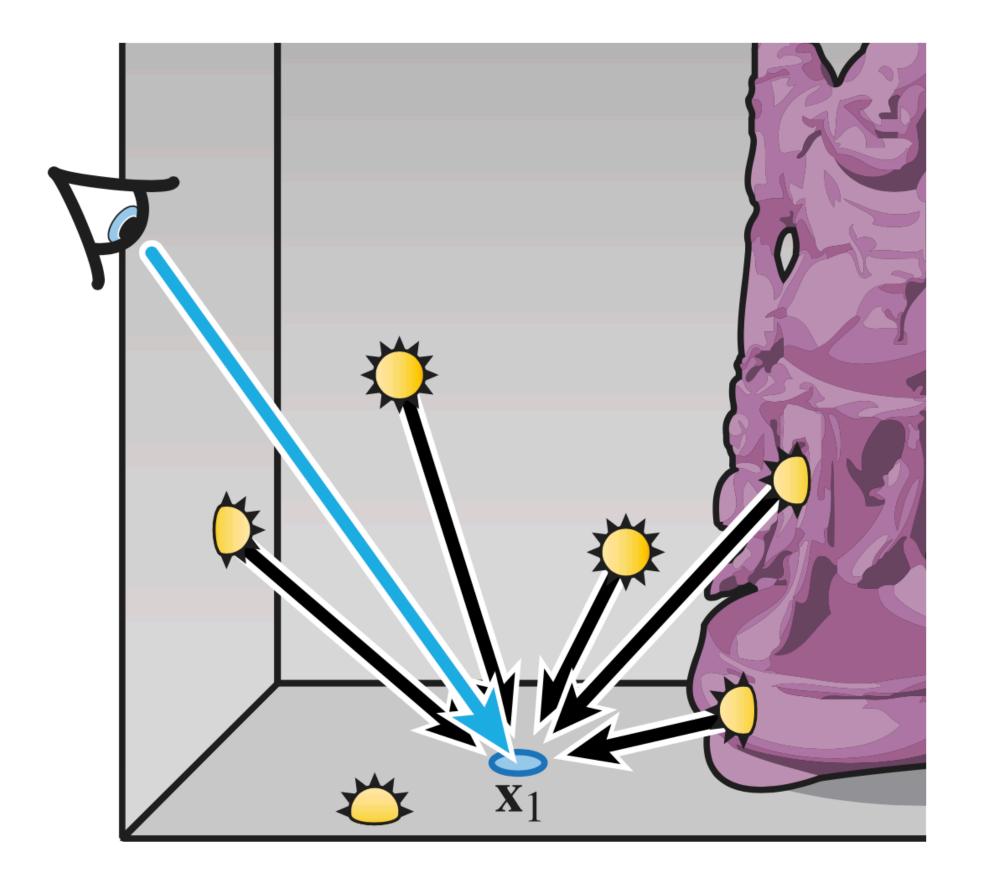
Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

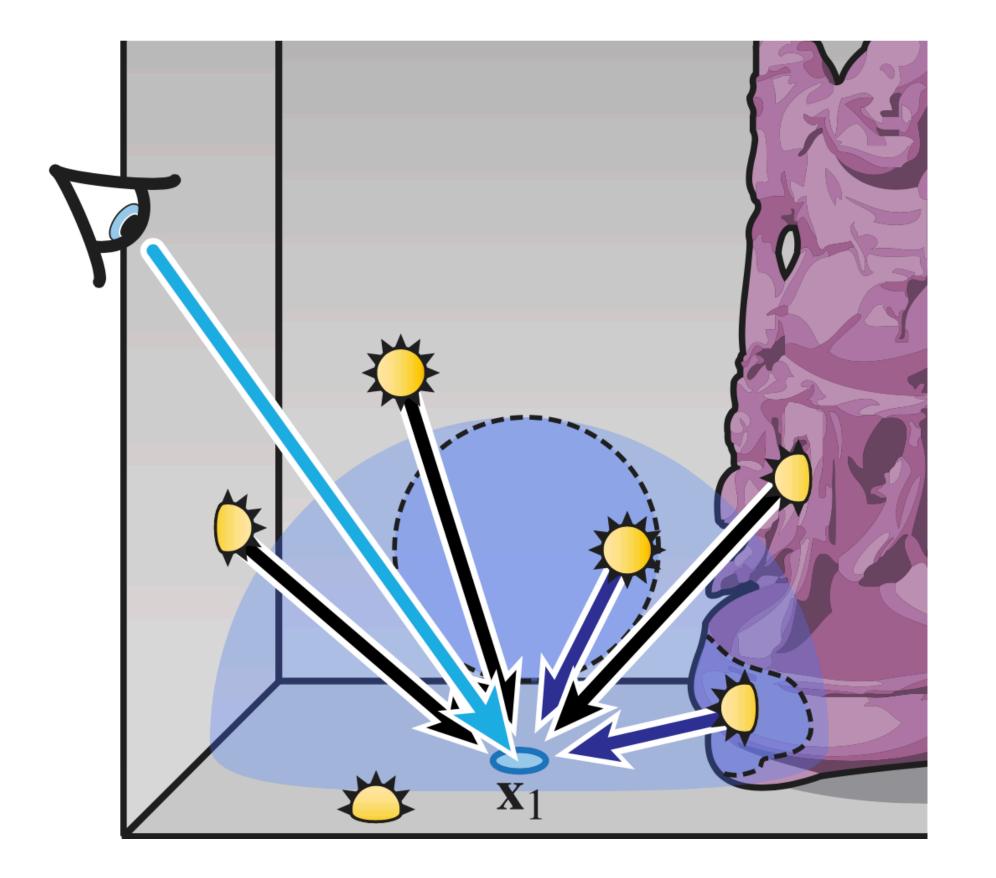
Trace paths to compute the compensation term (residual transport)



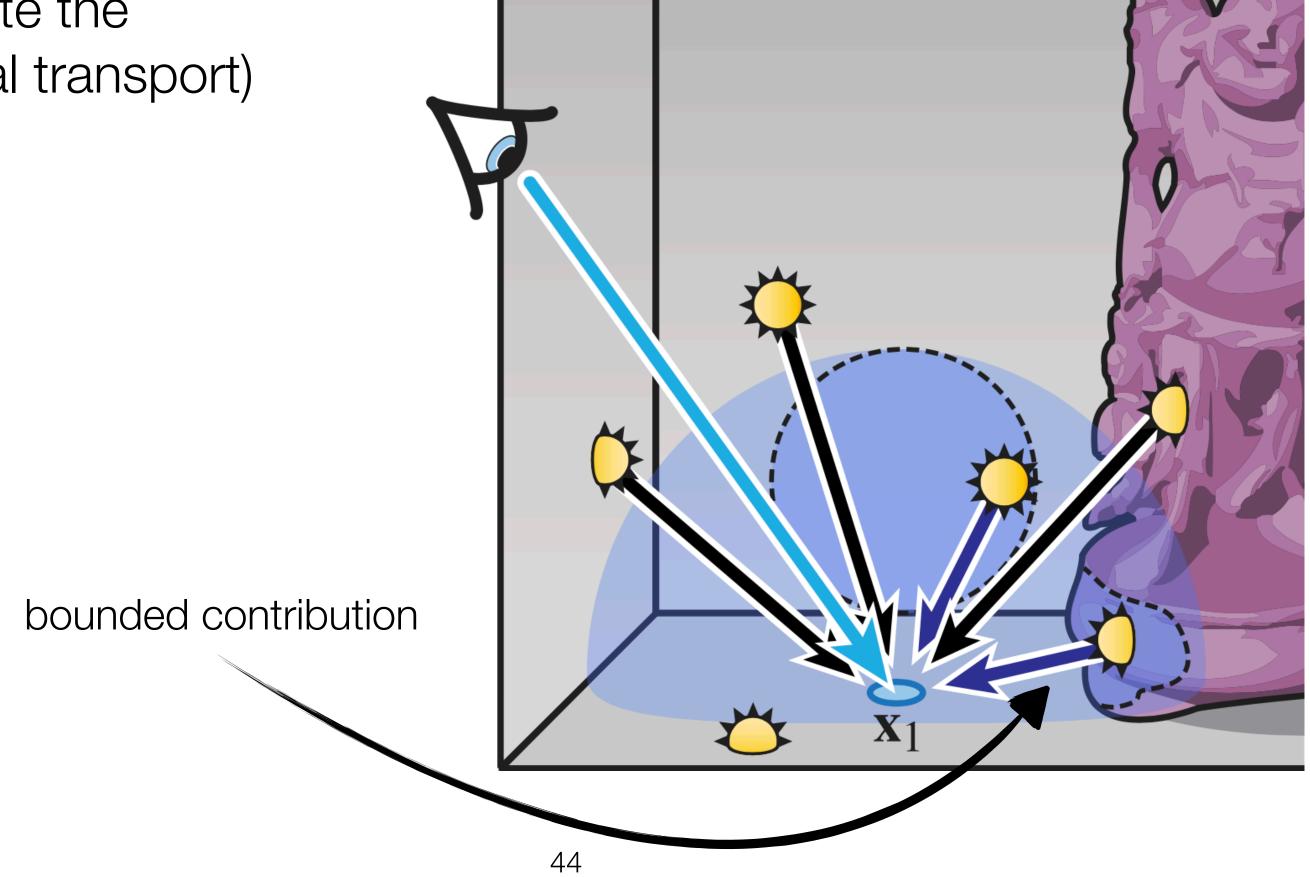


Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

Trace paths to compute the compensation term (residual transport)



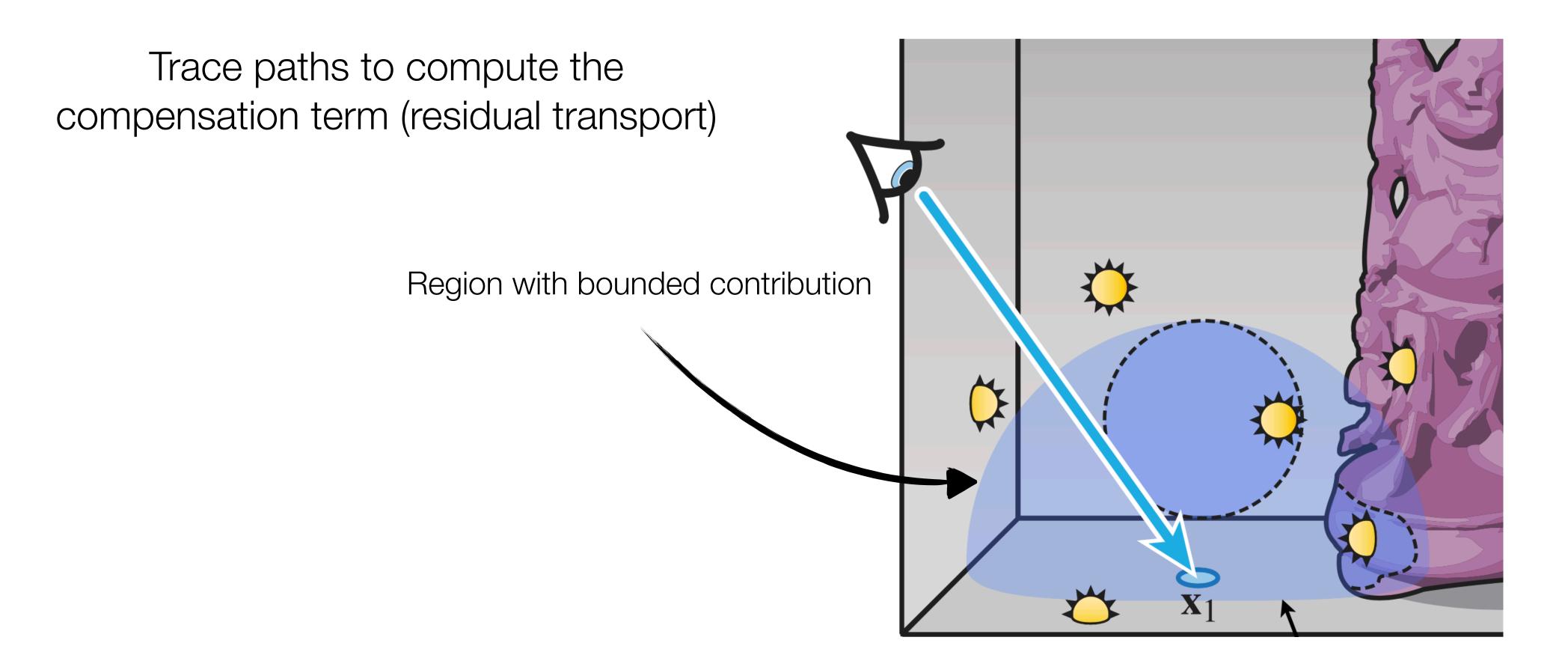
Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]







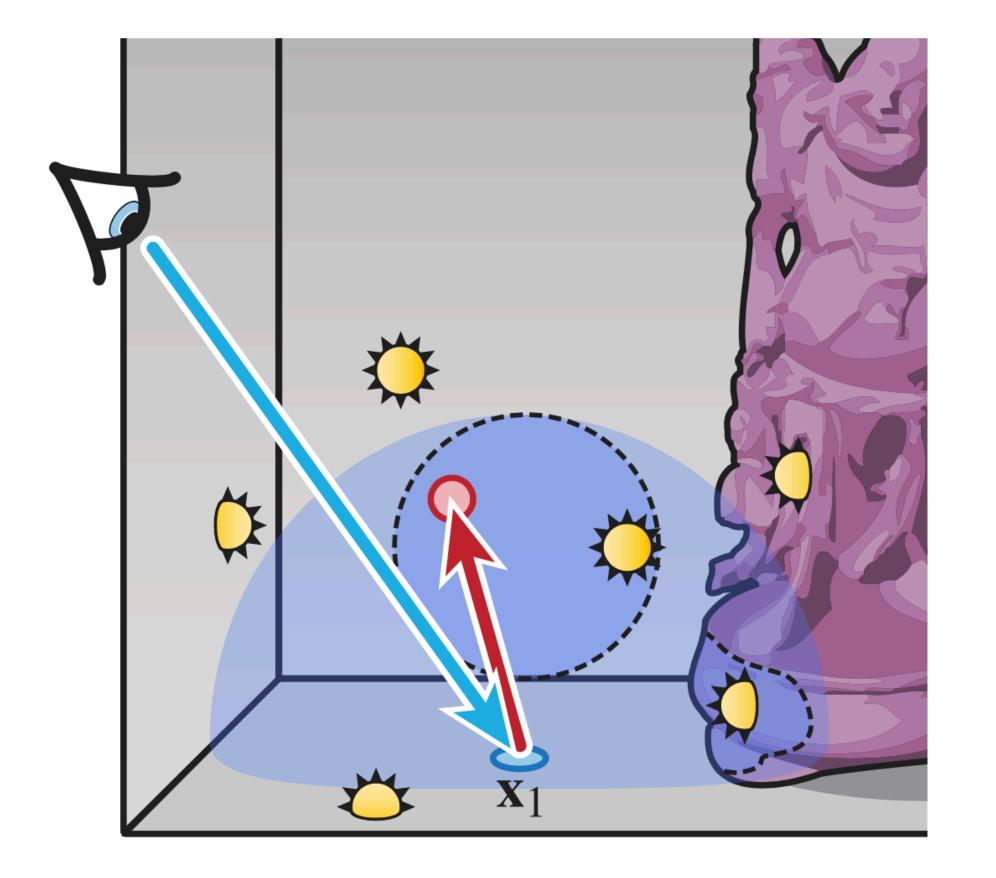
Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]





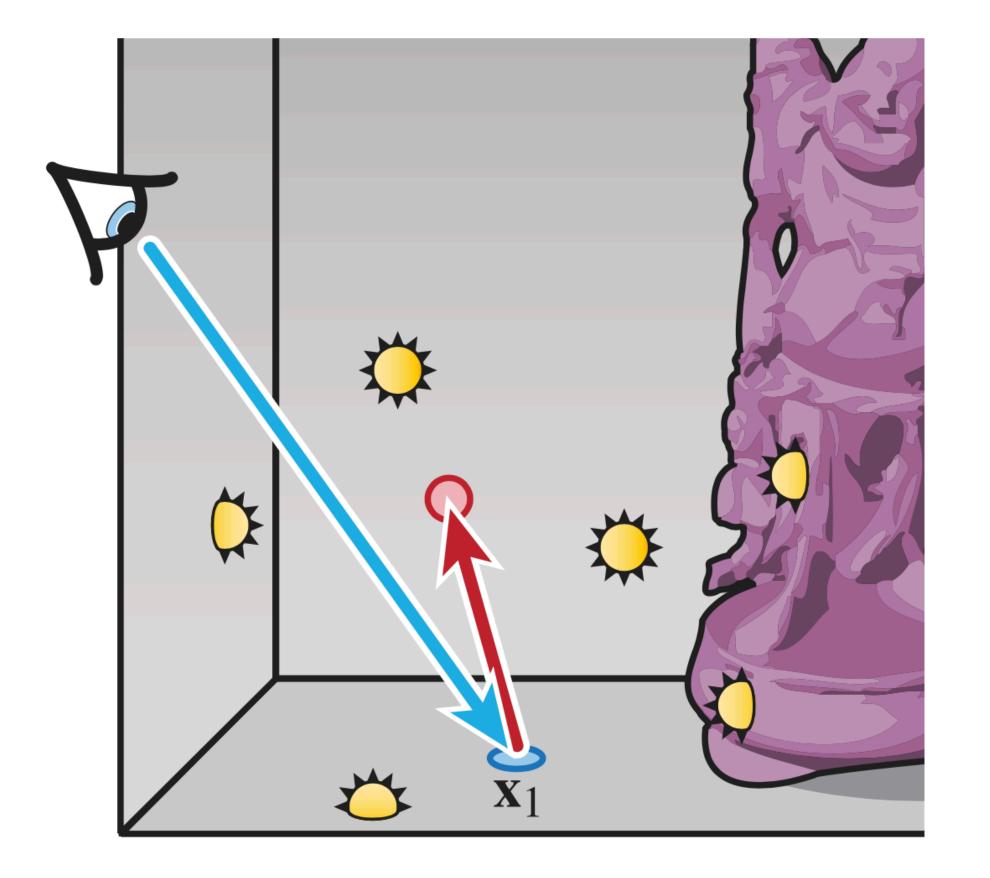


Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]



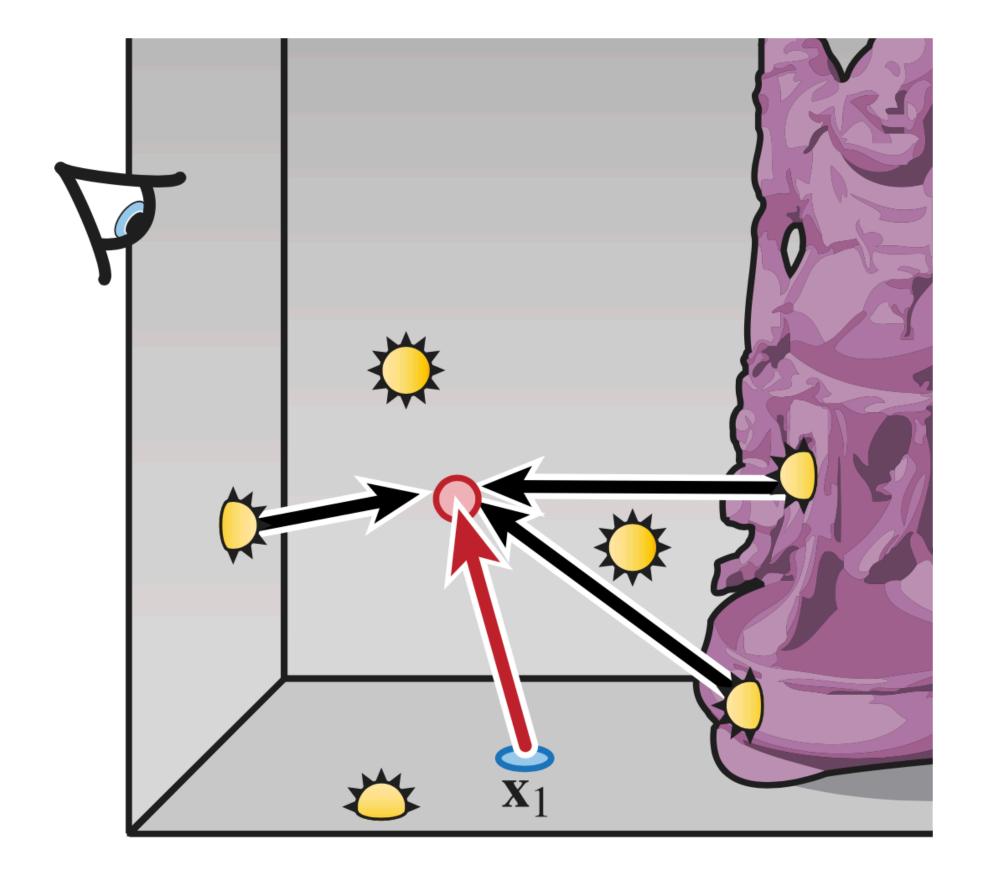


Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]





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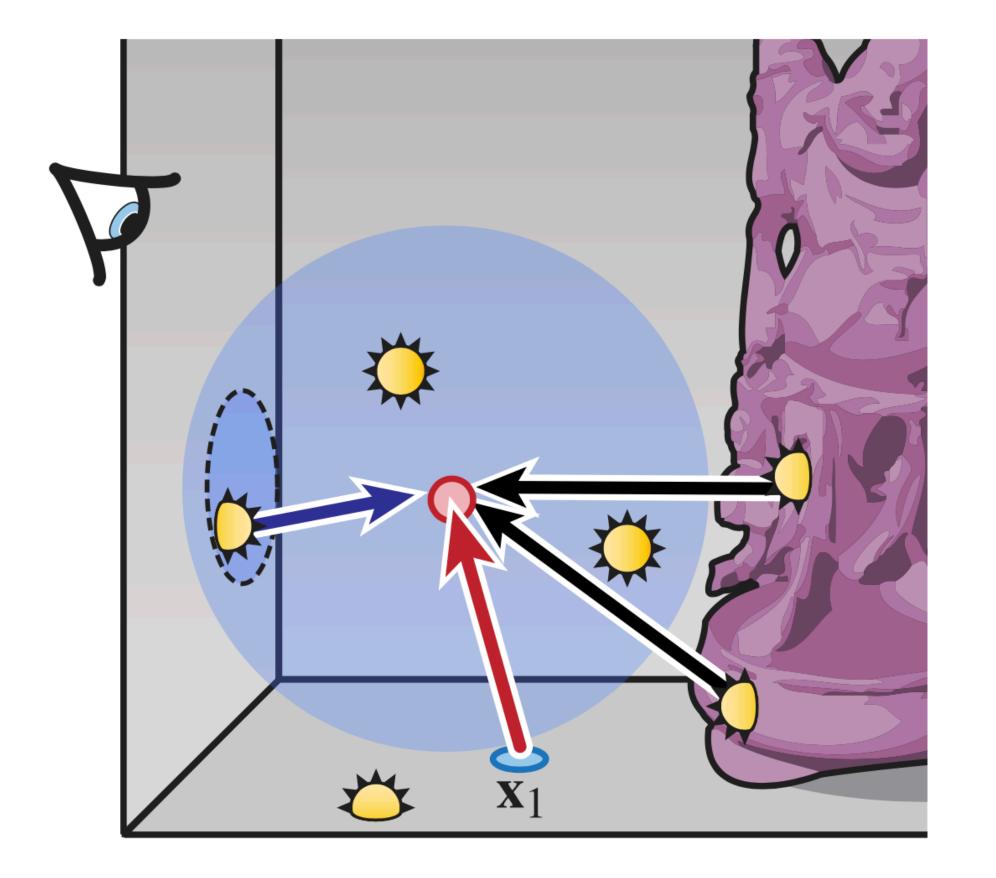






Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

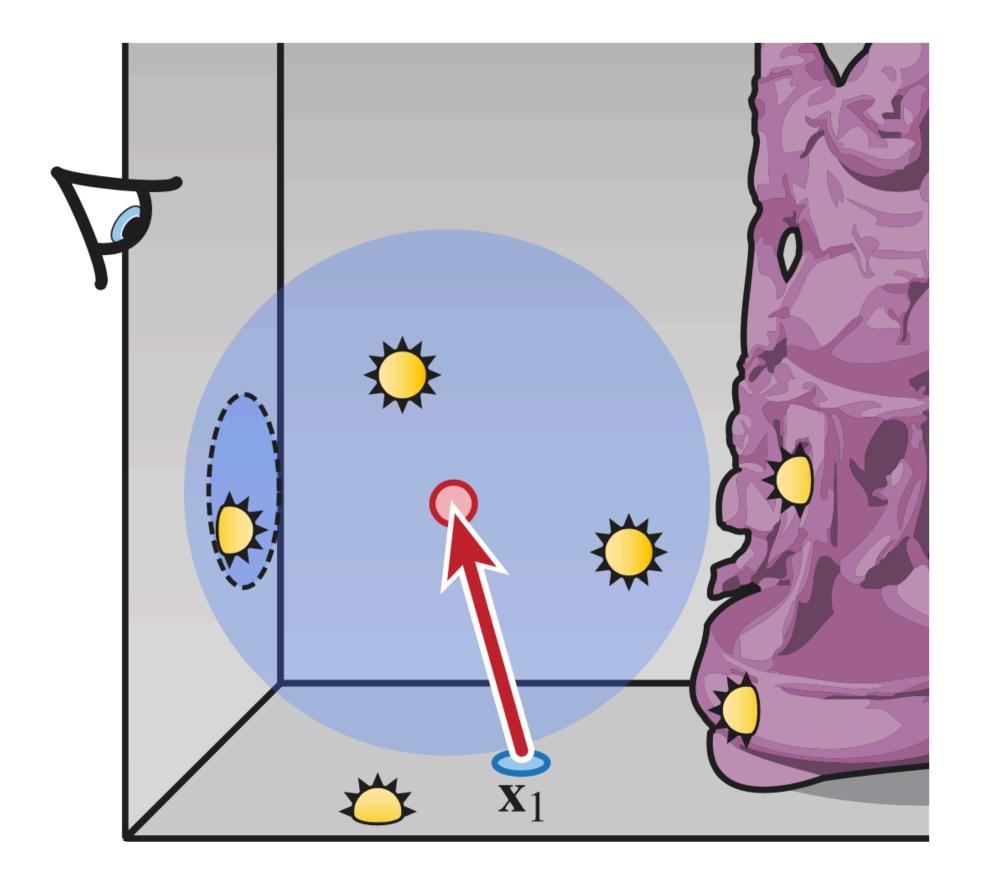
Trace paths to compute the compensation term (residual transport)





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

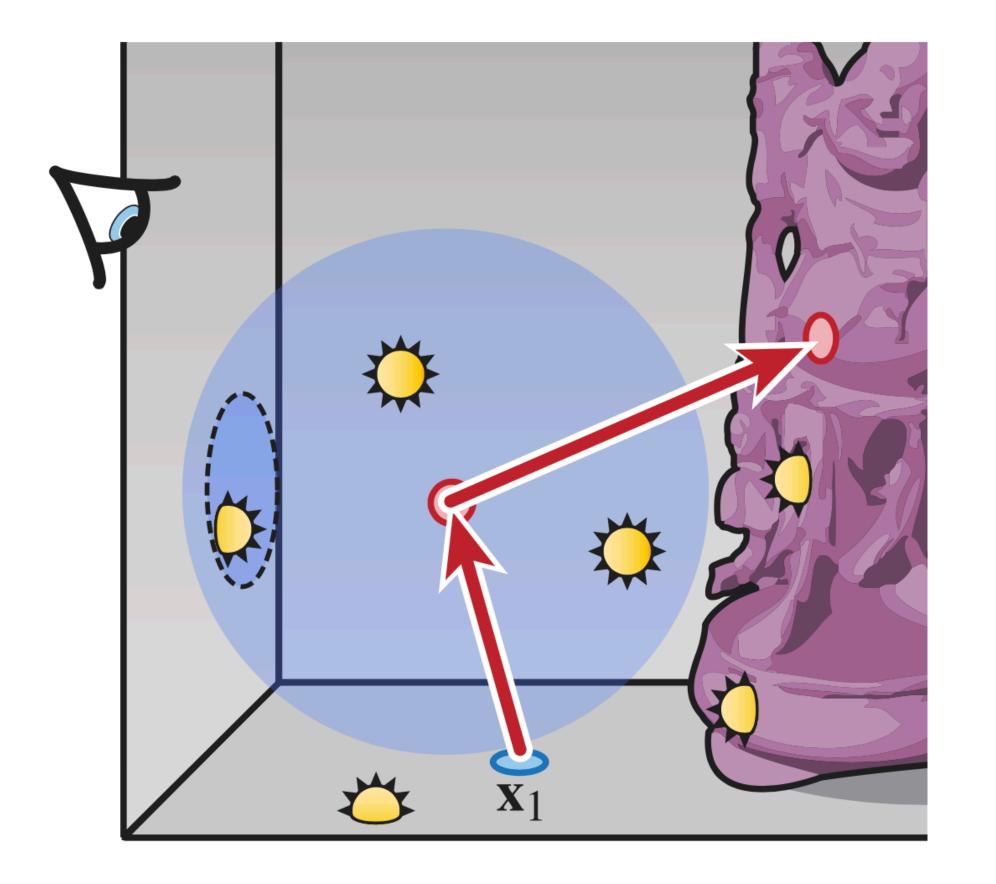
Trace paths to compute the compensation term (residual transport)





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

Trace paths to compute the compensation term (residual transport)





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

Advantages:

Recover all missing energy

Makes the algorithm unbiased

Disadvantages:

Recursive, degenerate to path tracing

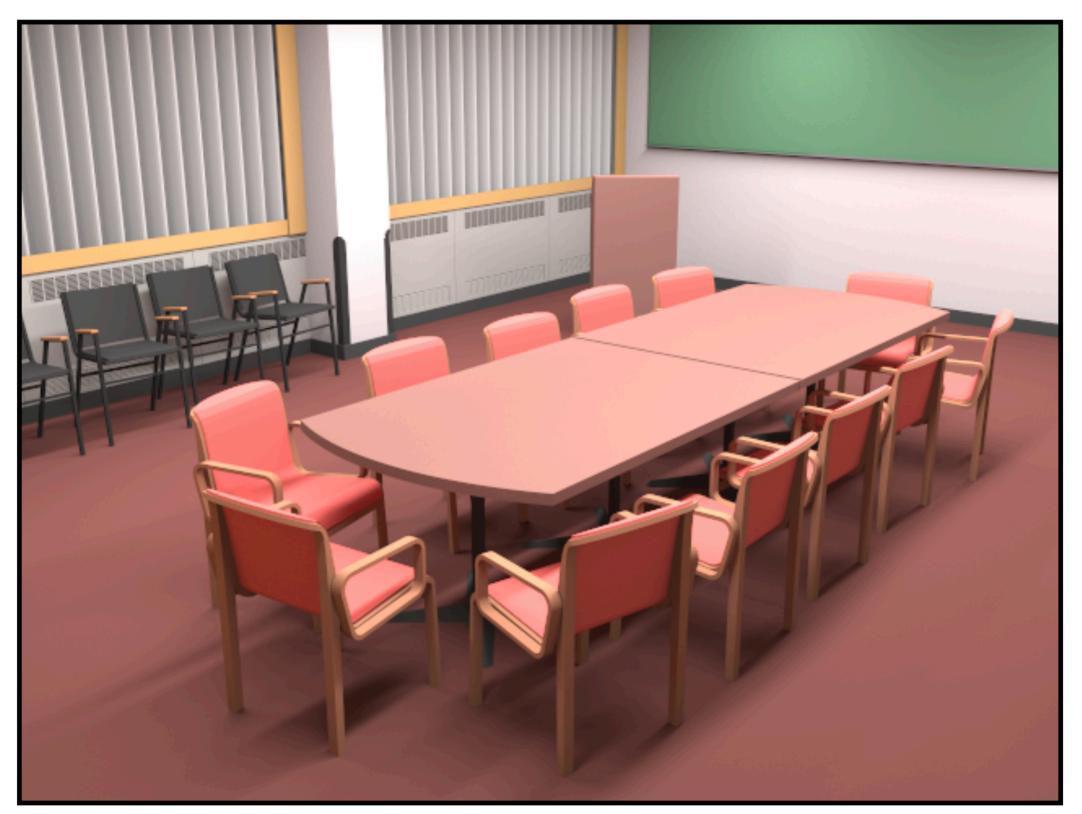
Very expensive: recovering 10% of energy may take 90% of the rendering time





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]





Bounding and Compensation

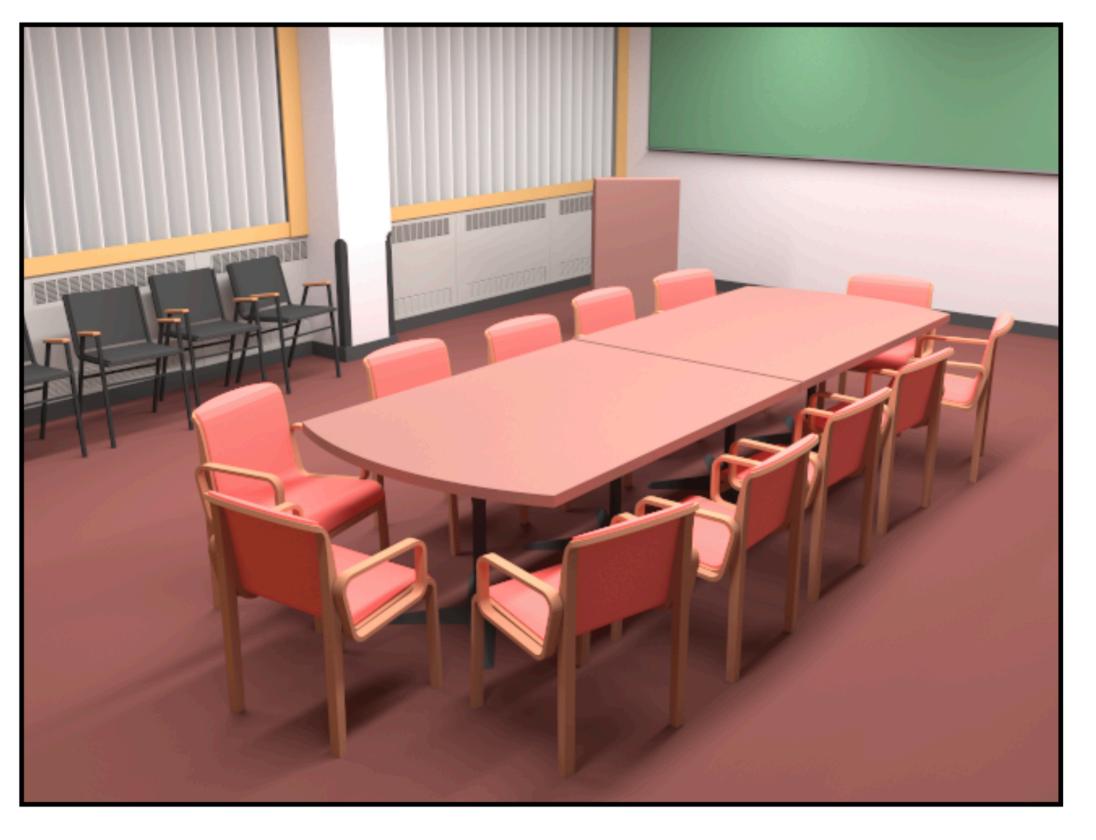


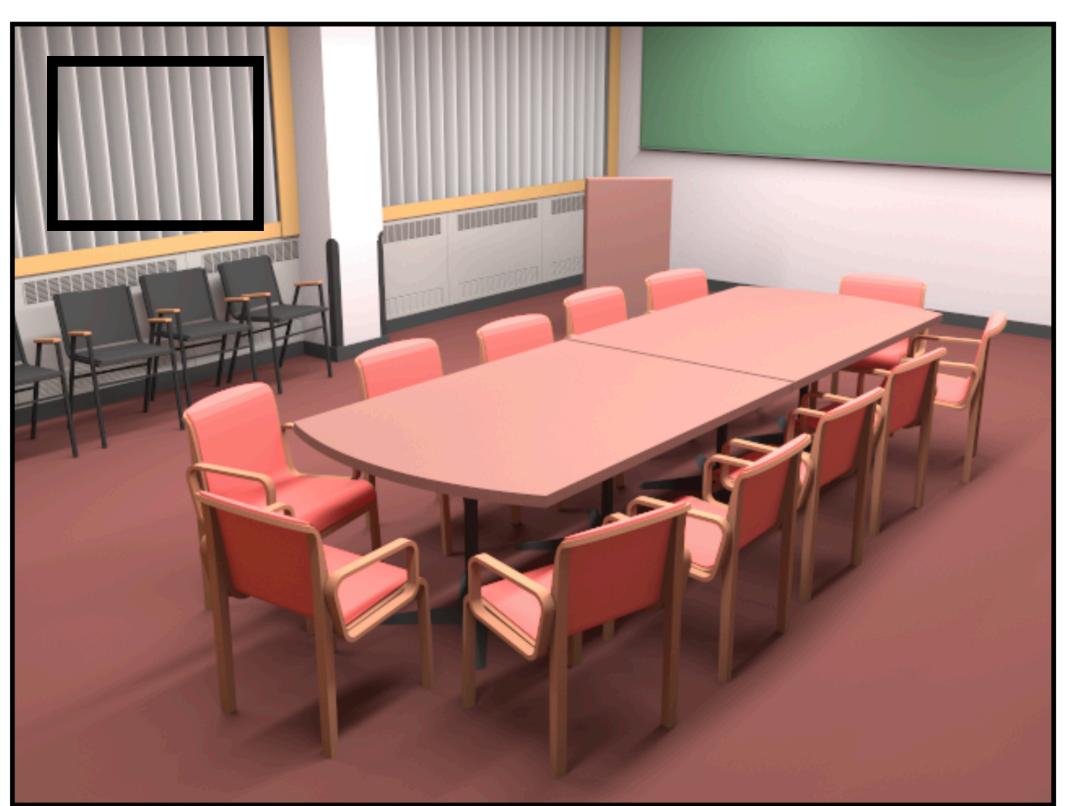
Image courtesy Kollig and Keller





Bias Compensation [Kollig and Keller 2004] [Raab et al. 2008]

Bounding only



Bounding and Compensation

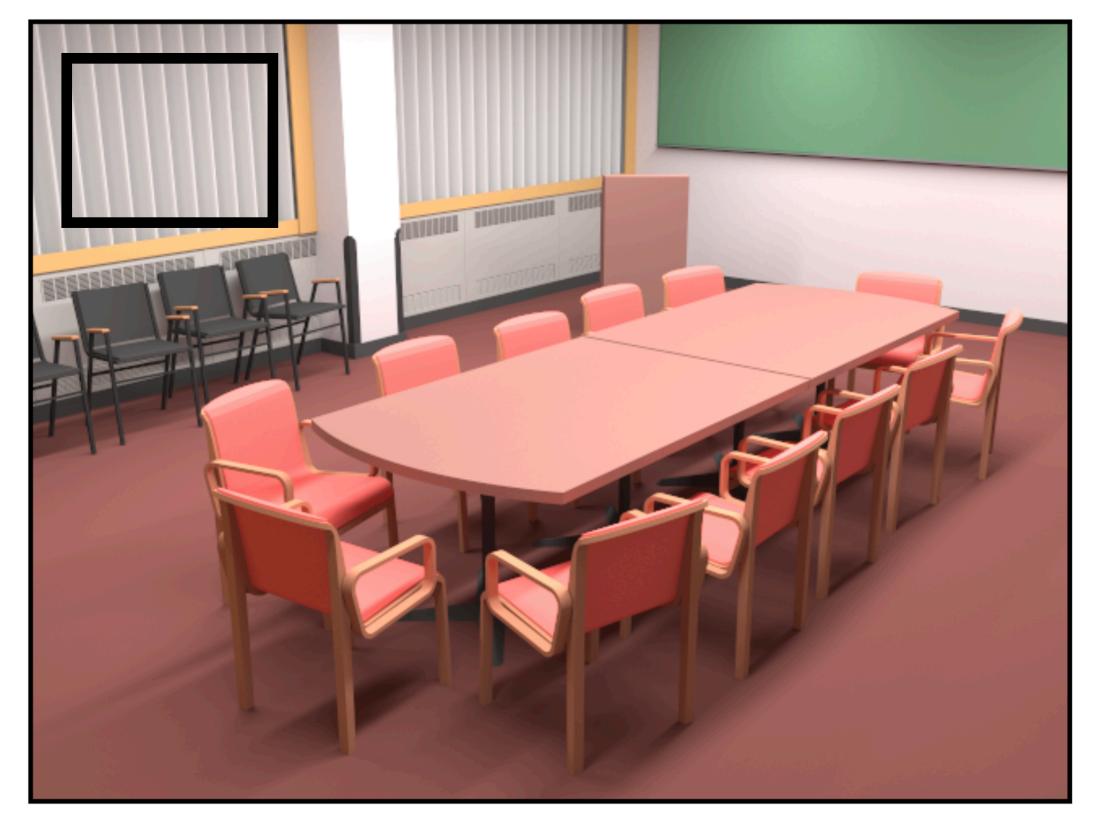


Image courtesy Kollig and Keller





Screen space bias compensation [Novak et al. 2011]

$$L = L_{\rm e} + TL$$





Screen space bias compensation [Novak et al. 2011]

$$L = L_{\rm e} + \mathbf{T}L$$

 $\approx L_{\rm e} + \mathbf{T}L_{\rm e} + \mathbf{T}\hat{L}$





Screen space bias compensation [Novak et al. 2011]

$$egin{aligned} L &= L_{
m e} + {f T}L \ &pprox L_{
m e} + {f T}L_{
m e} + {f T}\hat{L} \ &pprox L_{
m e} + {f T}L_{
m e} + {f T}_{
m b}\hat{L} + {f T}_{
m r}\hat{L} \end{aligned}$$





Screen space bias compensation [Novak et al. 2011]

$$L = L_{\rm e} + {f T}L$$
 $pprox L_{
m e} + {f T}L_{
m e} + {f T}\hat{L}$
 $pprox L_{
m e} + {f T}L_{
m e} + {f T}_{
m b}\hat{L} + {f T}_{
m r}\hat{L}$
 $pprox L_{
m e} + {f T}L_{
m e} + {f T}_{
m b}\hat{L} + {f T}_{
m r}(L - L_{
m e})$





Screen space bias compensation [Novak et al. 2011]

$$egin{aligned} L &= L_{
m e} + {f T}L \ &pprox L_{
m e} + {f T}L_{
m e} + {f T}\hat{L} \ &pprox L_{
m e} + {f T}L_{
m e} + {f T}_{
m b}\hat{L} + {f T}_{
m r}\hat{L} \ &pprox L_{
m e} + {f T}L_{
m e} + {f T}_{
m b}\hat{L} + {f T}_{
m r}(L - L_{
m e}) \ &pprox L_{
m e} + {f \sum}_{i=0}^{\infty} {f T}_{
m r}^i ({f T}L_{
m e} + {f T}_{
m b}\hat{L}) \ & ext{iteratively} & ext{compute once} \ & ext{apply $f T_{
m r}$} & ext{and store} \end{aligned}$$





Screen space bias compensation [Novak et al. 2011]

Rendering Equation:

$$pprox L_{\mathrm{e}} + \sum_{i=0}^{\infty} \mathbf{T}_{\mathbf{r}}^{i} (\mathbf{T}L_{\mathrm{e}} + \mathbf{T}_{\mathbf{b}}\hat{L})$$

direct + bounded indirect illumination

residual transport in screen-space

residual transport in screen-space



$$\mathbf{T}_{\mathbf{r}}^{0}(\mathbf{T}L_{\mathbf{e}}+\mathbf{T}_{\mathbf{b}}\mathbf{\hat{L}})$$





Screen space bias compensation [Novak et al. 2011]

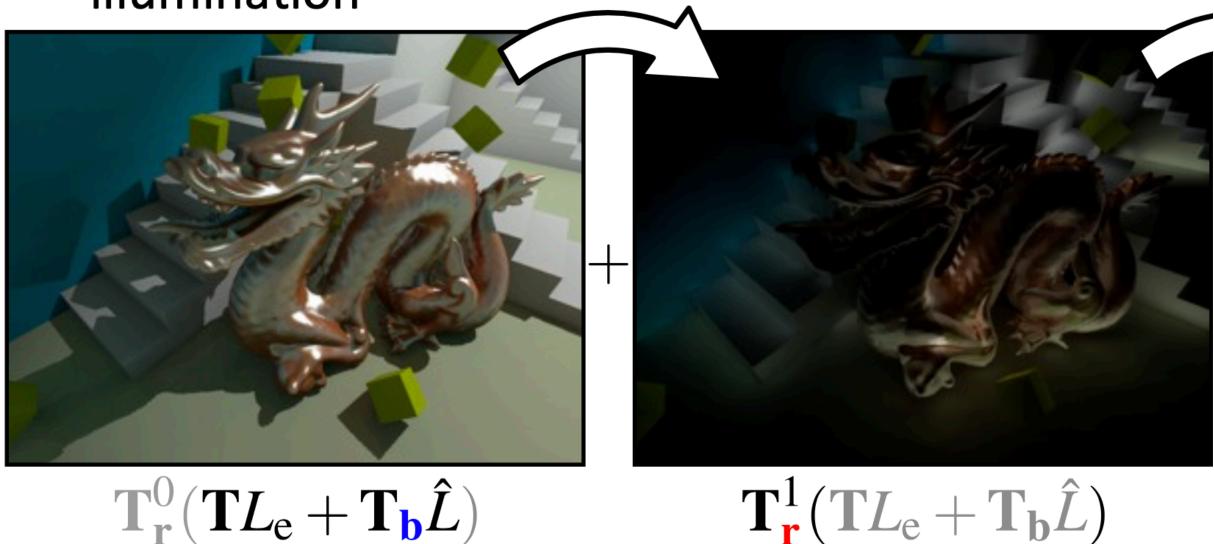
Rendering Equation:

$$pprox L_{
m e} + \sum_{i=0}^{\infty} {f T}^i_{f r} ({f T} L_{
m e} + {f T}_{f b} \hat{L})$$

direct + bounded indirect illumination

residual transport in screen-space

residual transport in screen-space







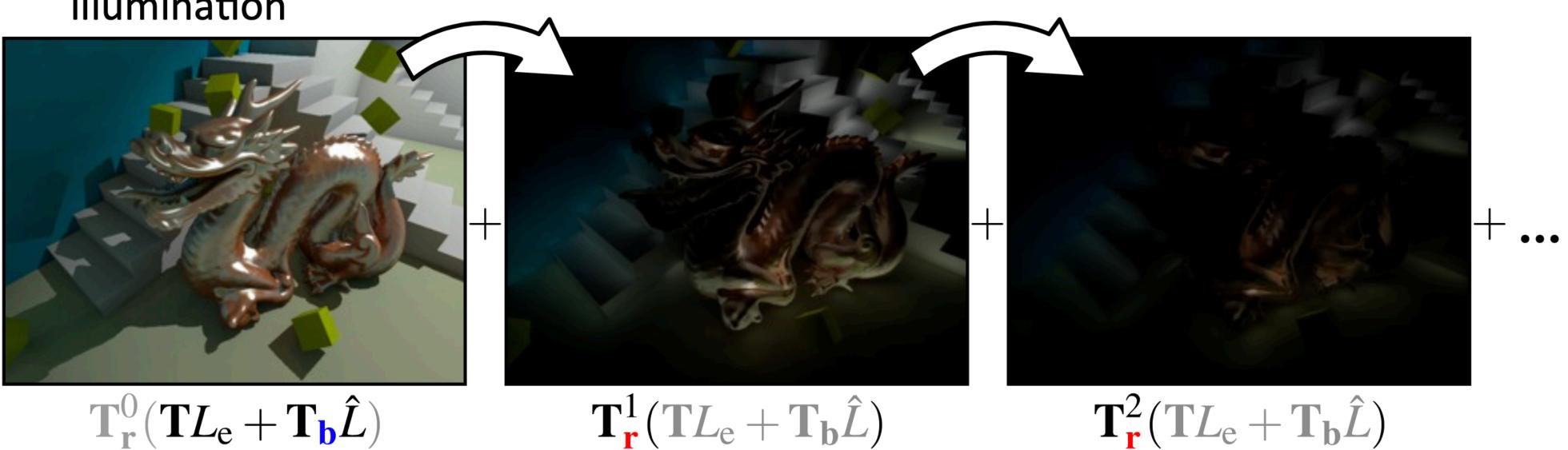
Screen space bias compensation [Novak et al. 2011]

Rendering Equation: $\approx L_{\rm e} + \sum_{i=0}^{\infty} {\bf T}_{\bf r}^i ({\bf T} L_{\rm e} + {\bf T}_{\rm b} \hat{L})$

direct + bounded indirect illumination

residual transport in screen-space

residual transport in screen-space







Screen space bias compensation [Novak et al. 2011]

Direct + bounded indirect

1- and 2-bounce residual





Screen space bias compensation [Novak et al. 2011]

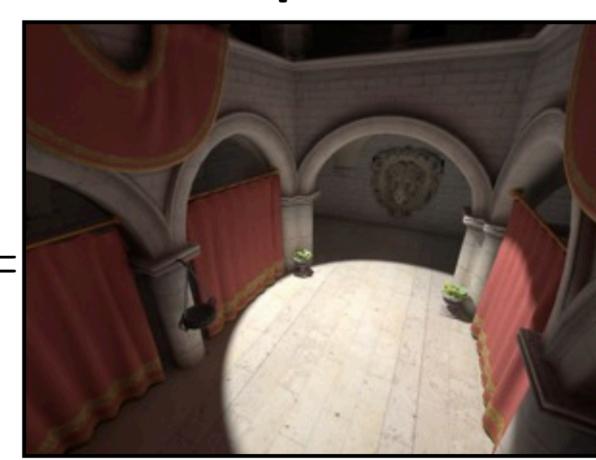
Direct + bounded indirect

1- and 2-bounce residual

Composited









Screen space bias compensation [Novak et al. 2011]

Direct + bounded indirect

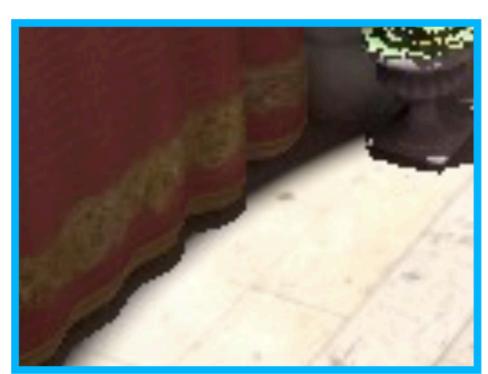
1- and 2-bounce residual

Composited







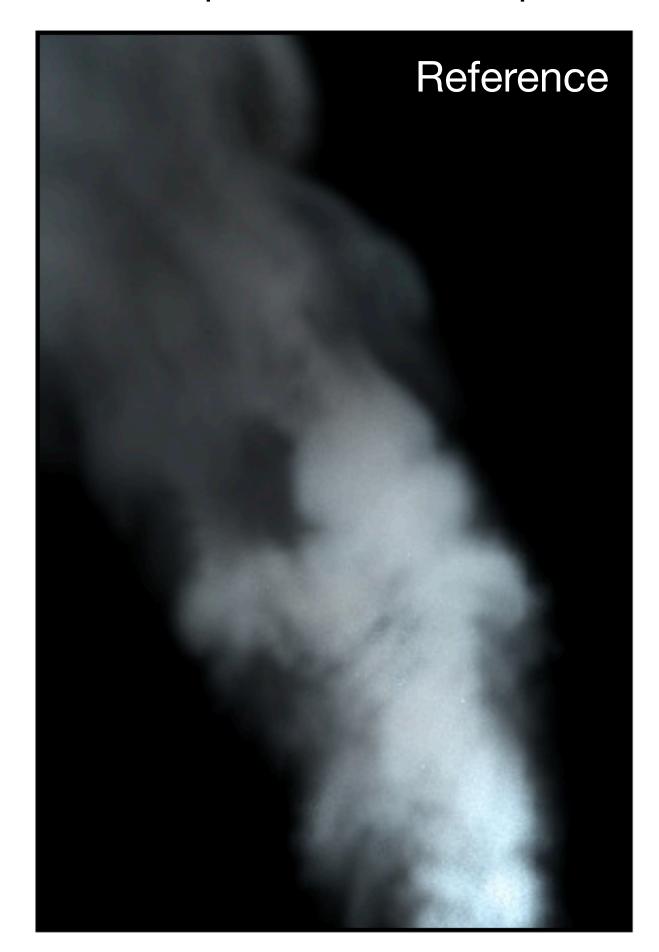








Approximate bias compensation [Engelhardt et al. 2012]



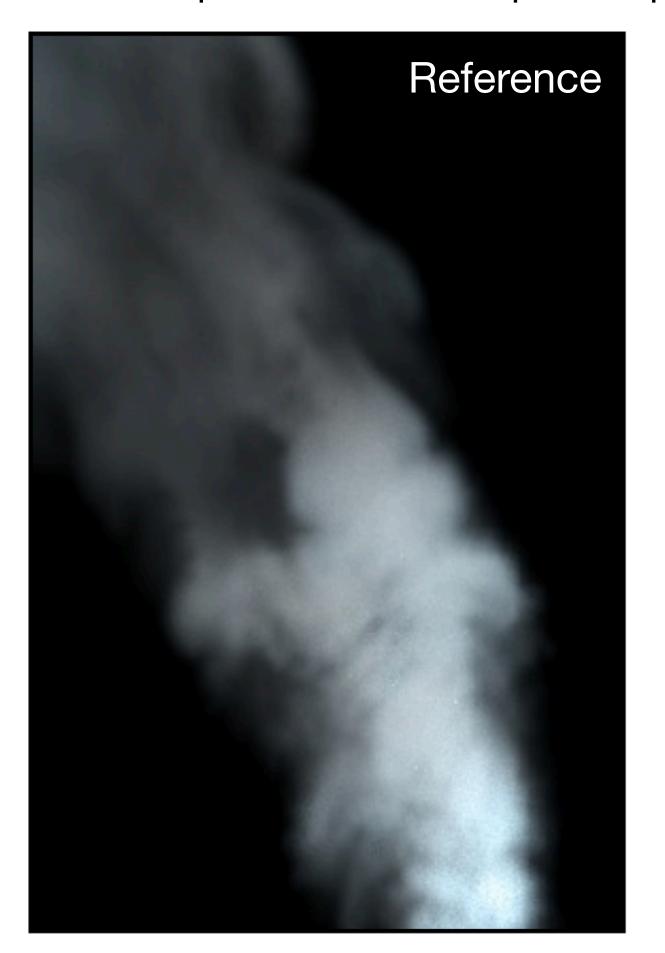


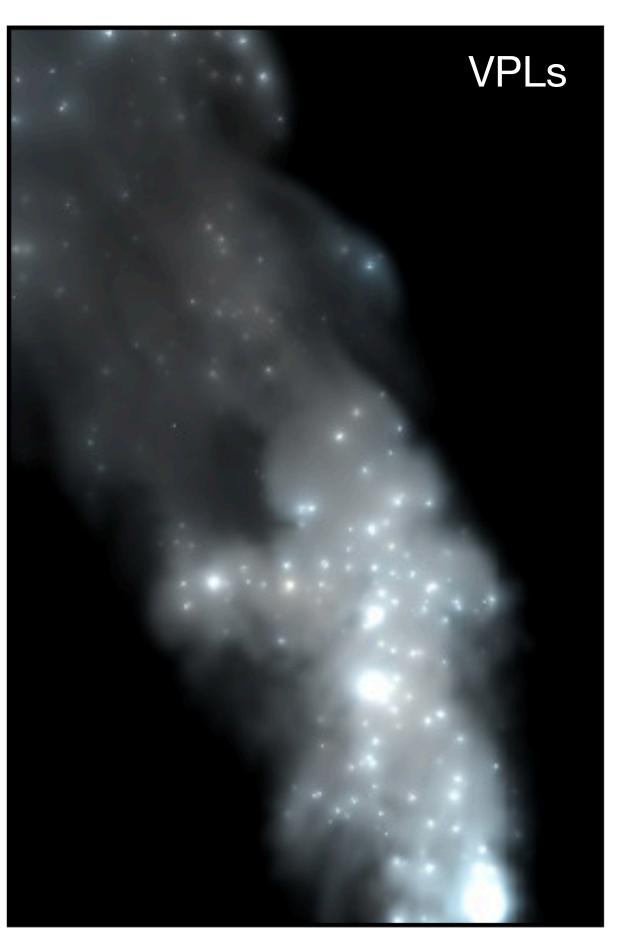


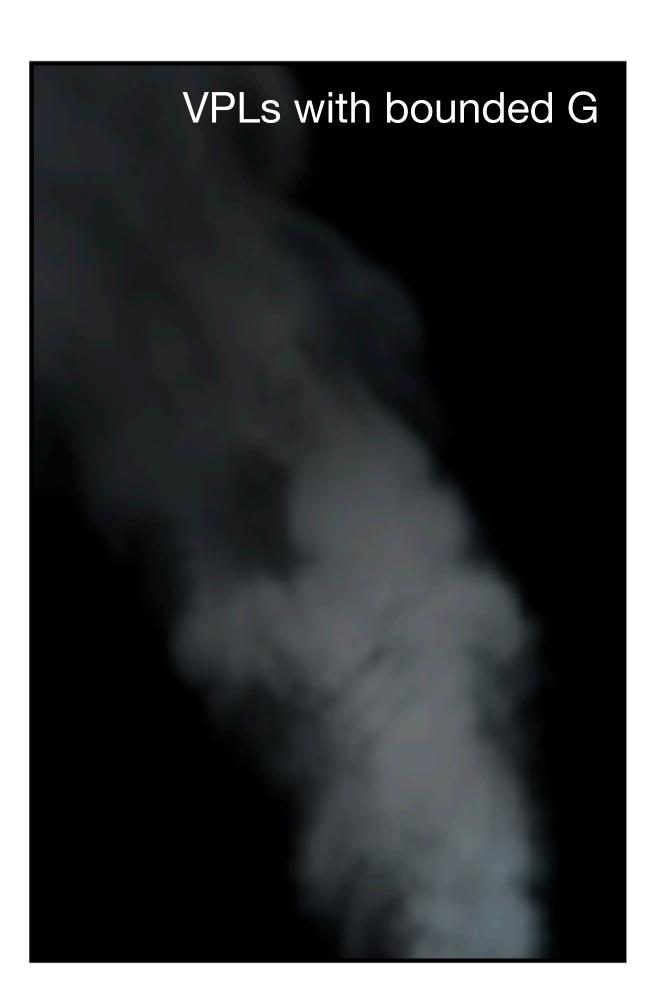
Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]

- efficient compensation for participating media



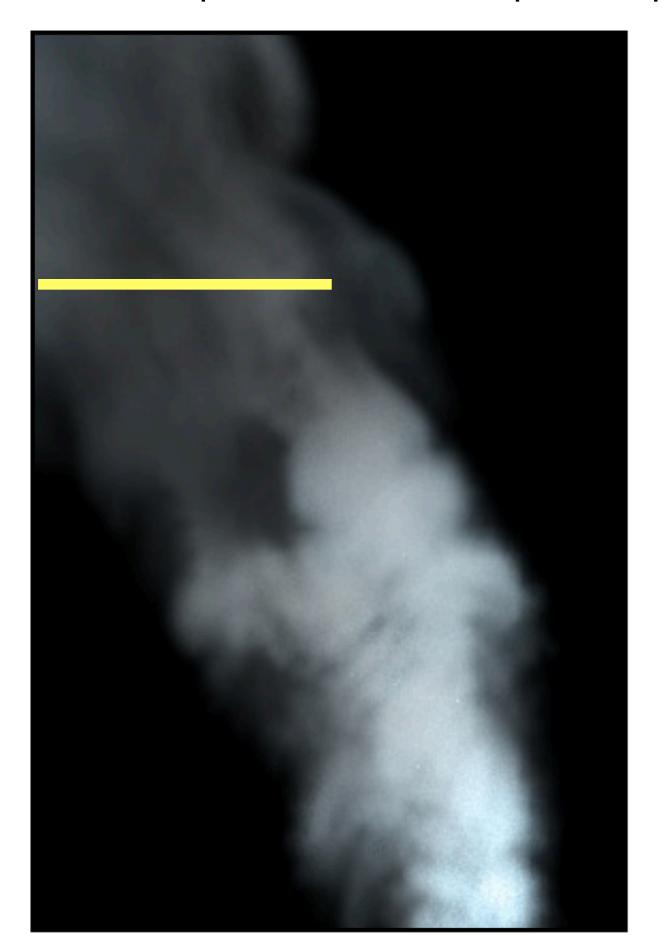


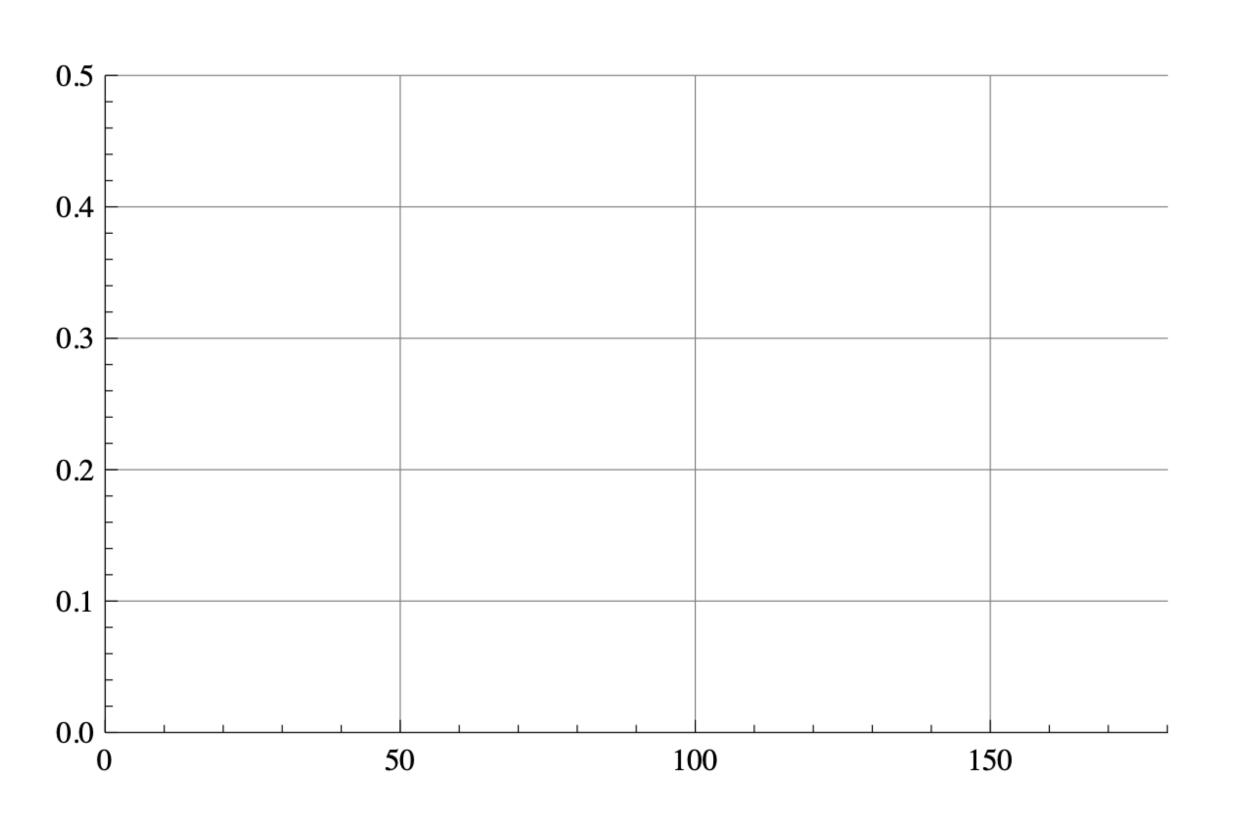




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]



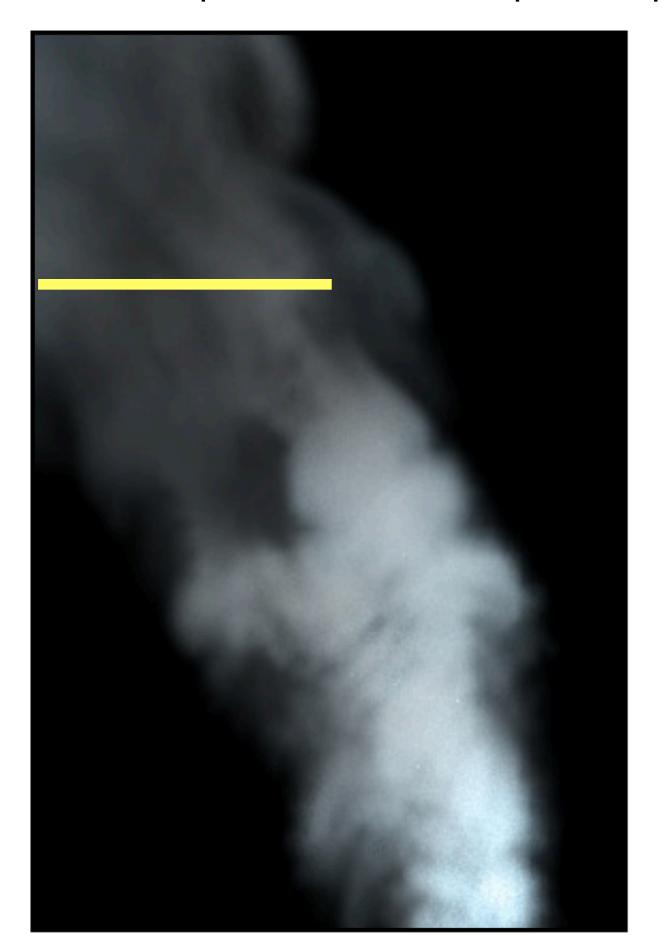


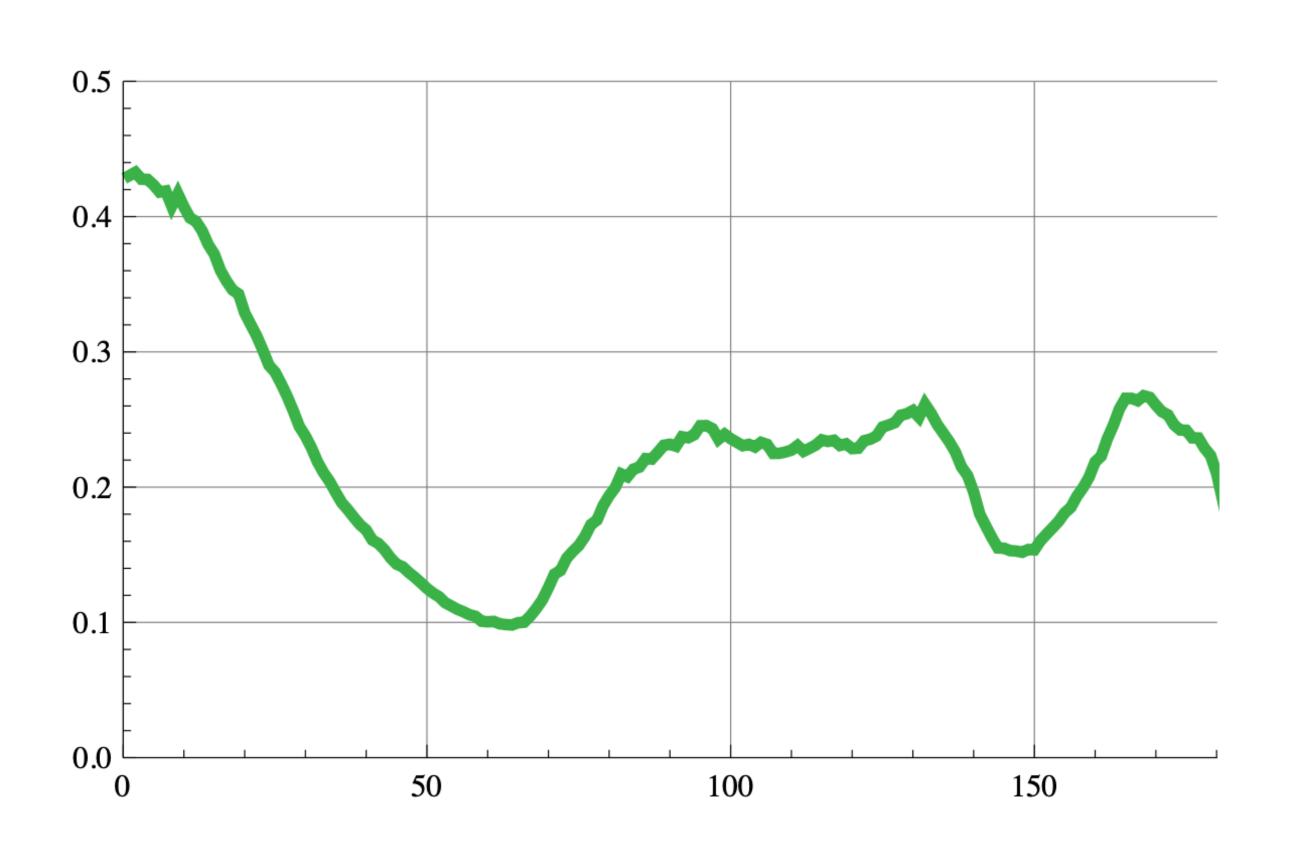




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]



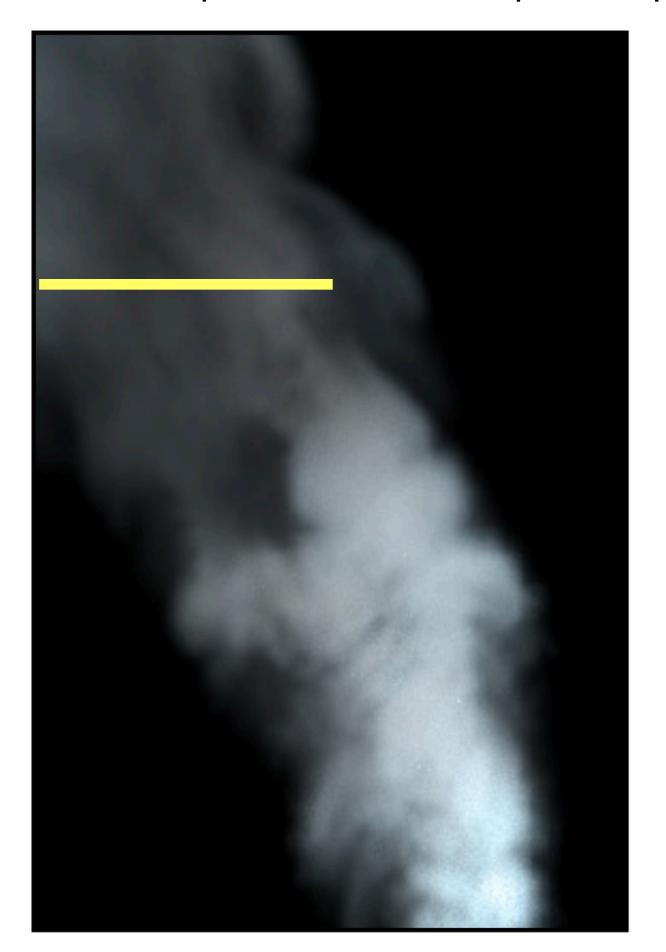


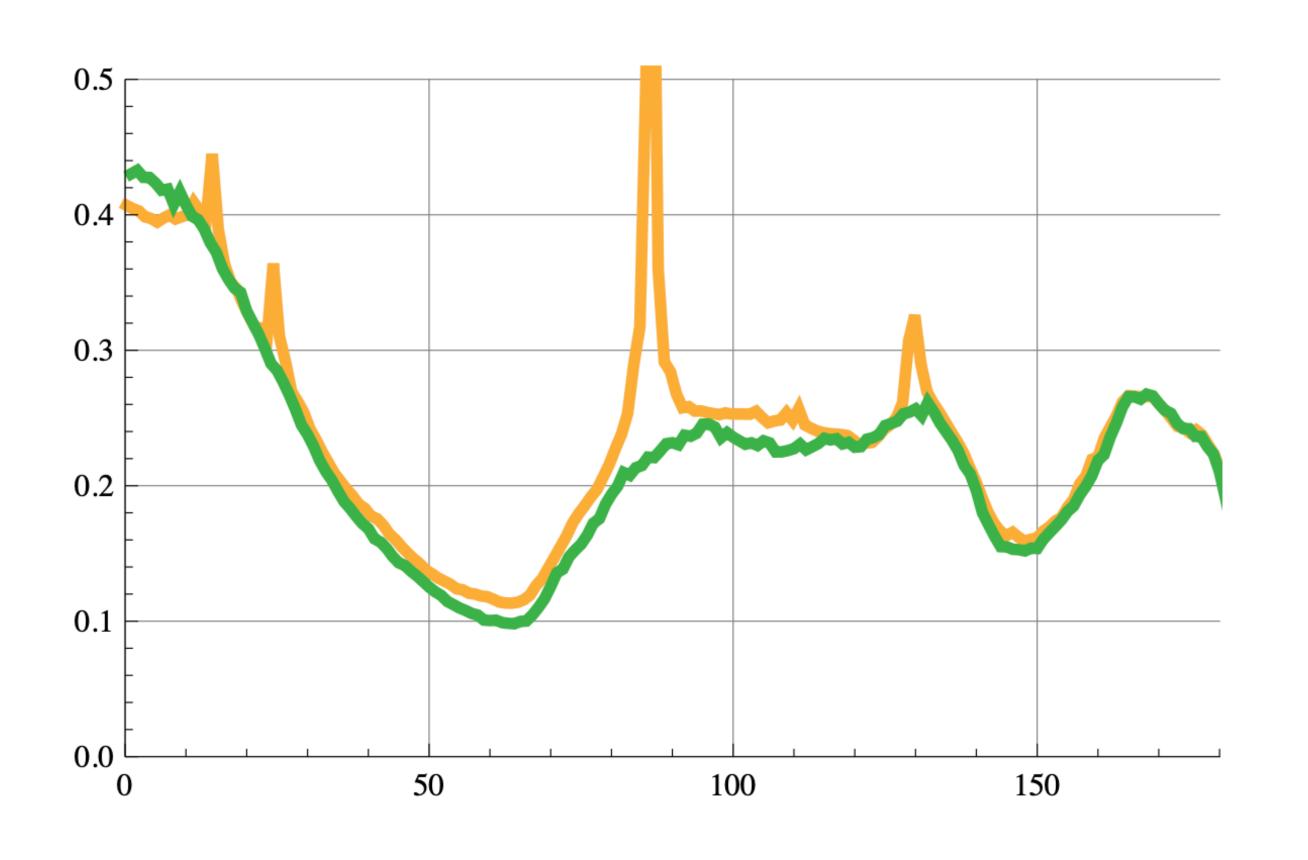




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]



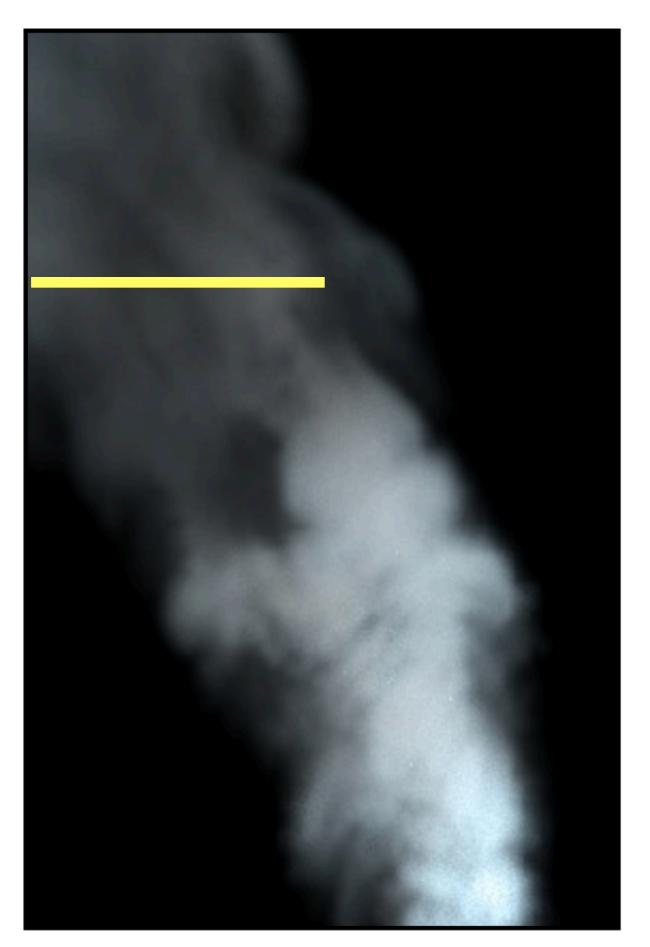


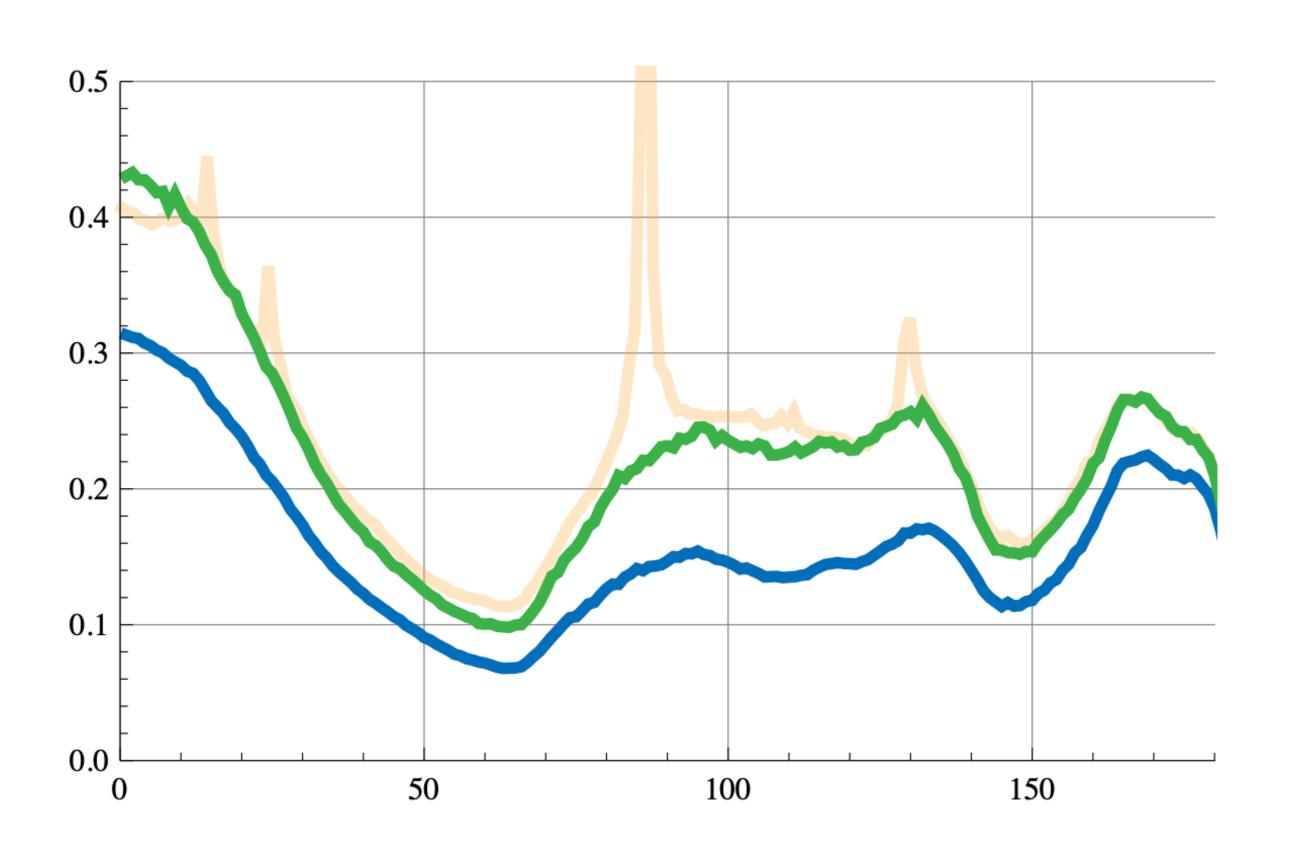




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]



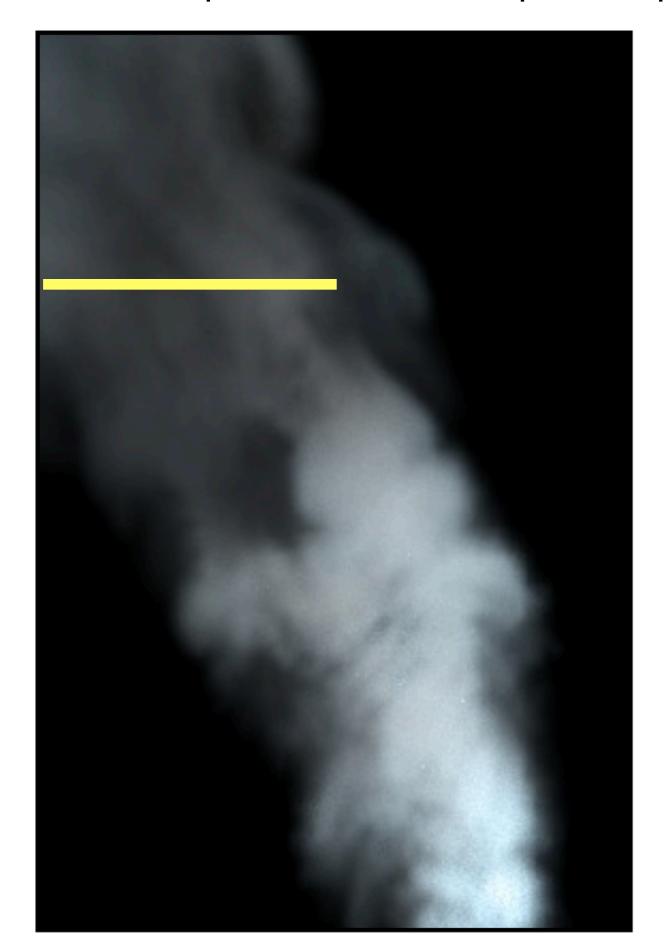


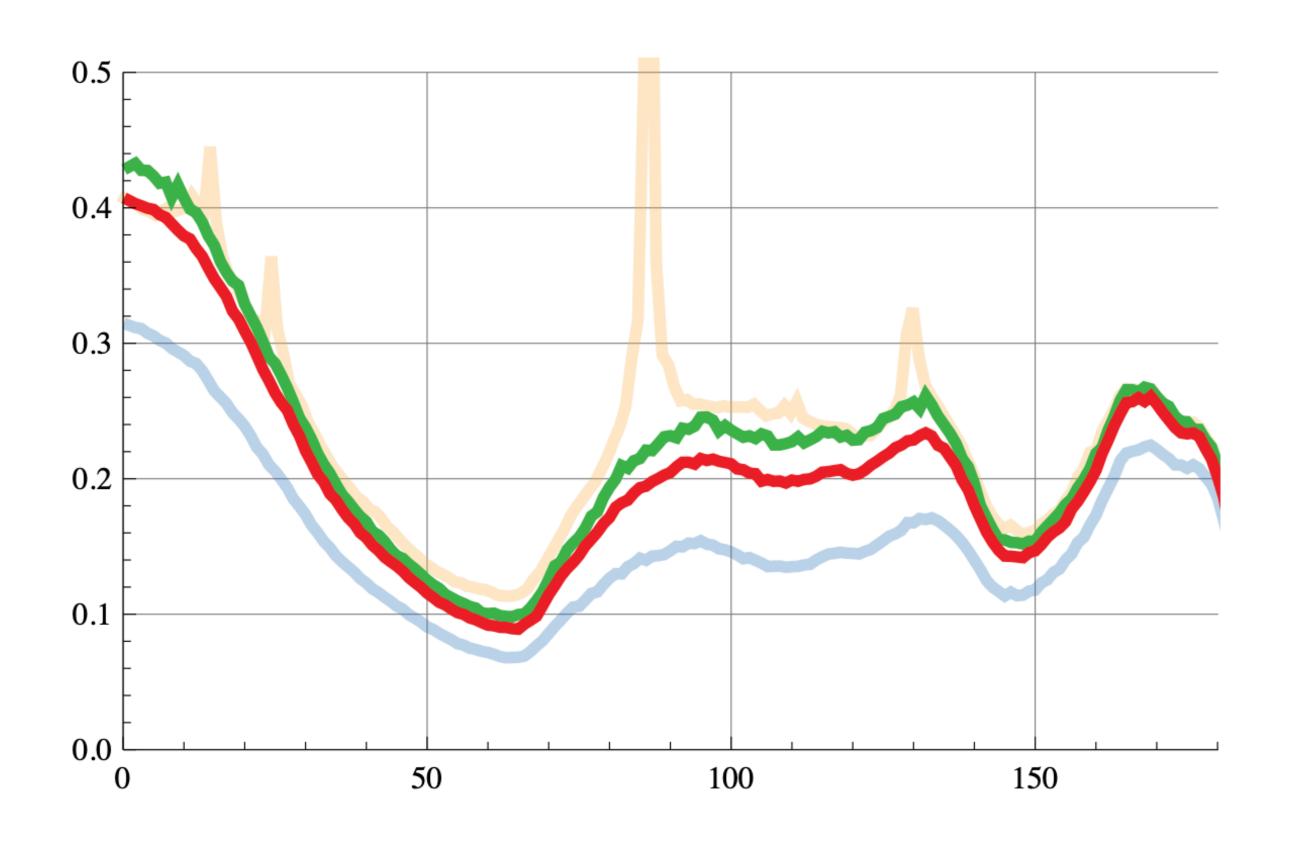




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]



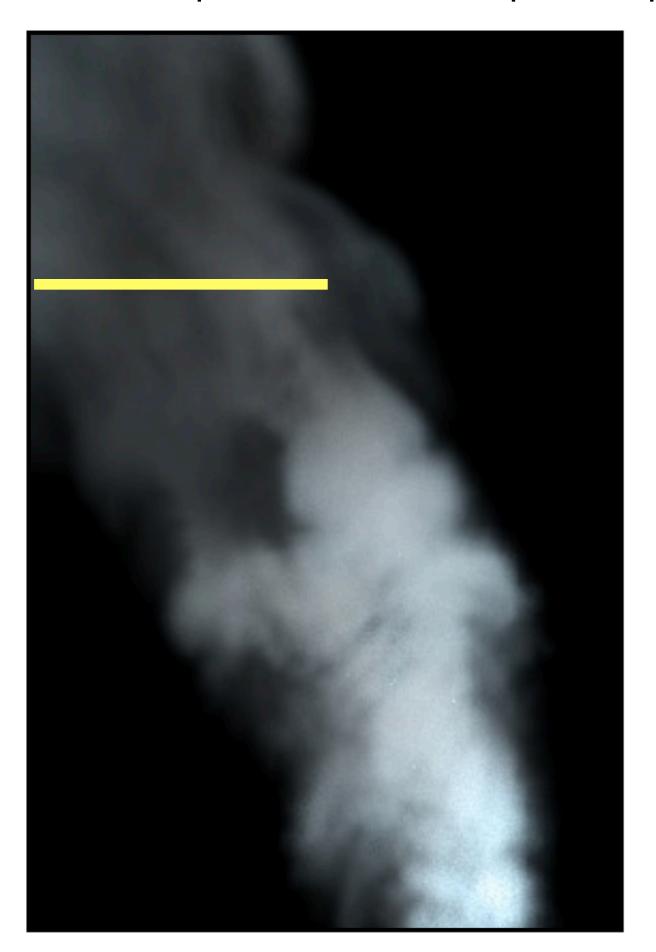


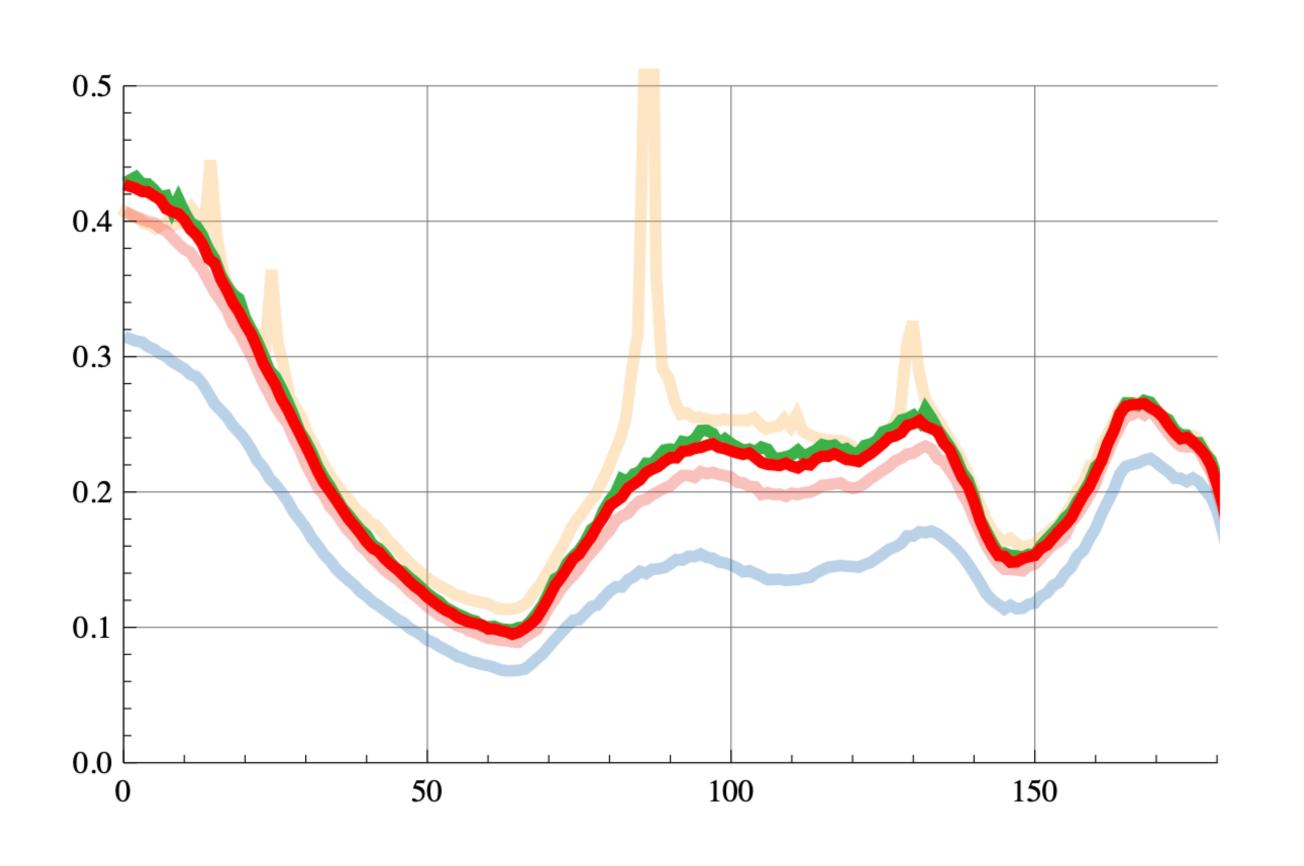




Bounding and Compensation

Approximate bias compensation [Engelhardt et al. 2012]









Approximate bias compensation [Engelhardt et al. 2012]

- efficient compensation for participating media

Optimizations used for Bias Compensation:

- Assumes locally homogeneous media
- Omit testing local visibility

Advantages:

Fast, GPU friendly

Disadvantages:

Approximate, complicated





Approximate bias compensation [Engelhardt et al. 2012]



Bounded: 39 mins

Approximate bias compensation: 13 mins





Lighting with VPLs

How to avoid Splotches?

Solutions:

- 1) Bound the geometry
 - remove energy, darkens the image
 - to get unbiased results, we need to compensate for the bounding
- 2) Distribute the flux of a VPL over area (volume)
 - redistributes energy, blurs the illumination
 - to get consistent results, progressively reduce the blurring





Lighting with VPLs

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Spreading the Energy





Problems with Instant Radiosity

Instant Radiosity does not handle glossy surfaces





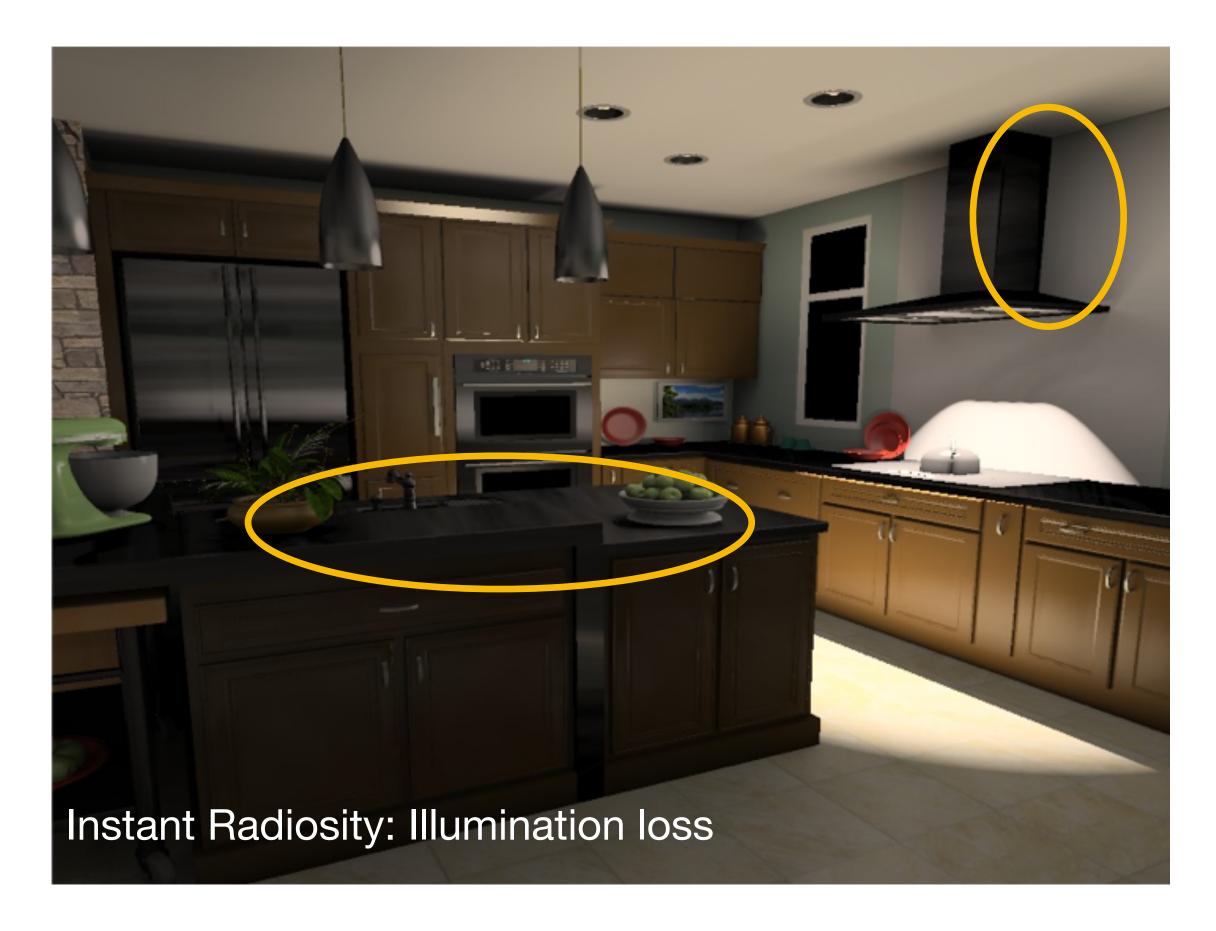




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Problems with Instant Radiosity

Instant Radiosity does not handle glossy surfaces







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Compute the missing components by path tracing [Kollig and Keller 2004]











Virtual Spherical Lights





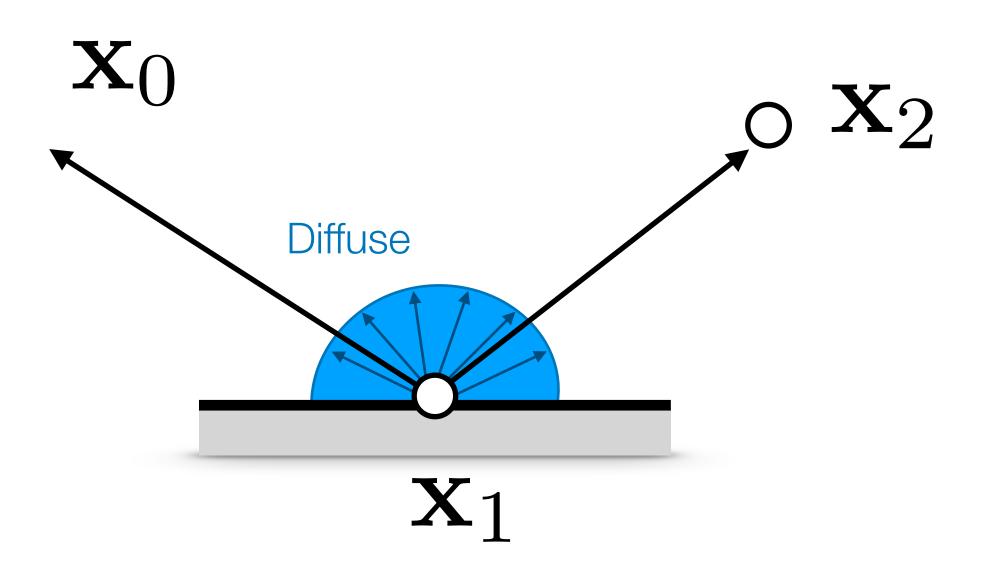




69

Instant Radiosity: Only Diffuse

Instant radiosity assumes all surfaces are diffuse



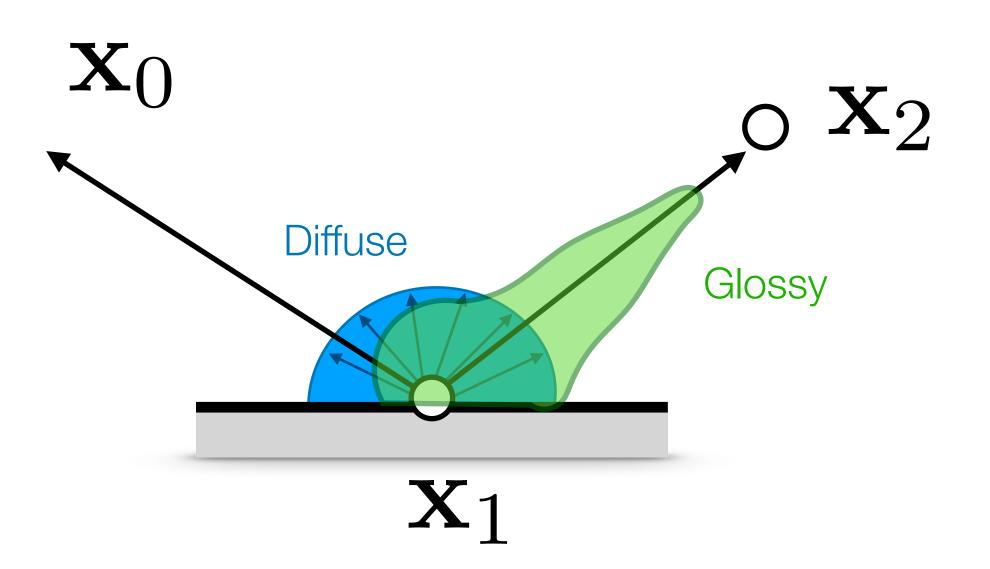




Hasan et al. 2009

Instant Radiosity: Only Diffuse

Instant radiosity assumes all surfaces are diffuse



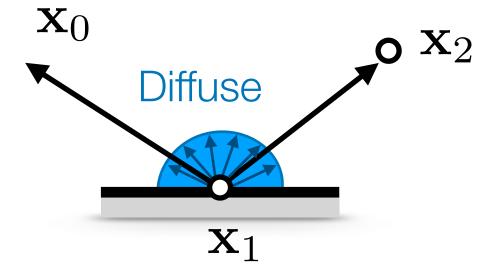




Hasan et al. 2009

Glossy VPL Emission: Illumination Spikes





Common solution:

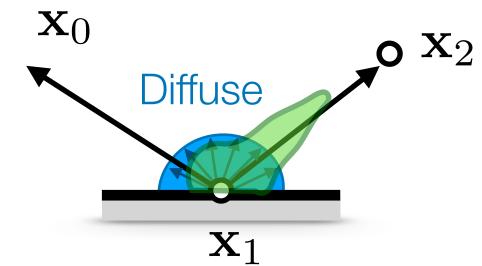
Only diffuse BRDF at light locations

Streak on the ceiling is caused by a VPL located on a highly anisotropic glossy surface.



Glossy VPL Emission: Illumination Spikes





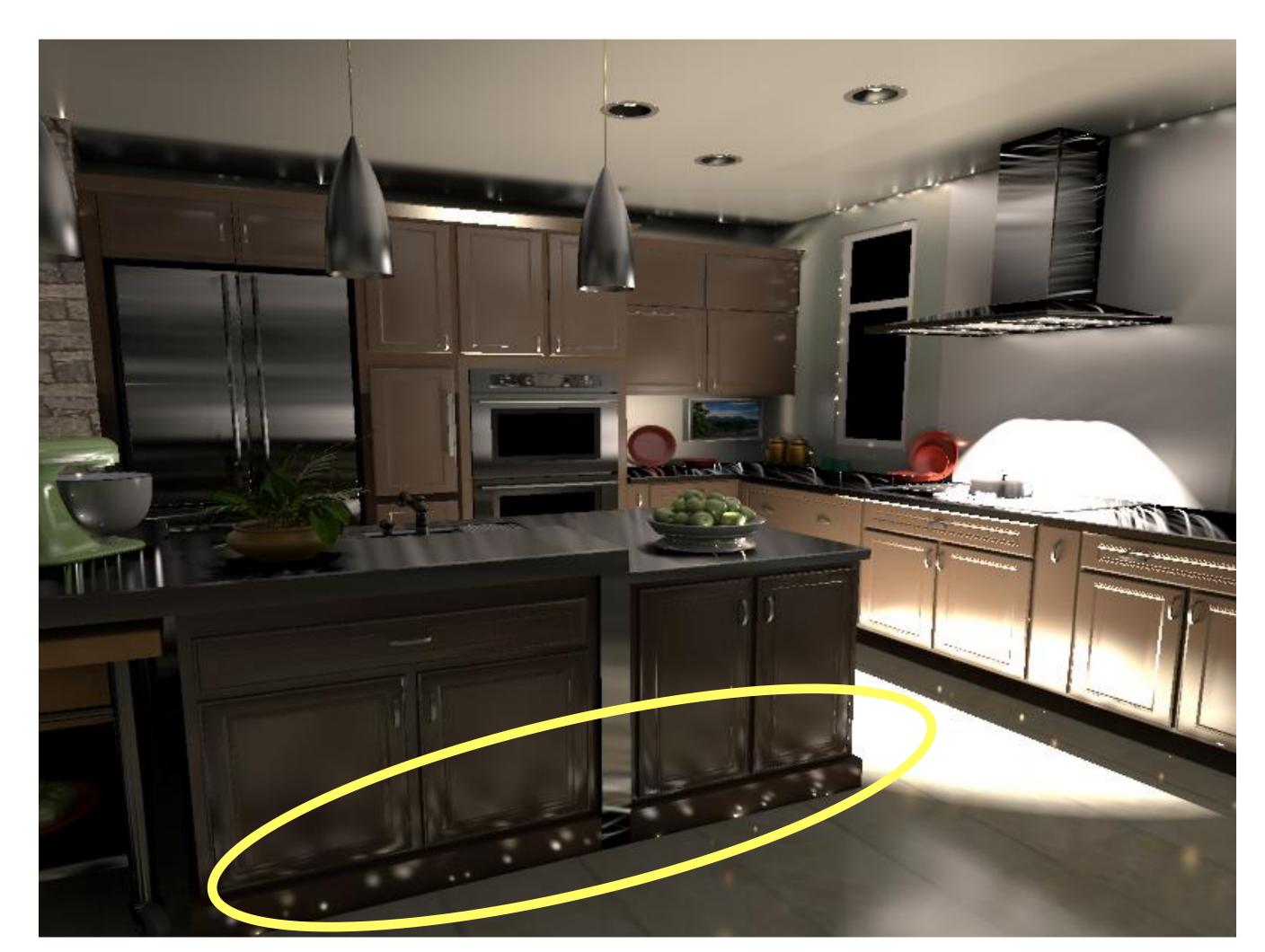
Common solution:

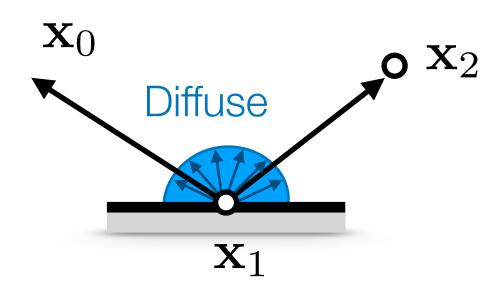
Only diffuse BRDF at light locations

Streak on the ceiling is caused by a VPL located on a highly anisotropic glossy surface.



Remaining Spikes

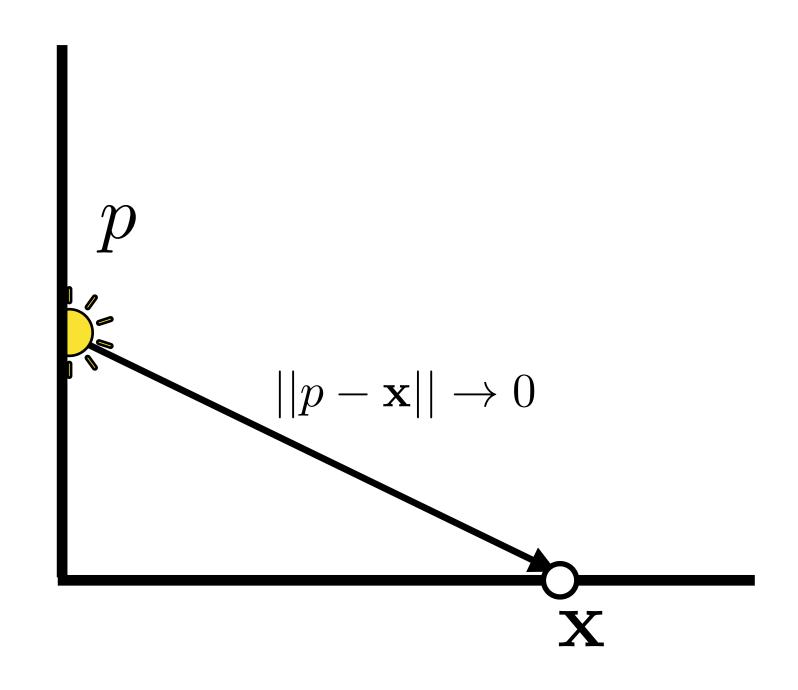




Common solution:

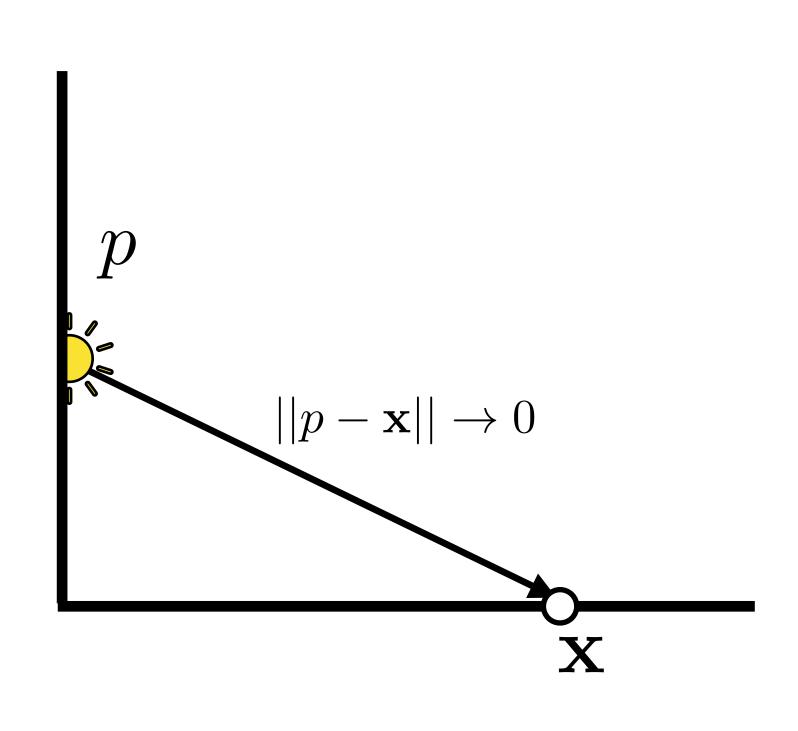
Only diffuse BRDF at light locations

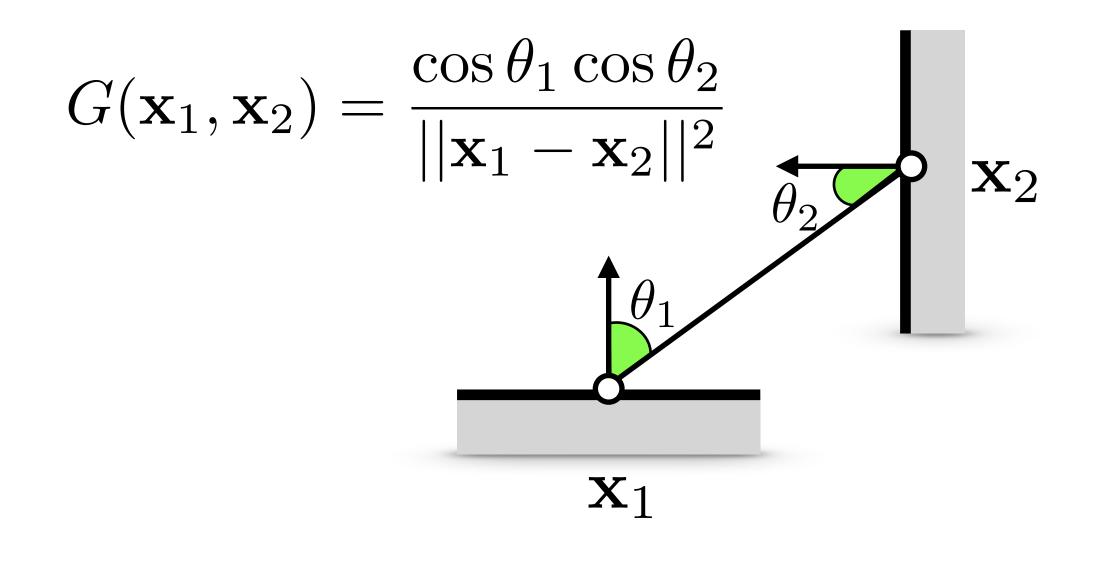






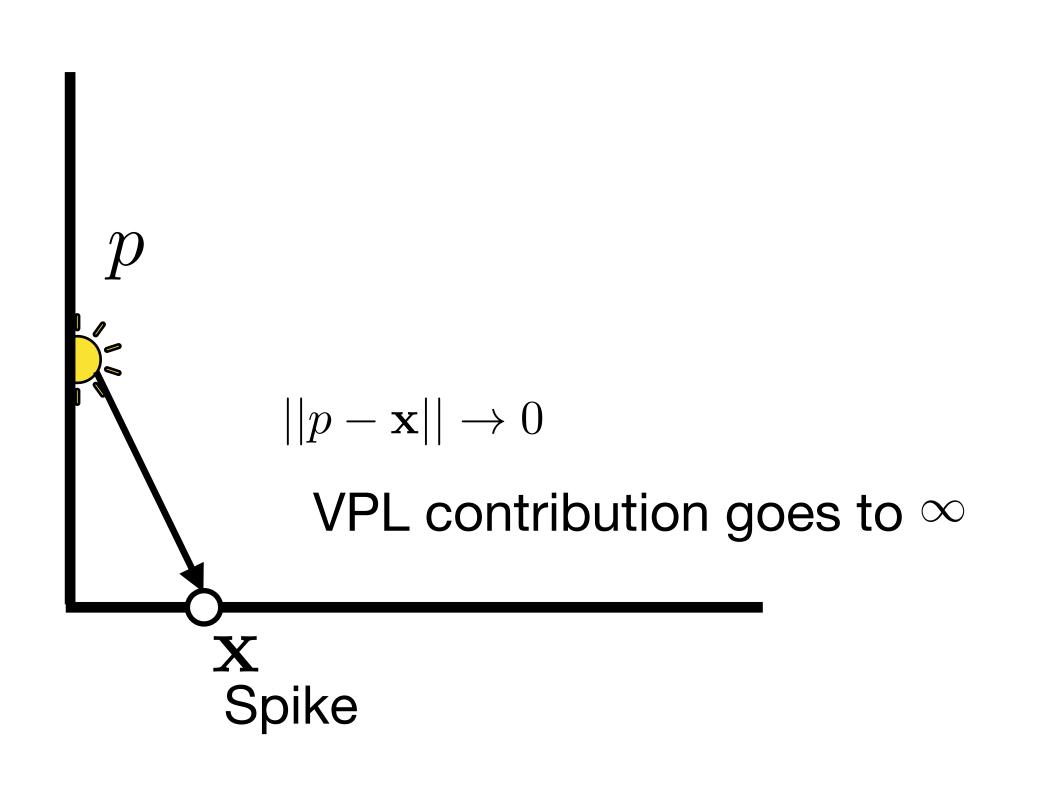


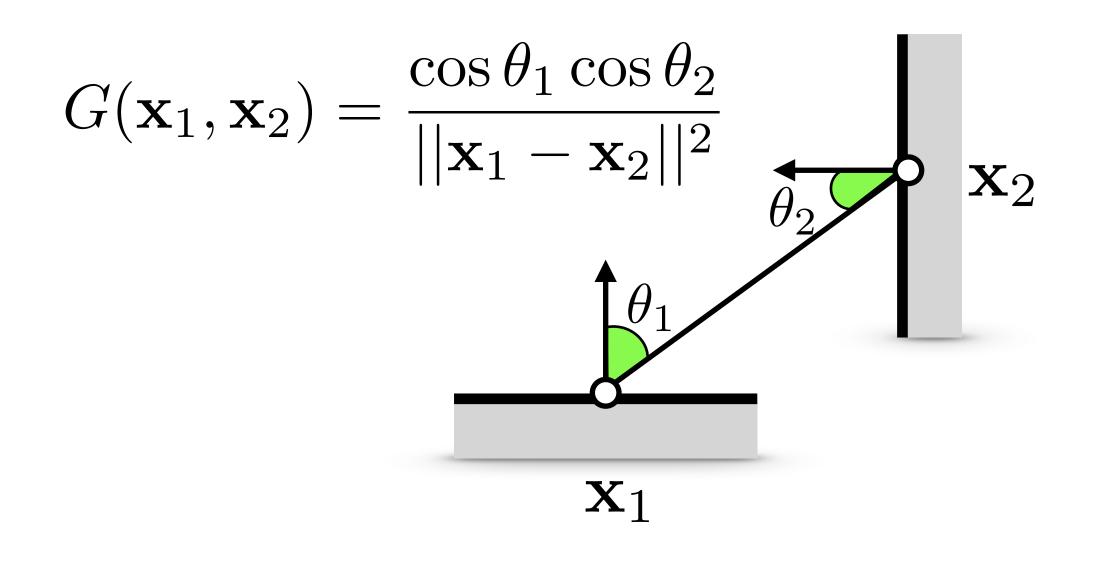






Hasan et al. 2009

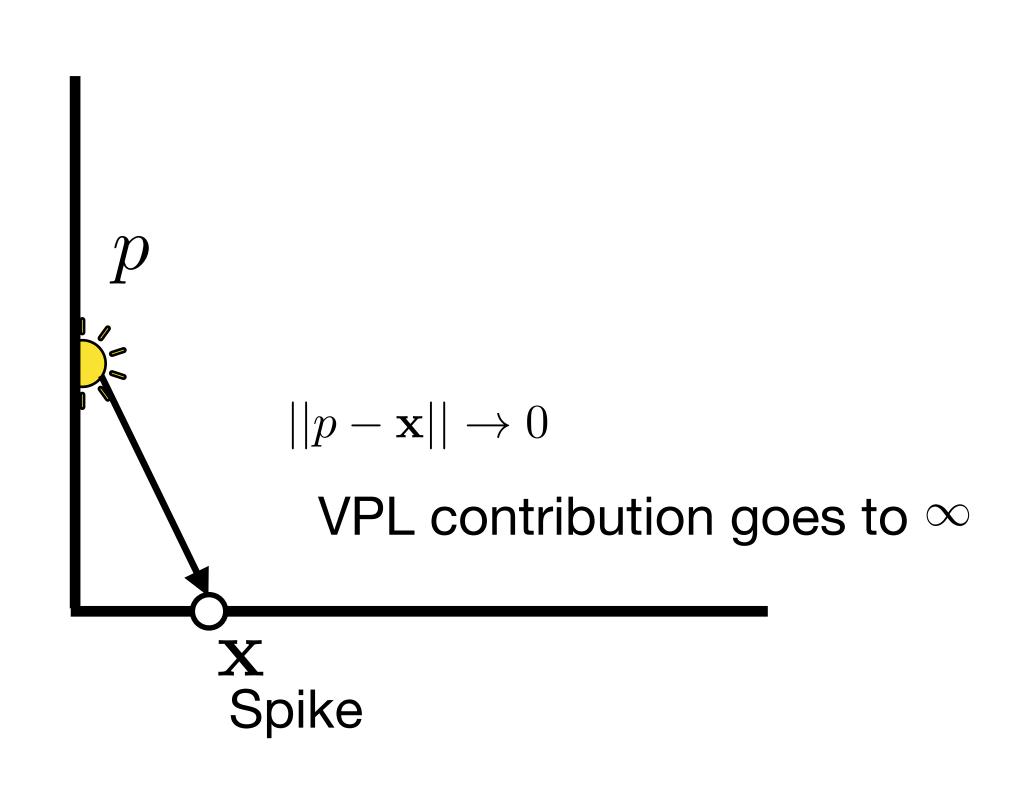


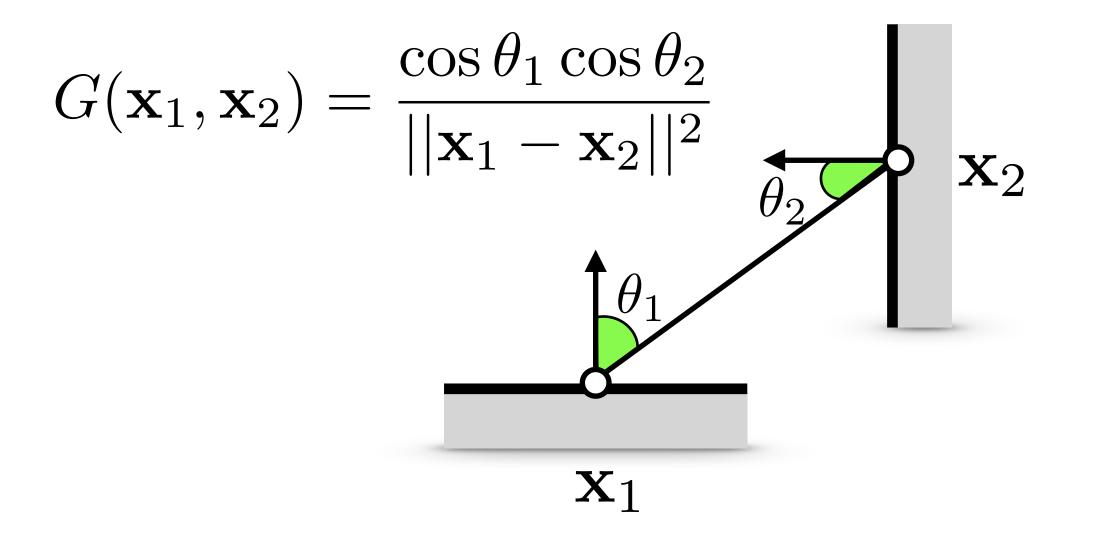




Hasan et al. 2009

Common solution: Clamp the VPL contribution







Hasan et al. 2009

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VPLs: Image splotches due to

- Spikes in the VPL emission distribution

-
$$\frac{1}{||\mathbf{x}_1 - \mathbf{x}_2||^2}$$
 term

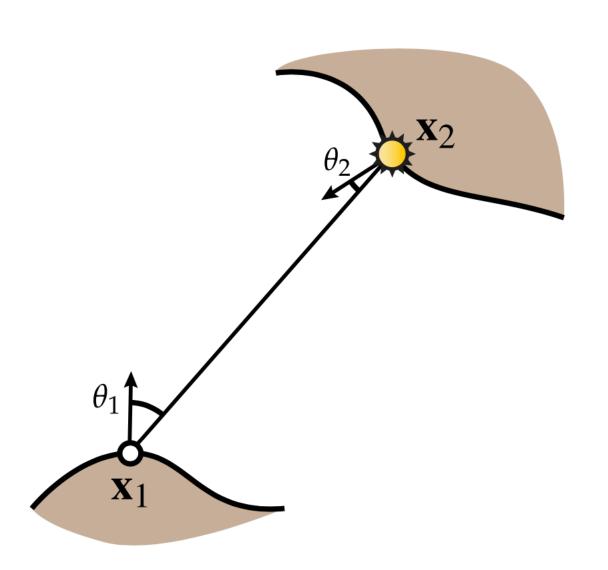
Virtual Spherical lights (VSL):

- Spread the energy of the infinitesimal VPL over a finite surface
- Computer contribution as a solid angle integral





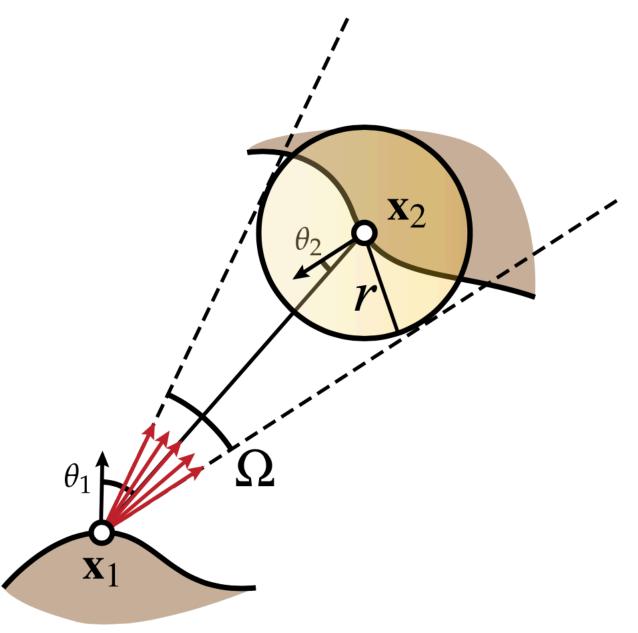
Spread the energy of the infinitesimal VPL over a finite surface







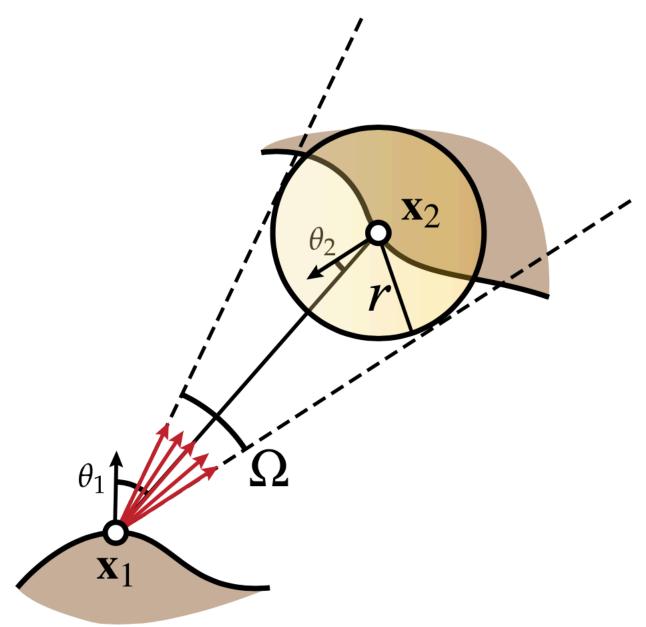
Spread the energy of the infinitesimal VPL over a finite surface







Spread the energy of the infinitesimal VPL over a finite surface



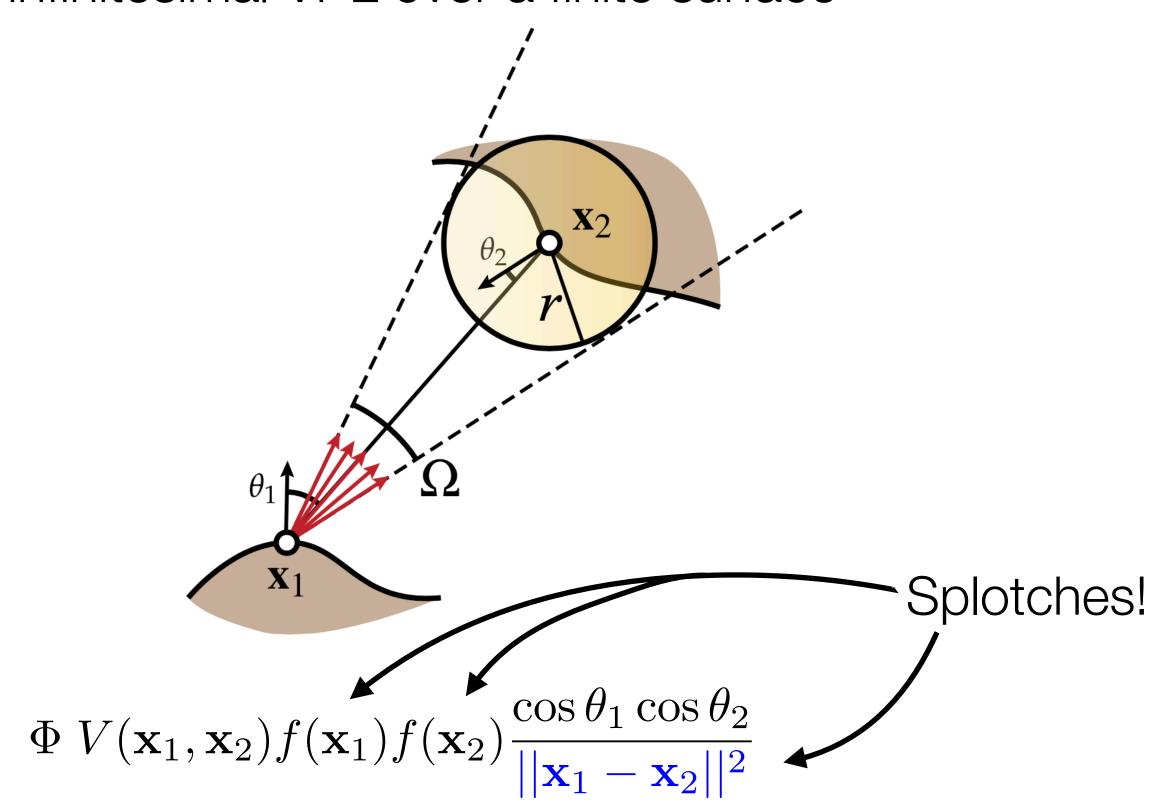
point-to-point:

$$\Phi V(\mathbf{x}_1, \mathbf{x}_2) f(\mathbf{x}_1) f(\mathbf{x}_2) \frac{\cos \theta_1 \cos \theta_2}{||\mathbf{x}_1 - \mathbf{x}_2||^2}$$





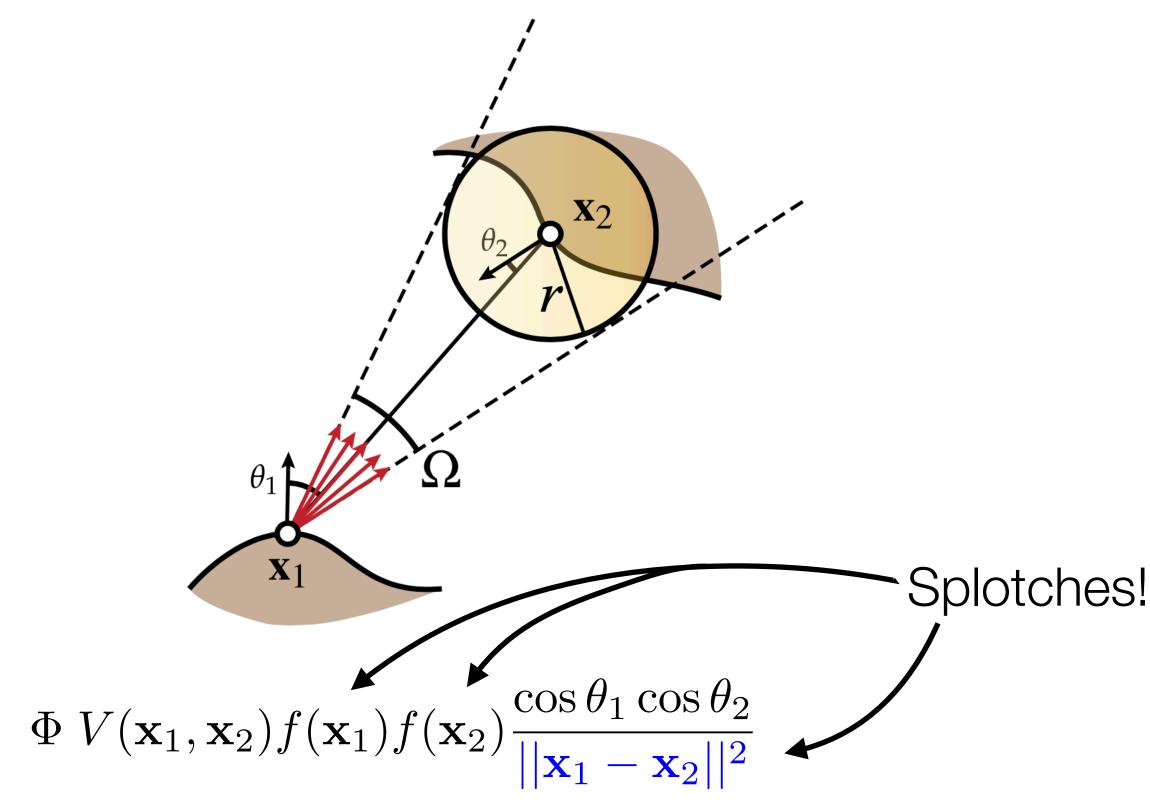
Spread the energy of the infinitesimal VPL over a finite surface



point-to-point:

UNIVERSITÄT DES SAARLANDES Hasan et al. 2009

Spread the energy of the infinitesimal VPL over a finite surface

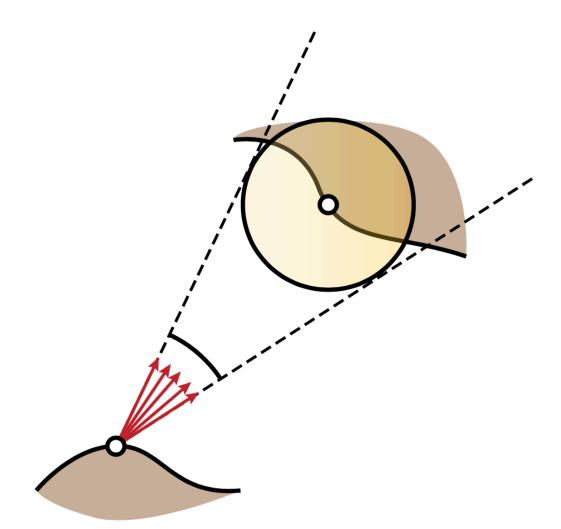


point-to-point:

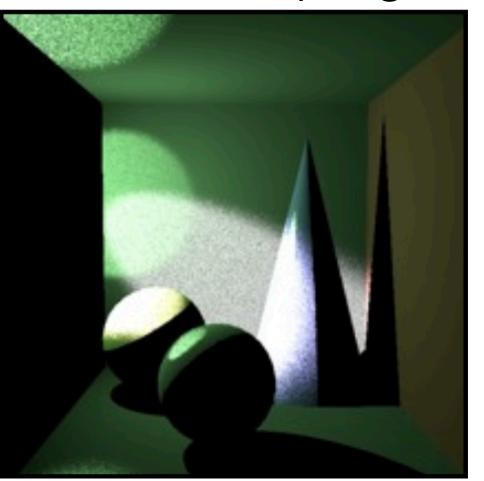
Approx. sphere-to-point:
$$\frac{\Phi}{\pi r^2} V(\mathbf{x}_1, \mathbf{x}_2) \int_{\mathcal{H}^2} f(\mathbf{x}_1) f(\mathbf{x}_2) \cos \theta_1 \cos \theta_2 d\vec{\omega}$$



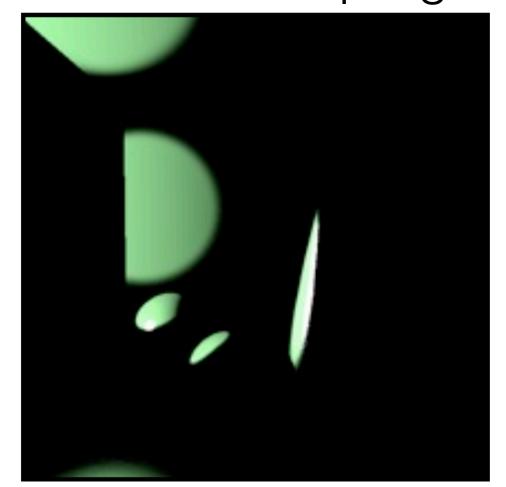


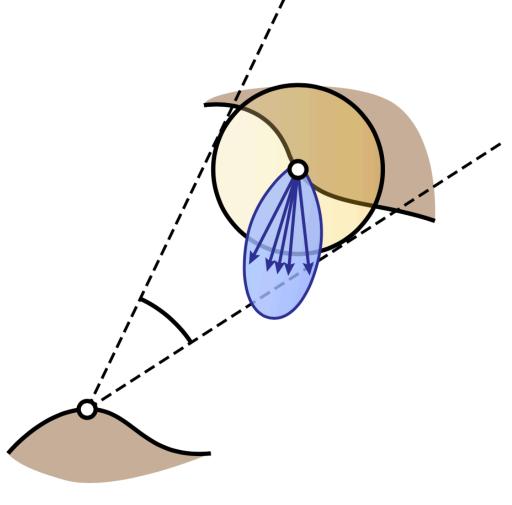


Cone sampling

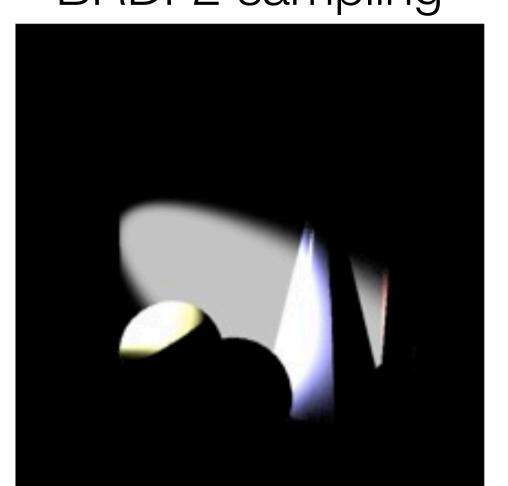


BRDF1 sampling

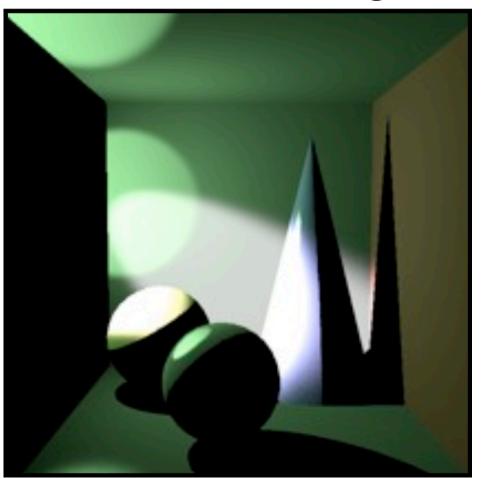




BRDF2 sampling



MIS sampling



Hasan et al. 2009





Advantages:

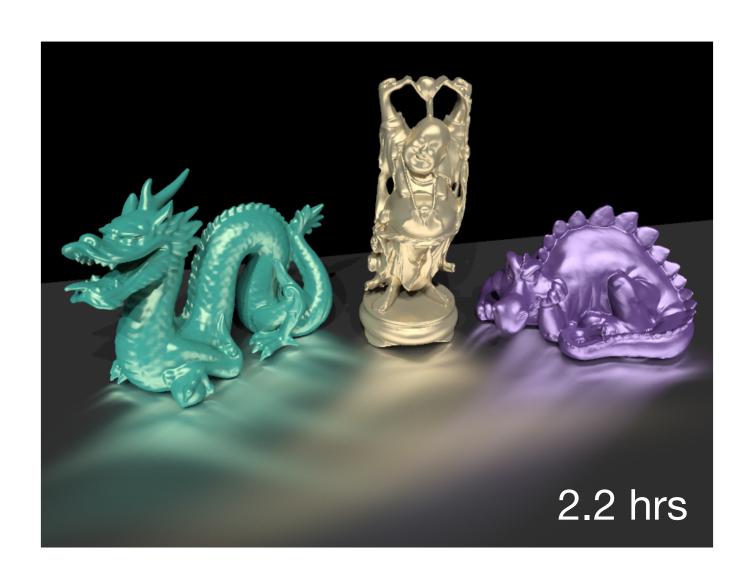
- Energy is blurred, not clamped

Disadvantages:

- Introduces bias
- Requires an extra integration over the solid angle











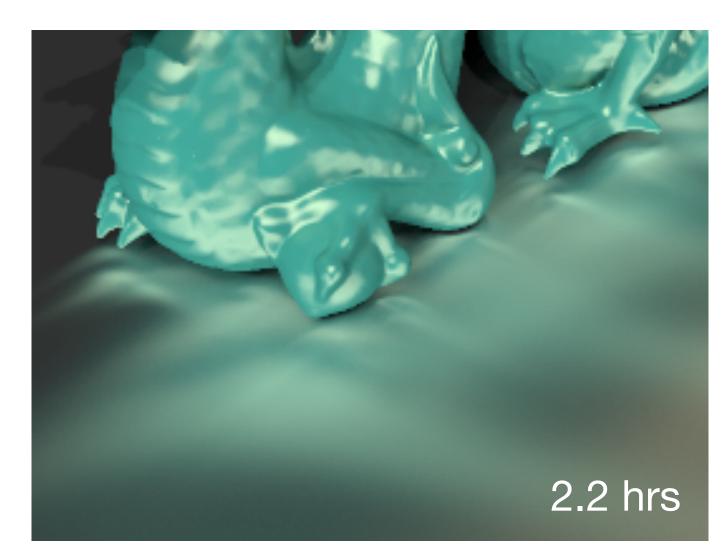
Clamped

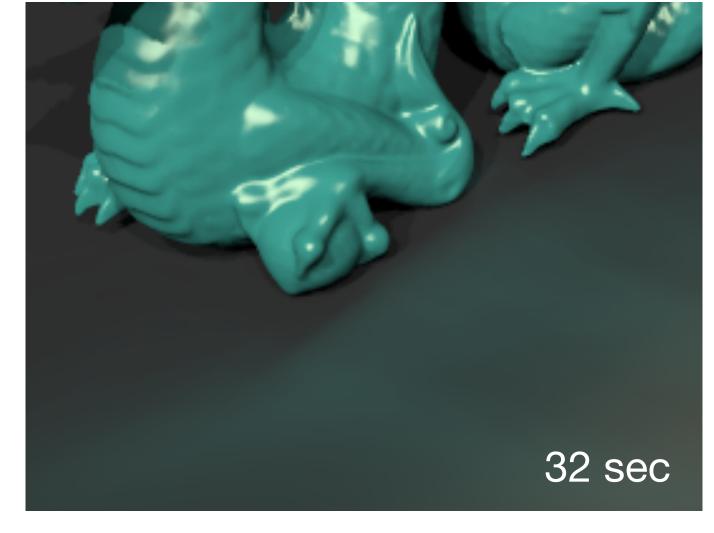


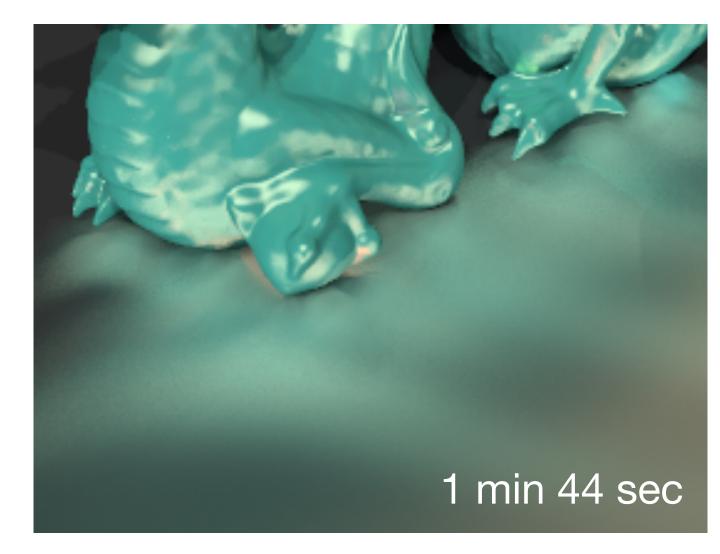
VSLs











Reference

Clamped

VSLs





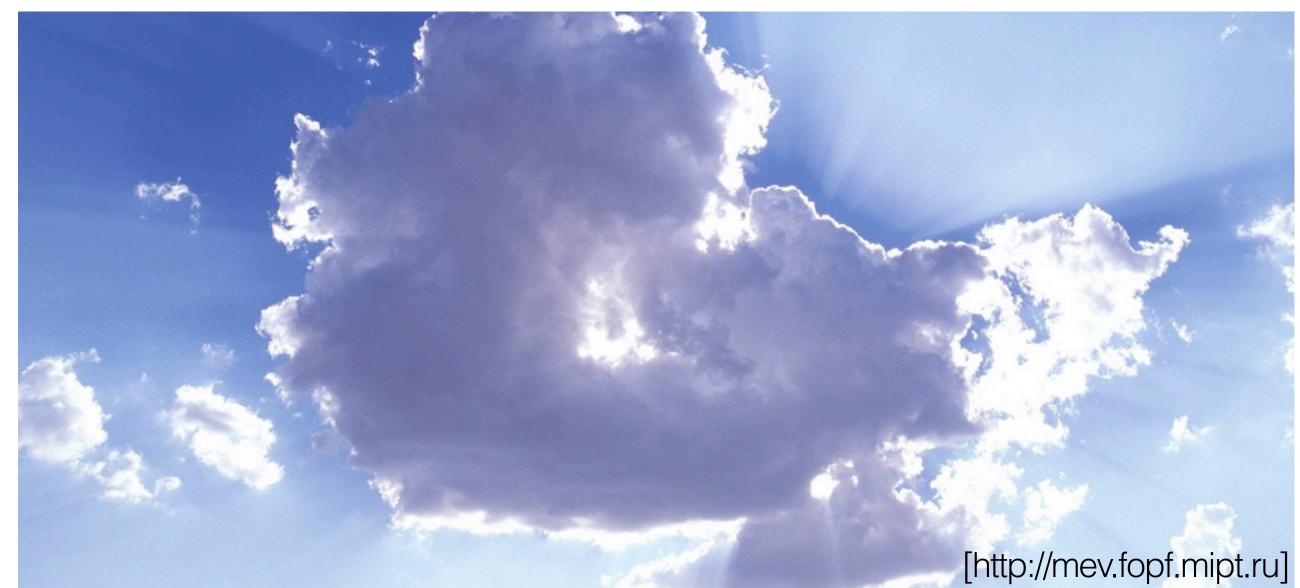
Virtual Ray Lights



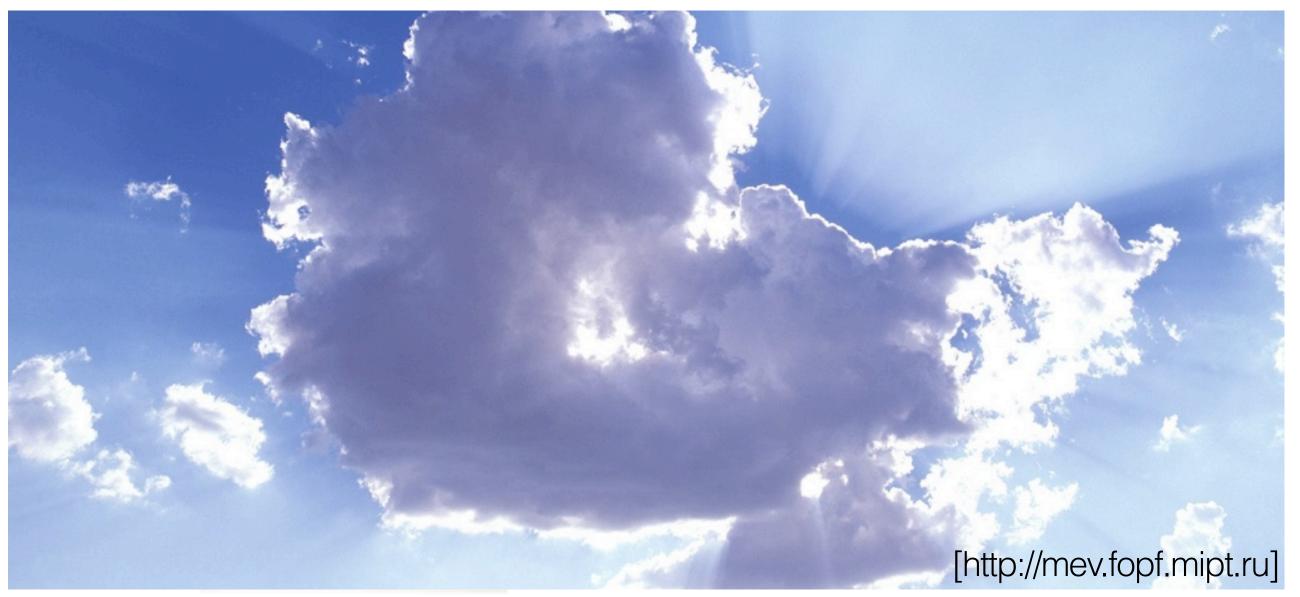


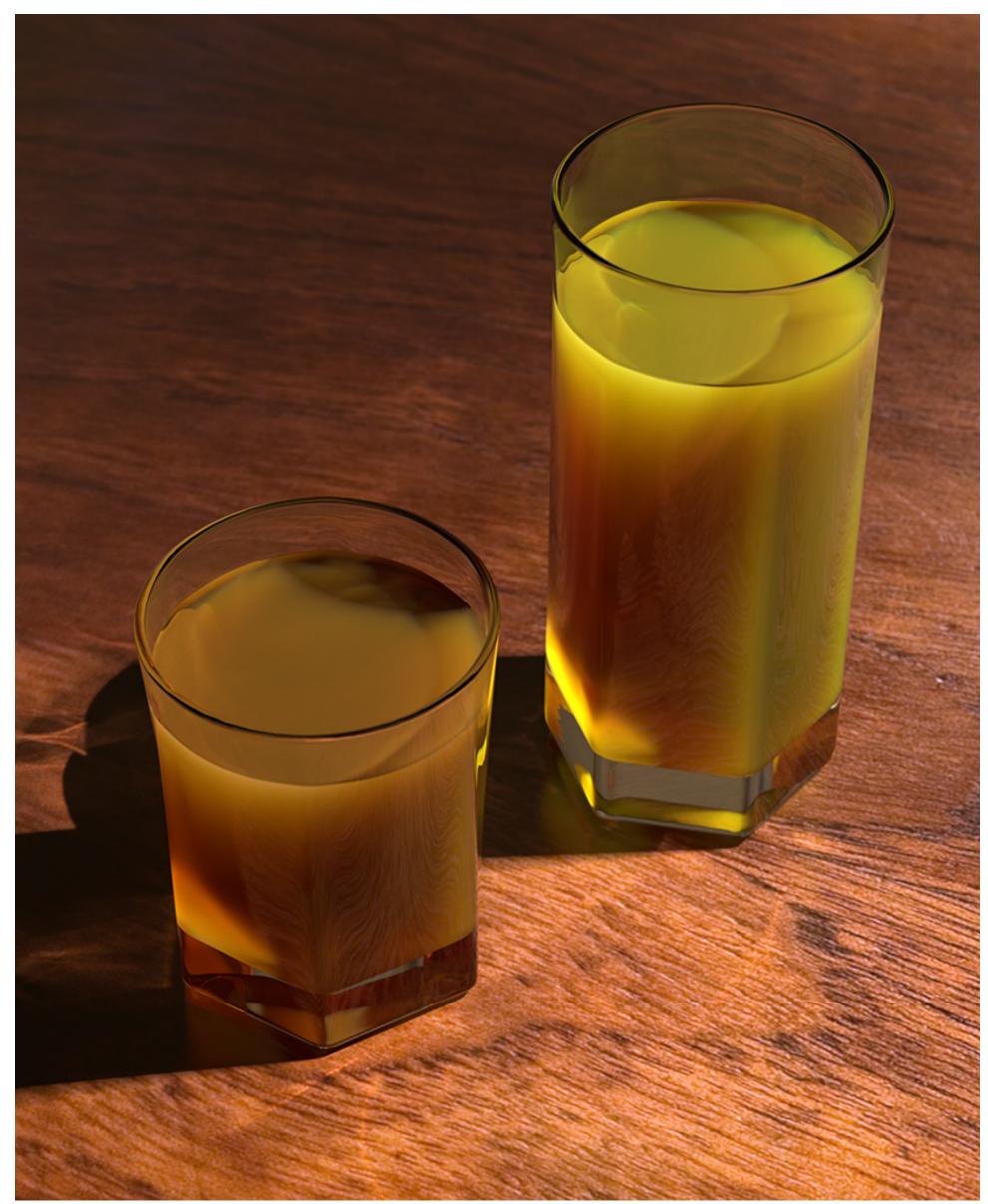




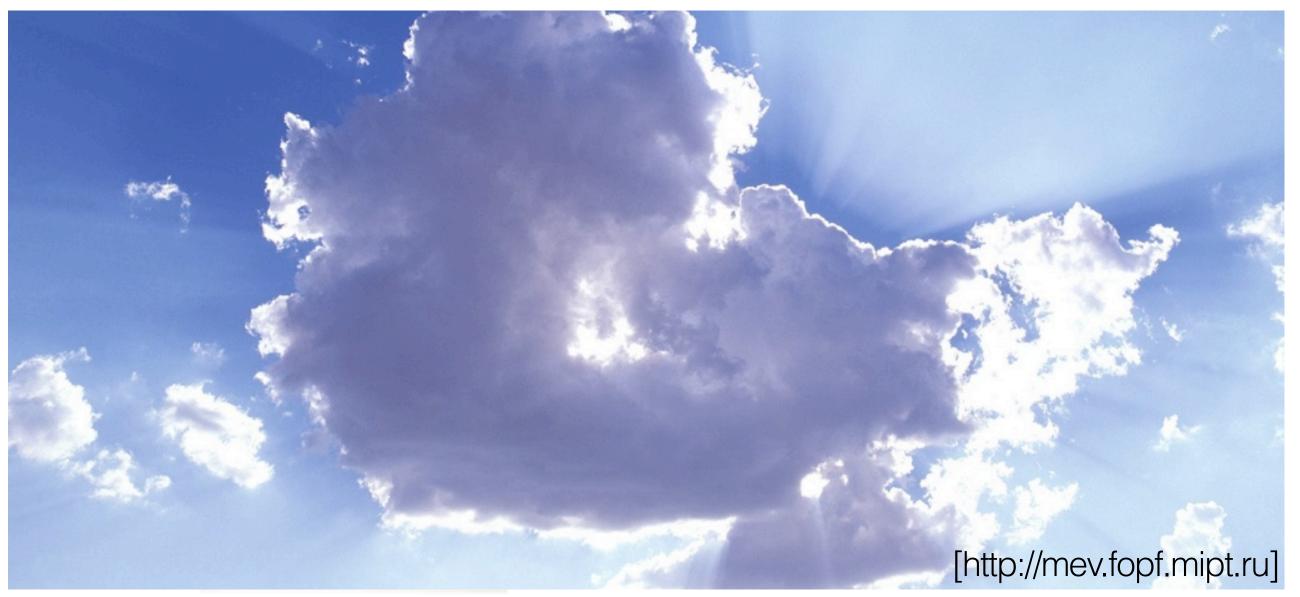




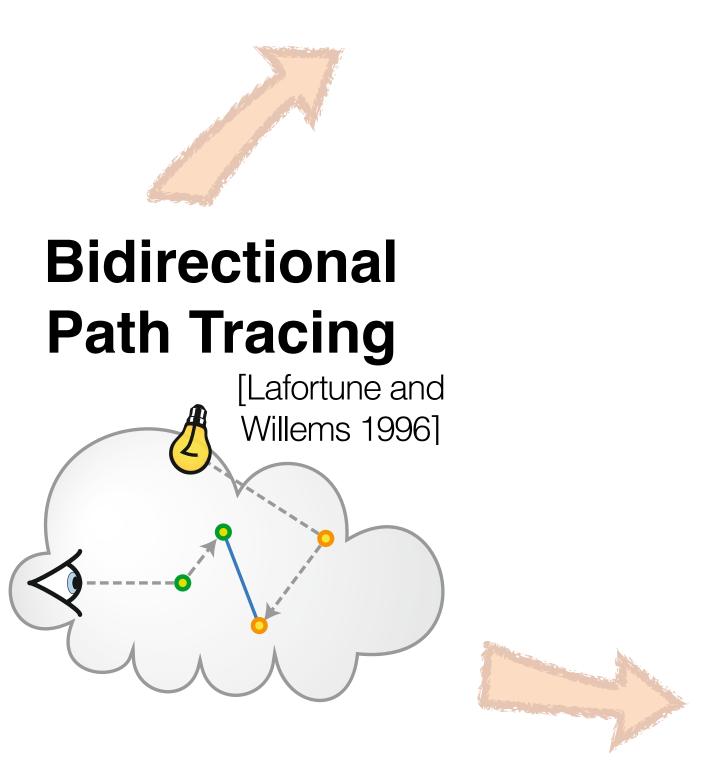


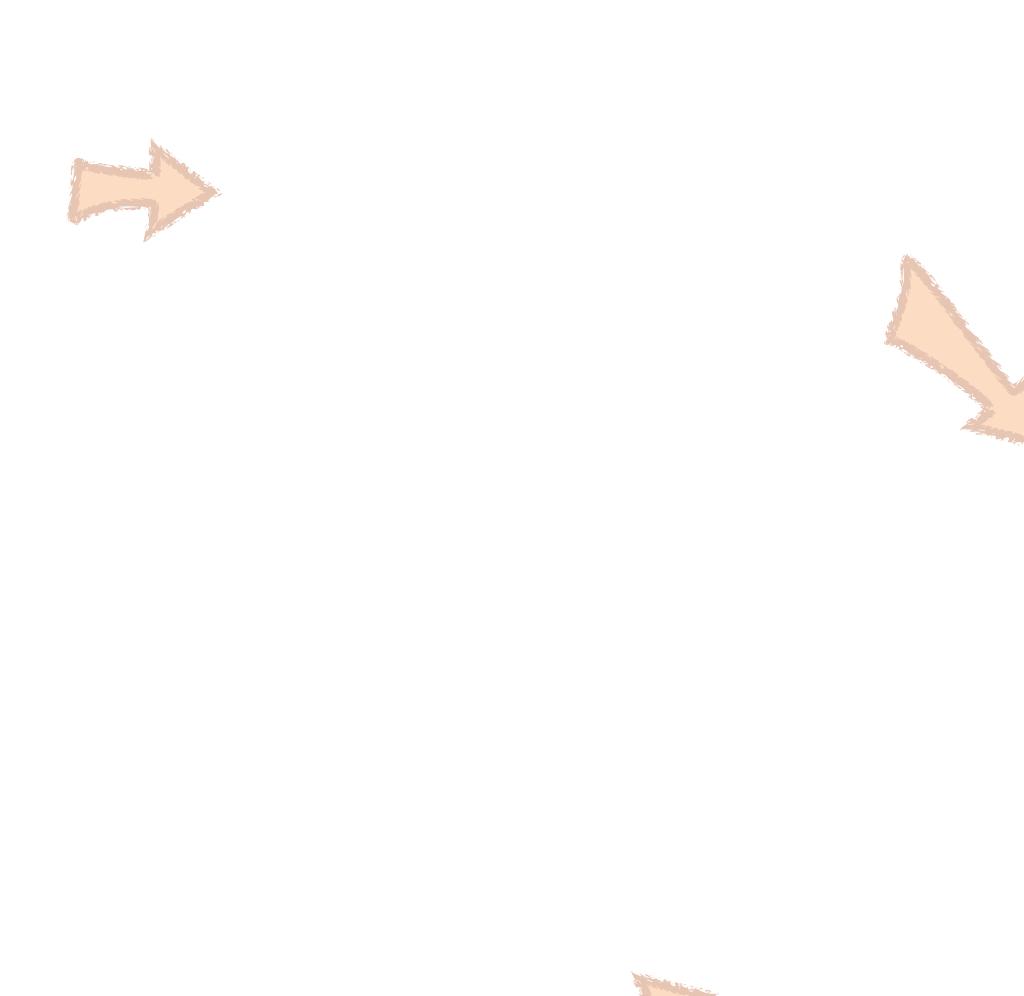




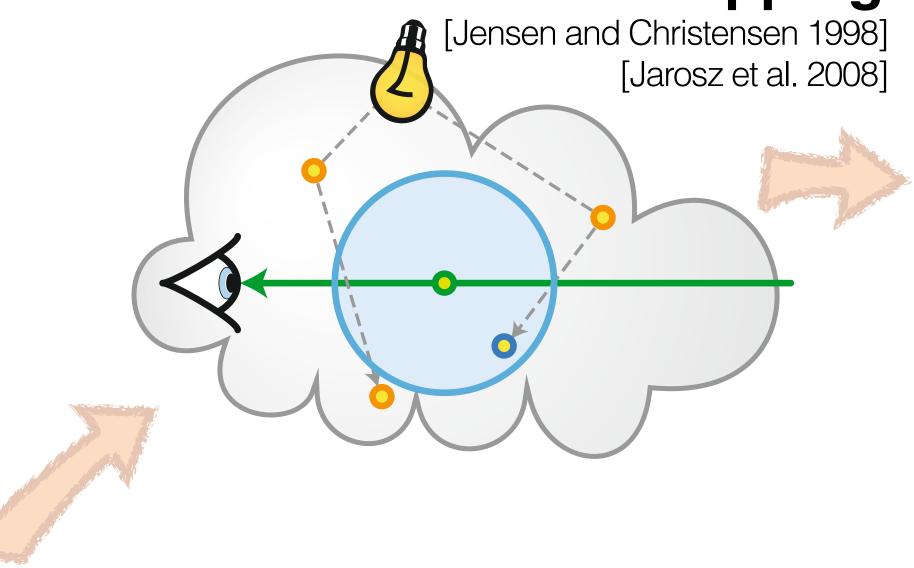


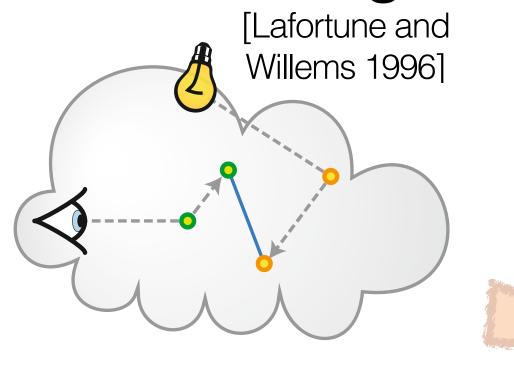


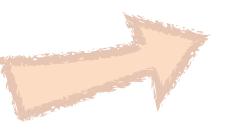




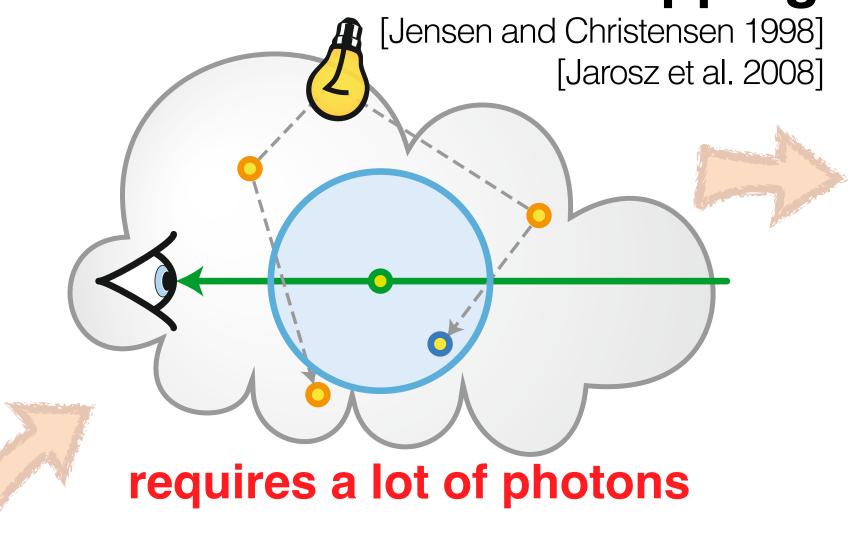
Volumetric Photon Mapping

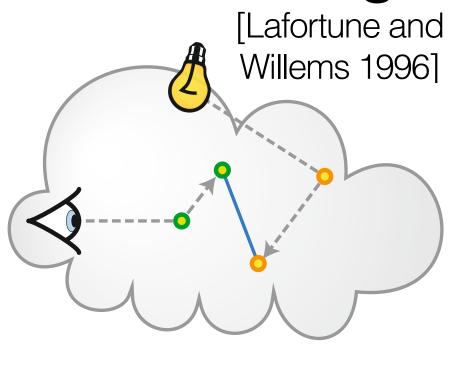


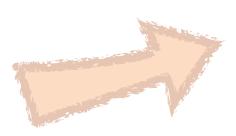


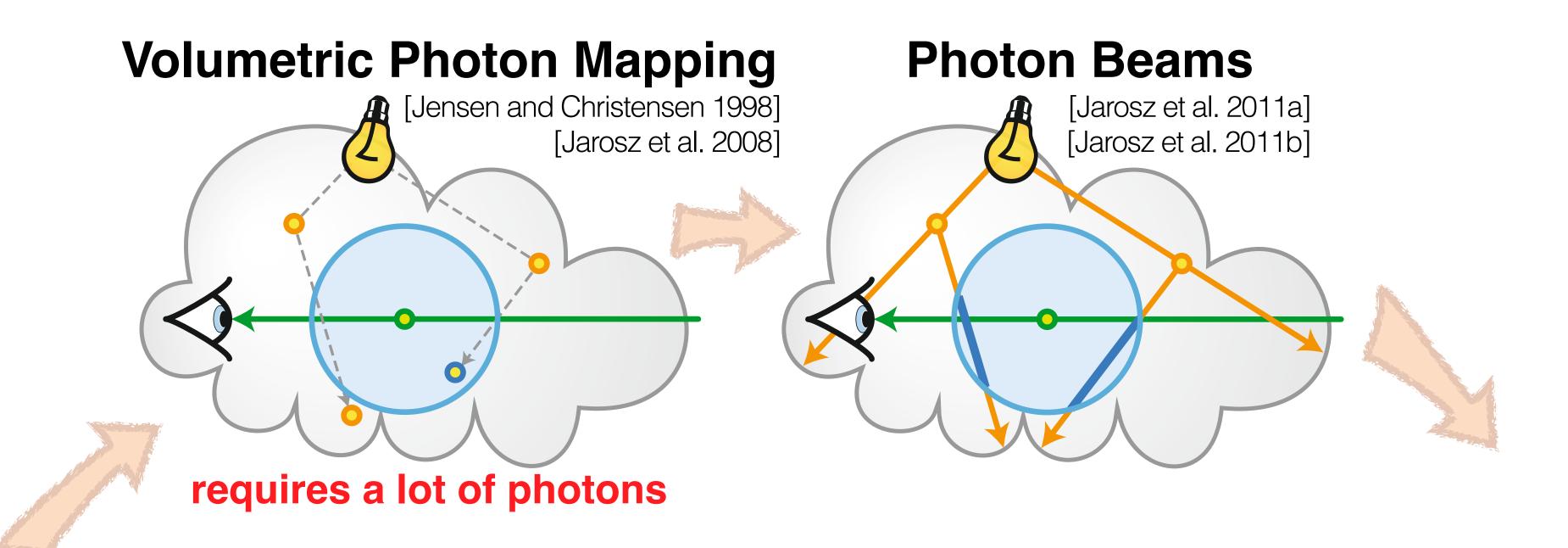


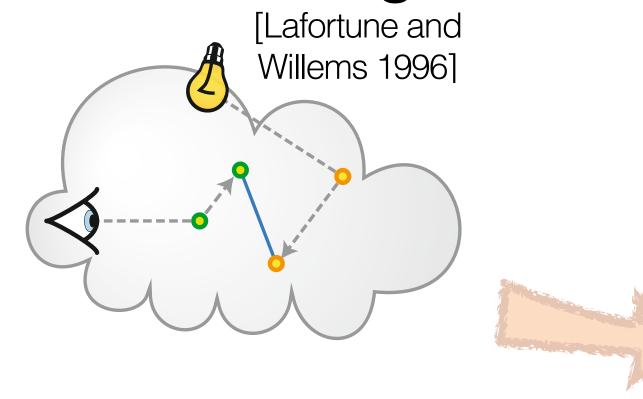
Volumetric Photon Mapping

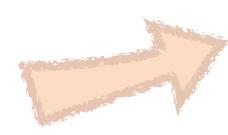


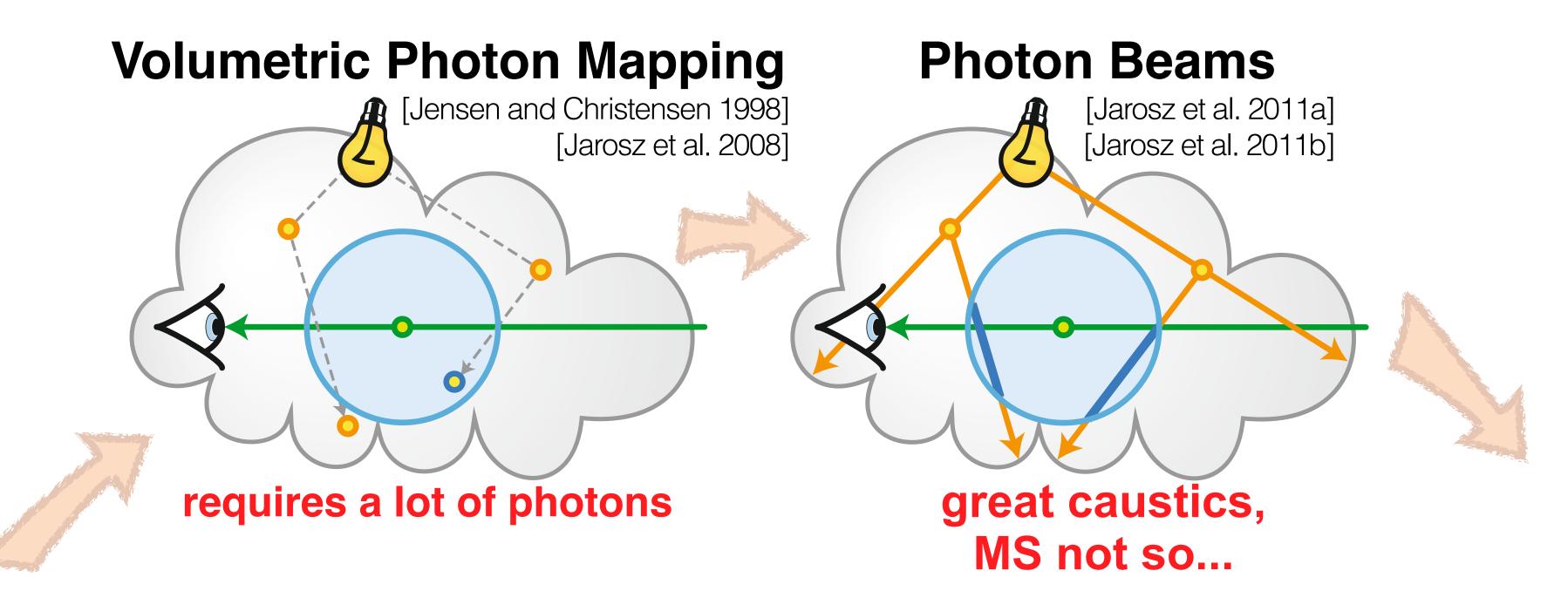


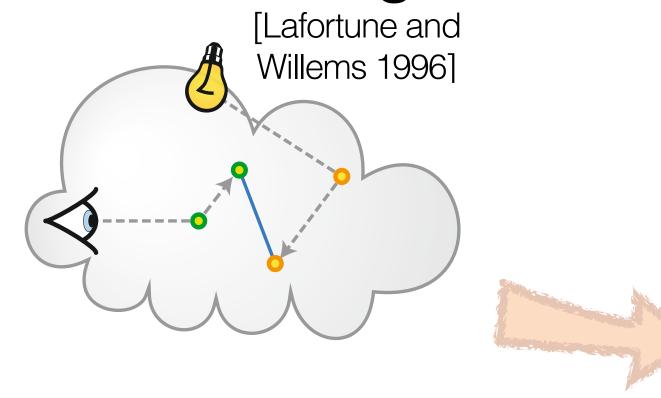


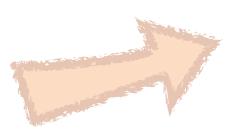


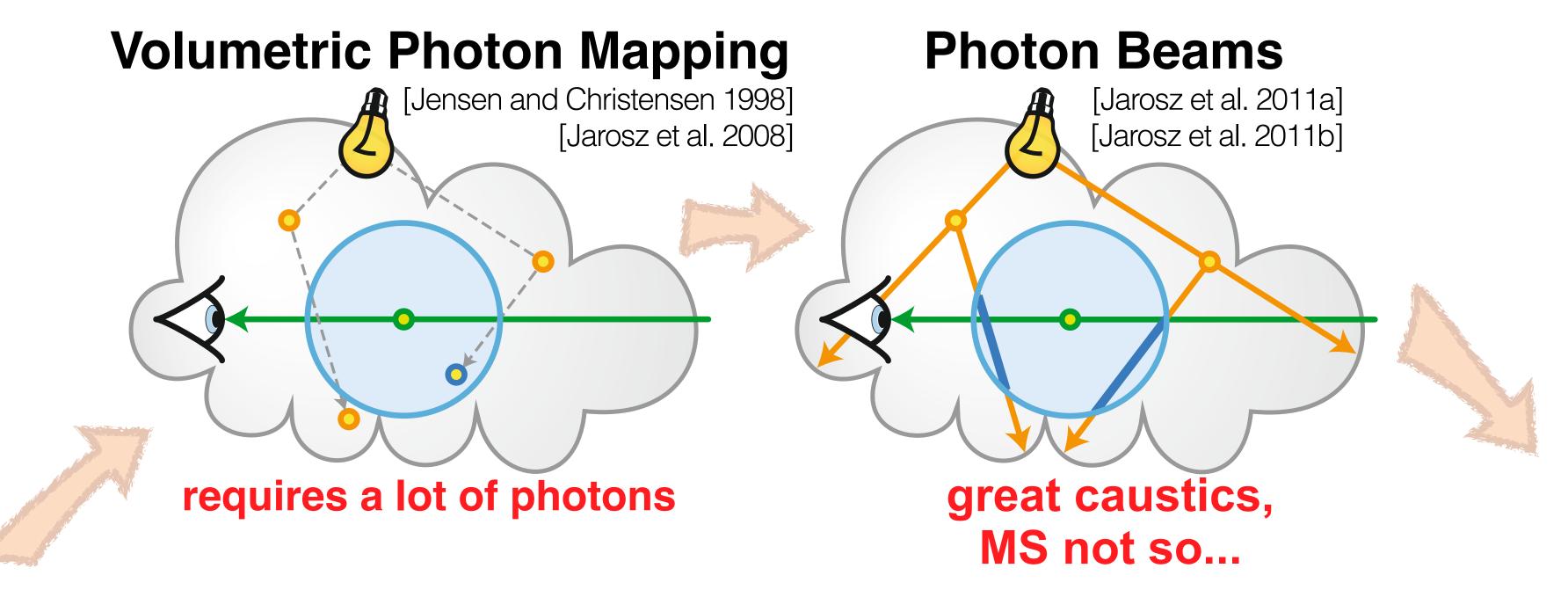




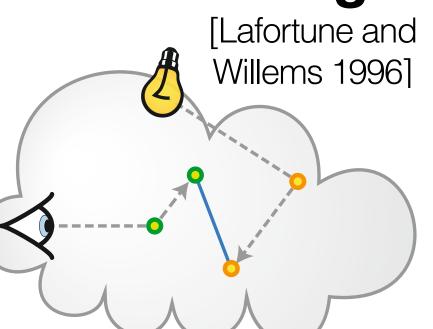




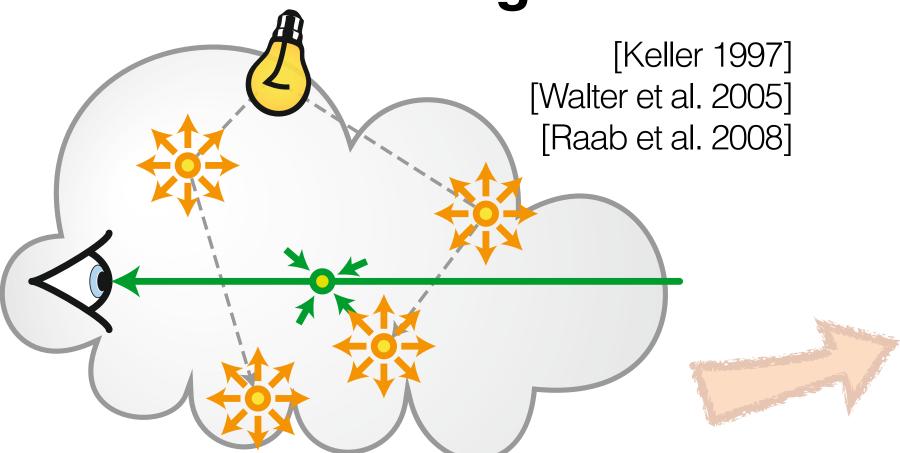


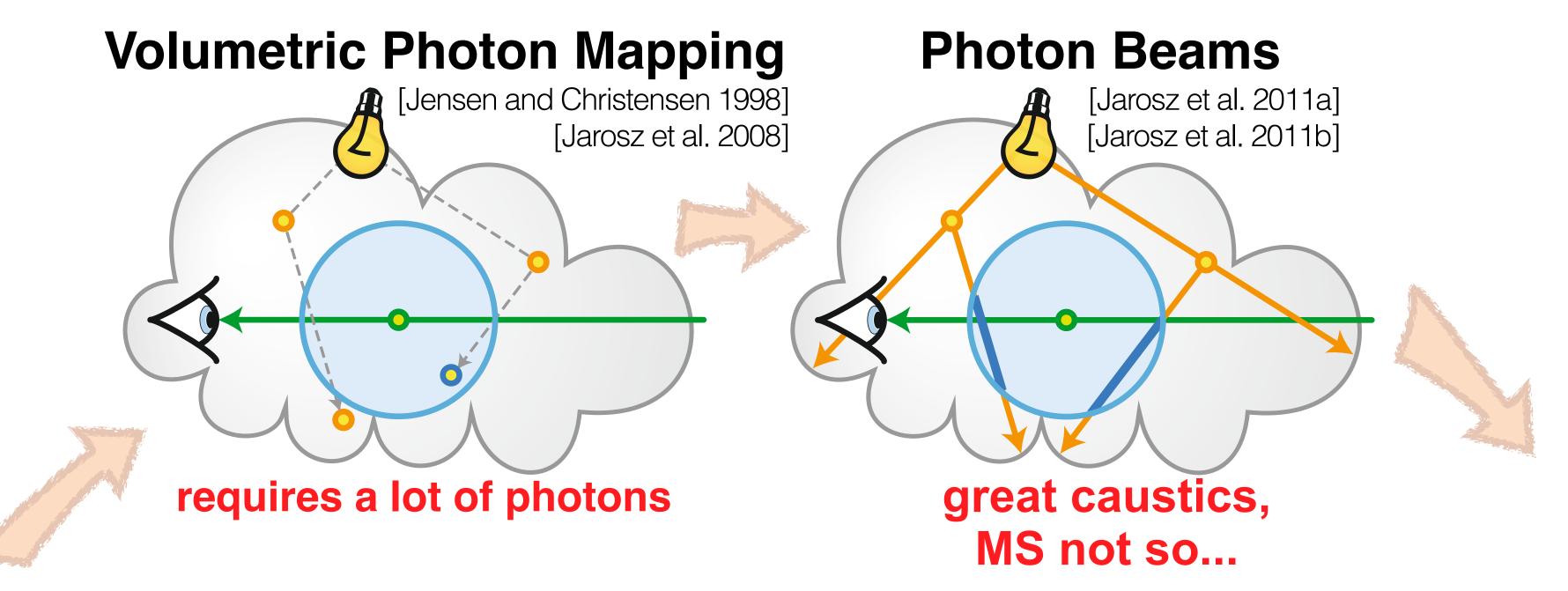


Bidirectional Path Tracing

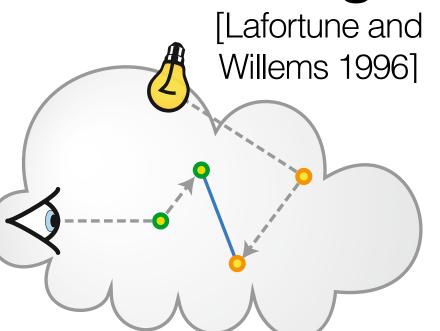


Virtual Point Lights

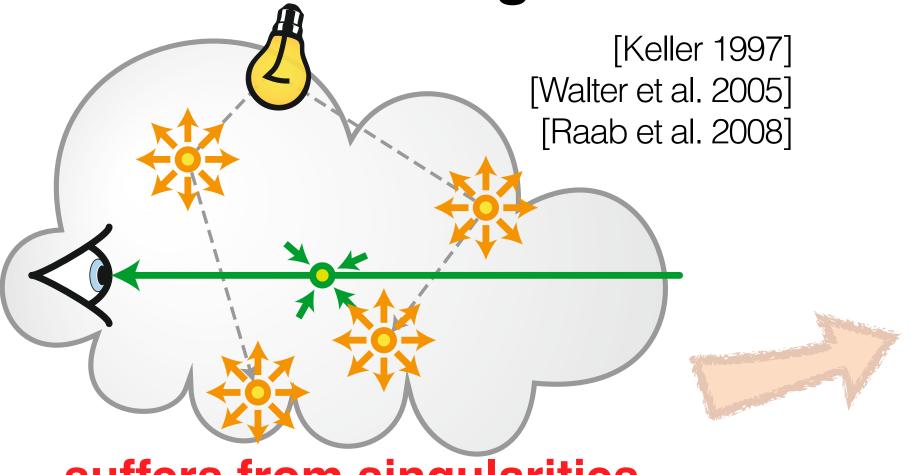


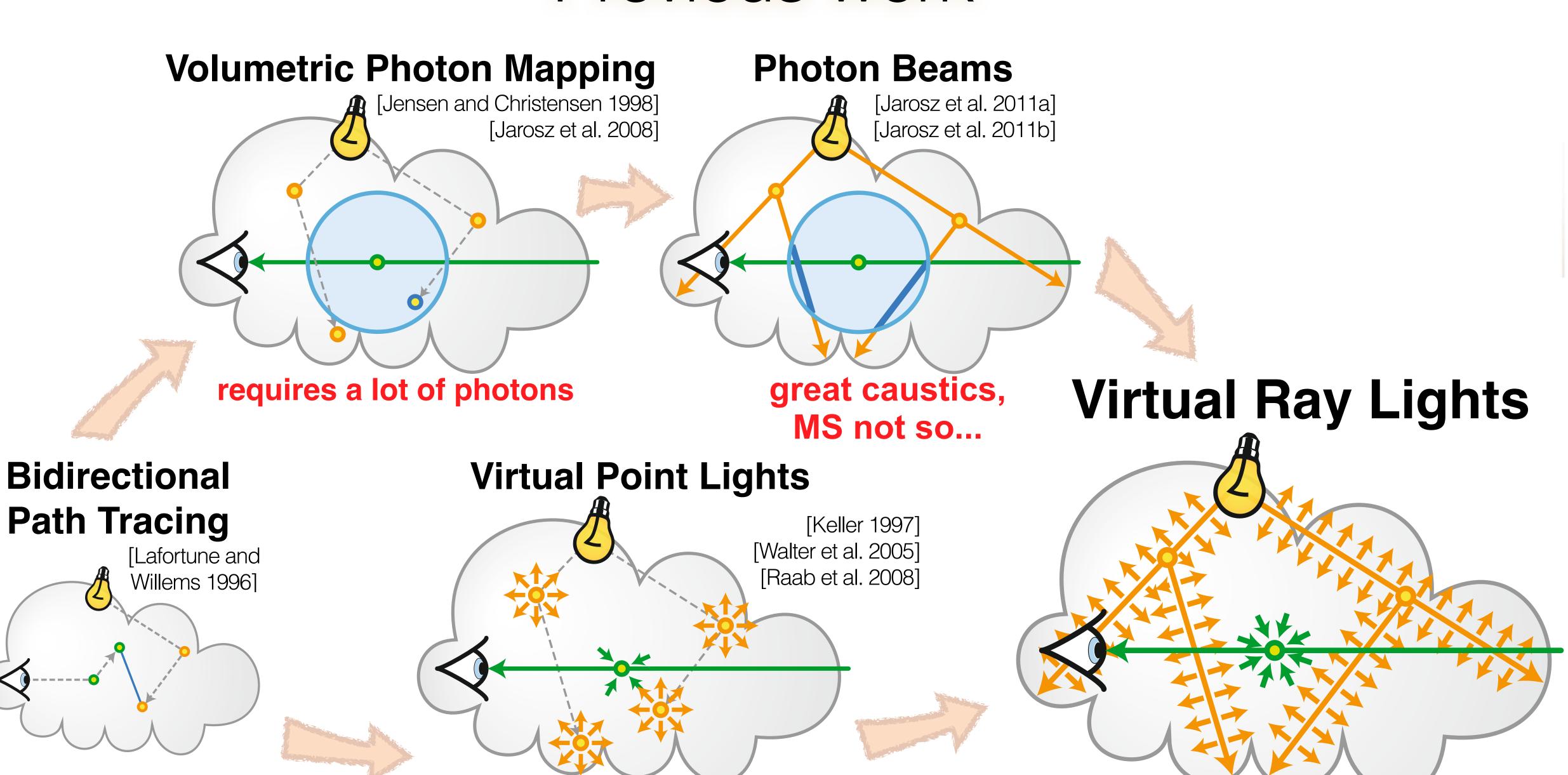


Bidirectional Path Tracing

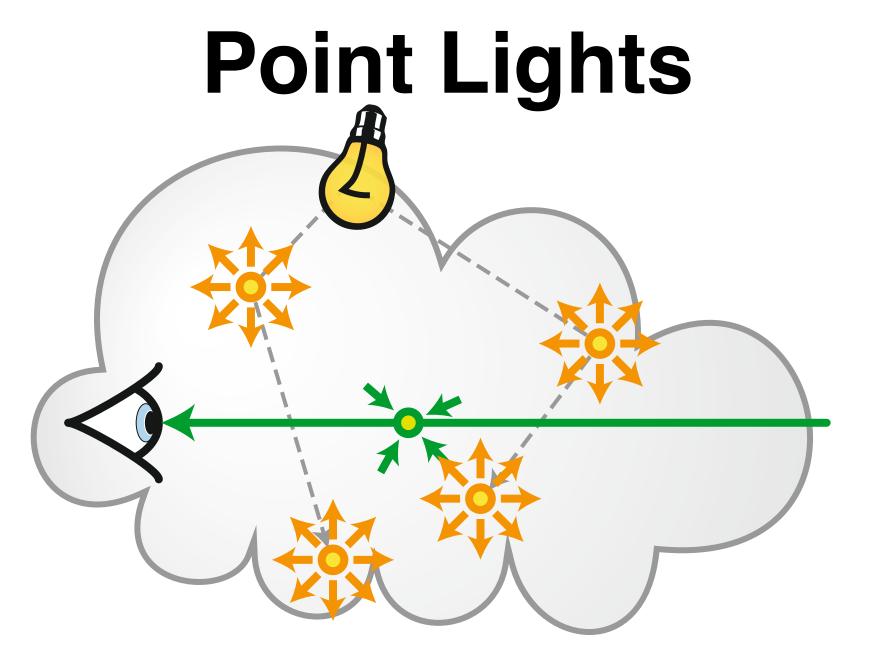


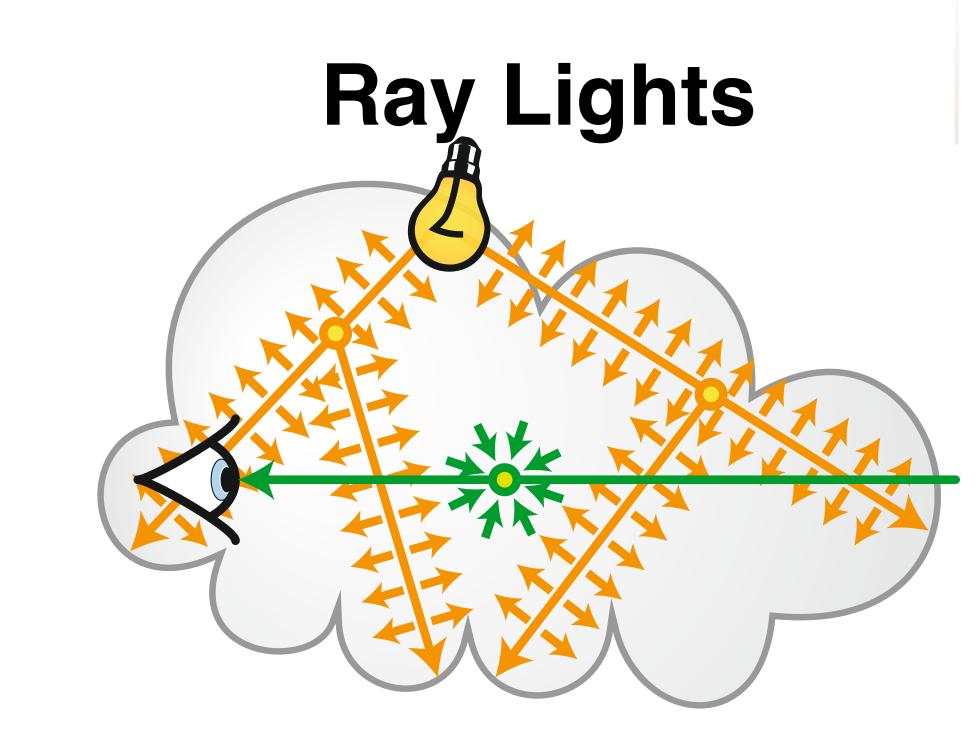
Virtual Point Lights

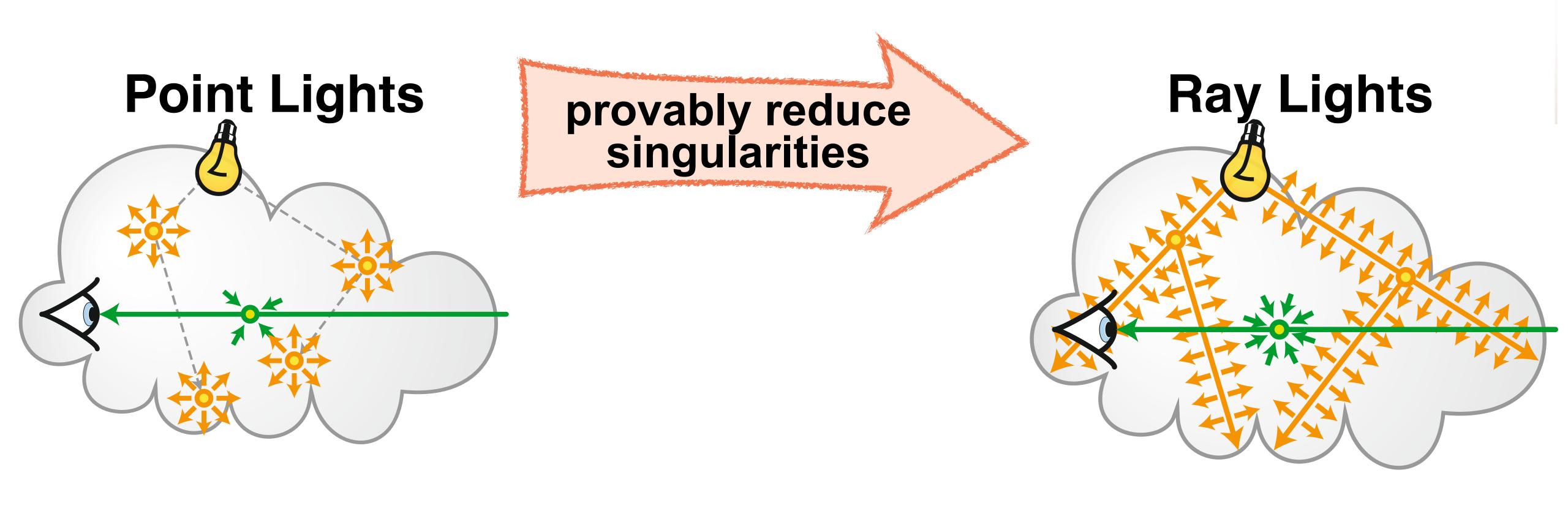


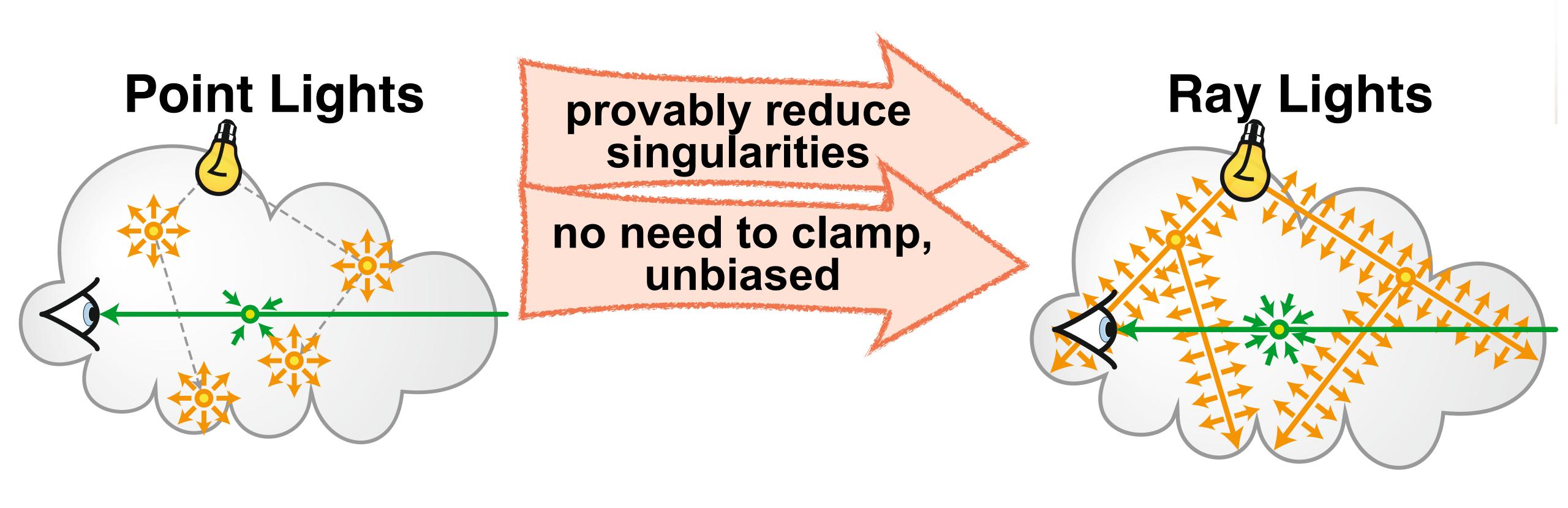


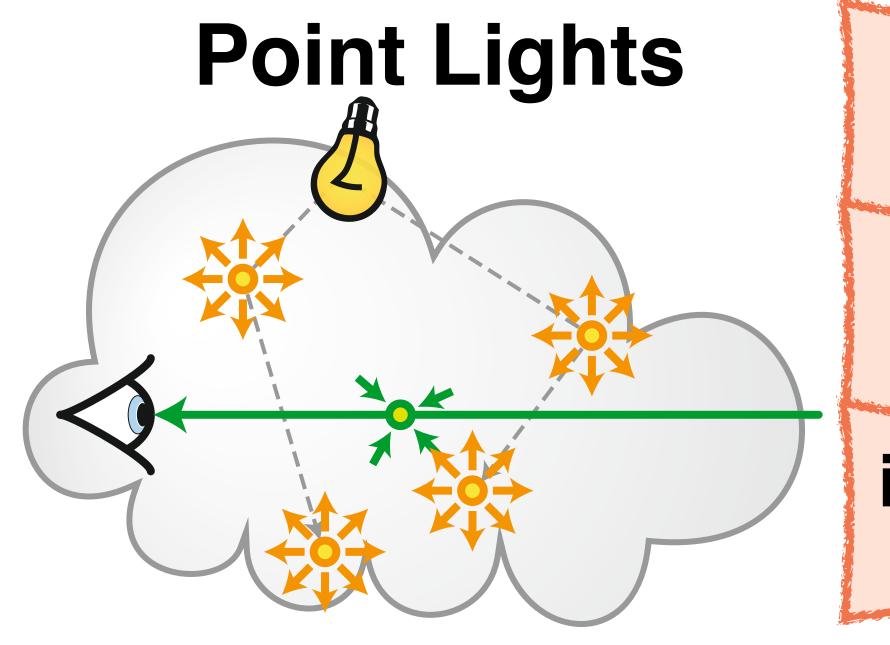
suffers from singularities







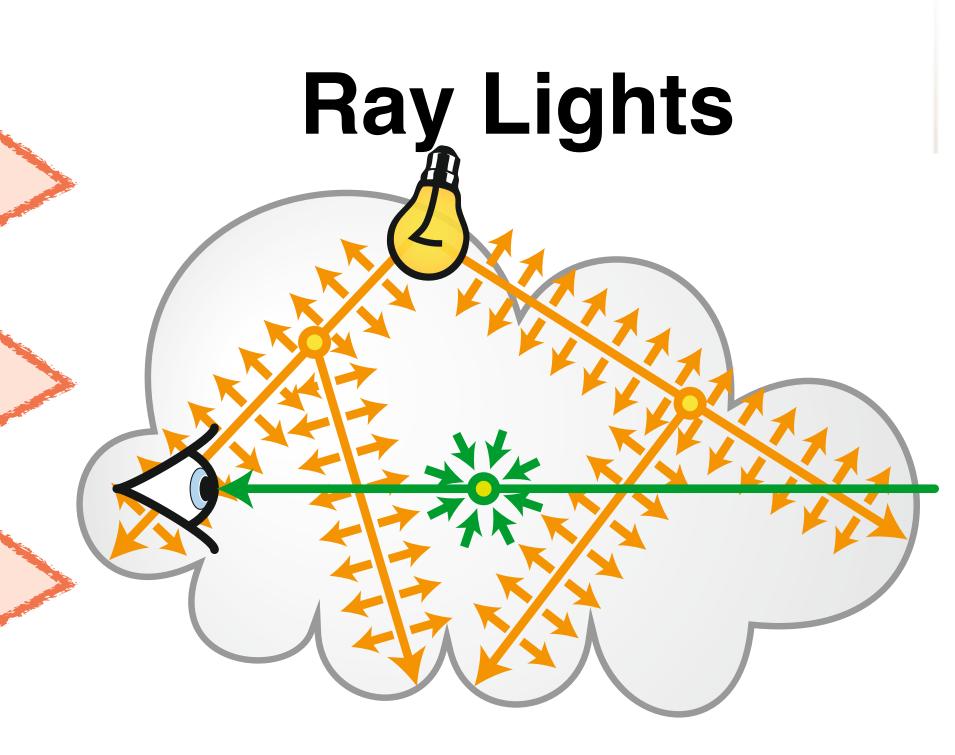




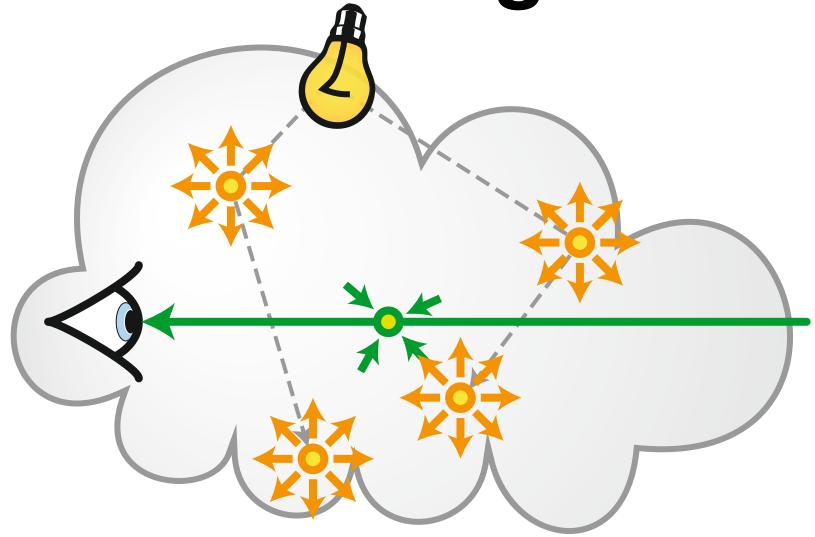
provably reduce singularities

no need to clamp, unbiased

increase sampling of path space





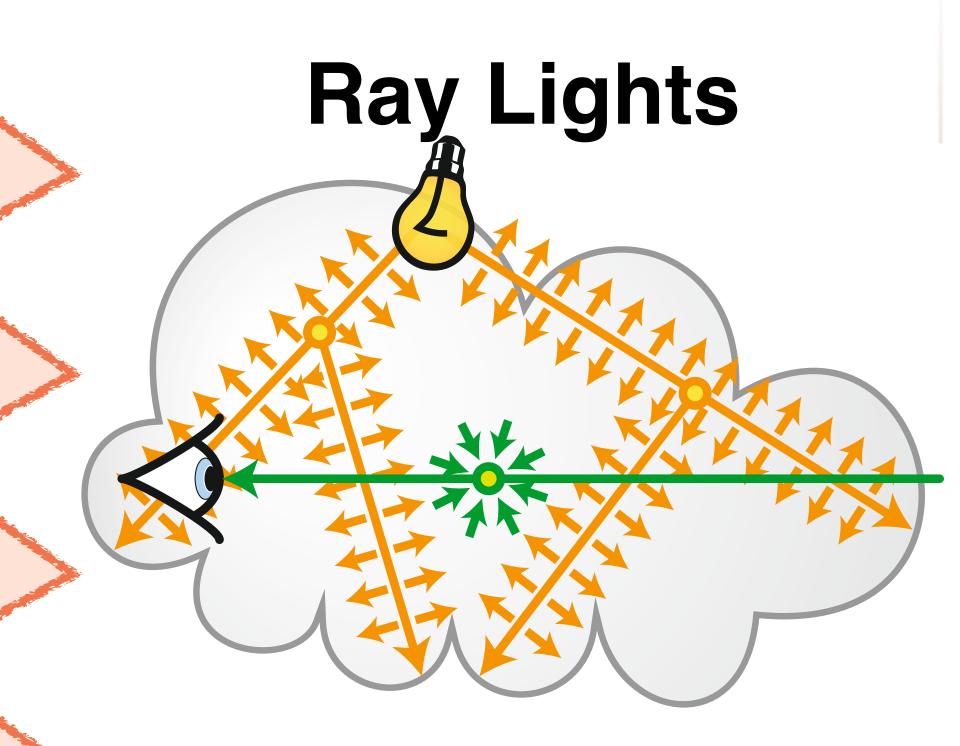


provably reduce singularities

no need to clamp, unbiased

increase sampling of path space

handle anisotropic "glossy" media



Quick demo

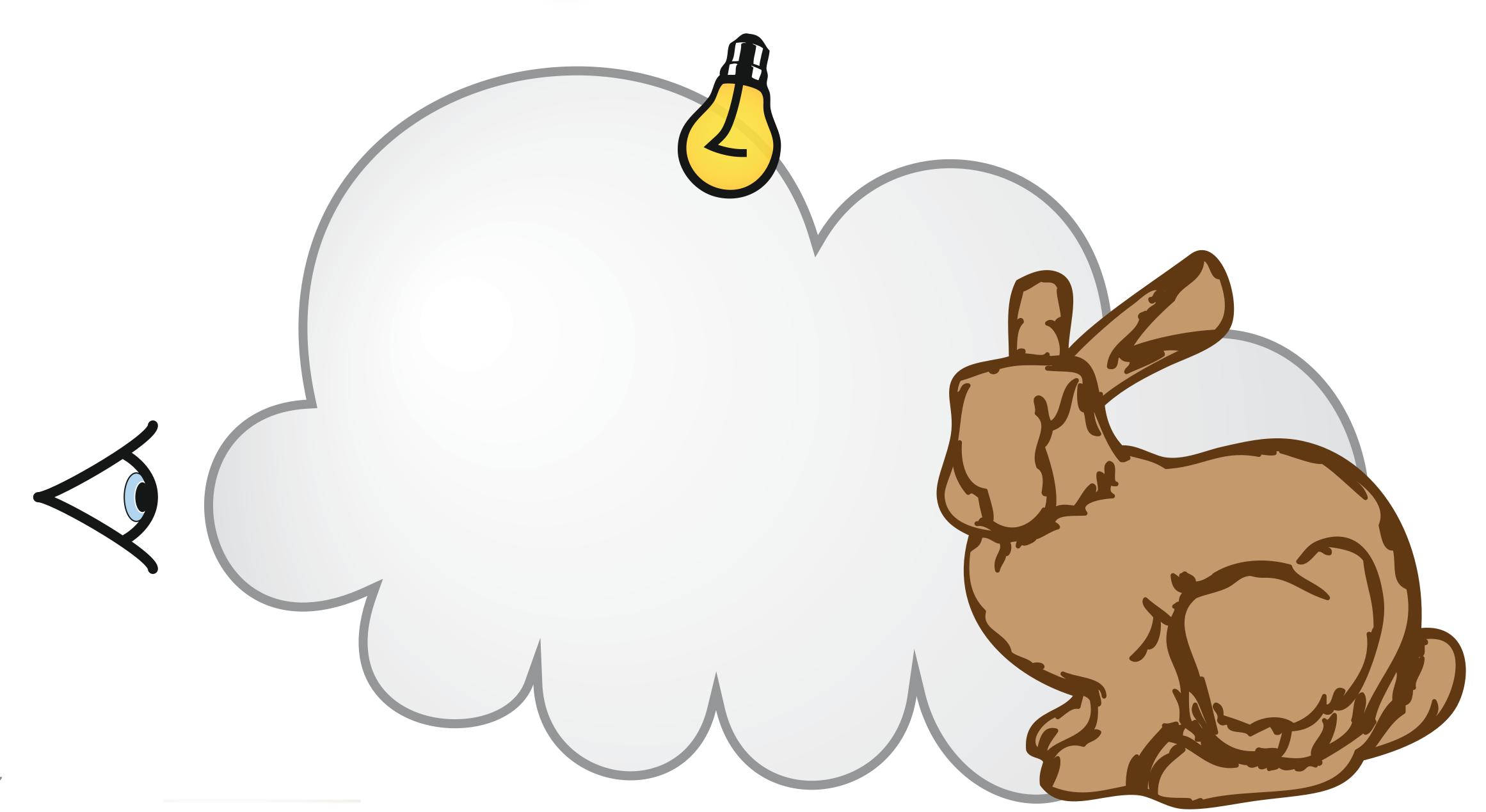


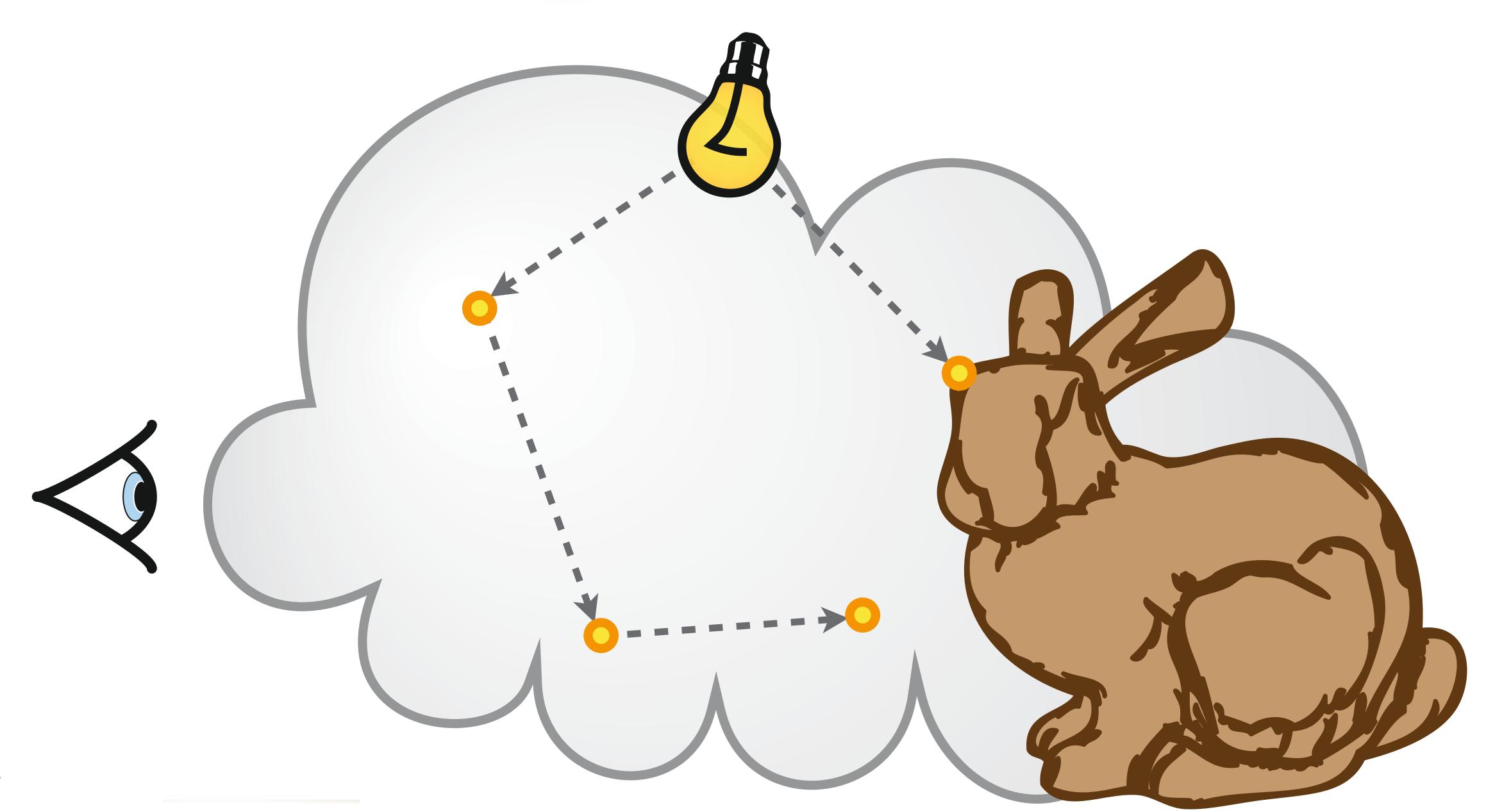
6 VRLs

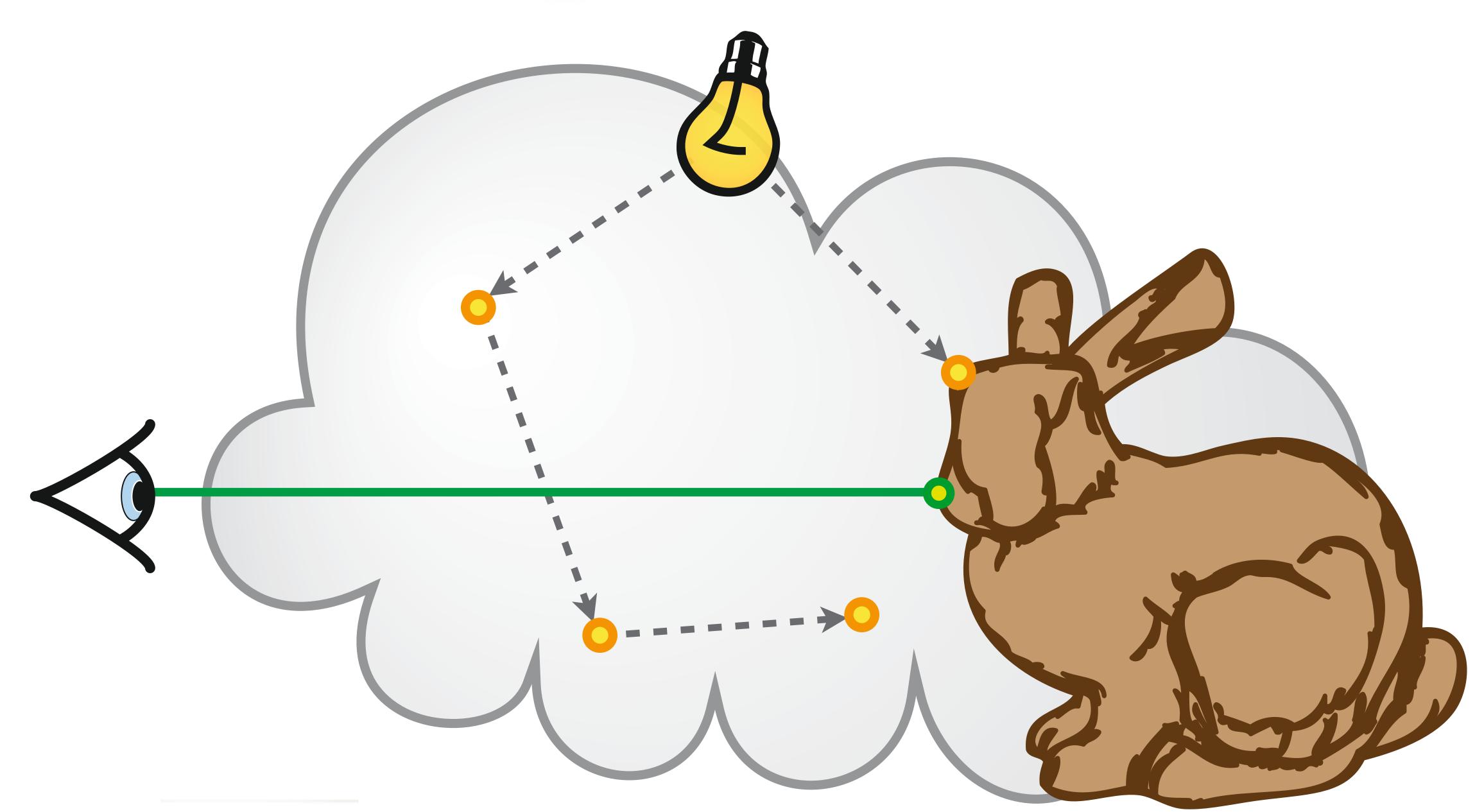
Quick demo

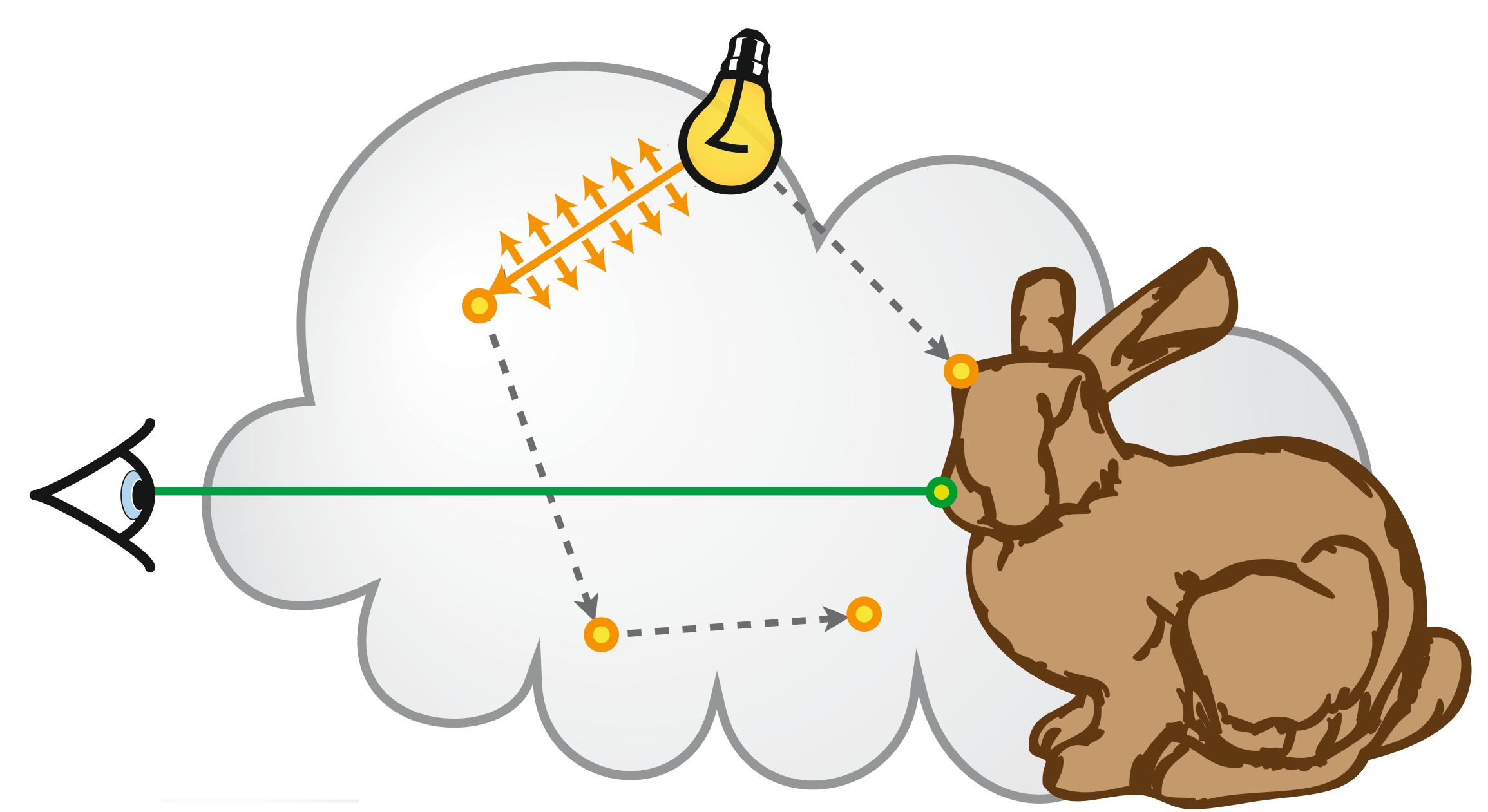


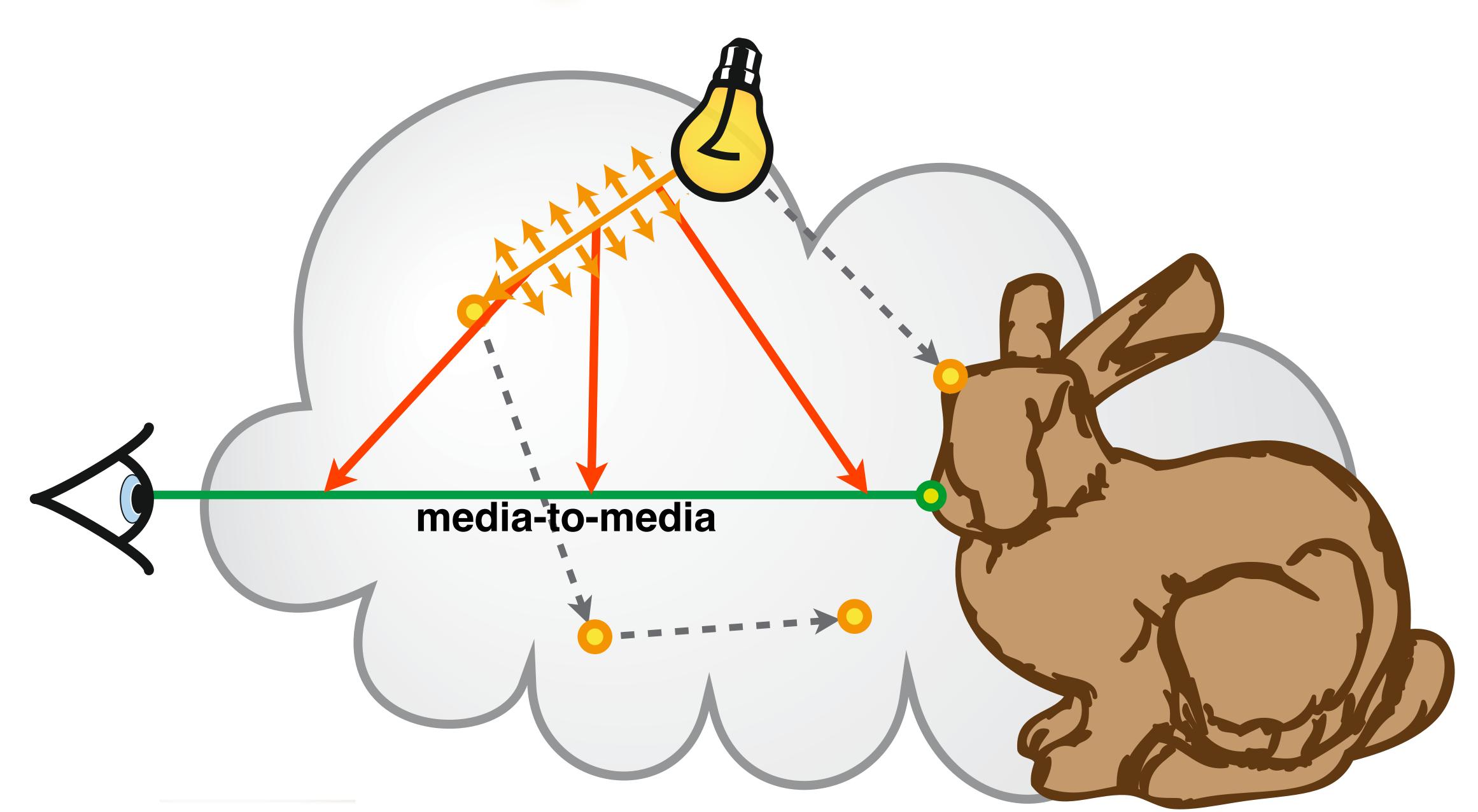
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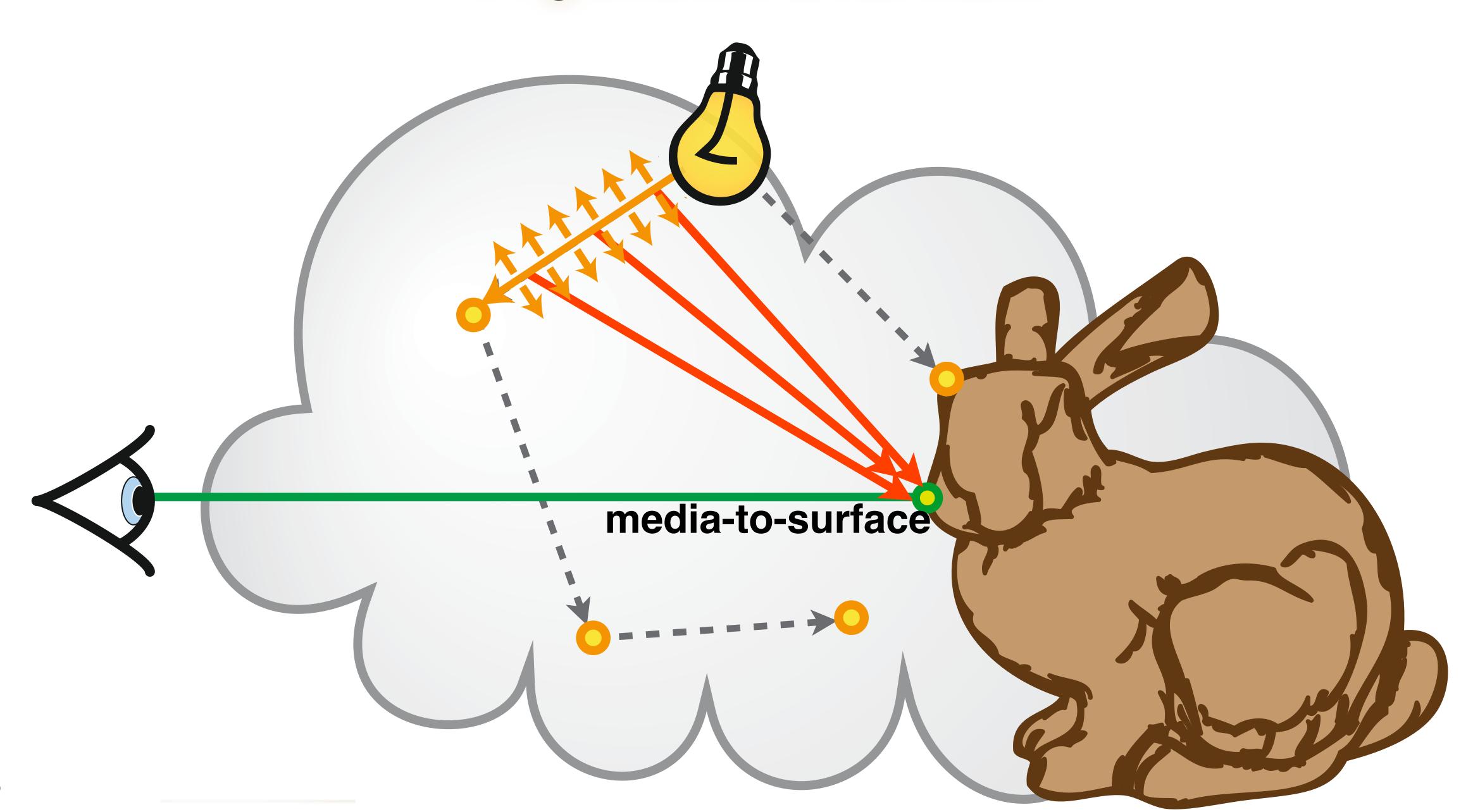


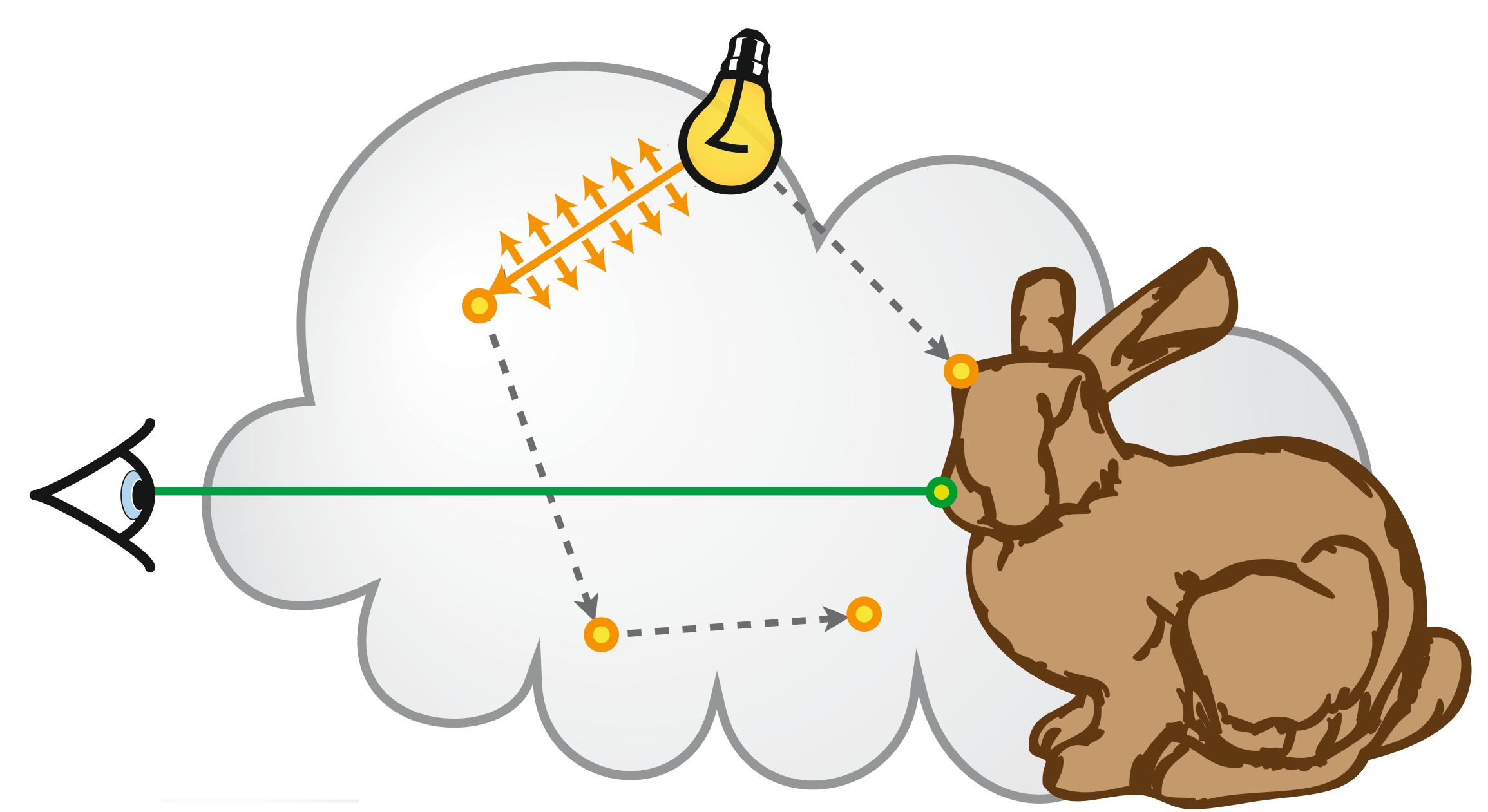


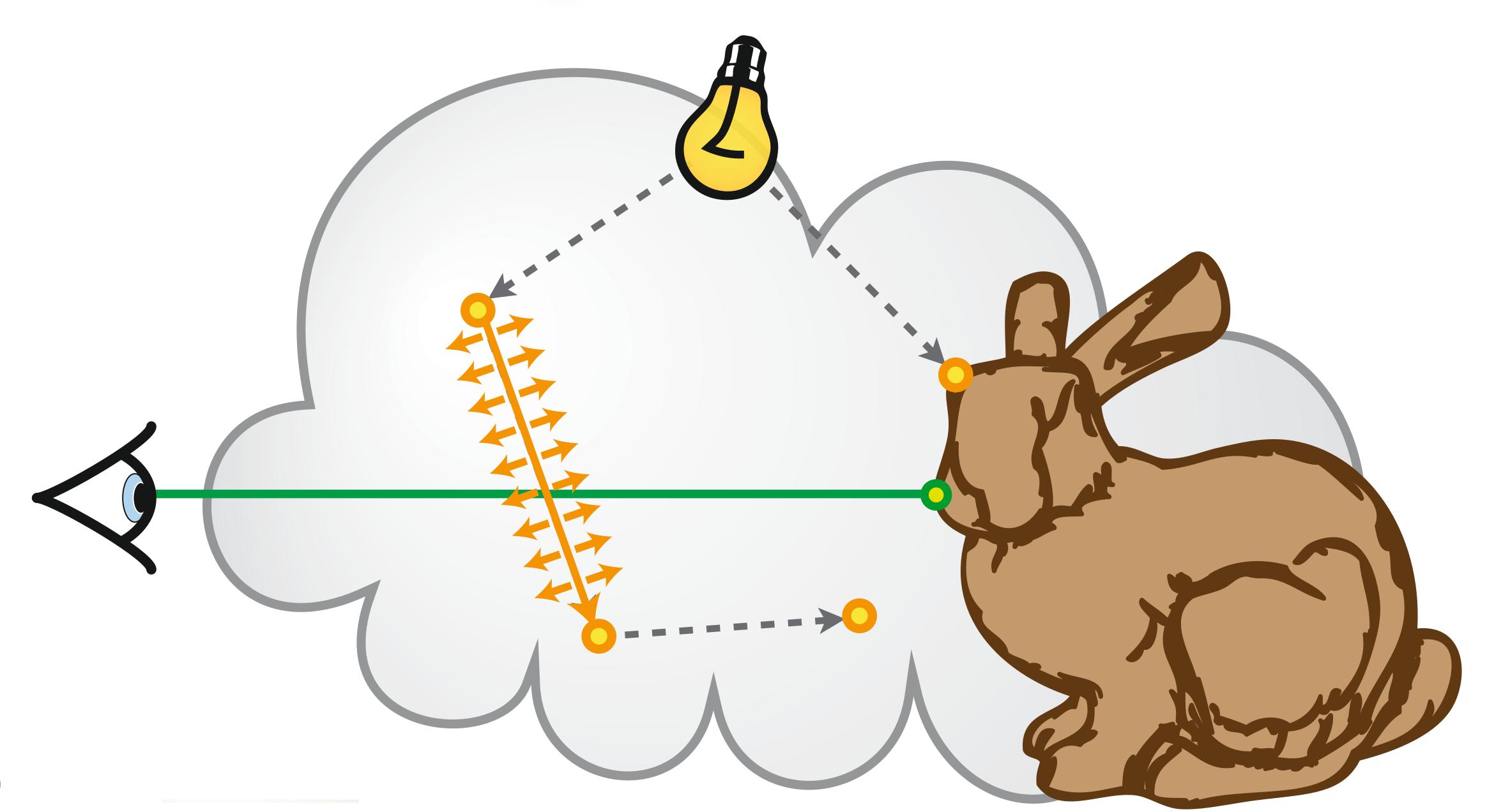


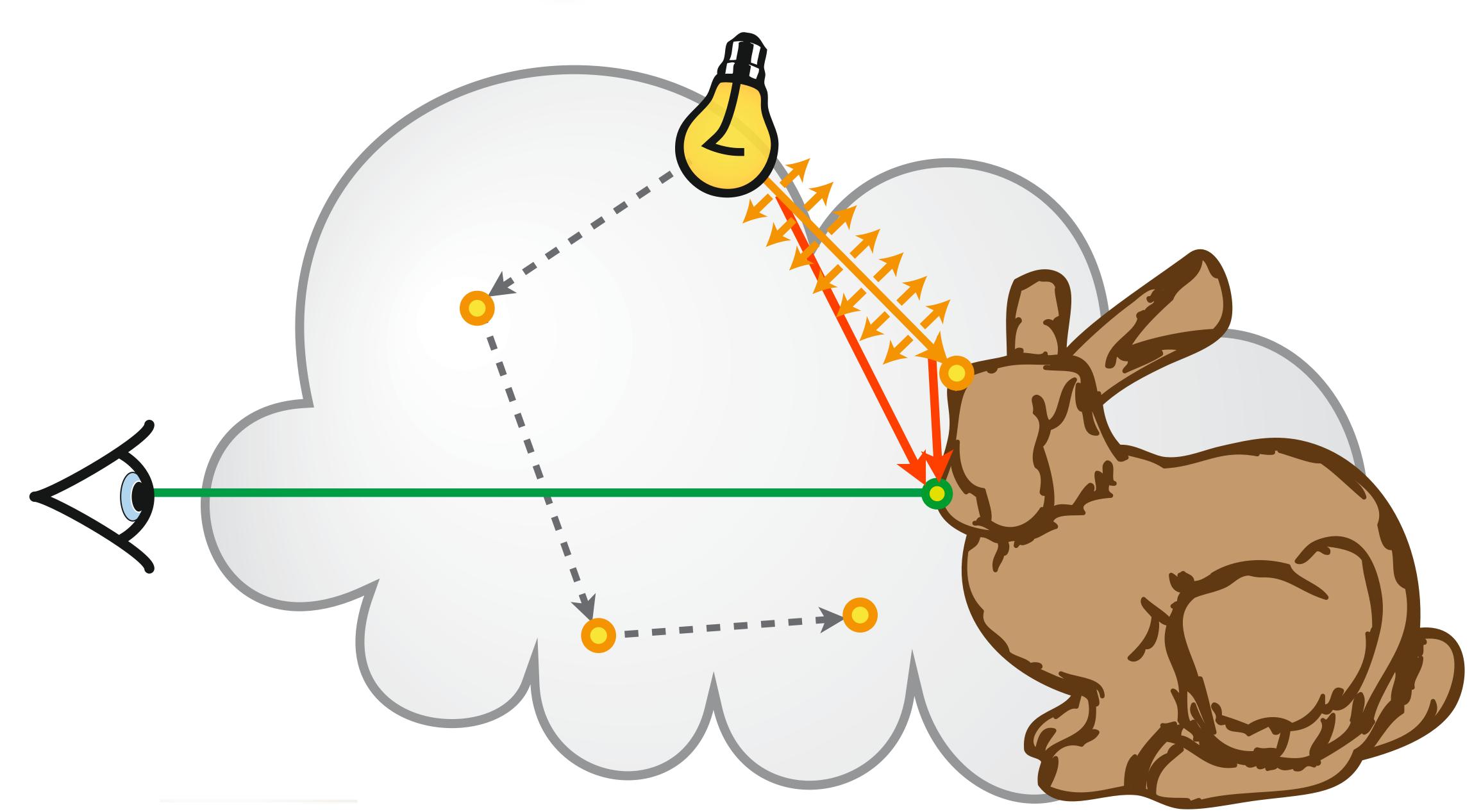


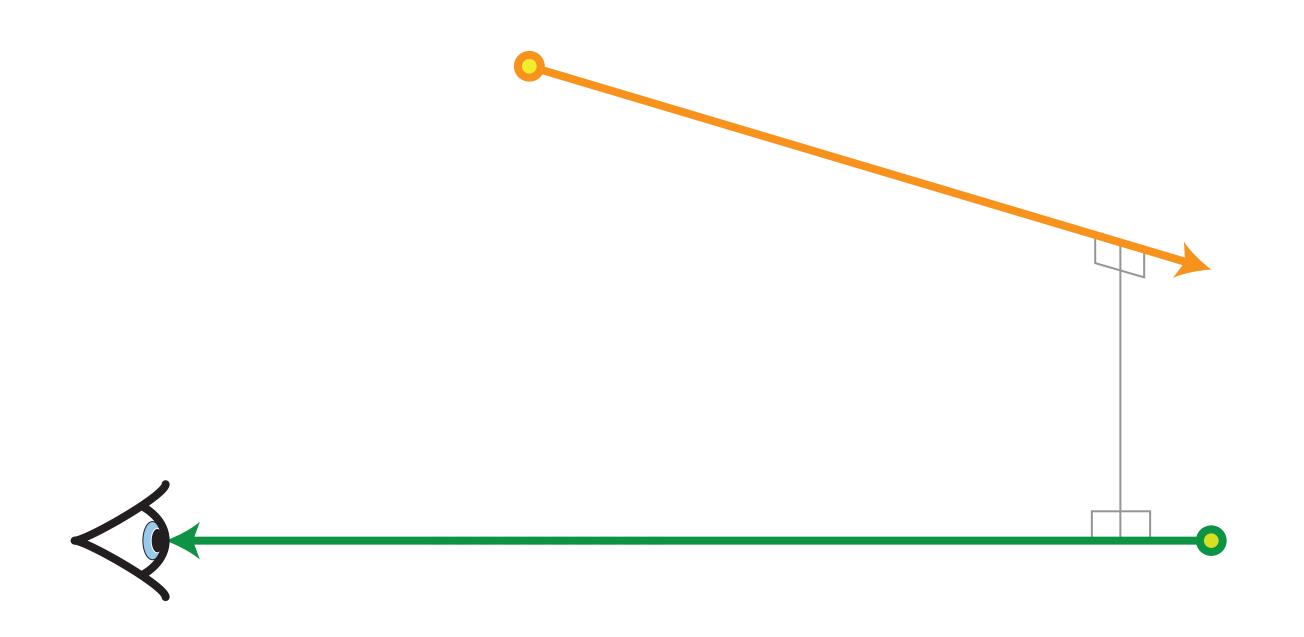


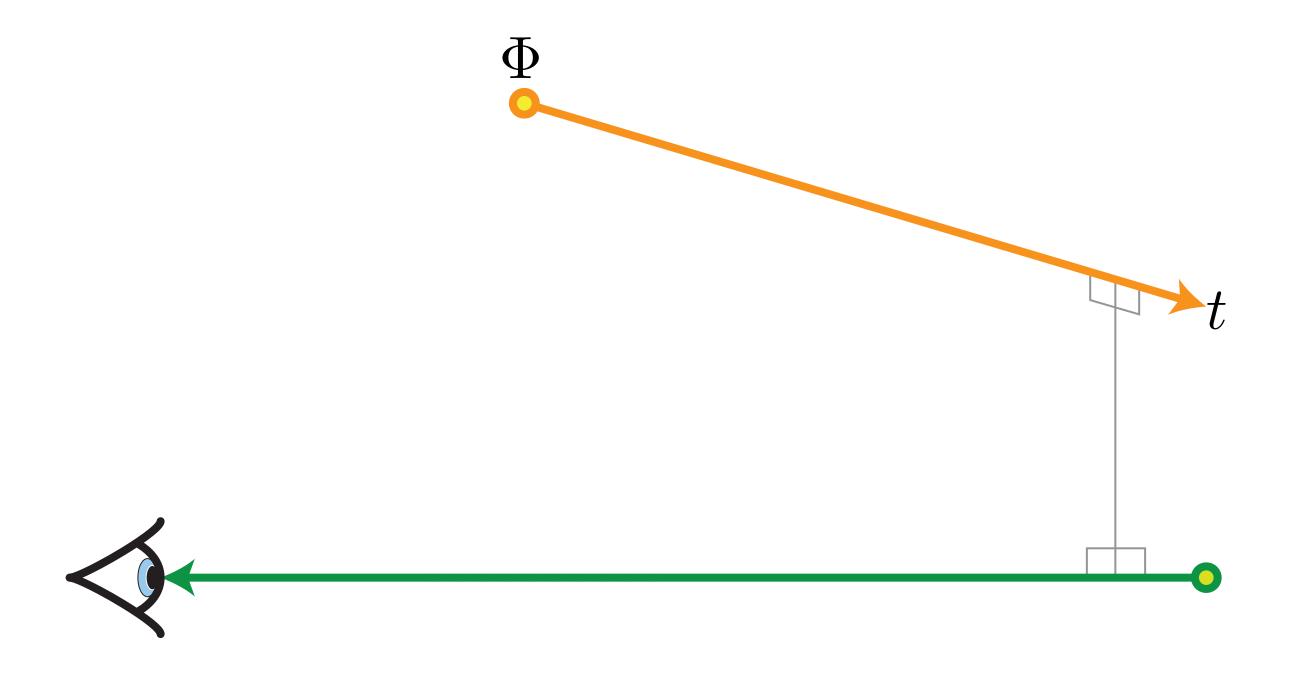


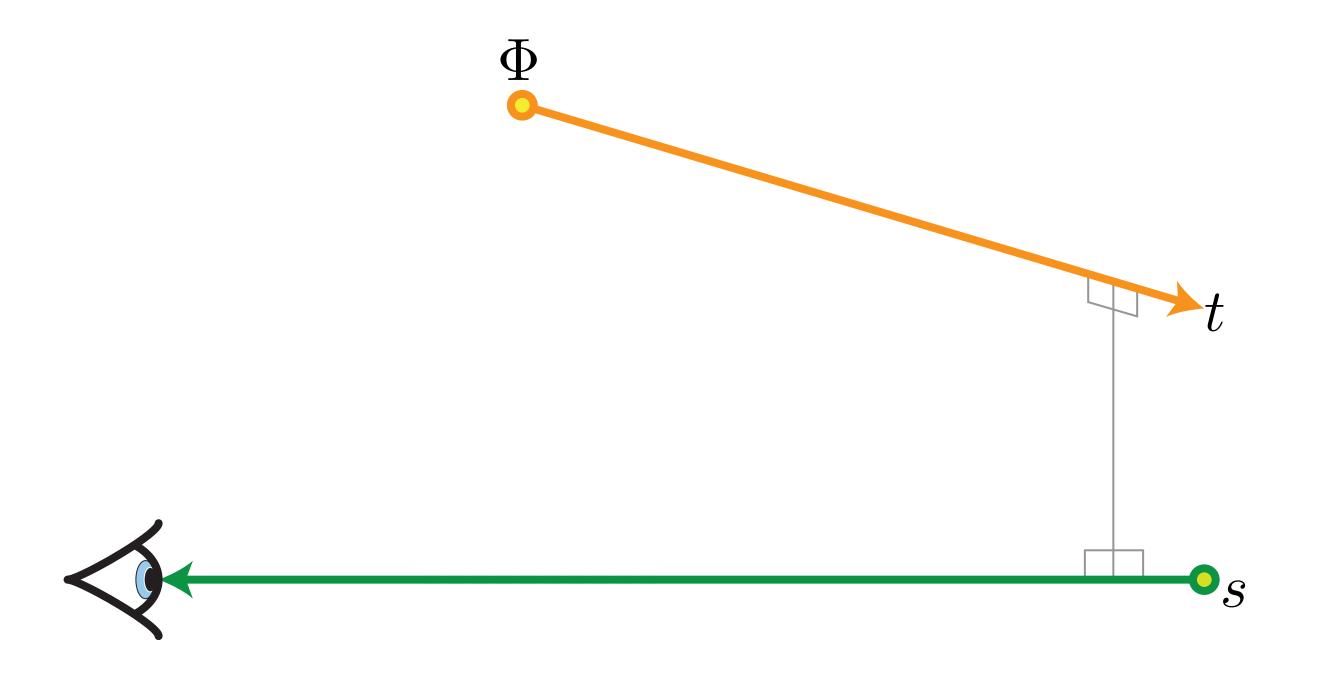


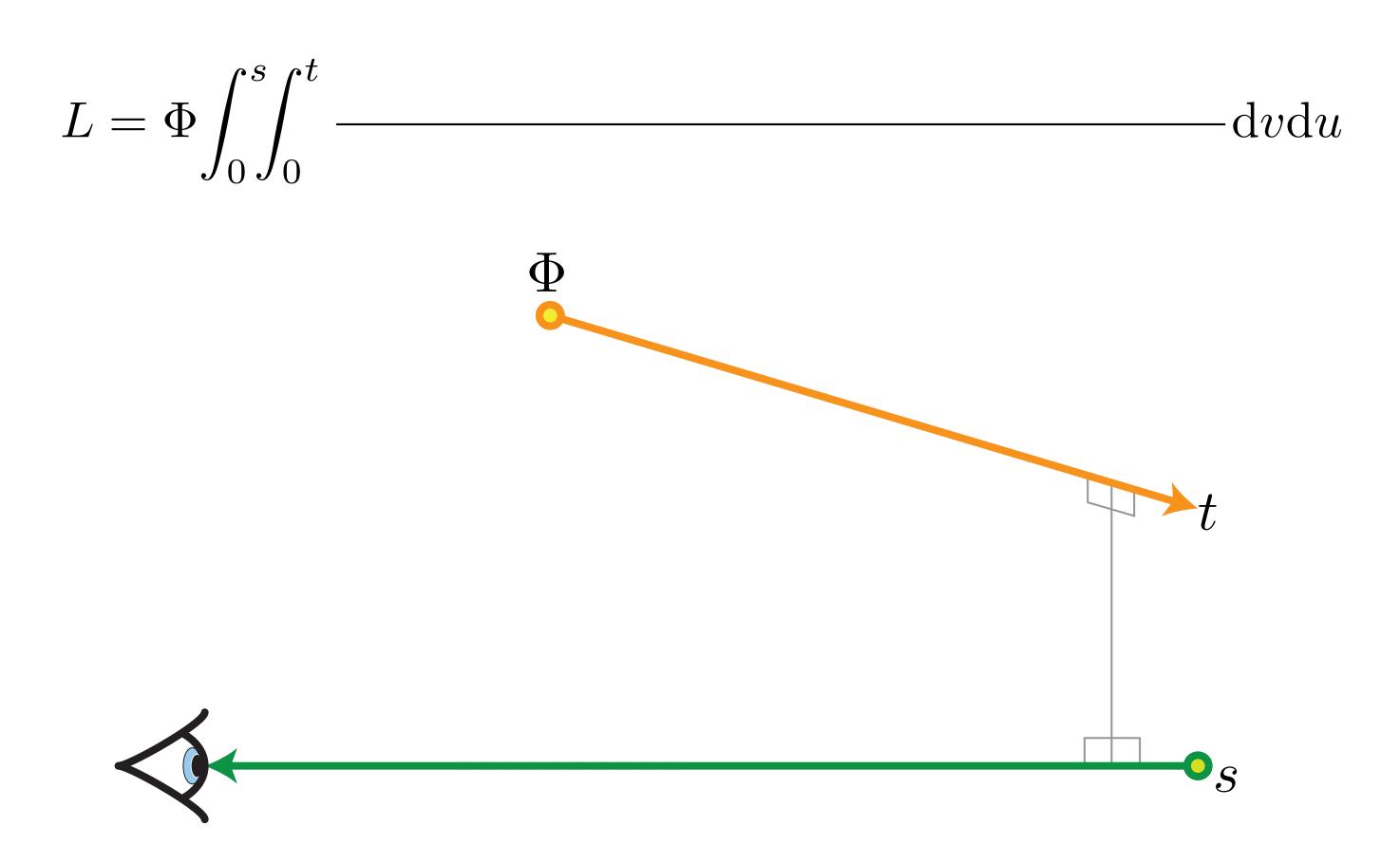


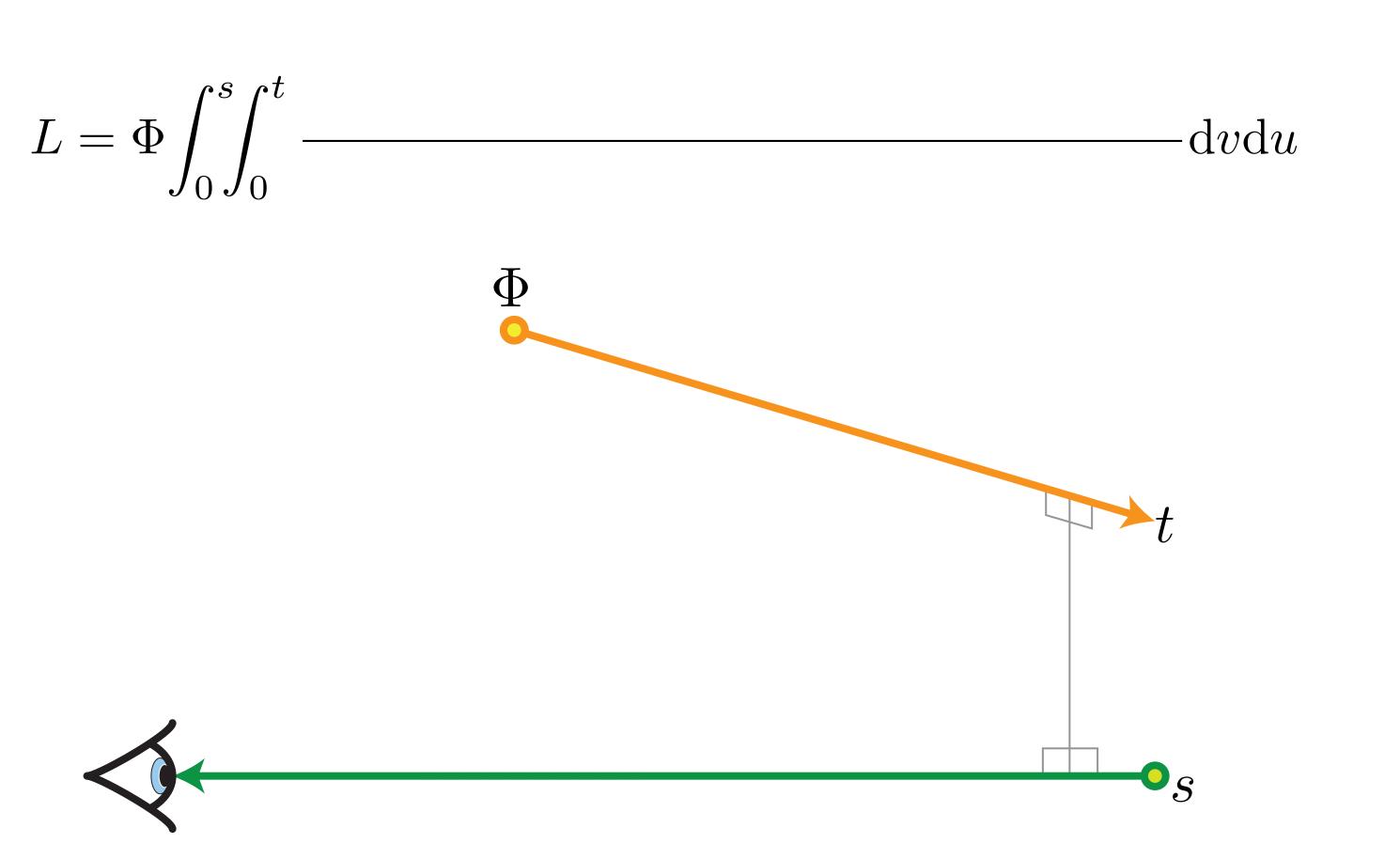


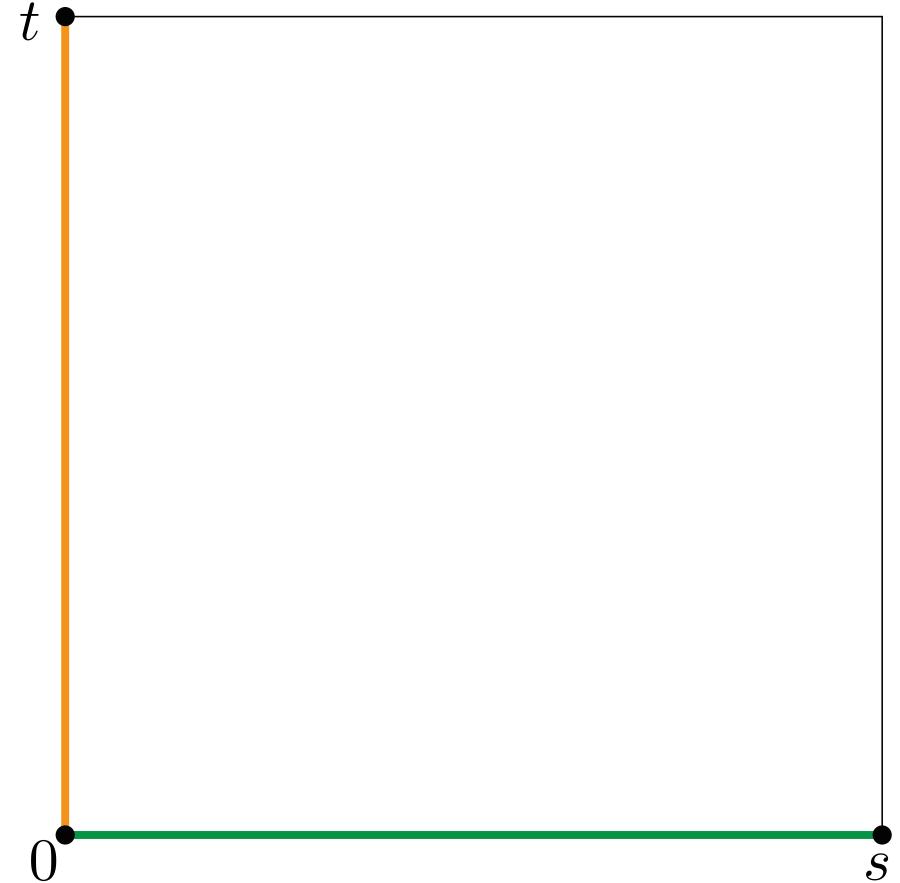


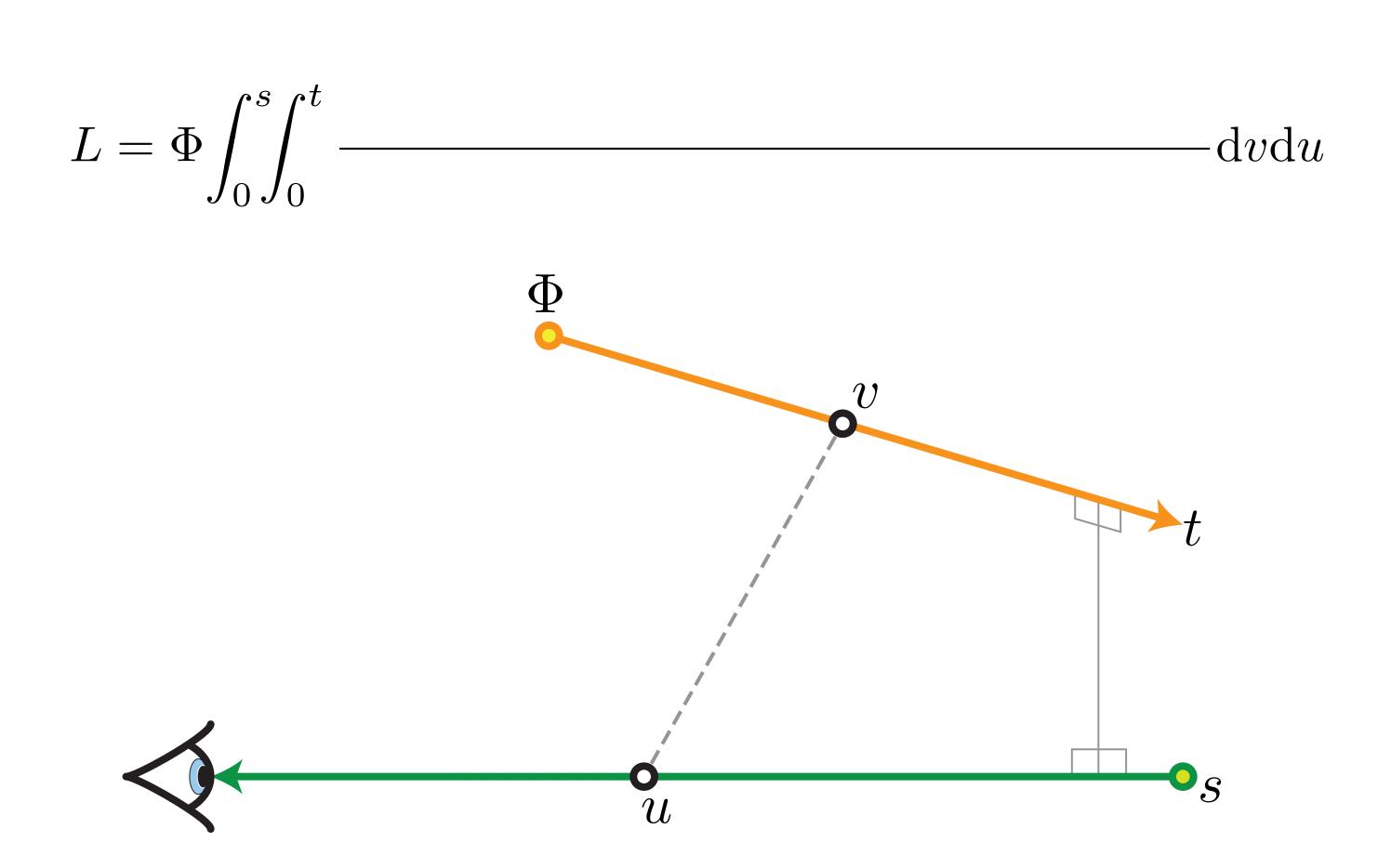


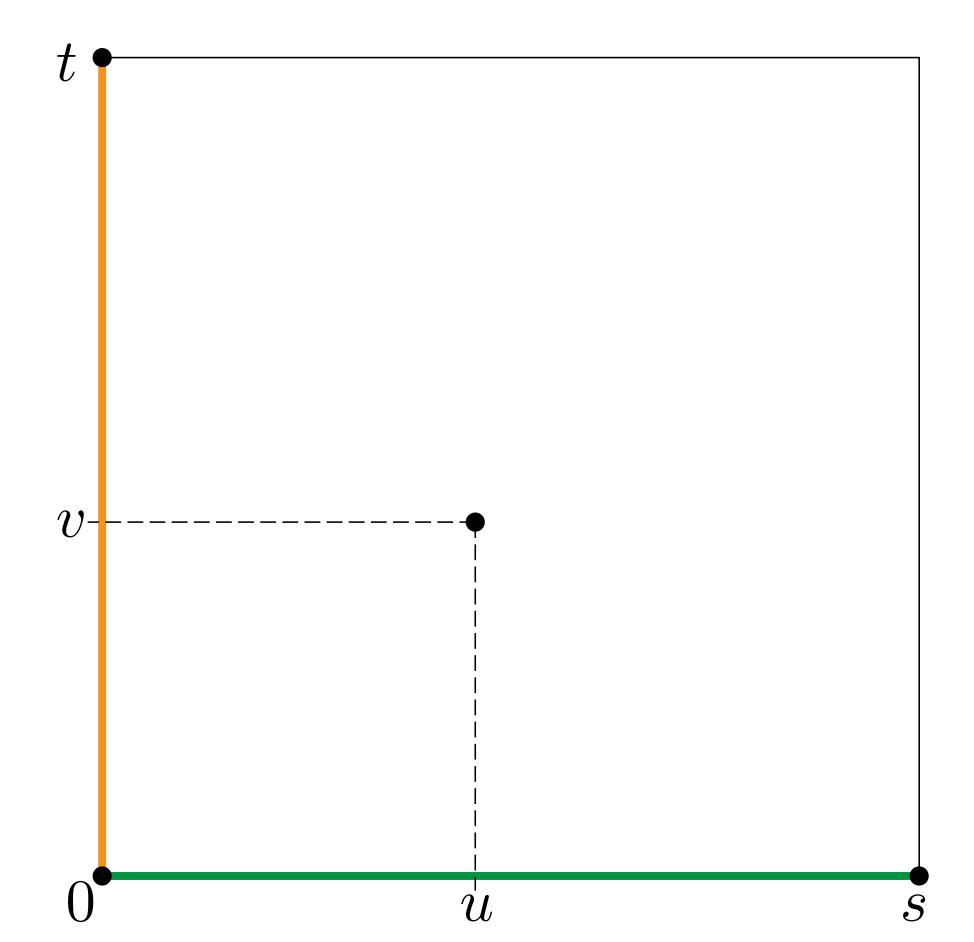


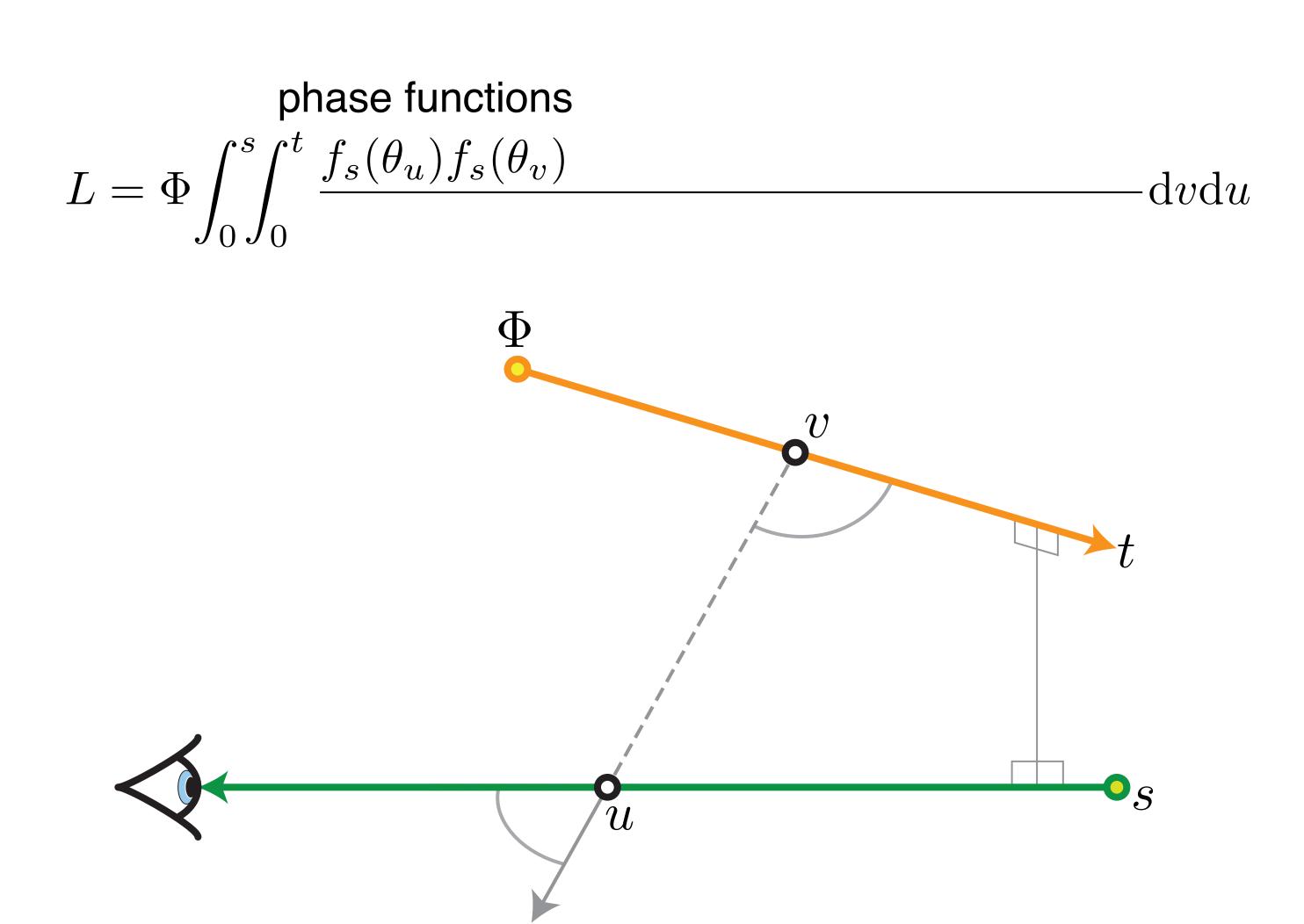


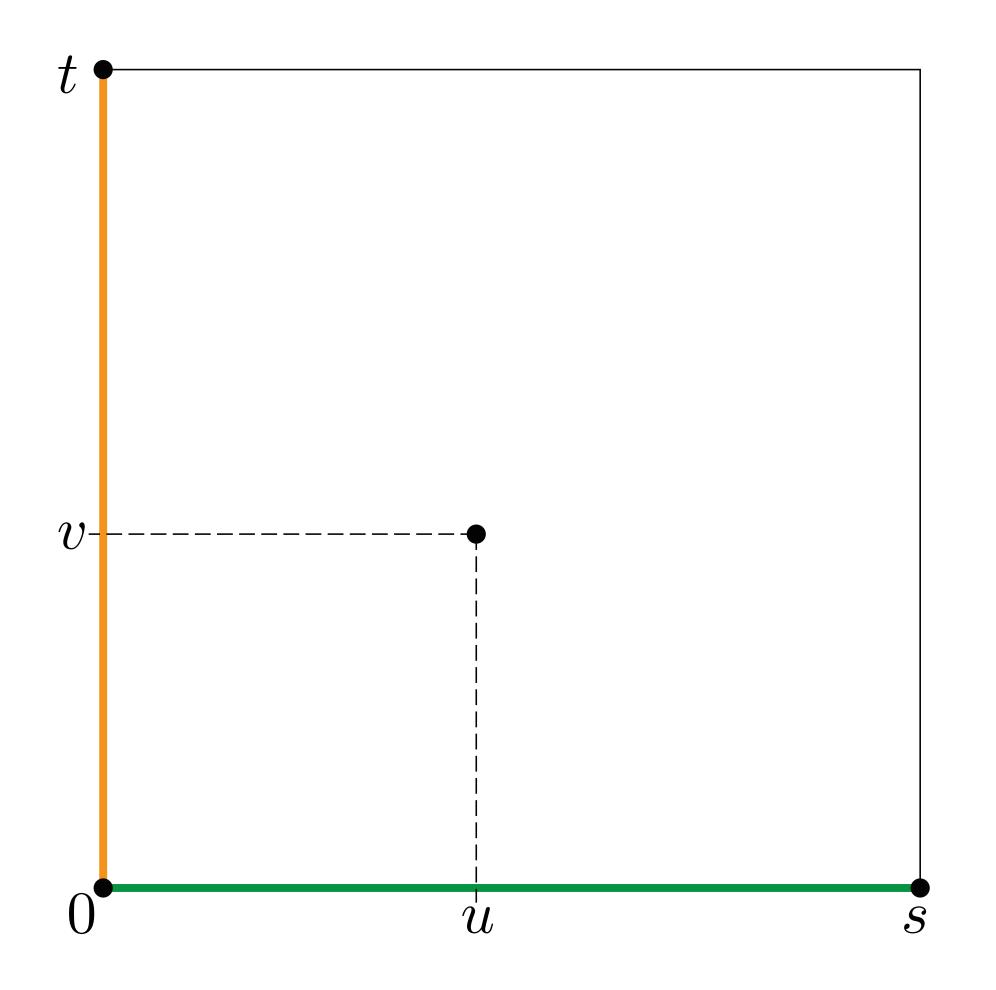


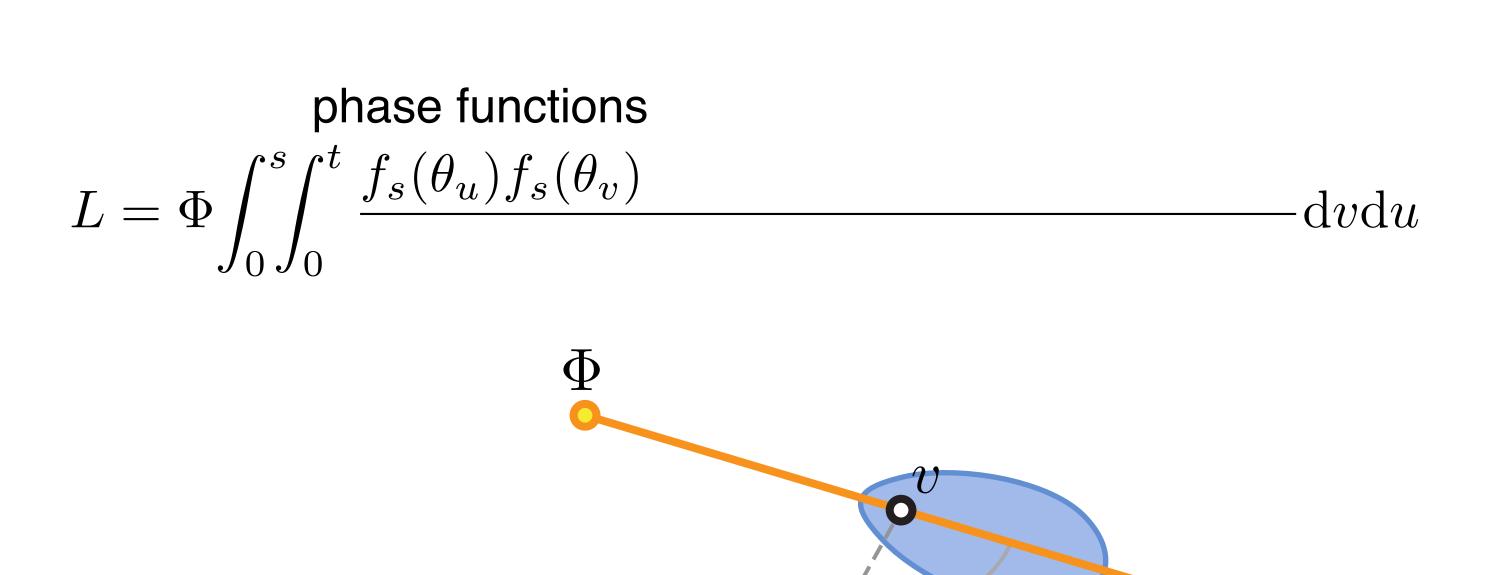


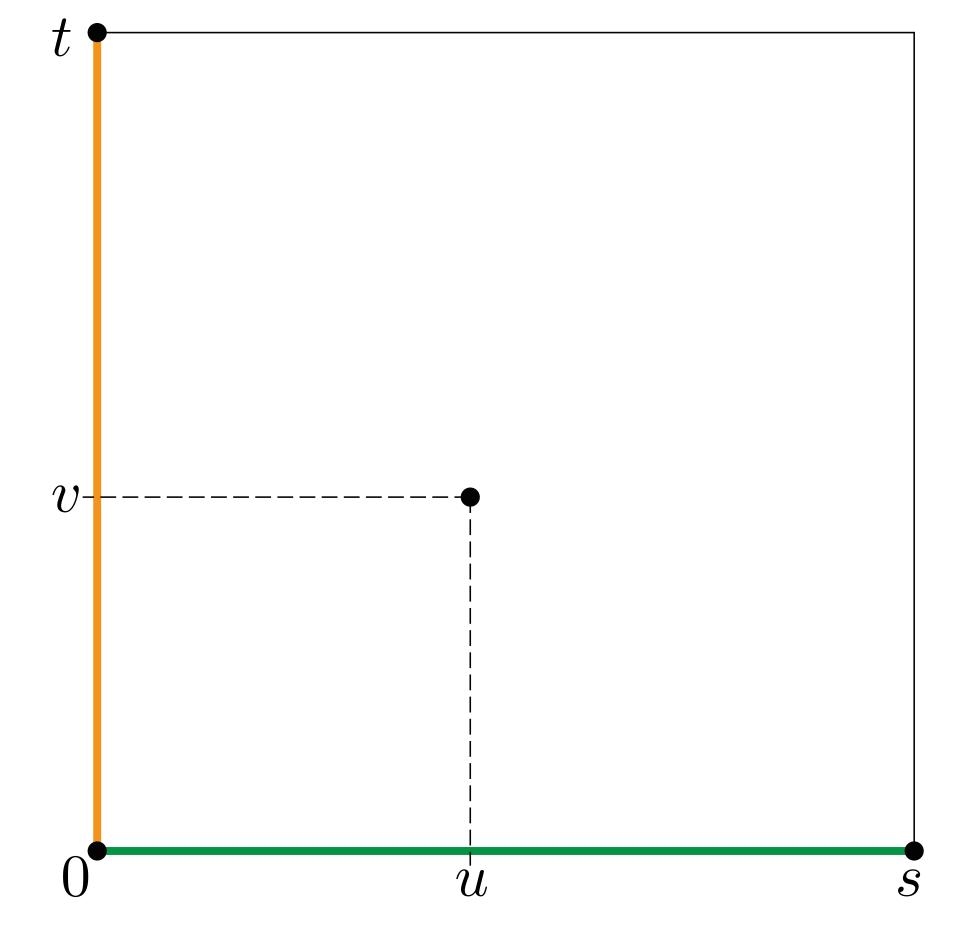


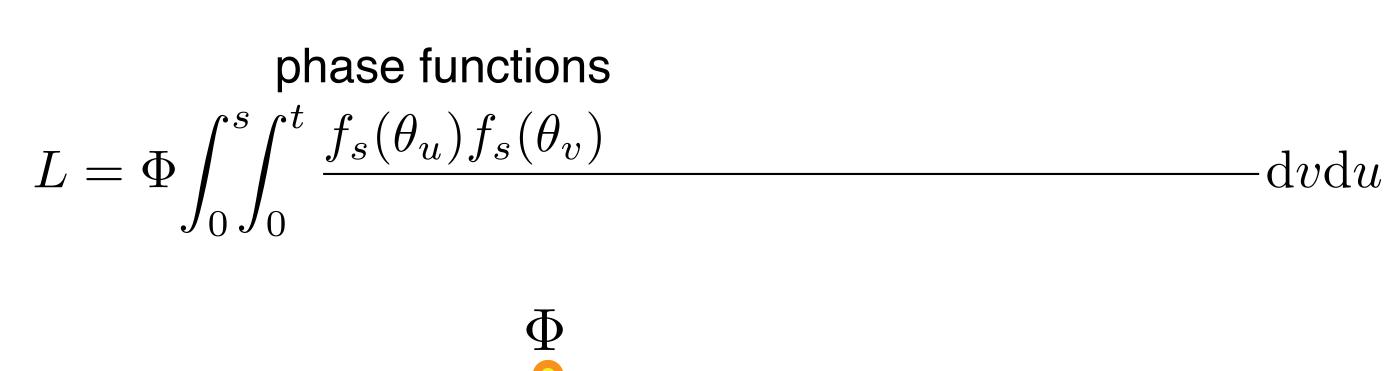


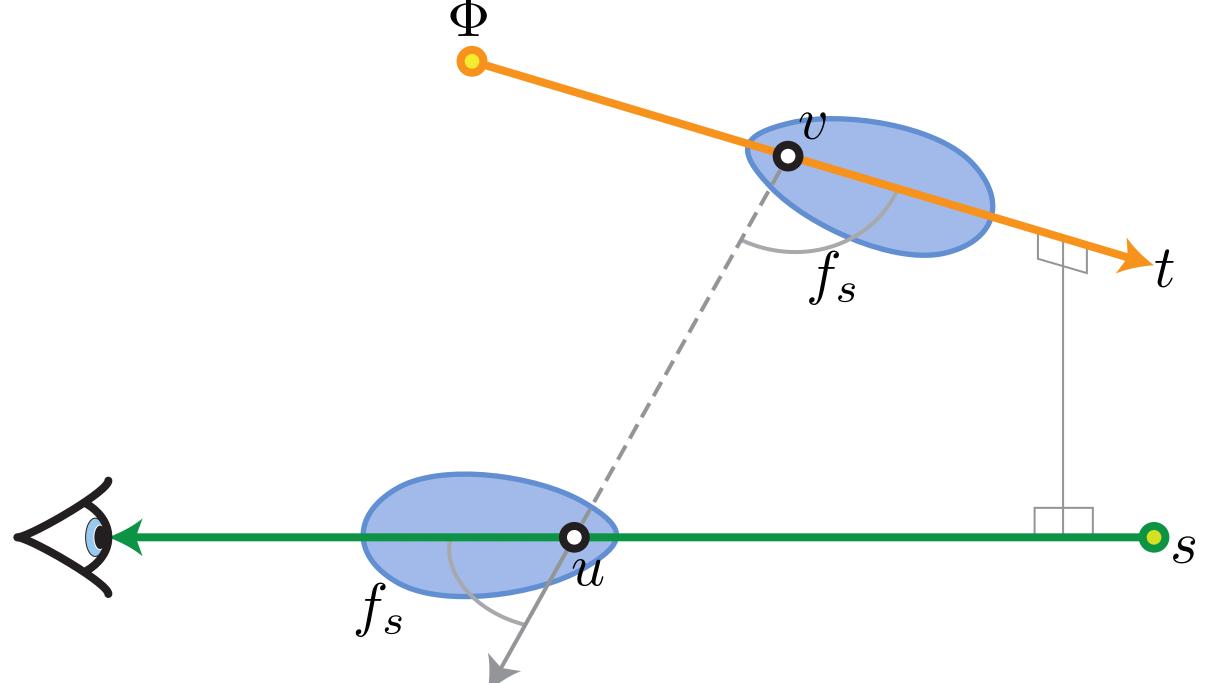


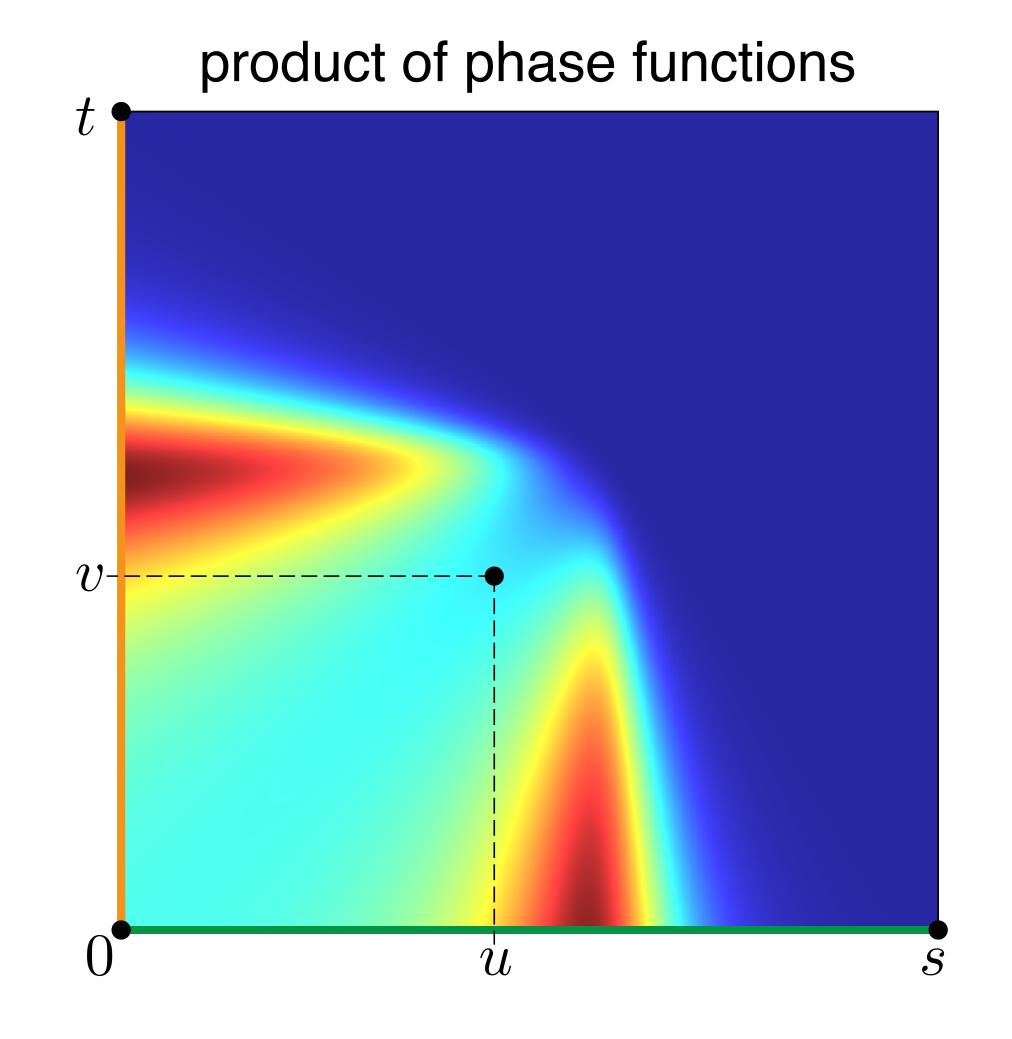




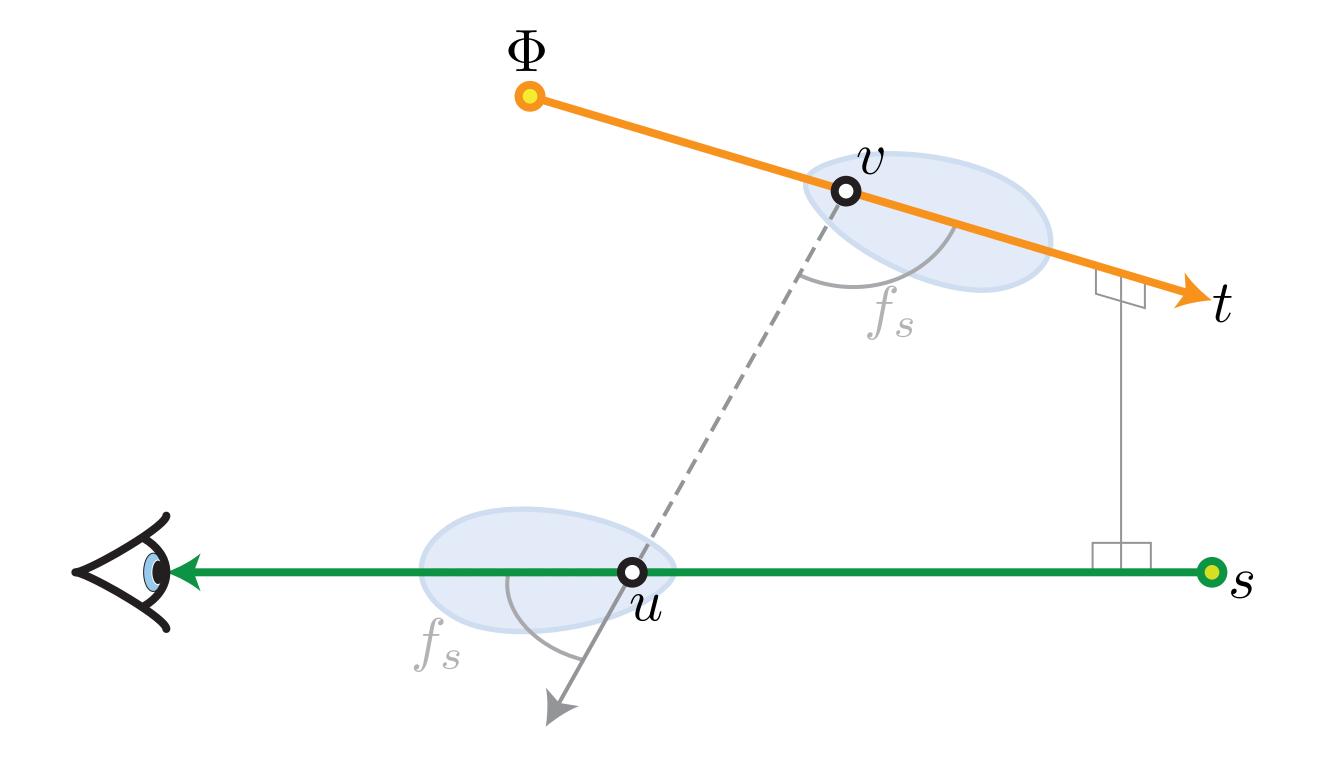


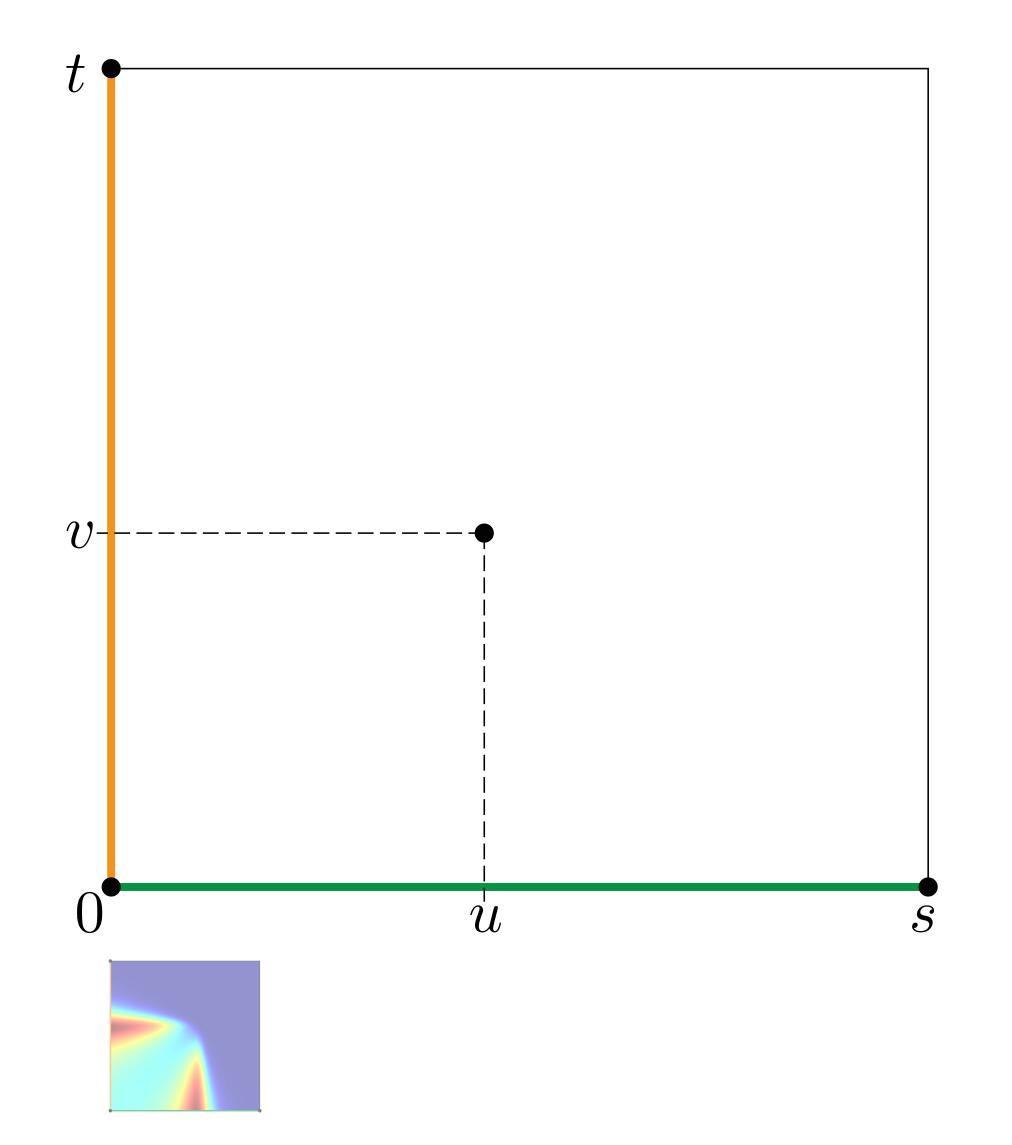


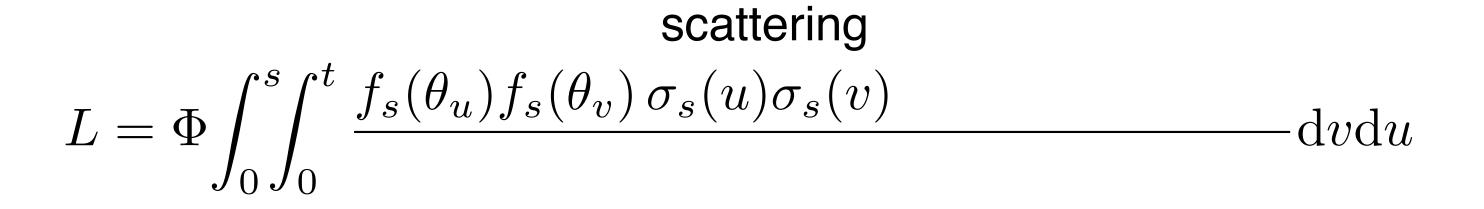


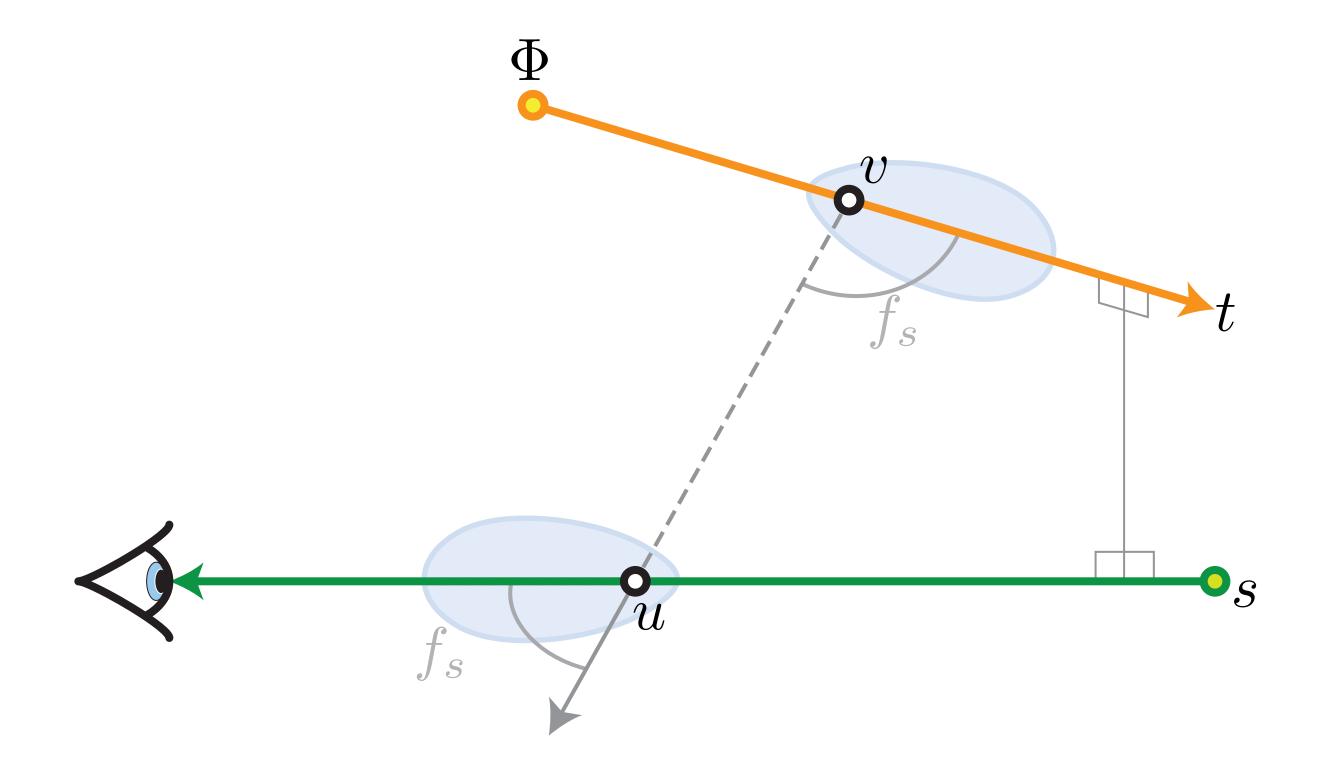


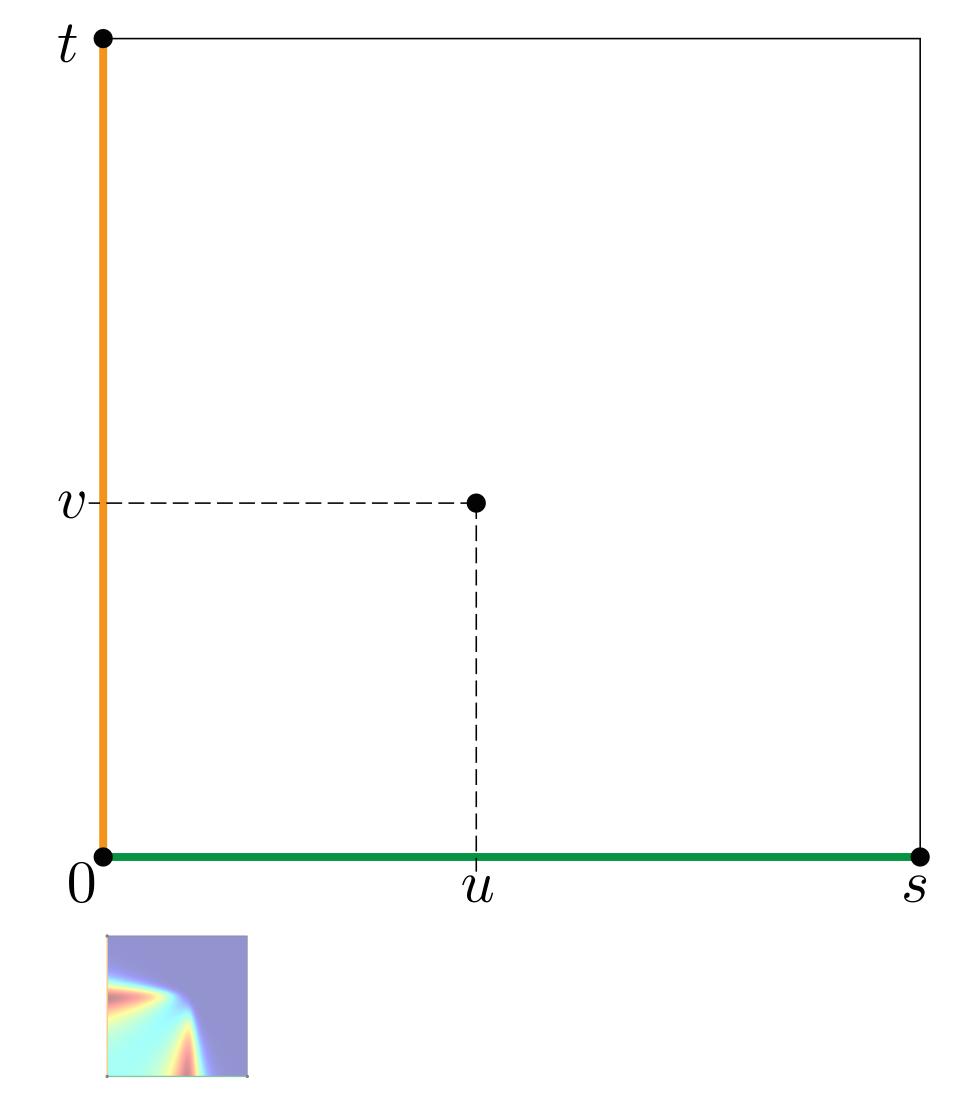
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v)}{\mathrm{d}v \mathrm{d}u} - \mathrm{d}v \mathrm{d}u$$

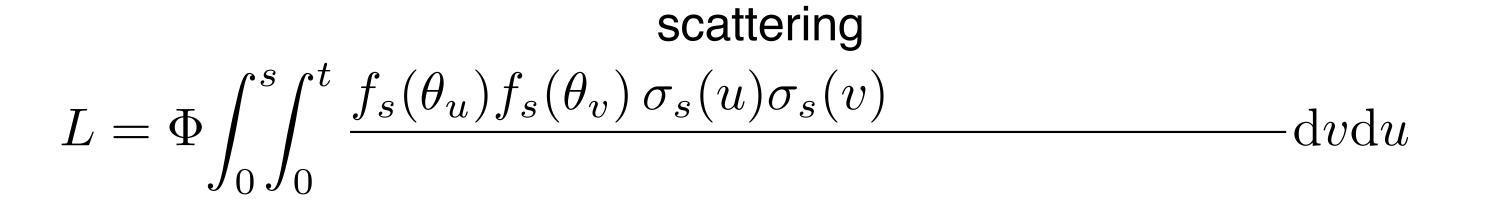


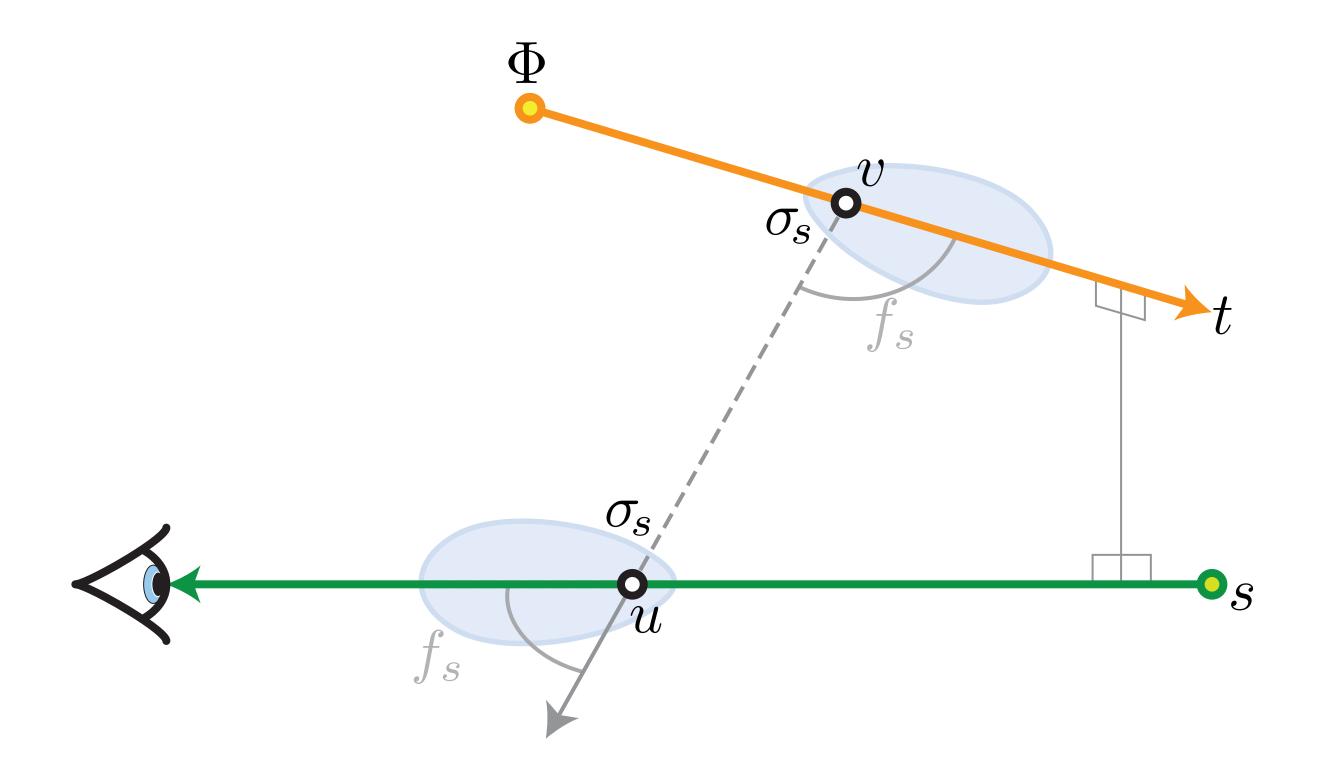


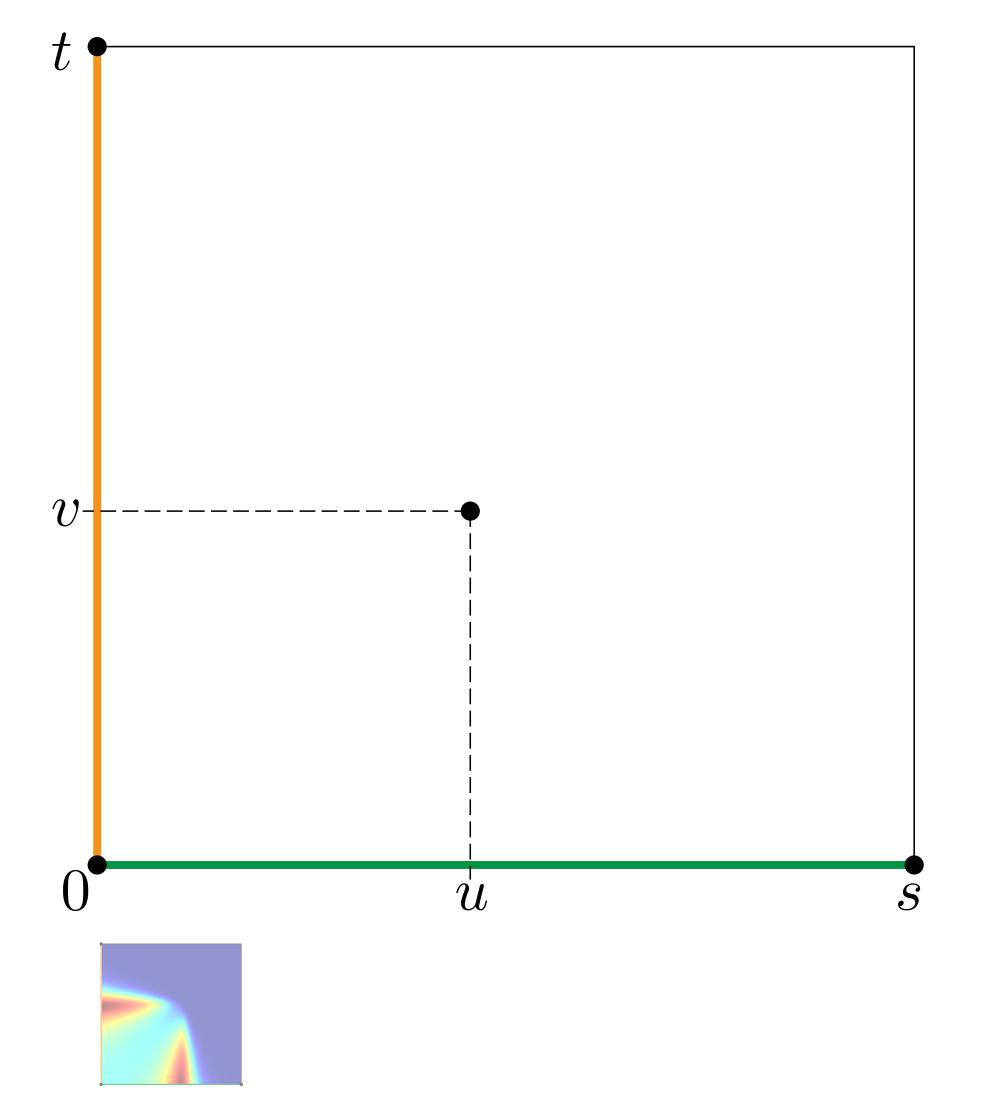




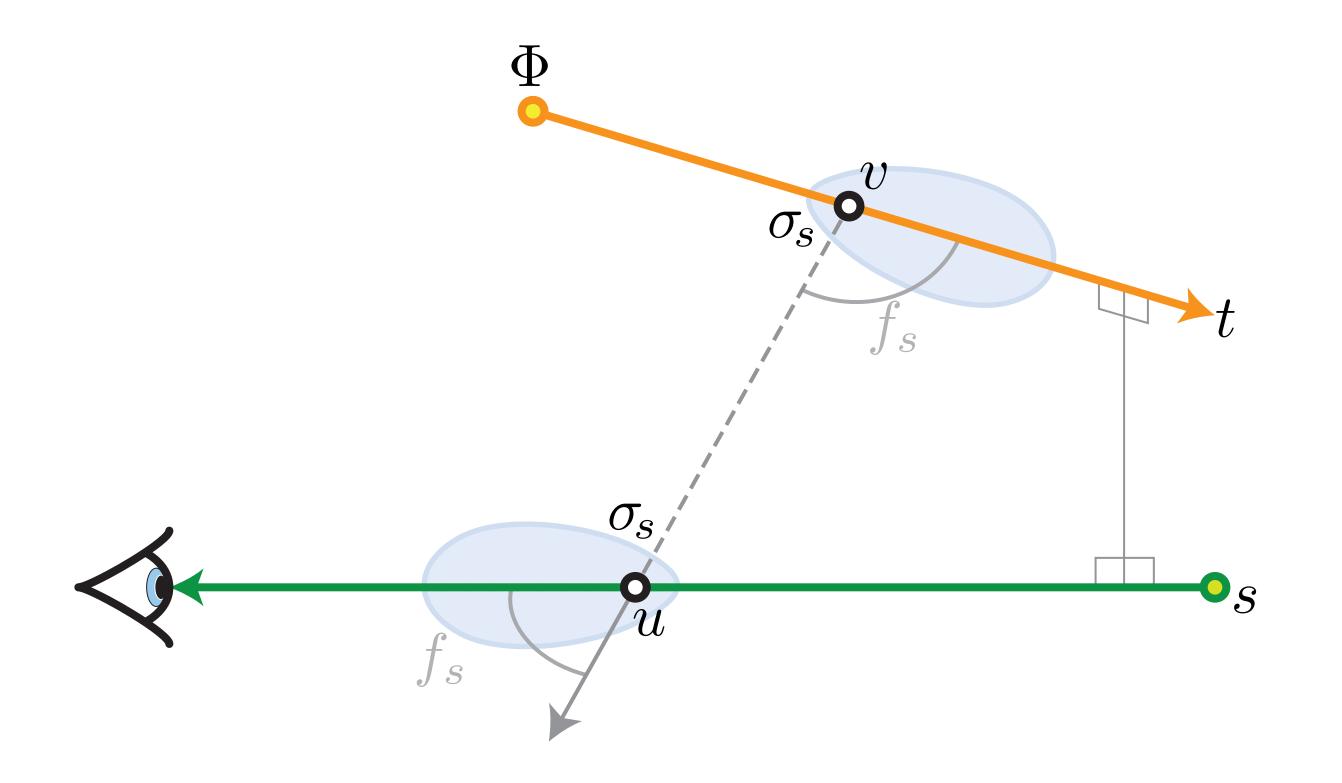


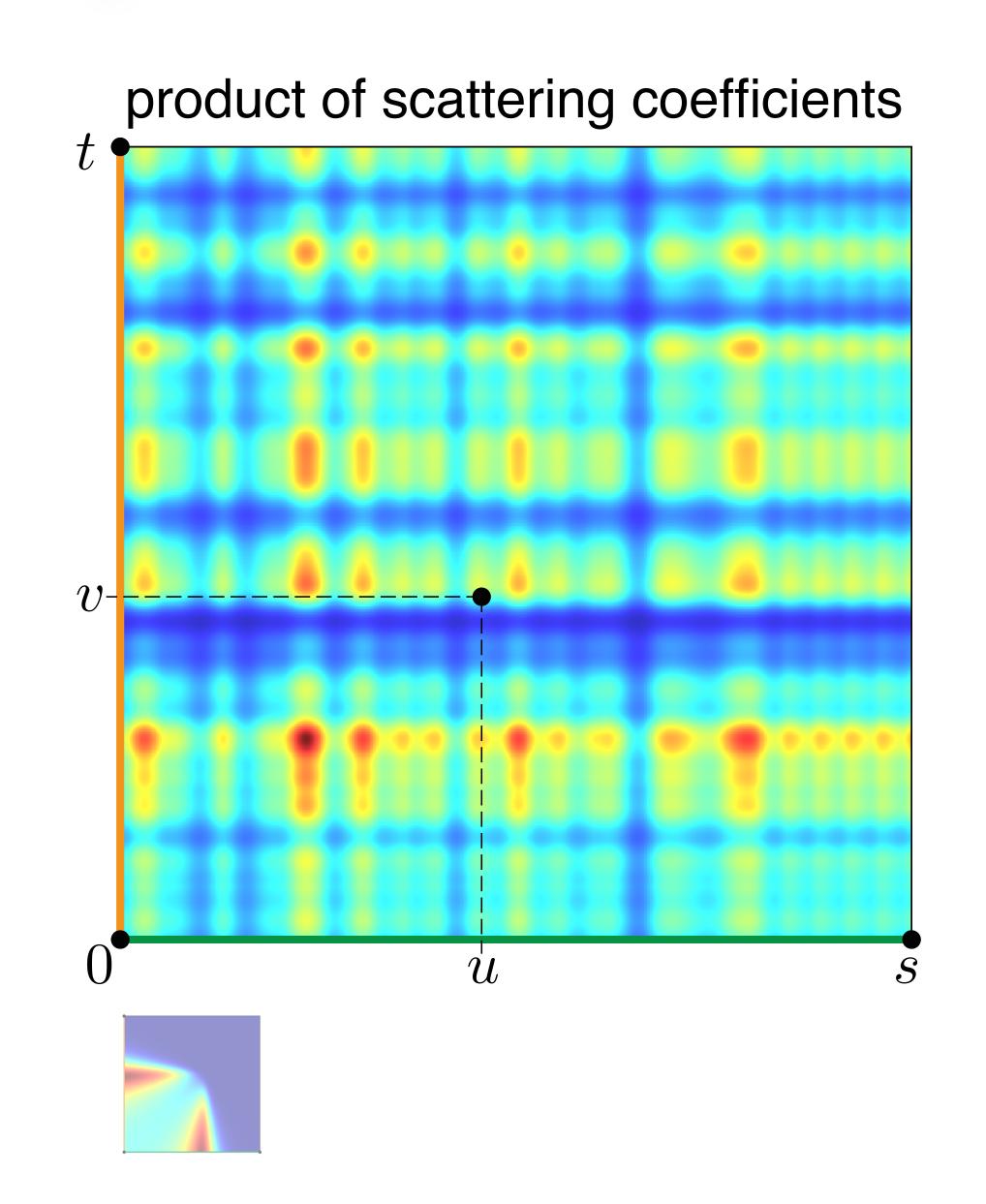




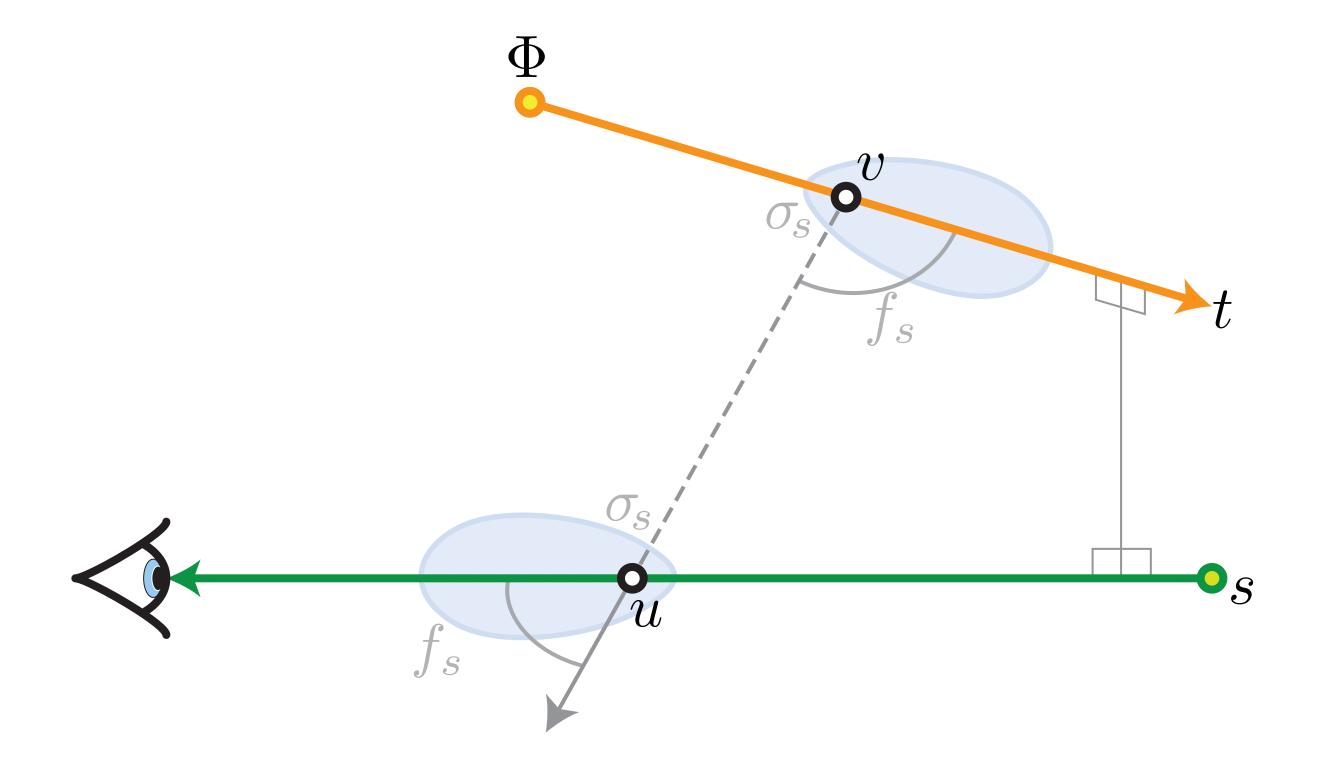


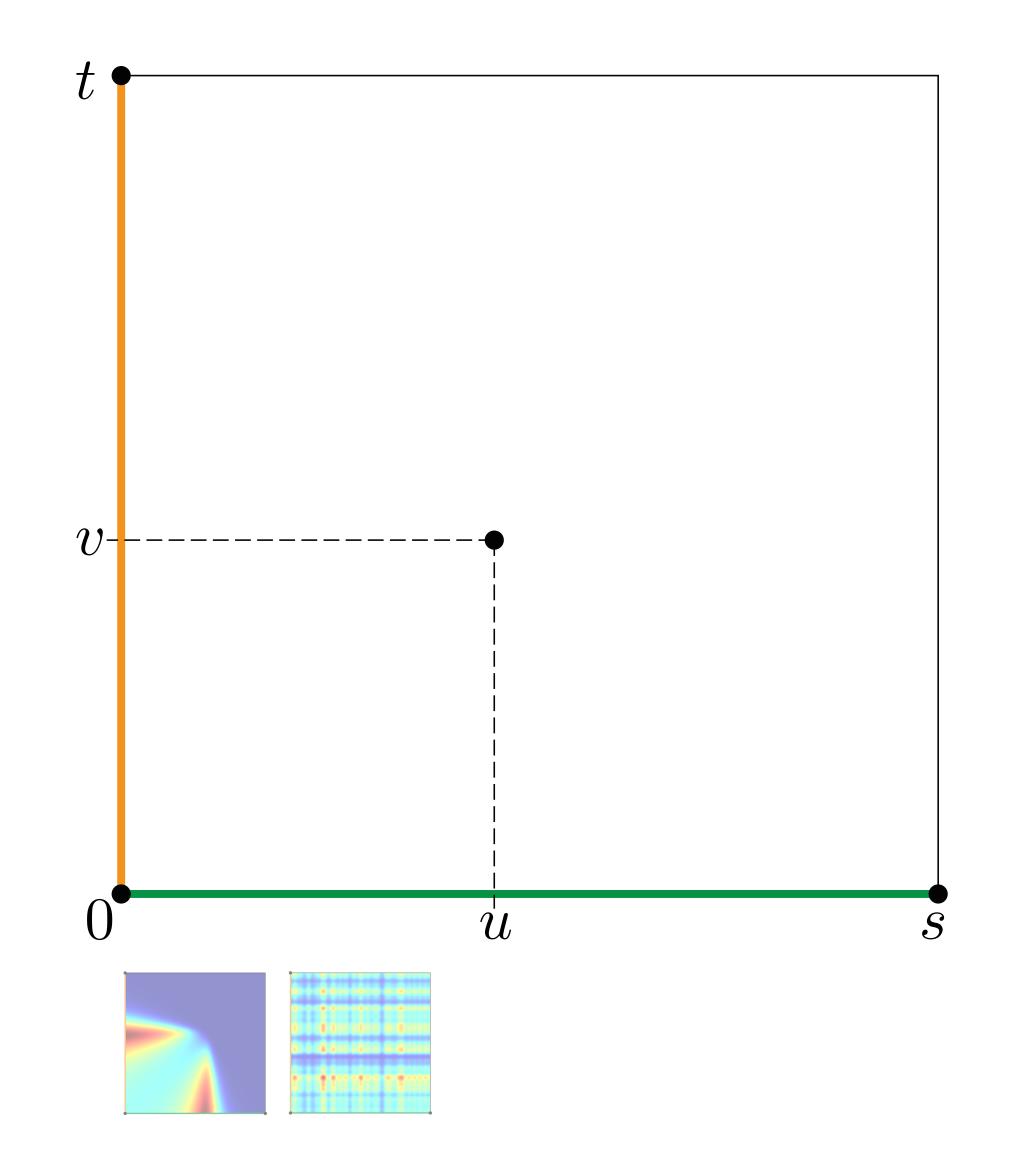
scattering
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \, \sigma_s(u) \sigma_s(v)}{\mathrm{d}v \mathrm{d}u} \mathrm{d}v \mathrm{d}u$$





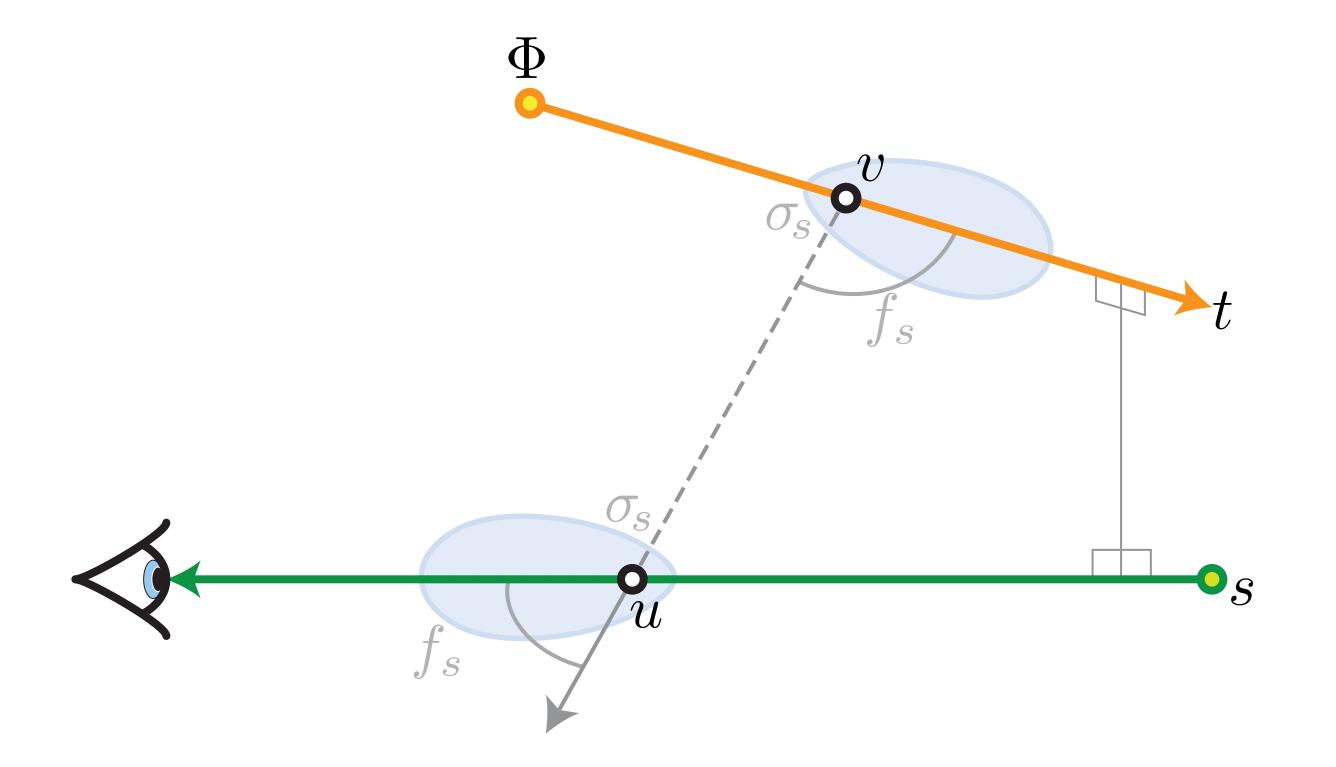
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v)}{dv du} dv du$$

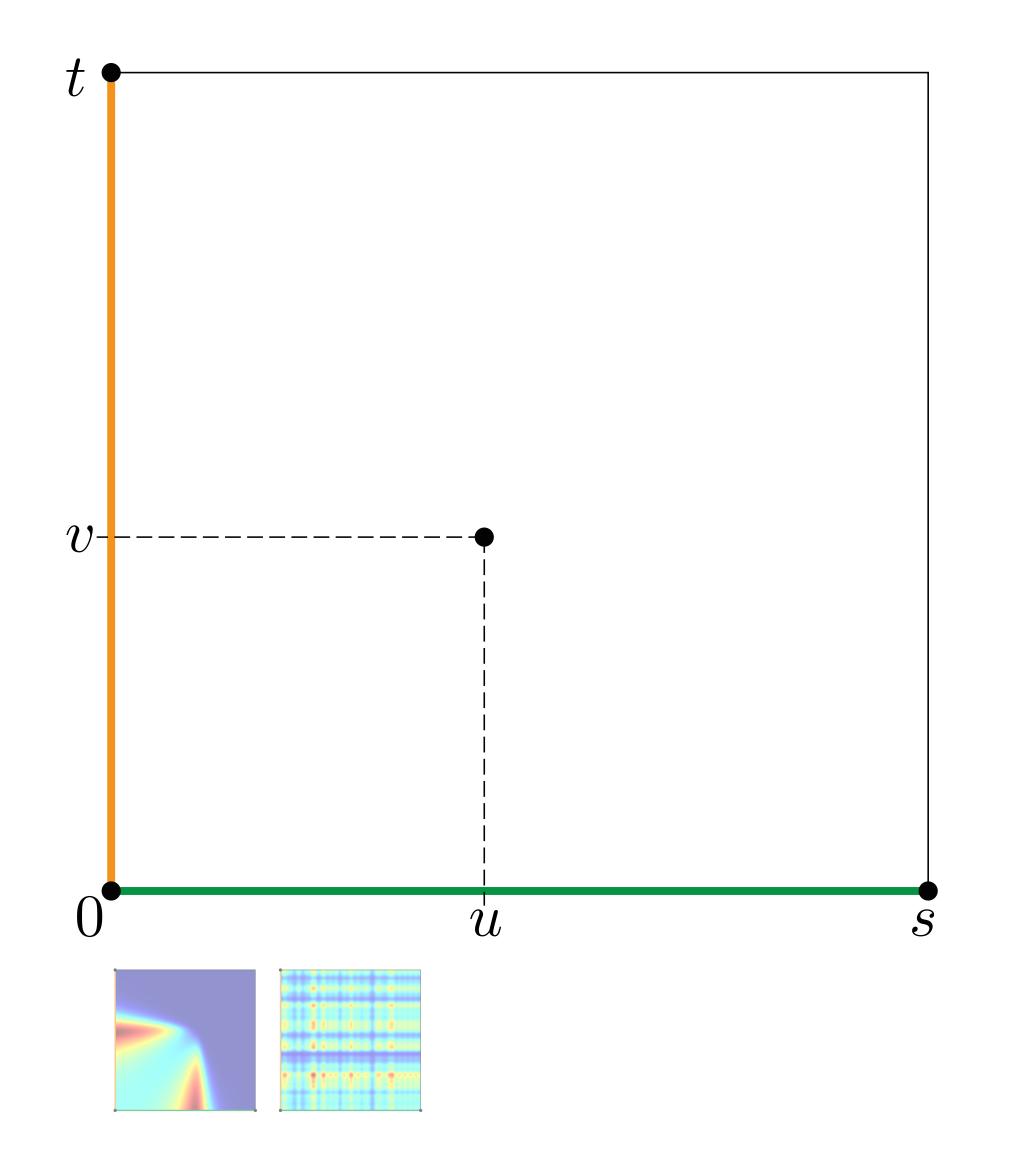




transmittance

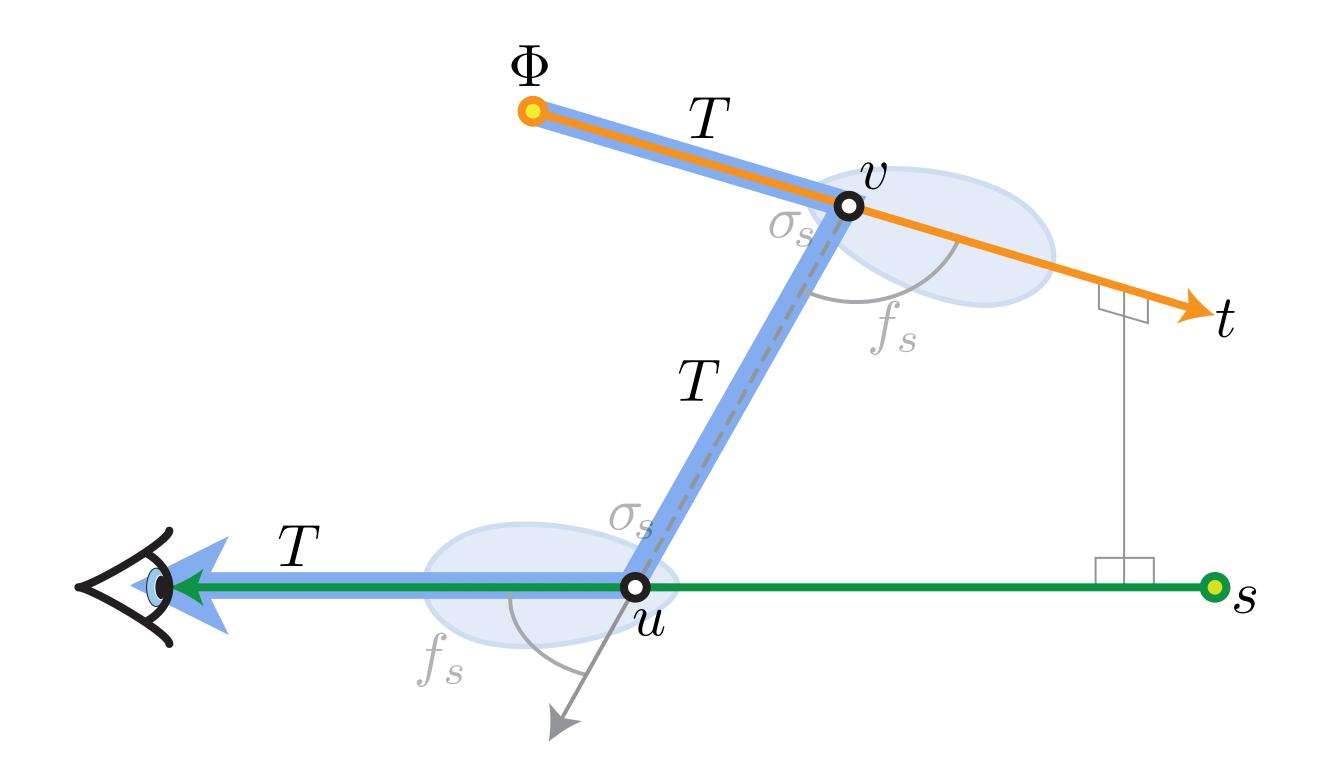
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w)}{\mathrm{d}v \mathrm{d}u}$$

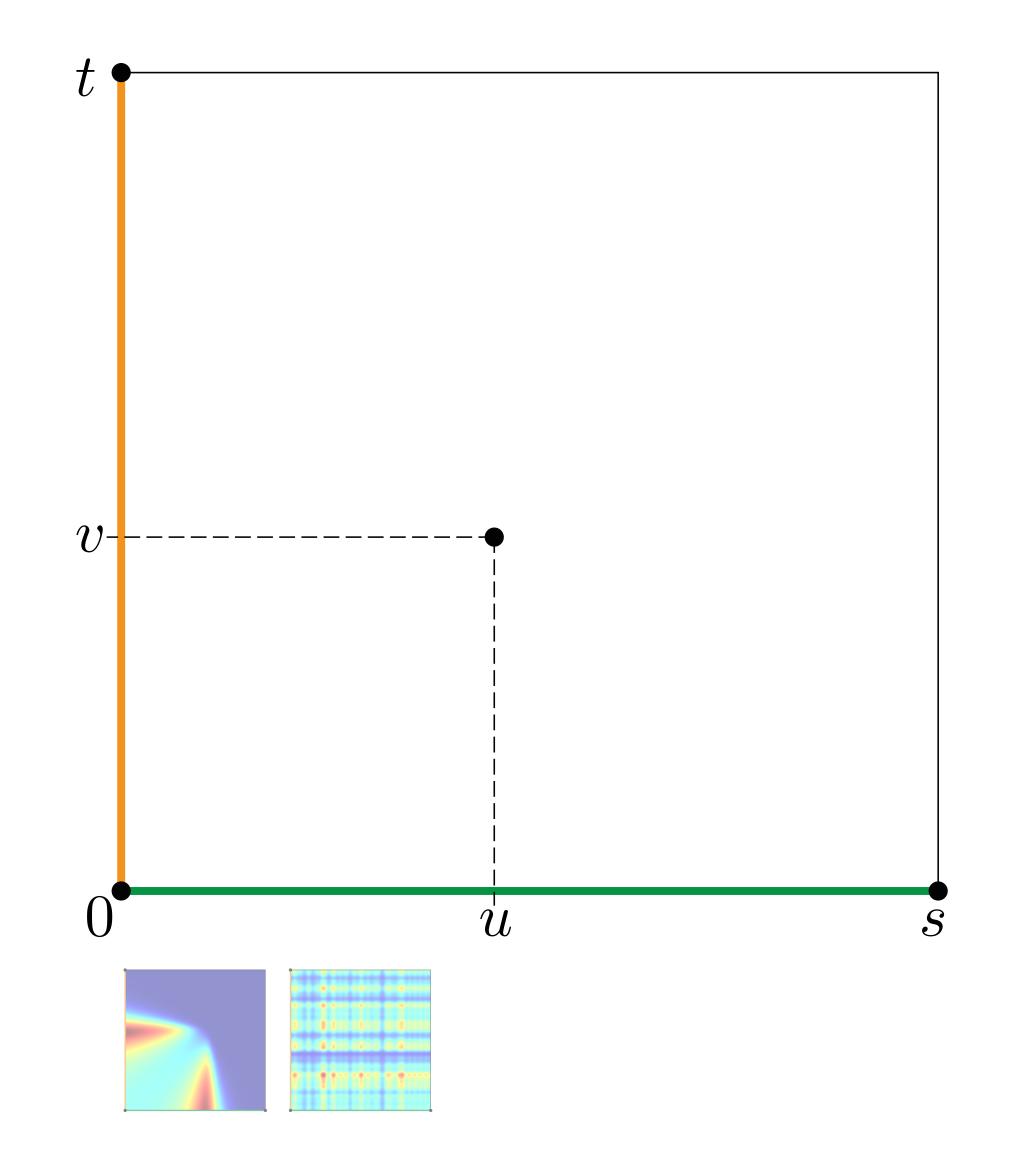




transmittance

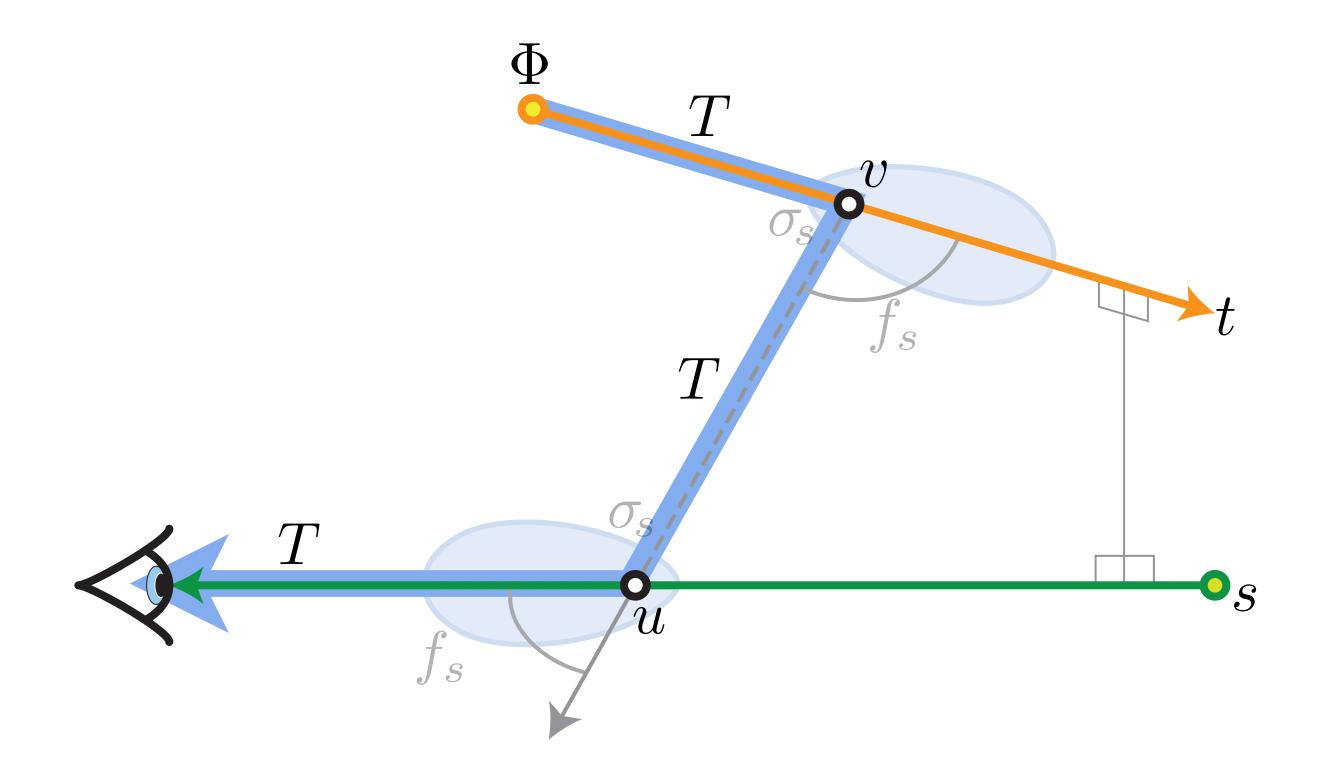
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w)}{\mathrm{d}v \mathrm{d}u}$$

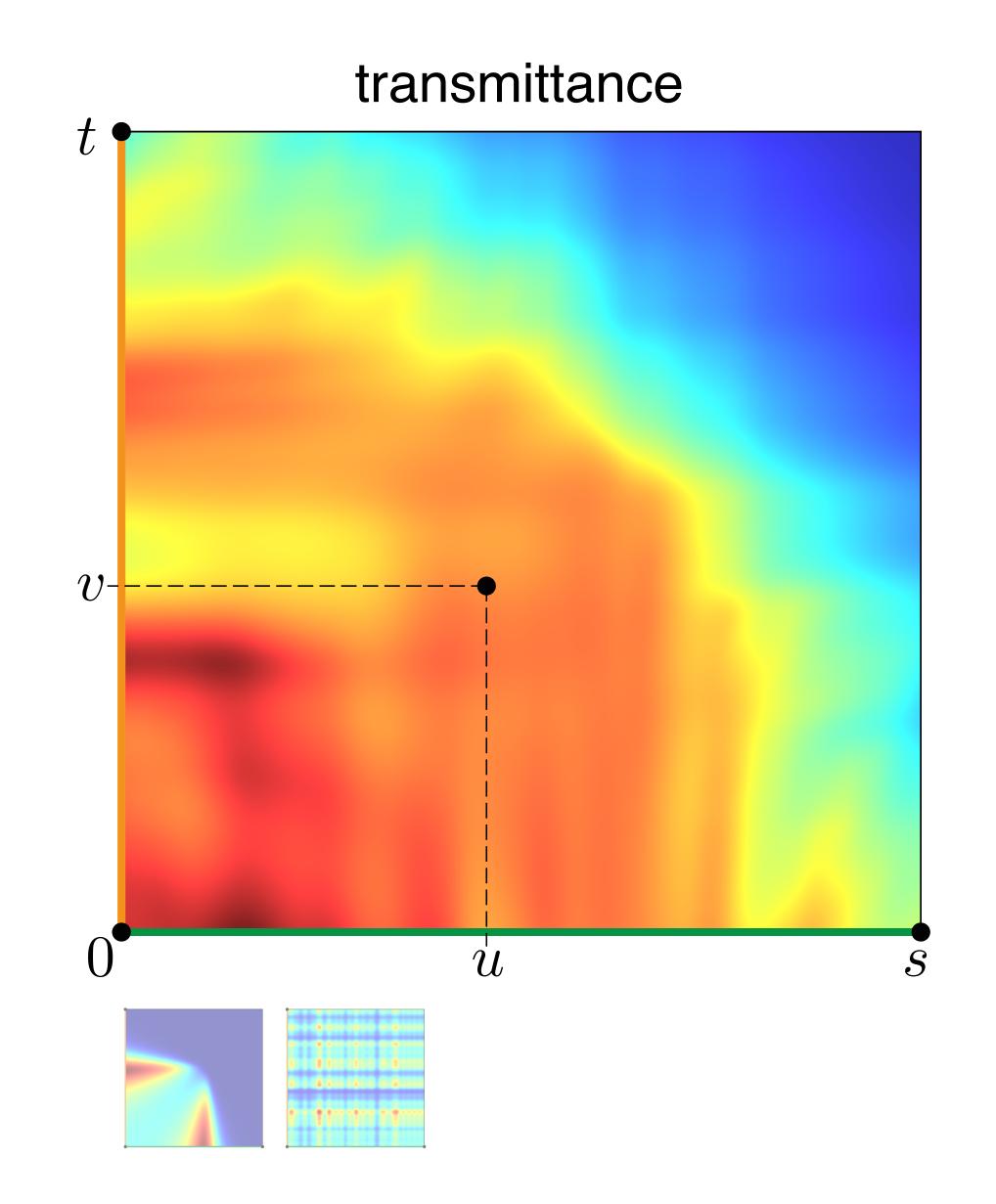




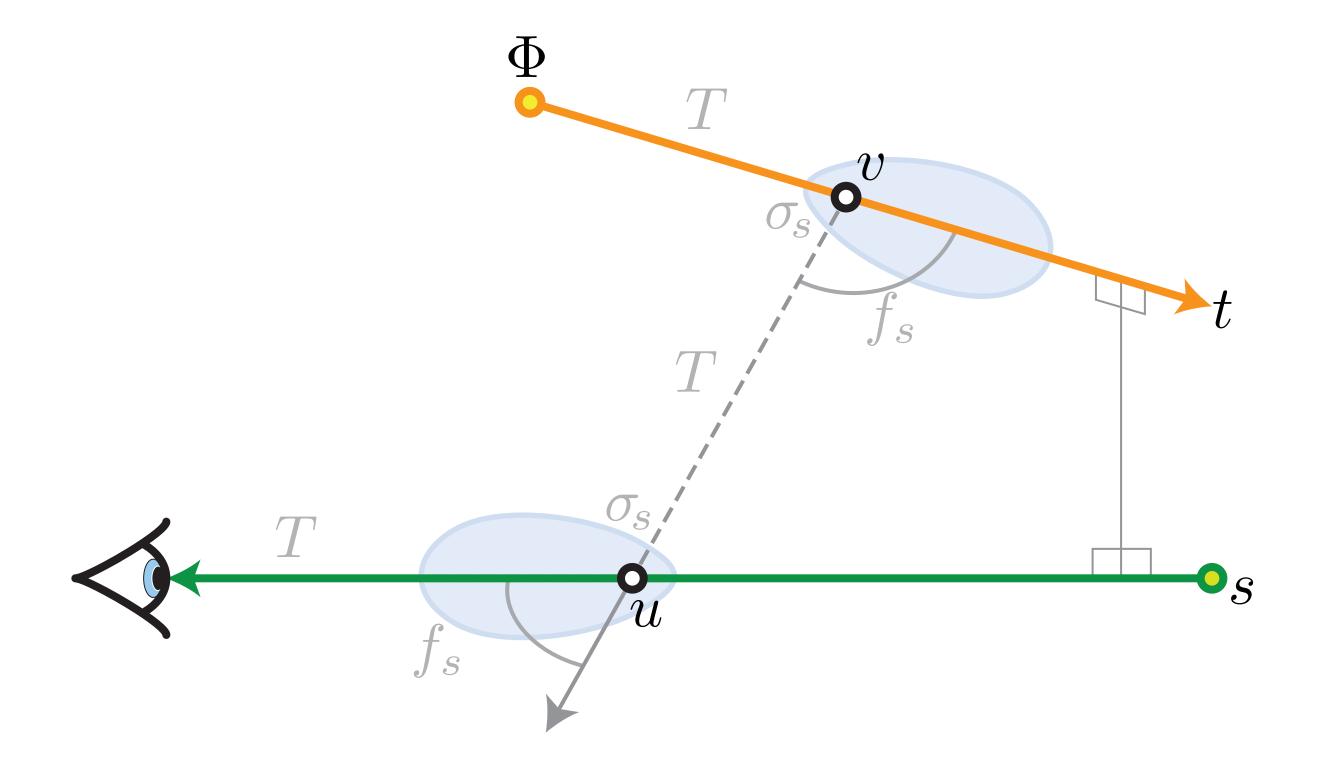
transmittance

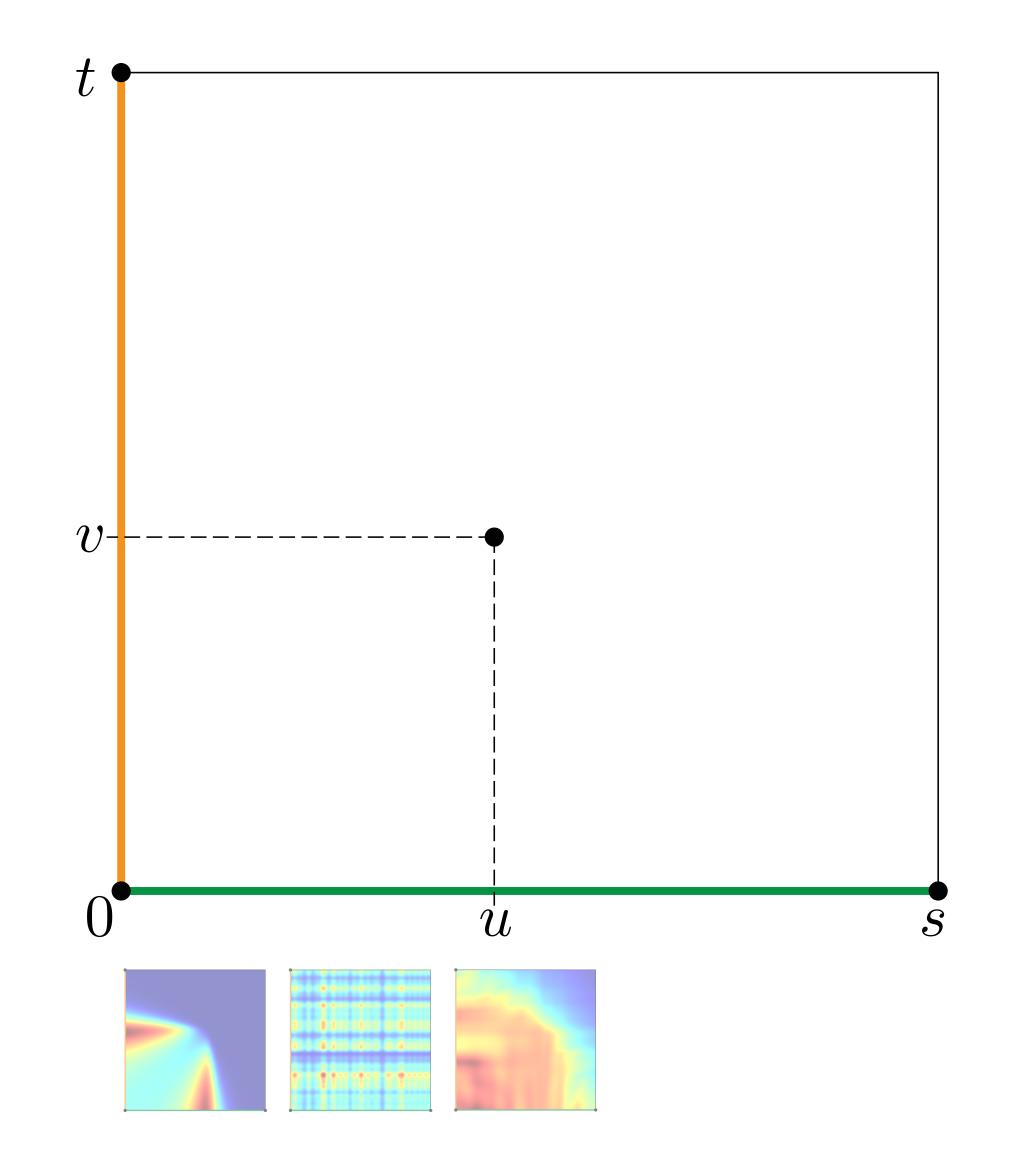
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w)}{dv du} dv du$$



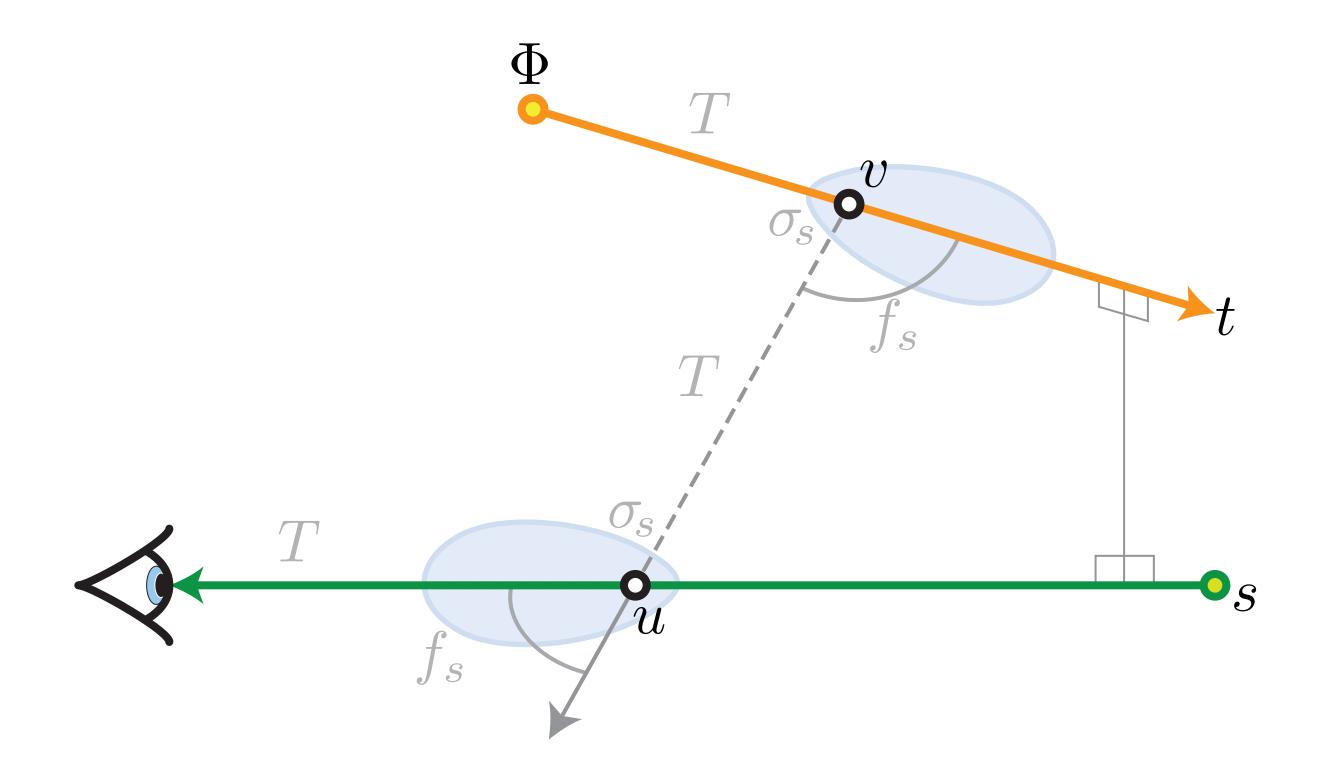


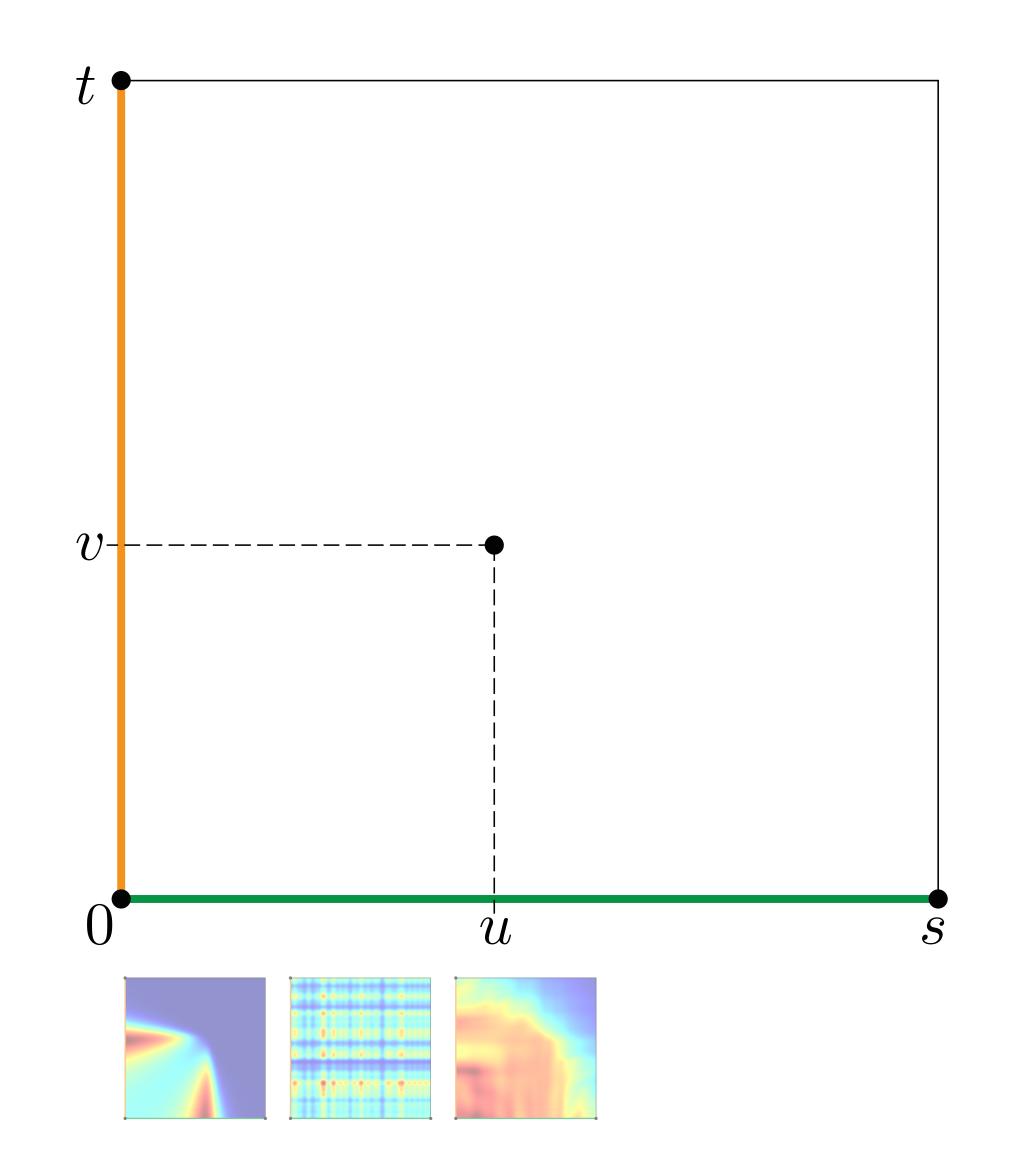
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w)}{\mathrm{d}v \mathrm{d}u}$$



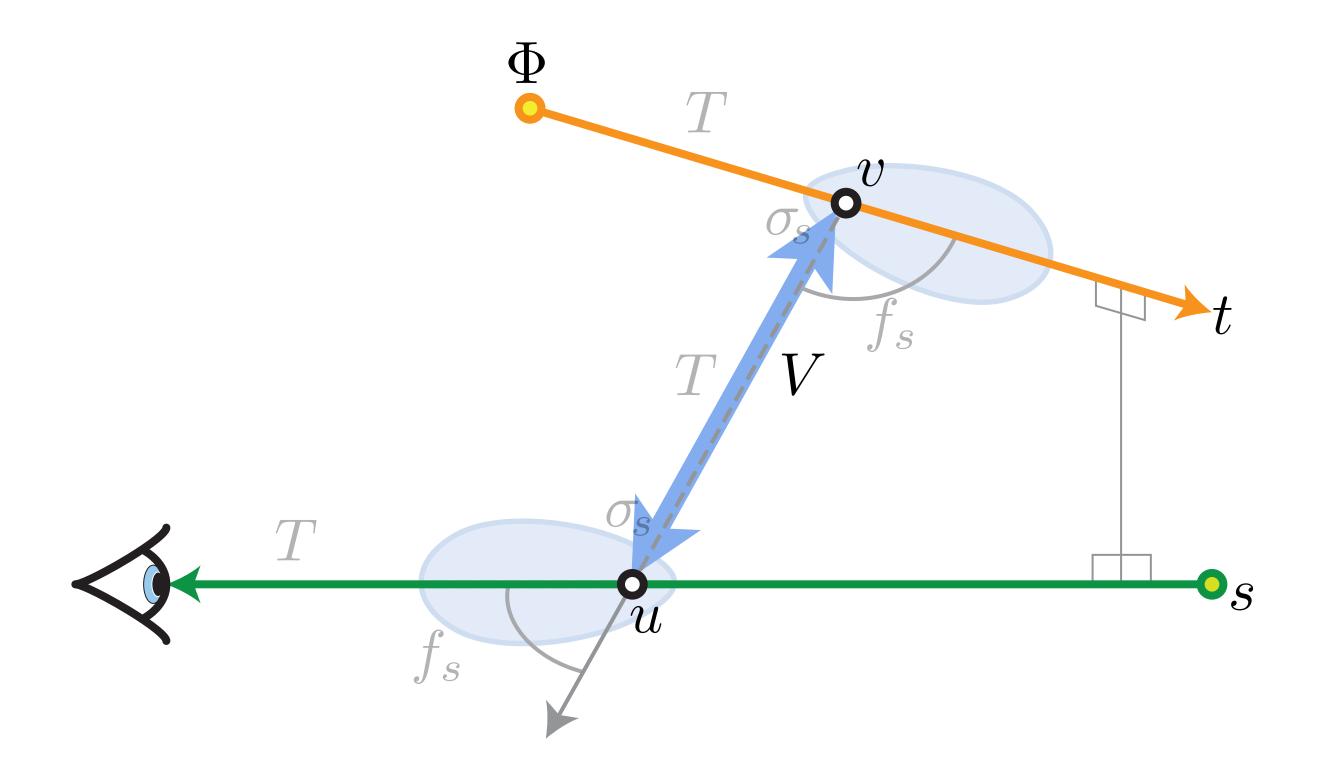


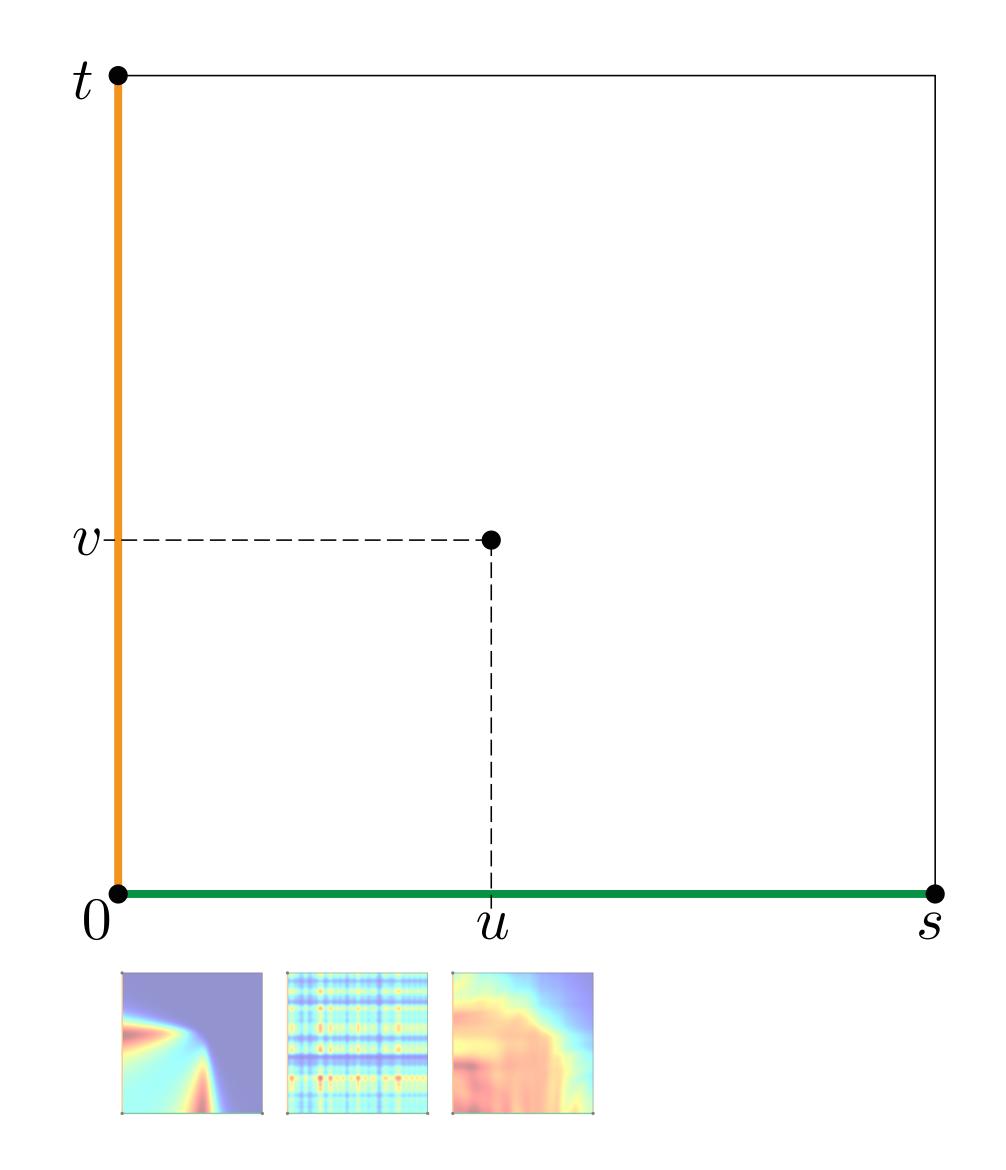
visibility
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \, \sigma_s(u) \sigma_s(v) \, T(u) T(v) T(w) \, V}{\mathrm{d}v \mathrm{d}u}$$



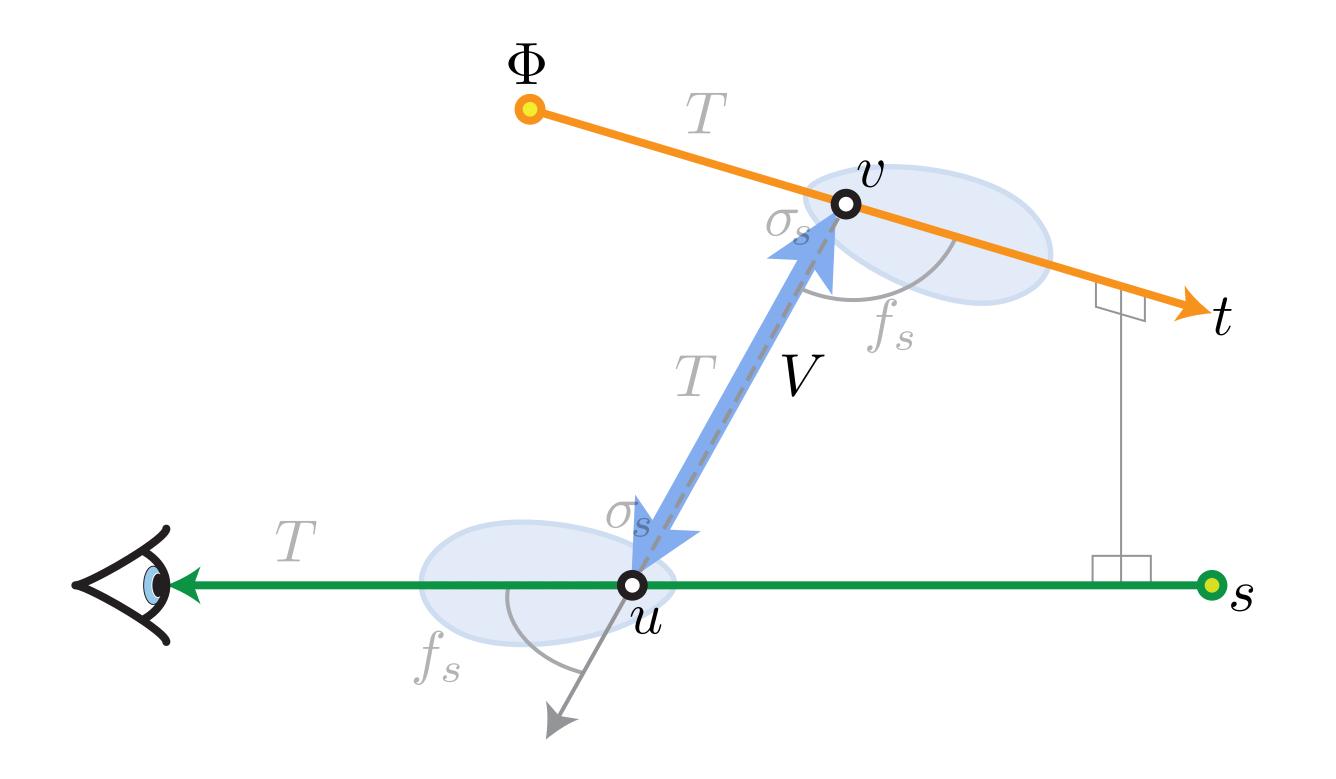


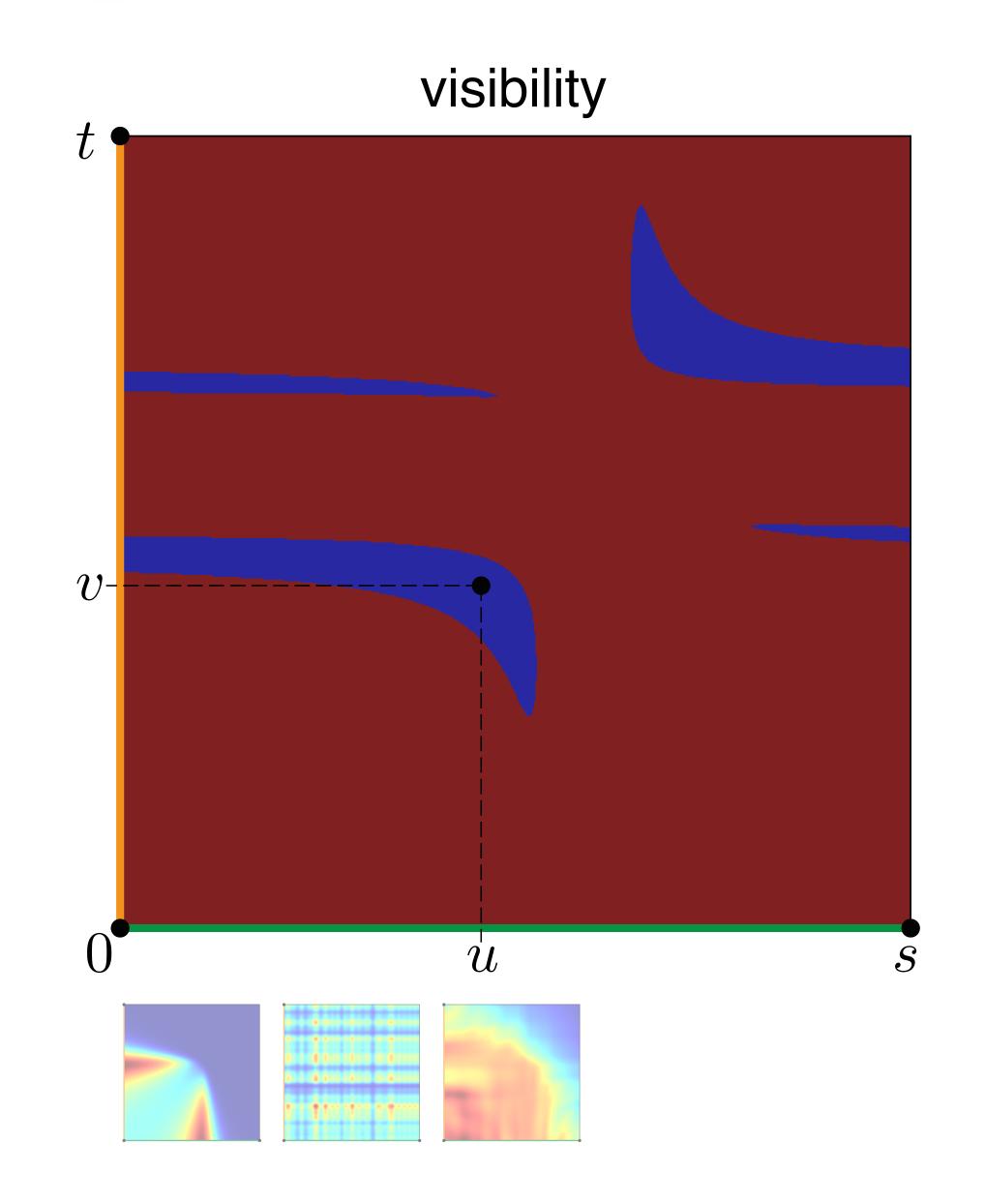
visibility
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \, \sigma_s(u) \sigma_s(v) \, T(u) T(v) T(w) \, V}{\mathrm{d}v \mathrm{d}u}$$



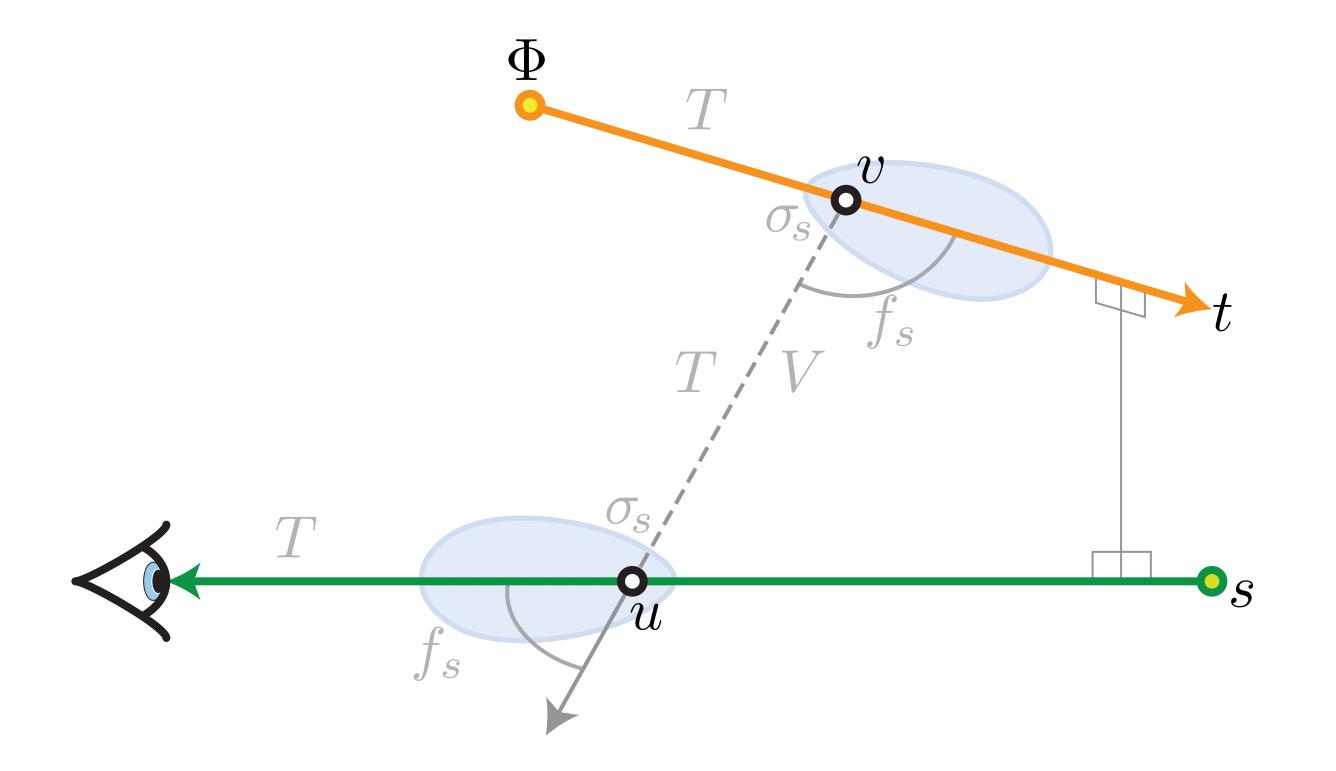


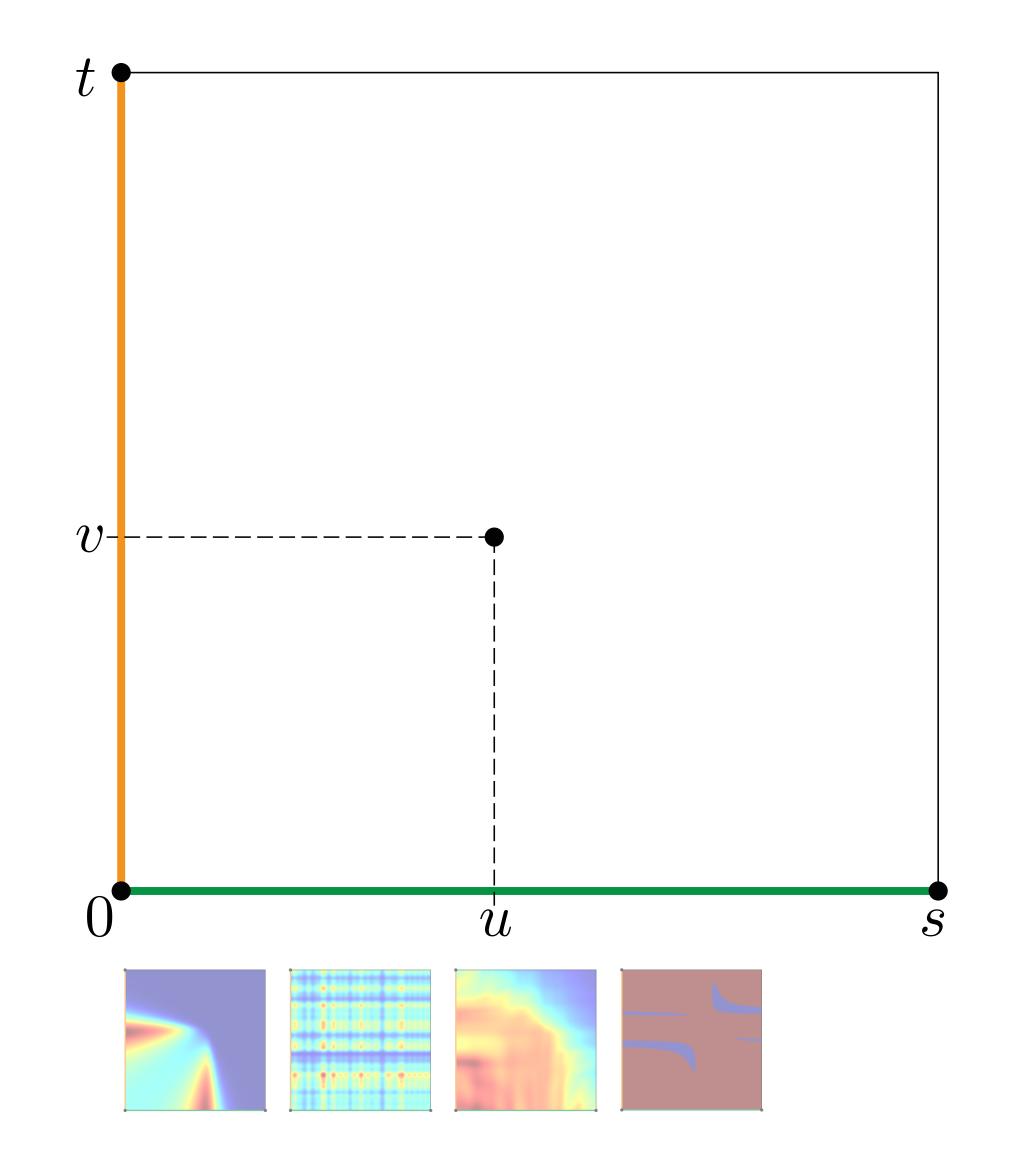
visibility
$$L = \Phi \int_0^s \!\! \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \, \sigma_s(u) \sigma_s(v) \, T(u) T(v) T(w) \, V}{\mathrm{d}v \mathrm{d}u}$$

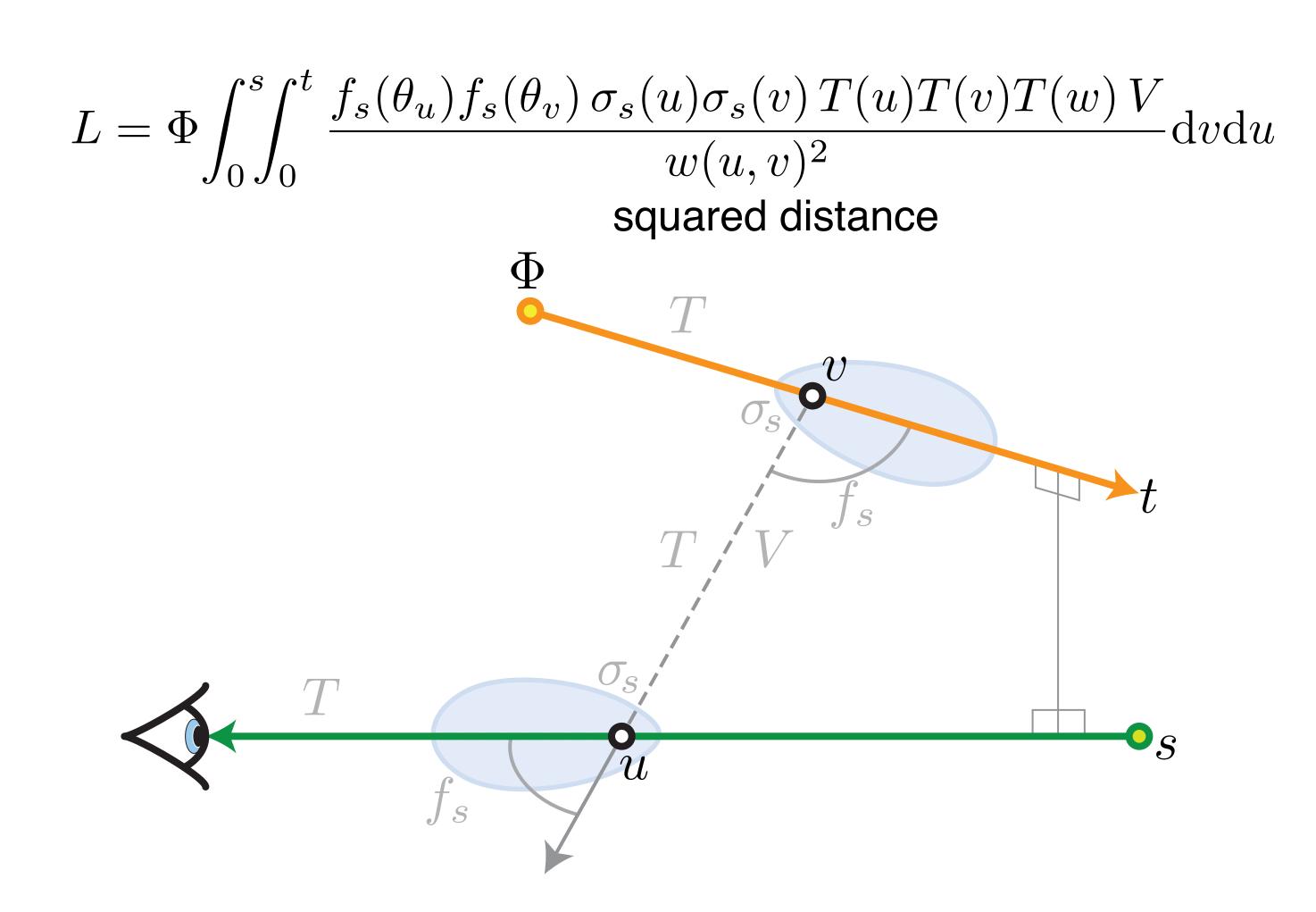


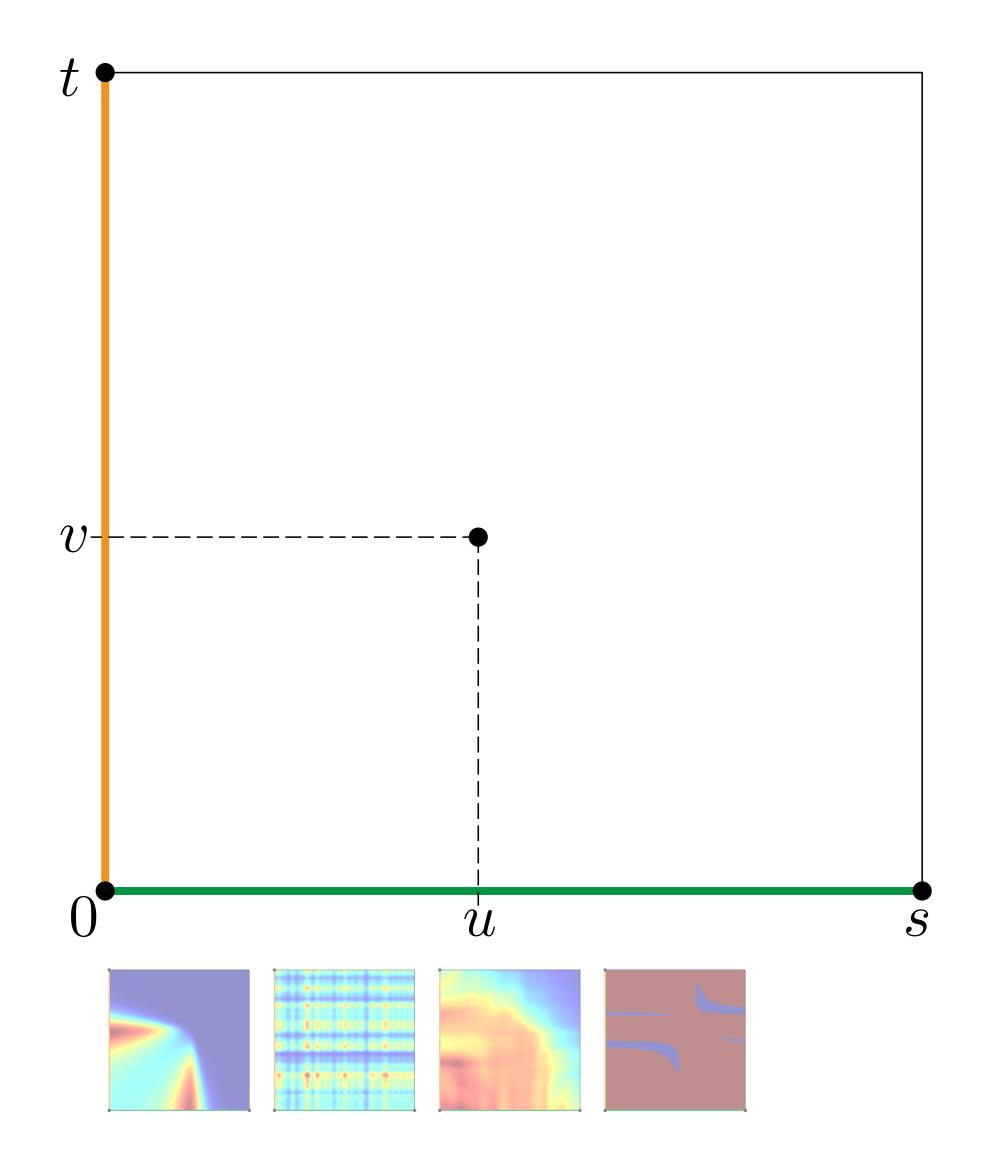


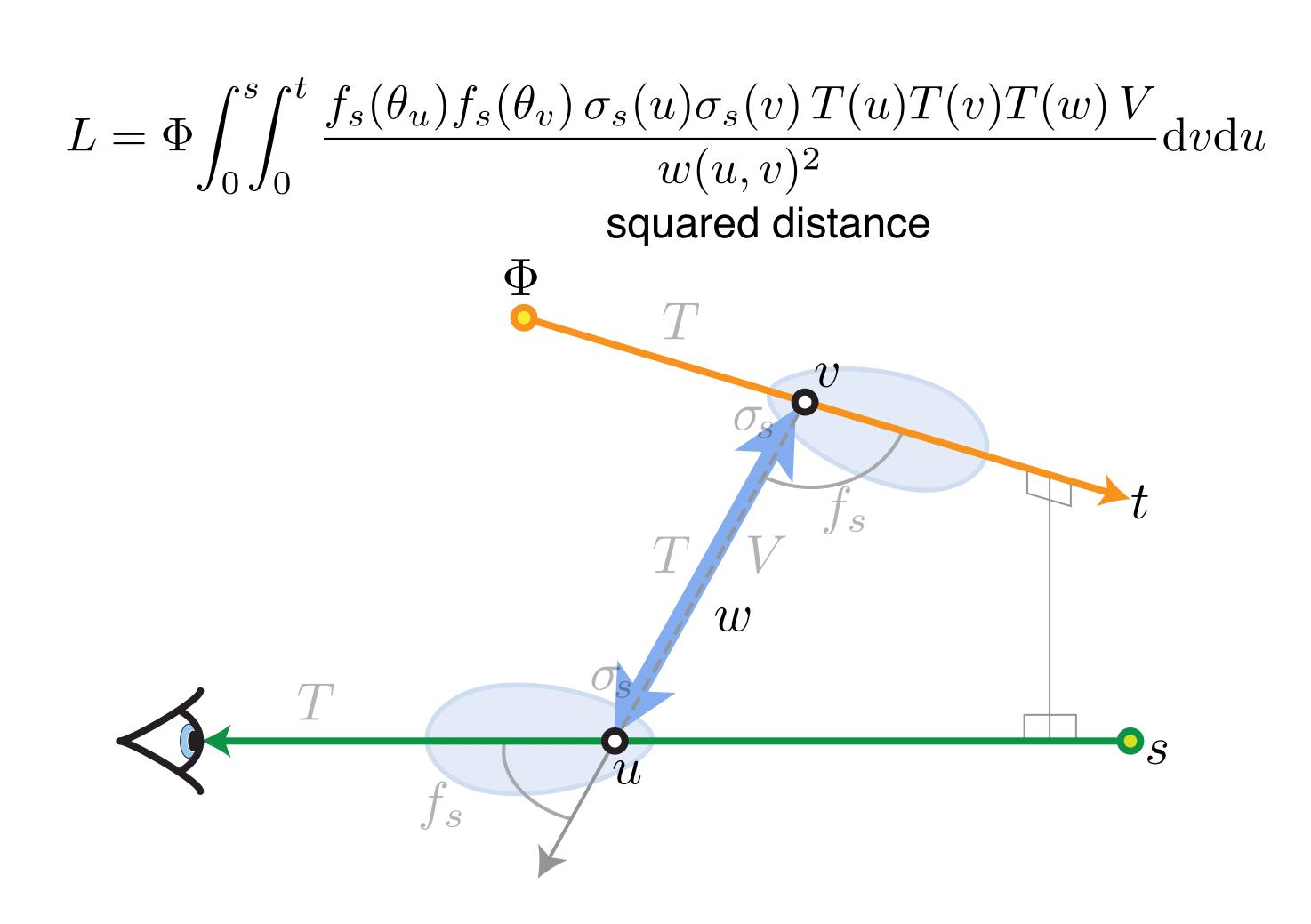
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w) V}{\mathrm{d}v \mathrm{d}u}$$

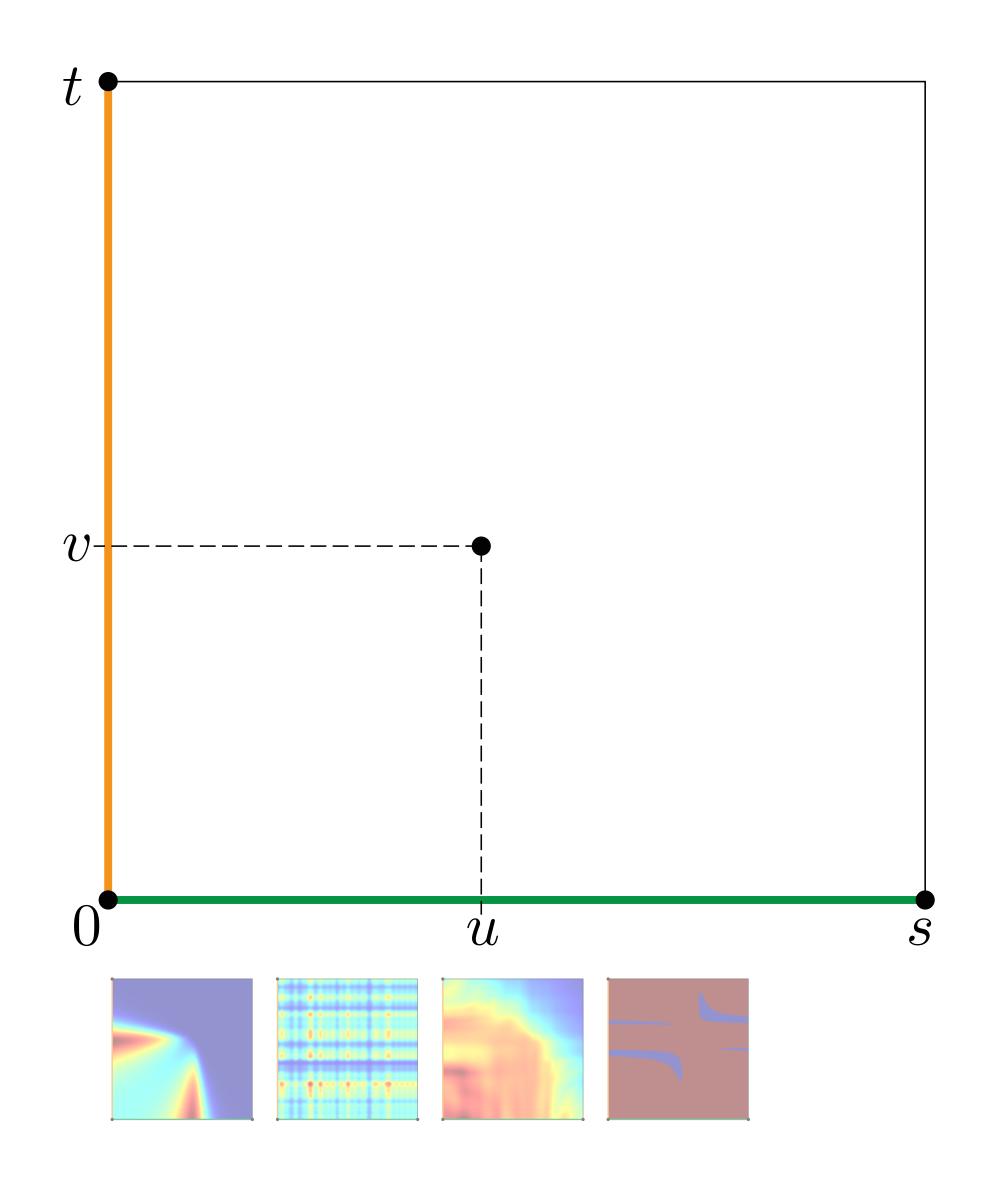


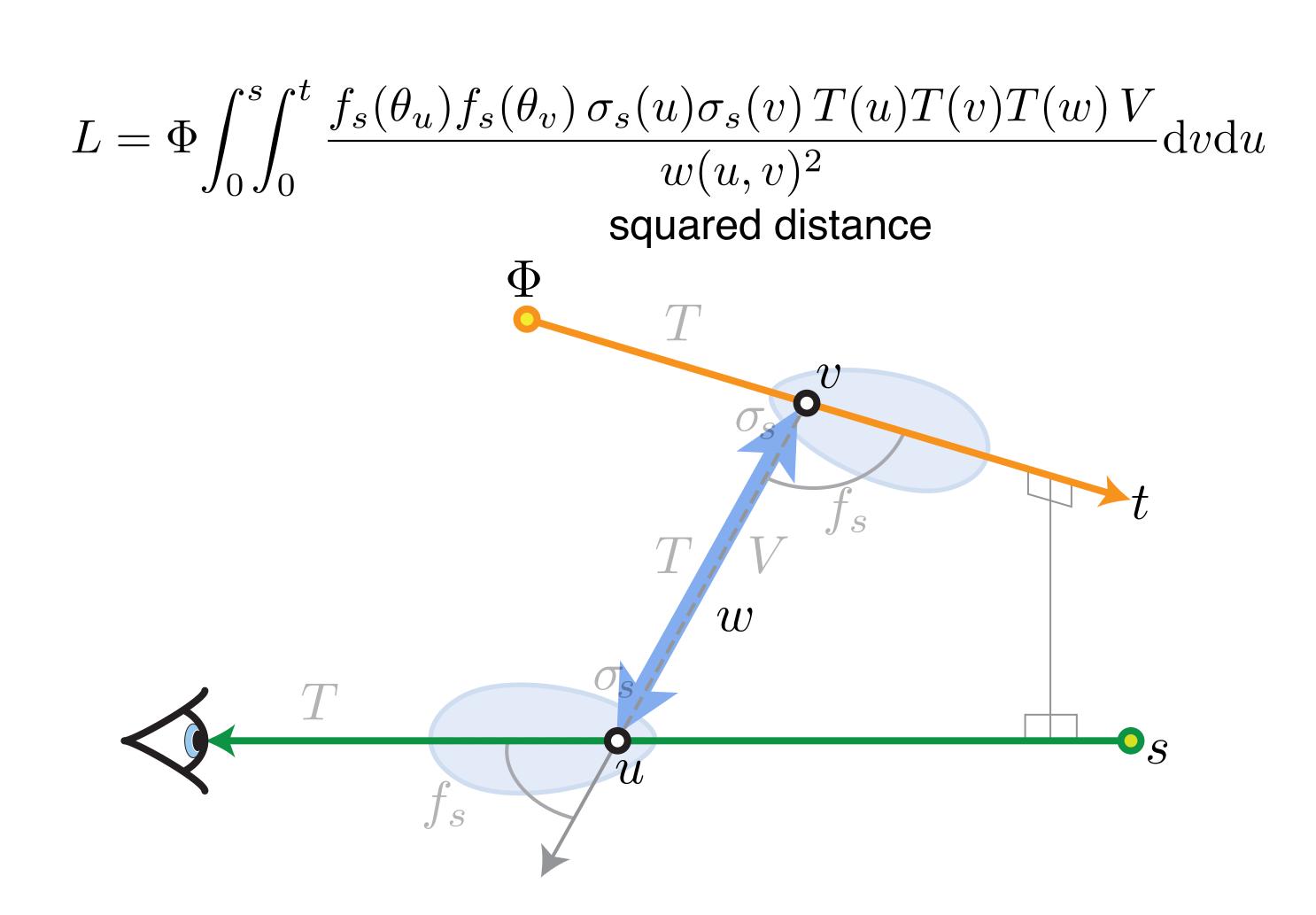


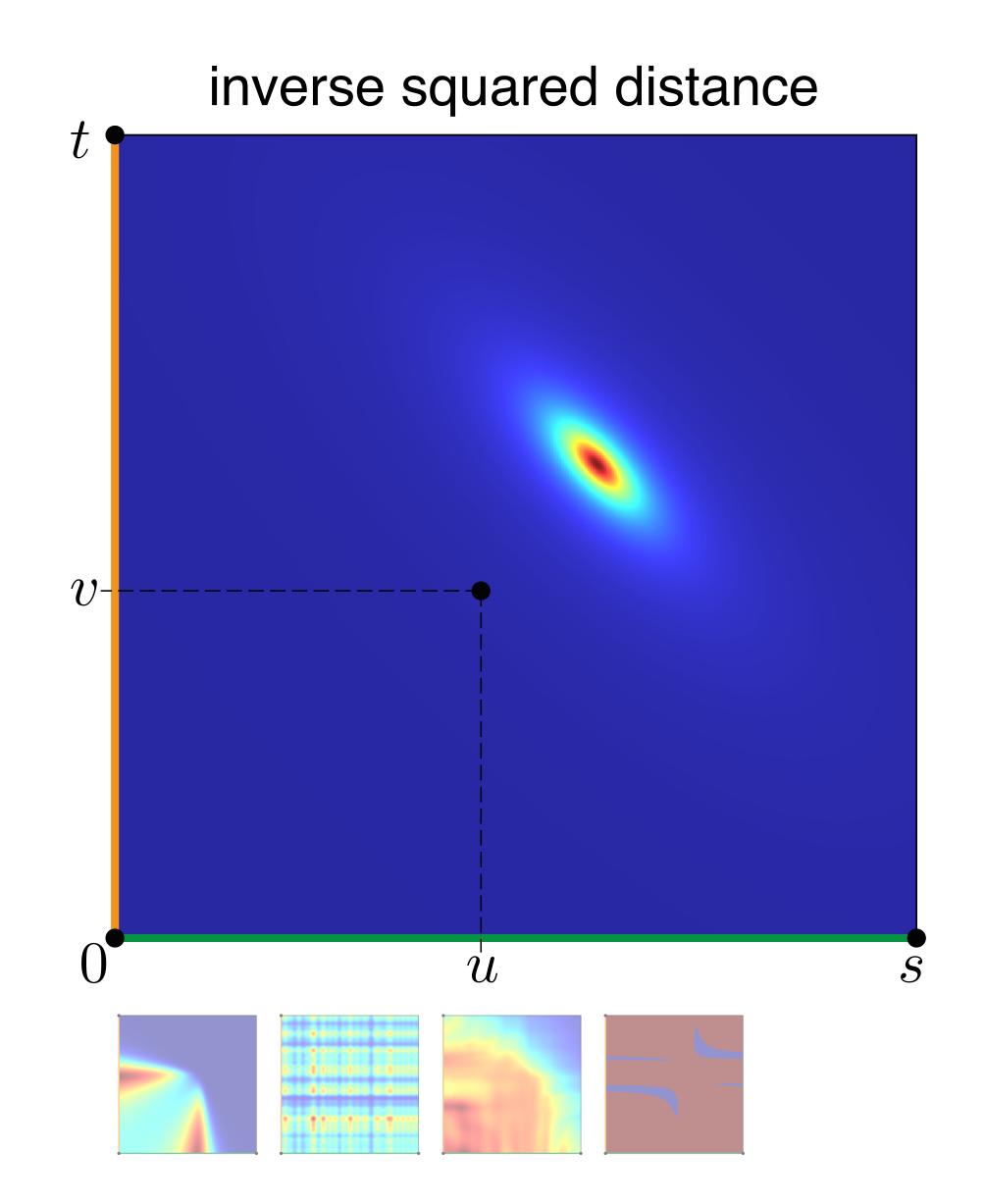




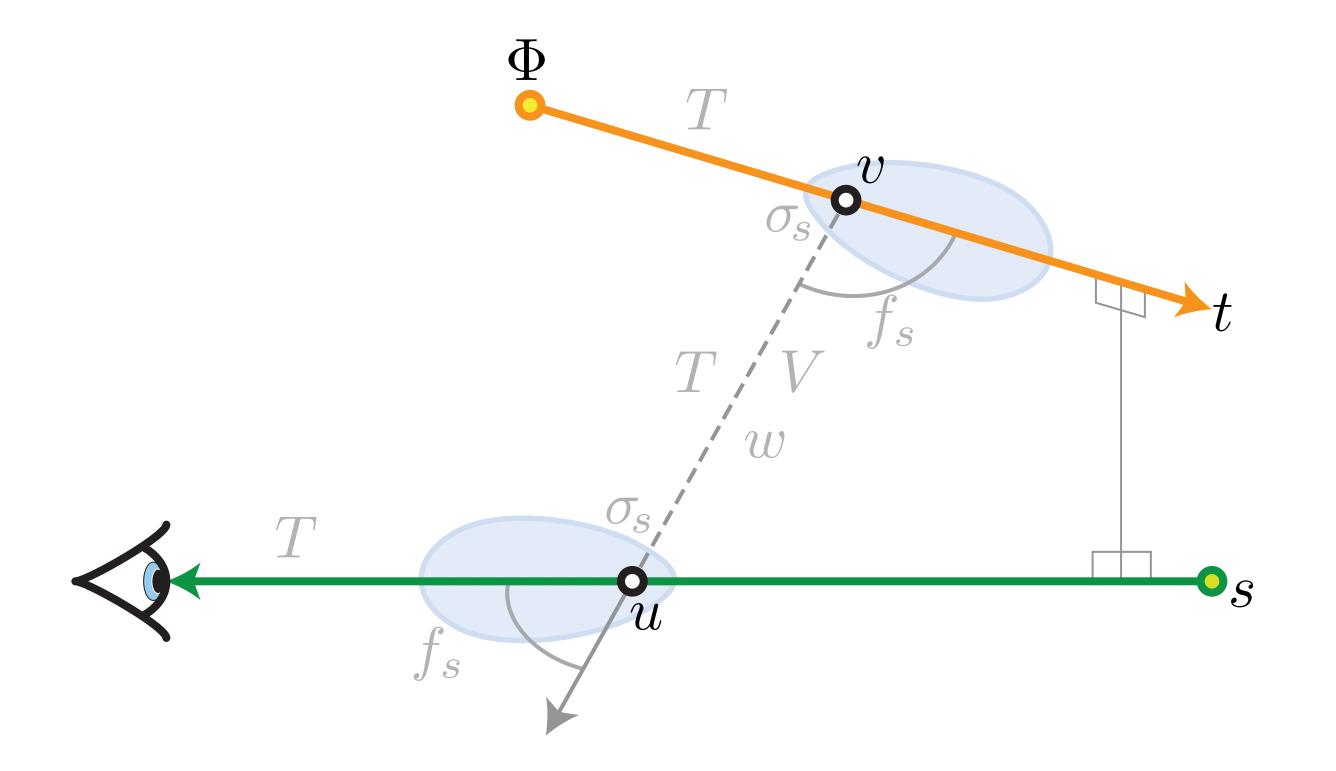


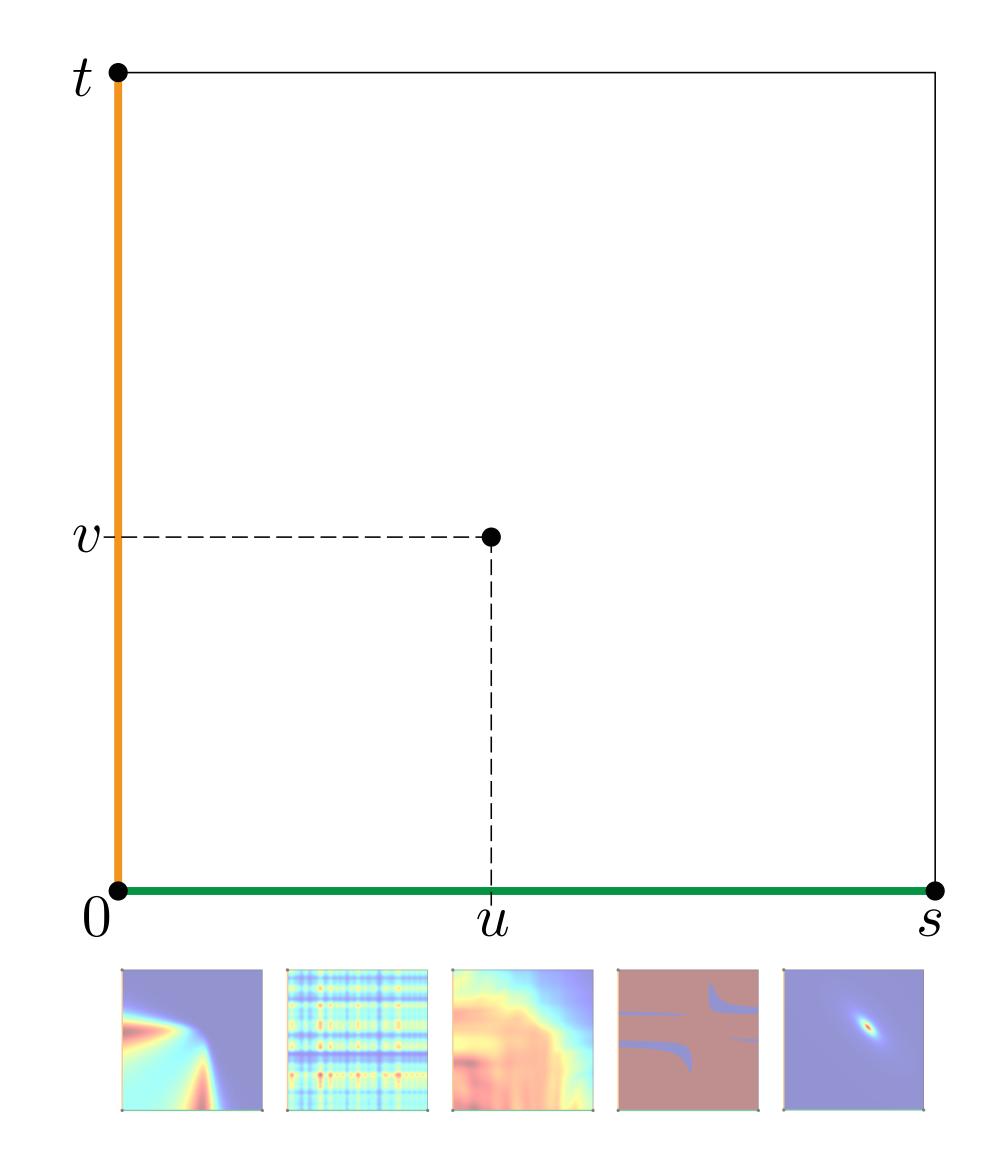




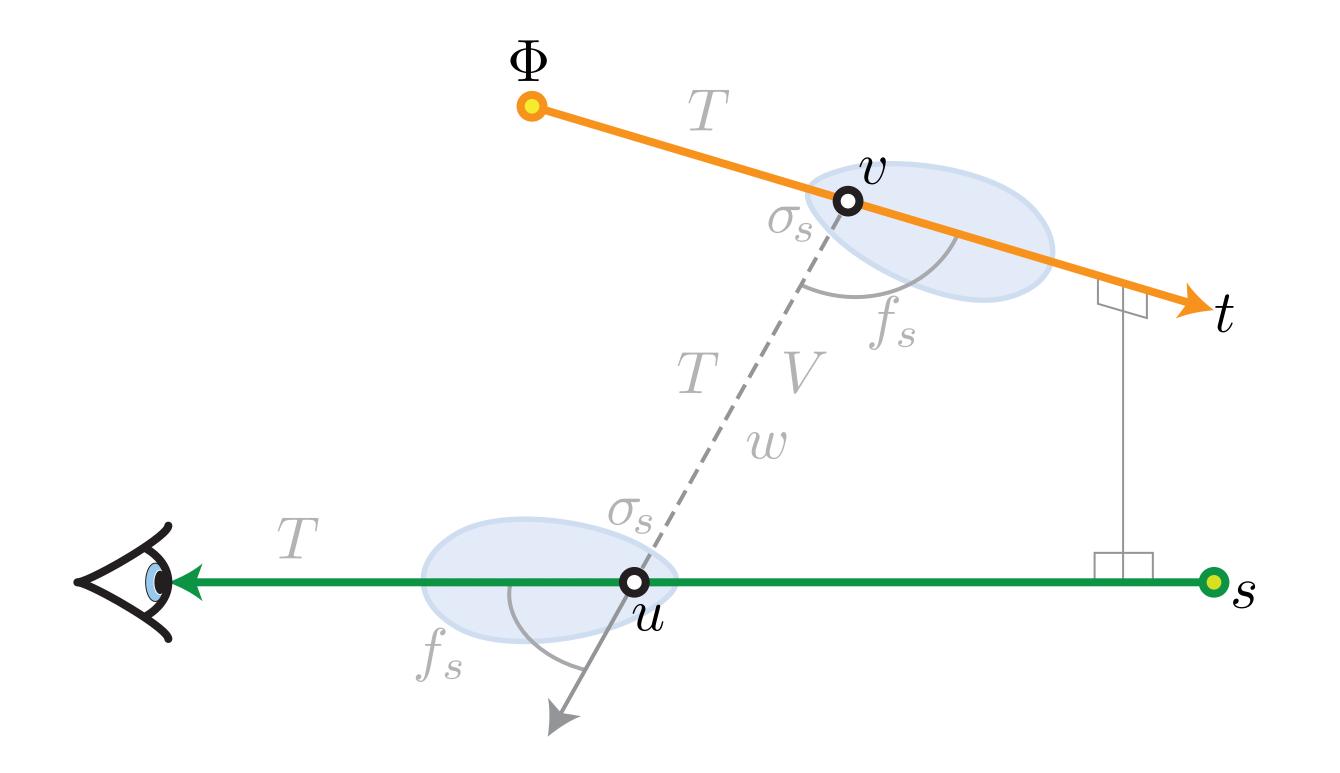


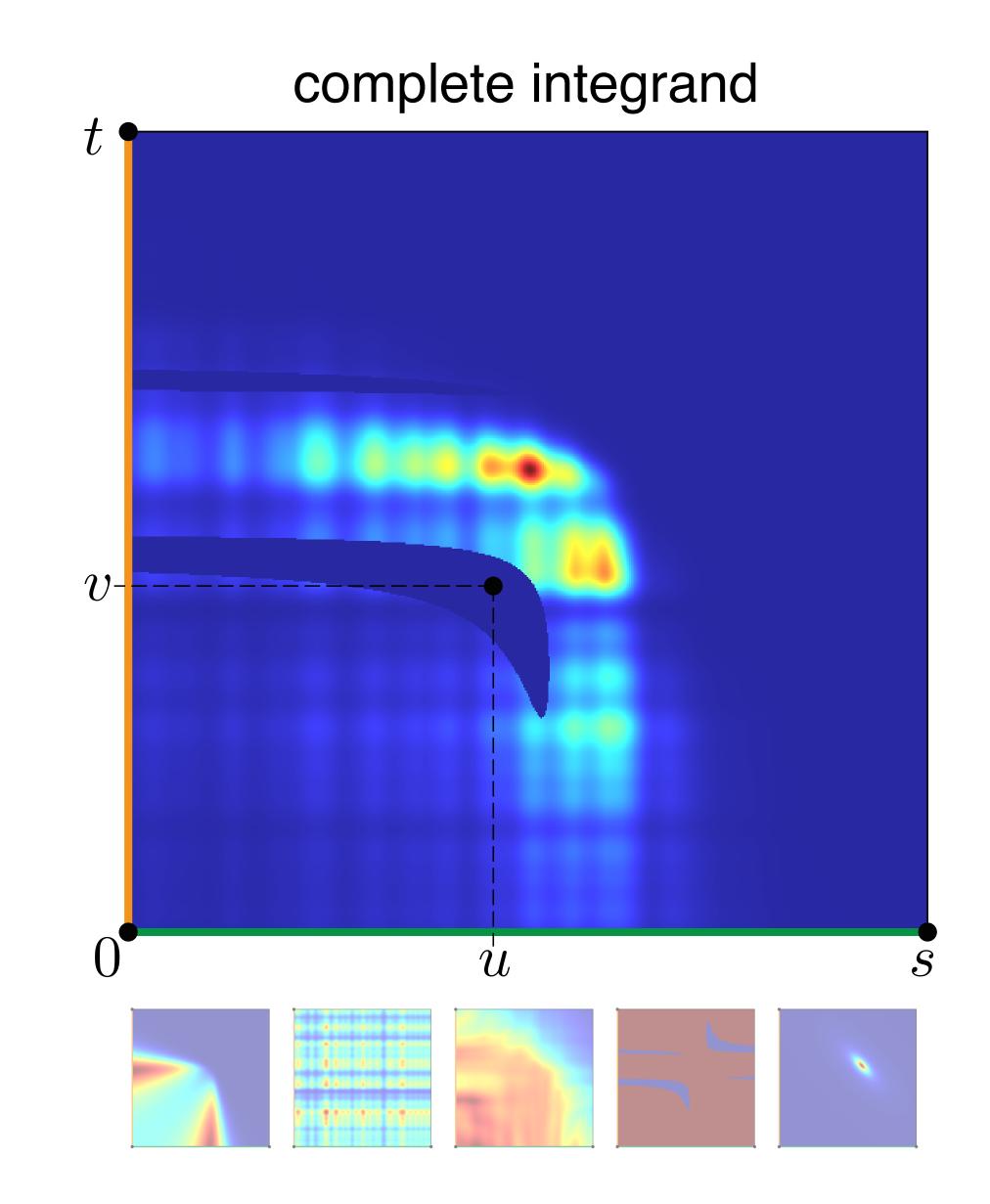
$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w) V}{w(u, v)^2} dv du$$



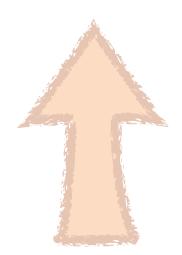


$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w) V}{w(u, v)^2} dv du$$

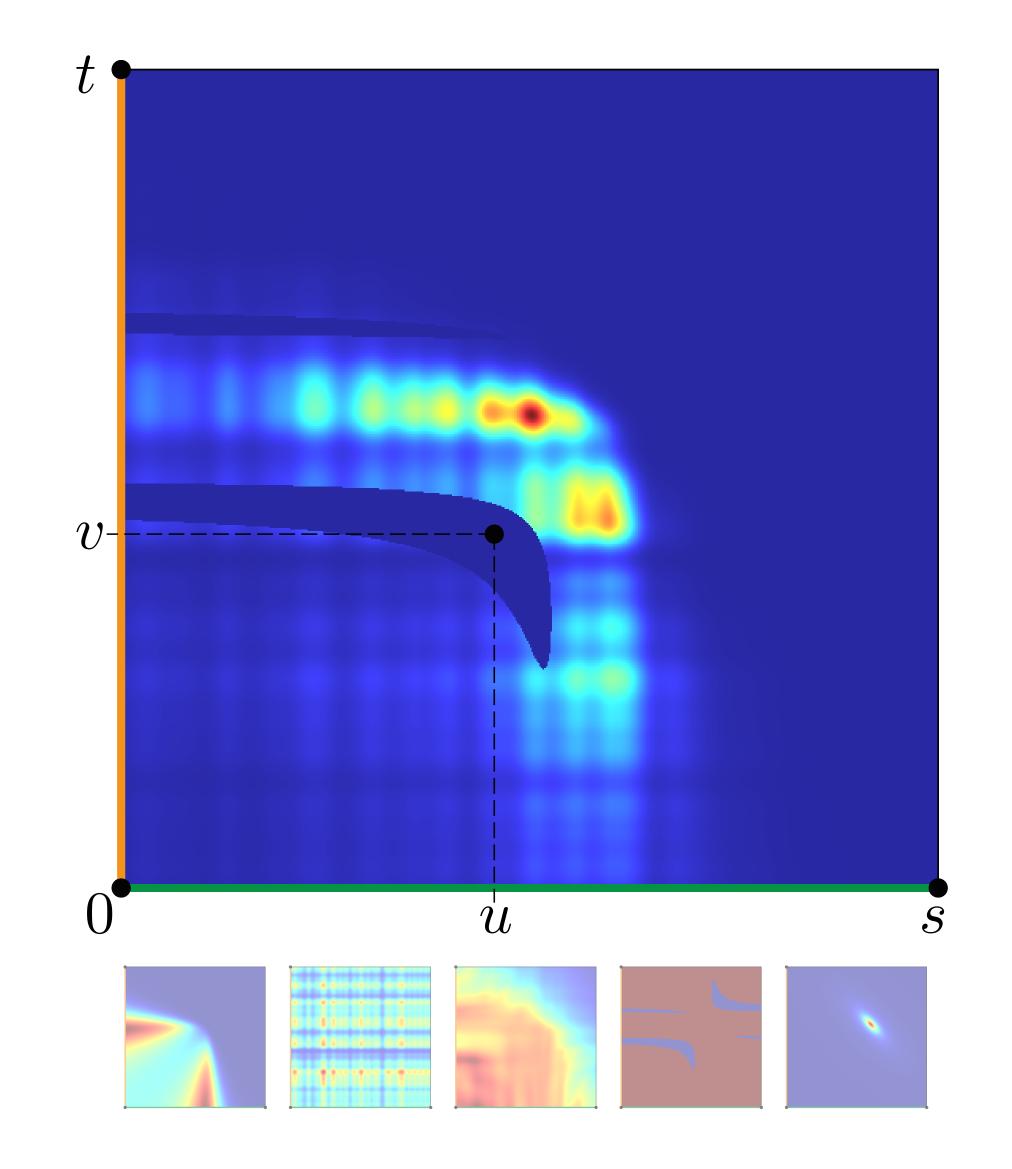


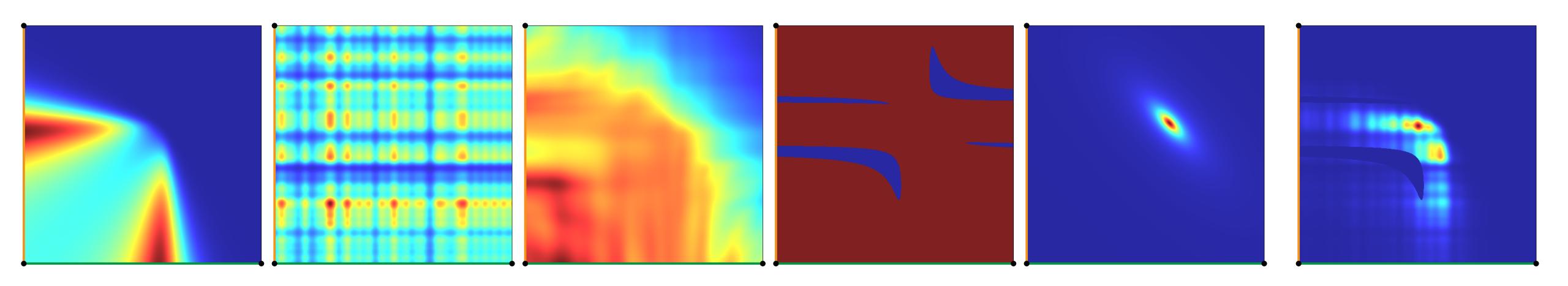


$$L = \Phi \int_0^s \int_0^t \frac{f_s(\theta_u) f_s(\theta_v) \sigma_s(u) \sigma_s(v) T(u) T(v) T(w) V}{w(u, v)^2} dv du$$

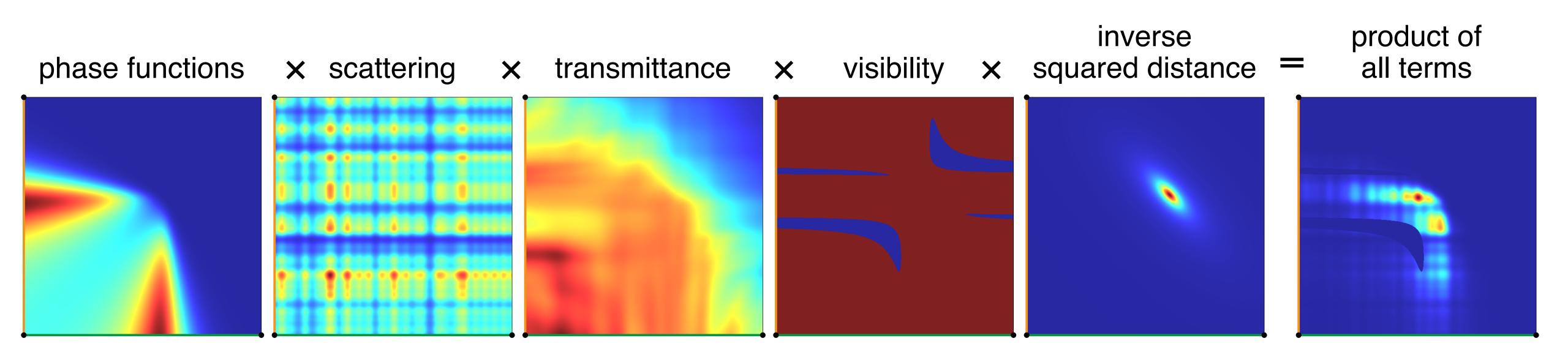


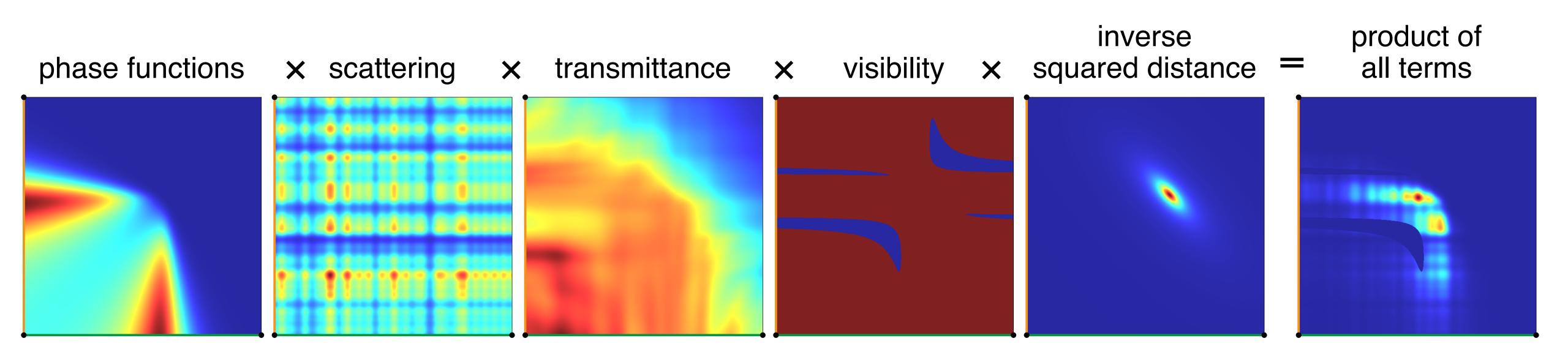
approximate using Monte Carlo with importance sampling

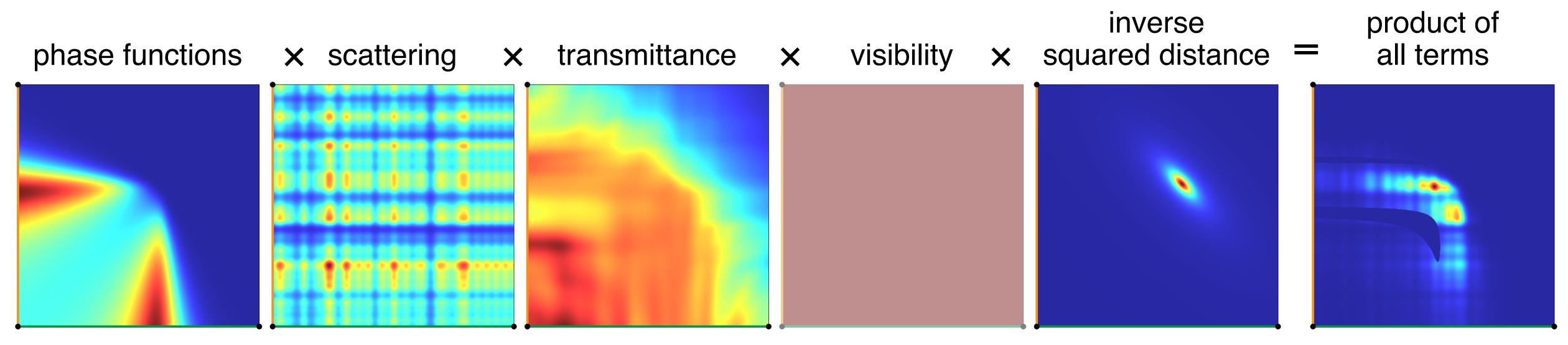


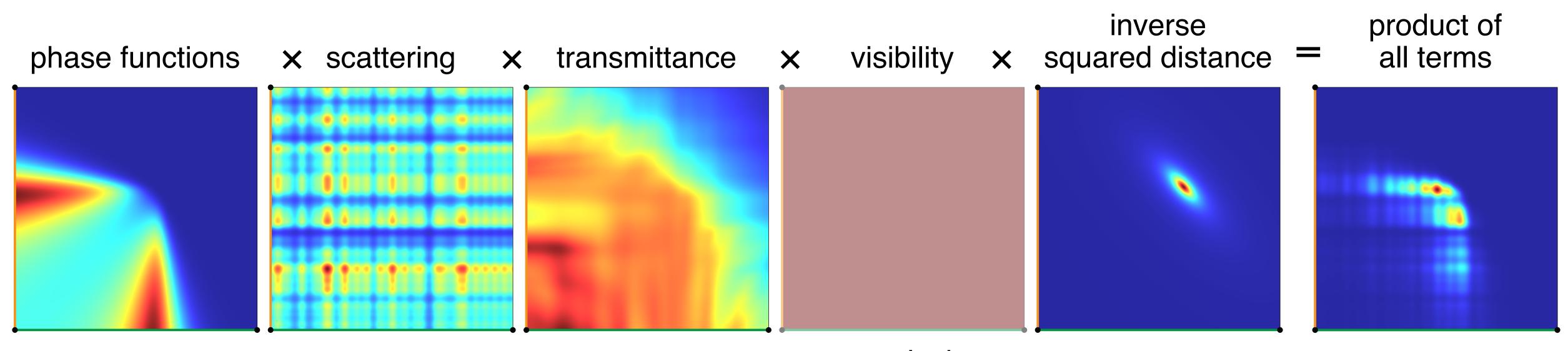


How to (importance) sample?

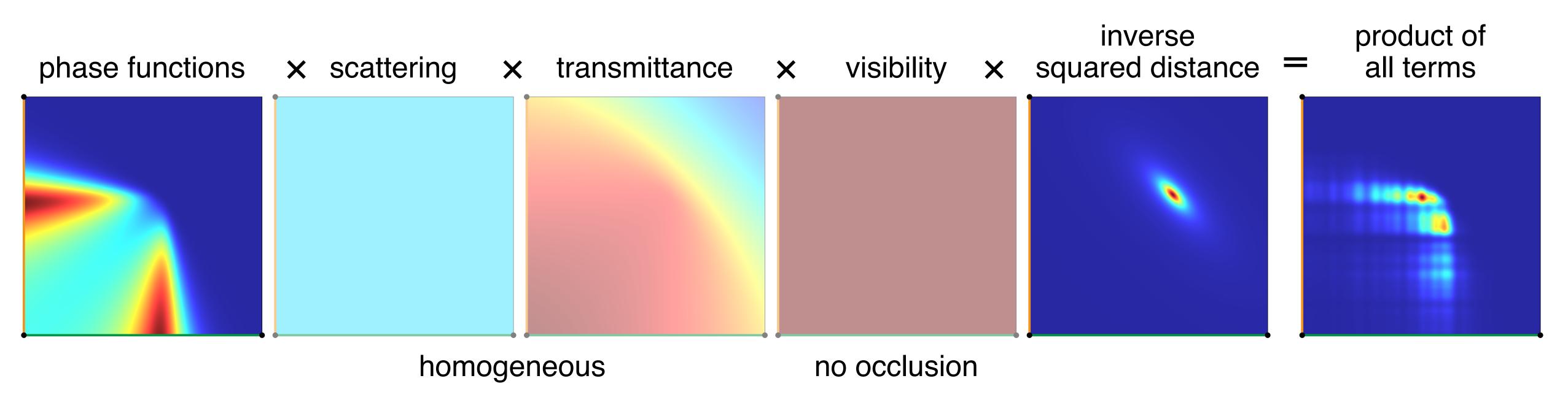


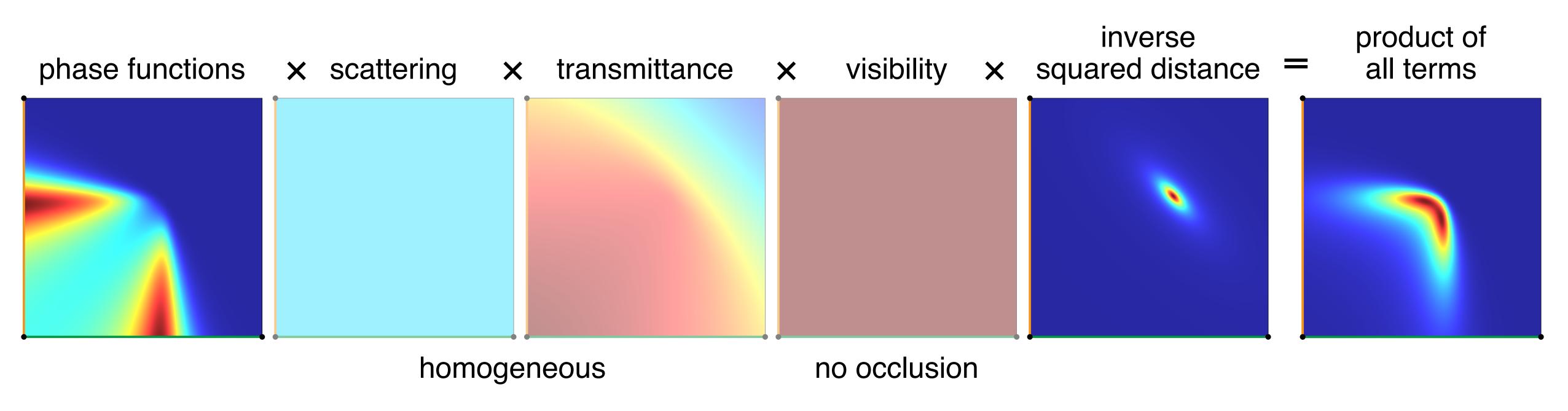


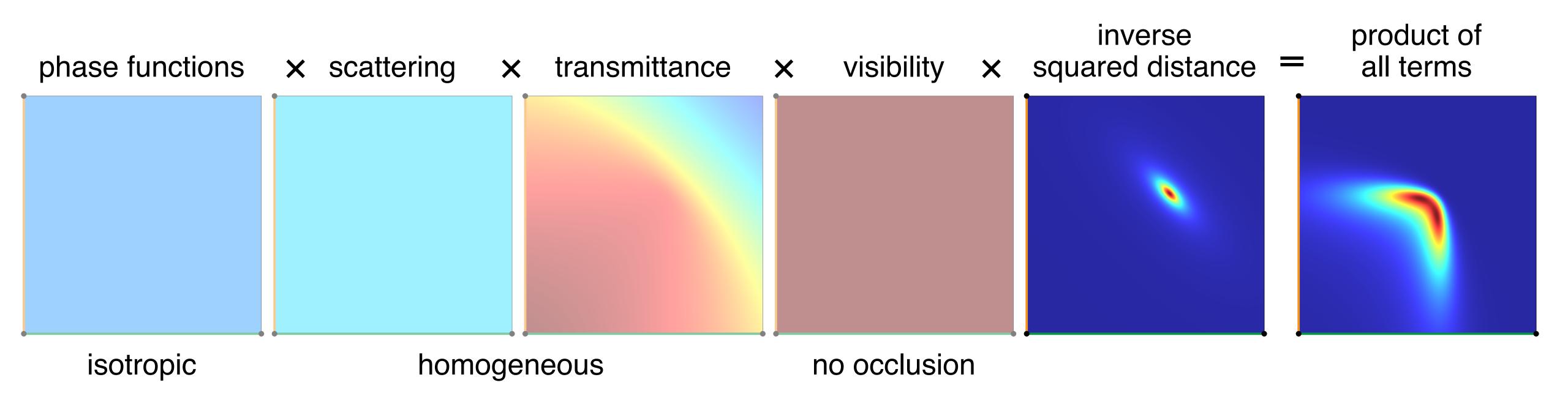




no occlusion

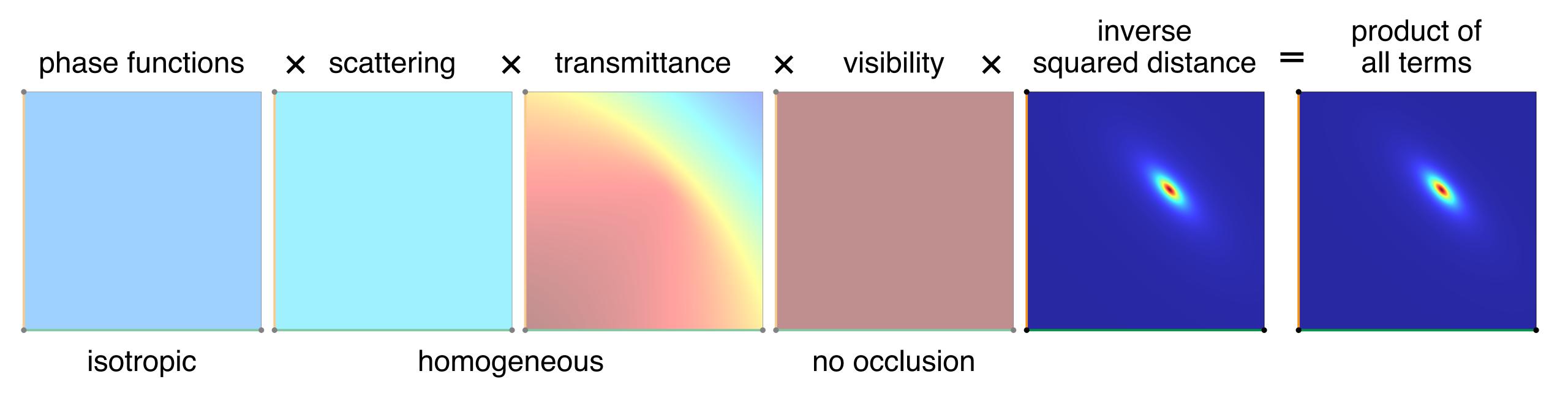






How to (importance) sample?

Simple cases first!

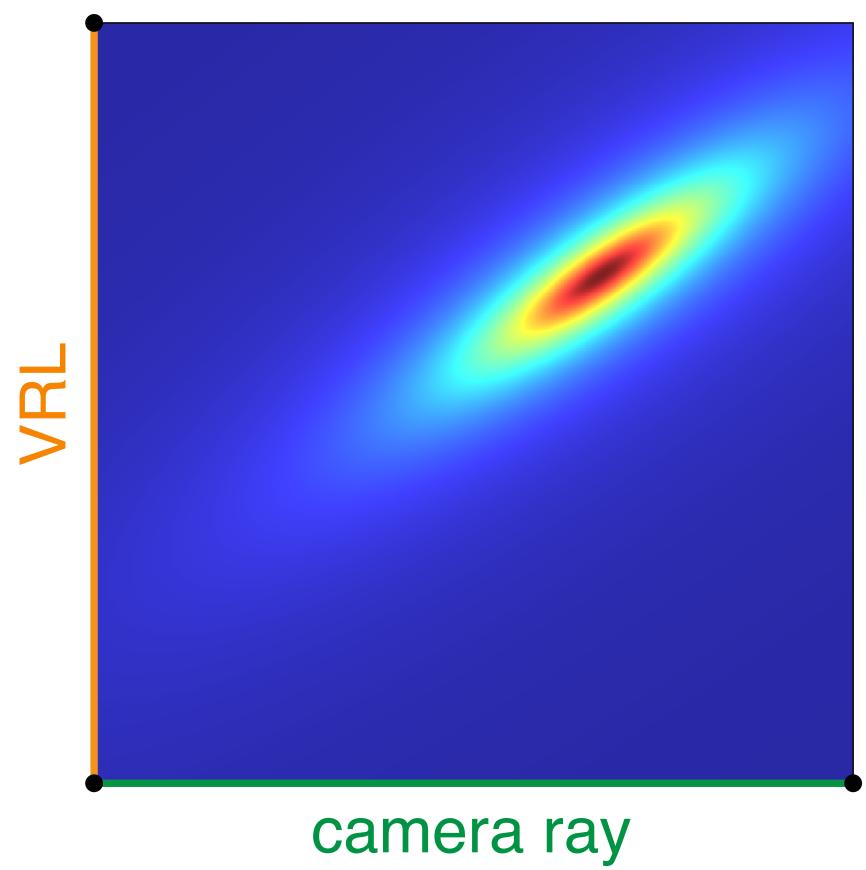




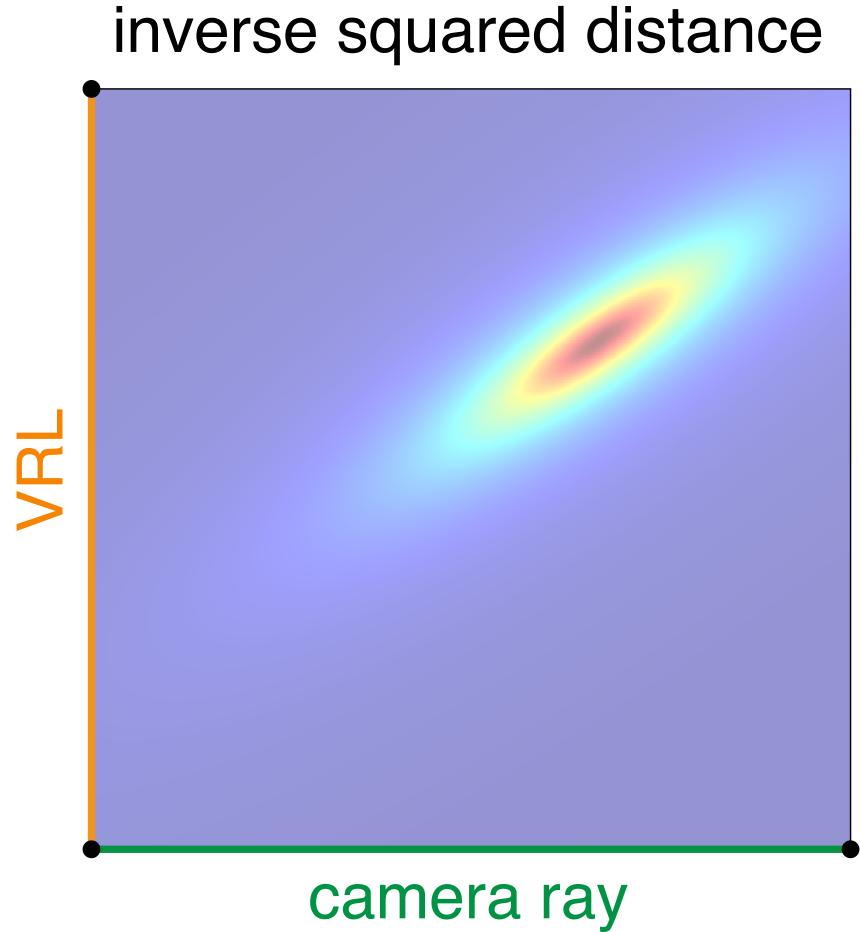




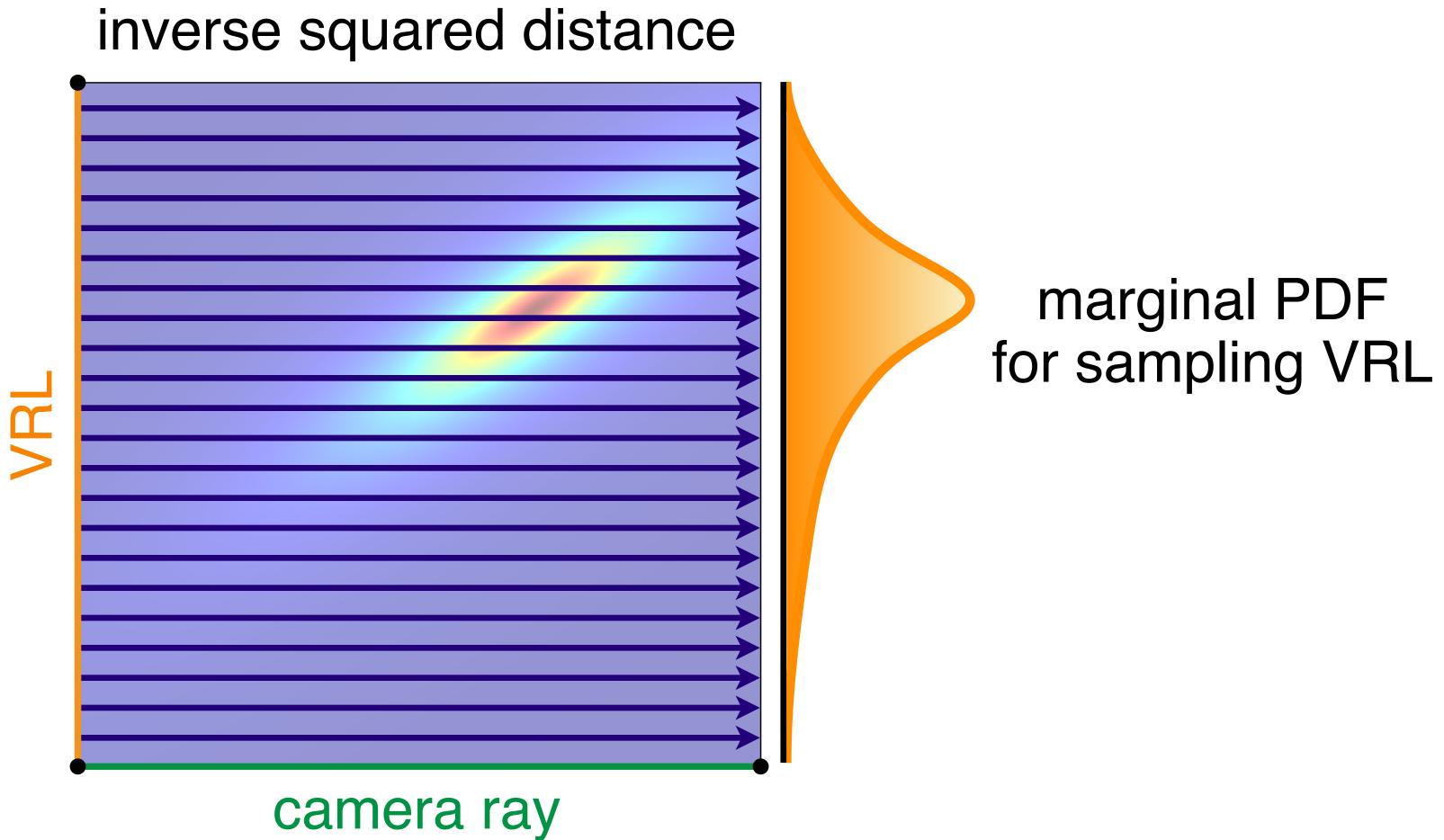
inverse squared distance



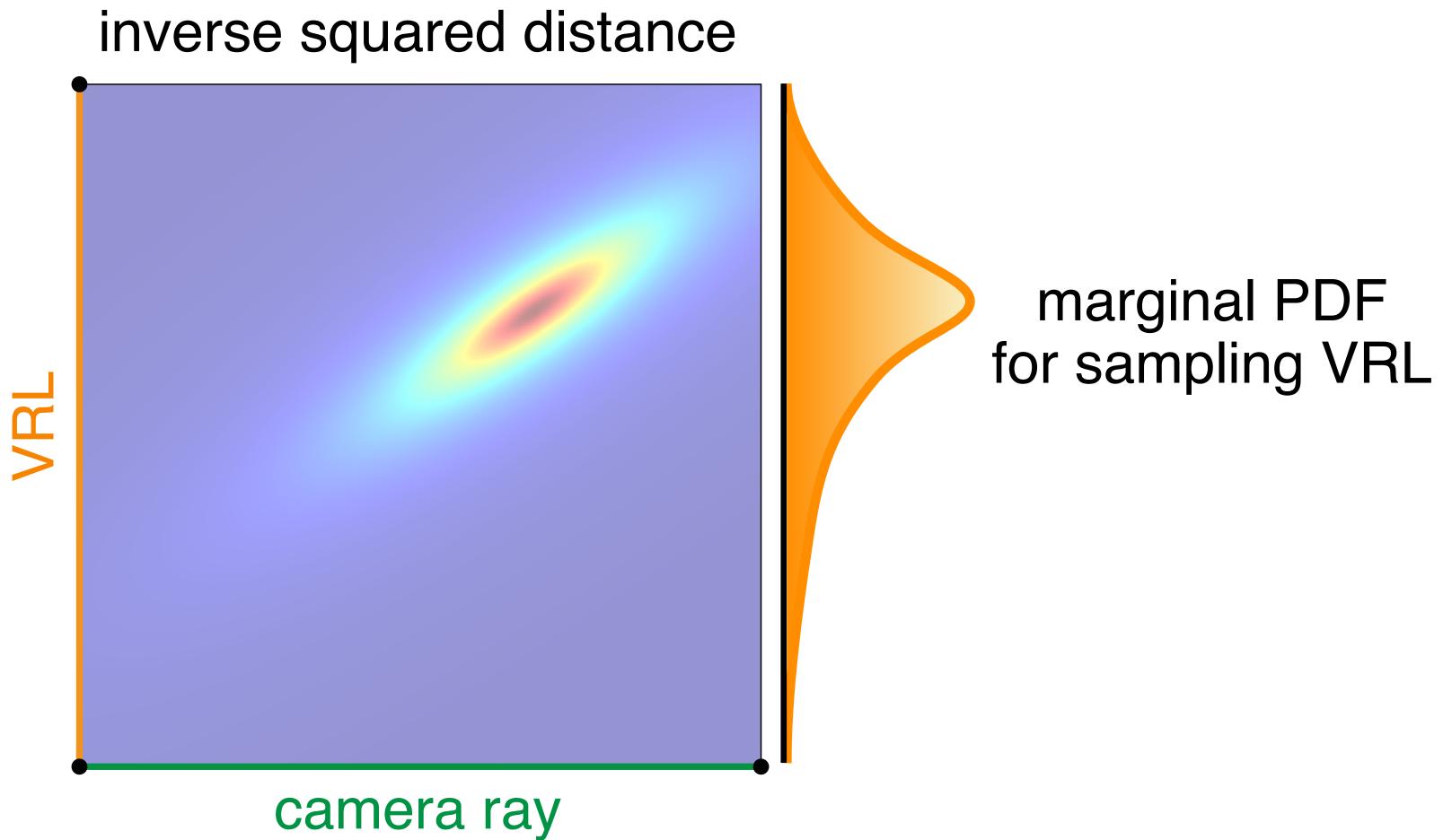




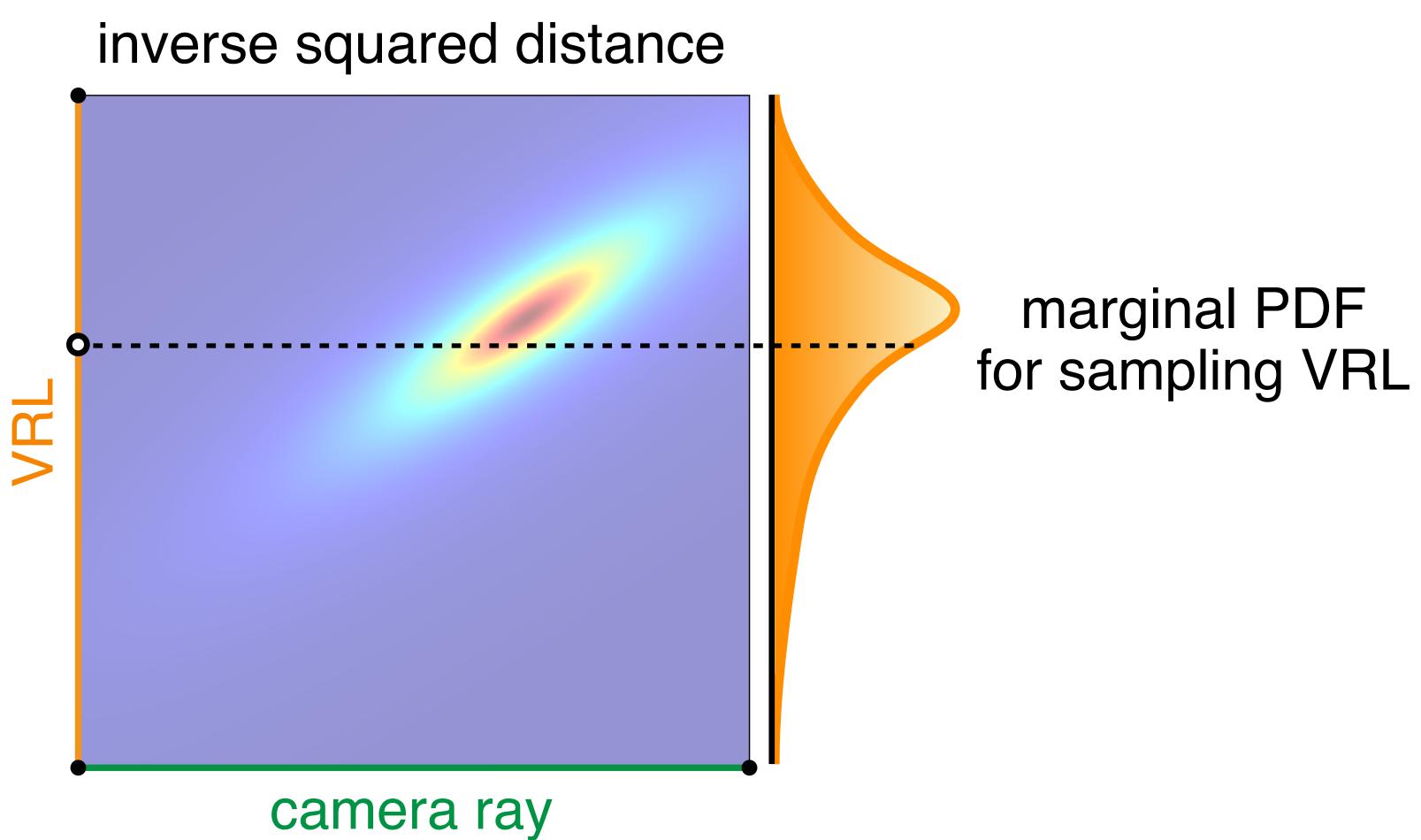




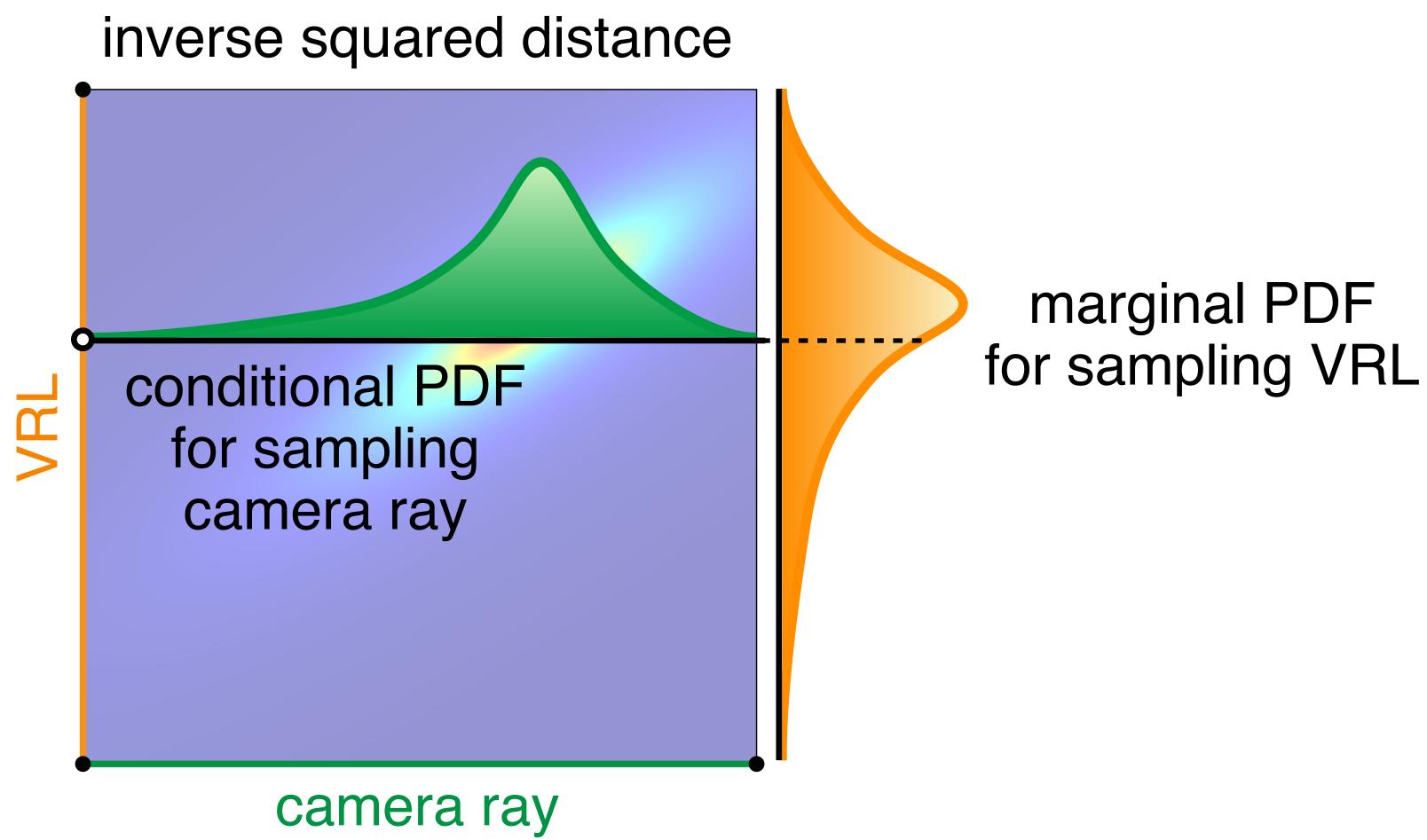


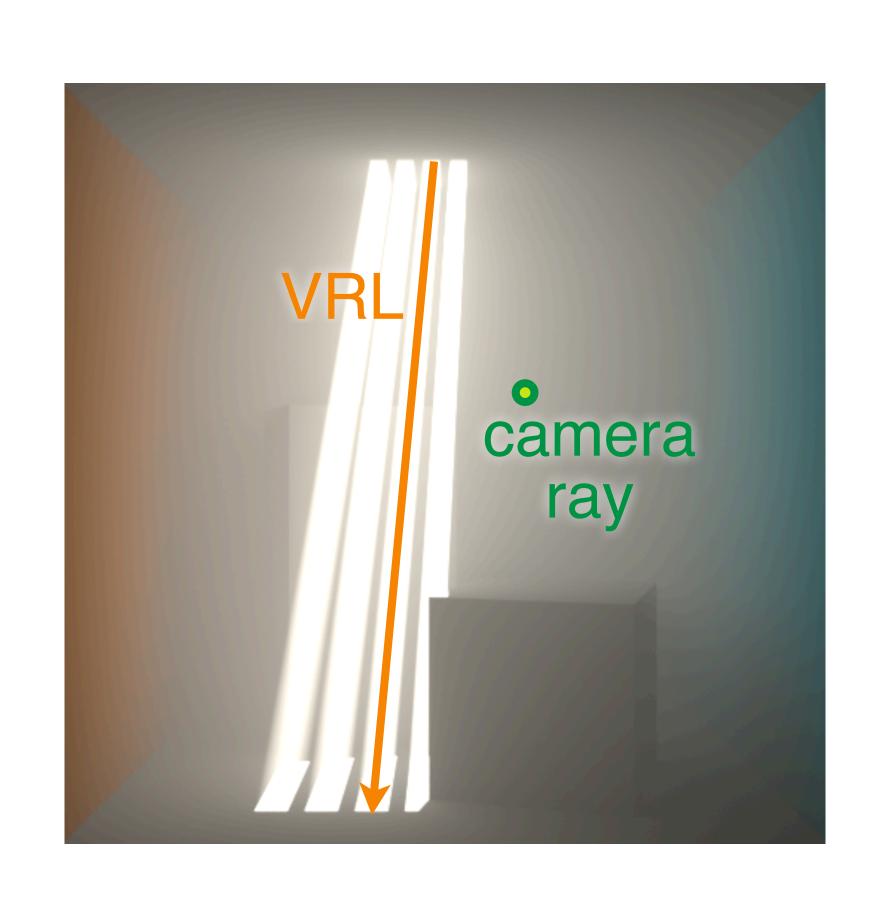


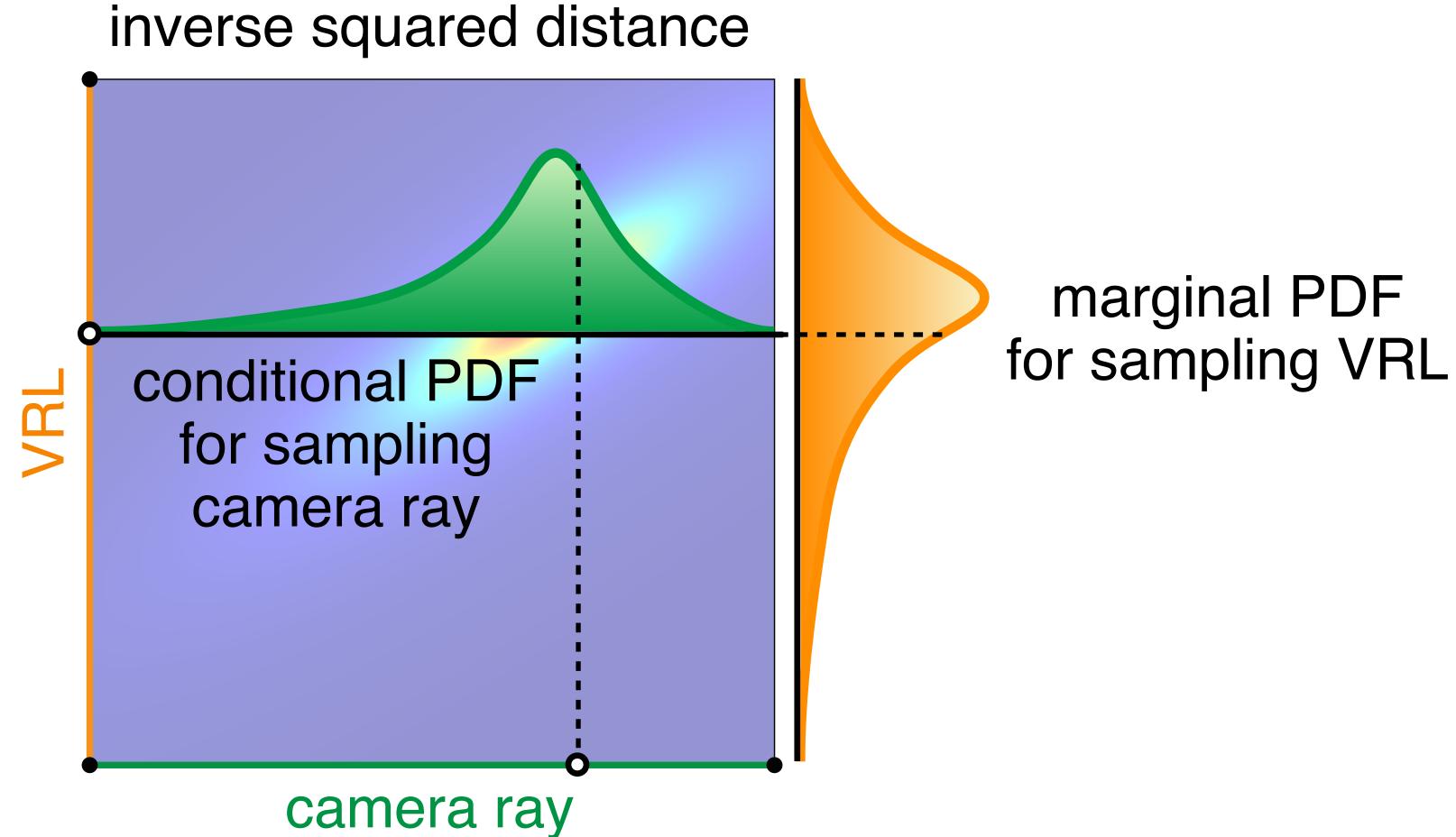


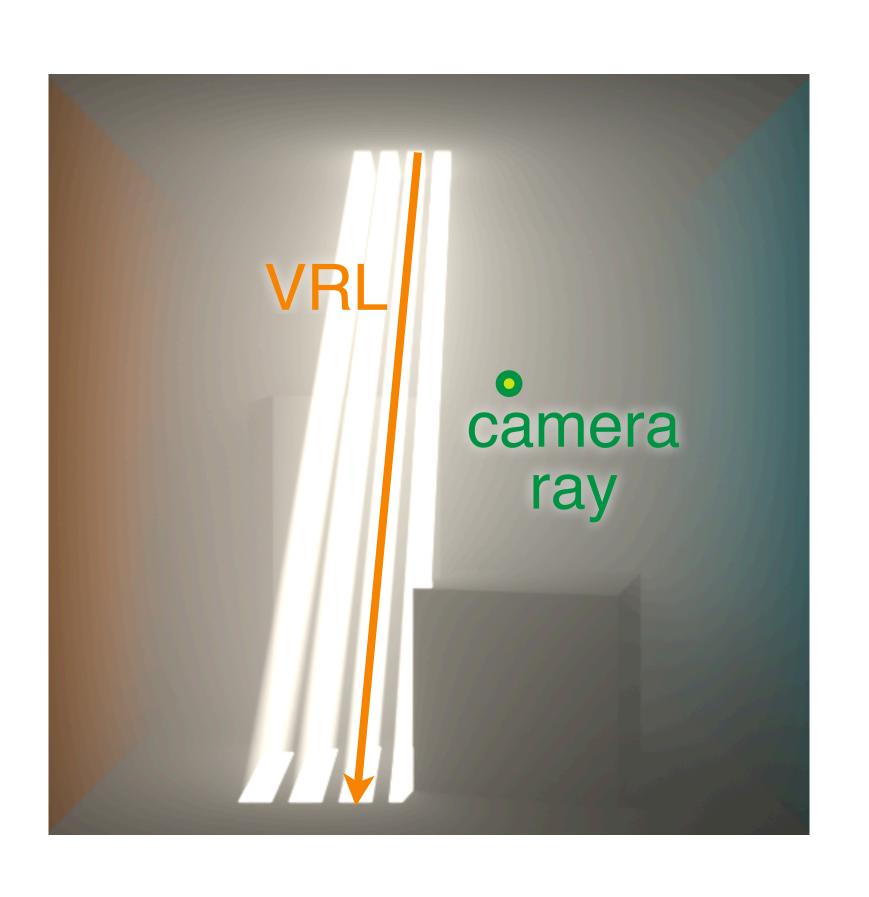




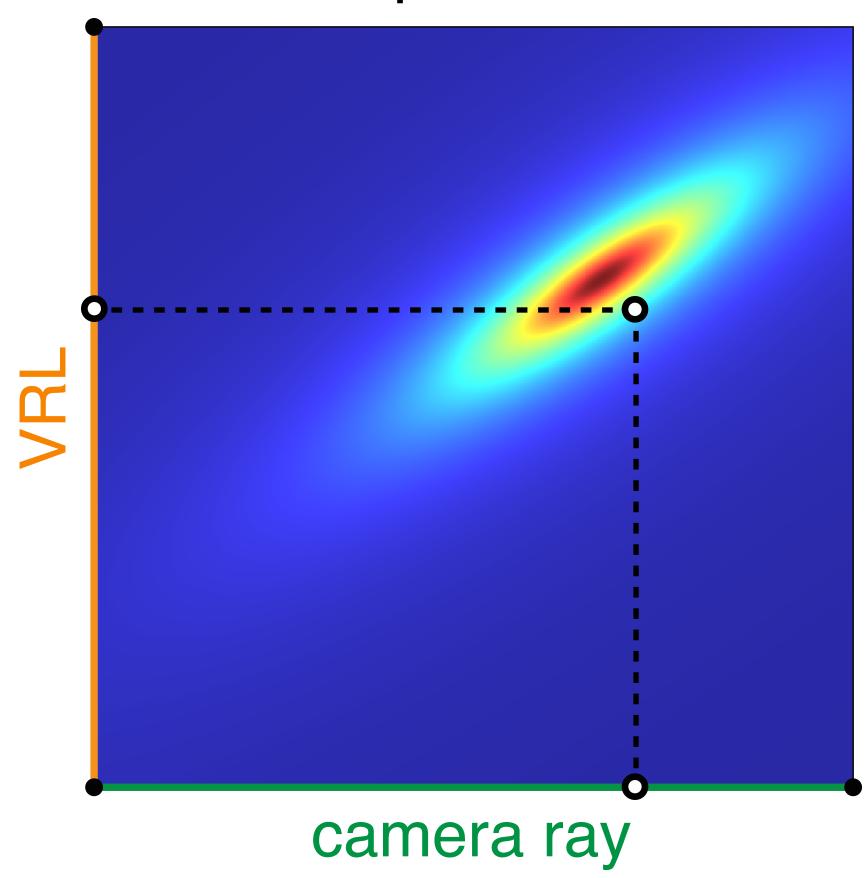




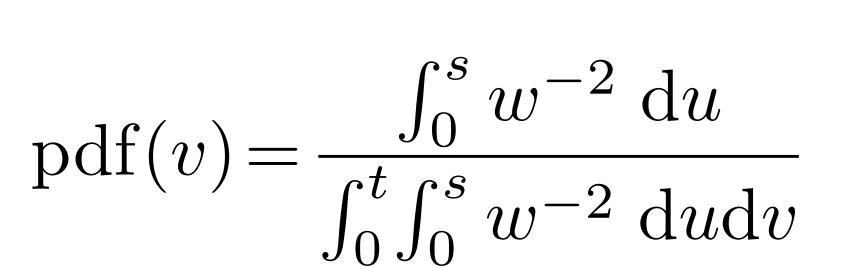




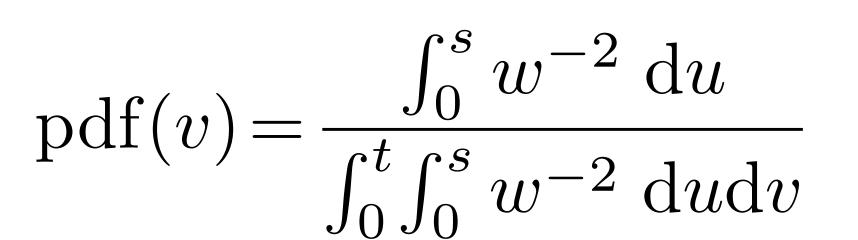
inverse squared distance



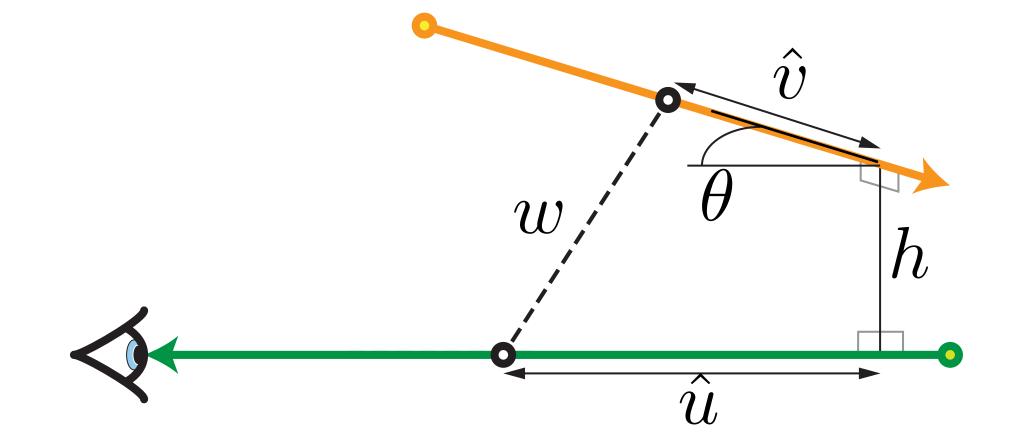
• Marginal PDF



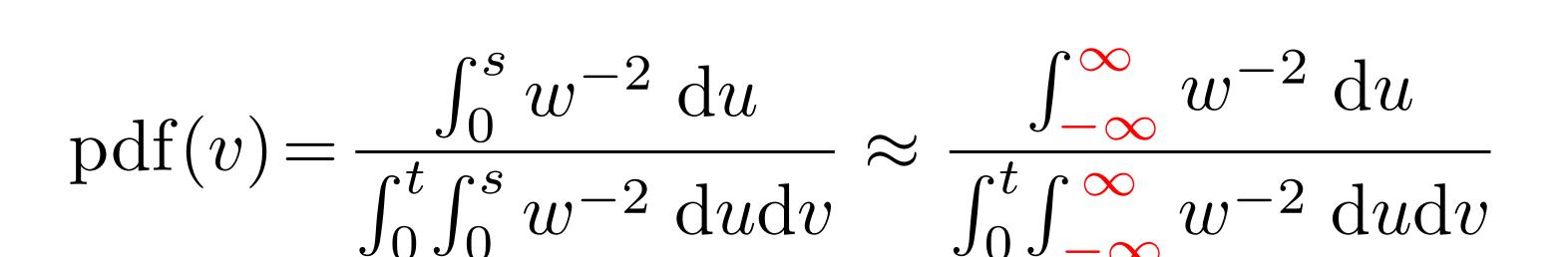
• Marginal PDF

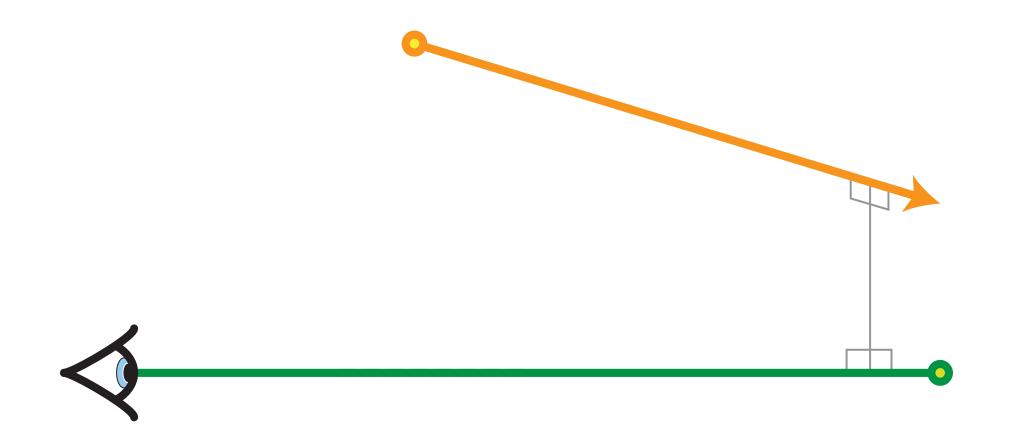


$$w = \sqrt{h^2 + \hat{u}^2 + \hat{v}^2 - 2\hat{u}\hat{v}\cos\theta}$$

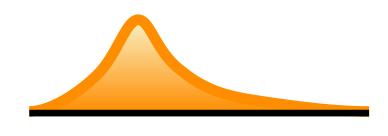


• Marginal PDF



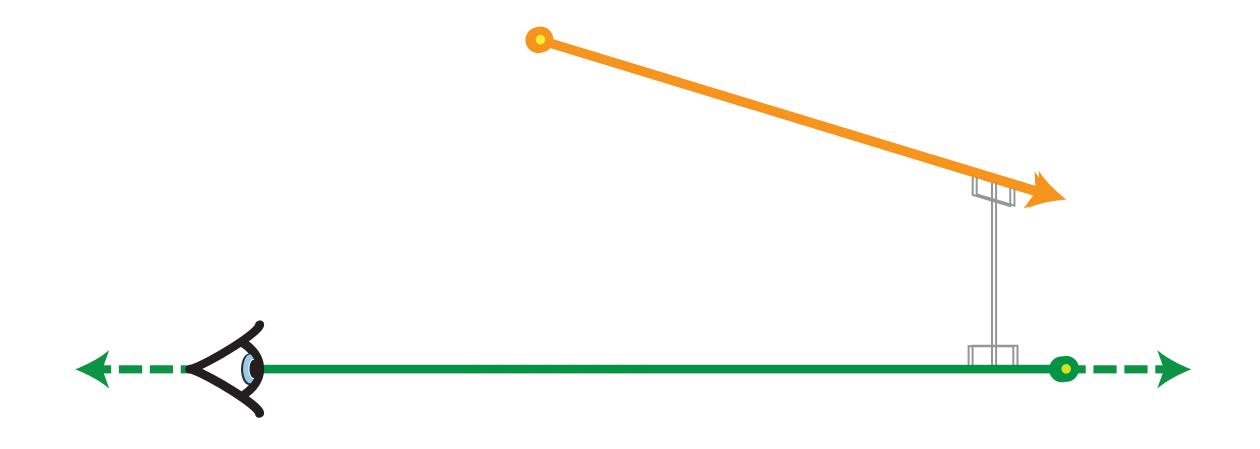


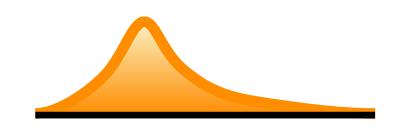
• Marginal PDF



$$pdf(v) = \frac{\int_0^s w^{-2} du}{\int_0^t \int_0^s w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv}$$

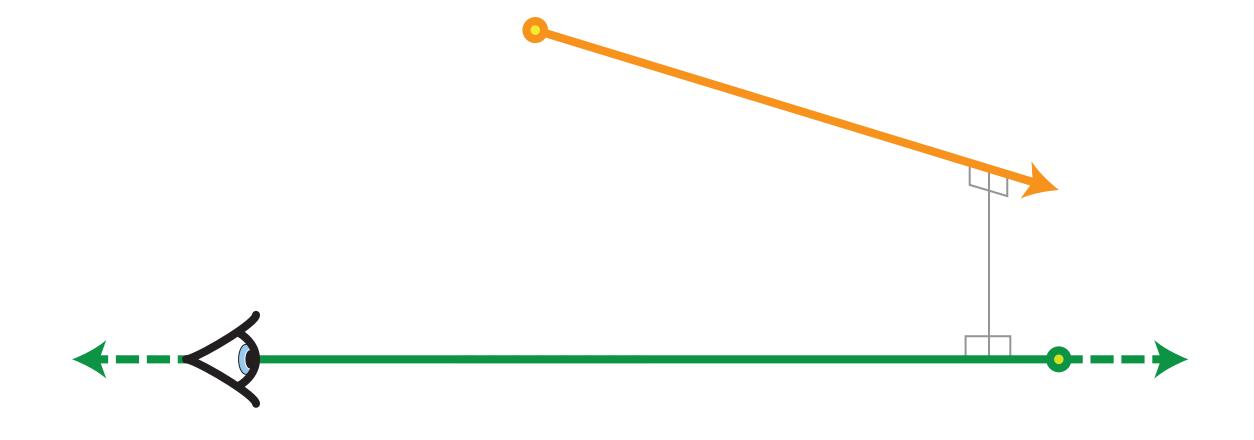
assume infinite camera ray

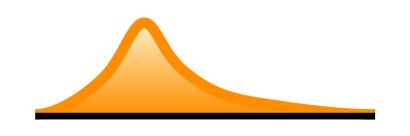




$$pdf(v) = \frac{\int_0^s w^{-2} du}{\int_0^t \int_0^s w^{-2} du dv} \approx \frac{\int_{-\infty}^{\infty} w^{-2} du}{\int_0^t \int_{-\infty}^{\infty} w^{-2} du dv} = \frac{\sin \theta}{(A(\hat{v}_1) - A(\hat{v}_0))\sqrt{h^2 + v^2 \sin^2 \theta}}$$

assume infinite camera ray

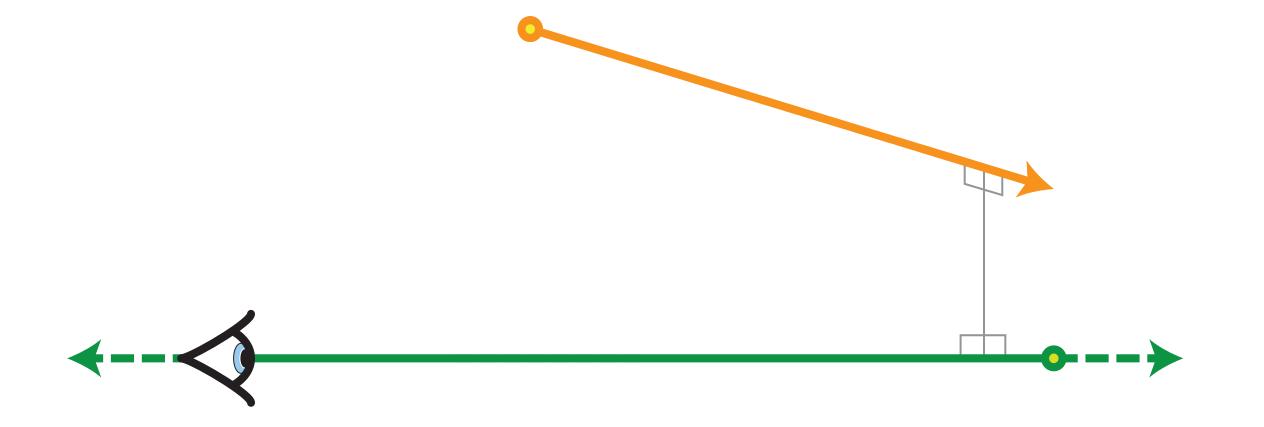




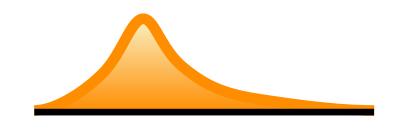
$$pdf(v) = \frac{\int_0^s w^{-2} du}{\int_0^t \int_0^s w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv} = \frac{\sin \theta}{(A(\hat{v}_1) - A(\hat{v}_0))\sqrt{h^2 + v^2 \sin^2 \theta}}$$

assume infinite camera ray

$$A(x) = \sinh^{-1}\left(\frac{x}{h}\sin\theta\right)$$



• Marginal PDF



$$pdf(v) = \frac{\int_0^s w^{-2} du}{\int_0^t \int_0^s w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv} = \frac{\sin \theta}{(A(\hat{v}_1) - A(\hat{v}_0))\sqrt{h^2 + v^2 \sin^2 \theta}}$$

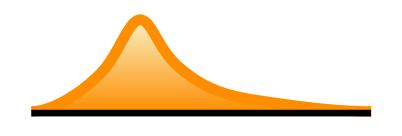
assume infinite camera ray

$$A(x) = \sinh^{-1}\left(\frac{x}{h}\sin\theta\right)$$

$$\operatorname{cdf}^{-1}(\xi) = \frac{h \sinh(\operatorname{lerp}(A(\hat{v}_0), A(\hat{v}_1), \xi))}{\sin \theta}$$



• Marginal PDF

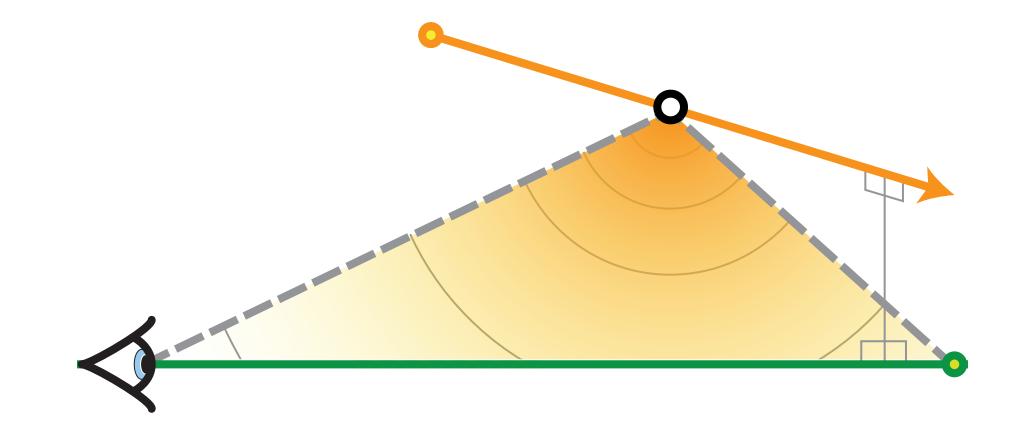


$$pdf(v) = \frac{\int_0^s w^{-2} du}{\int_0^t \int_0^s w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv} = \frac{\sin \theta}{(A(\hat{v}_1) - A(\hat{v}_0))\sqrt{h^2 + v^2 \sin^2 \theta}}$$

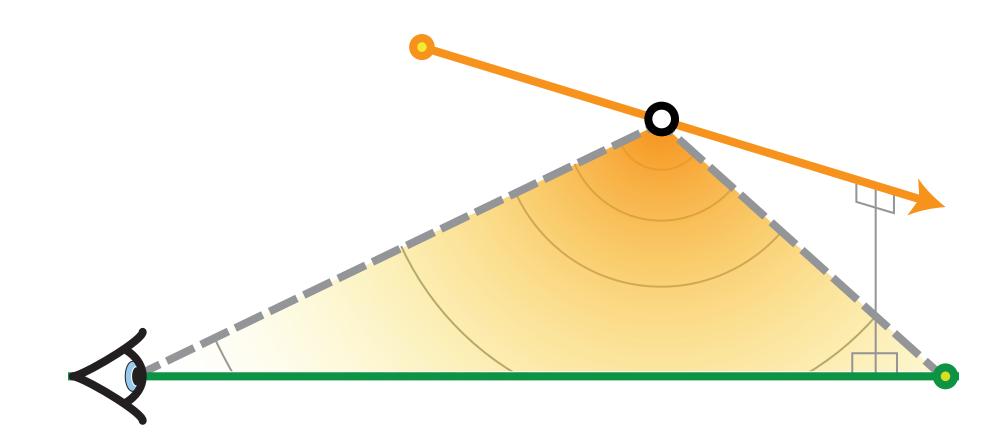
assume infinite camera ray

$$A(x) = \sinh^{-1}\left(\frac{x}{h}\sin\theta\right)$$

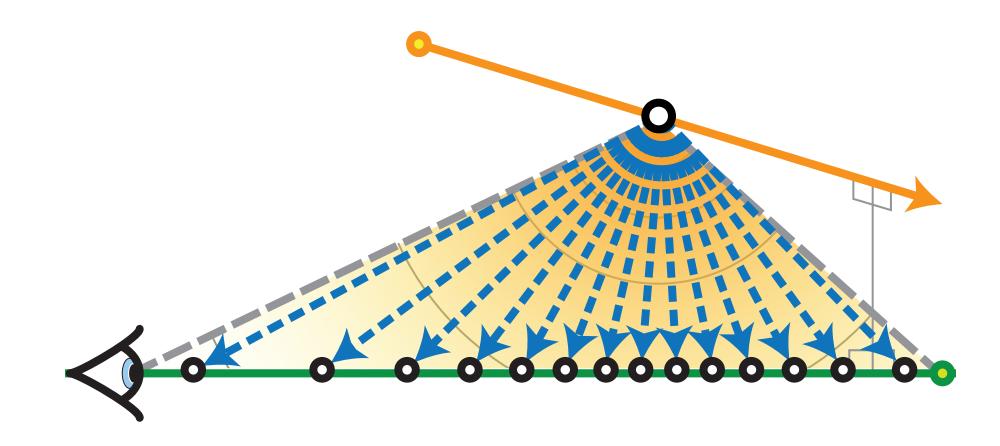
$$\operatorname{cdf}^{-1}(\xi) = \frac{h \sinh(\operatorname{lerp}(A(\hat{v}_0), A(\hat{v}_1), \xi))}{\sin \theta}$$



• Conditional PDF

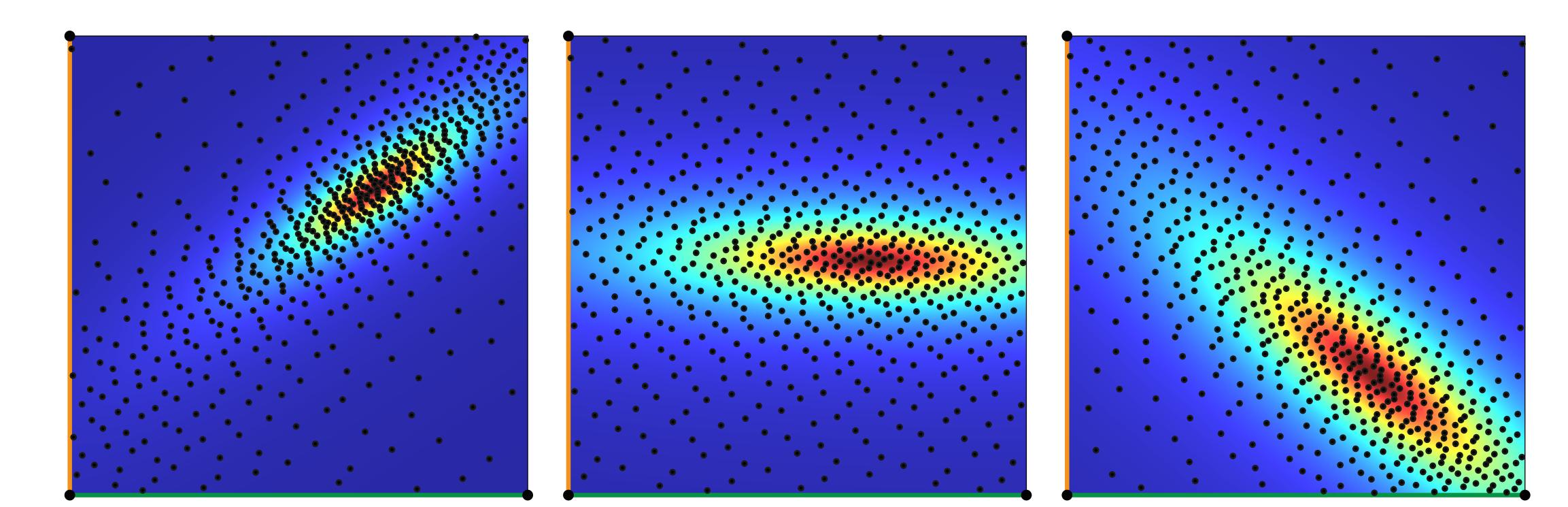


• Conditional PDF

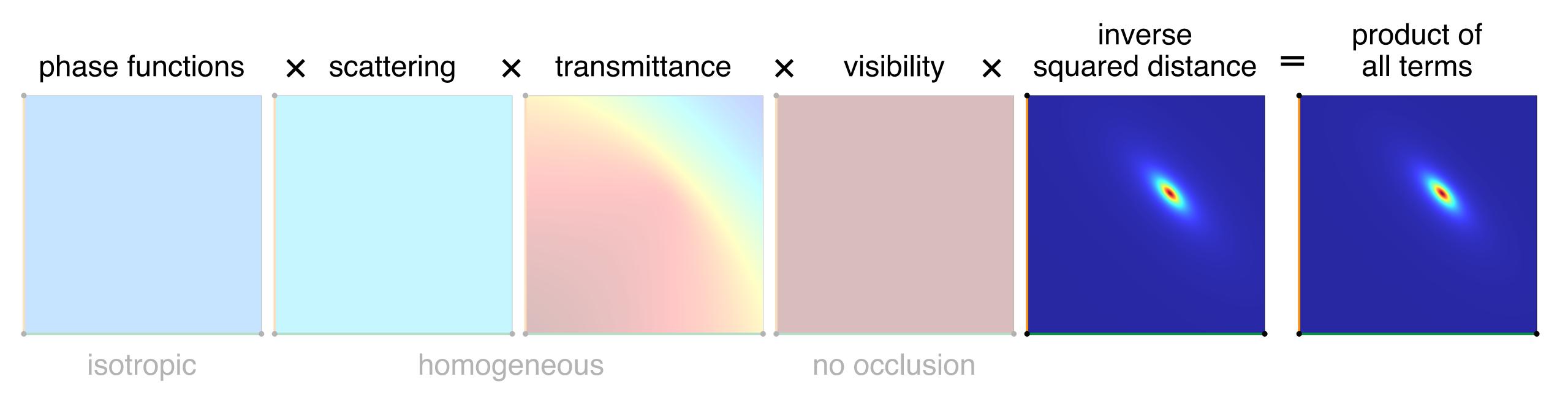


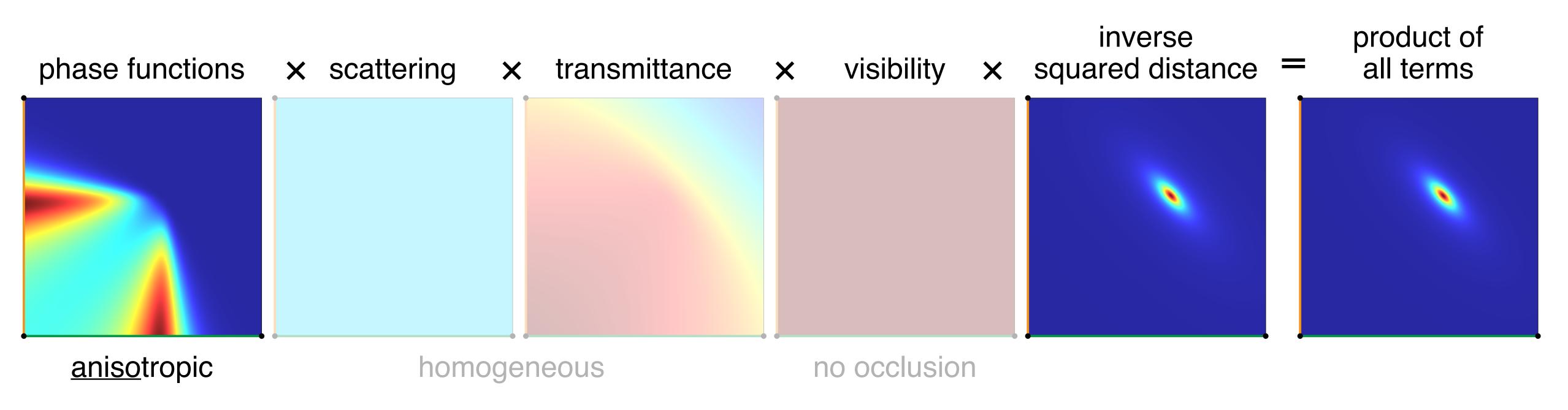
Summary of isotropic media:

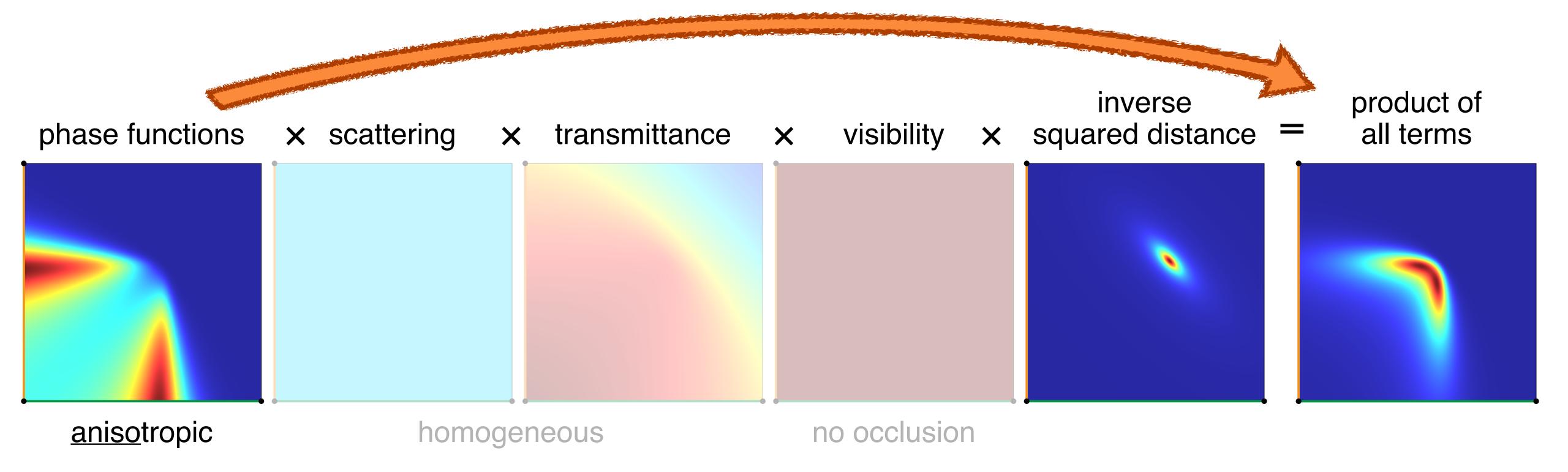
Summary of isotropic media:



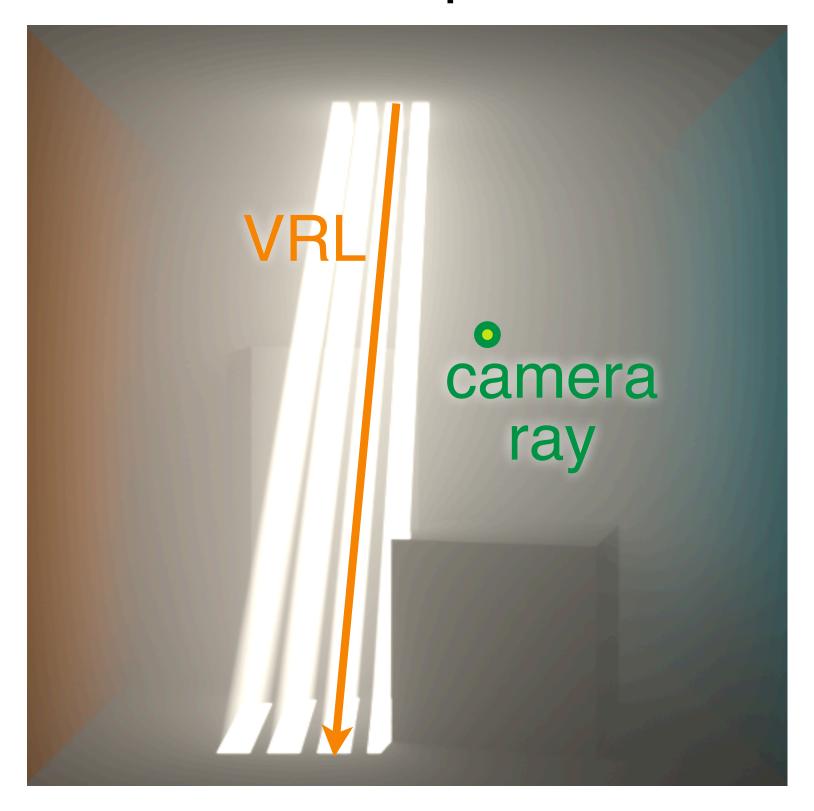
How to (importance) sample?



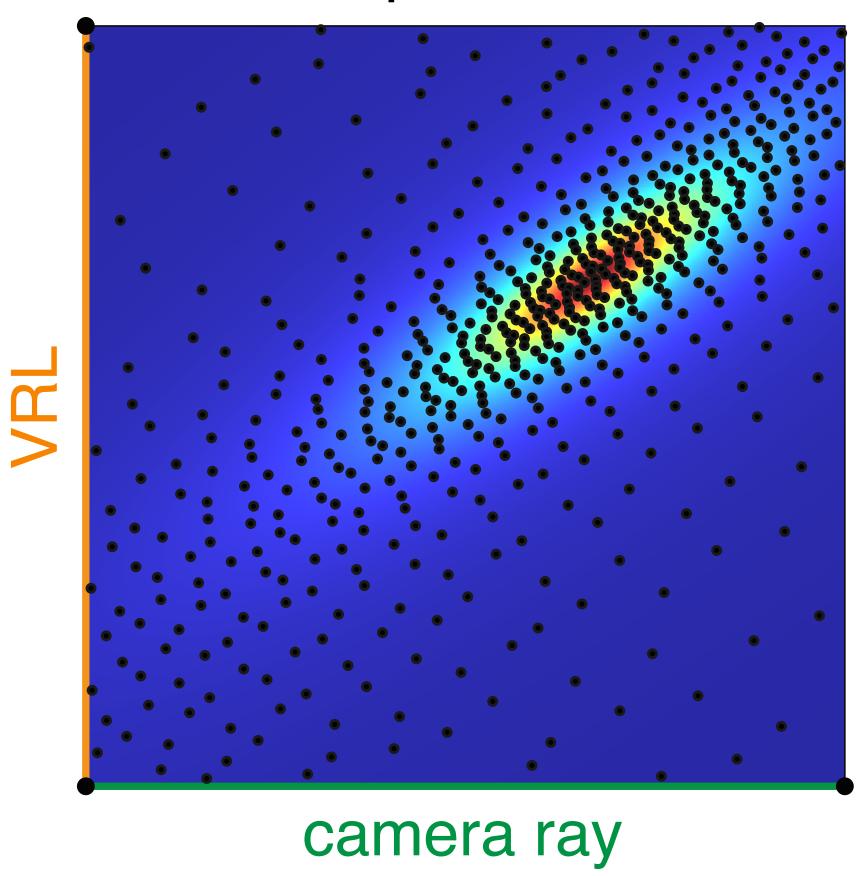




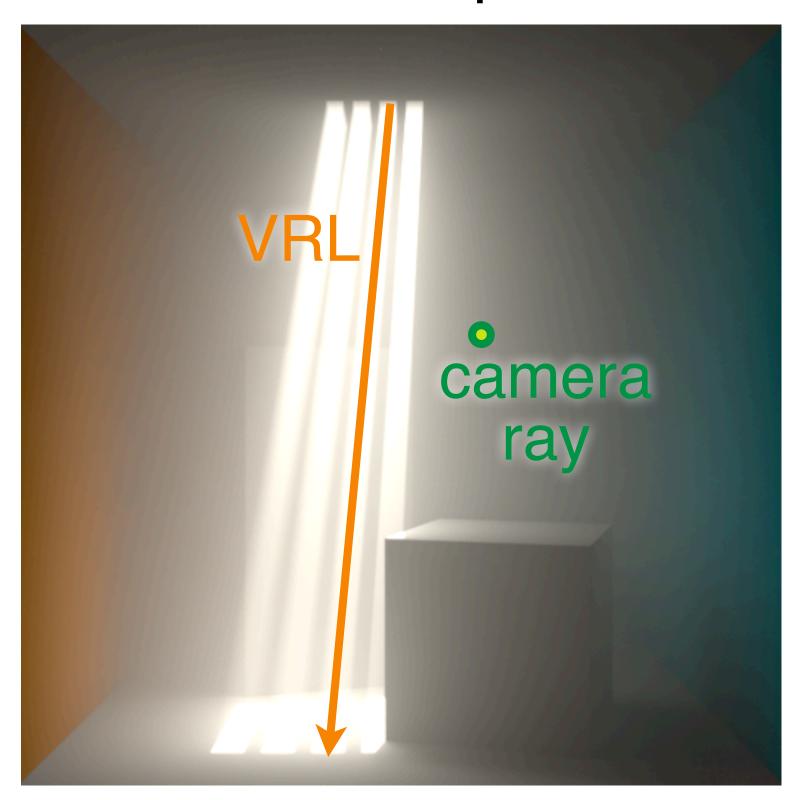
isotropic



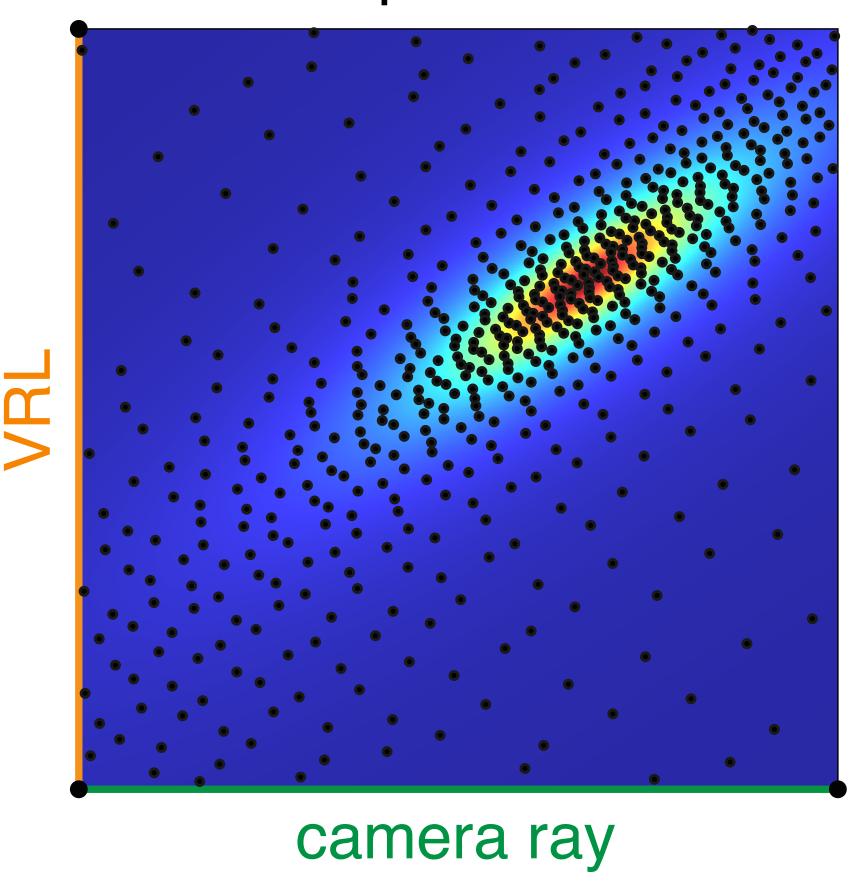
inverse squared distance



anisotropic



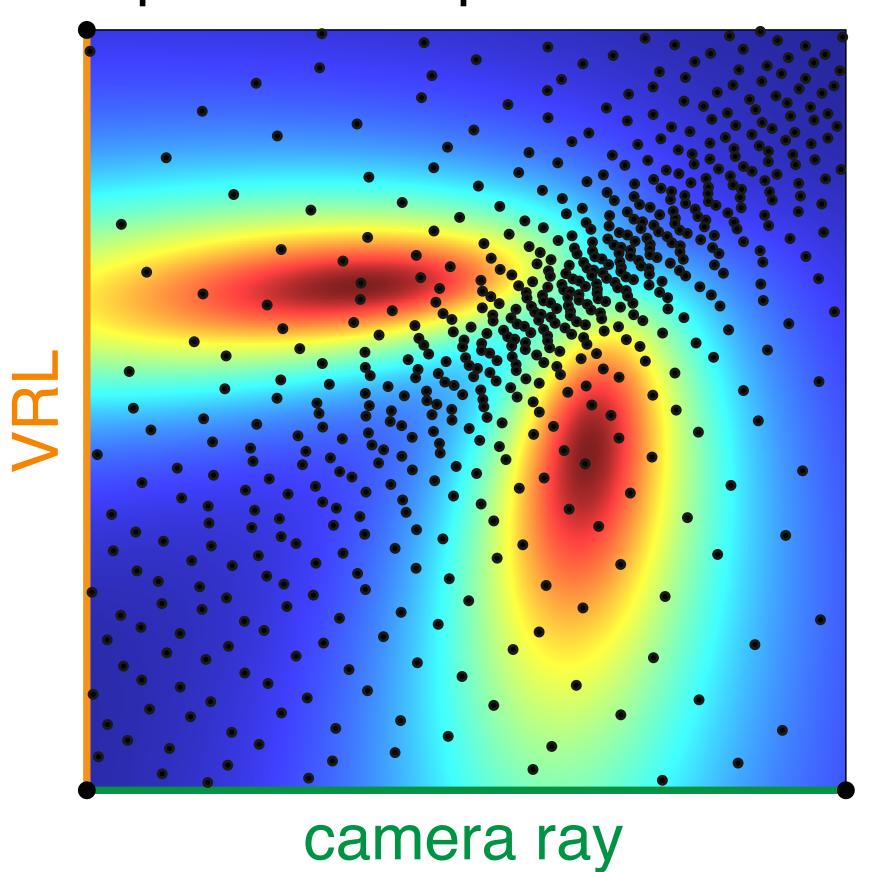
inverse squared distance



anisotropic

camera ray

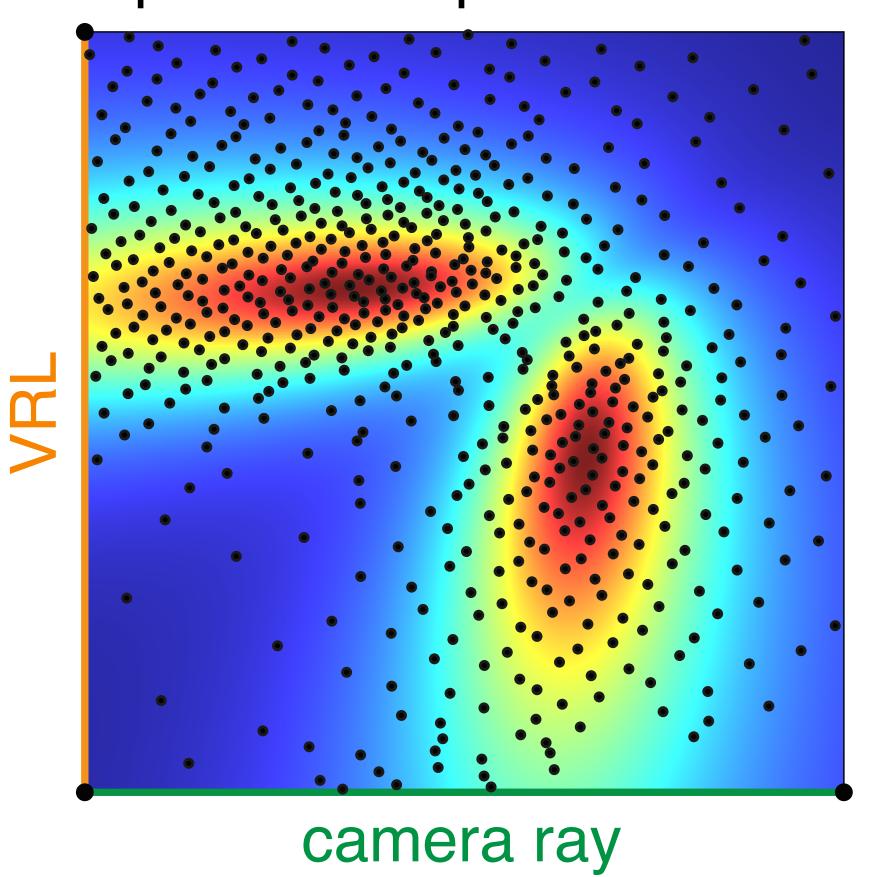
PF product / squared distance



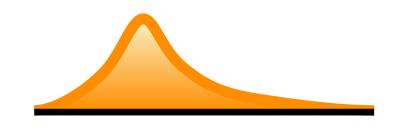
anisotropic

camera ray

PF product / squared distance

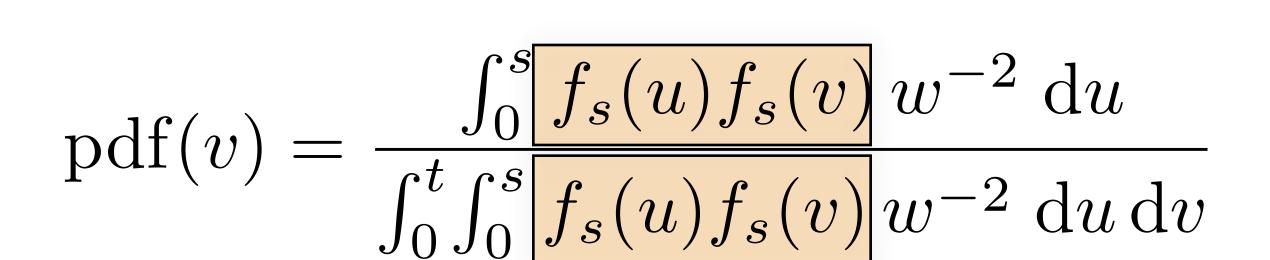


• Marginal PDF

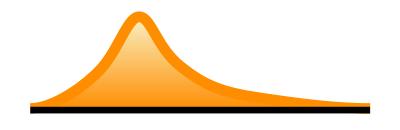


$$pdf(v) = \frac{\int_0^s f_s(u) f_s(v) w^{-2} du}{\int_0^t \int_0^s f_s(u) f_s(v) w^{-2} du dv}$$

• Marginal PDF

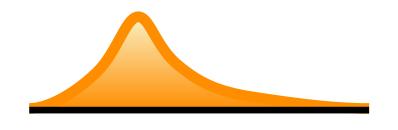


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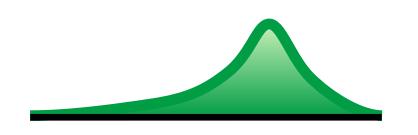
$$pdf(v) = \frac{\int_0^s f_s(u) f_s(v) w^{-2} du}{\int_0^t \int_0^s f_s(u) f_s(v) w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv}$$

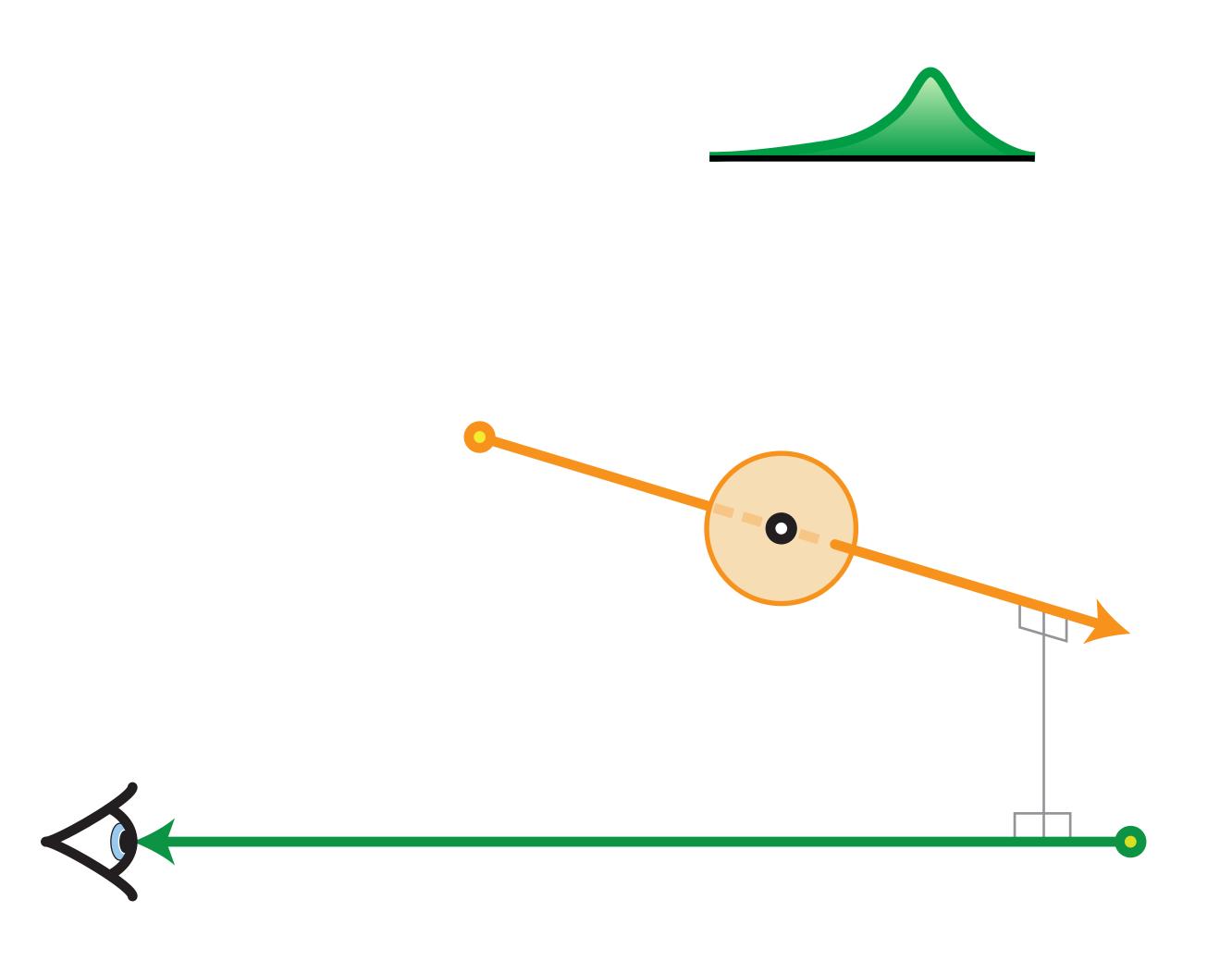
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$$pdf(v) = \frac{\int_0^s f_s(u) f_s(v) w^{-2} du}{\int_0^t \int_0^s f_s(u) f_s(v) w^{-2} du dv} \approx \frac{\int_{-\infty}^\infty w^{-2} du}{\int_0^t \int_{-\infty}^\infty w^{-2} du dv}$$

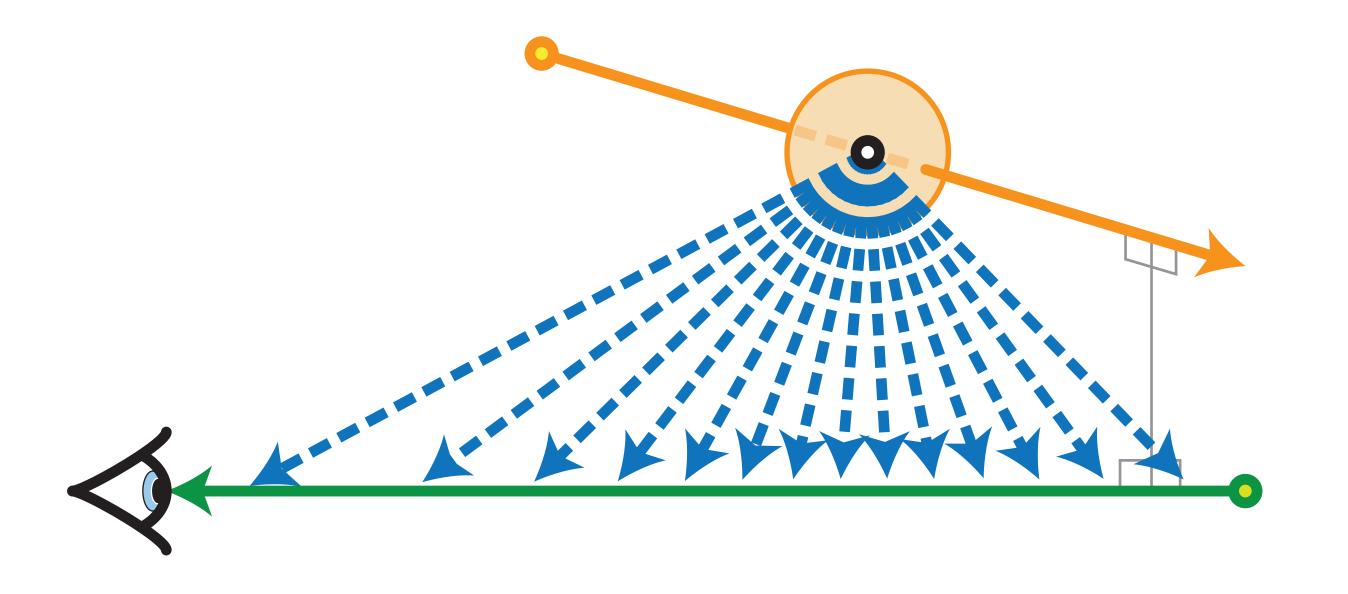
identical to isotropic medium





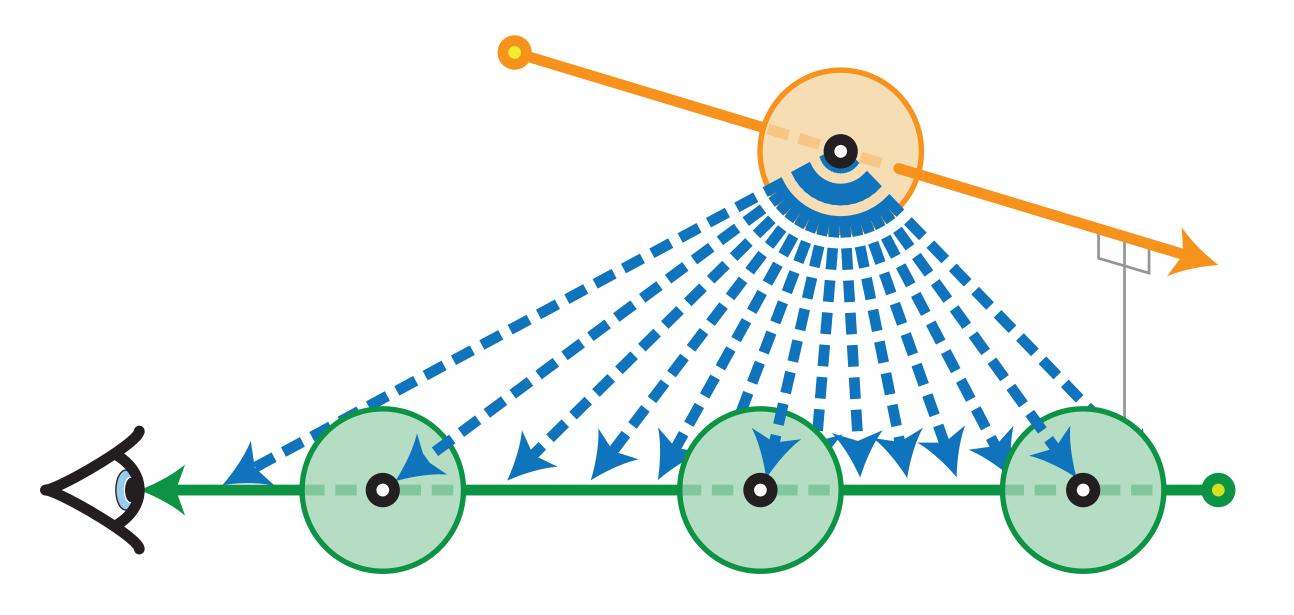
isotropic ~ equi-angular





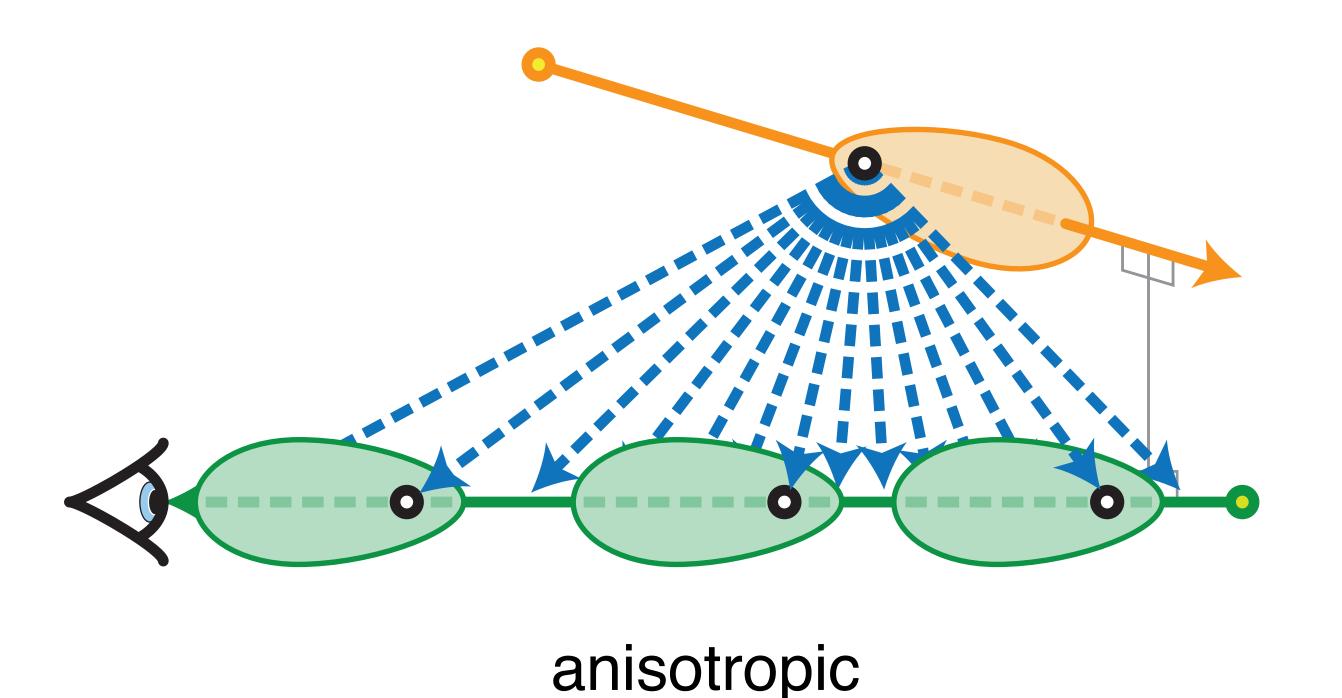
isotropic ~ equi-angular



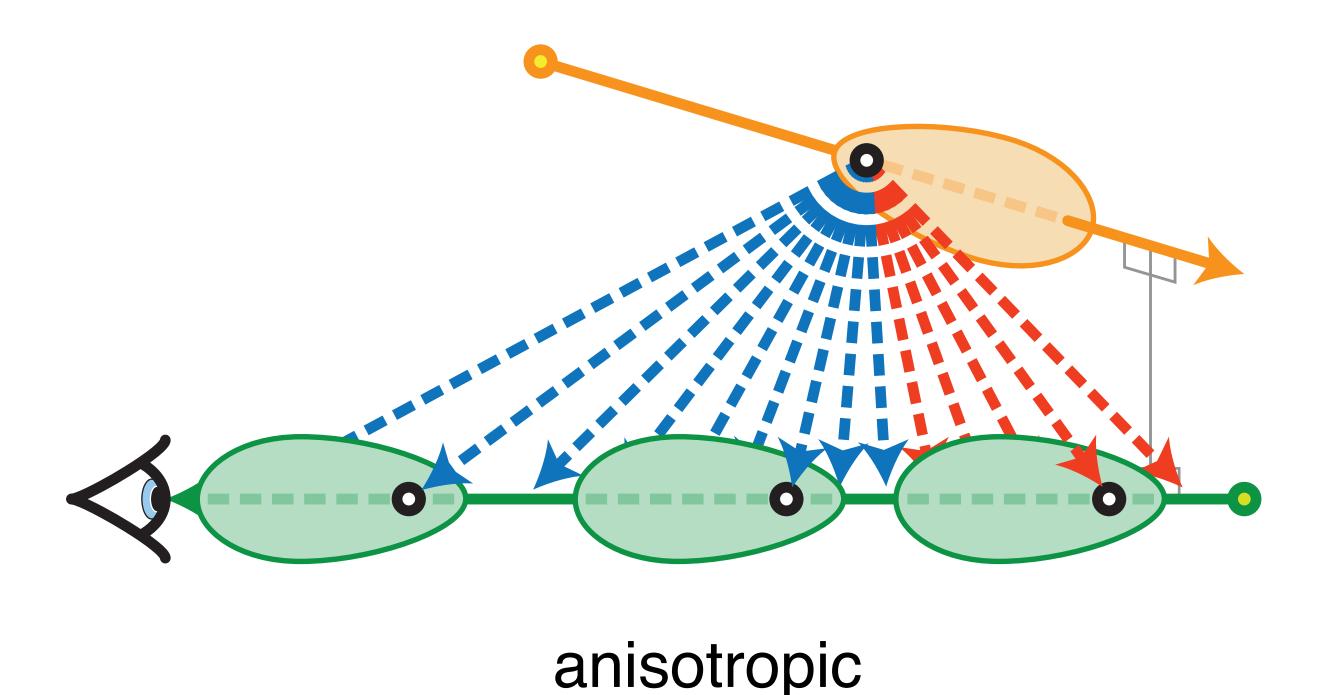


isotropic ~ equi-angular

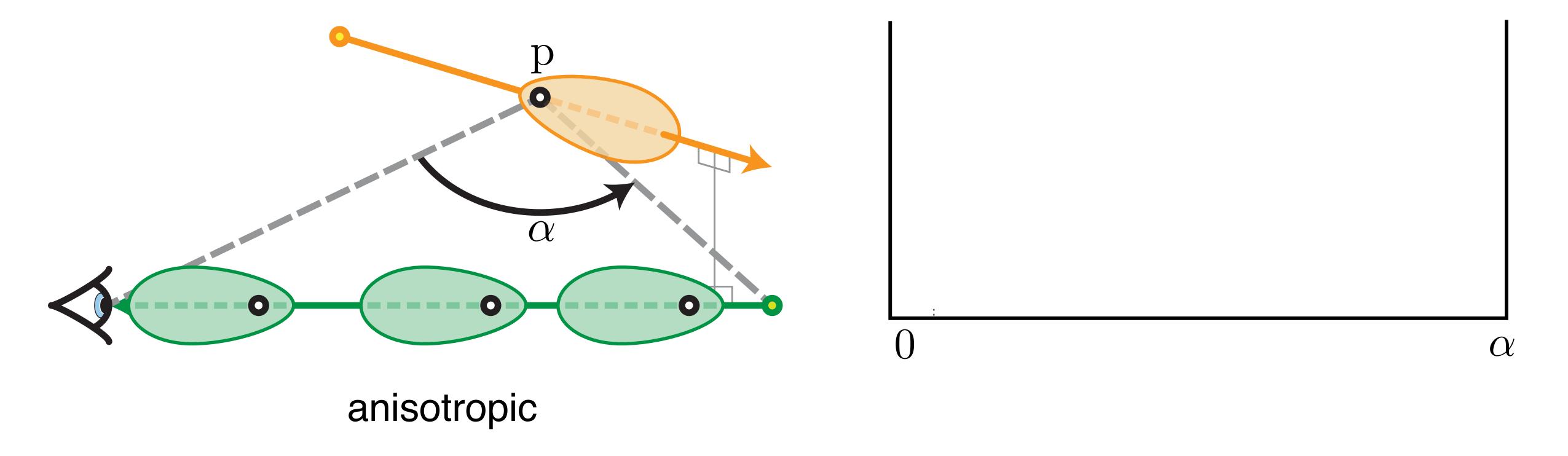
- Conditional PDF
 - replace equi-angular sampling by importance sampling the PF product



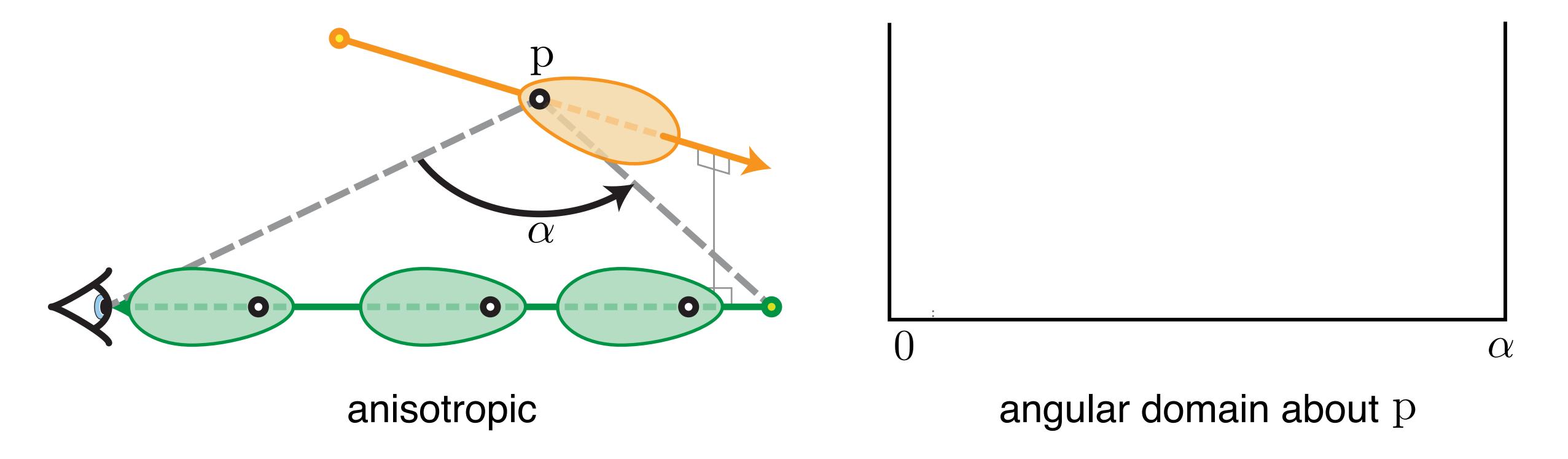
- Conditional PDF
 - replace equi-angular sampling by importance sampling the PF product



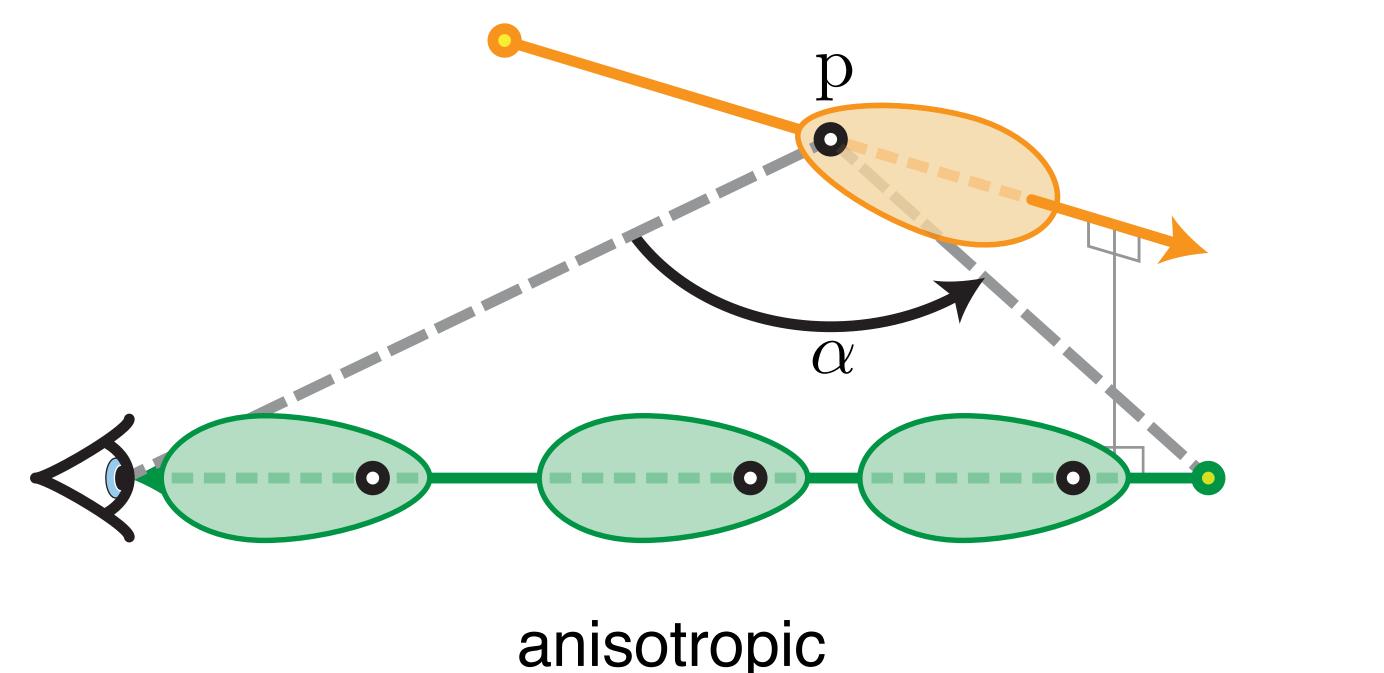
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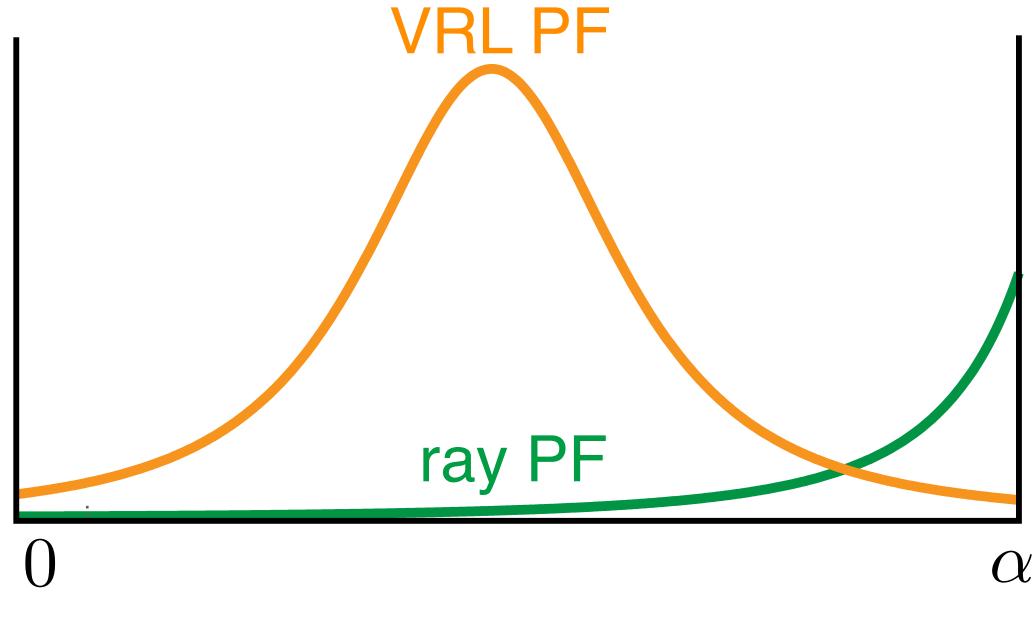


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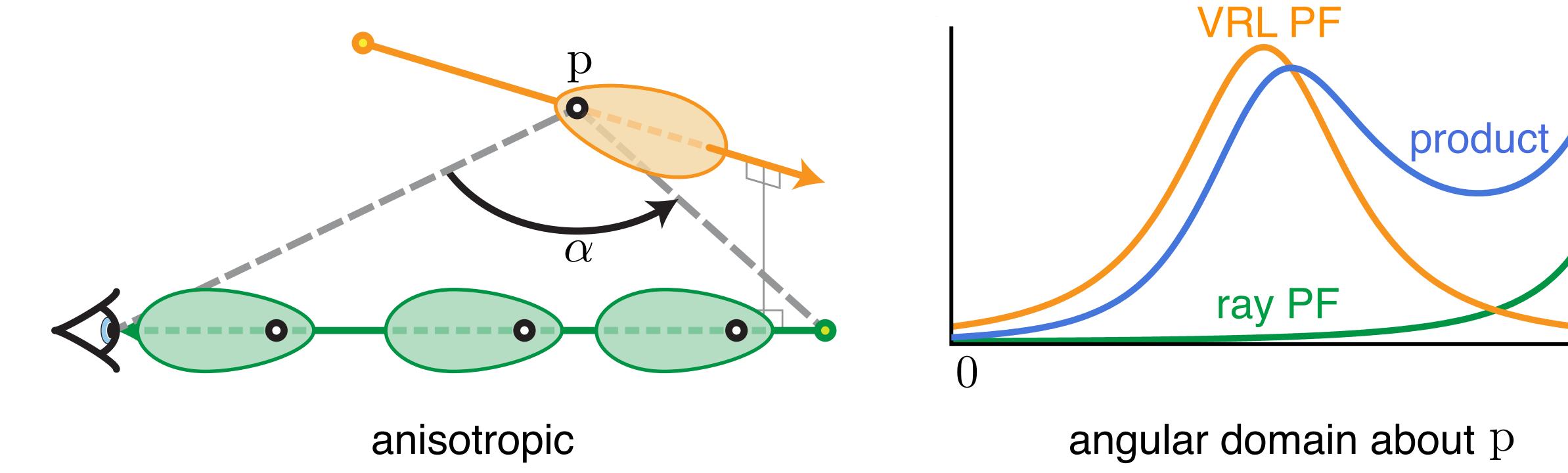
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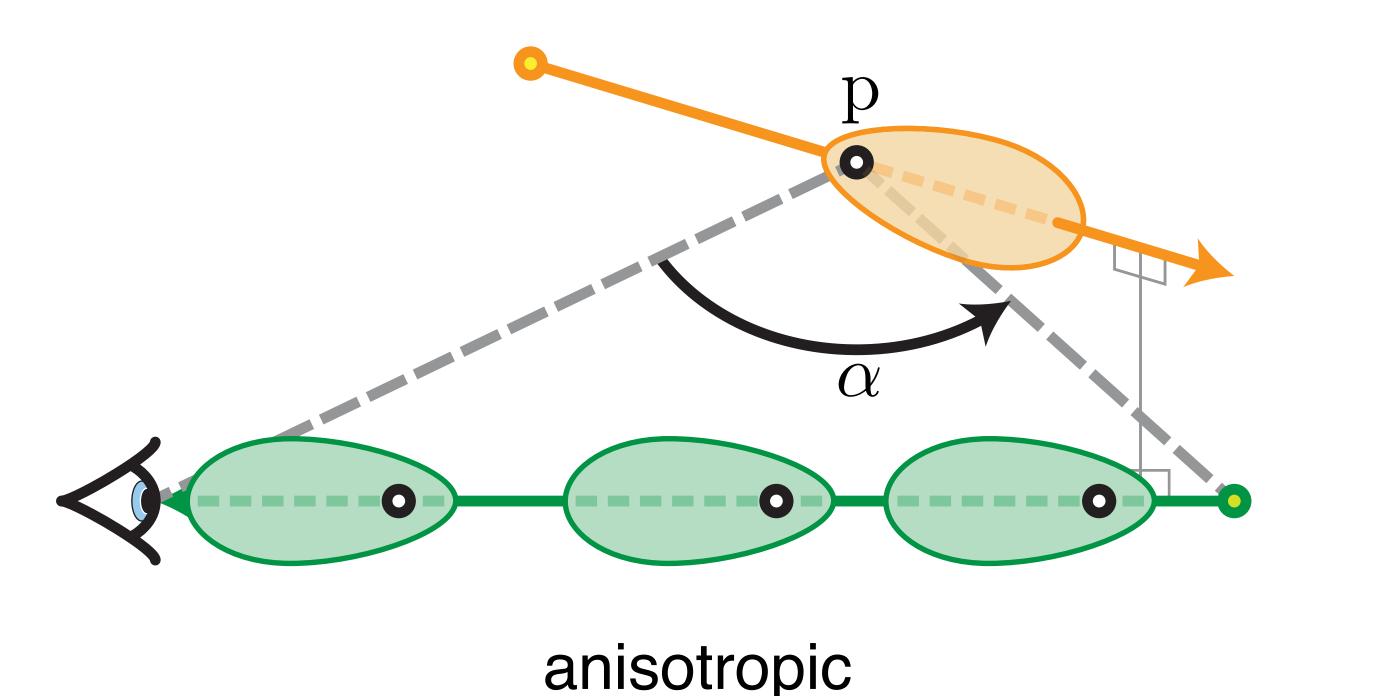


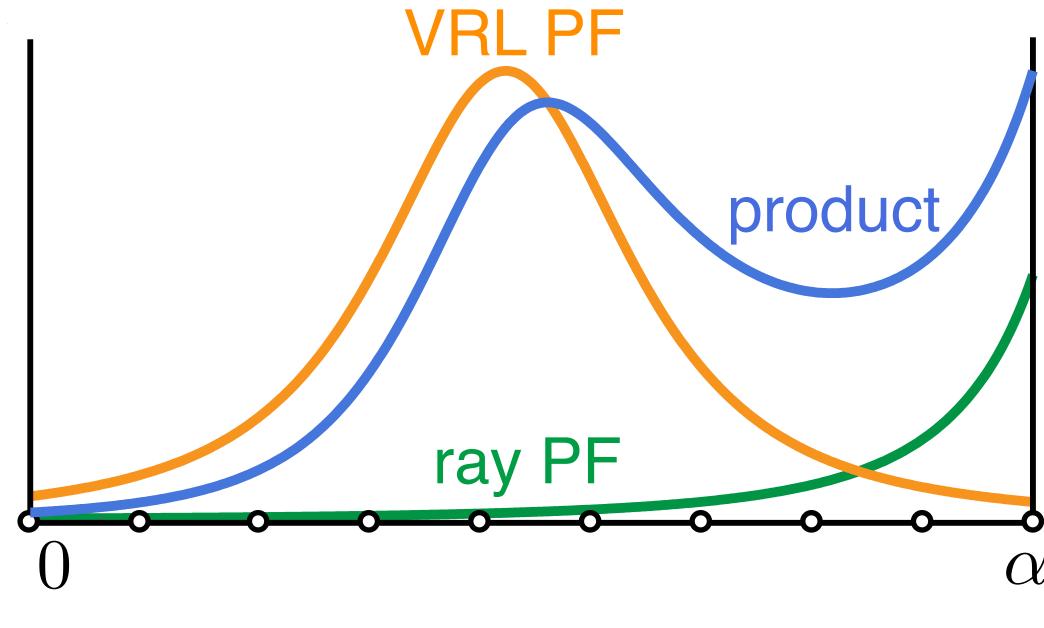
angular domain about p

• Conditional PDF



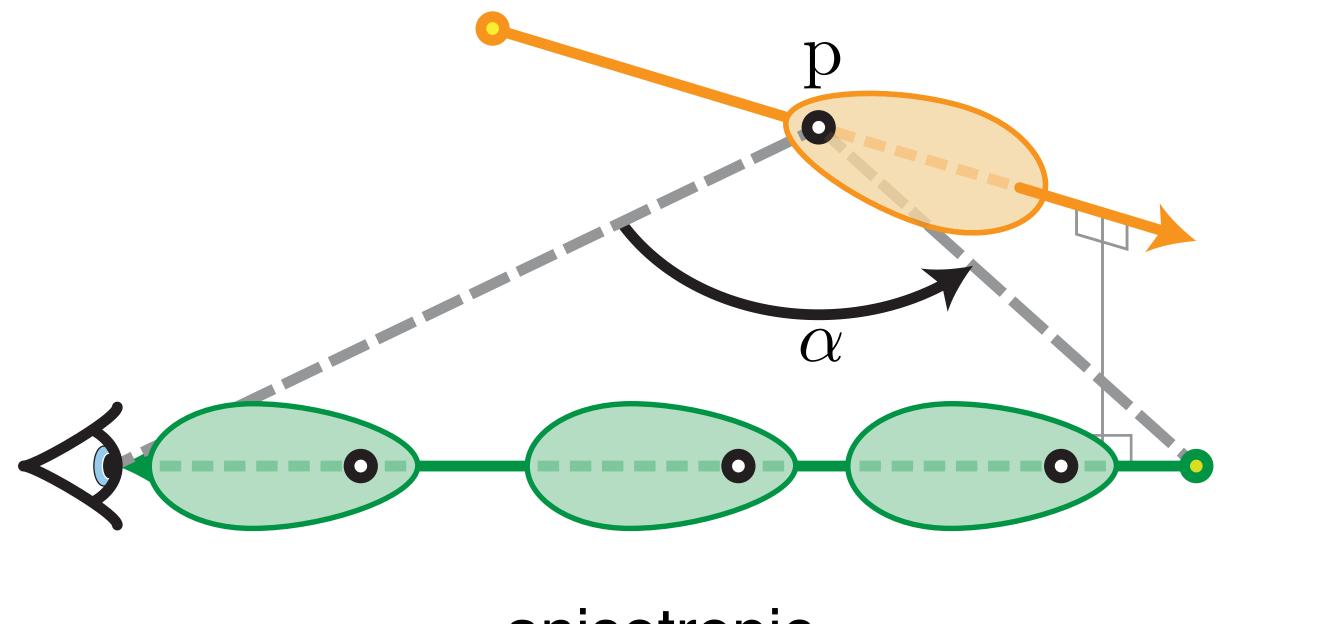
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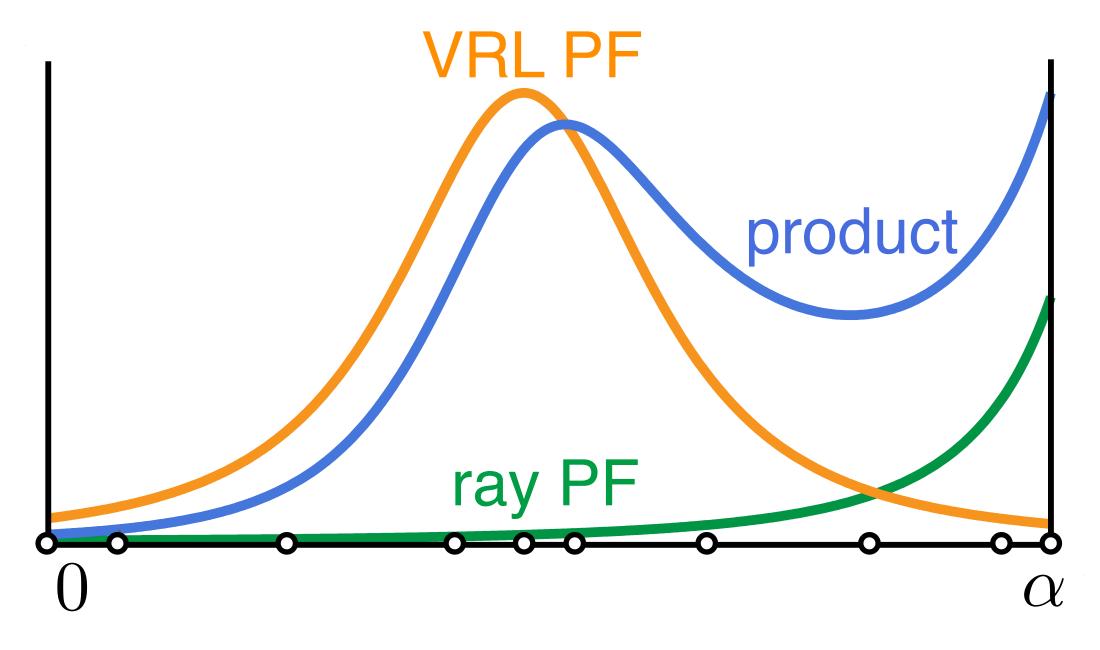


angular domain about p

Conditional PDF

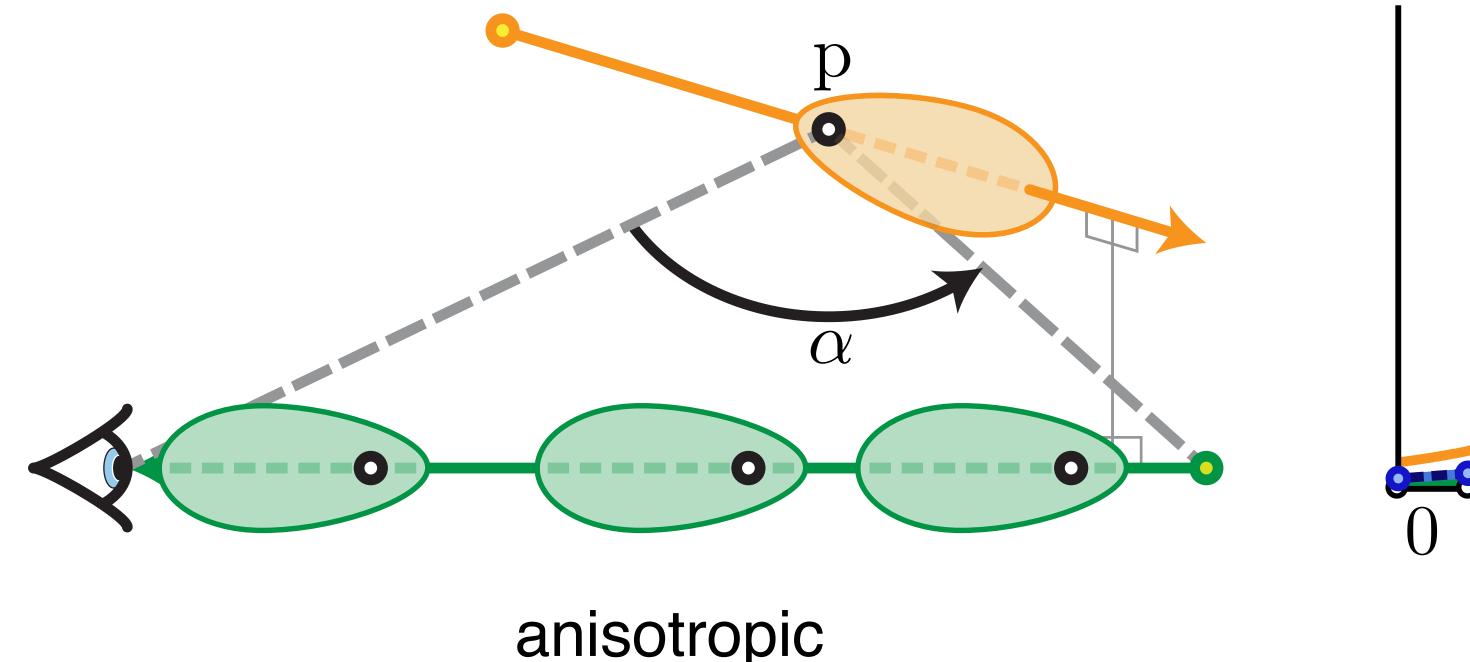


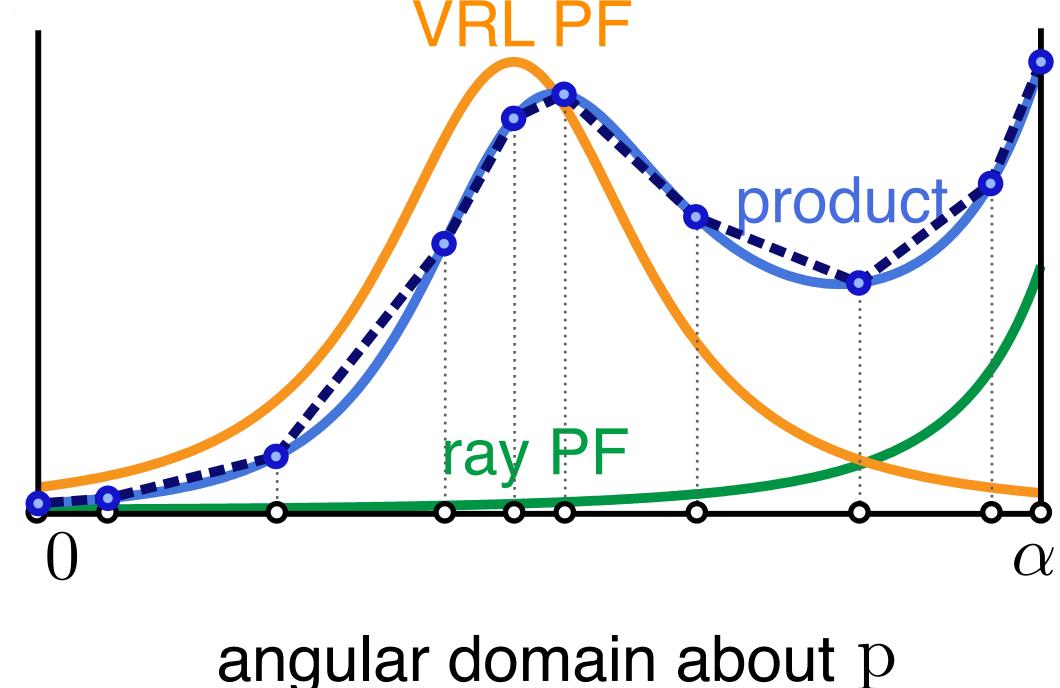
anisotropic



angular domain about p

Conditional PDF

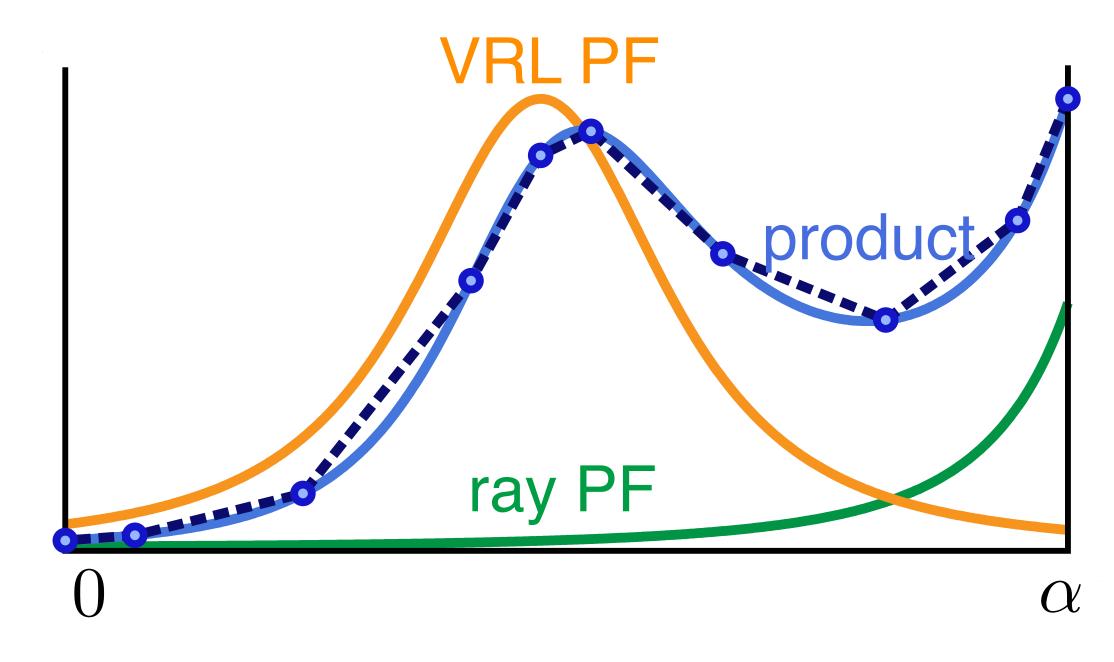




Conditional PDF

replace equi-angular sampling by importance sampling the PF product

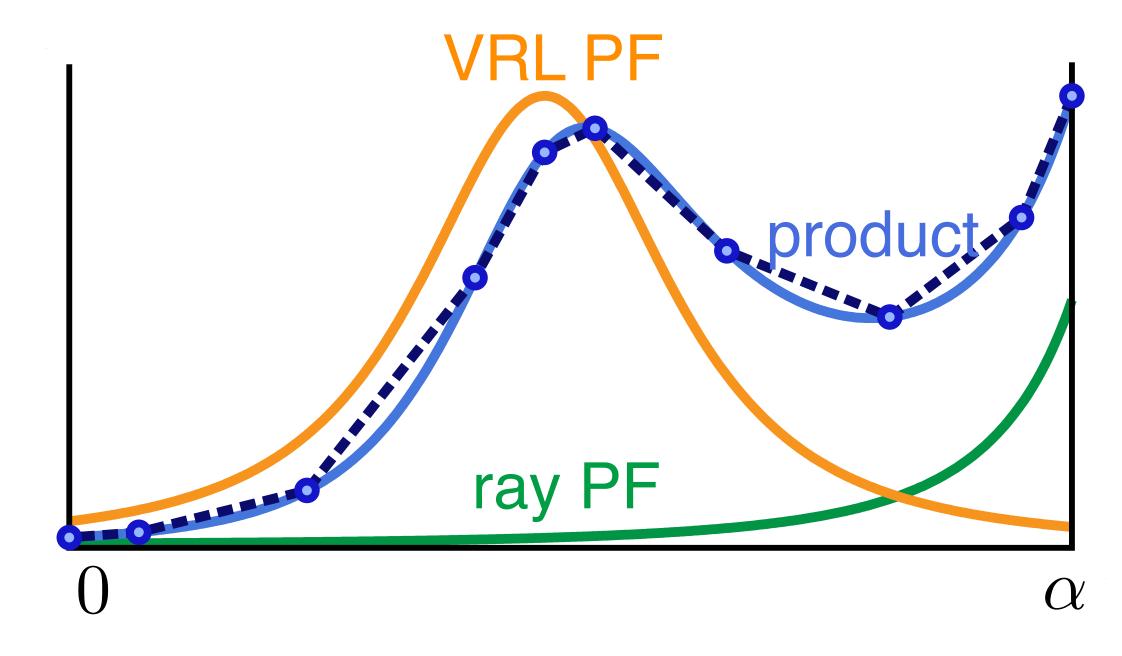
piece-wise linear PDF



angular domain about p

Conditional PDF

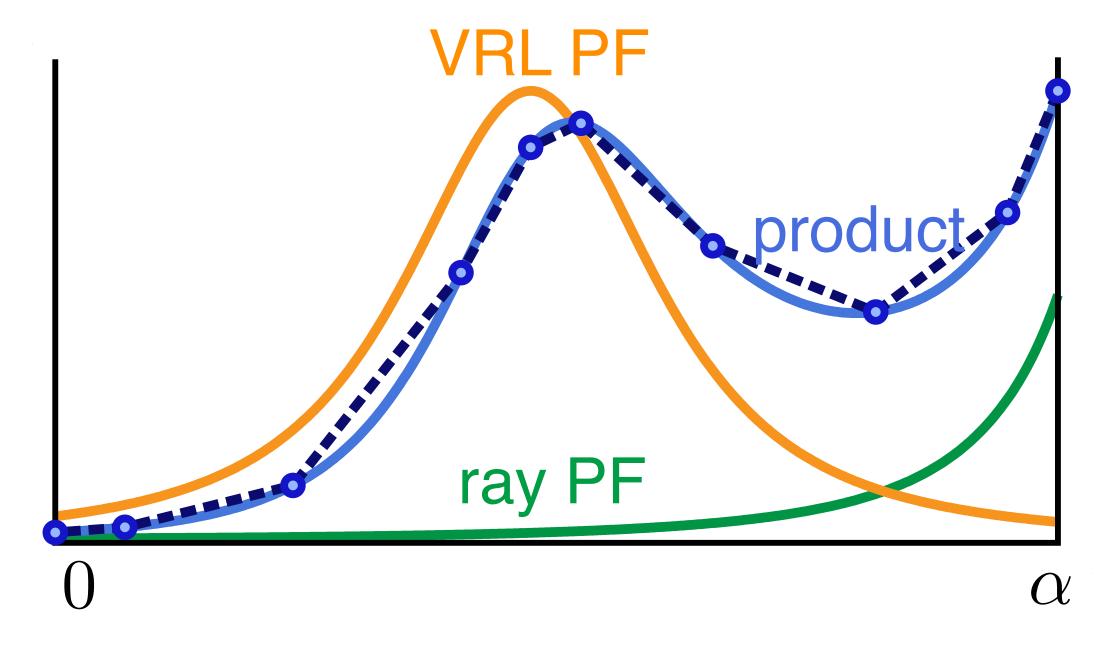
- piece-wise linear PDF
- piece-wise quadratic CDF



angular domain about p

Conditional PDF

- piece-wise linear PDF
- piece-wise quadratic CDF
- ▶ 10 adaptively distributed vertices balance between speed and quality



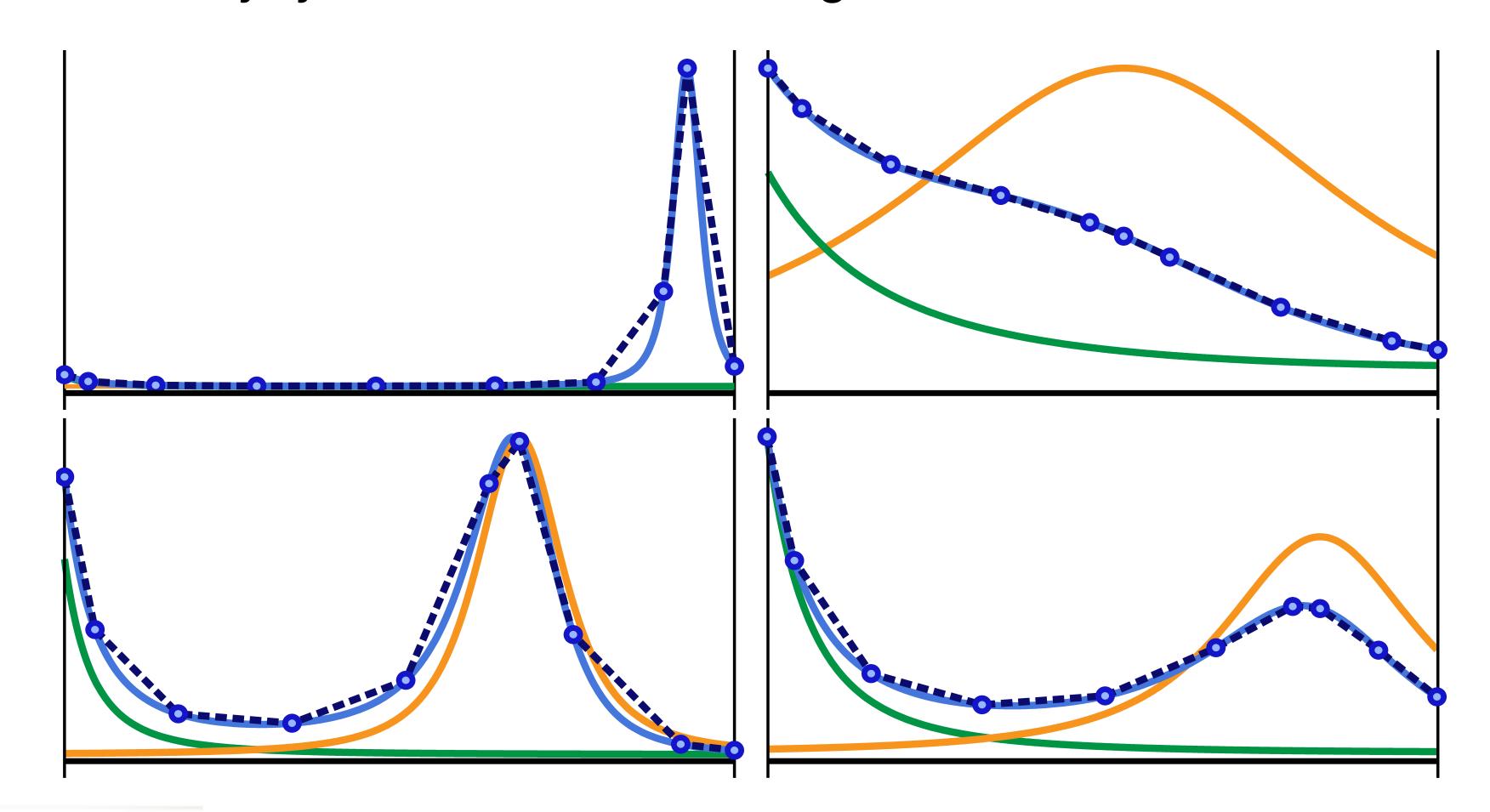
angular domain about p

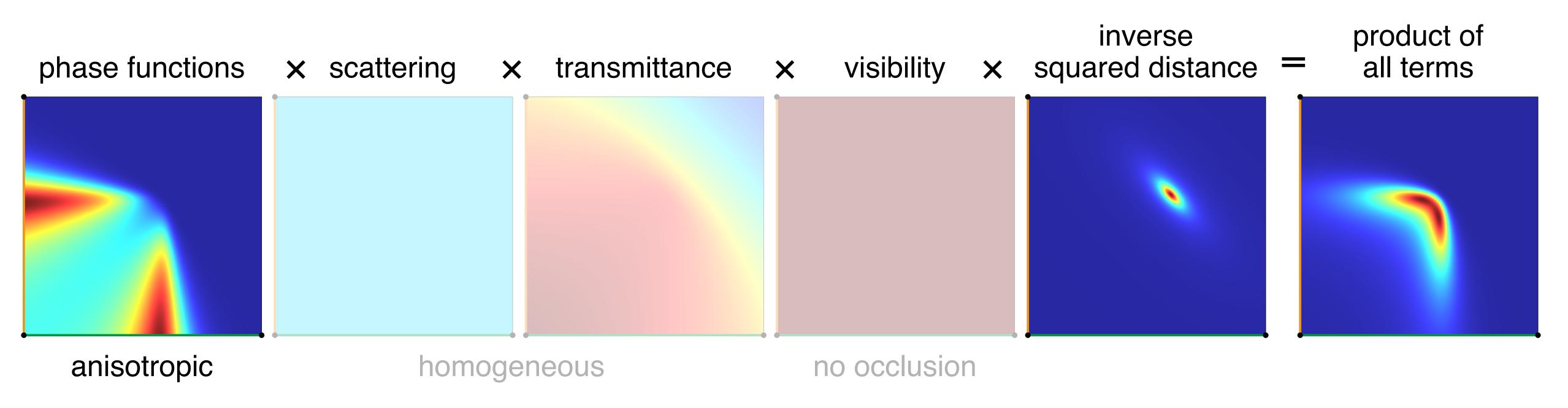
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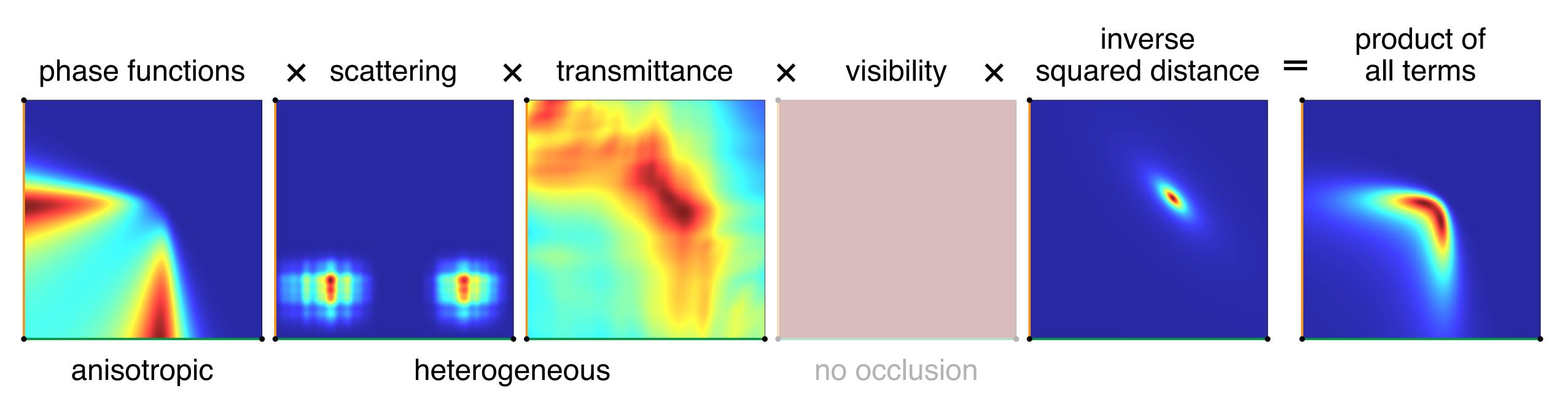
examples for Henyey-Greenstein PF with g = 0.95

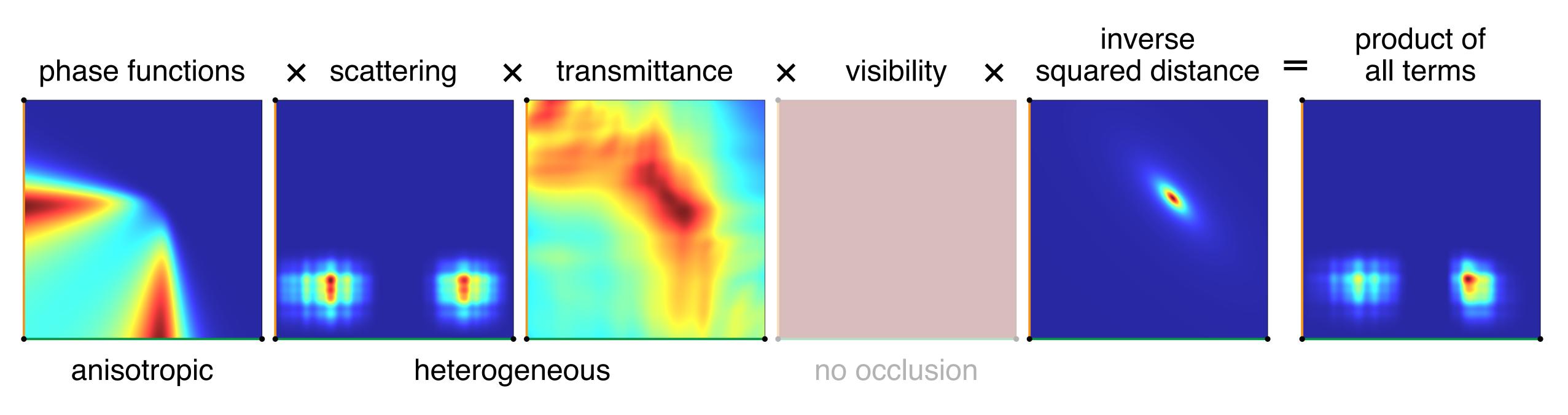
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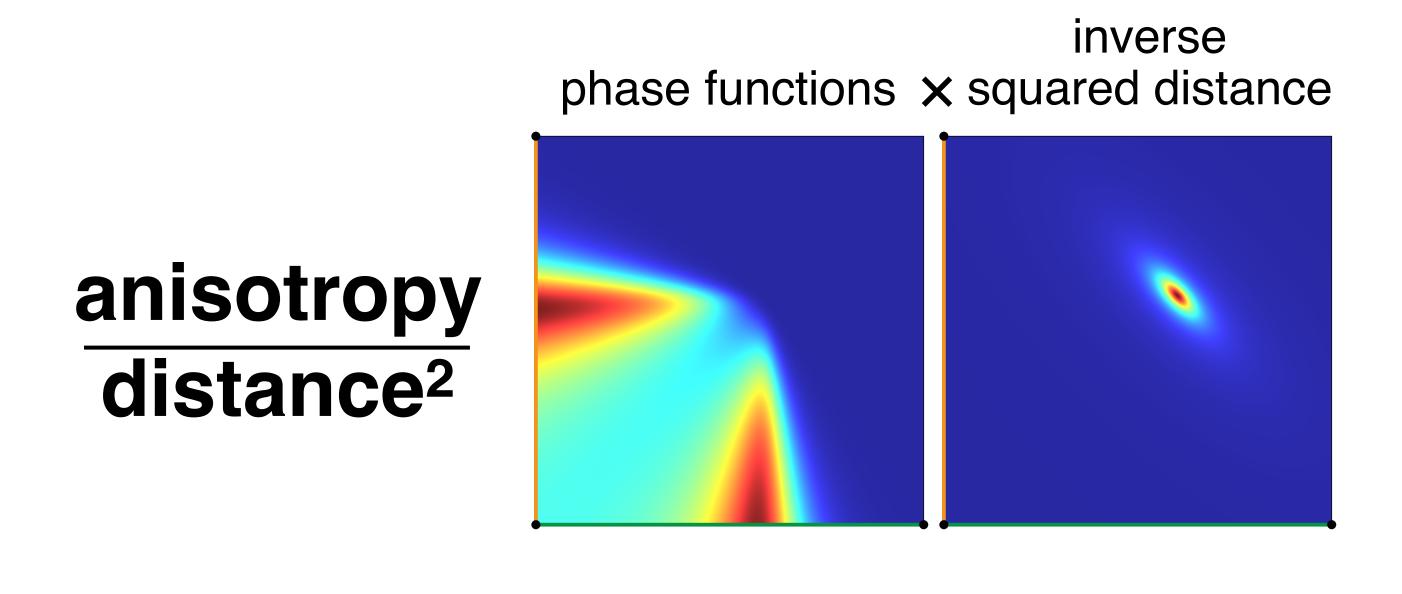
examples for Henyey-Greenstein PF with g = 0.95



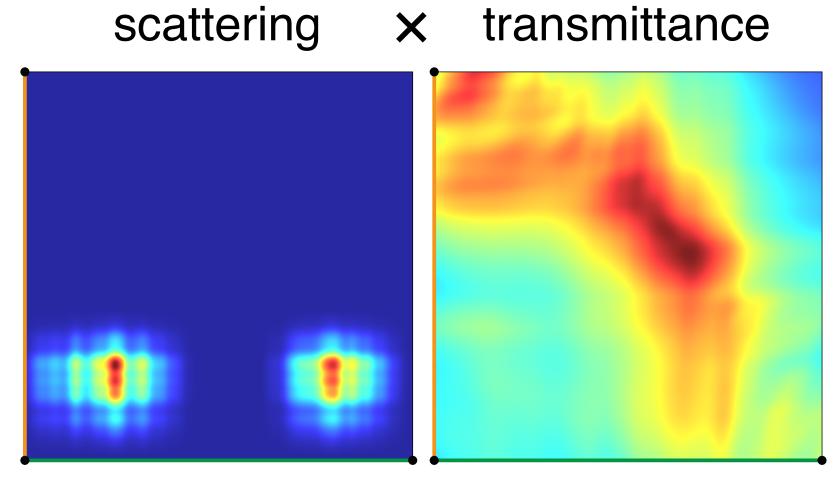


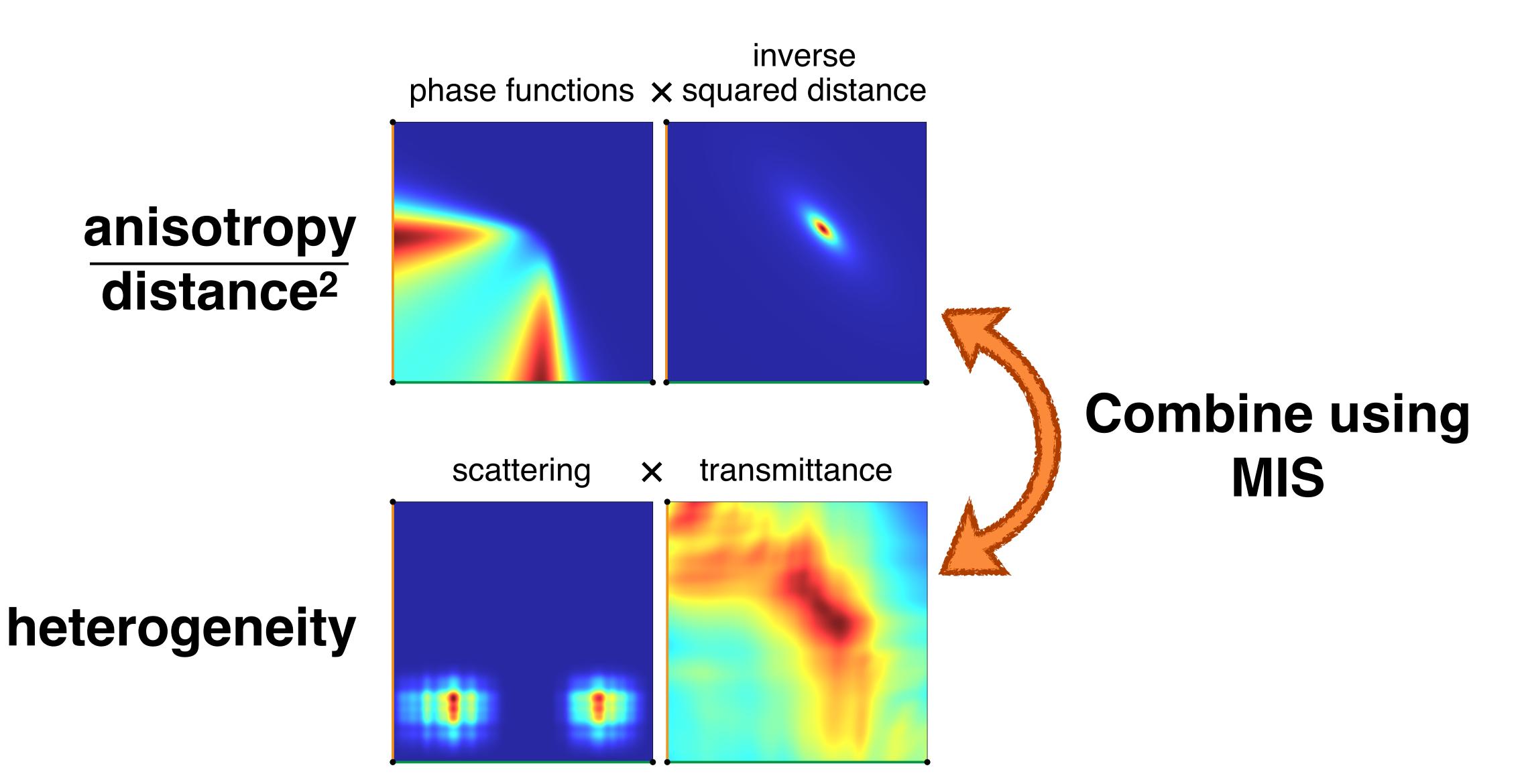












Heterogeneity

Heterogeneity

$$pdf(u, v) = \sigma_s(u) T(u) \sigma_s(v) T(v) T(u, v)$$

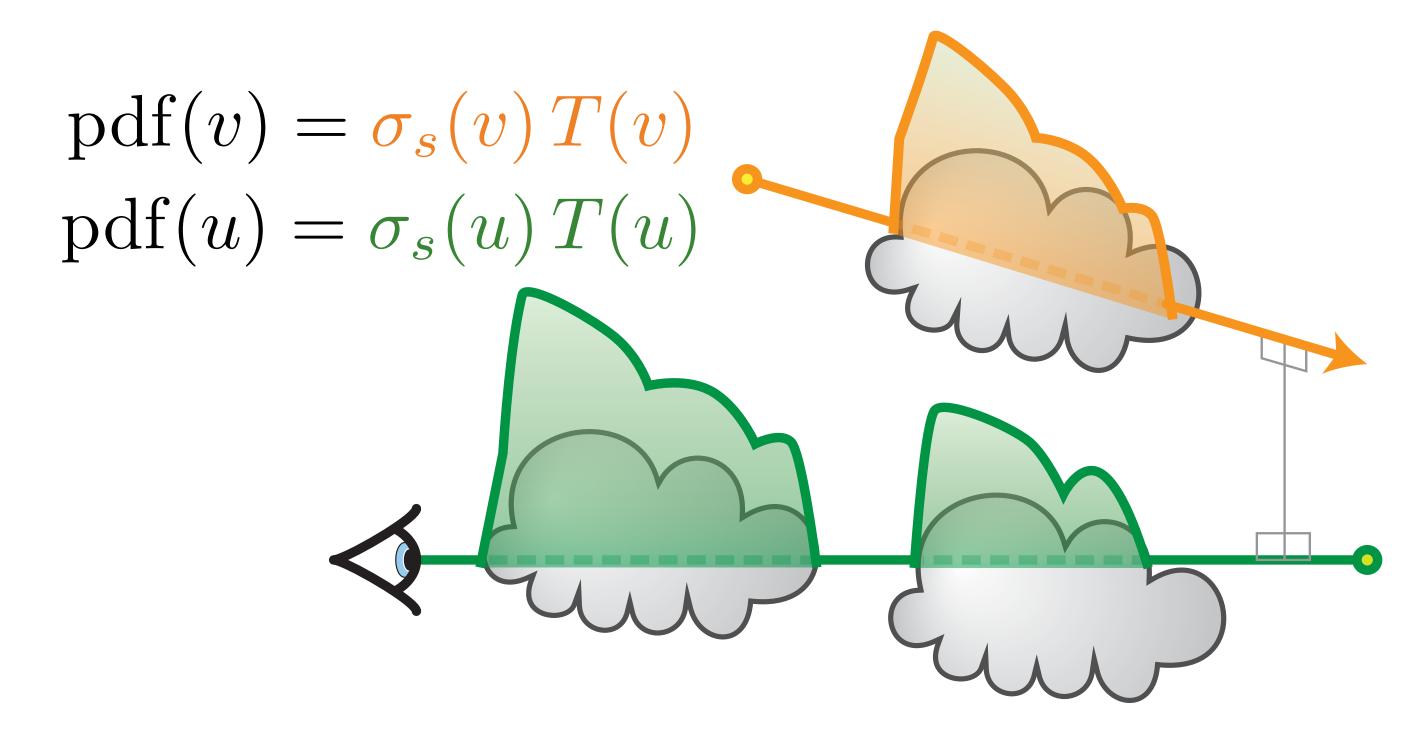
$$pdf(u,v) = \sigma_s(u) T(u) \sigma_s(v) T(v) T(u,v)$$

$$\mathrm{pdf}(u,v) = \underbrace{\sigma_s(u)\,T(u)}_{\text{along camera}} \underbrace{\sigma_s(v)\,T(v)}_{\text{along VRL}} T(u,v)$$

Separable!

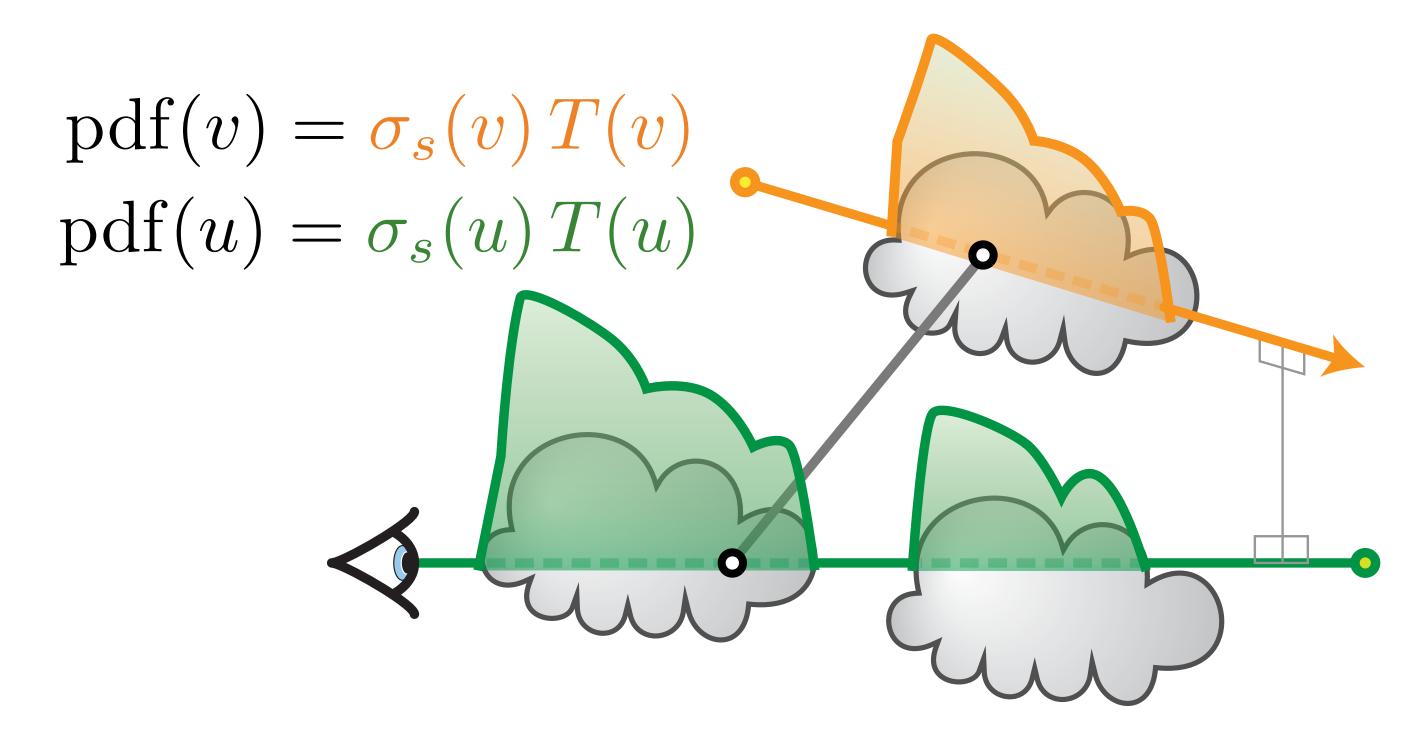
$$\mathrm{pdf}(u,v) = \underbrace{\sigma_s(u)\,T(u)}_{\text{along camera}} \underbrace{\sigma_s(v)\,T(v)}_{\text{along VRL}} T(u,v)$$

Separable!



$$\mathrm{pdf}(u,v) = \underbrace{\sigma_s(u)\,T(u)}_{\text{along camera}} \underbrace{\sigma_s(v)\,T(v)}_{\text{along vRL}} T(u,v)$$

Separable!



Analysis and Results



Fruit Juice

homogeneous anisotropic (HG g = 0.55) 512x512



Multiple Scattering

Virtual Ray Lights



4K VRLs



4K VPLs

Multiple Scattering

Virtual Ray Lights



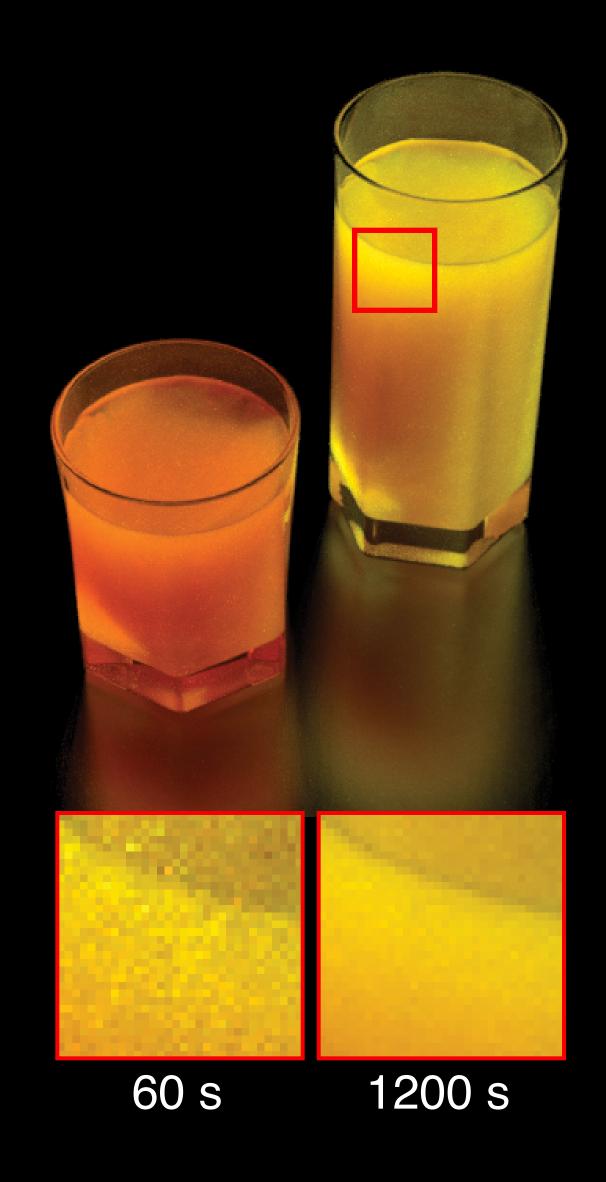
4K VRLs

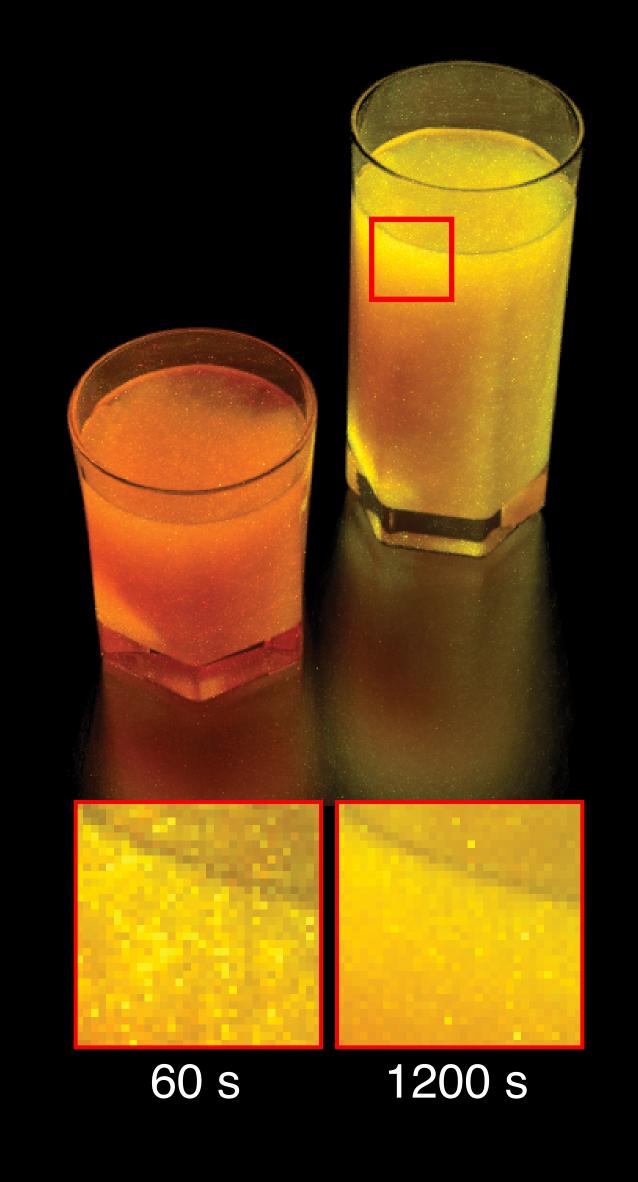


4K VPLs

Multiple Scattering

Virtual Ray Lights







Smoky Room

heterogeneous 1280x720

Virtual Ray Lights



6K VRLs



8K VPLs

Virtual Ray Lights



6K VRLs



8K VPLs

Virtual Ray Lights

101 seconds

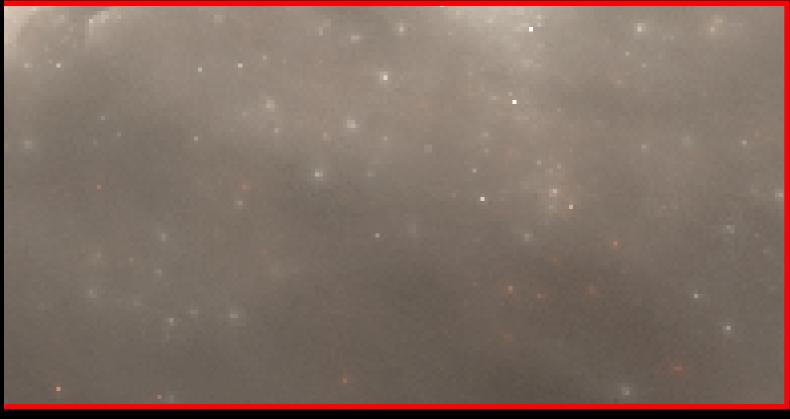




Virtual Point Lights

102 seconds





Media-to-Surface

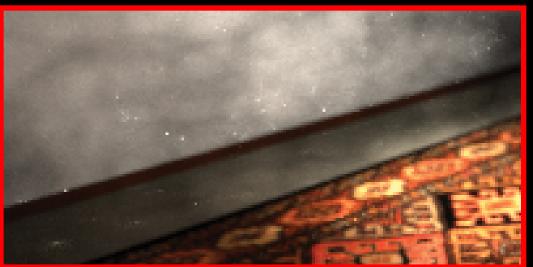
Virtual Ray Lights

600 seconds

Virtual Point Lights

600 seconds







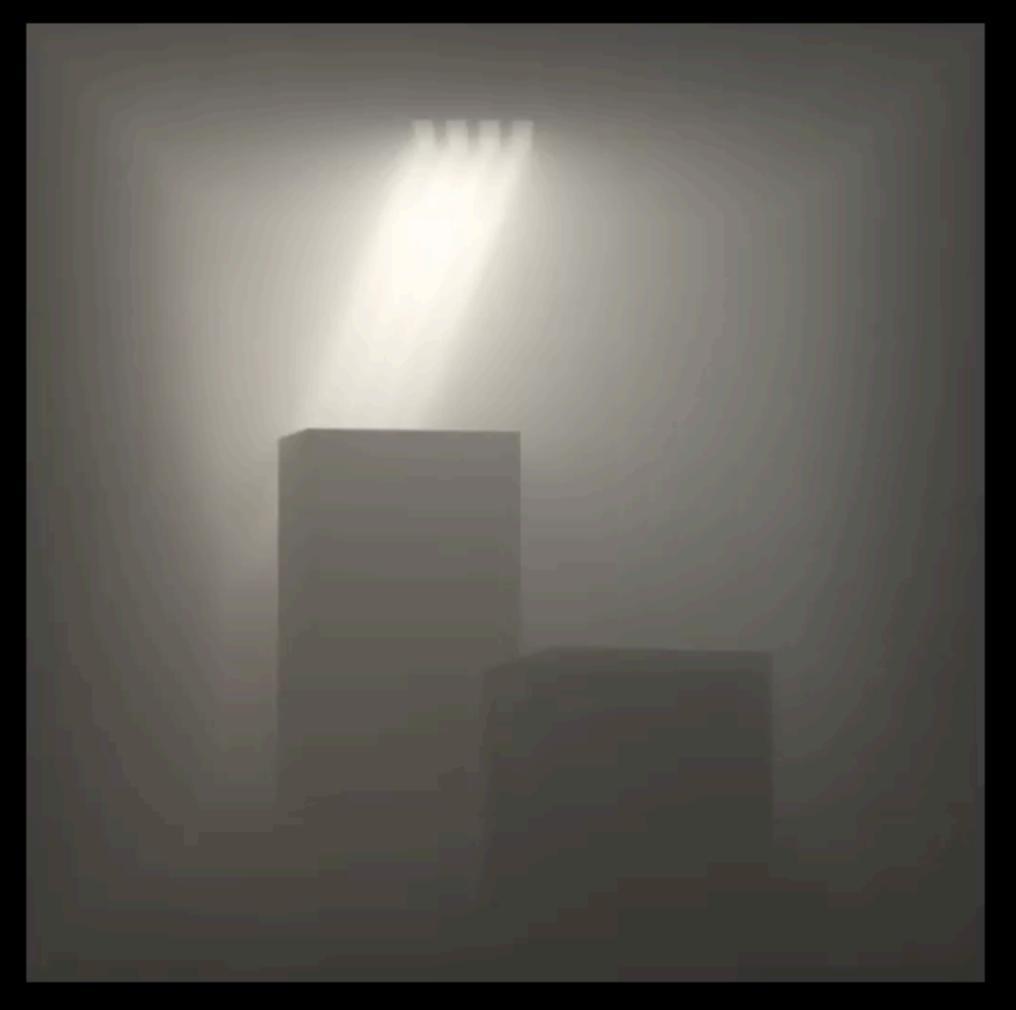




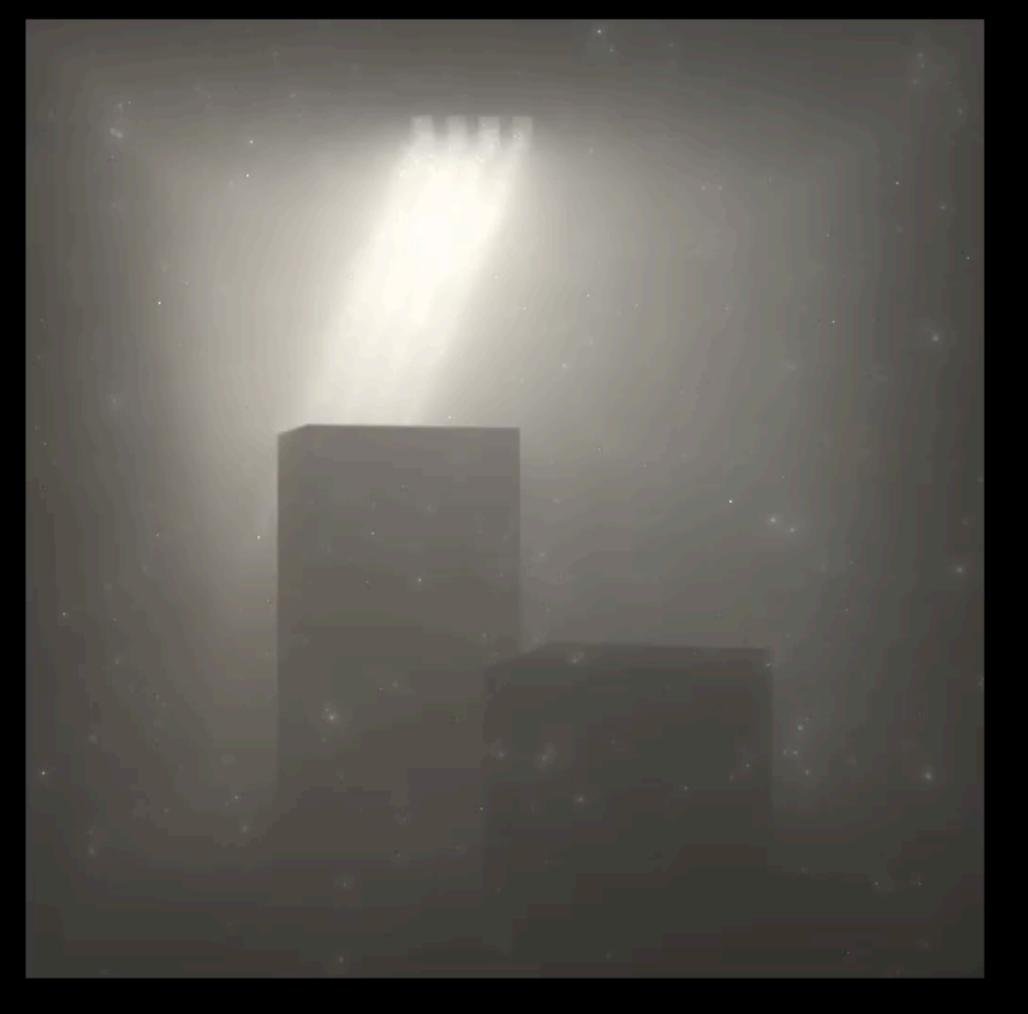


Temporal coherence VPLs vs. VRLs

Virtual Ray Lights

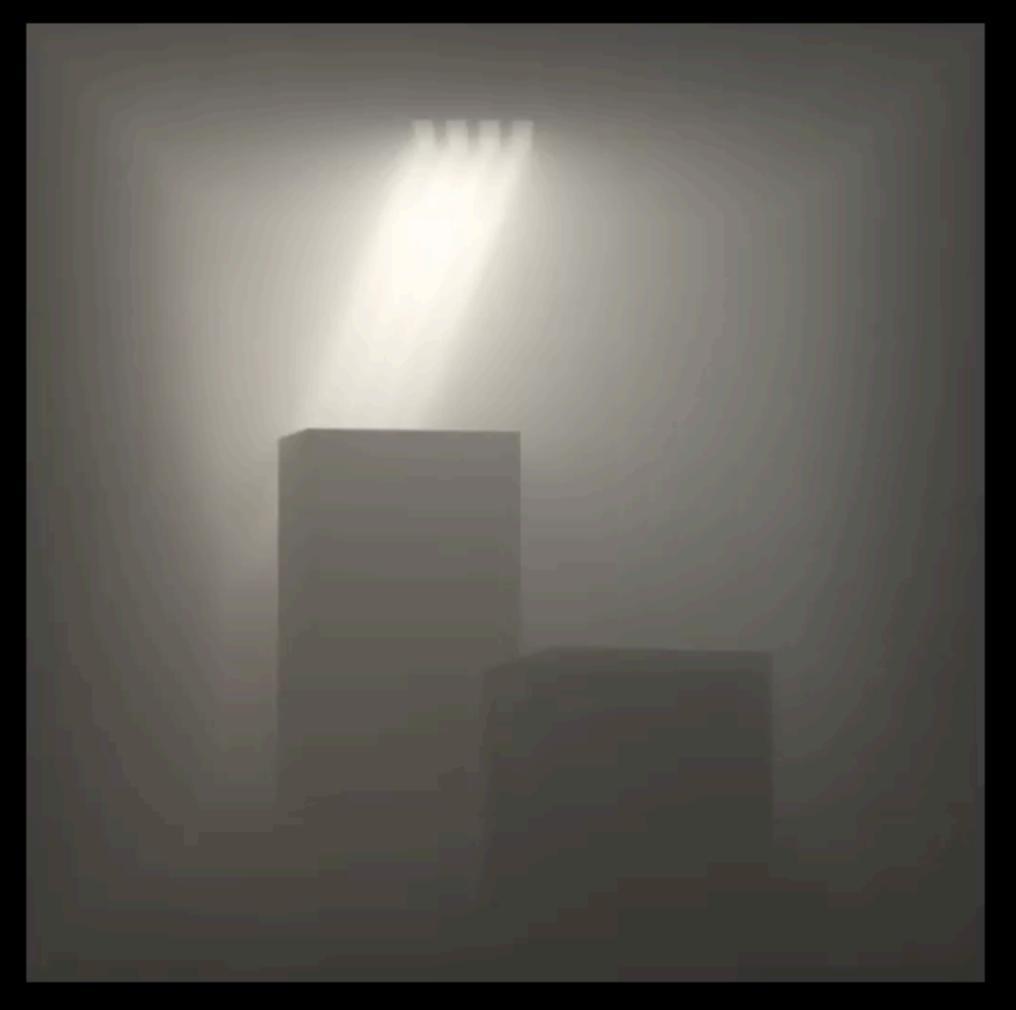


1 minute/frame

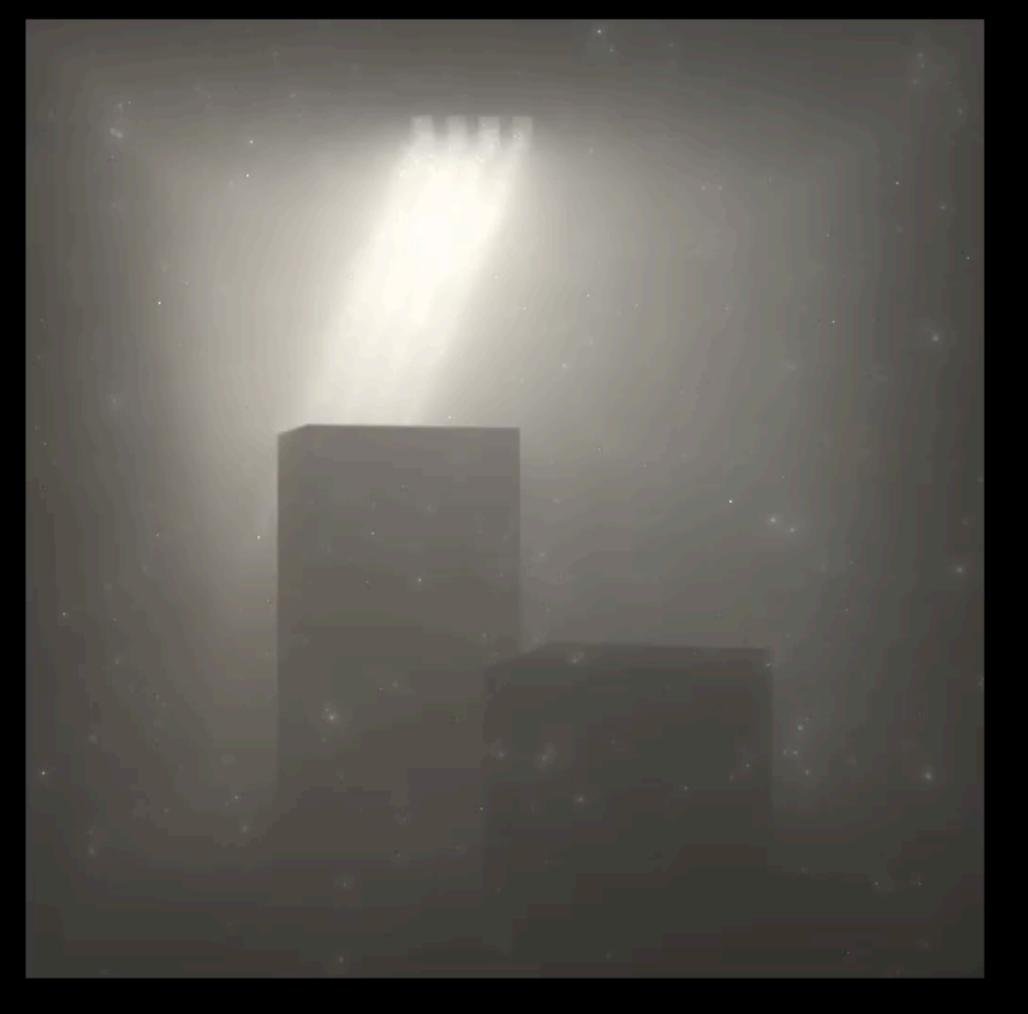


1 minute/frame

Virtual Ray Lights



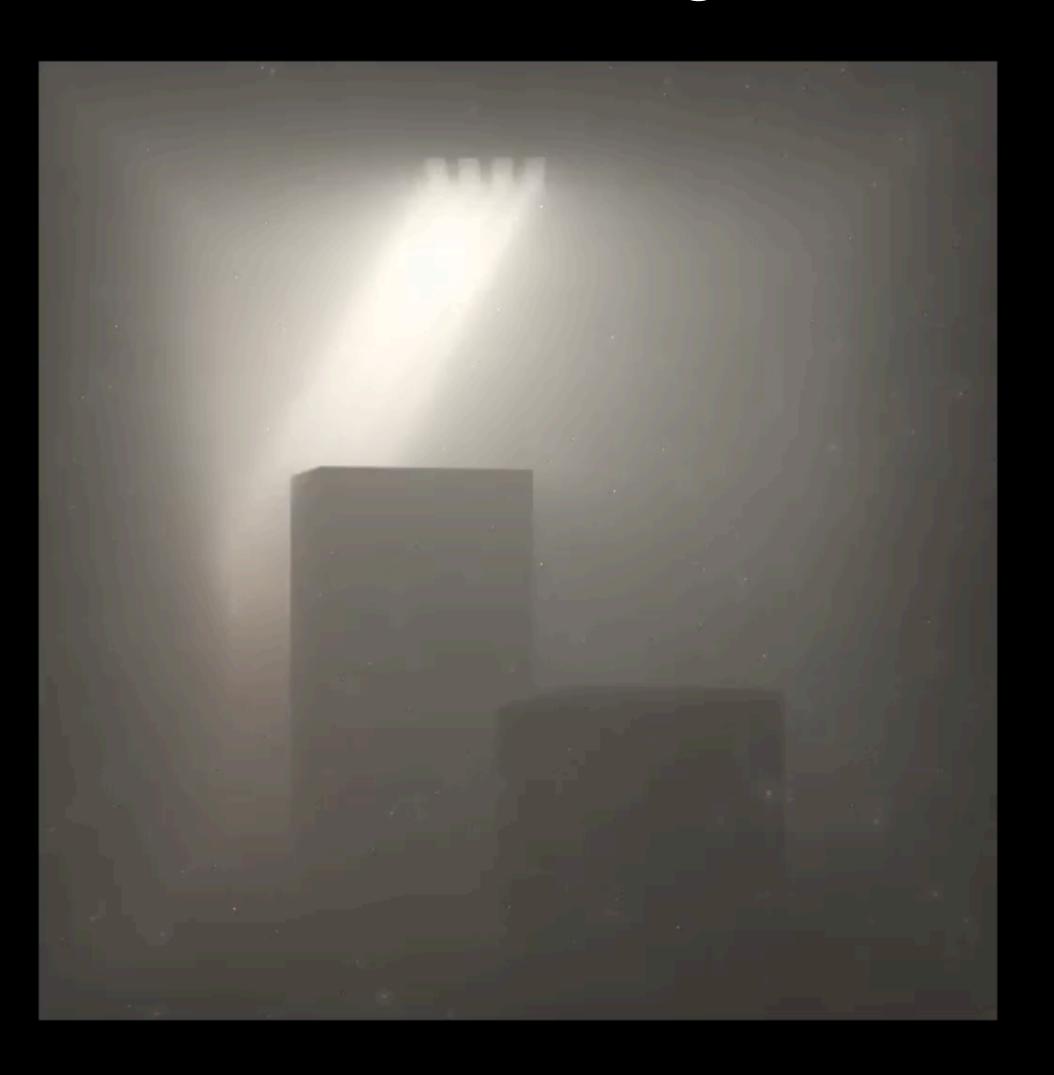
1 minute/frame



1 minute/frame

Virtual Ray Lights

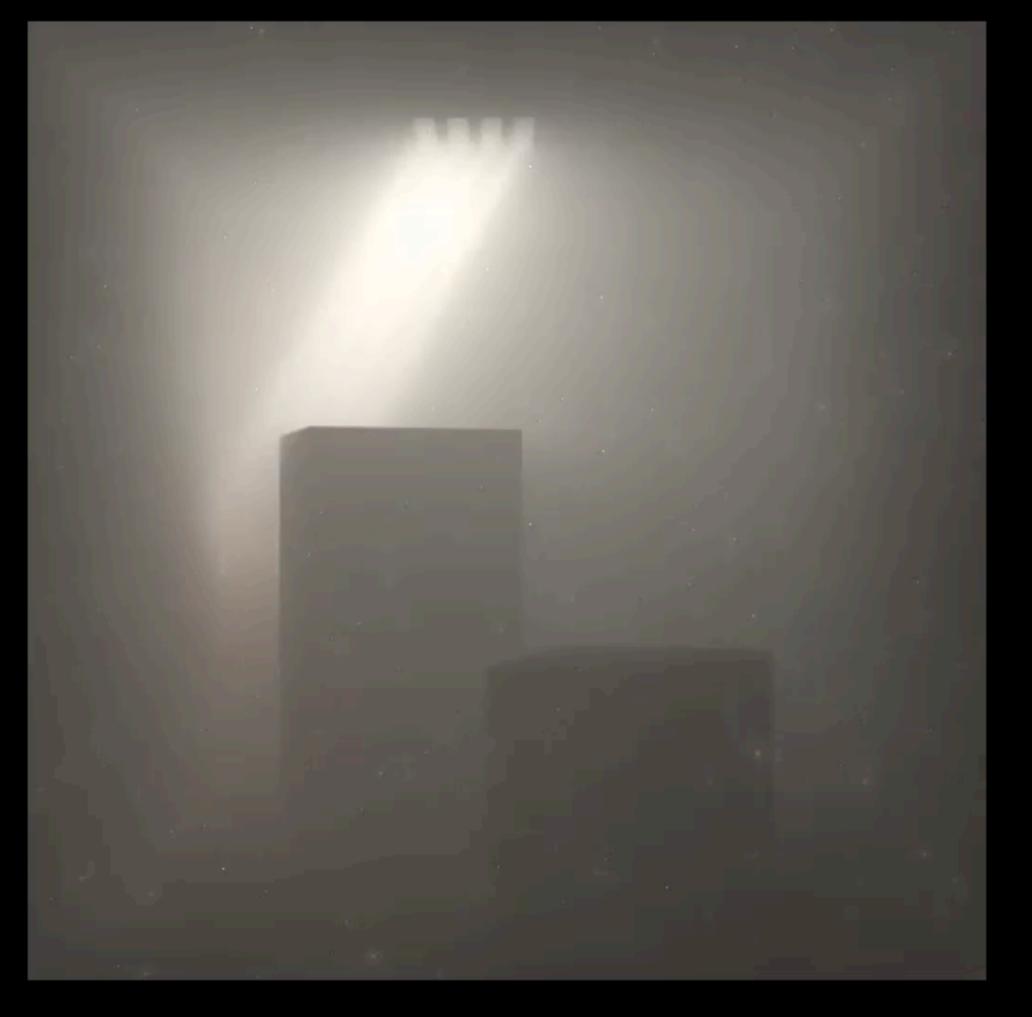




1 minute/frame

Virtual Ray Lights





1 minute/frame

3 minutes/frame

Spreading the Energy

Turn segments of paths into light sources

Advantages:

- energy is spread along lines, singularity is reduced (not removed)
- unbiased, temporally stable

Disadvantages:

- requires 2D integration (along both rays)

LightCuts

Only for Virtual Point Lights

Walter et al. [2005]







Many-Lights Problem

Brute Force: Consider all lights

Too Slow





Many-Lights Problem

Brute Force: Consider all lights

Too Slow

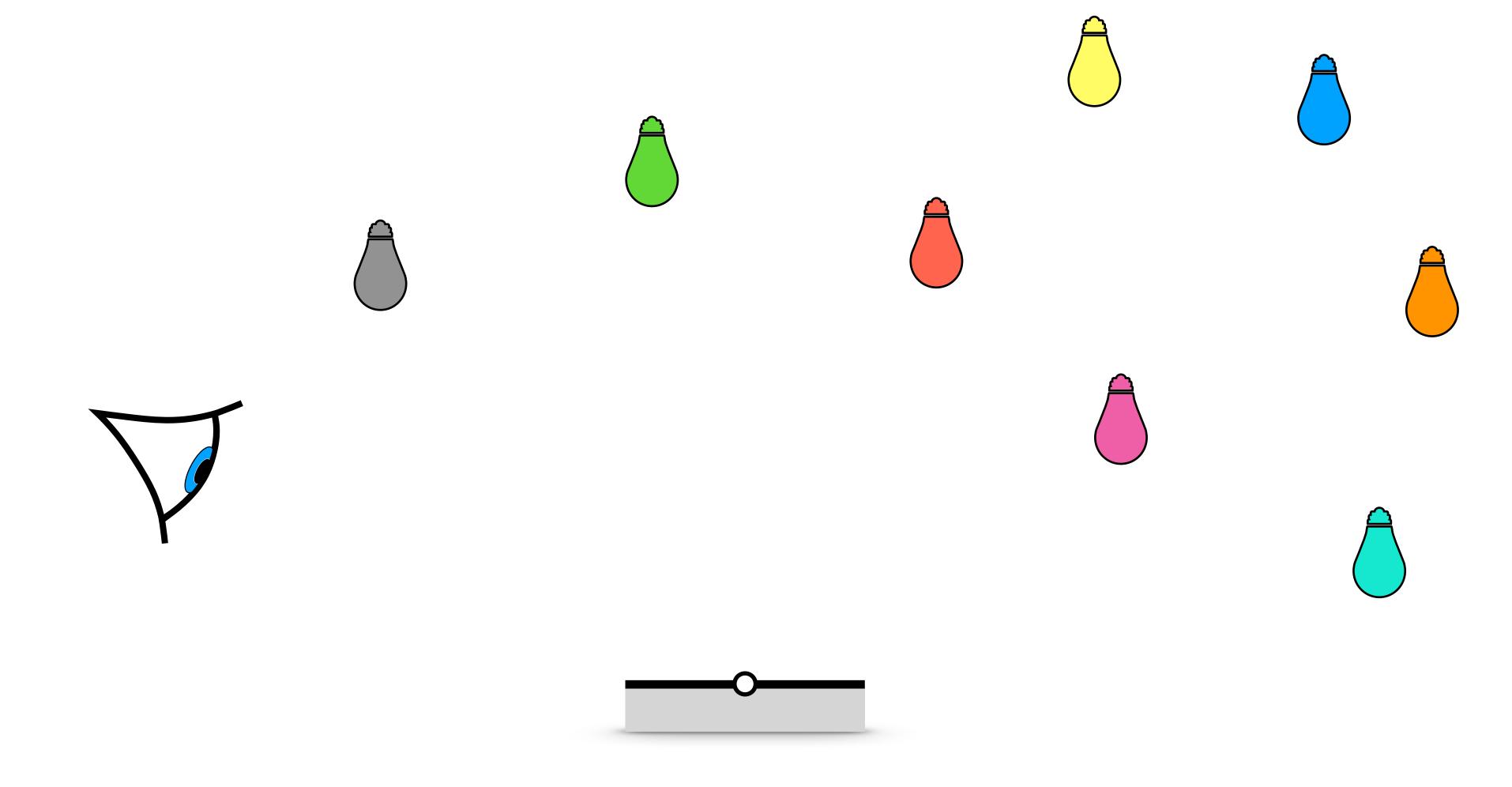
Importance Sampling: Shirley et al. [1996]

- By considering only a fraction of lights.
- The method suggests using light intensities as importance weights
- Does not give different set of weights, on different parts of the scenes.

Too Noisy

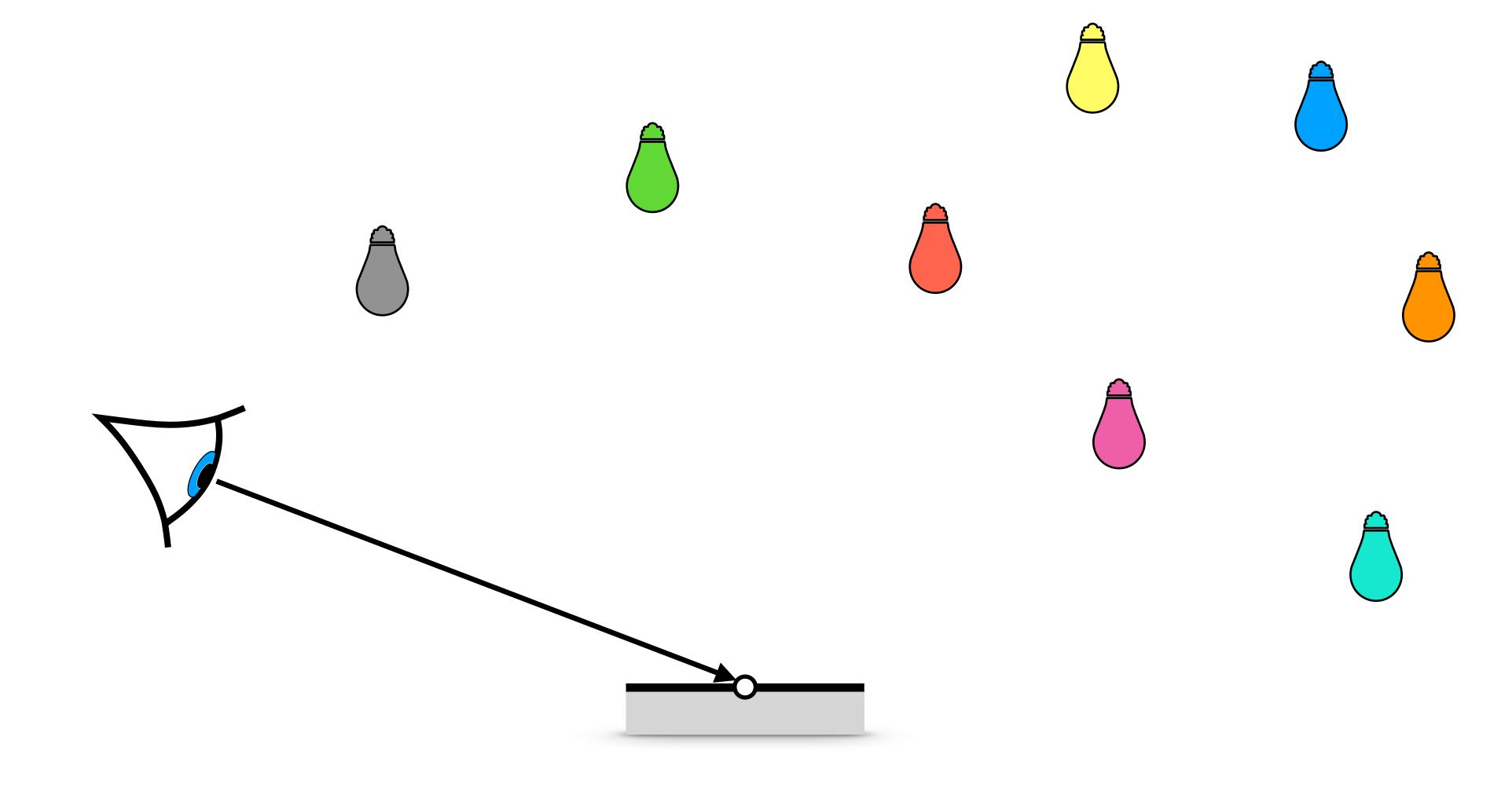






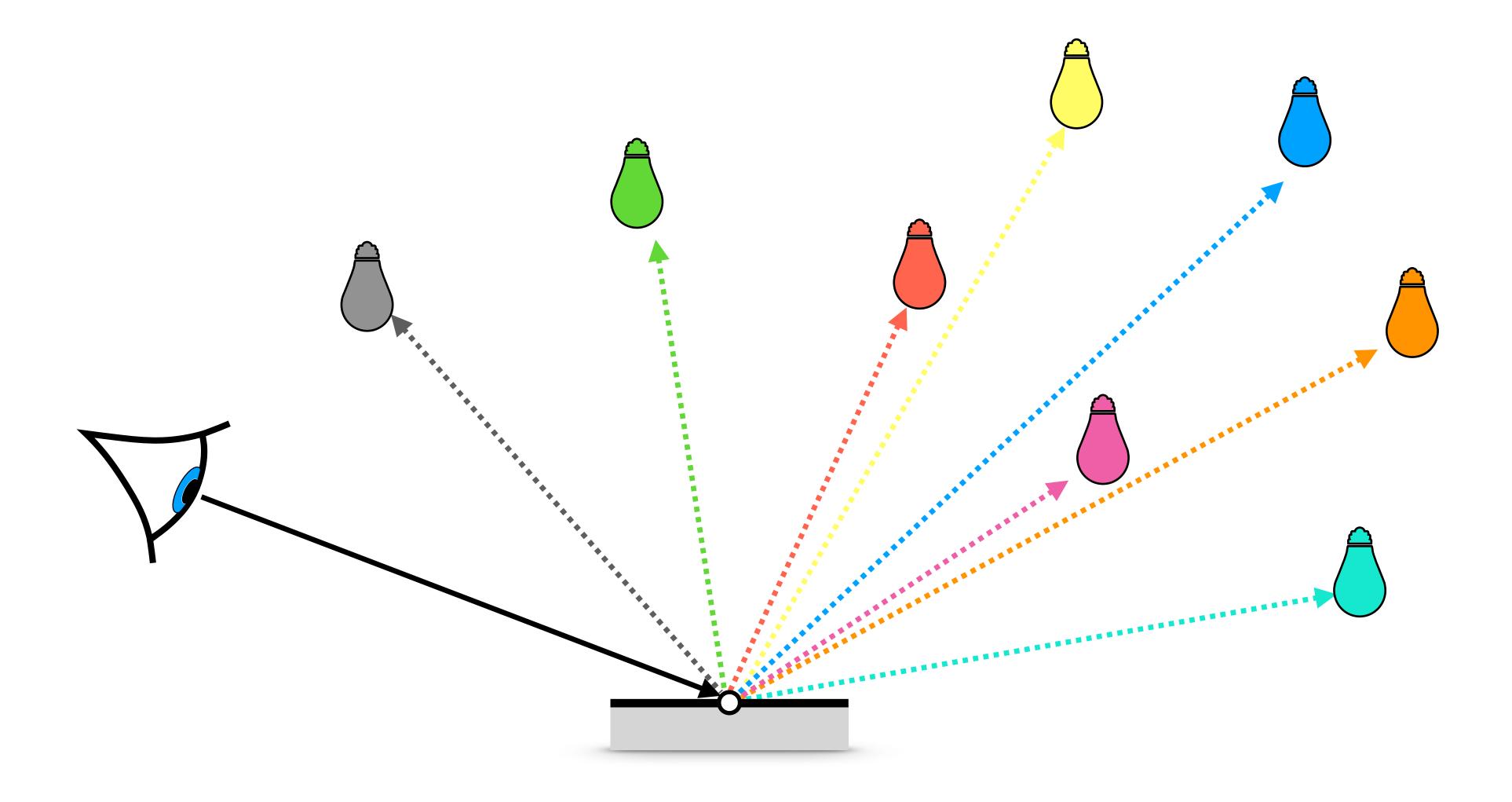




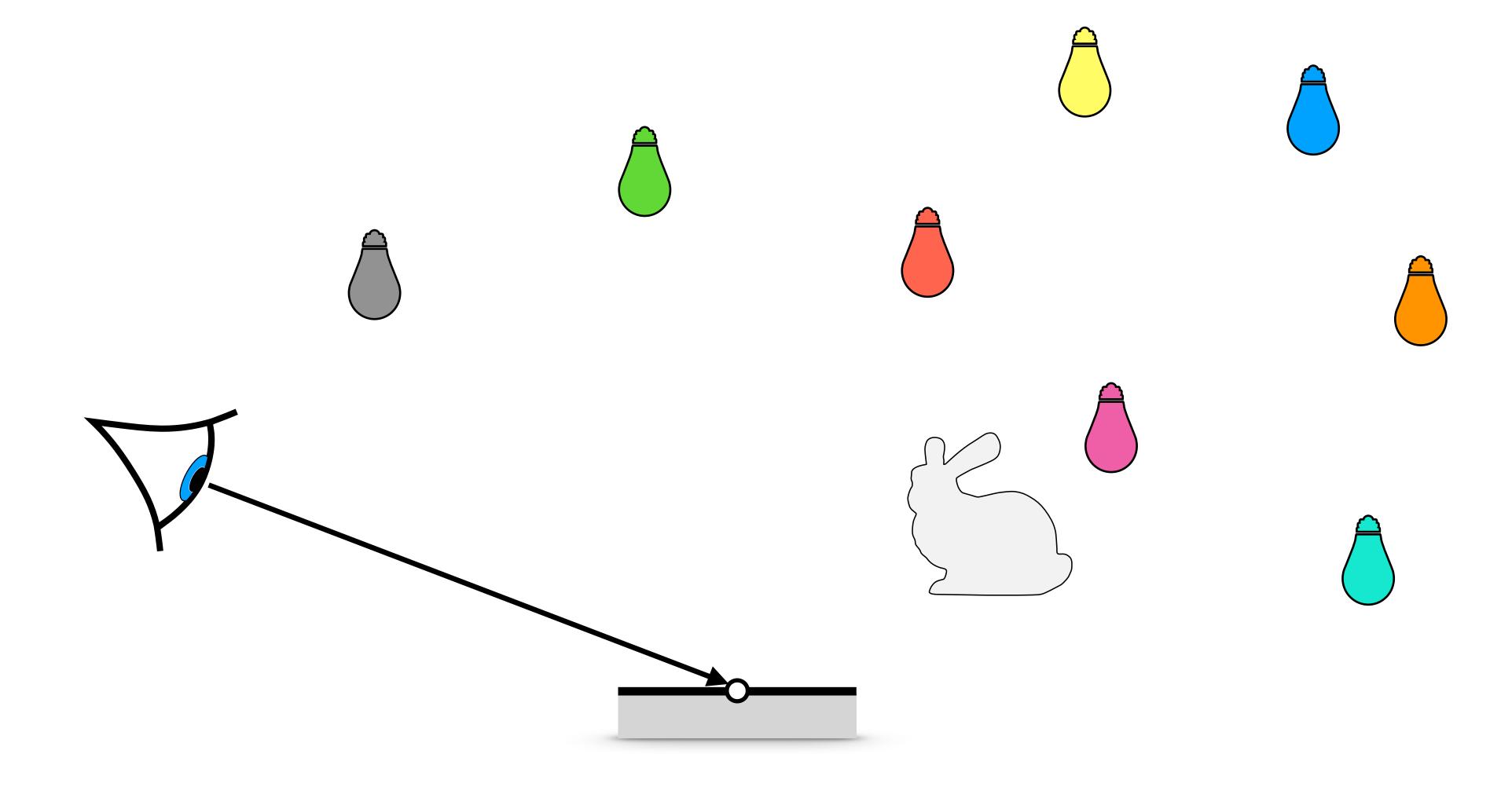






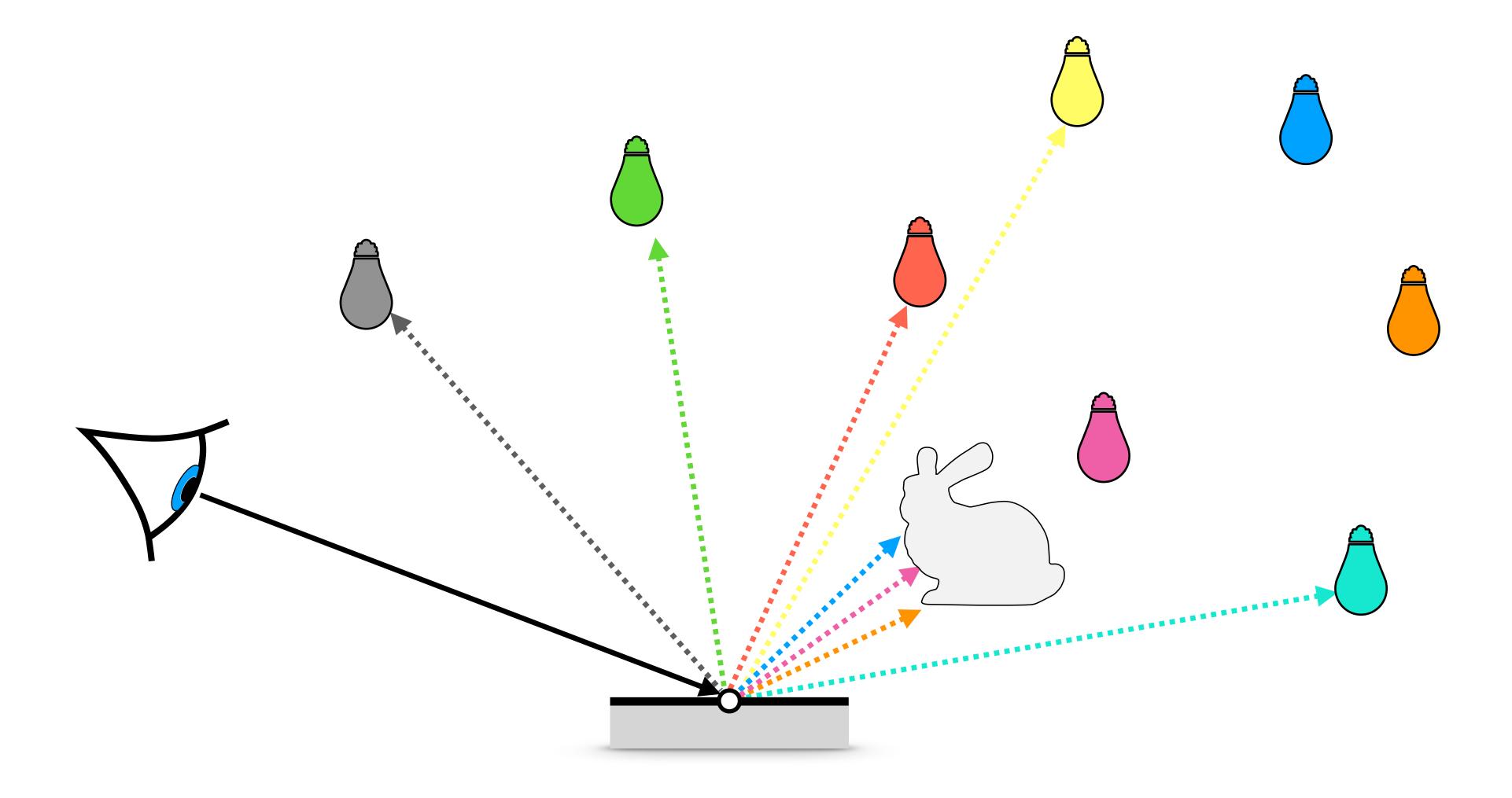










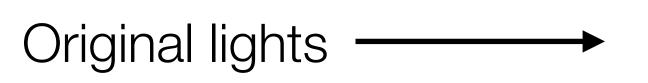






Light Clusters

Light Tree





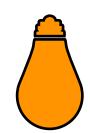
















Light Clusters Light Tree Light Clusters Original lights





Light Clusters Light Tree Light Clusters Original lights





Light Clusters Light Tree Light Clusters Original lights

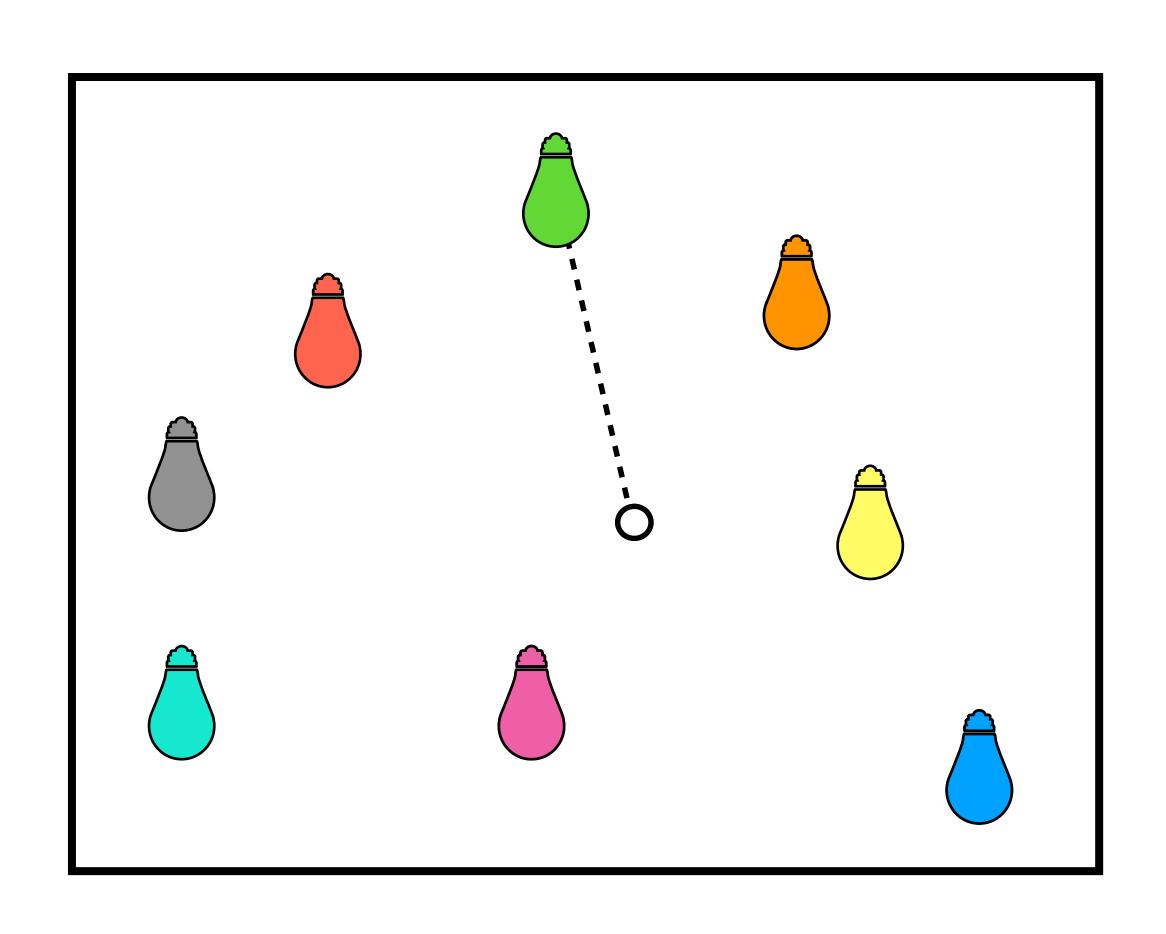


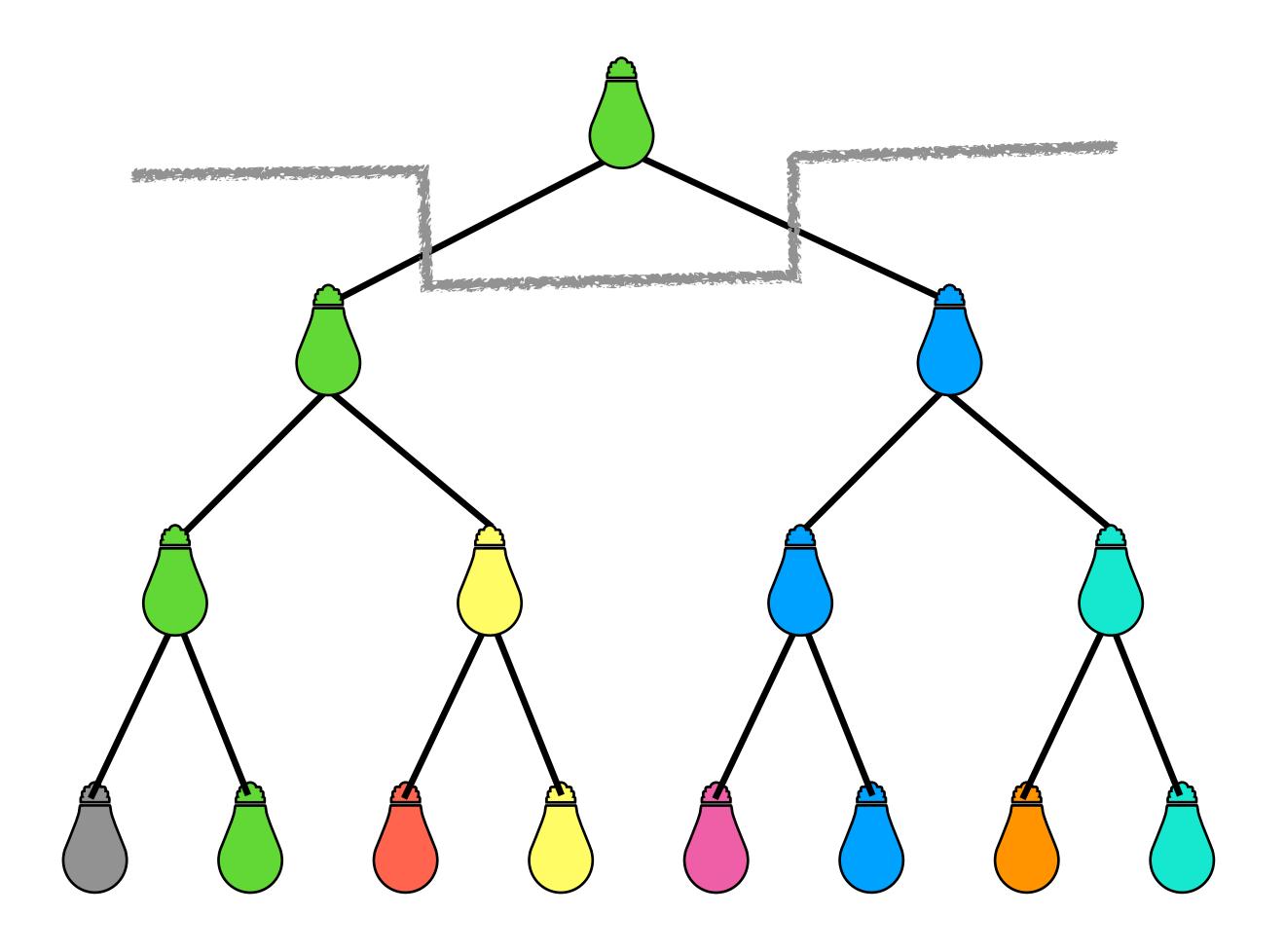


Light Clusters Light Tree Light Clusters Original lights



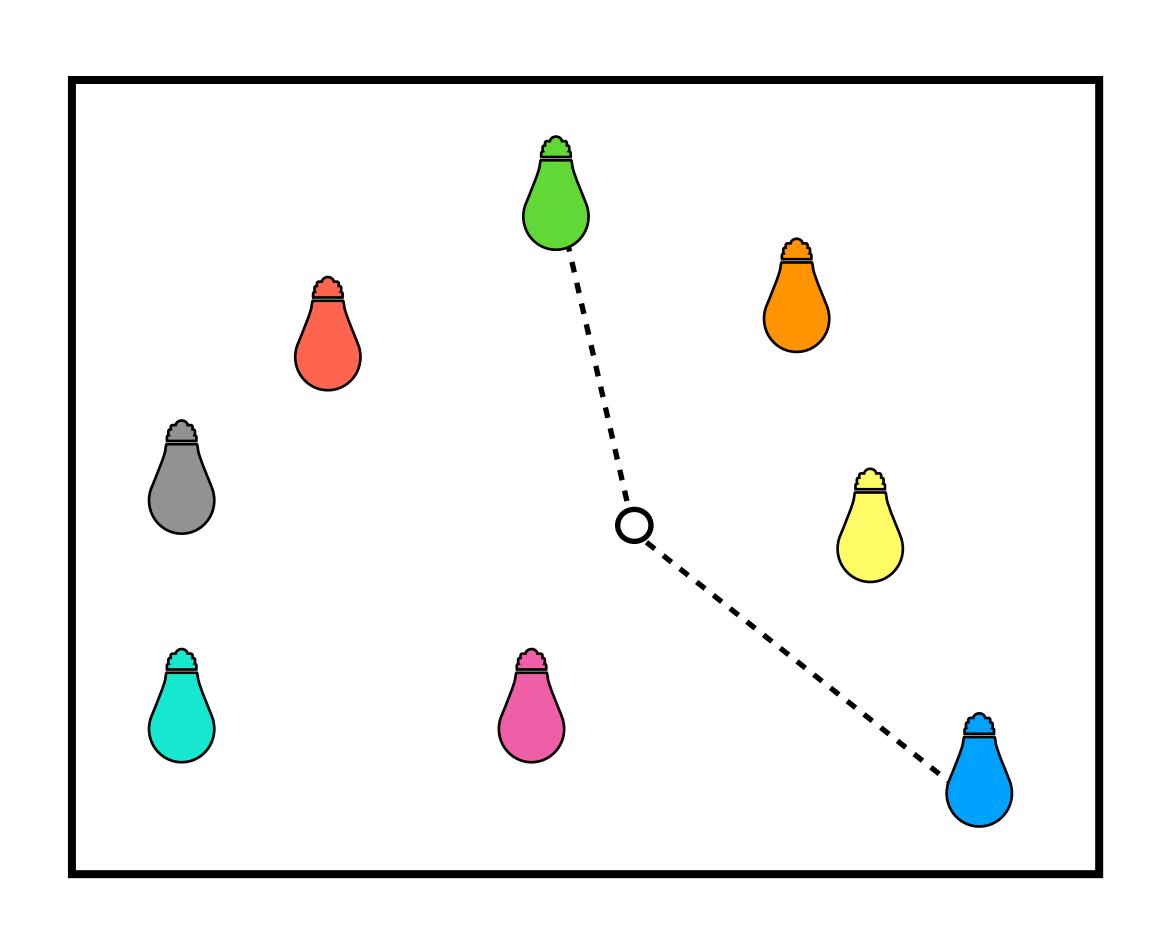


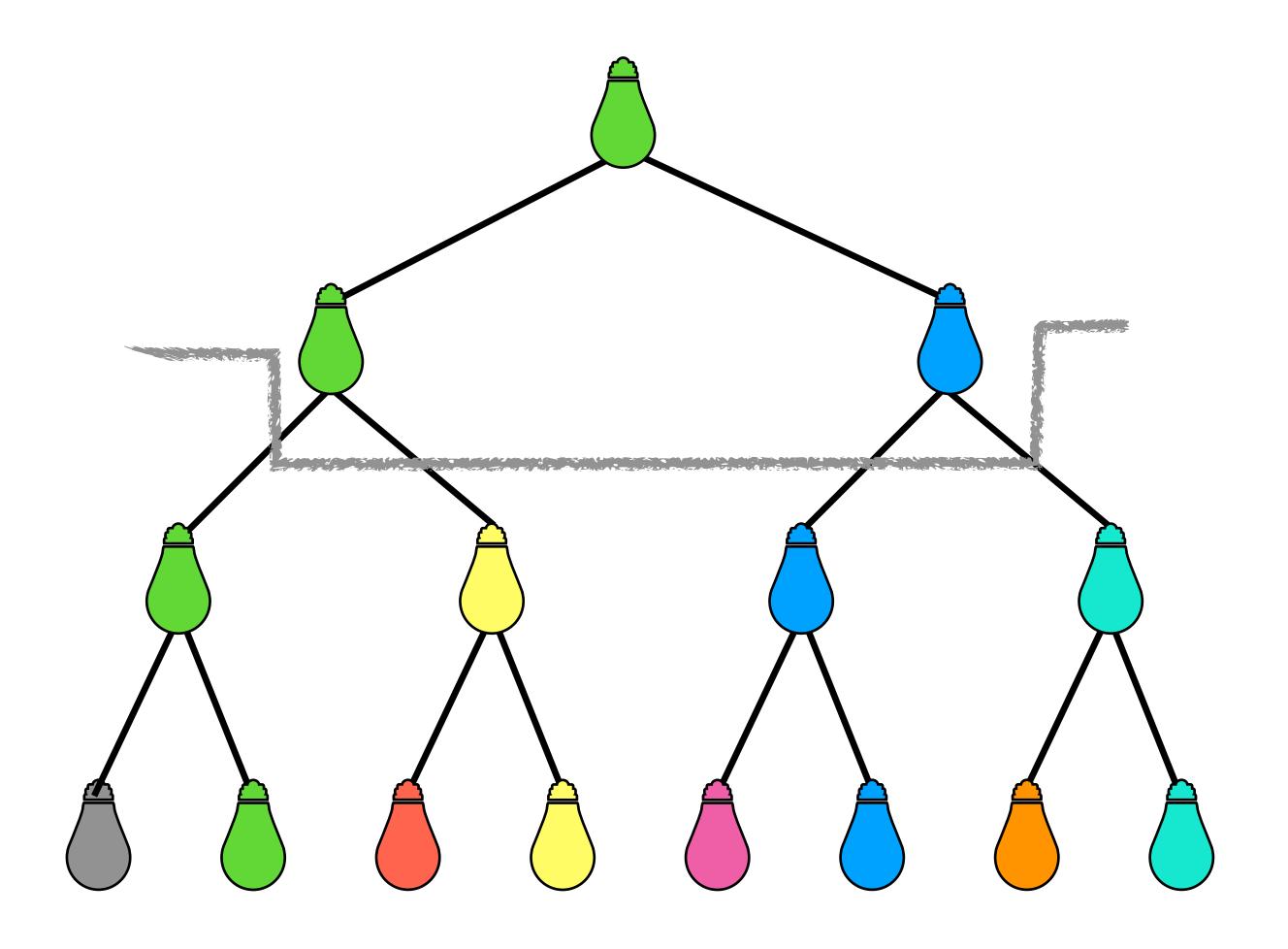






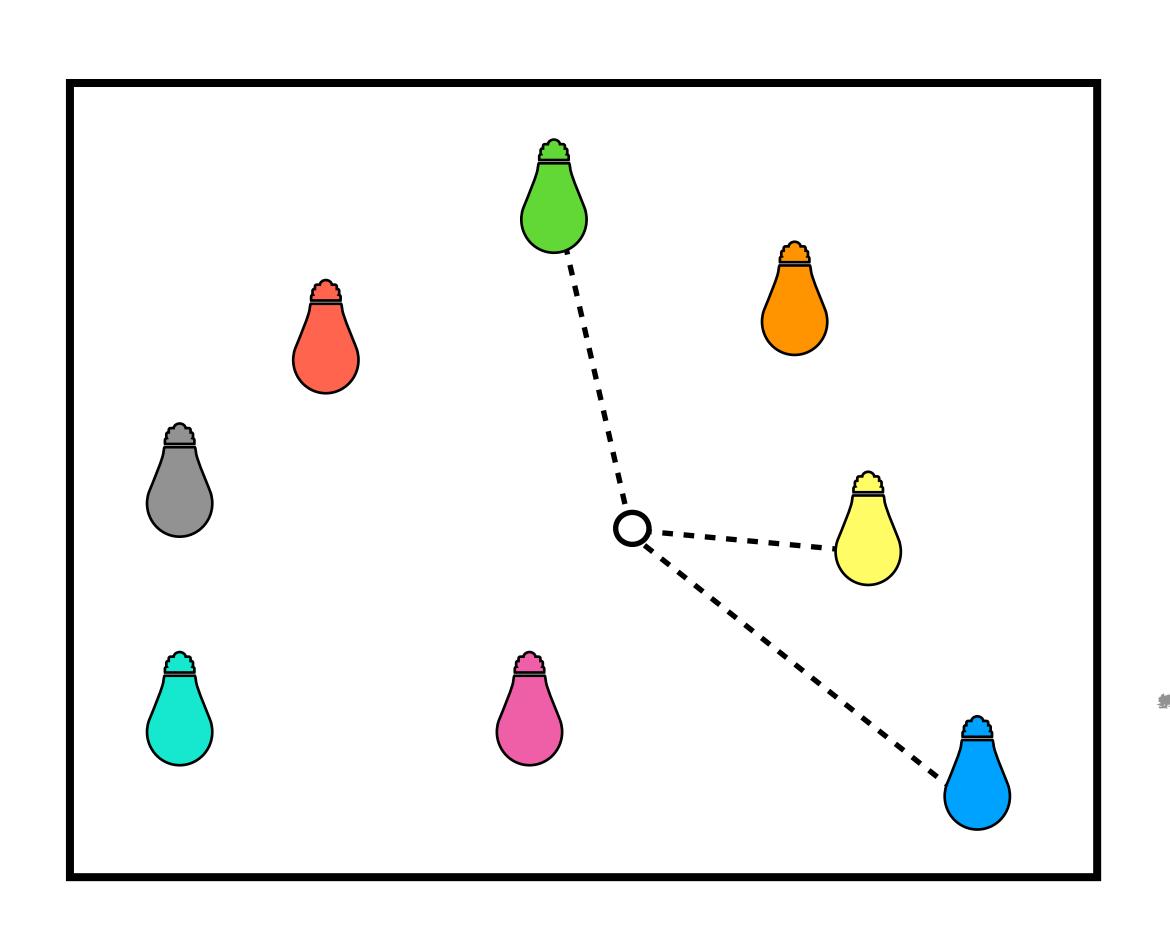


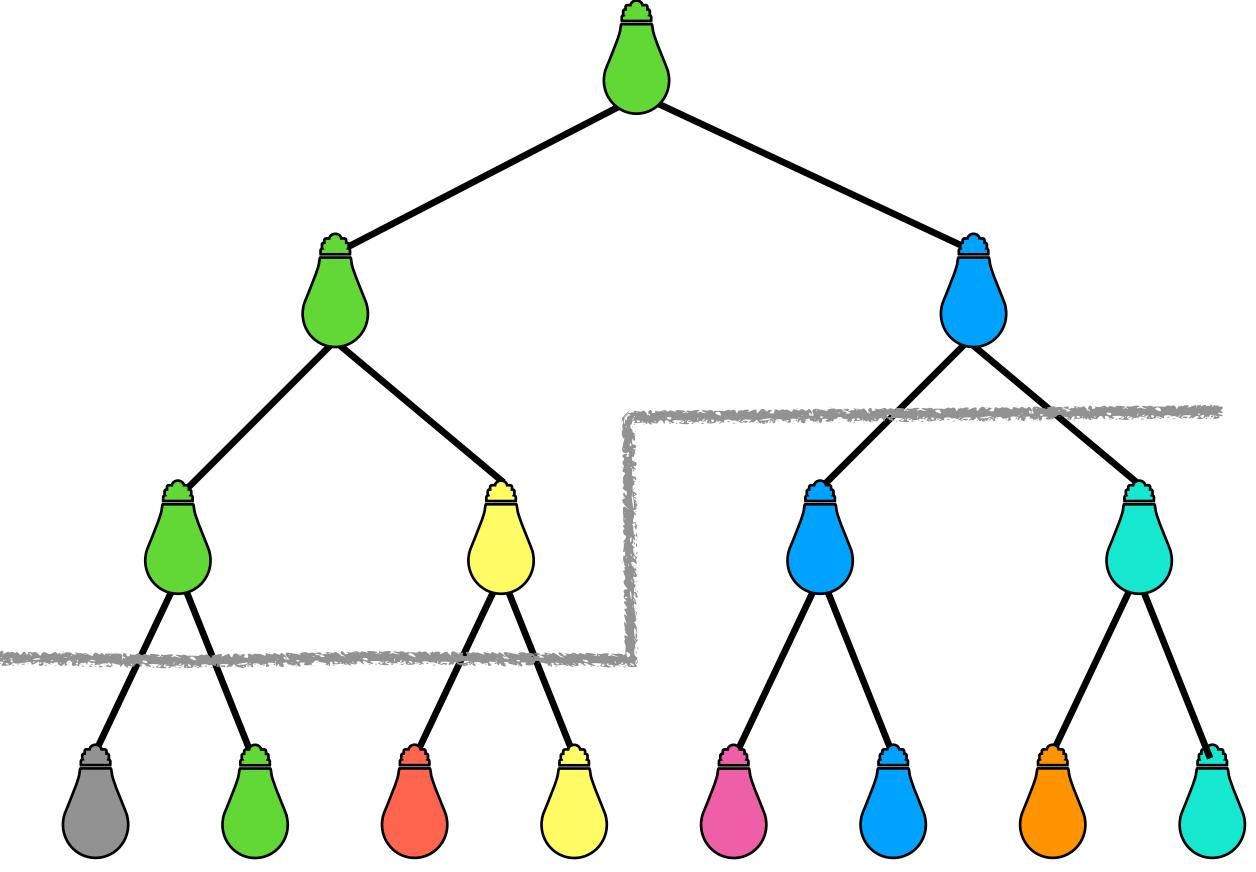






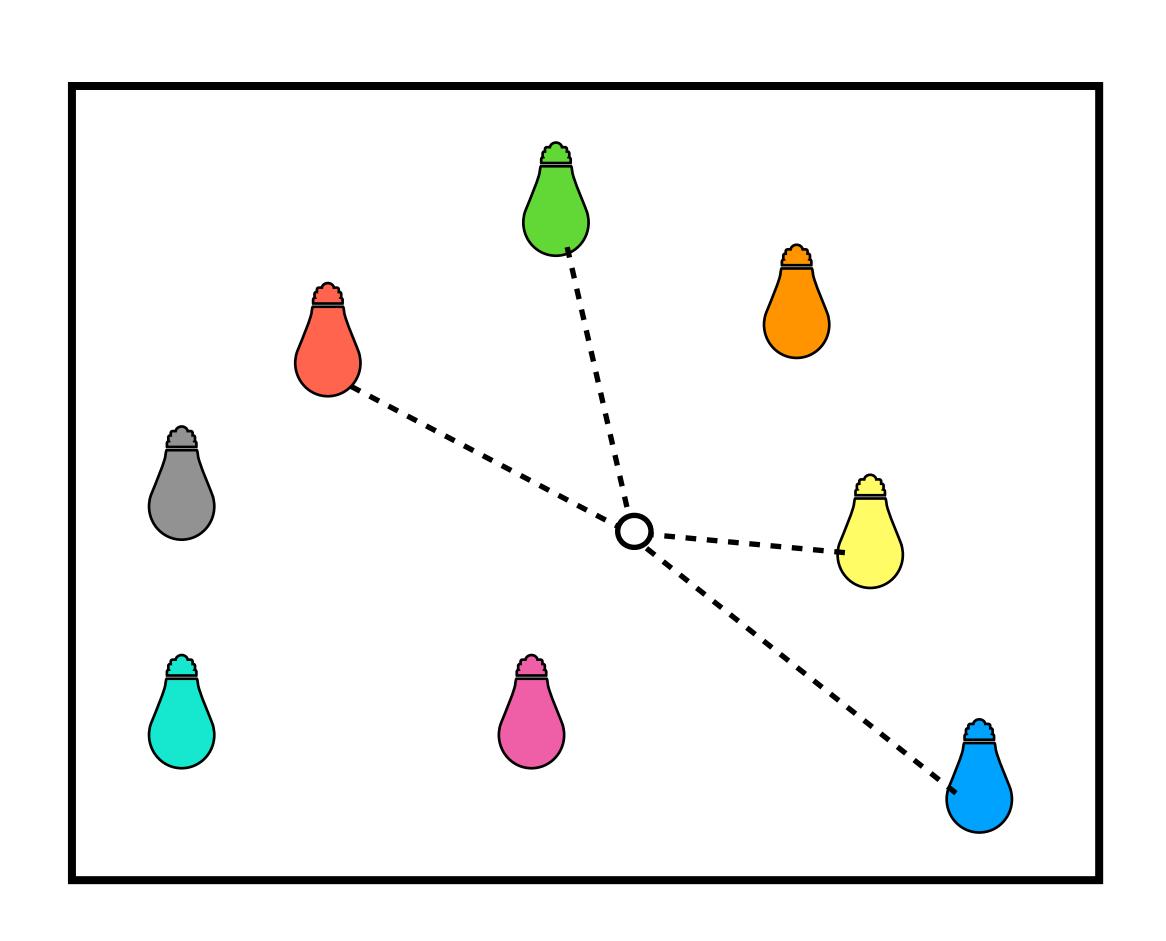


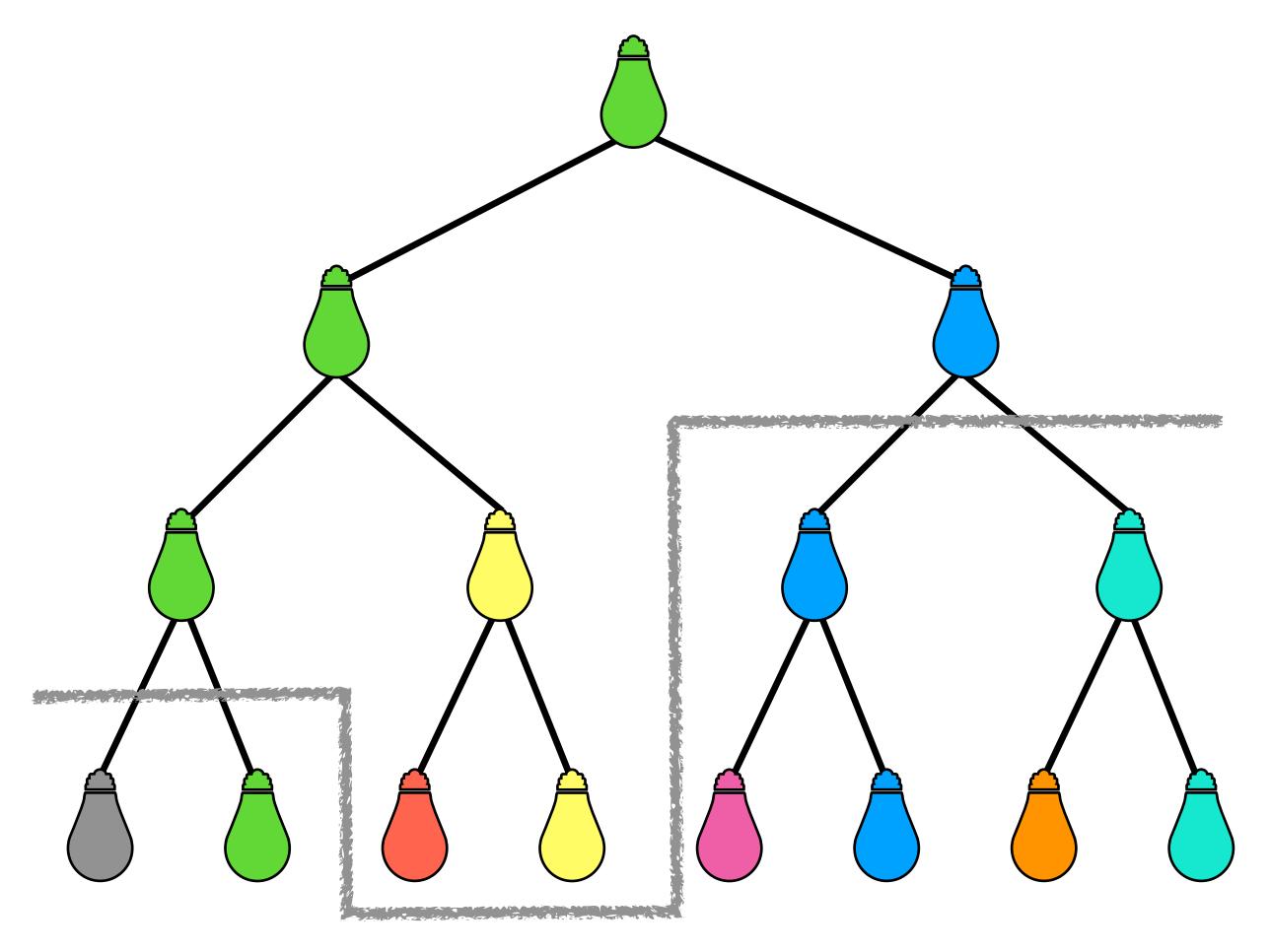






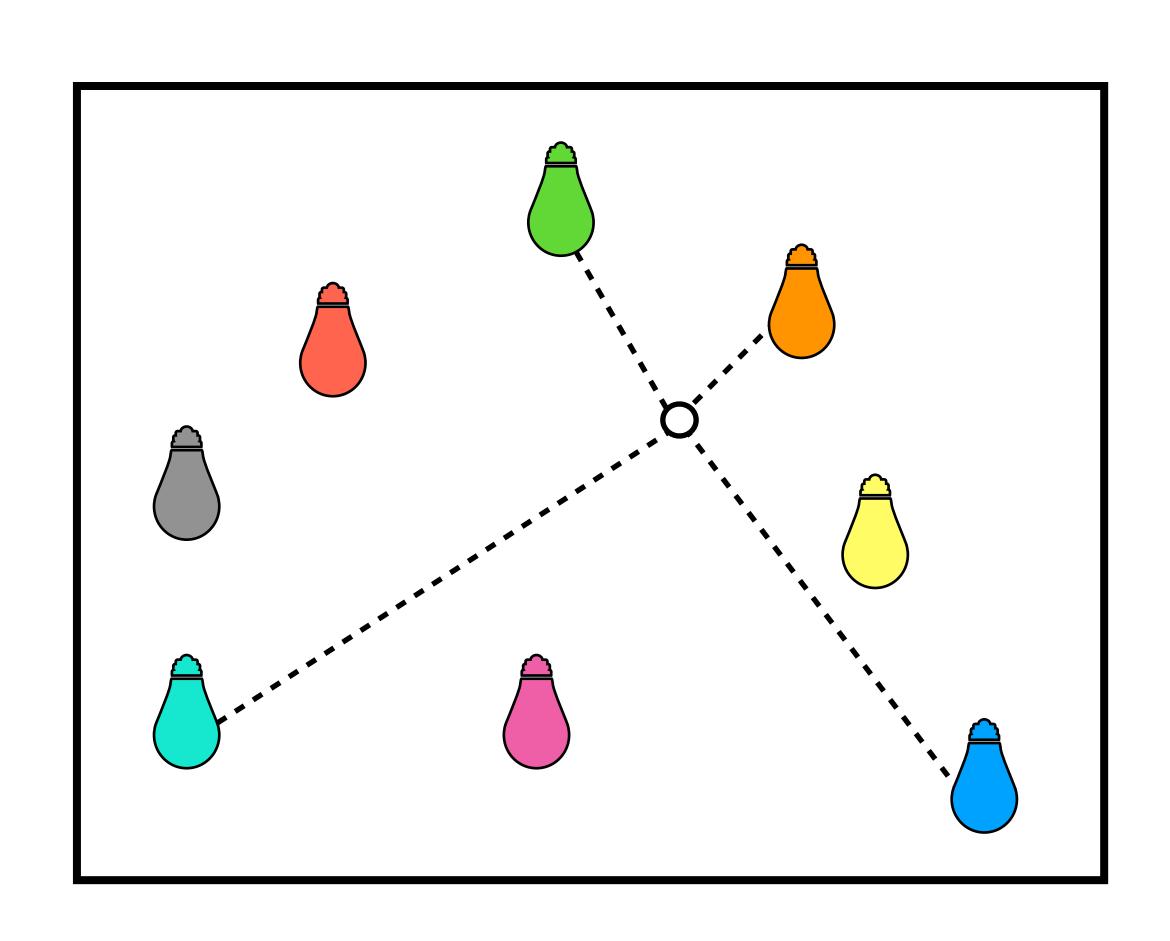


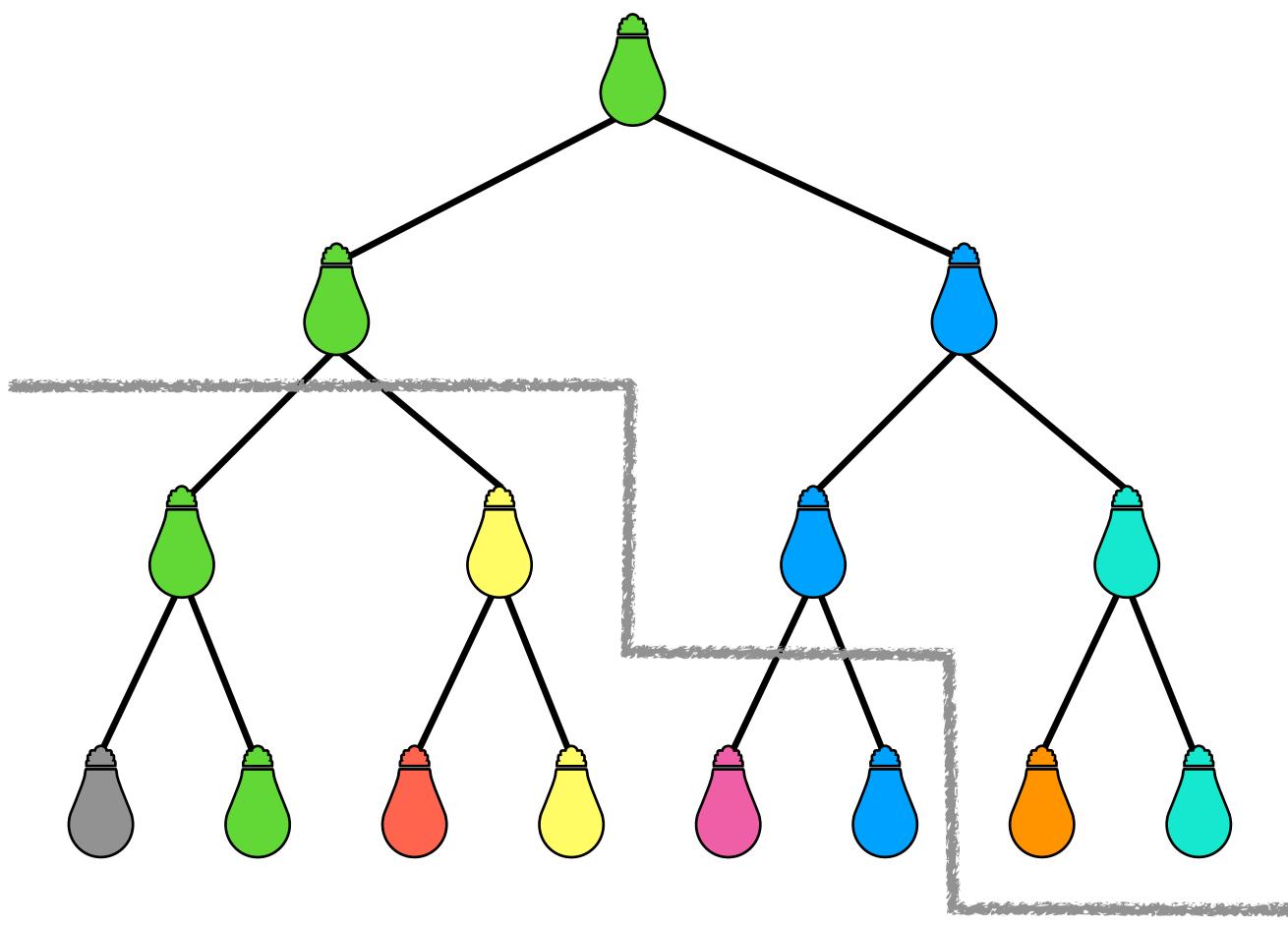






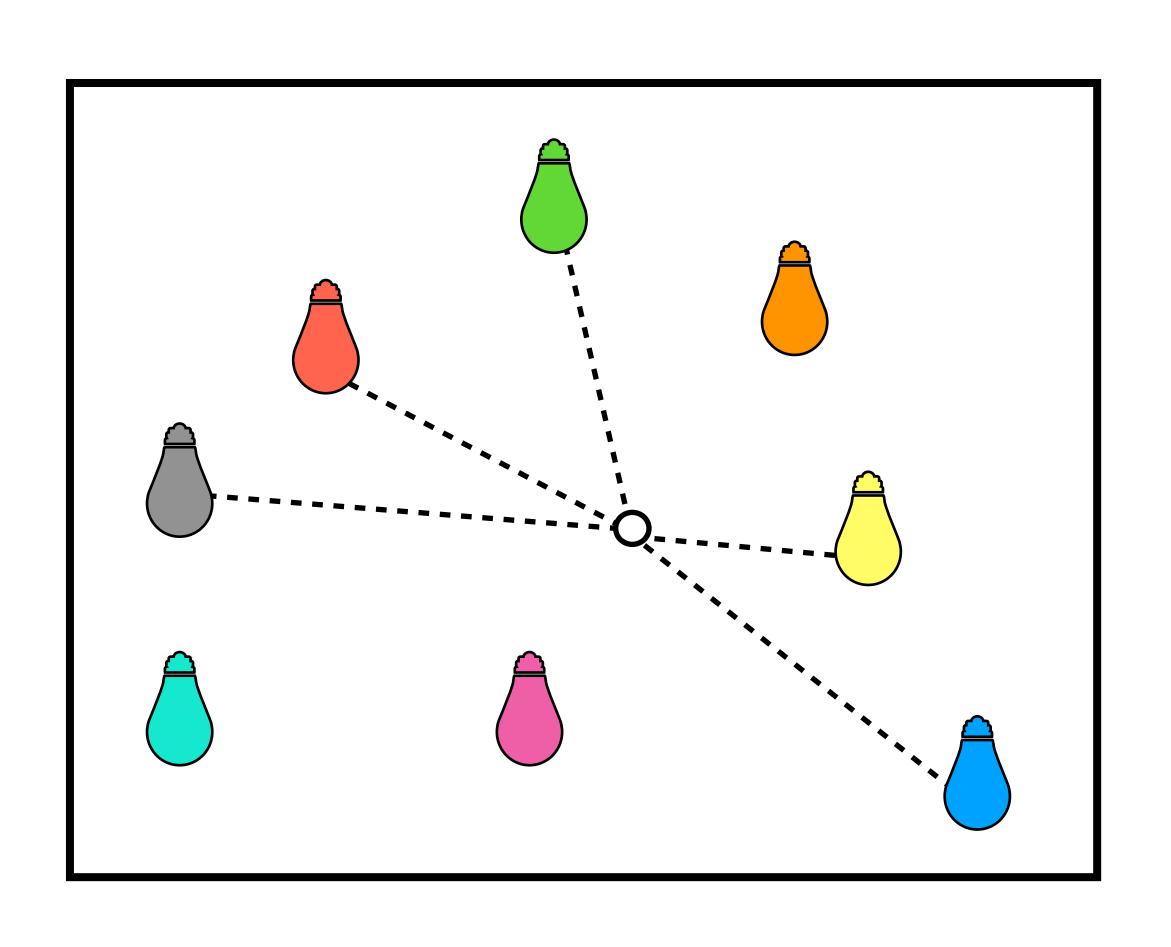


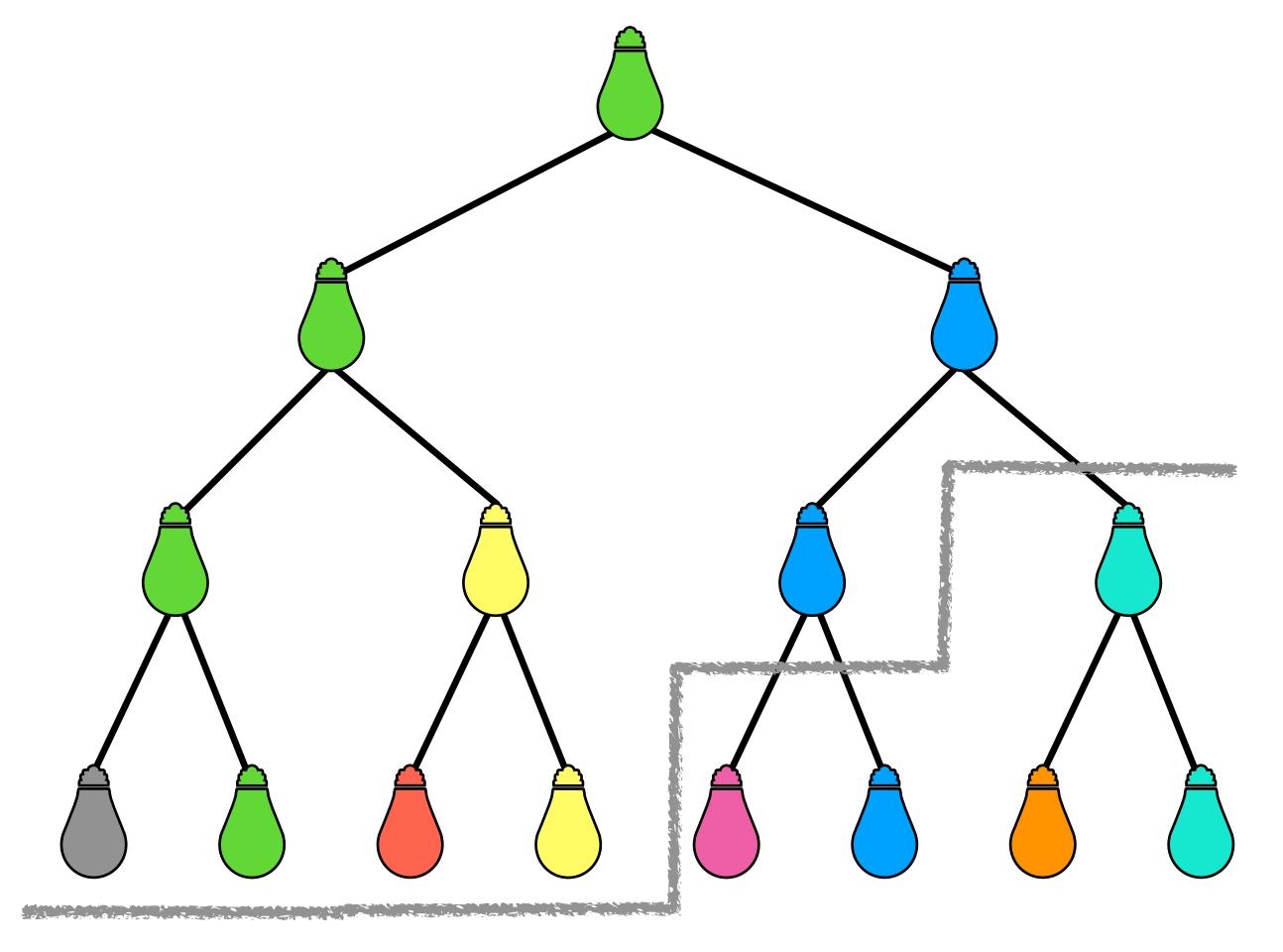
















LightCuts: Algorithm Overview

Pre-process:

- Convert illumination to virtual point light sources
- Build light tree

For each ray:

- Choose a cut to approximate the illumination



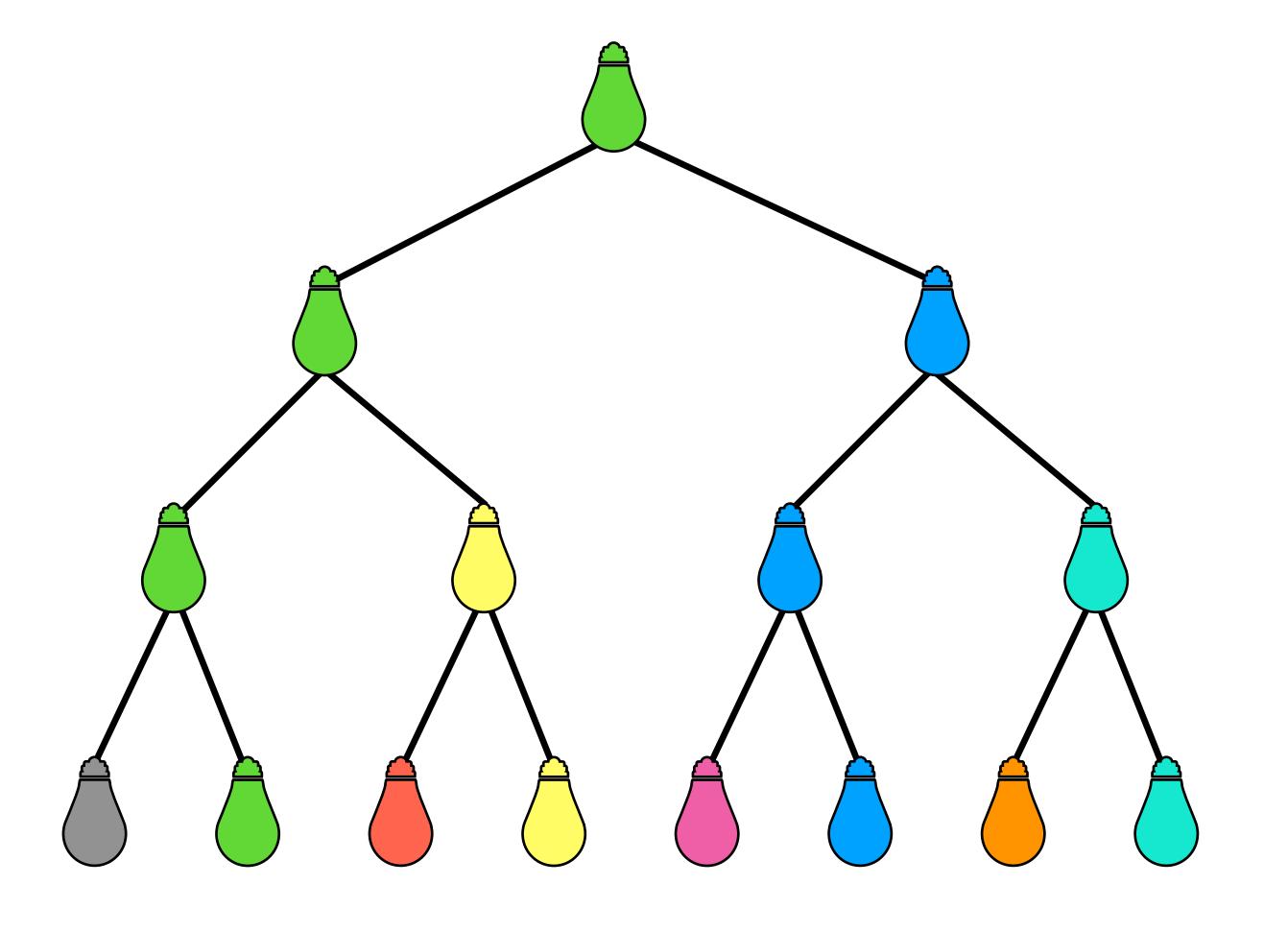


LightCuts: Issues

Light near the top of the tree more likely to be sampled

Sampling correlations

Different light trees favor different lights









Stochastic LightCuts

Only for Virtual Point Lights

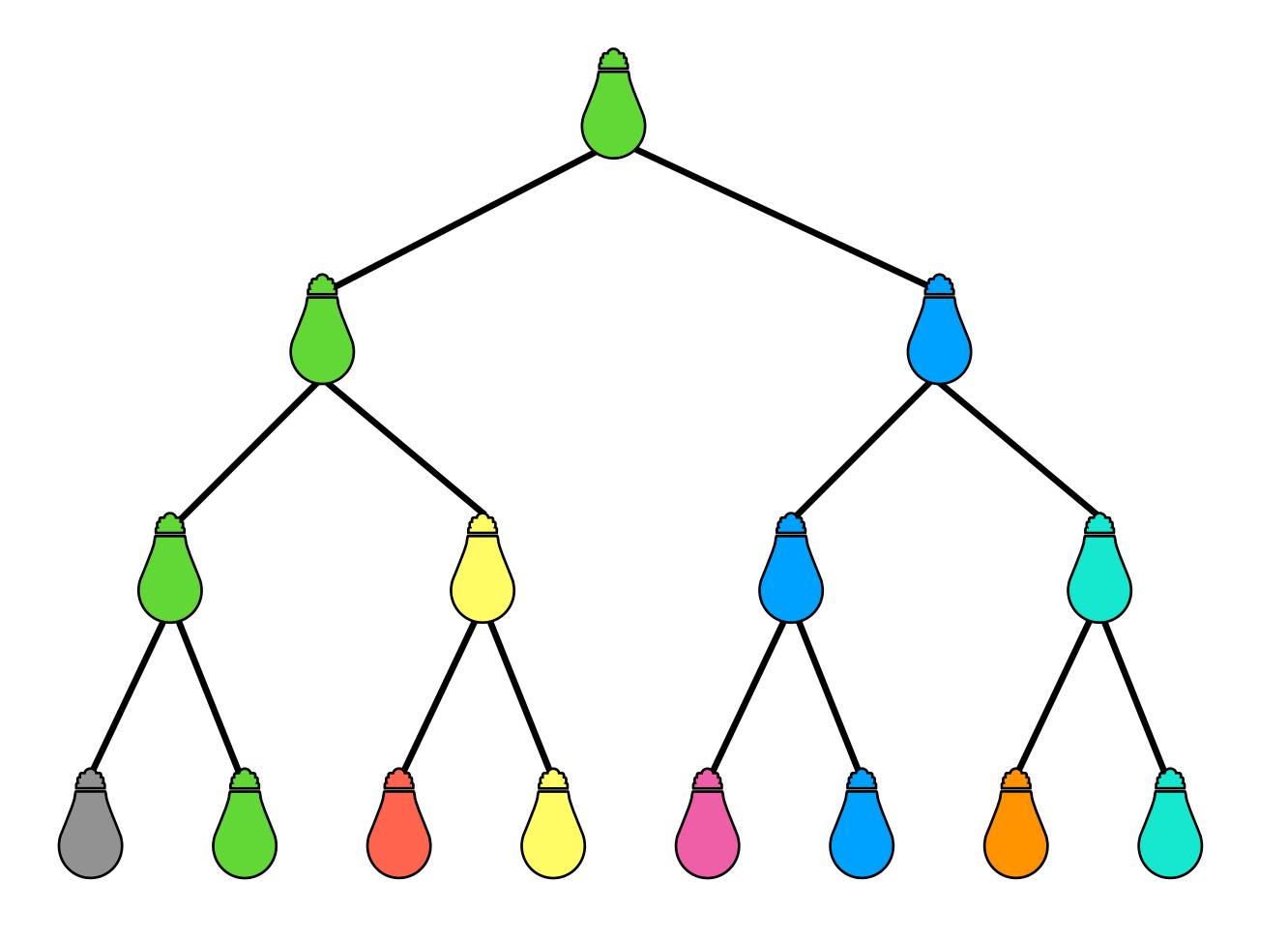
Cem Yuksel [2019]





Stochastic LightCut

No representative lights



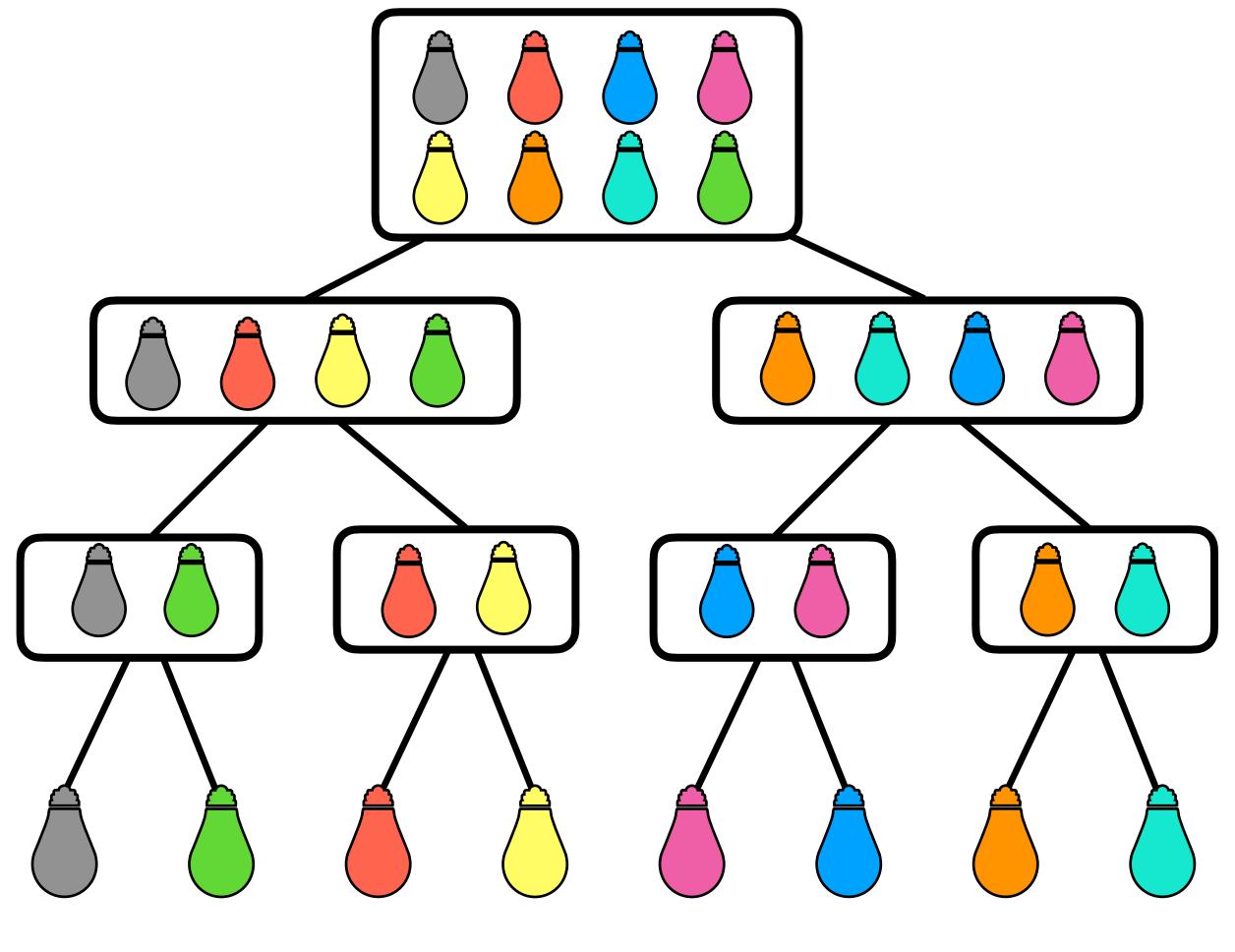




Stochastic LightCut

No representative lights

Randomly select lights







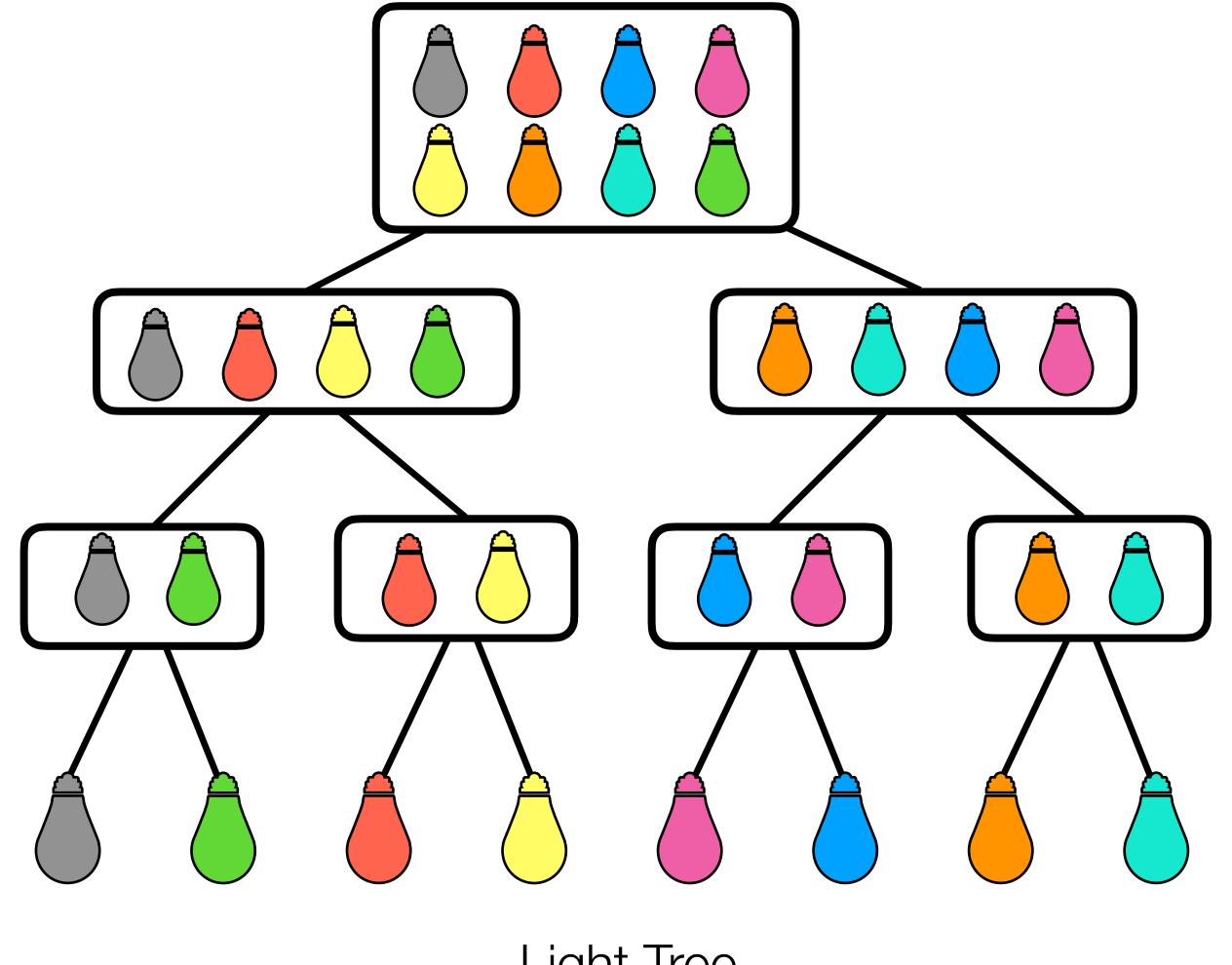


Stochastic LightCut

No sampling correlation (i.e. no flickering)

Any type of light (no representative light)

Light sample count can be limited



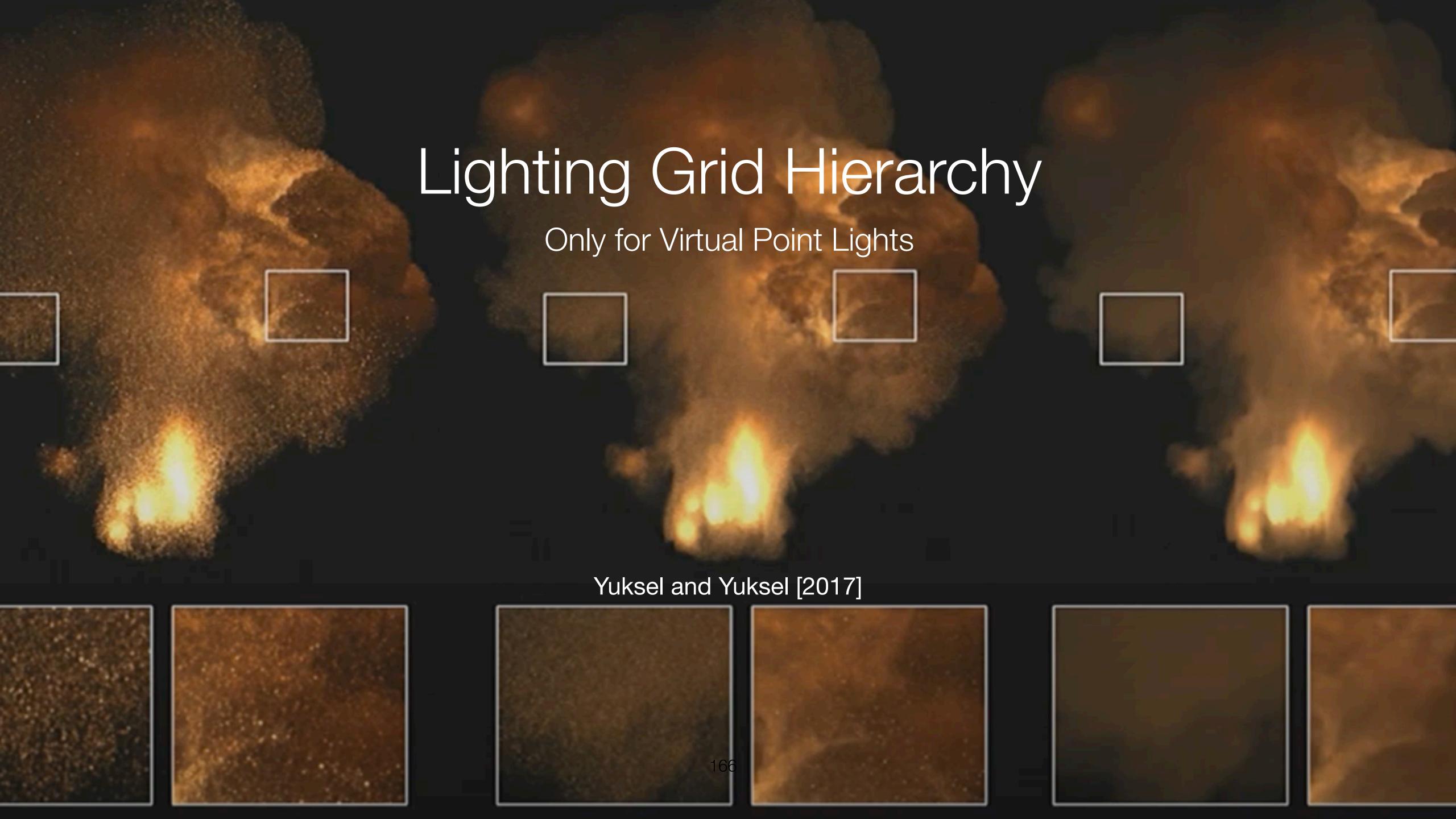
Light Tree











Explosion Rendering

Challenges:

- Animated volumetric light source
- High variation in fire
- Shadows from a heavy smoke layer





Industry Practices

Smoke illumination by the fire of the explosion:

- No illumination
- 2D filtering in compositing
- 3D filtering / diffusion techniques

Environment illumination by the fire of the explosion:

- No illumination
- Few hand-placed lights to approximate illumination
- Some inferior representation of the fire volume as light source





- Convert volume to many point lights
- Build Lighting Grid Hierarchy
- Pre-compute volumetric shadows
- Add multiple scattering, if desired
- Finally, render using hierarchy





A multi-resolution representation of lighting

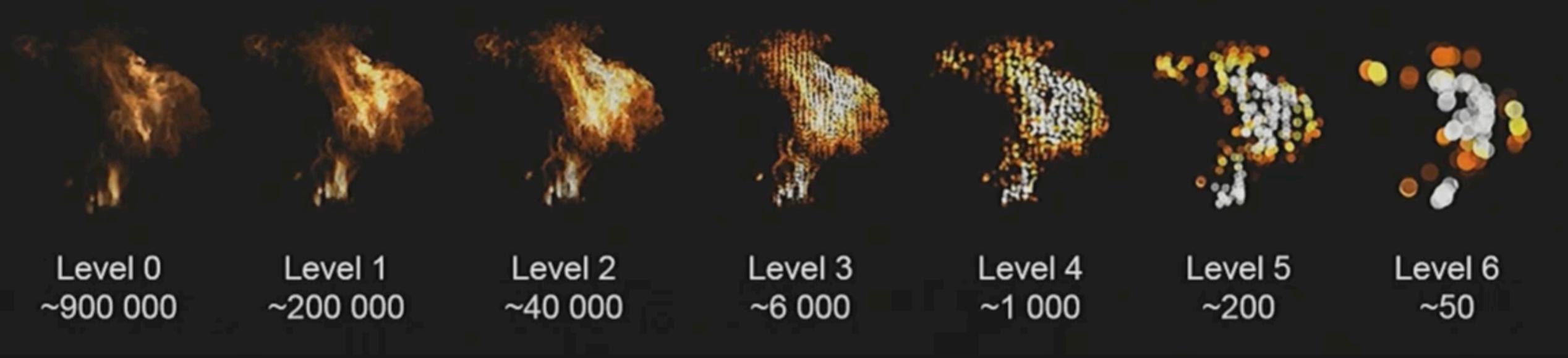
Temporary coherency:

- No clustering
- No binary decisions
- No sharp thresholds





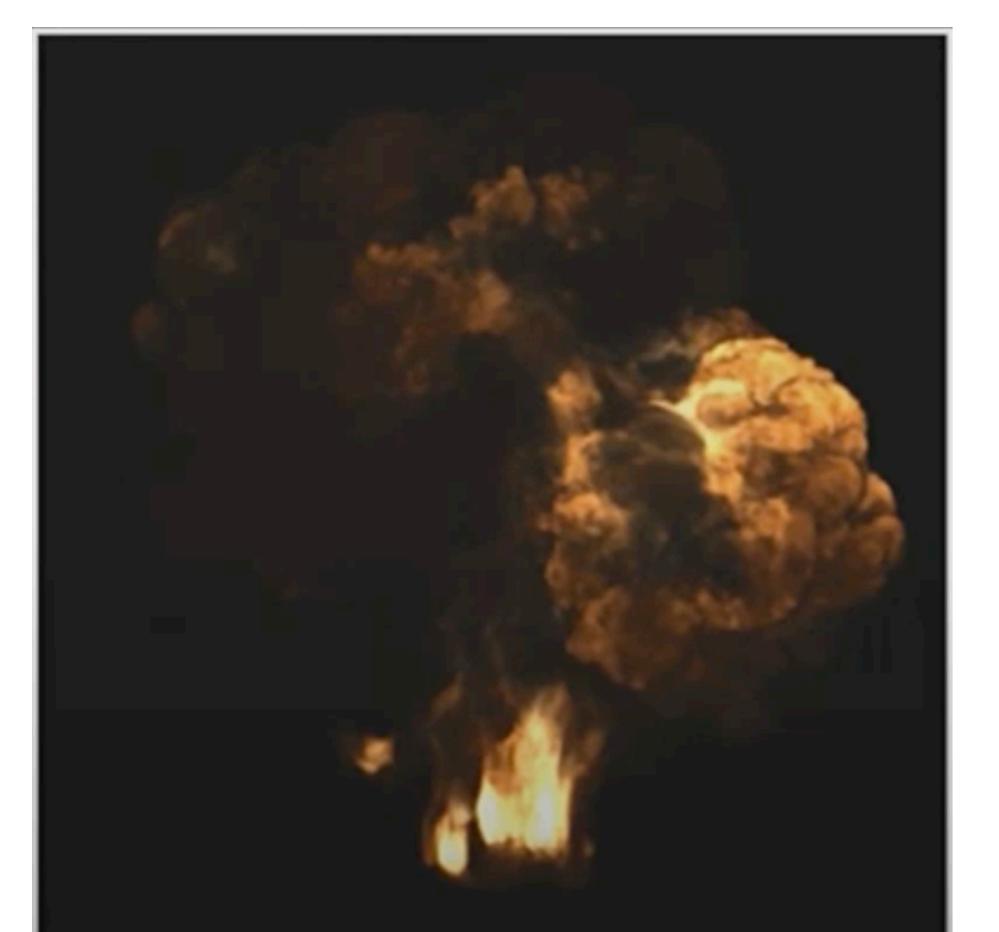
Multi-resolution representation:







Volume Data



Point Lights

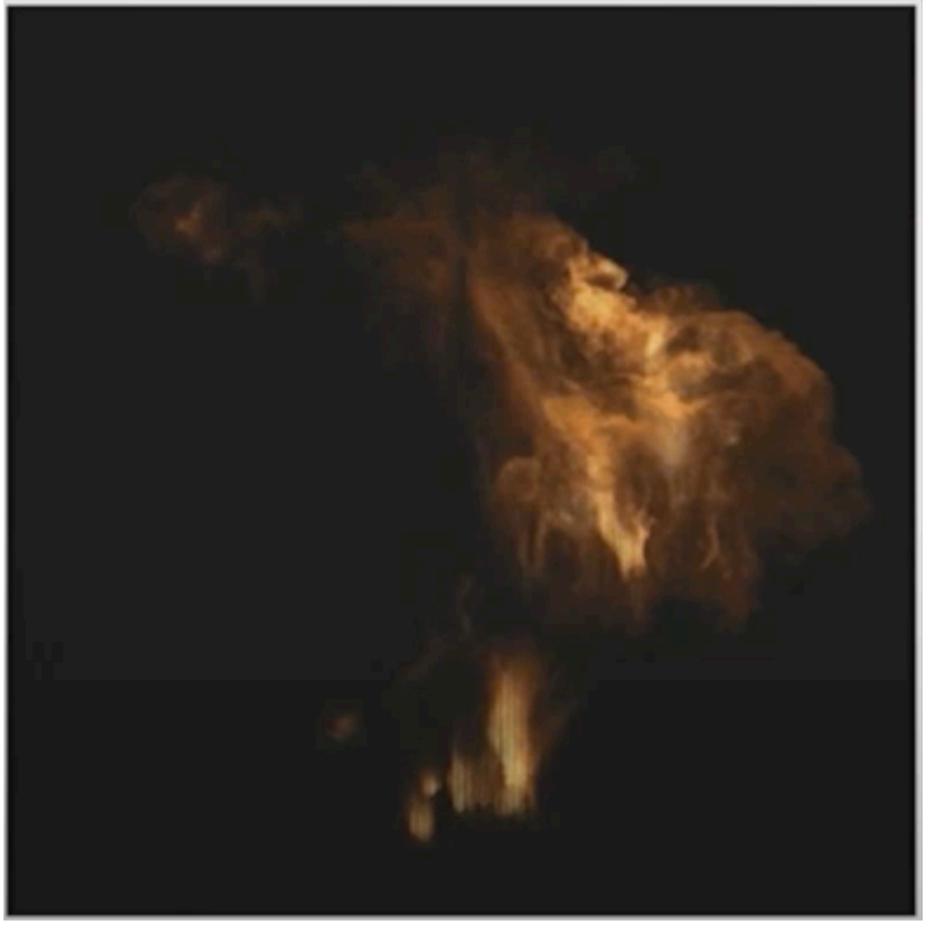
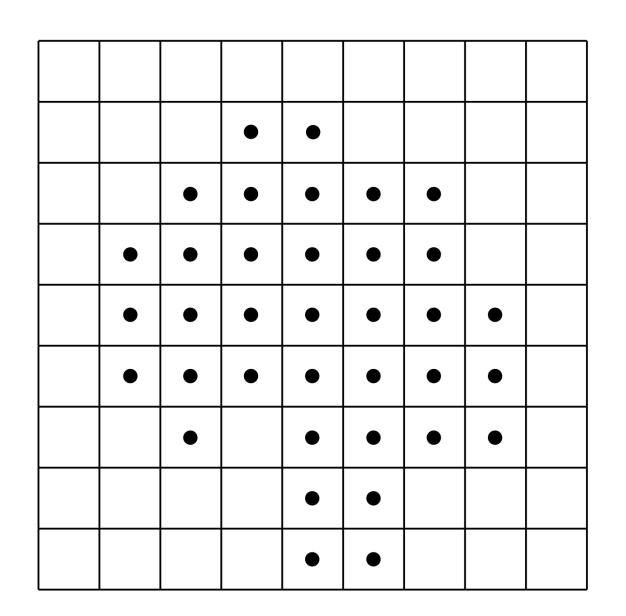


Image courtesy Yuksel and Yuksel [2017]

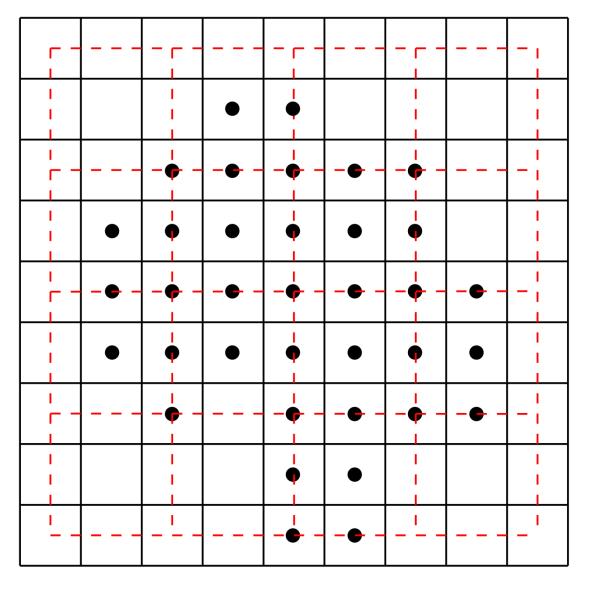




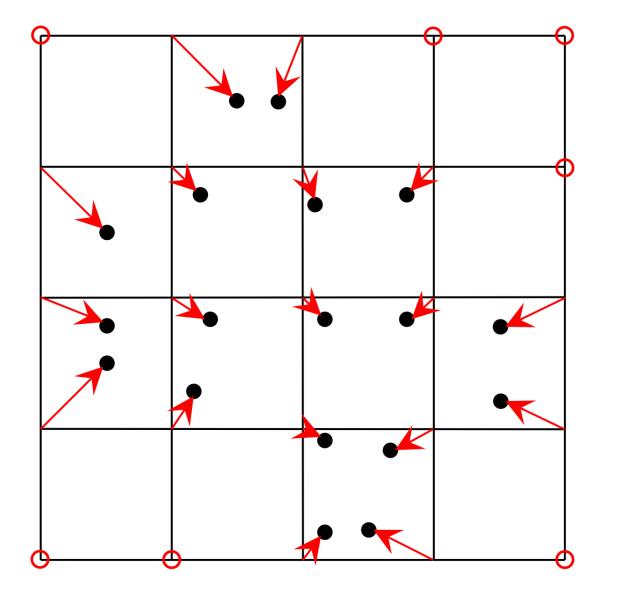
Generation



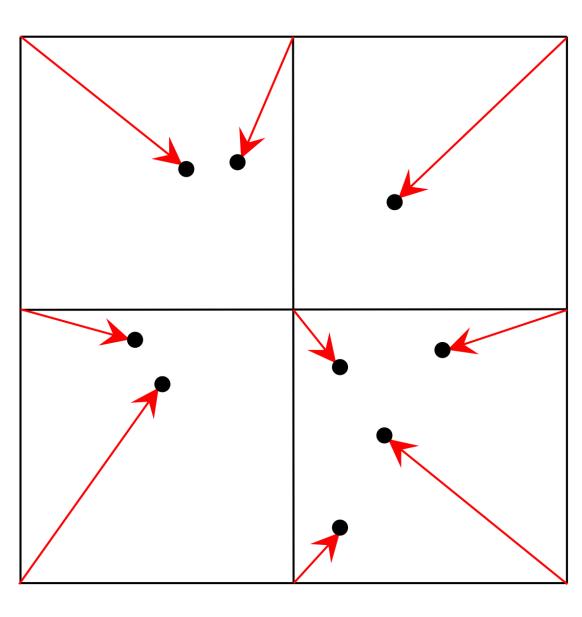
Simulation Grid 0-th level



Level-1 grid over simulation grid



Level-1 grid

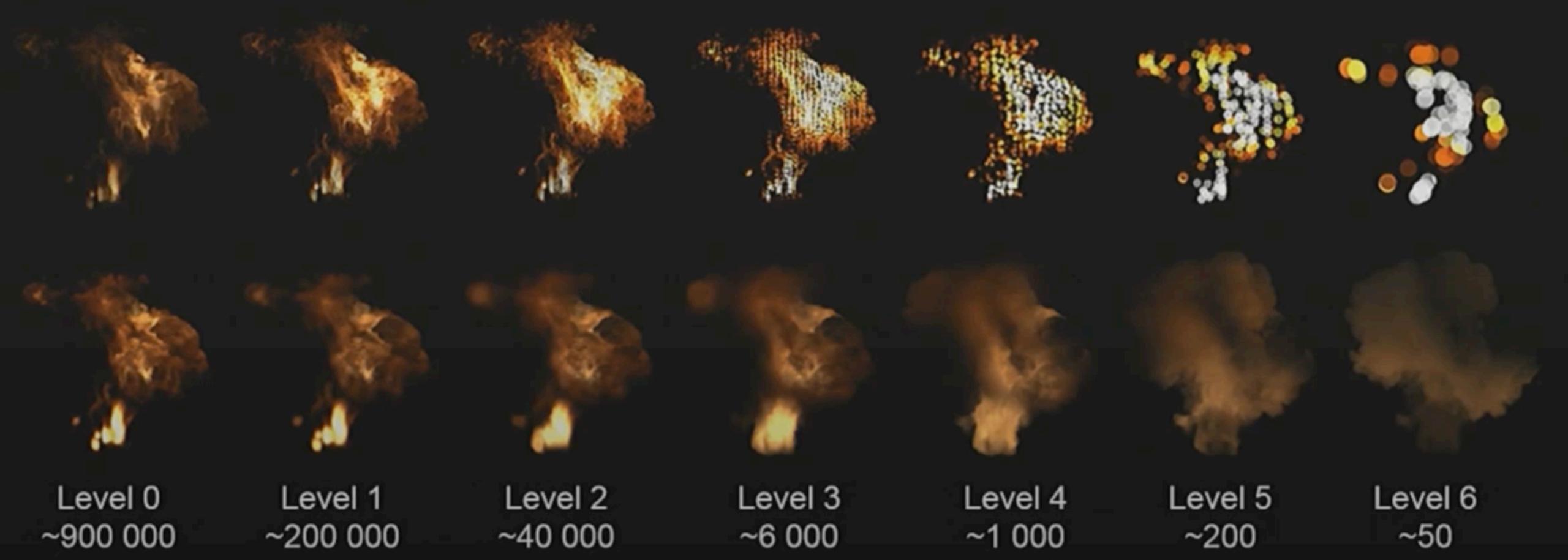


Level-2 grid

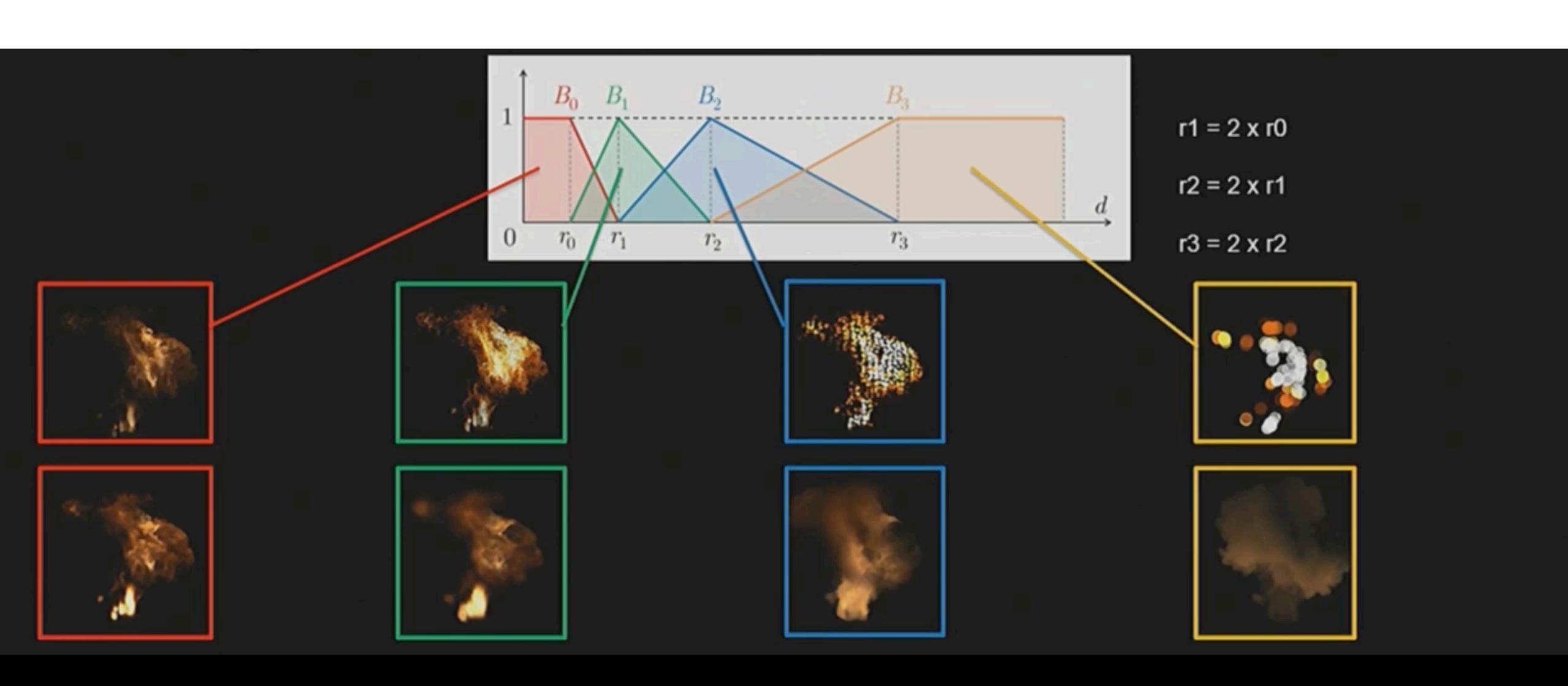




Multi-resolution representation of illumination:

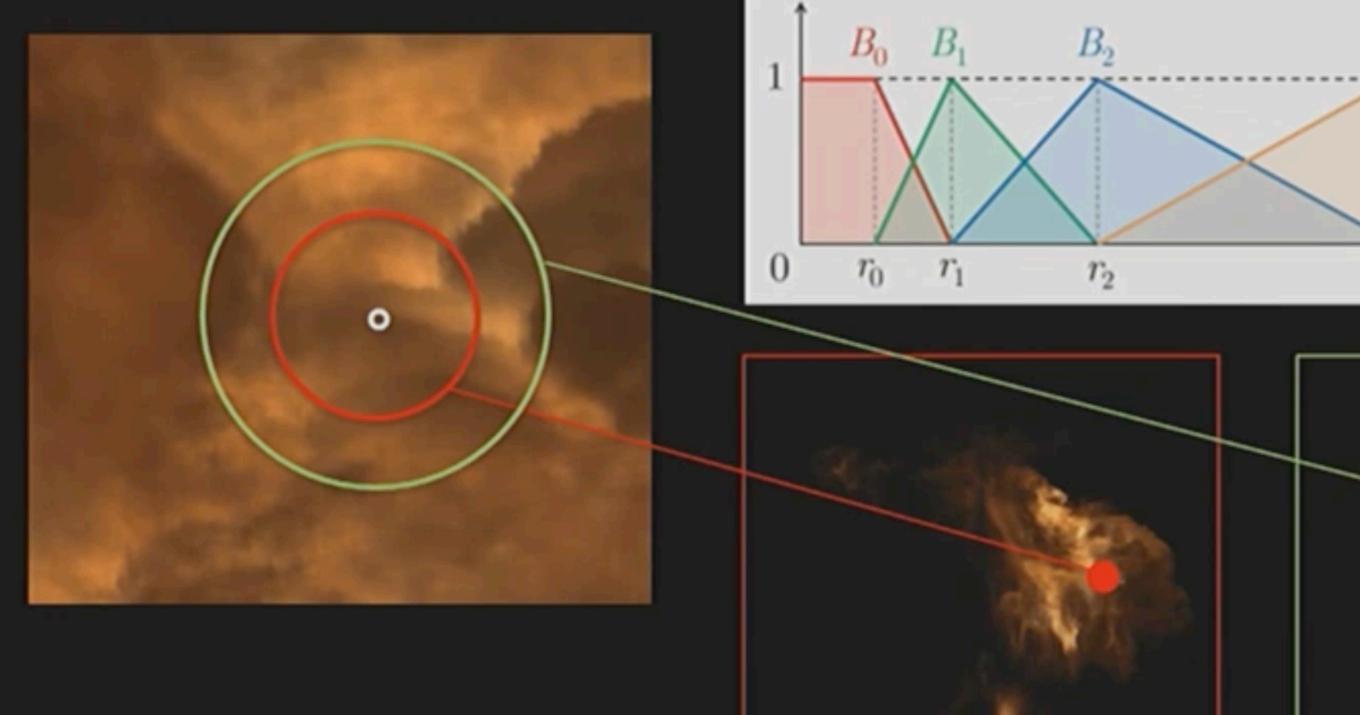


Blending Function



Shading

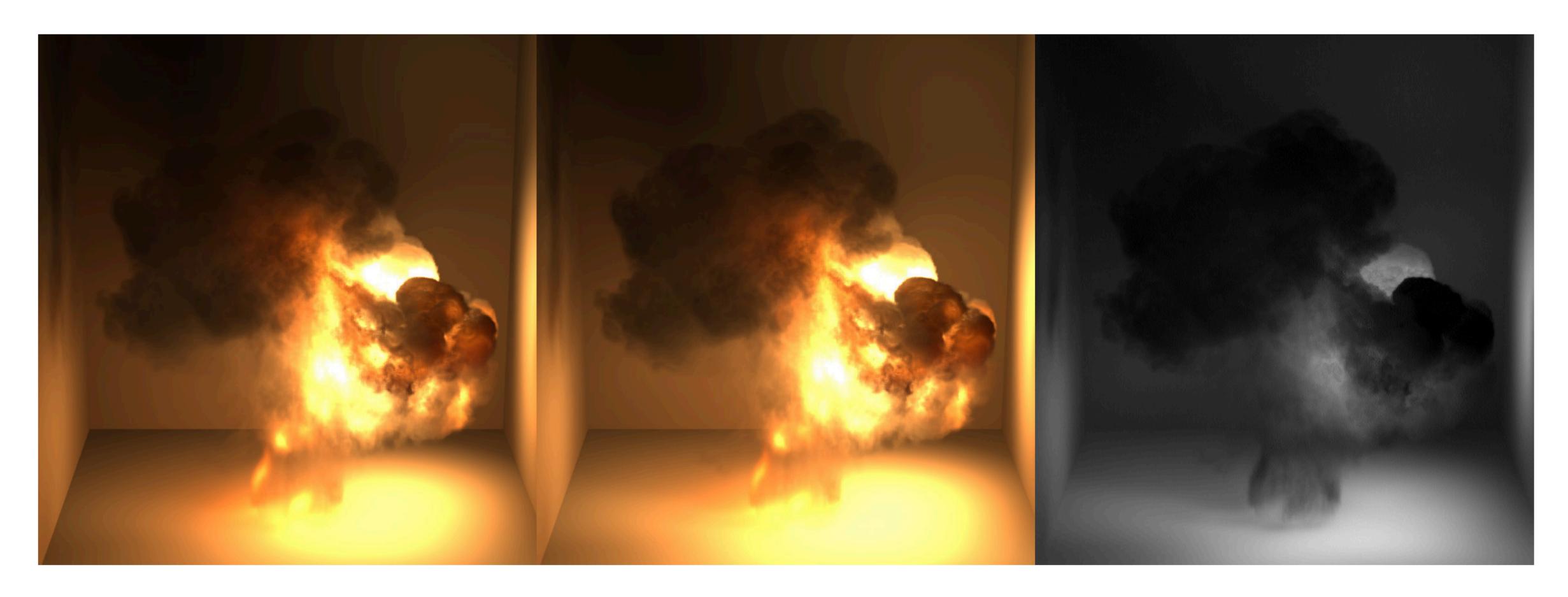
Shading a point:



Level 0



Multiple Scattering



Single Scattering

Multiple Scattering

Luminance Diff. ×2





Path Tracing

Lighting Grid Hierarchcy

Path Tracing

Lighting Grid Hierarchcy

Acknowledgements

We thank all the people who make their work available online which helped shape these slides.

Special thanks to Walter et al., Hasan et al., Carsten Dachsbacher et al., Jan Novak et al. for making their EG STAR Many-Light methods slides online, Can Yuksel and Cem Yuksel for making their slides and videos available online.

Scalable Realistic Rendering with Many-Light Methods [Dachsbacher et al. 2013]

Virtual Ray Lights for Rendering Scenes with Participating Media [Jan Novak et al. 2012]

LightCuts [Walter et al. 2005]

Stochastic LightCuts [Cem Yuksel 2019]



