

# Manifold Path Guiding for Importance Sampling Specular Chains

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## Connecting STORIES

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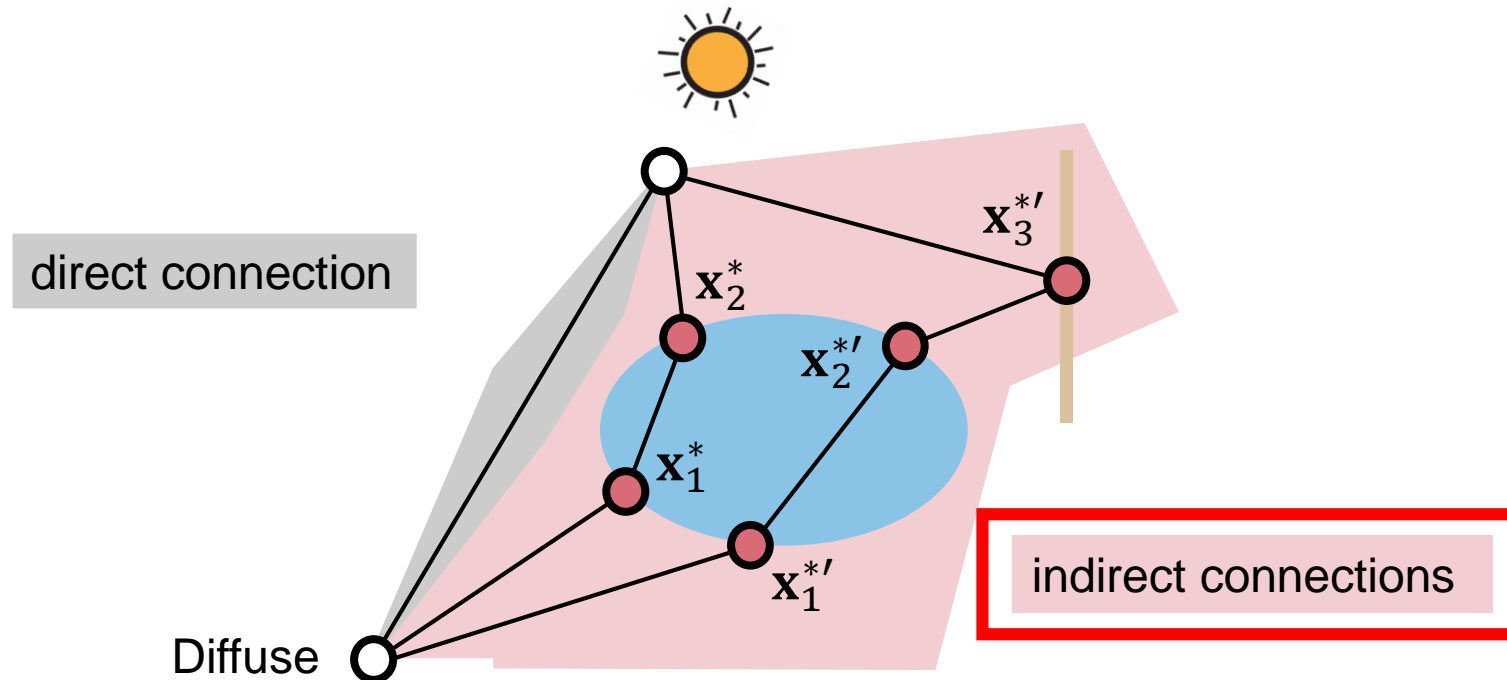
# Caustics

- **Complex visual effects**
  - high-frequency details
  - curved specular surfaces
- **Produced by specular paths**
  - containing multiple specular vertices
- **A long-standing challenge** to unbiased Monte Carlo rendering



# Specular chains connecting two endpoints

- Given a **point light** and a **shading point**
- How to connect them through **specular vertices**?





# Path tracing

- **Most popular technique**
  - in physically-based rendering
- **Hard to generate caustics**
  - extremely high variance
  - fireflies



# Path guiding

- **Prevalent technique in recent years**
- **Refining sampling distributions**
  - fitted from historical samples
- **Not good even if trained properly**
  - easily miss patterns

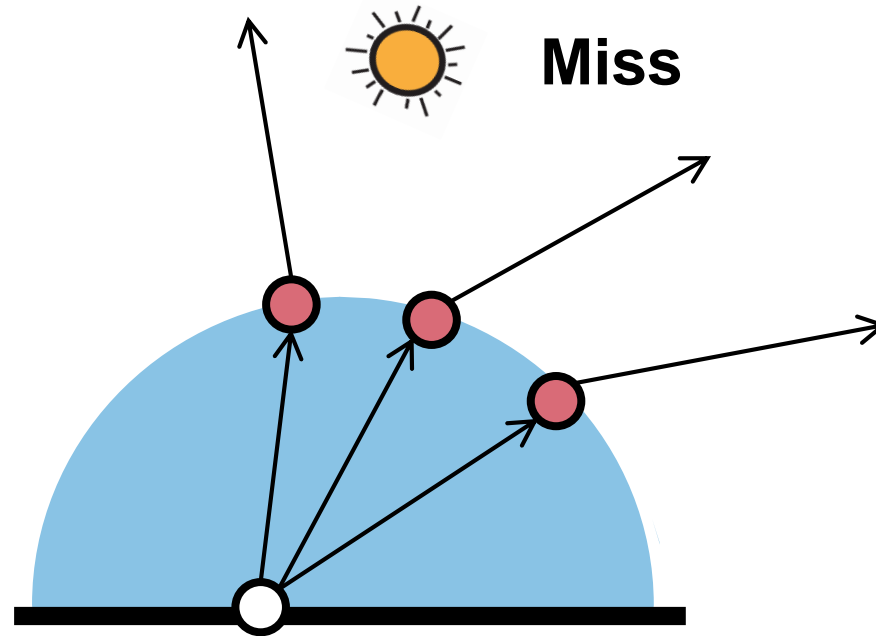




Ours (equal time)



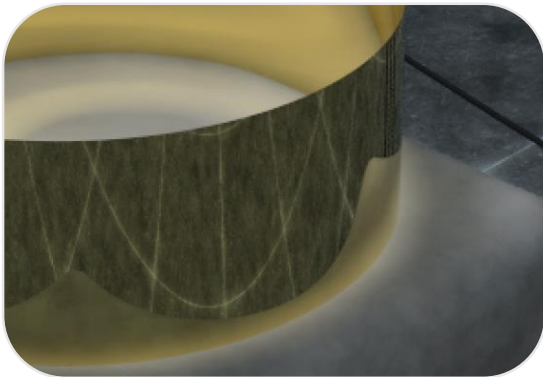
# Local sampling methods



**Nearly zero probability to sample a valid path!**

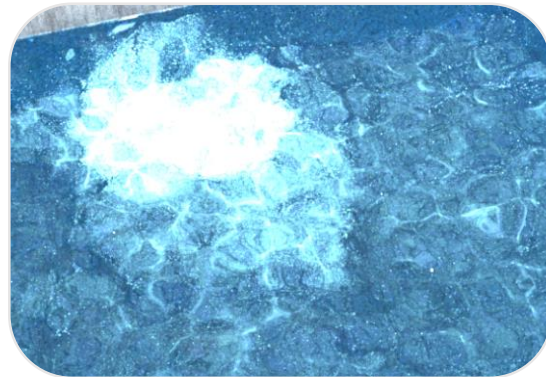


# Prior works



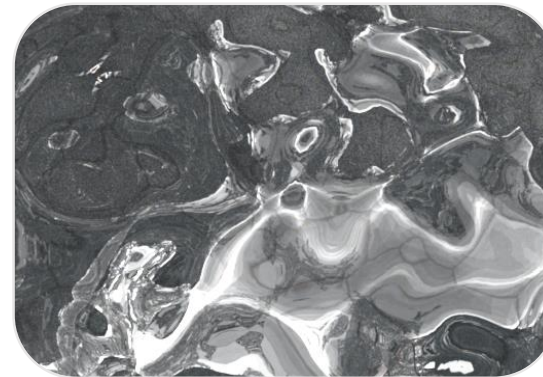
**SPPM**

[Hachisuka et al. 2009]



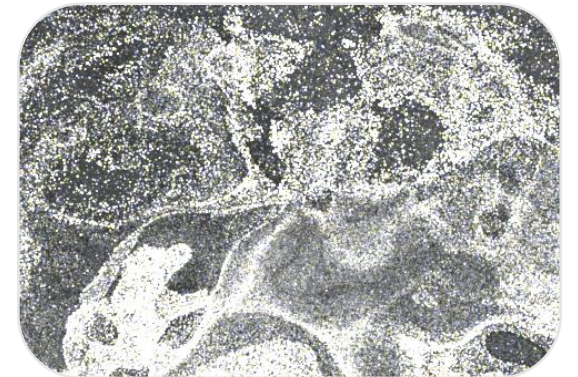
**MEMLT**

[Jakob and Marschner 2012]



**MNEE**

[Hanika et al. 2015]



**SMS**

[Zeltner et al. 2020]



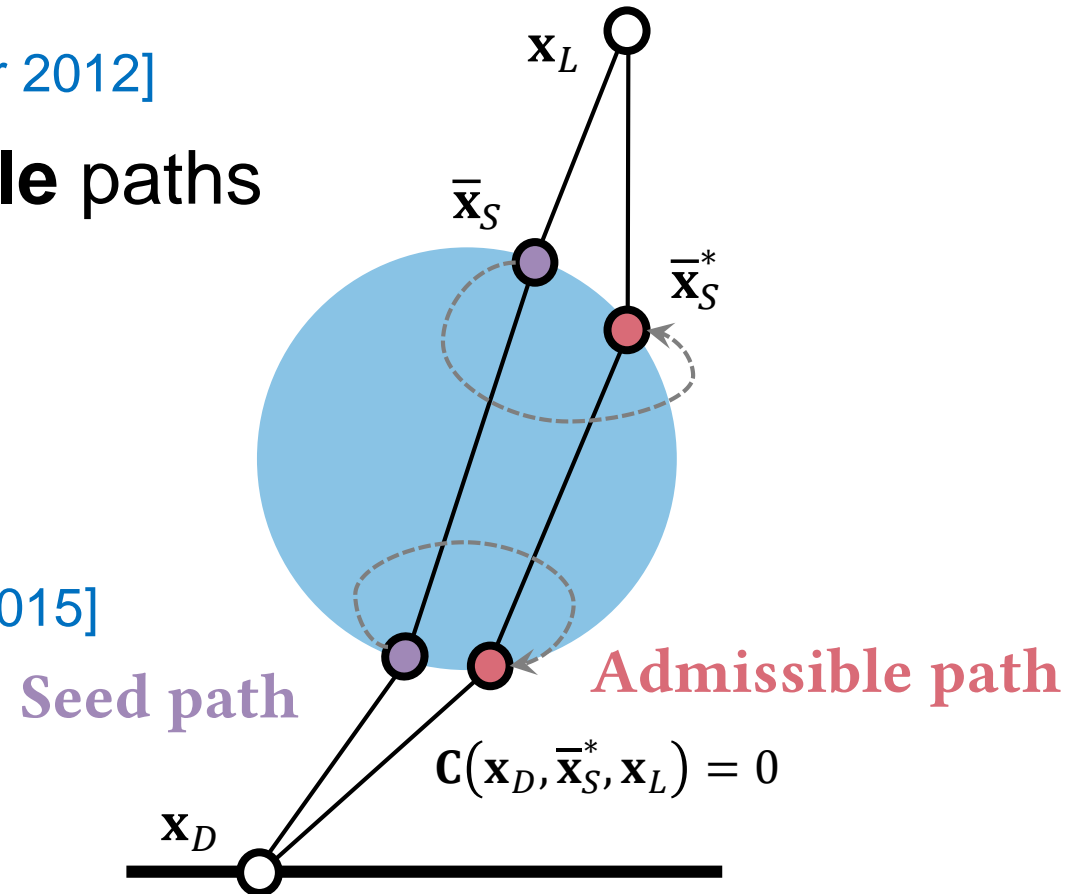
# Manifold sampling

## Manifold Walks [Jakob and Marschner 2012]

- Convert **seed** paths into **admissible** paths
- Work like Newton's iteration

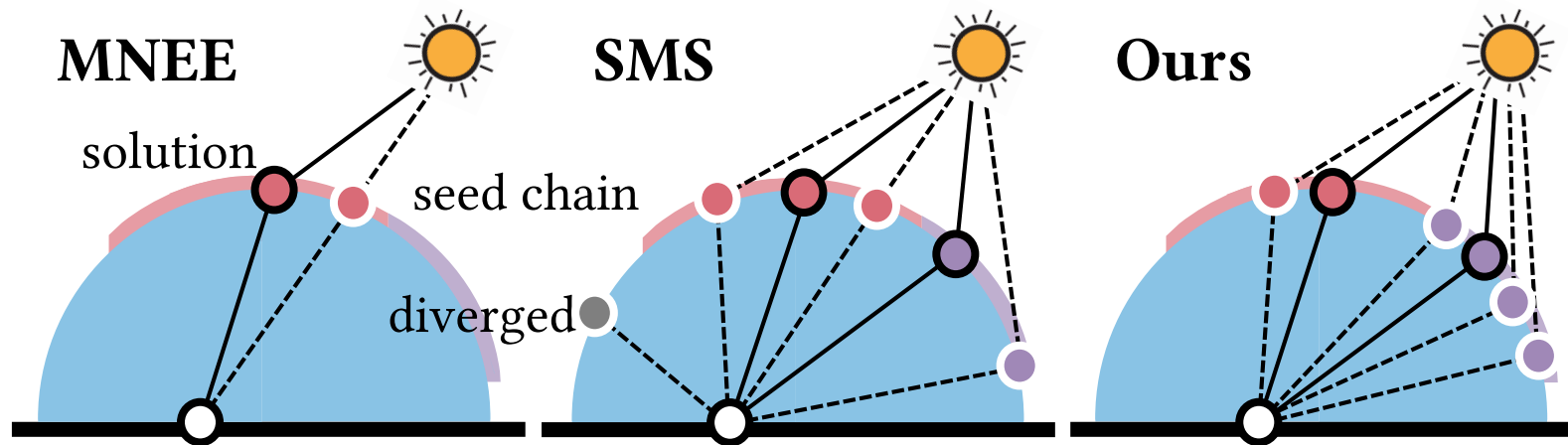
## How to sample seed paths?

- heuristically generated [Hanika et al. 2015]
- uniformly sampled [Zeltner et al. 2020]



# Motivation

- MNEE finds **at most one solution**, resulting in **energy loss**
- SMS **uniformly** samples the seeds, which leads to **high variance**
- Goal: find paths that are not only **admissible** but also “**important**”





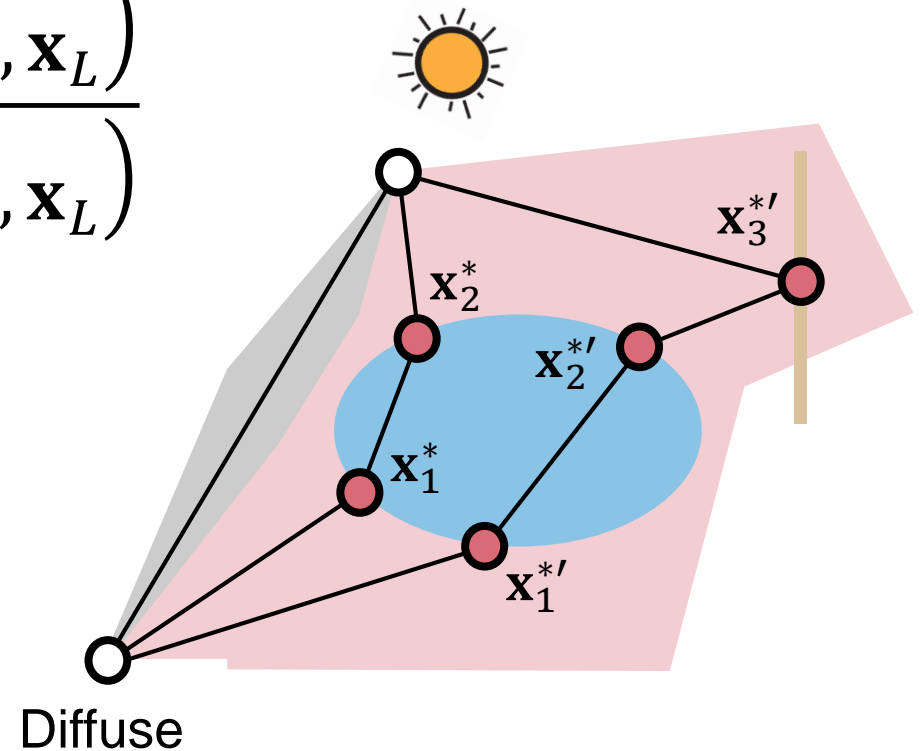
# Problem formulation

- MC estimator

$$\left\langle \sum_{\bar{\mathbf{x}}_S^* \in \mathcal{P}_S^*} T(\mathbf{x}_D, \bar{\mathbf{x}}_S^*, \mathbf{x}_L) \right\rangle = \frac{1}{N} \sum_{i=1}^N \frac{T(\mathbf{x}_D, \bar{\mathbf{x}}_S^{*(i)}, \mathbf{x}_L)}{P(\bar{\mathbf{x}}_S^{*(i)} | \mathbf{x}_D, \mathbf{x}_L)}$$

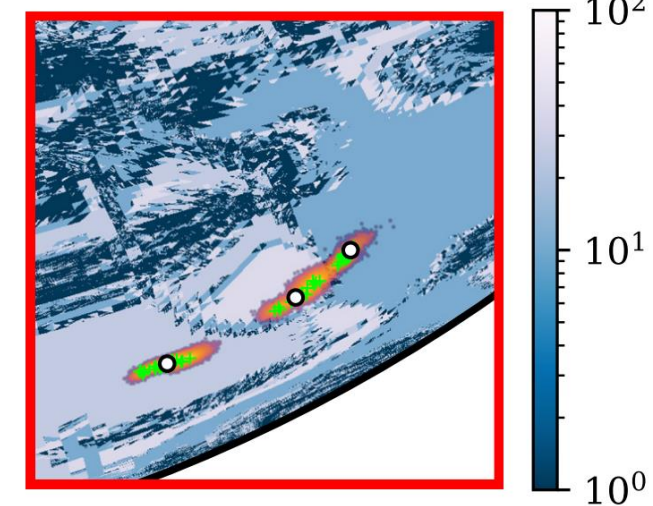
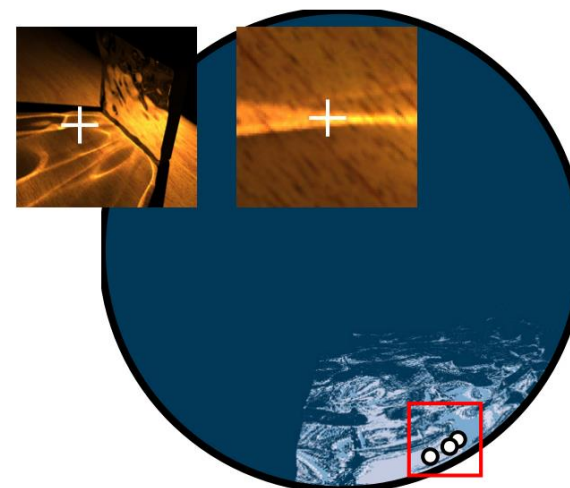
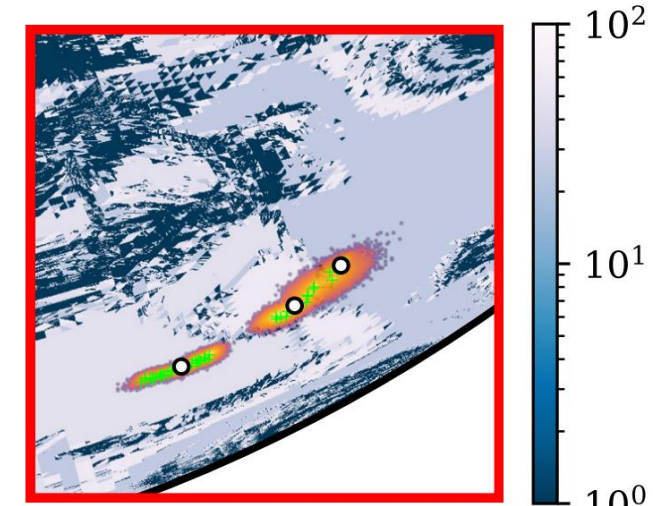
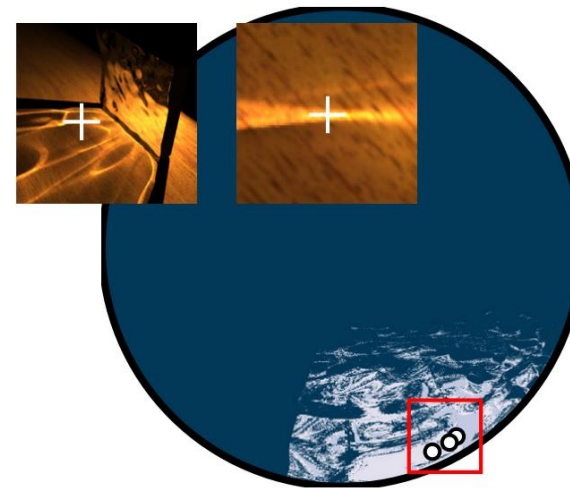
- Optimal importance sampling

$$P(\bar{\mathbf{x}}_S^* | \mathbf{x}_D, \mathbf{x}_L) \propto T(\mathbf{x}_D, \bar{\mathbf{x}}_S^*, \mathbf{x}_L)$$



# Exploitation of coherence

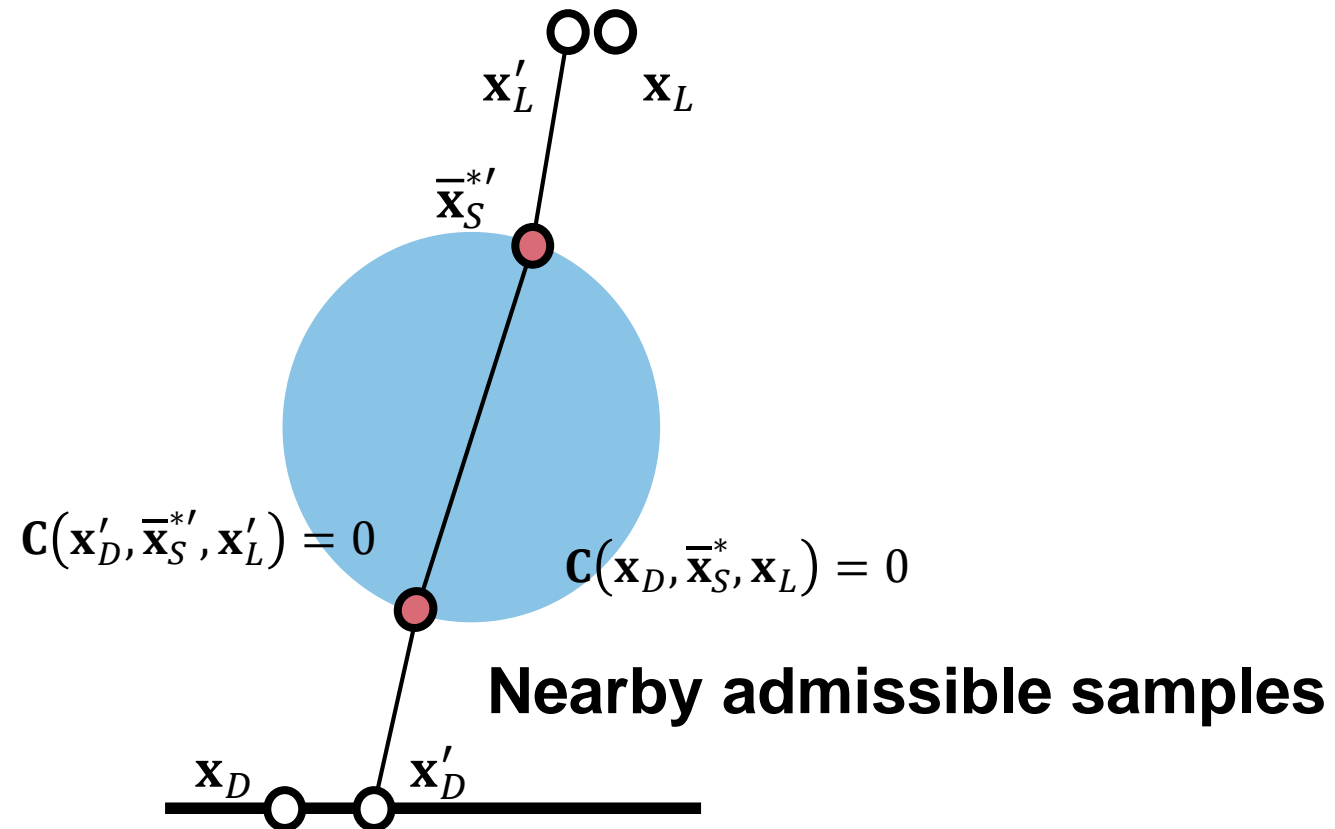
- Admissible chain of nearby configurations can be used as seeds to converge to an admissible chain of current configuration
  - Throughputs of those two admissible chains are close
- Importance sampling nearby solutions achieves importance sampling current ones!





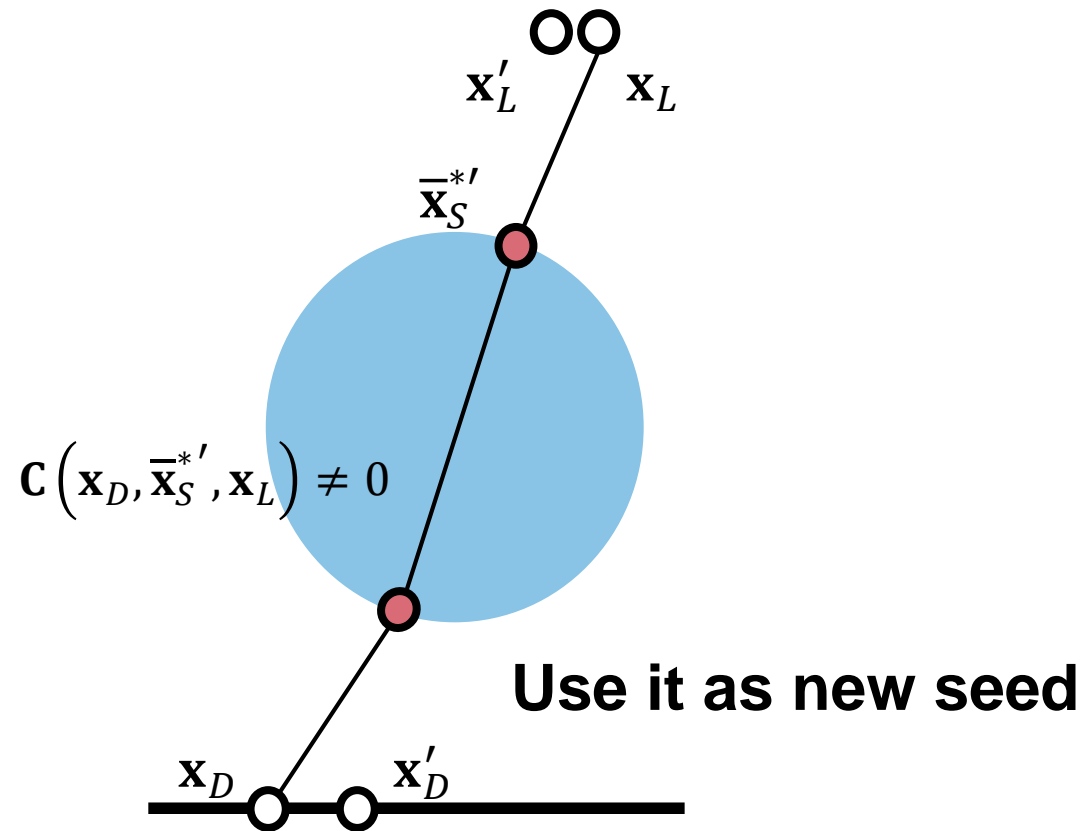
# Importance seed distributions

Utilize nearby admissible samples to generate the current seed



# Importance seed distributions

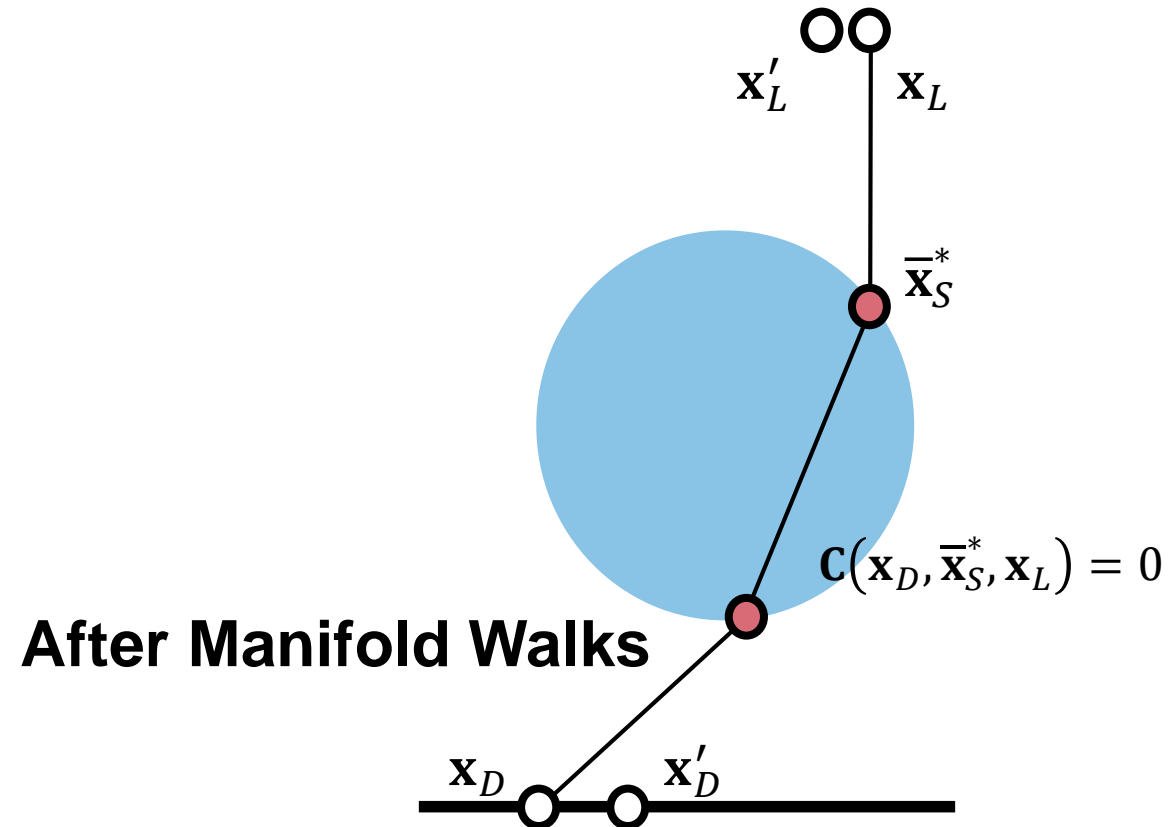
Utilize nearby admissible samples to generate the current seed



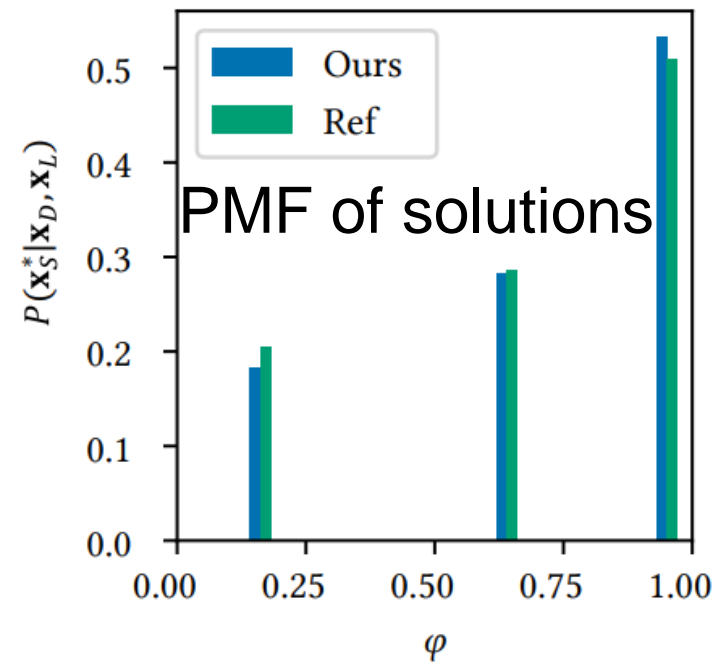
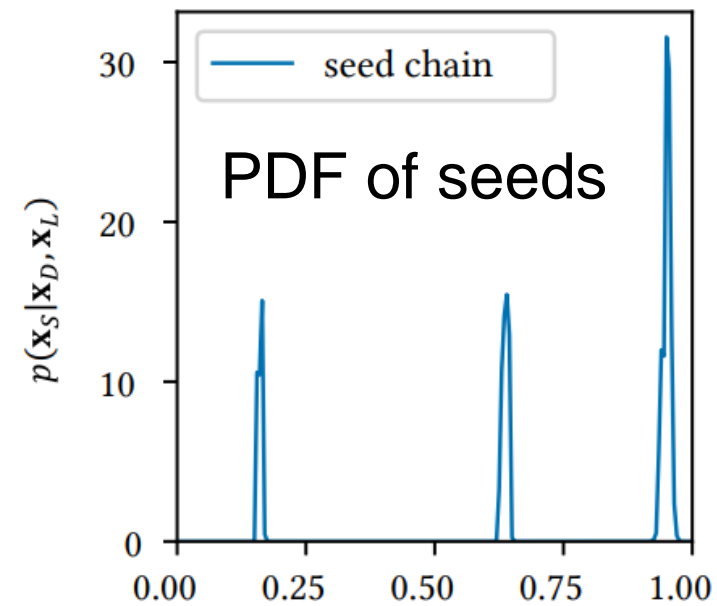
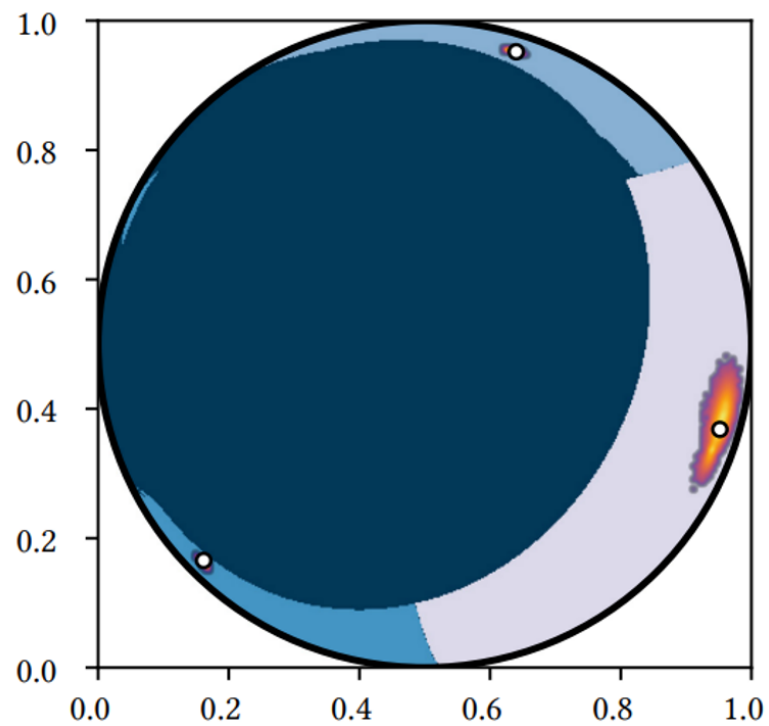
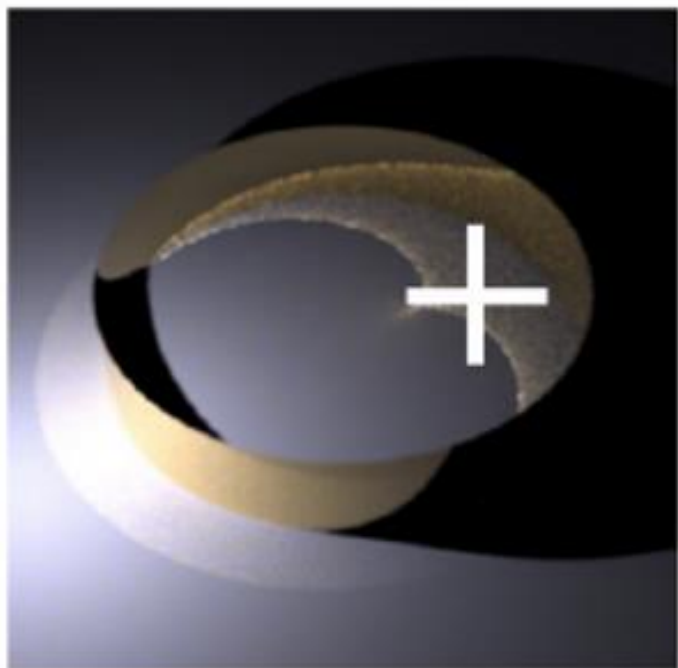


# Importance seed distributions

Utilize nearby admissible samples to generate the current seed

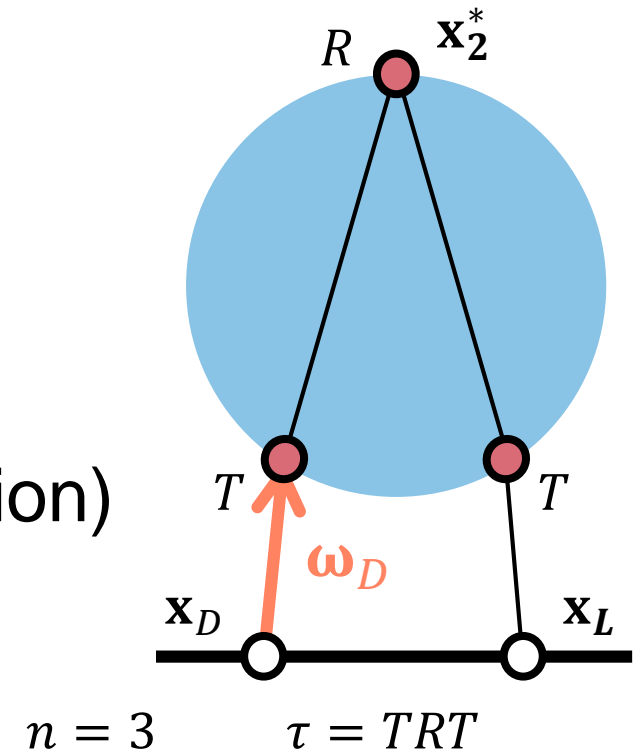


# Validation



# Dimensionality reduction

- Use all vertices to represent a specular chain?
- The dimensionality is too high!
- What we only need:
  - The length  $n$  (for ending)
  - The scattering type  $\tau$  (reflection / refraction)
  - The first direction  $\omega_D$



# Importance sampling seed chains

- We can sample the three terms in order

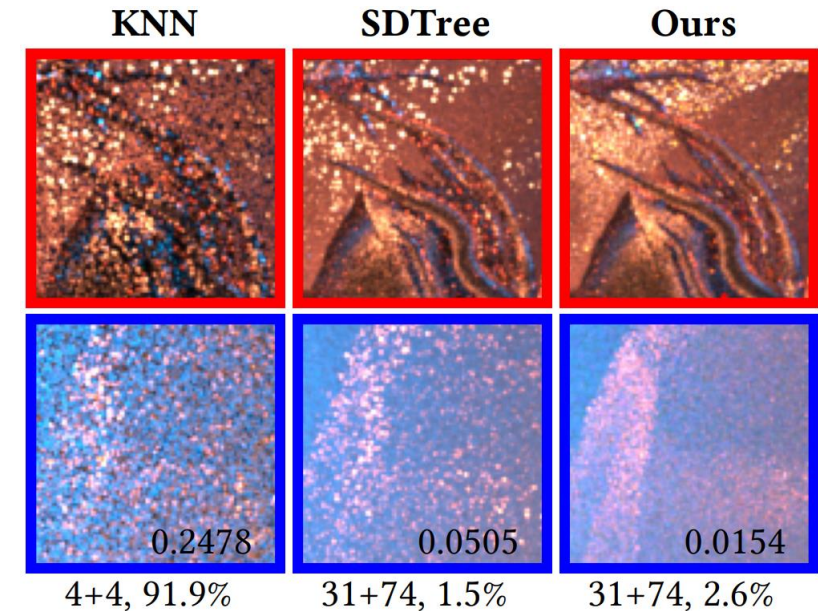
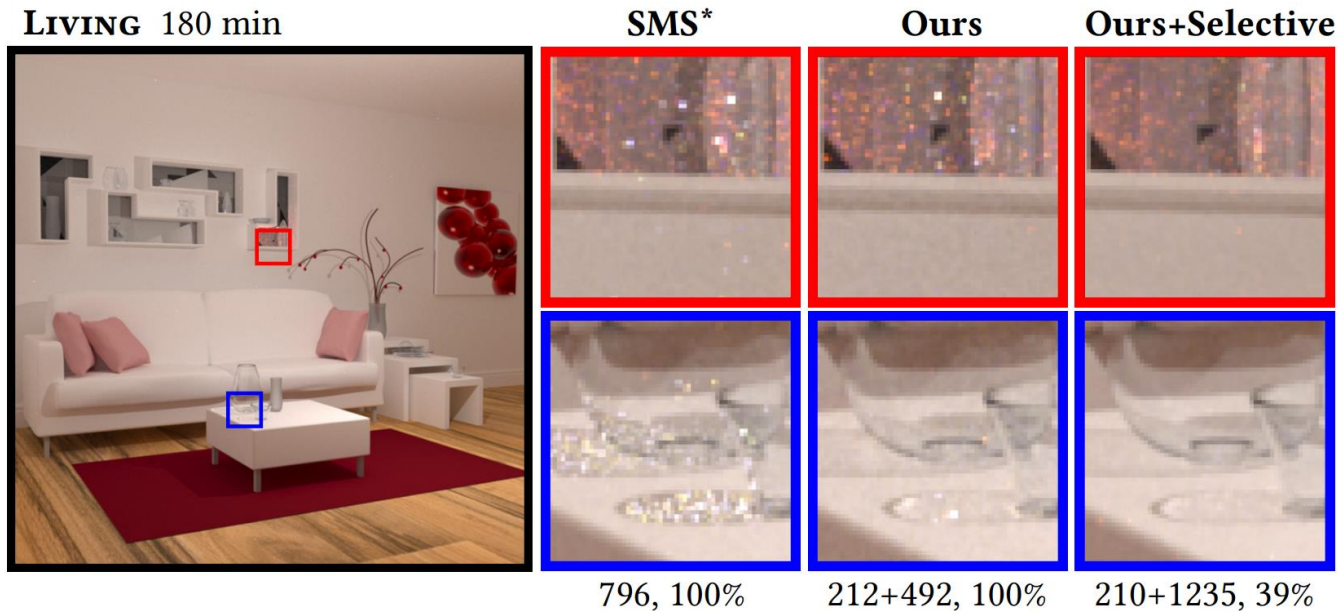
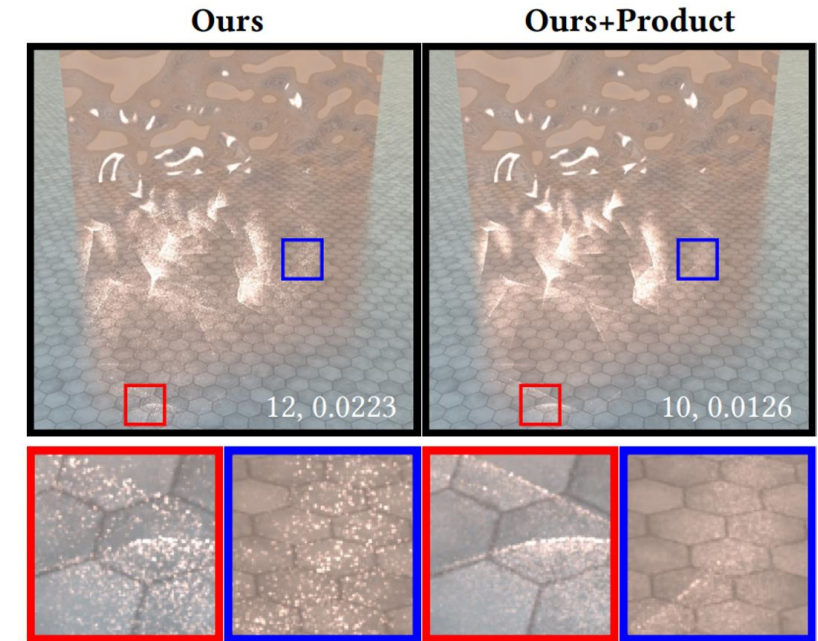
$$p(\mathbf{x}_S | \mathbf{x}_D, \mathbf{x}_L) = P(n | \mathbf{x}_D, \mathbf{x}_L) P(\tau | \mathbf{x}_D, \mathbf{x}_L, n) p(\boldsymbol{\omega}_D | \mathbf{x}_D, \mathbf{x}_L, \tau)$$

- Fitting distributions from nearby sub-path samples
- **Sampling the chain length  $n$  and the scattering type  $\tau$** 
  - Building discrete distributions
- **Sampling the direction  $\boldsymbol{\omega}_D$** 
  - Density estimation
  - No analytic PDF eval required! (Diff. from conventional guiding)

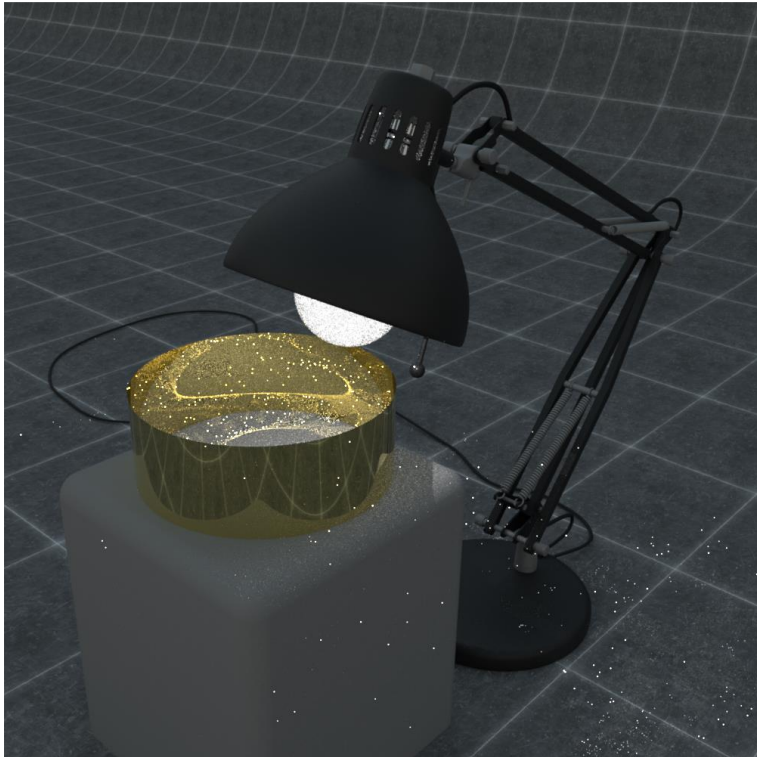


# Practical considerations

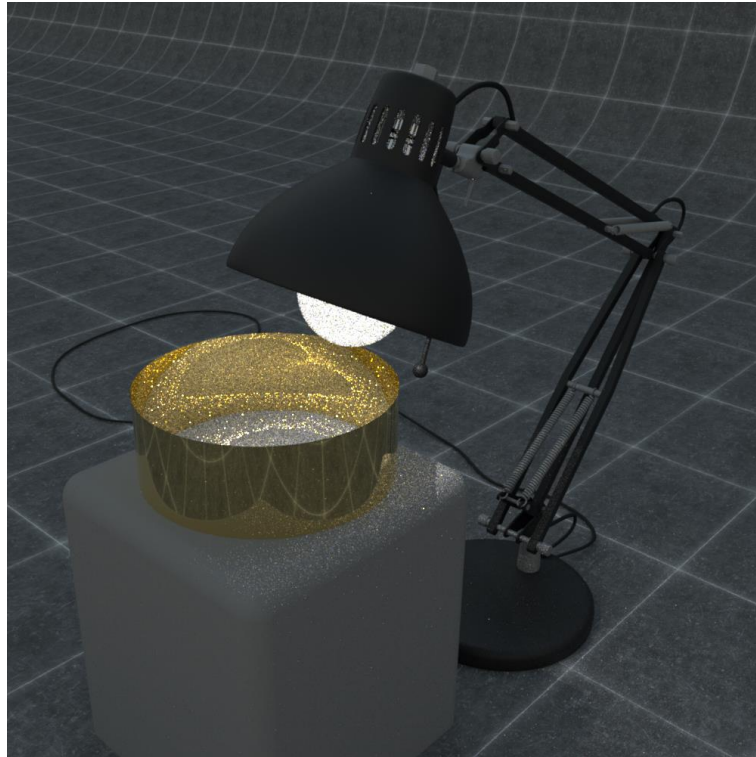
- Spatio-directional structures
- Product importance sampling
- Selective activation



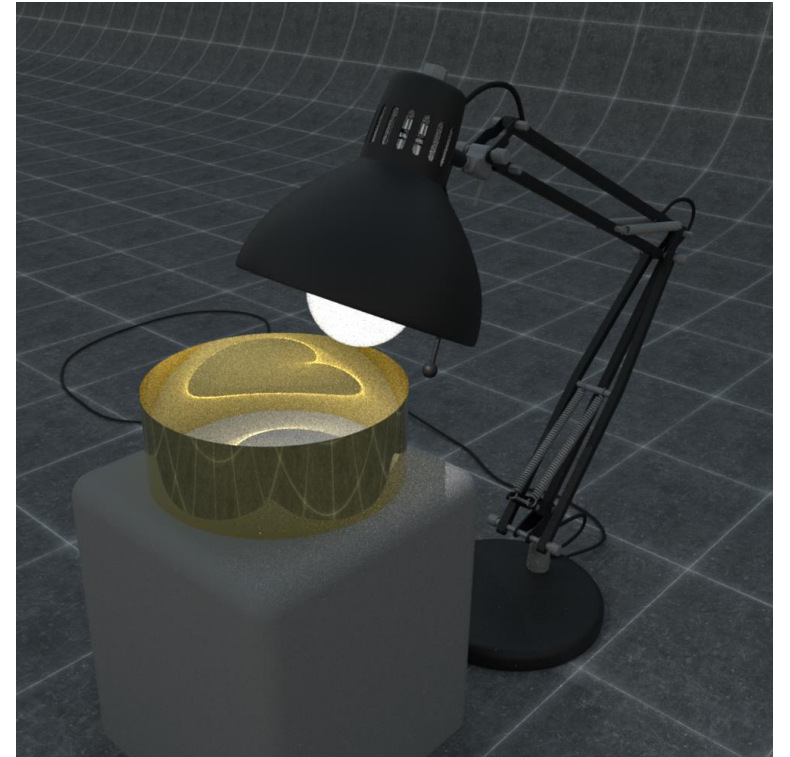
# Equal-time comparisons



PAVMM Path Guiding  
[Ruppert et al. 2020]



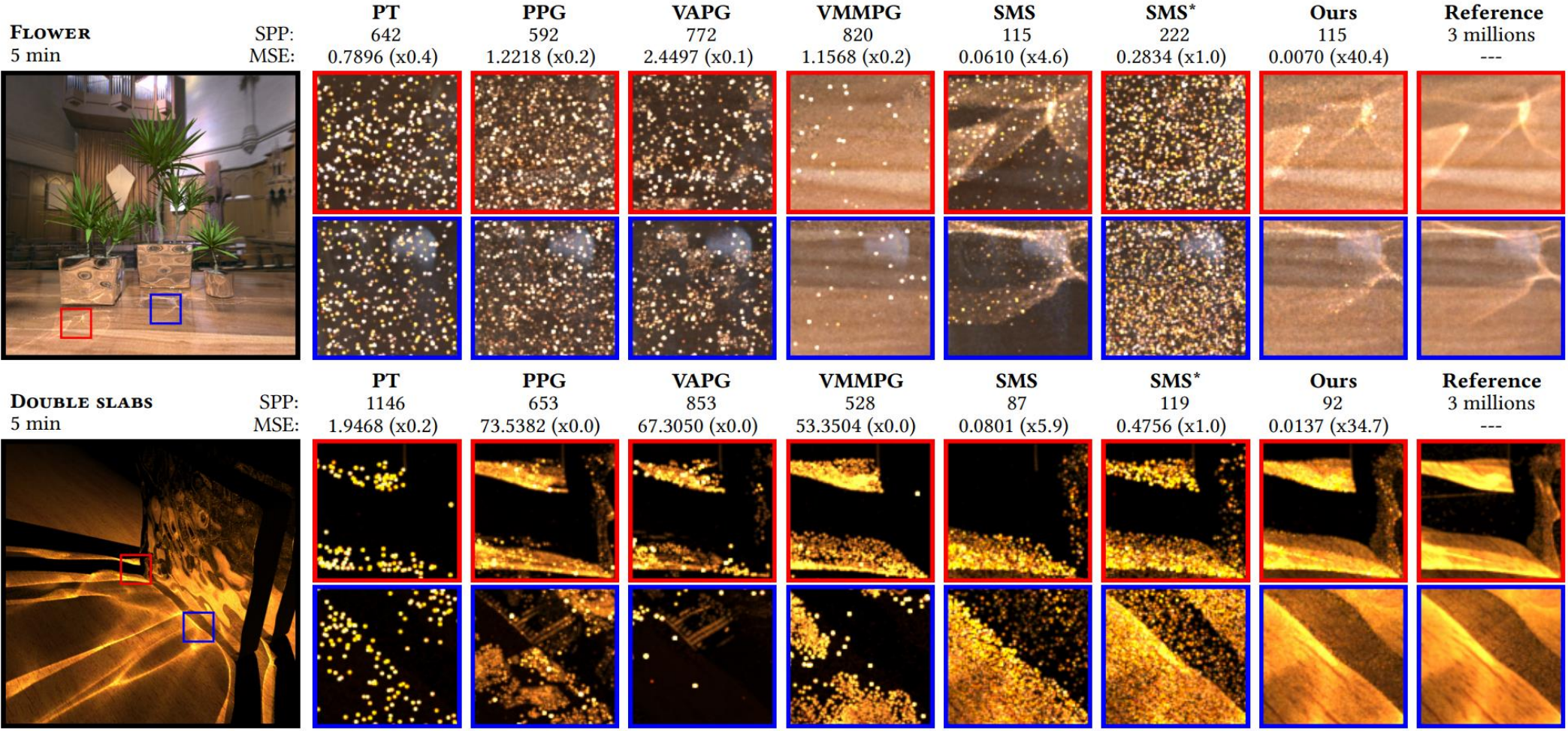
SMS\*  
[Zeltner et al. 2020]



Ours



# Equal-time comparisons



# Conclusion

- The first generic method for importance sampling specular chains
- Seed chain sampling plays an important role
- Make manifold-based methods more efficient
- Promote new interests in MC rendering
  
- Source code and scripts available on GitHub



GitHub Repo





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