# Rendering specular light paths 

UCSD CSE 272
Advanced Image Synthesis

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## Back to SDS light paths



## Photon mapping fails when light source is very far away



## Can we directly find this light path?



## Idea: find light paths by solving a non-linear equation

diffuse
point light


# Illumination from Curved Reflectors 

## Idea: find light paths by solving a non-linear equation

diffuse
point light


# Illumination from Curved Reflectors 

## Idea: find light paths by solving a non-linear equation

- solve $x_{2}$ using Newton's method: start from an initial guess, iteratively improve


$$
C\left(x_{2}\right)=0
$$



## Idea: find light paths by solving a non-linear equation

- solve $x_{2}$ using Newton's method: start from an initial guess, iteratively improve

$$
C\left(x_{2}+\Delta x_{2}\right) \approx C\left(x_{2}\right)+J_{C}\left(x_{2}\right) \Delta x_{2}=0
$$

$$
C\left(x_{2}\right)=0
$$

$$
\Delta x_{2}=-J_{C}\left(x_{2}\right)^{-1} C\left(x_{2}\right)
$$



## Idea: find light paths by solving a non-linear equation

- solve $x_{2}$ using Newton's method: expand constraint C using first-order Taylor expansion

$$
C\left(x_{2}^{\prime}+\Delta x_{2}\right) \approx C\left(x_{2}^{\prime}\right)+J_{C}\left(x_{2}^{\prime}\right) \Delta x_{2}=0
$$

$$
\Delta x_{2}=-J_{C}\left(x_{2}^{\prime}\right)^{-1} C\left(x_{2}^{\prime}\right)
$$

$$
\begin{aligned}
& \text { start from an initial guess x_2' } \\
& \text { while }\left\|C\left(x_{2}^{\prime}\right)\right\|>\epsilon \text { : } \\
& \mathrm{x}^{\prime} 2^{\prime}=\mathrm{x}, 2^{\prime}-J_{C}\left(x_{2}^{\prime}\right)^{-1} C\left(x_{2}^{\prime}\right) \\
& \mathrm{x} \_2=\mathrm{x} \text {-2 }{ }^{\text {' }}
\end{aligned}
$$

$$
C\left(x_{2}\right)=0
$$



## A single triangle case without shading normal

- $n$ is fixed, find a point $x_{2}$ on the plane s.t. the constraint is satisfied
- unique solution exists



## A single triangle case with shading normal

- $n$ interpolates $n_{a^{\prime}} n_{b^{\prime}} n_{c}$ based on the position of $x_{2}$
- may have zero, one, or multiple solutions



## Easily generalizable to multiple specular surfaces



$$
\begin{aligned}
& \text { given } x_{1} \text { and } x_{5} \\
& \text { find } x_{2}, x_{3}, x_{4} \text { s.t. } \\
& \frac{-\omega_{1}+\omega_{3}}{\left\|-\omega_{1}+\omega_{3}\right\|}=n_{2} \\
& \frac{-\omega_{2}+\omega_{4}}{\left\|-\omega_{2}+\omega_{4}\right\|}=n_{3} \\
& \frac{-\omega_{3}+\omega_{5}}{\left\|-\omega_{3}+\omega_{5}\right\|}=n_{4} \\
& C\left(x_{2}, x_{3}, x_{4}\right)=0
\end{aligned}
$$

Theory and Application of Specular Path Perturbation

## Challenge: incorporate Newton's method in a Monte Carlo renderer

- how do we generate initial guesses?
- how do we handle a large number of triangles?
- what is the probability density?

$$
\begin{aligned}
& \text { start from an initial guess x_2' } \\
& \text { while }\left\|C\left(x_{2}^{\prime}\right)\right\|>\epsilon: \\
& \quad \mathrm{x} \_2,=\mathrm{x}^{\prime} 2^{\prime}-J_{C}\left(x_{2}^{\prime}\right)^{-1} C\left(x_{2}^{\prime}\right) \\
& \mathrm{x} 2=\mathrm{x}-2^{\prime} \\
& p\left(x_{2}\right)=?
\end{aligned}
$$



# Three strategies to incorporate Newton's method into a renderer 

use new data structure to enumerate roots
Single Scattering in Refractive Media with Triangle Mesh Boundaries


Path Cuts: Efficient Rendering of Pure Specular Light Transport
BEIBEI WANG, School of Computer Science and Engineering, Nanjing University of Science and Technolog
MILOS̃ HAŠAN, Adobe Research
LING-QI YAN, University of California, Santa Barbara


Manifold Exploration: A Markov Chain Monte Carlo technique for rendering scenes with difficult specular transport

Wenzel Jakob
Con


Specular Manifold Sampling for Rendering High-Frequency Caustics and Glints
TIZIAN ZELTNER, École Polytechnique Fédérale de Lausanne (EPFL), Switzerland
WENZEL JAKOB, Ecole Polytechniquue Fedederale de Lausanne (EPFL), Switzerland

randomized initialization using Monte Carlo sampling

# Three strategies to incorporate Newton's method into a renderer 



Manifold Exploration: A Markov Chain Monte Carlo technique for rendering scenes with difficult specular transport

$$
\text { Wenzel Jakob } \quad \text { Steve Marschner }
$$

Cornell University


Metropolis light transport

Specular Manifold Sampling for Rendering High-Frequency Caustics and Glints
TIZIAN ZELTNER, École Polytechnique Fédérale de Lausanne (EPFL), Switzerland
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randomized initialization using Monte Carlo sampling

## Idea: enumerate roots using a data structure

- observation: most triangles contain no solution given $x_{1}$ and $x_{3}$

note: there are only countably many roots if the scene is made of triangle meshes


## Hierarchical pruning using a 6D tree

- skip the whole subtree if it is impossible that half-vector would be the same as the normal


Single Scattering in Refractive Media with Triangle Mesh Boundaries
$\qquad$ Bruce Walter
Cornell University INRIA - LJK
$x_{1}$
all possible $n$

all possible $x_{2}$

## Hierarchical pruning generalizes to multiple bounces



Path Cuts: Efficient Rendering of Pure Specular Light Transport
BEIBEI WANG, School of Computer Science and Engineering, Nanjing University of Science and Technology

## Optional: subdivide triangles

## with shading normals for more accurate results

- can subdivide until the constraint is provably convex



## In practice, triangle subdivision is usually not worth it

- subdivision can find a few more paths, but usually gives visually similar results

w/o Interval Netwon (TT) w/ Interval Netwon (TT)

Glint count: 1036,
Time: 0.62 s

Glint count: 1052,
Time: 23 h

## Fancy animations



## Fancy images



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## Fancy images



Path tracing (TTTT) Time: $7.92 \mathrm{~h}, \mathrm{spp}: 64 \mathrm{~K}$

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Metropolis light transport

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randomized initialization using Monte Carlo sampling

## Let's solve a slightly relaxed problem

diffuse
pinhole camera
mirror (specular)

small area light


# Manifold exploration with Metropolis sampling 

- use bidirectional path tracing to find an initial path



## Manifold exploration with Metropolis sampling

- mutate the camera subpath until a diffuse hit



## Manifold exploration with Metropolis sampling

- given $x_{3}^{\prime}$ and $x_{5}$, perturb $x_{4}$ using Newton's method to satisfy the constraint



## Manifold exploration with Metropolis sampling

- works for arbitrary number of specular vertices



## Satisfying detailed balance

- in Metropolis, we only need to compute the ratio of PDFs, making PDF calculation much easier



## Extension to glossy surfaces

- probabilistically determine whether a surface is specular or not based on roughness
- use the sampled micro-normal as the specular normal



## Extension to volumetric light transport

- Henyey-Greenstein with high g can be seen as near-specular phase functions



## Metropolis light transport in Mitsuba

- first open source implementation of Veach-style MLT 15 years after Veach's publication!

| 8. Plugin reference |  | 8.10. Integrators |
| :---: | :---: | :---: |
| 8.10.11. Path Space Metropolis Light Transport (mlt) |  |  |
| Parameter | Type | Description |
| maxDepth | integer | Specifies the longest path depth in the generated output image (where -1 corresponds to $\infty$ ). A value of 1 will only render directly visible light sources. 2 will lead to singlebounce (direct-only) illumination, and so on. (Default: -1) |
| directSamples | integer | By default, the implementation renders direct illumination component separately using the direct plugin, which uses low-discrepancy number sequences for superior performance (in other words, it is not handled by MLT). This parameter specifies the number of samples allocated to that method. To force MLT to be responsible for the direct illumination component as well, set this to -1 . (Default: 16) |
| luminanceSamples | integer | MLT-type algorithms create output images that are only relative. The algorithm can e.g. determine that a certain pixel is approximately twice as bright as another one, but the absolute scale is unknown. To recover it, this plugin computes the average luminance arriving at the sensor by generating a number of samples. (Default: 100000 samples) |
| twoStage | boolean | Use two-stage MLT? See pssmlt for details. (Default: false) |
| bidirectional ${ }^{2}$ <br> Mutation, <br> [lens,multiChain, caustic, manifold] $\kappa$ Perturbation | boolean | These parameters can be used to pick the individual mutation and perturbation strategies that will be used to explore path space. By default, the original set by Veach and Guibas is enabled (i.e. everything except the manifold perturbation). It is possible to extend this integrator with additional custom perturbations strategies if needed. |
| lambda | float | Jump size of the manifold perturbation (Default 50) |


(a) Lens perturbation

(c) Multi-chain perturbation

(b) Caustic perturbation

(d) Manifold perturbation

## Fancy images


(a) MLT
(b) ERPT

(c) PSSMLT
(d) MEPT

## Fancy images


(c) PSSMLT
(d) MEPT

## Fancy images



(a) MLT

(b) ERPT

(c) PSSMLT

(d) MEPT

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Metropolis light transport

## Specular Manifold Sampling for Rendering High-Frequency Caustics and Glints

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randomized initialization using Monte Carlo sampling

## Back to point lights

diffuse
pinhole camera

# Manifold exploration with normal Monte Carlo sampling 

- do normal path tracing before we hit the light

point light


## Manifold exploration with normal Monte Carlo sampling

- connect to the light source - quiz: what is the contribution of this light path?



## Manifold exploration with normal Monte Carlo sampling

- connect to the light source - contribution is zero since we are on a specular surface



## Manifold exploration with normal Monte Carlo sampling

- perturb $x_{4}$ to satisfy the specular constraint



## Manifold exploration with normal Monte Carlo sampling

- perturb $x_{4}$ to satisfy the specular constraint
- what is the PDF of the path $x_{1} x_{2} x_{3} x_{4}^{\prime} x_{5}$ ? diffuse



## PDF of a specular path is an integral

- the probability density of sampling $x_{1} x_{2} x_{3} x_{4}^{\prime} x_{5}$ is the sum of all probability densities of path that will perturb to it


$$
p\left(x^{\prime}\right)=\int p(x) p\left(x^{\prime} \mid x\right) \mathrm{d} x
$$

## PDF of a specular path is an integral

- the probability density of sampling $x_{1} x_{2} x_{3} x_{4}^{\prime} x_{5}$ is the sum of all probability densities of path that will perturb to it


$$
p\left(x^{\prime}\right)=\int p(x) p\left(x^{\prime} \mid x\right) \mathrm{d} x
$$



## Evaluating contribution

- need to use Monte Carlo sampling to estimate the PDF $p\left(x^{\prime}\right)$ itself

$$
\frac{f\left(x^{\prime}\right)}{p\left(x^{\prime}\right)}=\frac{f\left(x^{\prime}\right)}{\int p(x) p\left(x^{\prime} \mid x\right) \mathrm{d} x}
$$



$$
p\left(x^{\prime}\right)=\int p(x) p\left(x^{\prime} \mid x\right) \mathrm{d} x
$$

## Unbiased evaluation of reciprocal of integral

- same as the unbiased photon mapping paper

$$
\frac{1}{\int f(x) \mathrm{d} x}=\frac{1}{1-F}=1+F+F^{2}+\cdots
$$

can estimate using Russian roulette

## Pseudocode

```
ALGORITHM 2: Unbiased specular manifold sampling
    Input: Shading point \(\mathbf{x}_{1}\) and emitter position \(\mathbf{x}_{3}\) with density \(p\left(\mathbf{x}_{3}\right)\)
    Output: Estimate of radiance traveling from \(x_{3}\) to \(\mathbf{x}_{1}\)
    \(1 \mathrm{X}_{2} \leftarrow\) sample a specular vertex as initial position
    \(\mathbf{x}_{2}^{*} \leftarrow\) manifold_walk \(\left(\mathbf{x}_{1}, \mathbf{x}_{2}, \mathbf{x}_{3}\right)\)
    \(3\left\langle 1 / p_{k}\right\rangle \leftarrow 1 \quad \triangleright\) Estimate inverse probability of sampling \(\mathbf{x}_{2}^{*}\)
    while true do
        \(\mathbf{x}_{2} \leftarrow\) sample specular vertex as above
        \(\mathbf{x}_{2}^{\prime} \leftarrow\) manifold_walk \(\left(\mathbf{x}_{1}, \mathbf{x}_{2}, \mathbf{x}_{3}\right)\)
        if \(\left\|\mathbf{x}_{2}^{\prime}-\mathbf{x}_{2}^{*}\right\|<\varepsilon\) then
            break
        \(\left\langle 1 / p_{k}\right\rangle \leftarrow\left\langle 1 / p_{k}\right\rangle+1\)
    return \(f_{s}\left(\mathbf{x}_{2}^{*}\right) \cdot G\left(\mathbf{x}_{1} \leftrightarrow \mathbf{x}_{2} \leftrightarrow \mathbf{x}_{3}\right) \cdot\left\langle 1 / p_{k}\right\rangle \cdot L_{e}\left(\mathbf{x}_{3}\right) / p\left(\mathbf{x}_{3}\right)\)
```


## Fancy images



## Fancy images



# Manifold exploration is used in practice 



Manifold Next Event Estimation

## Manifold exploration is used in practice



Plausible Iris Caustics and Limbal Arc Rendering

## Connection to physical simulation

- Lagrangian mechanics $/$ Hamilton's least action principle $=$ finding shortest paths towards target
- a generalization of Fermat's principle
- specular light path rendering is a physical trajectory finding problem!



# Next: multiple importance sampling++ 

$$
\sum w_{i} \frac{f_{i}}{p_{i}}
$$

