

THE PREMIER CONFERENCE 8 EXHIBITION ON COMPUTER GRAPHICS & INTERACTIVE TECHNIQUES

#### FROM MICROFACETS TO PARTICIPATING MEDIA: A UNIFIED THEORY OF LIGHT TRANSPORT WITH STOCHASTIC GEOMETRY

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docs.unrealengine.com/5.0/en-US/nanite-virtualized-geometry-in-unreal-engine/

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disneyanimation.com/data-sets/?drawer=/resources/clouds/

UAV LiDAR Scanning System

Application in Historical Relics



The SGGX Microflake Distribution

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# Lay the *foundations* for a new *geometry* representation for *surfaces and volumes*.

#### GAUSSIAN PROCESS IMPLICIT SURFACES (GPIS)





## A SIDE NOTE ON GAUSSIAN PROCESSES (GPS)



#### A distribution over *functions*



- Completely described by **mean function** and **covariance kernel**
- Closed under linear transforms (including derivatives)
- No relation to **Gaussian Splatting™**



## Implicit Surface

## Stochastic Implicit Surface



#### **ENSEMBLE AVERAGE LIGHT TRANSPORT**



 $\mathbf{E}[\mathscr{L}]u$ 







Remember: Light Transport is not linear in geometry!

#### LIGHT TRANSPORT IN STOCHASTIC IMPLICIT SURFACES





## **COMPUTING RADIANCE**



$$L_i(x,\omega) = \int_{\Omega} L_i(x_s,\omega_s)\rho(x_s,\omega_s,n_s)d\omega_s$$

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 $\langle L_i(x,\omega) \rangle_{GP} = \int_{GP} \int_{\Omega} L_i(x_s^f,\omega_s) \rho(x_s^f,\omega_s,n_s^f) d\omega_s d\gamma(f)$ 

#### **COMPUTING ENSAMBLE RADIANCE**







$$\langle L_i(x,\omega) \rangle_{GP} = \int_{GP} \int_{\Omega} L_i(x_s^f, \omega_s) \rho(x_s^f, \omega_s, n_s^f) d\omega_s d\gamma(f)$$

#### **COMPUTING ENSAMBLE RADIANCE**





#### **COMPUTING ENSAMBLE RADIANCE**





#### **ENSAMBLE RADIANCE**







## Globally Consistent





## Globally Consistent Position Consistent +Normal Consistent



## **PROCESS MEMORY MODELS**







Position

## **Position+Normal**

## **PROCESS MEMORY MODELS**





## Global

## Position

## Position+Normal



0

à

4 ...

# APPEARANCE SPACE OF STOCHASTIC IMPLICITS





## 2D heightfields are the special case!







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## **CONNECTIONS TO MICROFACET THEORY**







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## **GENERATE SURFACES WITH GIVEN NDFS**





## OR LOOK AT MORE GENERAL GEOMETRY!





## VOLUME-TYPE GPIS APPROXIMATE CLASSICAL MEDIA





## **CONNECTIONS TO PARTICIPATING MEDIA**





## A NON-STATIONARY GPIS MODEL





## A NON-STATIONARY GPIS MODEL





Surface Volume "Moti

## "Motion Blur"

## MANUAL EDITING



## "posterior editing"



"prior editing"

## STOCHASTIC CONSTRUCTIVE SOLID GEOMETRY





## STOCHASTIC POISSON SURFACE RECONSTRUCTION



























# Original Downsampled



Ours

## LIMITATIONS











- Find closed-form approximations for transmittance.
- Differentiable rendering algorithm
- Wider range of stochastic processes
- Apply to Monte Carlo PDE solvers



# **THANK YOU!**







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