

Learning Generative Models for Rendering Specular Microgeometry

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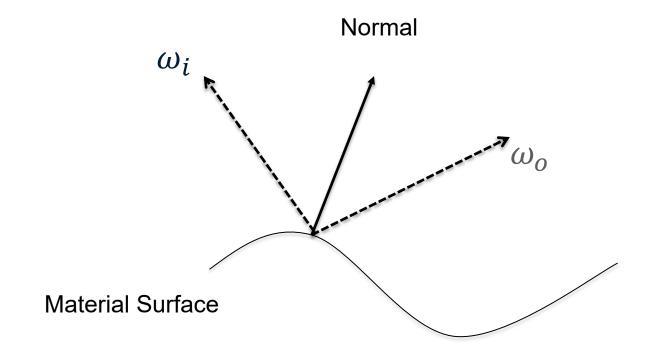
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sa2019.siggraph.org





Reflectance (BRDF)

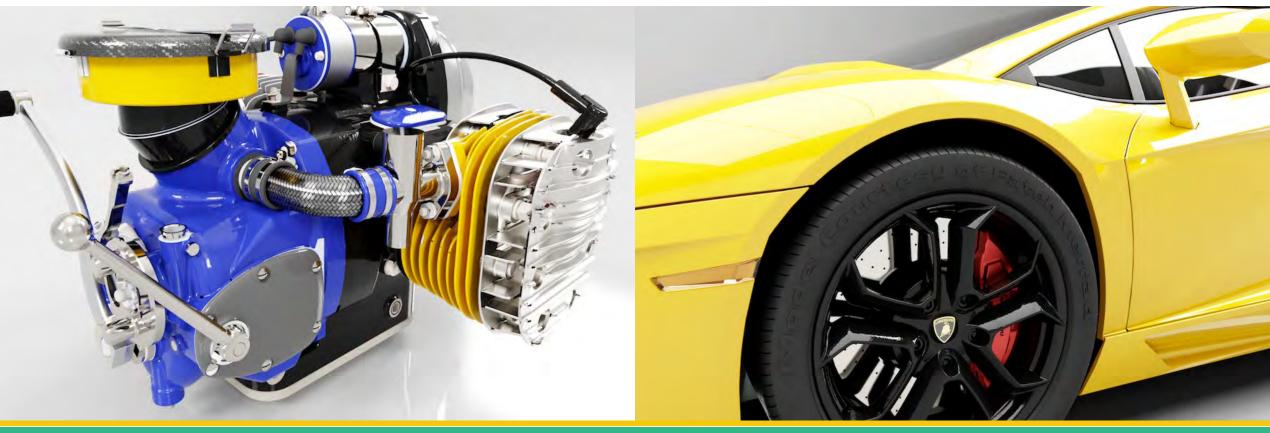






Microfacet BRDF

- SIGGRAPH Standard material model
- Microfacet BRDFs Fails to capture imperfections







Dual Brushed Metal

Sandblasted Finish



Specular Microgeometry Works



Rendering Glints on High-Resolution Normal-Mapped Specular Surfaces by Yan et al. 2014



Scratch Iridescence: Wave-Optical Rendering of Diffractive Surface Structure



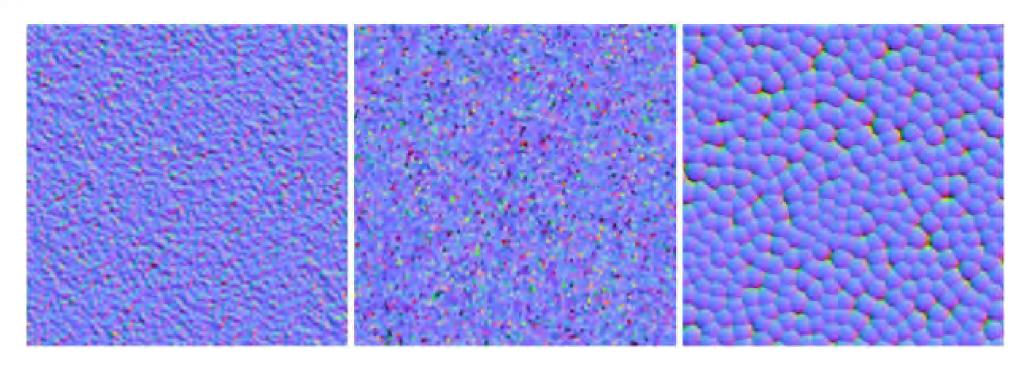


Rendering Specular
Microgeometry with Wave
Optics
by Yan et al. 2018



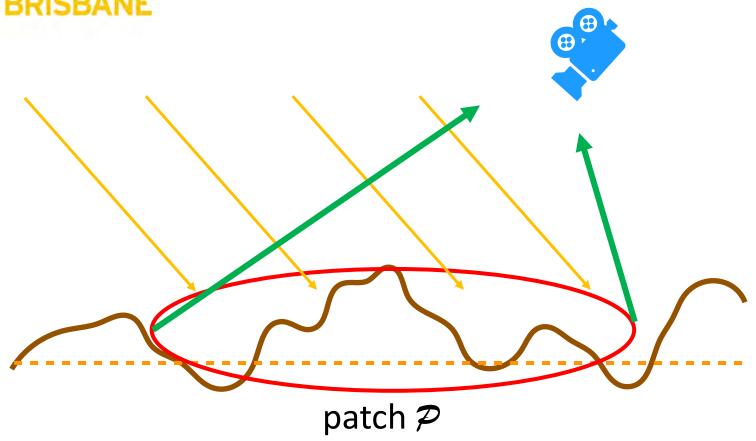


NDF for rendering



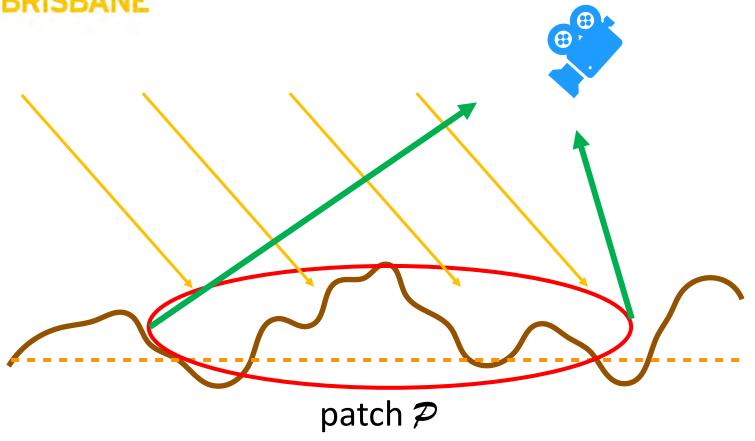


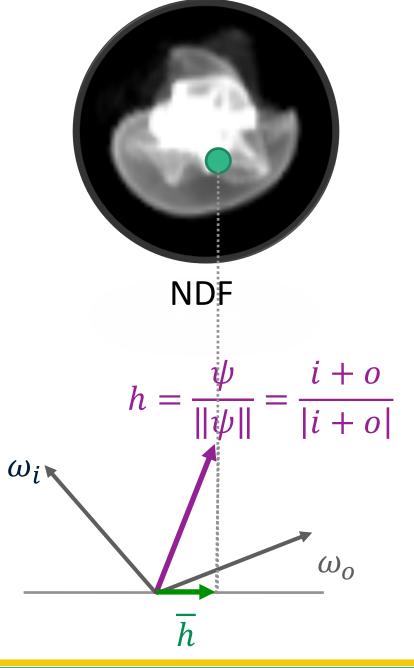
NDF for rendering





NDF for rendering







Limitations of Previous Works

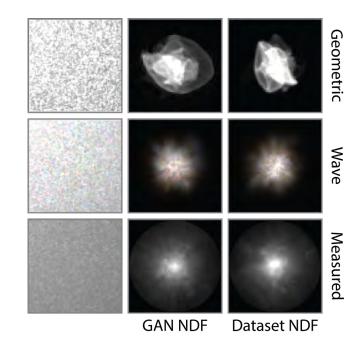
- Requires explicit heightfield
 - Explicit scratches
 - Hard to acquire
 - Memory Intensive
- Complexity of material
- Limited to simulations
 - No fabric
 - No captured materials





Main Contributions

- First to use GAN to render
 - Performance is independent of material complexity
 - Synthesize infinite surfaces
 - Measured data
 - Small storage requirement
- Unify different methods
- Partial Evaluation of CNN
 - 50x faster









Our Pipeline

SKISDANL

Data Acquisition

Simulation

- Geometric
- Wave
- FabricCapture



Training

Stochastic

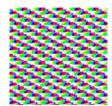
GAN

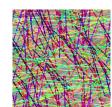
Spatial Correlation

CGAN

Feature Vector
Texture

Texture Synthesis

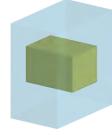




Rendering

Partial Evaluation of CNN

partial 4 x 4 x 64

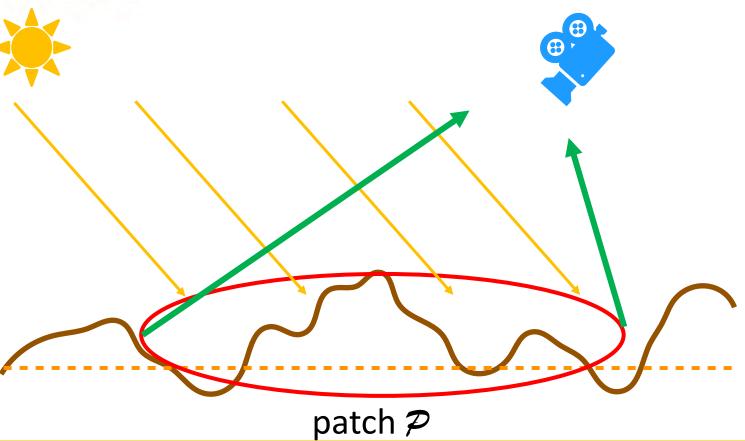


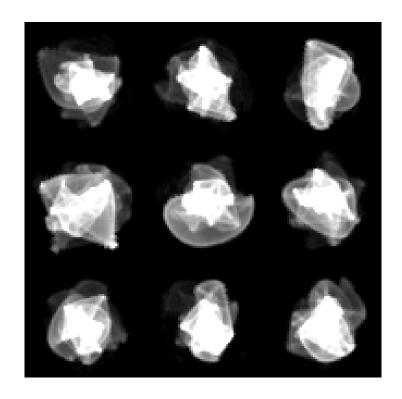




Geometric Optic NDFs

Contribution from patch P



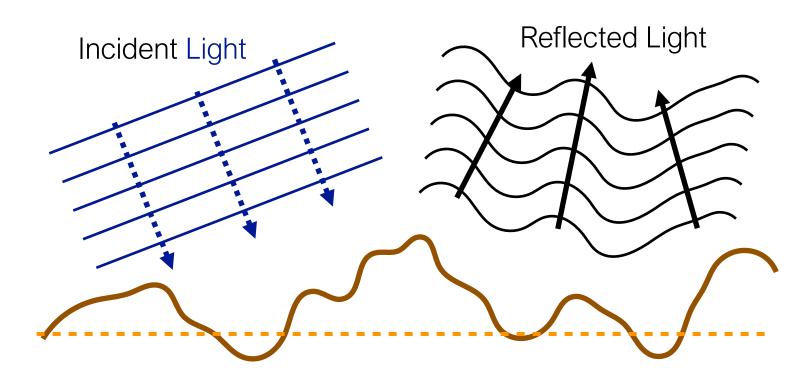


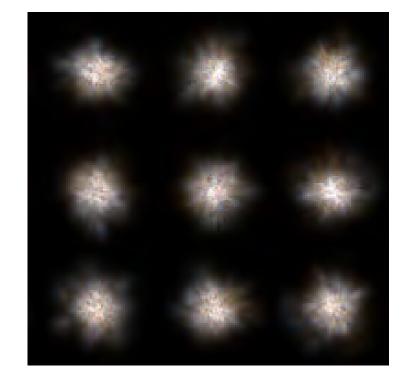




Wave Optic NDFs

Light is modeled as waves



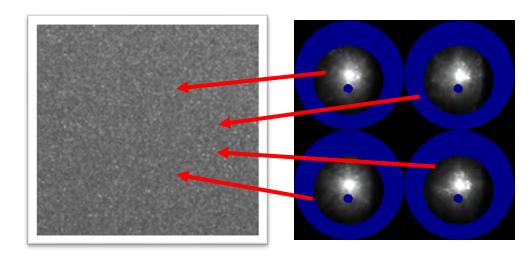






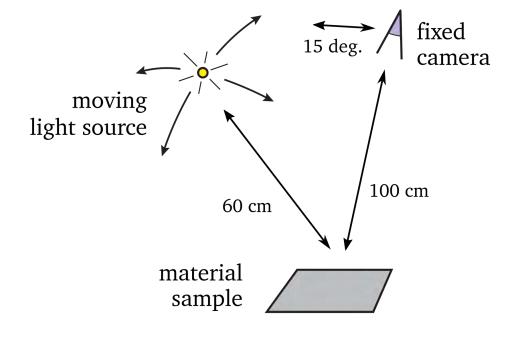
Measurement of NDFs

- Fixed camera
- RISBANE Moving Light Source
 - Blind Spots



Material Photo

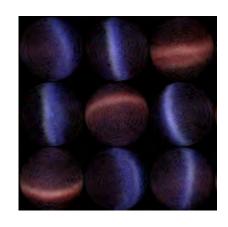
Captured NDF

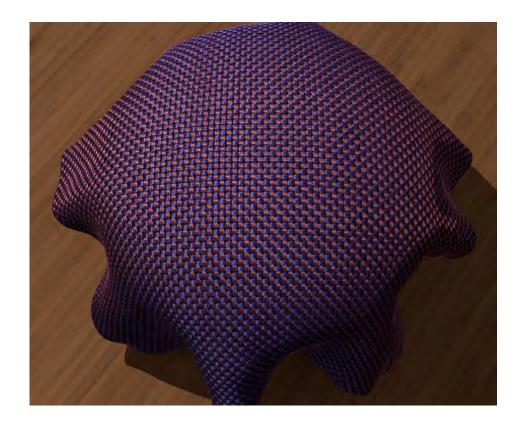




Fabric NDFs

- **BRISBANE** Render NDFs with hair BSDF
 - Use half-vector
 - Ignore difference vector

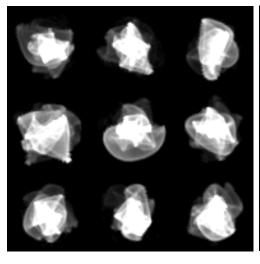


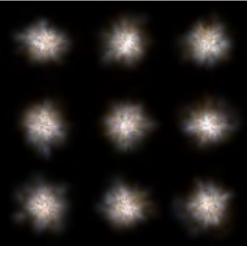


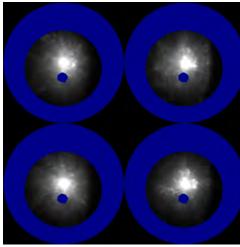




NDFs





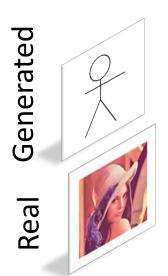






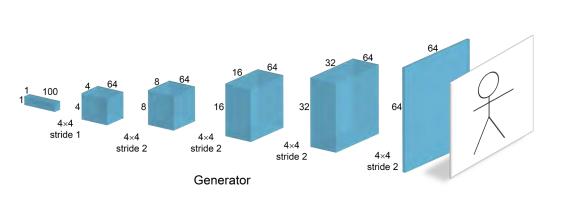
GAN Overview

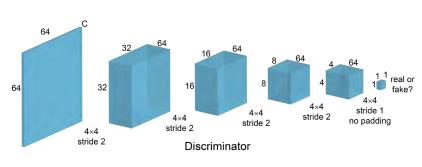
- Generator
 - Generates fake data
 - Noise vector as an input
- Discriminator
 - distinguish between the real data





Images generated by GAN

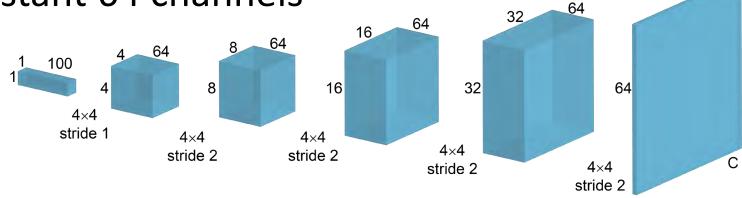


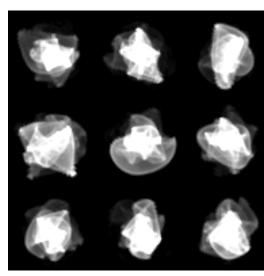


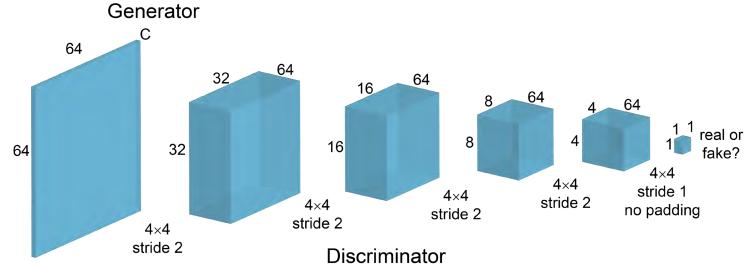




Constant 64 channels





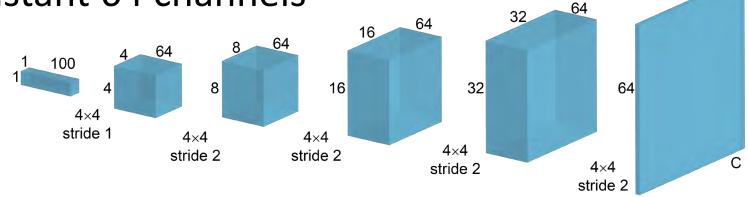


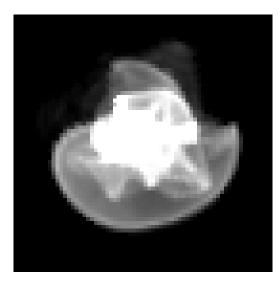
64





Constant 64 channels





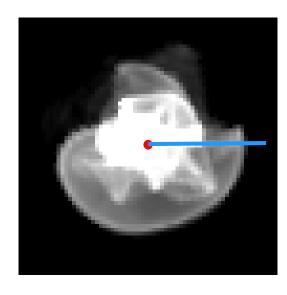


64

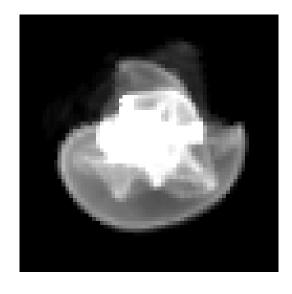


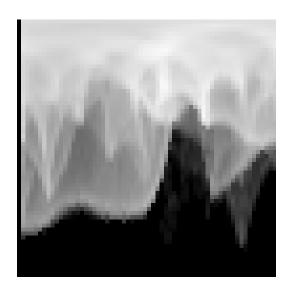


SIGGRAPH • Constant 64 channels



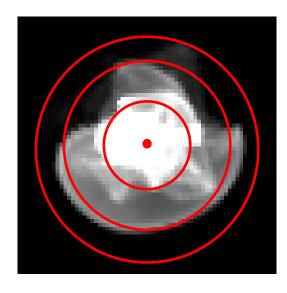


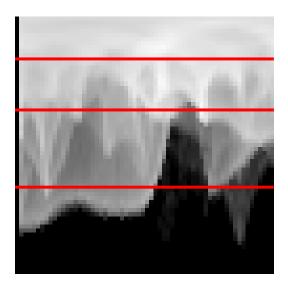










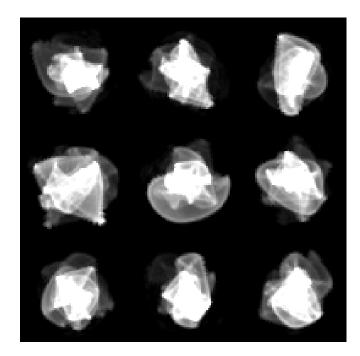


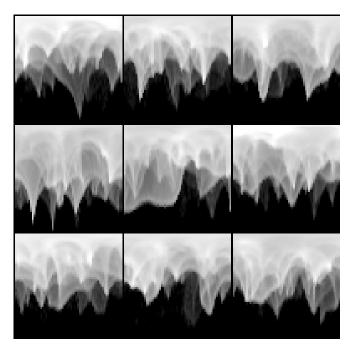




Polar Transformation

- **SIGGRAPH** Circular Structure
 - RISBANE Convolution reuses features
 - Horizontal wrapping

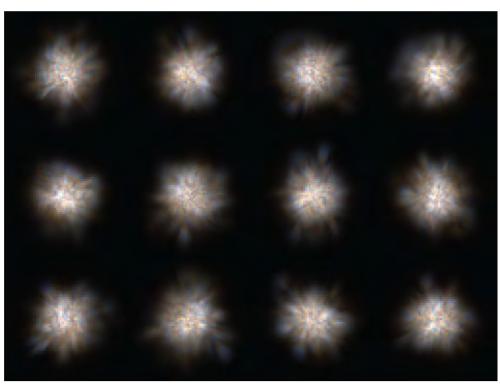








Comparison of NDFs Wave Optics



NDFs in training set

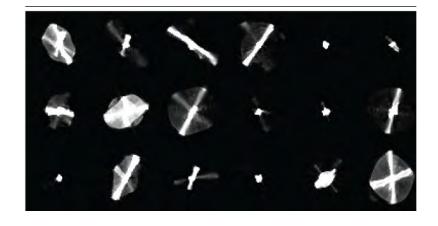
Our generated NDFs



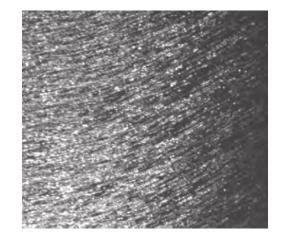


cGAN and Feature Vectors

- Large Scale Texture Variation
 - Scratched/brushed metal
 - Fabric
- Train autoencoder
 - Low dimensional feature vector
 - Blurry lack of details
- Use this feature vector as condition for training cGAN



NDFs of scratched metal



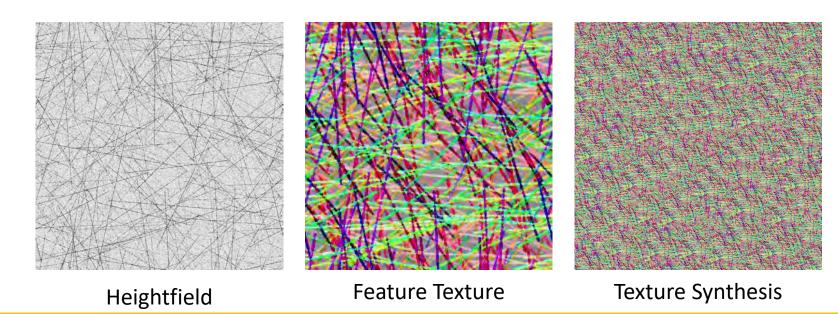
Scratched metal





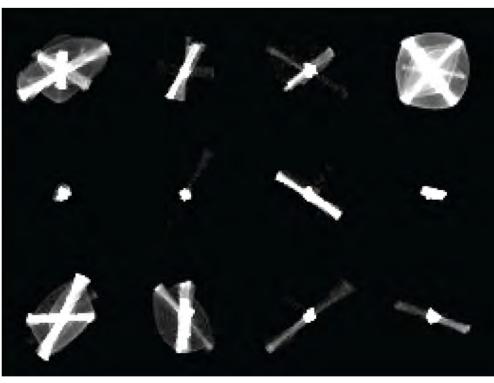
cGAN and Feature Vectors

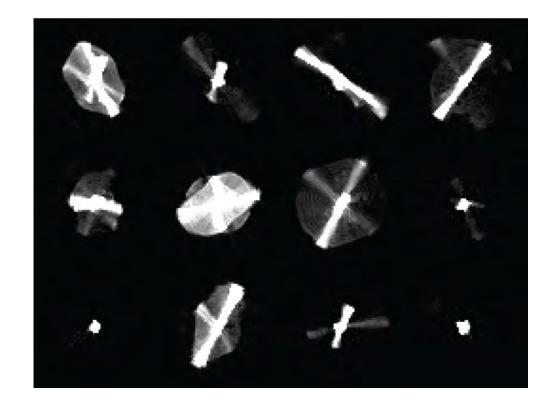
- Convert real NDFs into 2D grid of feature vectors
 - 256x256
- Use corresponding feature vector when rendering
- Use texture synthesis to increase size





Comparison of NDFs Scratched metal





NDFs in training set

Our generated NDFs



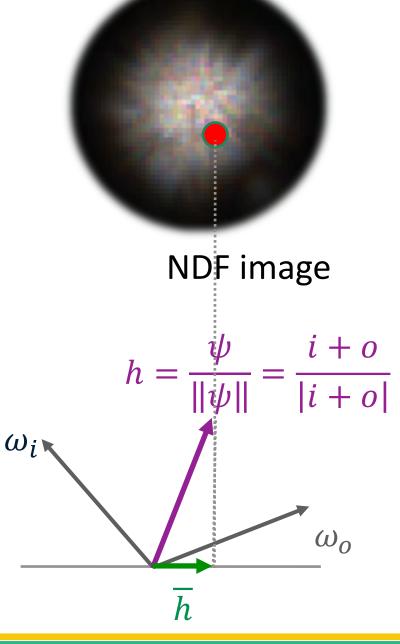






Rendering

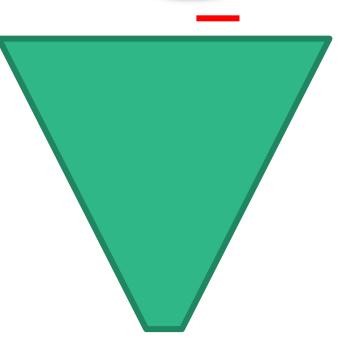
- Each intersection
- Calculate half-vector h
 - Light source and camera directions
- Project h to unit circle
- Look up value from the image
 - 2x2 image region if bilinear interpolation





Partial Evaluation in Details

- Each intersection
- Calculate half-vector h
 - Light source and camera directions
- Project h to unit circle
- Look up value from the image
 - 2x2 image region if bilinear interpolation



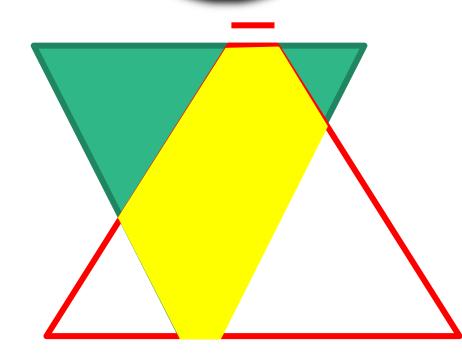
Input vector





Partial Evaluation in Details

- Calculate ranges bottom→top
- Calculate ranges top→bottom
- Find the intersection
- Evaluate only valid ranges



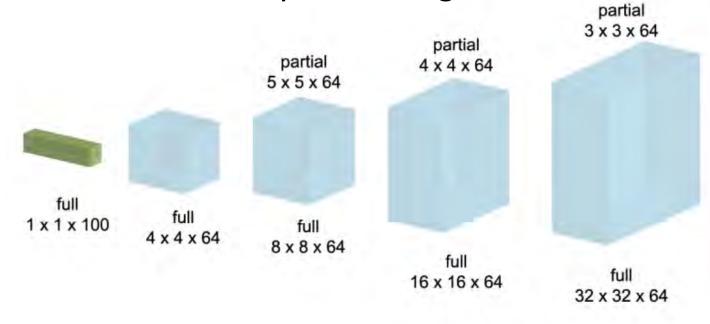
Input vector



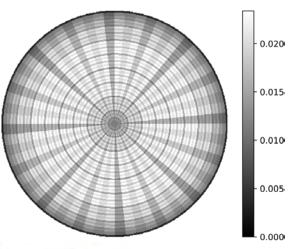


Partial Evaluation of CNN

- Evaluate only what you need
 - 1.75% computation vs full eval
 - Compute valid ranges for all hidden layers
 - Evaluate only those ranges



Fraction of Computation for Each Pixel









Our Pipeline

Training

Data Acquisition

Simulation

- Geometric
- Wave
- FabricCapture



Stochastic

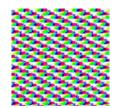
GAN

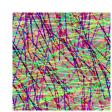
Spatial Correlation

Feature Vector
Texture

CGAN

Texture Synthesis





Rendering

Partial Evaluation of CNN

4 × 4 × 64

partial























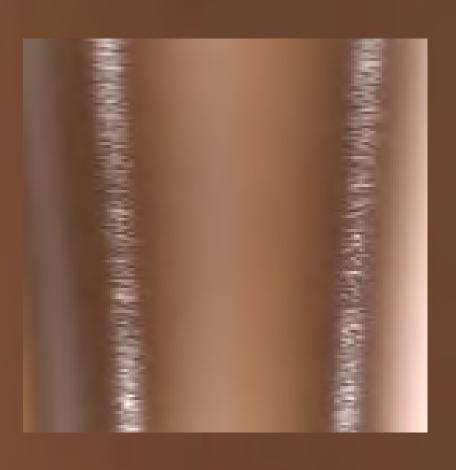


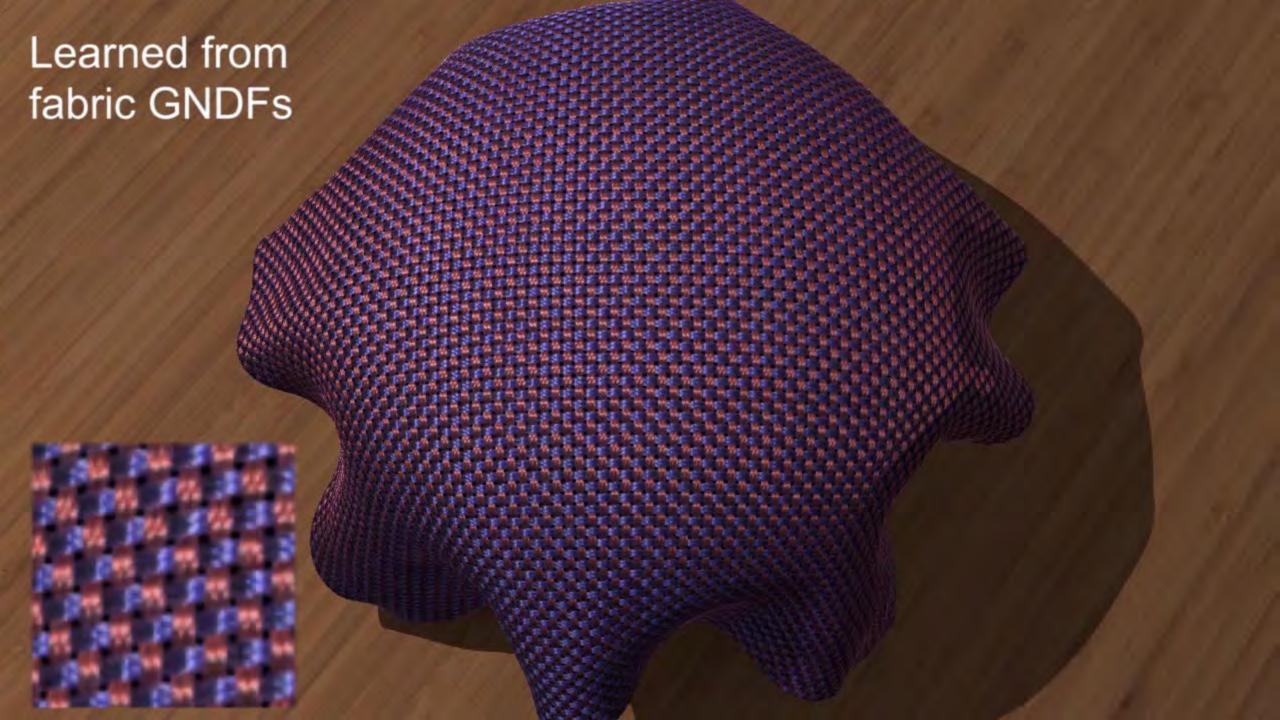




Learned from geometric optics

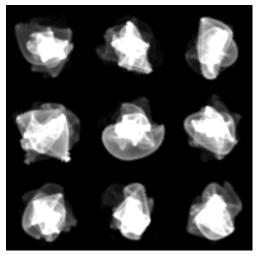


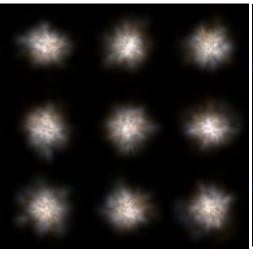


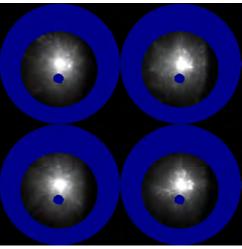




NDFs



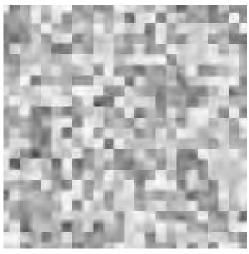




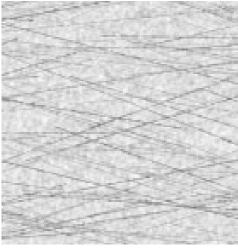


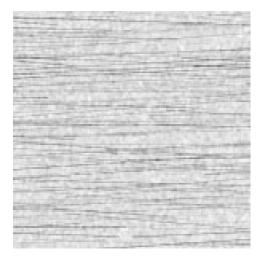


Additional high fields







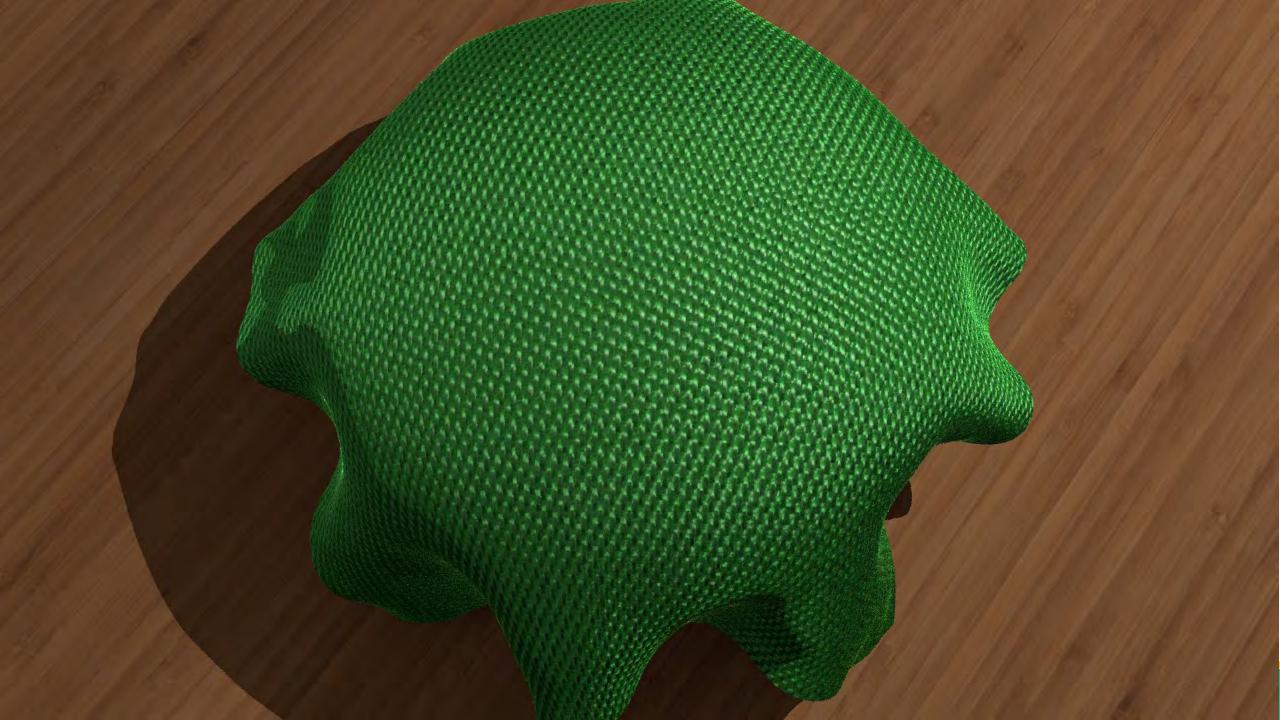
















Render Time (direct only)

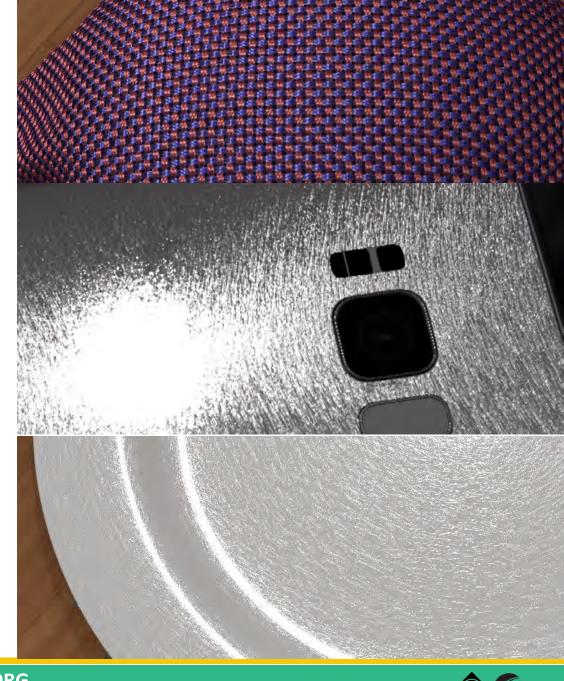
Scene	Type	Our Time	Prev. Time
Macbook	Geom.	4.5s	2.0s
Macbook	Wave	5.9s	234s
Plate	Geom.	12.3s	45.8s
Phone	Geom.	6.4s	3.3s
Cloth	Fabric	10.1s	n/a
	1,2		5





Conclusion

- Render specular microgeometry
 - GAN
 - Measured materials
 - Fabrics
 - No explicit heightfield
- Partial Evaluation of CNN
 - 50x faster







Thank you!

Acknowledgments

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Phone

point light + environment lighting

geometric optics, wave optics

moving light