



THE PREMIER CONFERENCE & EXHIBITION ON COMPUTER GRAPHICS & INTERACTIVE TECHNIQUES

ULTRA-HIGH RESOLUTION SVBRDF RECOVERY FROM A SINGLE IMAGE

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Specialized hardware

- High quality
- Dense sampling of light/view angle pairs
- Requiring a prohibitive amount of time



Dana and Wan Ags 20014 et al. 2020 C. Schwartz et al. Universitet al. 2023 Holroyd et al. 2010 of Bonn's Dome II

Light-weight setups

- A great trend in SVBRDF acquisition
- Fast, convenient, low-cost
- Generally only using a cell phone with flash
- Relying on deep neural networks (in particular CNNs)



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 Potential limitations of CNN-based SVBRDF acquisition methods when facing Ultra-High Resolution inputs:

- Memory constraints of GPUs
- Difficulties in capturing global features









HR Image cropping •

- Material Exemplar prediction ٠
- Exemplar-specific fine-tuning •
- Post-processing: image stitching •

inputs (> 200 seconds per-image)

Runs extremely slow even with 1K



Material Exemplar









- Single, casually captured image as input (without camera calibration)
- Supporting Ultra-high resolution inputs (2K or 4K)
- Light-weight setup
- Low computational cost
- Fast evaluation with end-to-end prediction









A divide-and-conquer solution







LOCAL FEATURE EXTRACTOR

Providing a "local view" of the underlying material



stacked convolutional blocks



Conv(k4s2)+LeakyReLU

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Simply stitching local feature maps generated by LFE will cause noticeable inconsistency among different patches.



GLOBAL FEATURE EXTRACTOR





Material Vision Transformers



MATERIAL VISION TRANSFORMER



Each MVT has a **global receptive field** that is beneficial for preserving globally coherent material properties





→ GLOBAL FEATURE EXTRACTOR





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COORDINATE-AWARE FEATURE ASSEMBLY MODULE





COORDINATE-AWARE FEATURE ASSEMBLY MODULE





(a) Simple stitching (b) Ours without overlaps (c) Ours with overlaps

Removing seams in a recovered diffuse map by introducing some overlaps between adjacent patches.



COMPLETE NETWORK ARCHITECTURE





DATASET AND LOSS FUNCTION

• UHR dataset

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Part2

	4K		2K		1K	
	Part 1	Part 2	Part 1	Part 2	Part 1	Part 2
Training	-	160	2,105	-	6,315	2,400
Testing	-	25	24	26	-	-

- Part1 Collected by Deschaintre et al. [2020]
 - https://ambientcg.com/
 - <u>https://polyhaven.com/</u>
 - https://www.sharetextures.com
 - https://www.cgbookcase.com/textures/

Loss Function

 $\mathcal{L} = \mathcal{L}_{map} + \mathcal{L}_{render} + \lambda_{percept} \mathcal{L}_{percept}$

$$\mathcal{L}_{all} = \lambda_1 \mathcal{L}_{global} + \lambda_2 \mathcal{L}_{patch}$$











We compare our method to

- Des18: a classic U-Net based end-to-end method that can recover
 SVBRDFs from a single image, proposed by Deschaintre et al. [2018],
- **Gao19**: a deep inverse rendering method that can recover high-resolution SVBRDFs from multiple images, proposedby Gao et al. [2019],
- HANet: a state-of-the-art method for singe-image (usually low-resolution)
 SVBRDF recovery, proposed by Guo et al.[2021],
- Guided: a recent method that can support 4K SVBRDF recovery by transferring low-resolution exemplar SVBRDFs to a target UHR image, proposed by Deschaintre et al. [2020].









GPU memory consumption

Runtime performance



COMPARISONS ON SYNTHETIC DATA 2K



























Input

SVBRDF maps

Rendering







Guided









Guided







The choice of exemplars to Guided

For **Guided**, we show the prediction results from two input images of the same material taken under environment lighting (Env.) and a flash point light (Point), respectively.







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- A consistent and implicit neural representation for SVBRDF recovery that can preserve both global structures.
- Material vision Transformer, a convolution-augmented vision transformer, to extract rich global features from the UHR input, providing the "global environment" of the material.
- A coordinate-aware feature assembly module to assemble "local views" of the underlying material in the feature space, guaranteeing spatial coherency.
- Our method is able to recover material maps as large as 4K, which is a challenge for previous learning-based methods.

• Increasing the normal map diversity in the dataset would further improve the quality of reconstructed normal maps.

- Incorporating some special designs, e.g. highlight-aware convolution.
- Extending our method to curved surfaces.

Input

SVBRDF maps











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THANK YOU

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