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SIGGRAPH 2018

EFFICIENT REFLECTANCE CAPTURE USING AN AUTOENCODER

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Sinovation Ventures



Introduction

INTRODUCTION

- Realistic Material Appearance is **Important**

INTRODUCTION

- Realistic Material Appearance is **Important**



Visual Effects

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e-Commerce

<https://www.coinsshopy.com/product/classic-brand-shoes-women-casual-pointed-toe-black-oxford-shoes-for-women-flats-comfortable-slip-on-women-shoes/>



Cultural Heritage

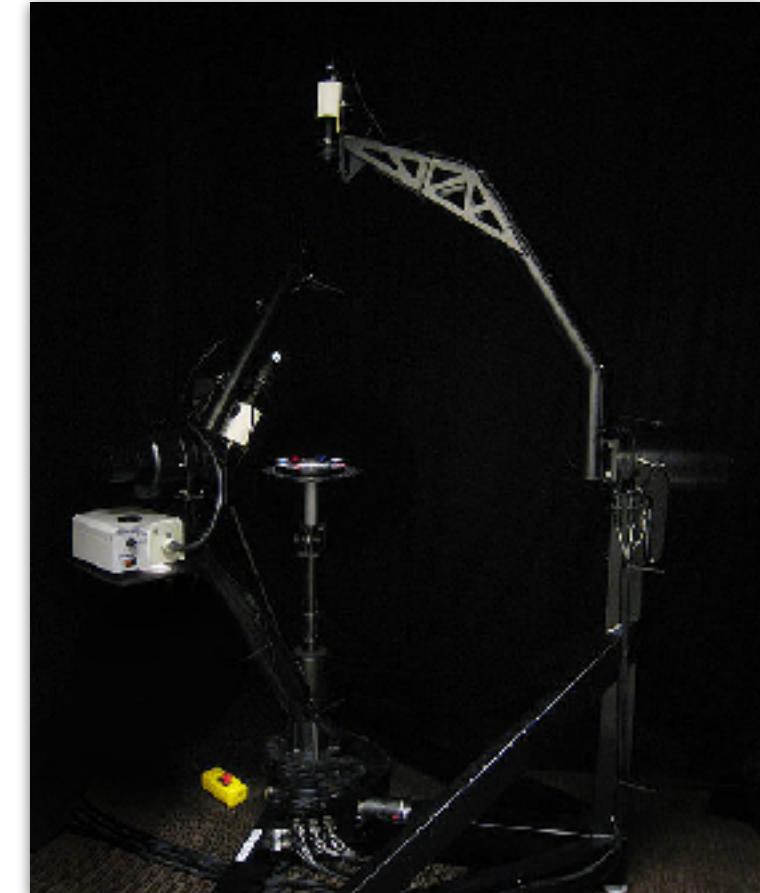
<https://pro.europeana.eu/post/inspire-makers-to-creatively-transform-europe-s-digital-cultural-heritage>

INTRODUCTION

- Realistic Material Appearance is **Important**
 - **6D** SVBRDF (Location, Lighting & View Directions)

INTRODUCTION

- Realistic Material Appearance is Important
 - 6D SVBRDF (Location, Lighting & View Directions)
- Reflectance Acquisition is Challenging
 - Directly Sampling the 6D Domain is **Inefficient**



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LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

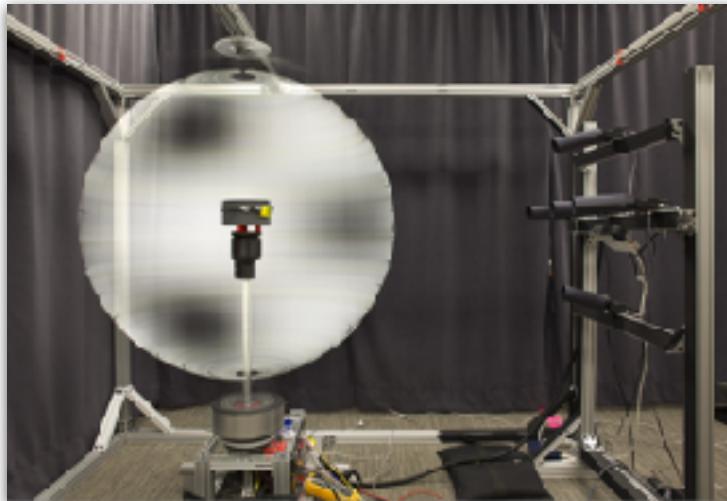
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- One Successful Class for High-Quality Capture
- Program Many Lights Simultaneously
 - Forming Different Lighting Patterns

LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

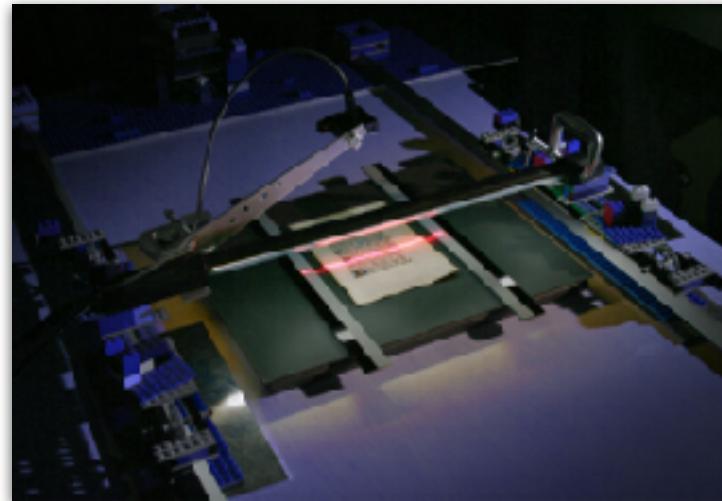
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- One Successful Class for High-Quality Capture
- Program Many Lights Simultaneously
 - Forming Different Lighting Patterns



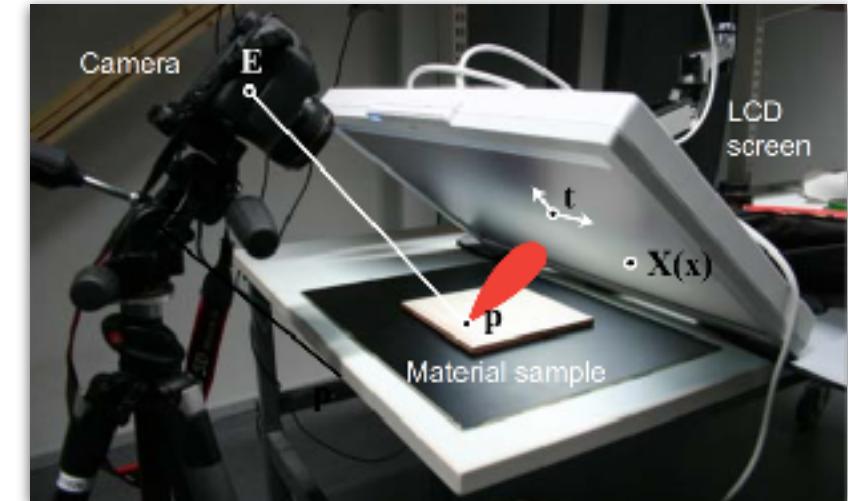
Lightstage

[Ghosh et al. 2009; Tunwattanapong et al. 2013]



Linear Light Source

[Gardner et al. 2003; Chen et al. 2014]



LCD-based Setup

[Aittala et al. 2013]

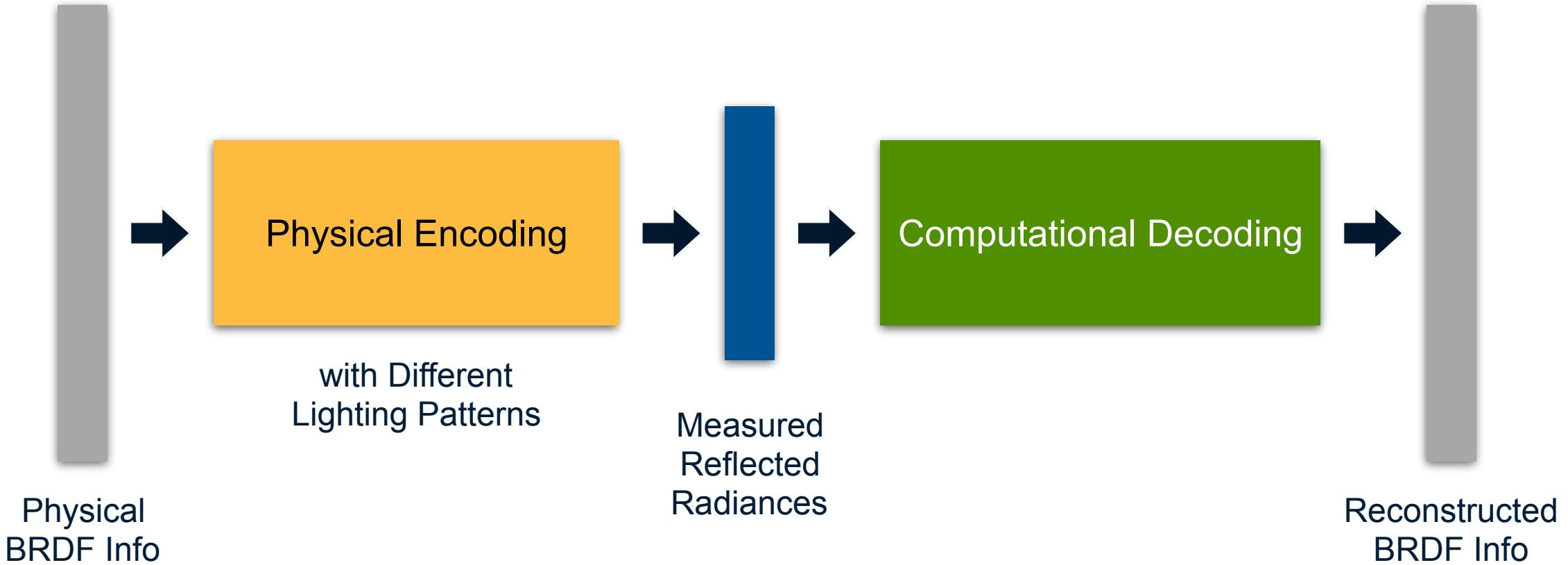
LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

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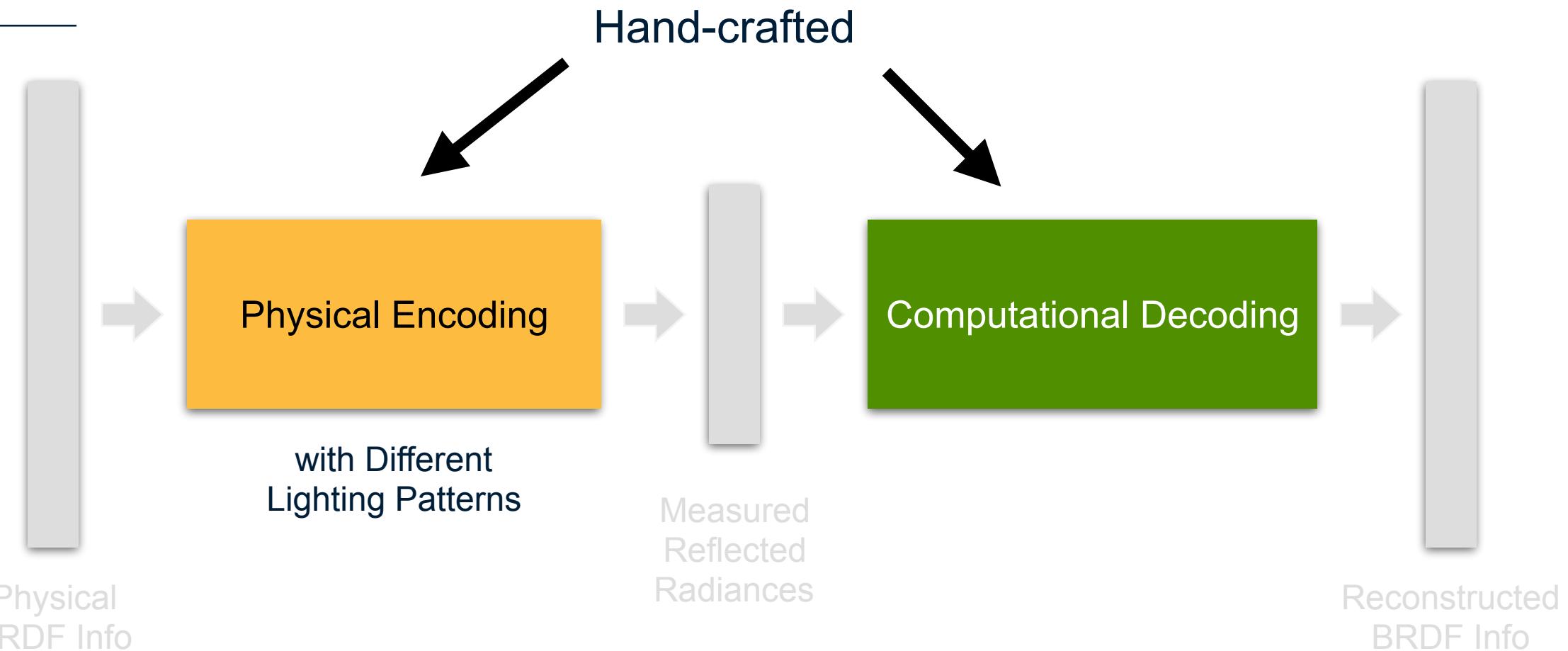
- One Successful Class for High-Quality Capture
- Program Many Lights Simultaneously
 - Forming Different Lighting Patterns
- 240 Lighting Patterns / Photos Needed [Chen et al. 2014]
- Our Goal
 - **To Improve Efficiency for SVBRDF Acquisition**

REFLECTANCE ACQUISITION = ENCODING + DECODING

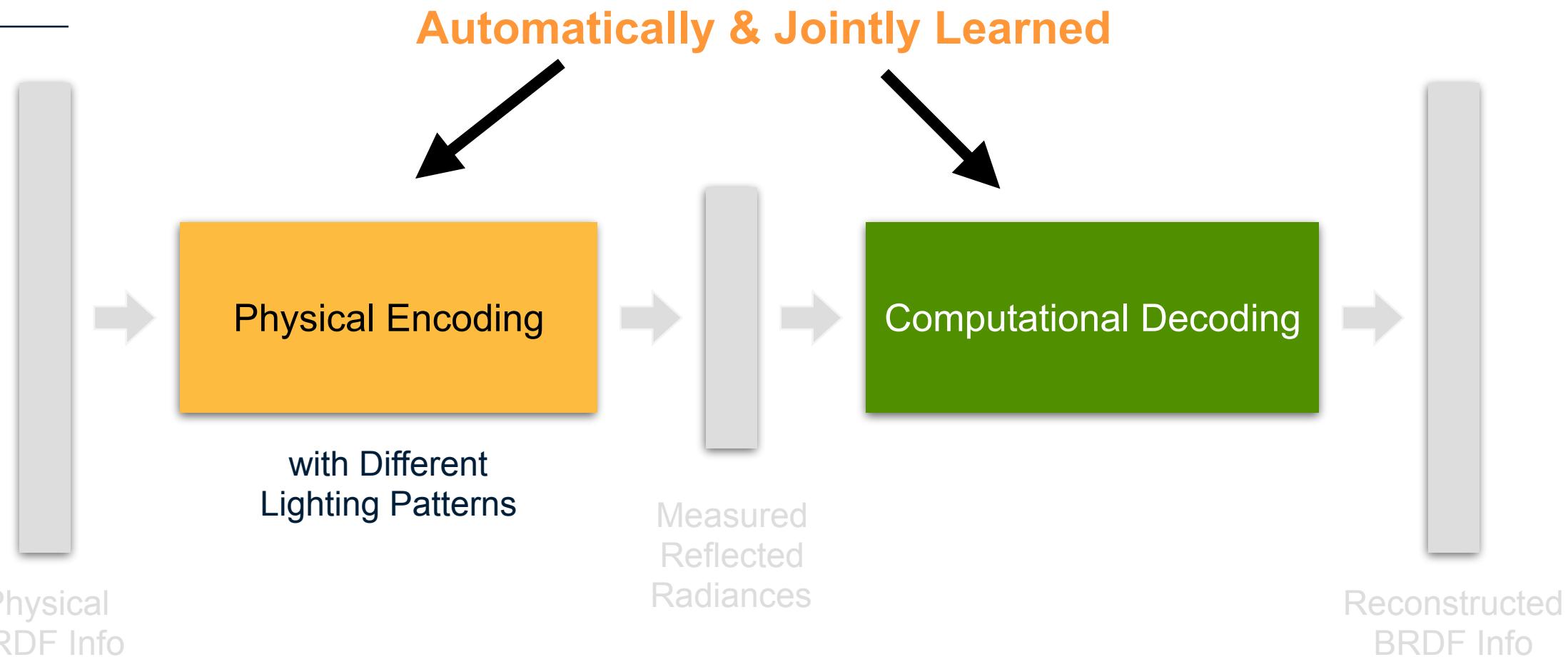
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PREVIOUS WORK



OUR APPROACH



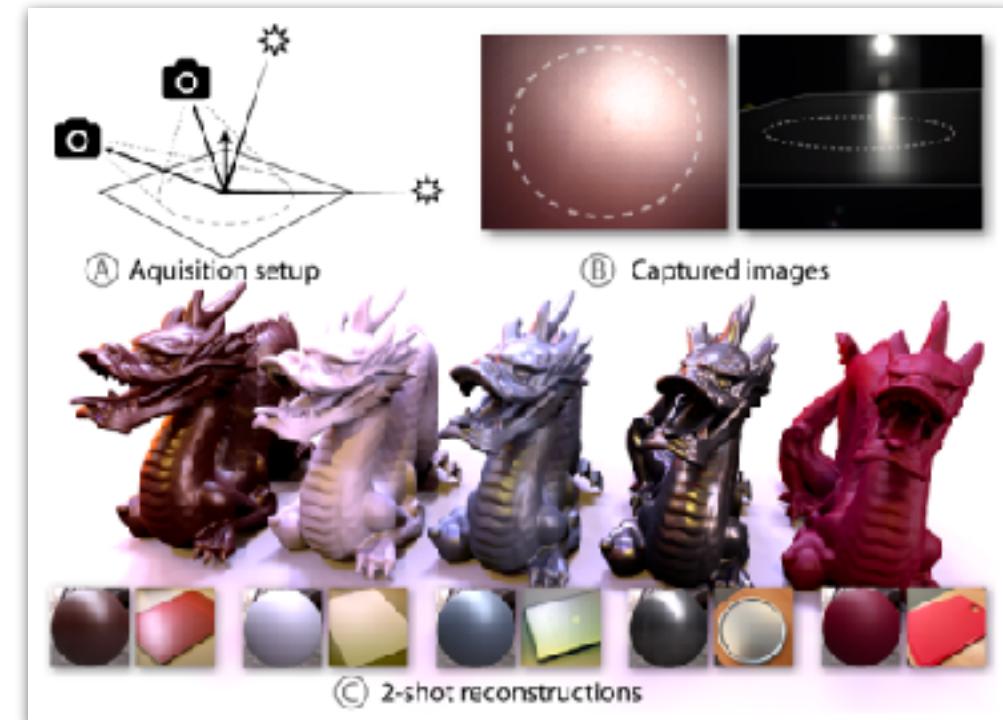
OUR APPROACH

- Lighting Pattern #: 16~32
- Acquisition Time: 12~25s
- Faithfully Recovers **General** Reflectance under **Near-field Lighting**
- Automatically Adapt to Various Factors
 - Setup's Geometry
 - Properties of Appearance

Related Work

RELATED WORK

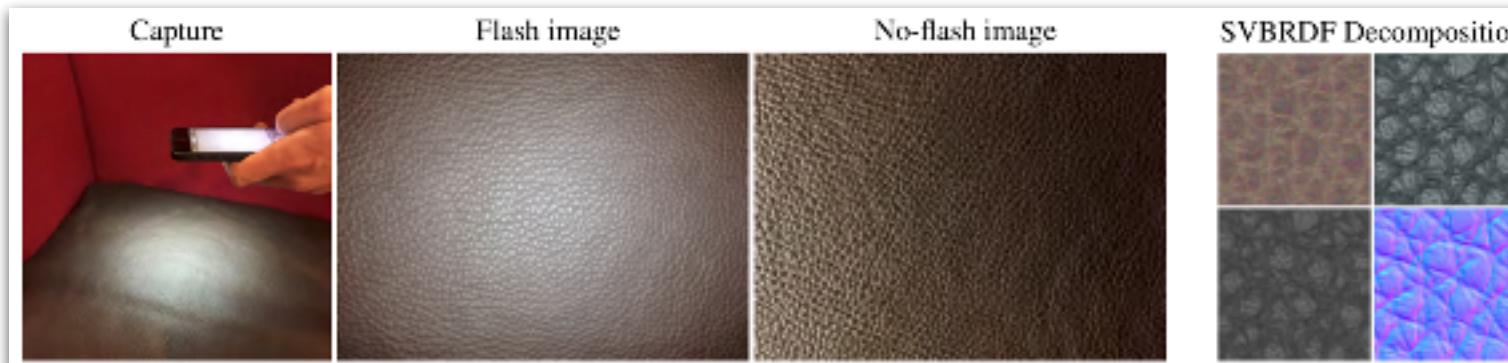
- Optimal BRDF Sampling
 - [Matusik et al. 2003; Nielsen et al. 2015; Xu et al. 2016]
 - Distant Lighting
 - Isotropic BRDF
 - Limited Support for Spatial / Normal Variations



[Xu et al. 2016]

RELATED WORK

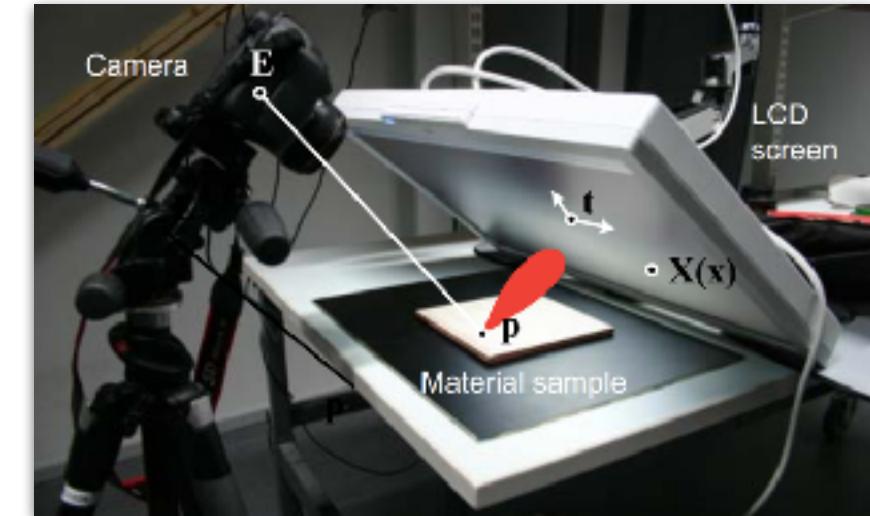
- Optimal BRDF Sampling
- Direct Sampling
 - Gonioreflectometer [Dana et al. 1999; Lawrence et al. 2006]
 - Prior over Reflectance [Lensch et al. 2003; Aittala et al. 2015]



[Aittala et al. 2015]

RELATED WORK

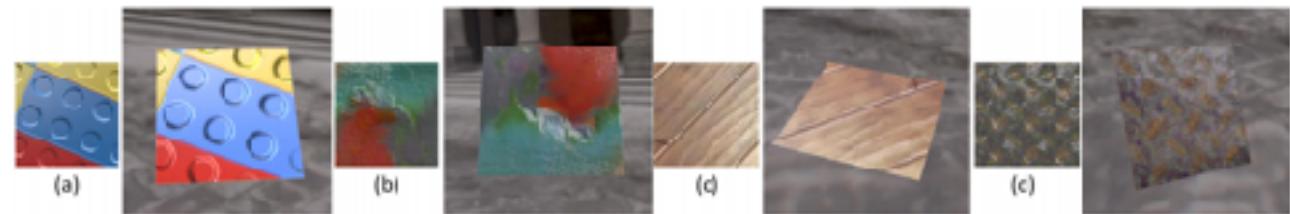
- Optimal BRDF Sampling
- Direct Sampling
- Complex Lighting Patterns
 - [Gardener et al. 2003; Ghosh et al. 2009; Aittala et al. 2013]
 - **Manually** Derive Lighting Patterns & Reconstruction Algorithms



[Aittala et al. 2013]

RELATED WORK

- Optimal BRDF Sampling
- Direct Sampling
- Complex Lighting Patterns
- Deep-Learning-Assisted Reflectance Modeling
 - [Aittala et al. 2016; Li et al. 2017]
 - Single Input Image
 - Less controlled Lighting
 - Assumptions over Reflectance



[Li et al. 2017]

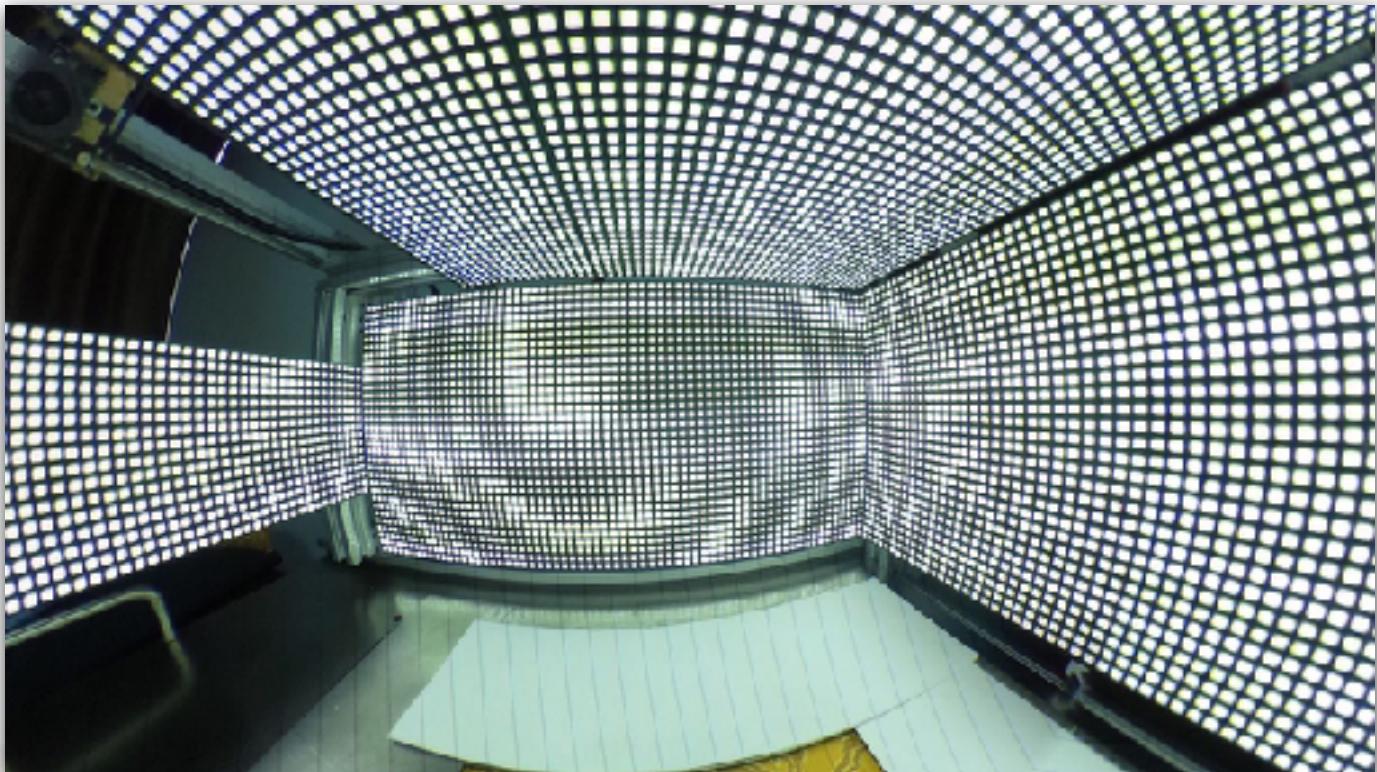
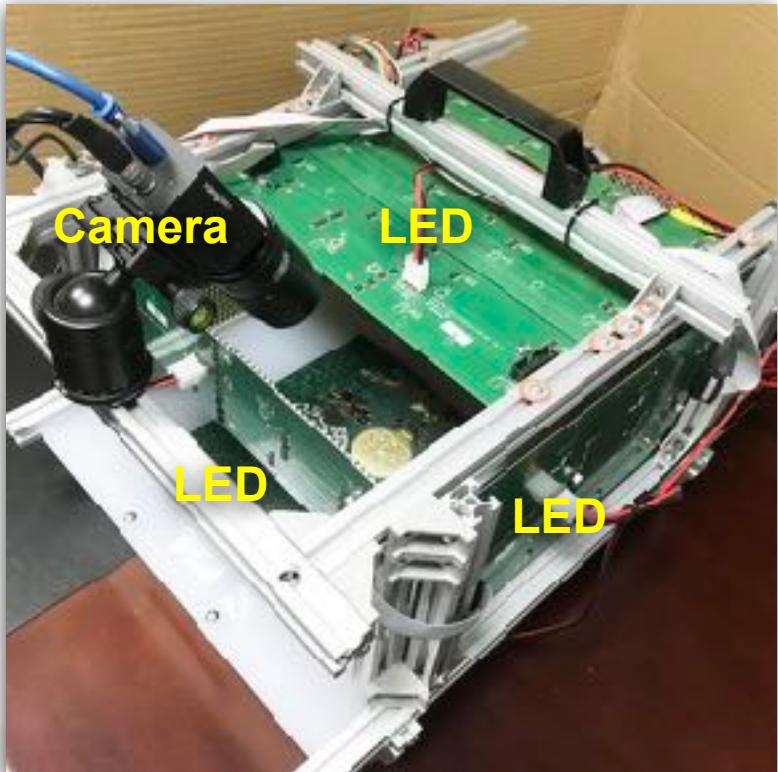
Our Framework

ASSUMPTIONS

- A Mostly Planar Sample
- Independent Reflectance Reconstruction on Each Location
- For a Single Channel

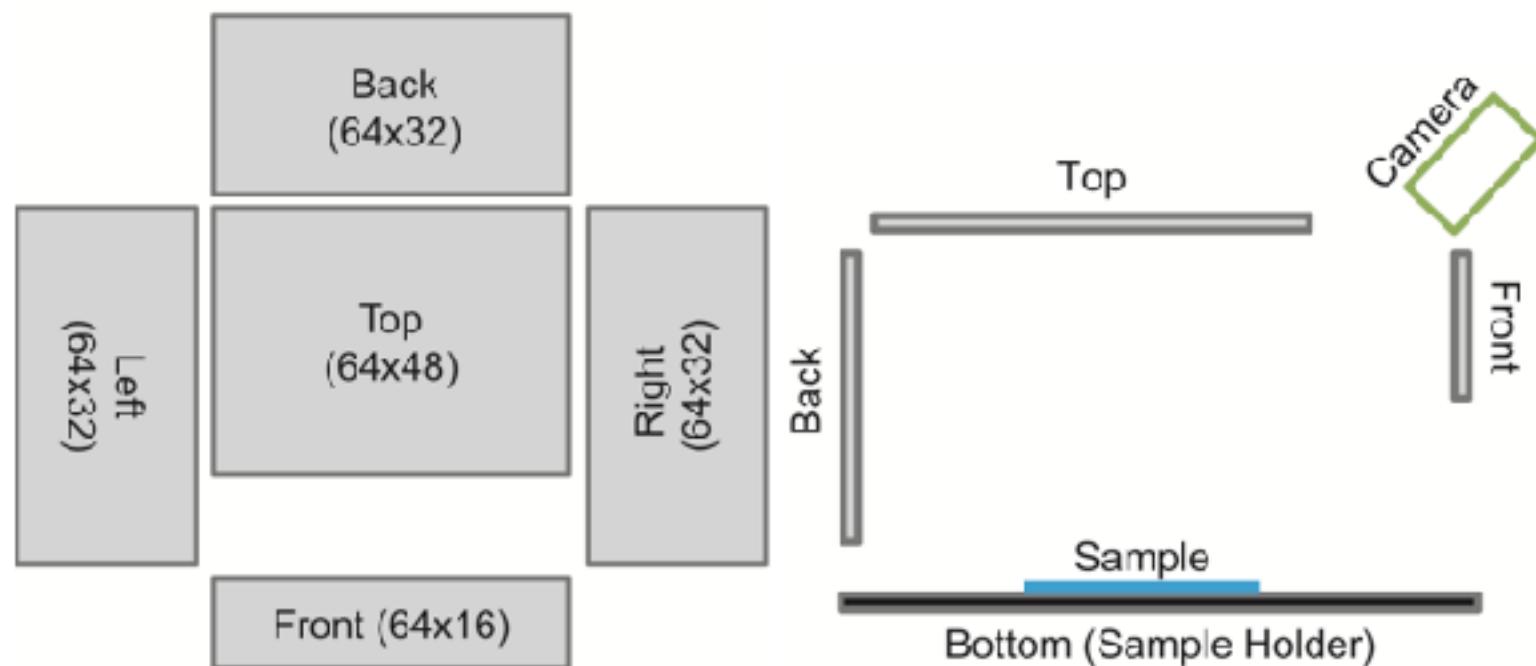
OUR DEVICE

- Mini, Near-field Lightstage

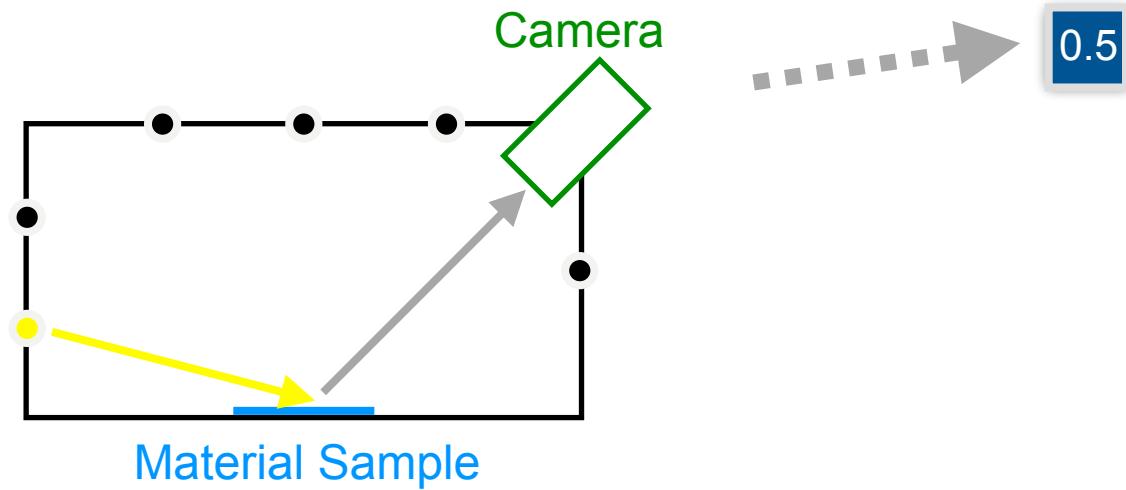


OUR DEVICE

- Mini, Near-field Lightstage
 - 420mm*360mm*210mm
 - Sample Size: 120mm*120mm
 - Single Camera
 - 10,240 LEDs
 - High-Precision Control via FPGA

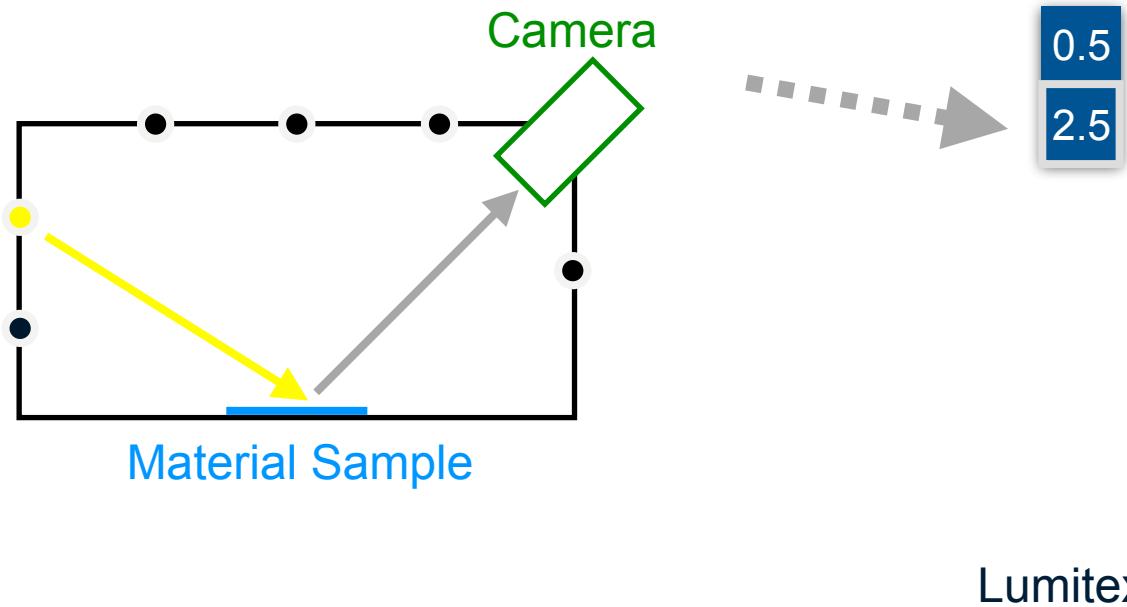


LUMITEXEL

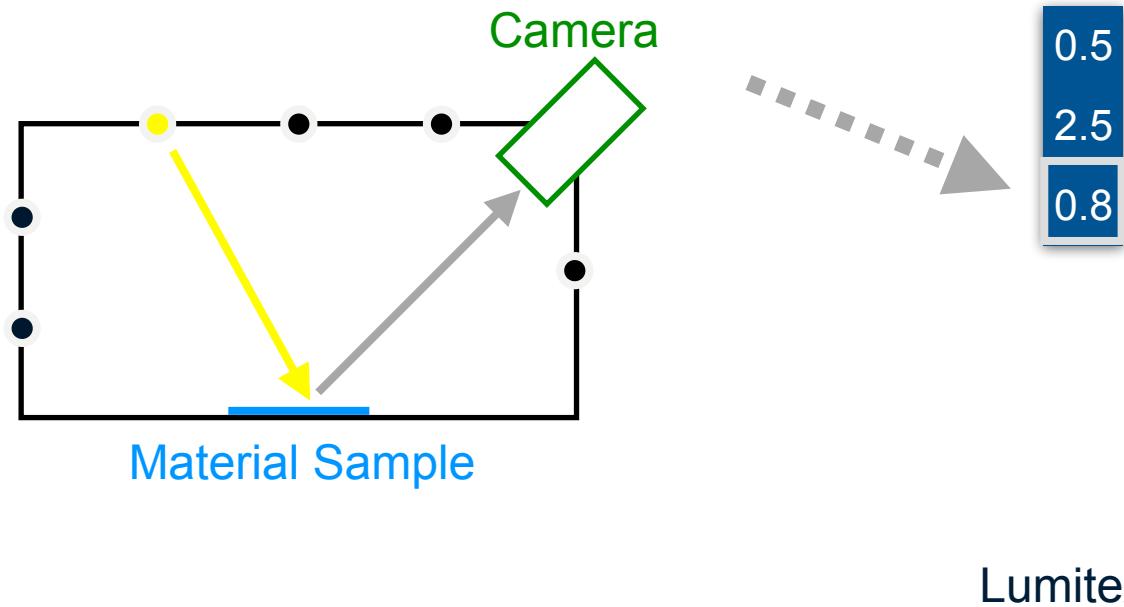


Lumitexel

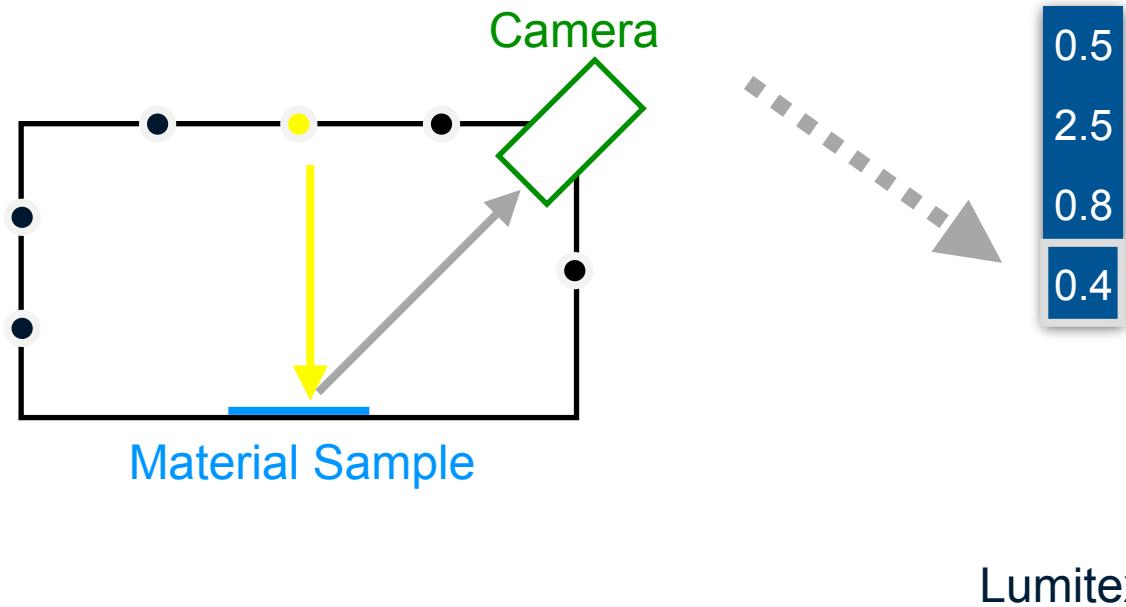
LUMITEXEL



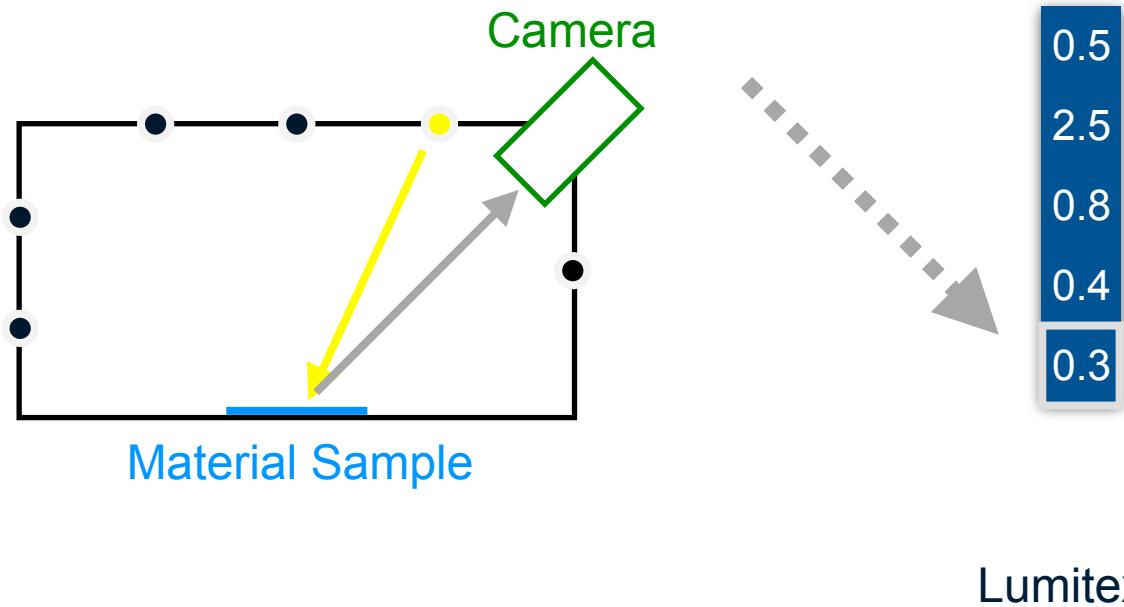
LUMITEXEL



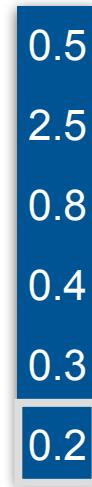
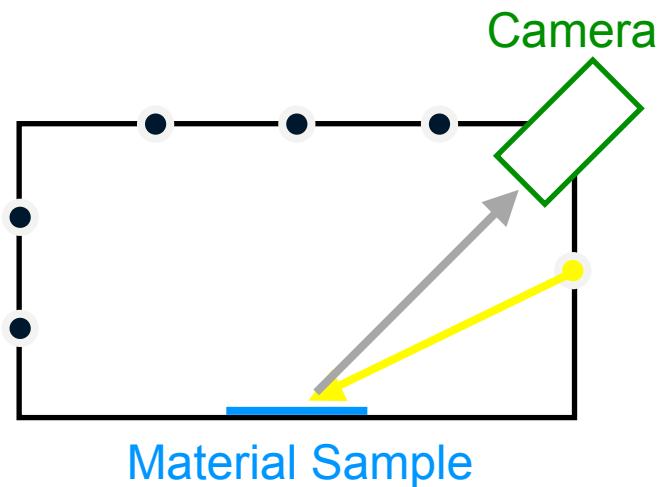
LUMITEXEL



LUMITEXEL



LUMITEXEL



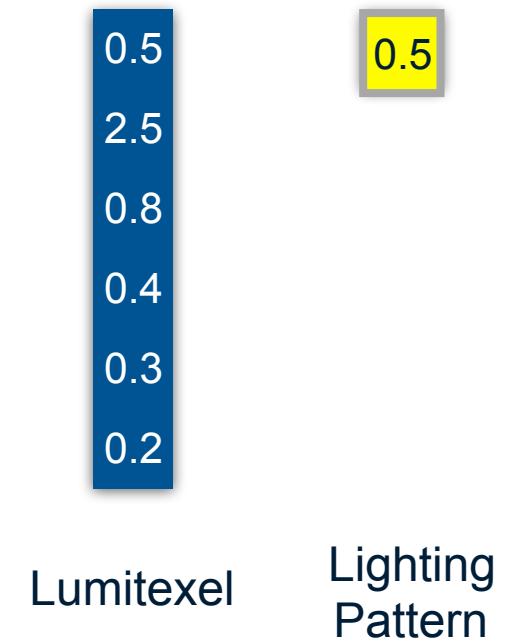
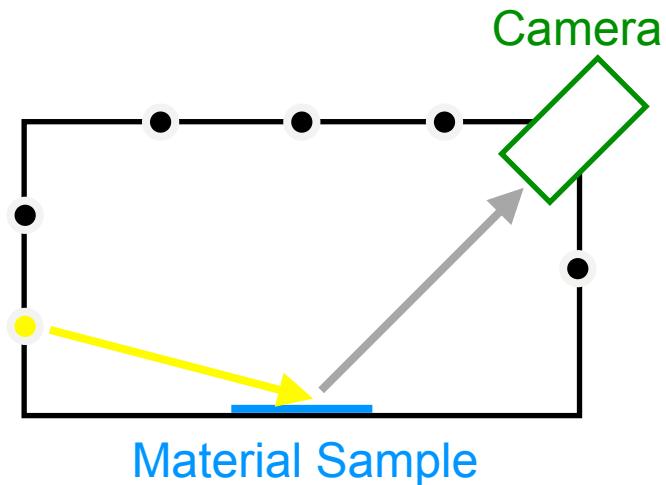
Lumitexel

Most Informative

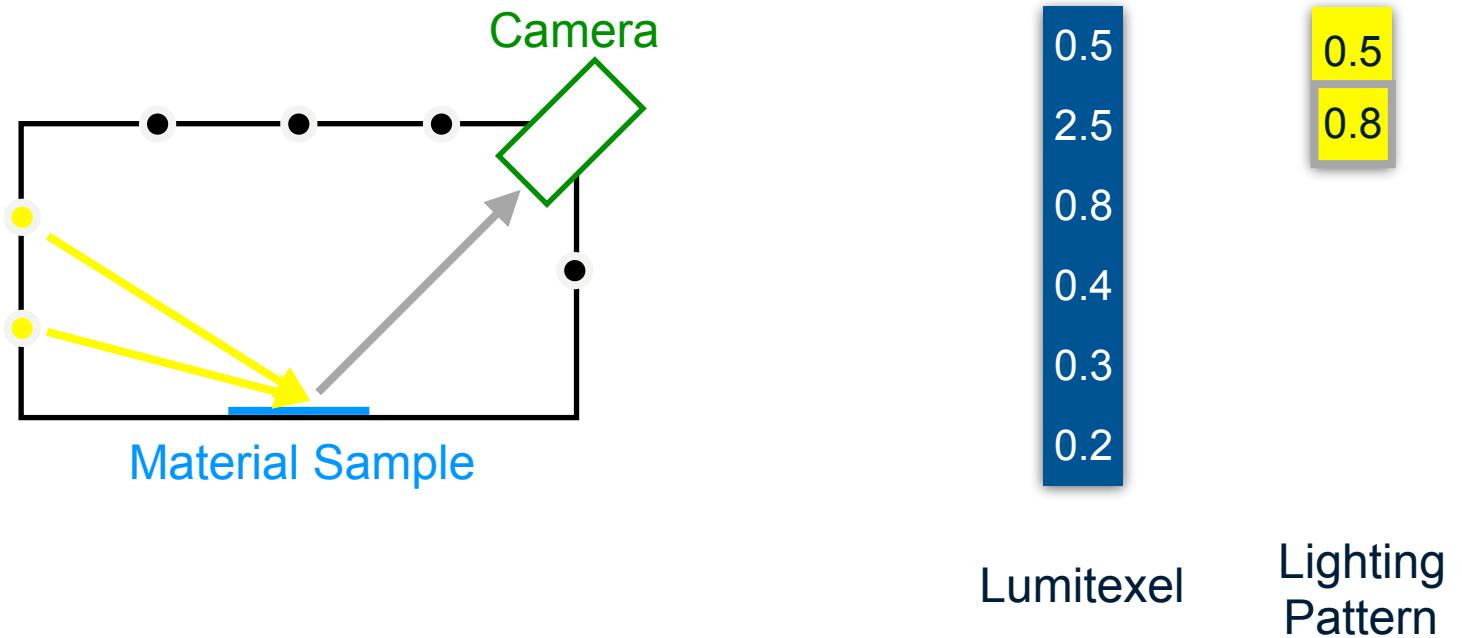
Time Consuming

1. Many Lights
2. Long Exposure

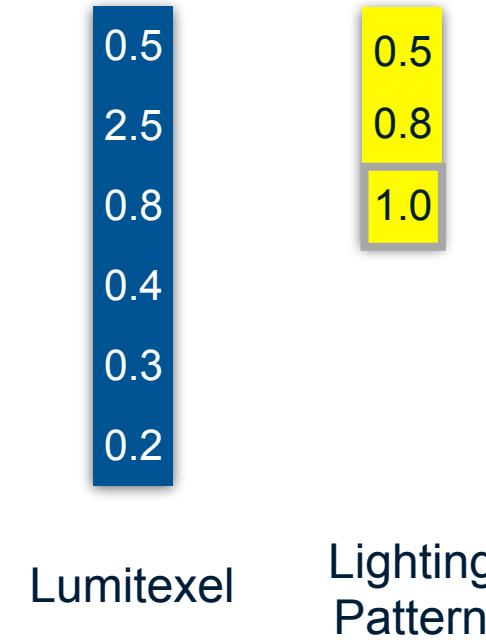
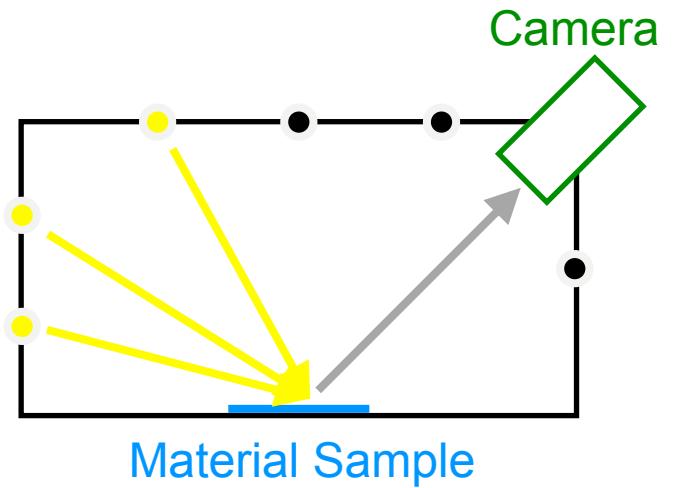
LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



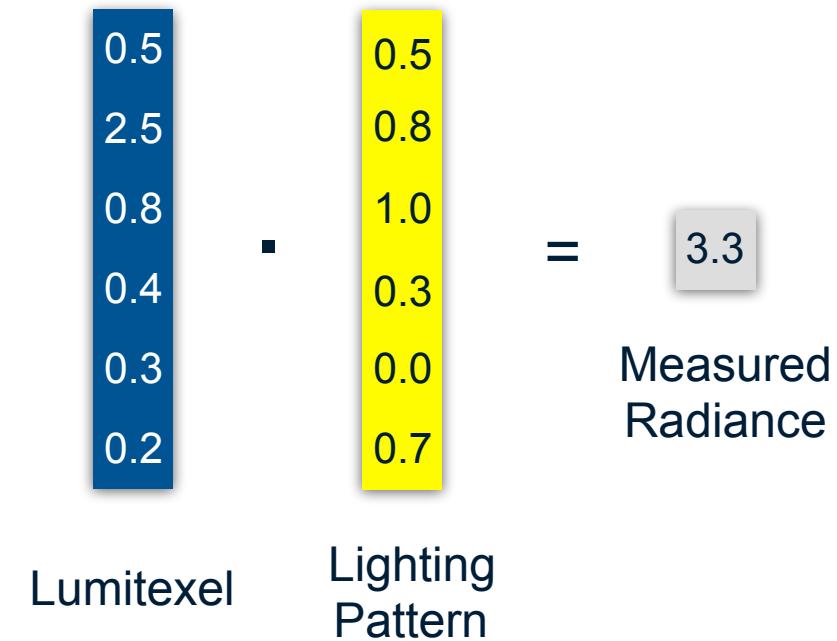
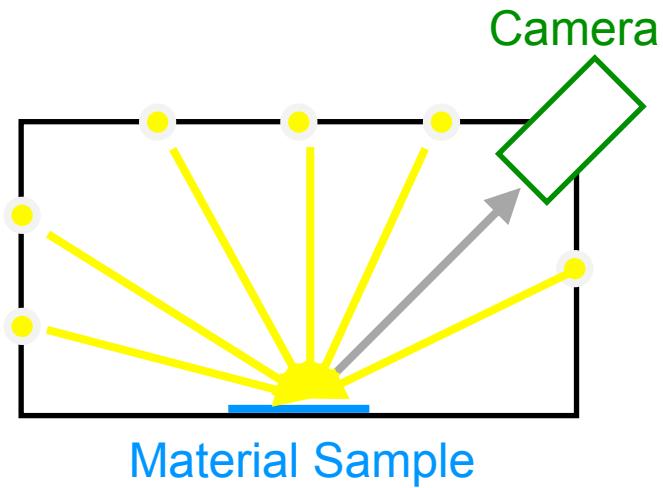
LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



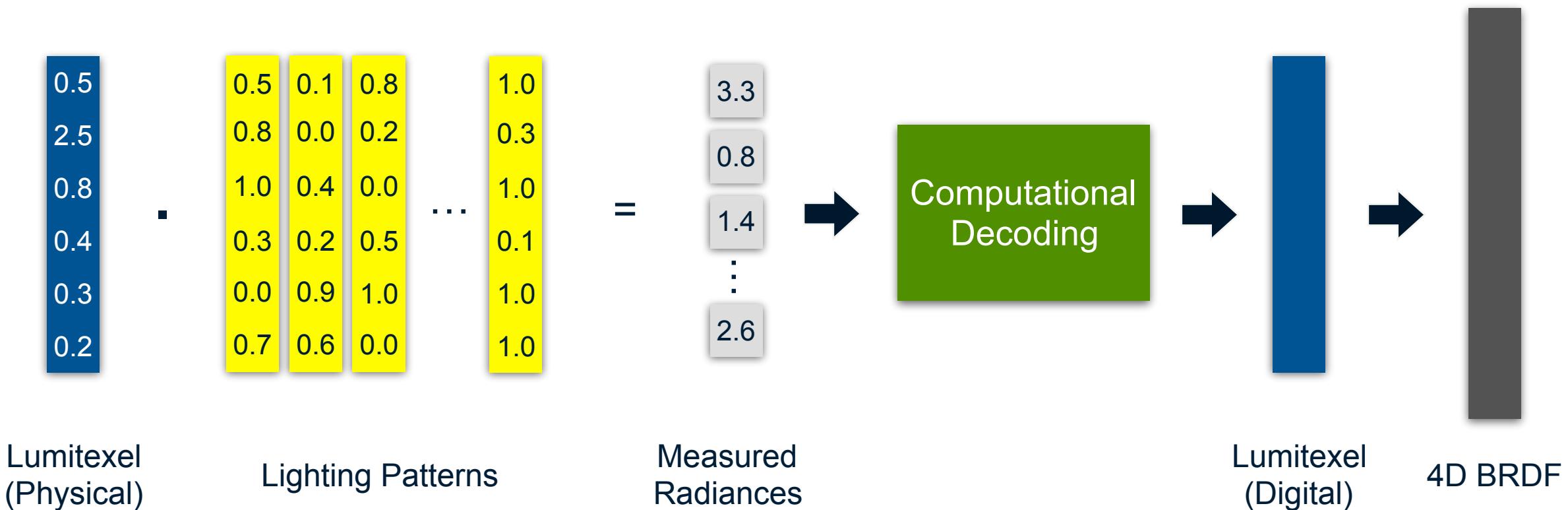
LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



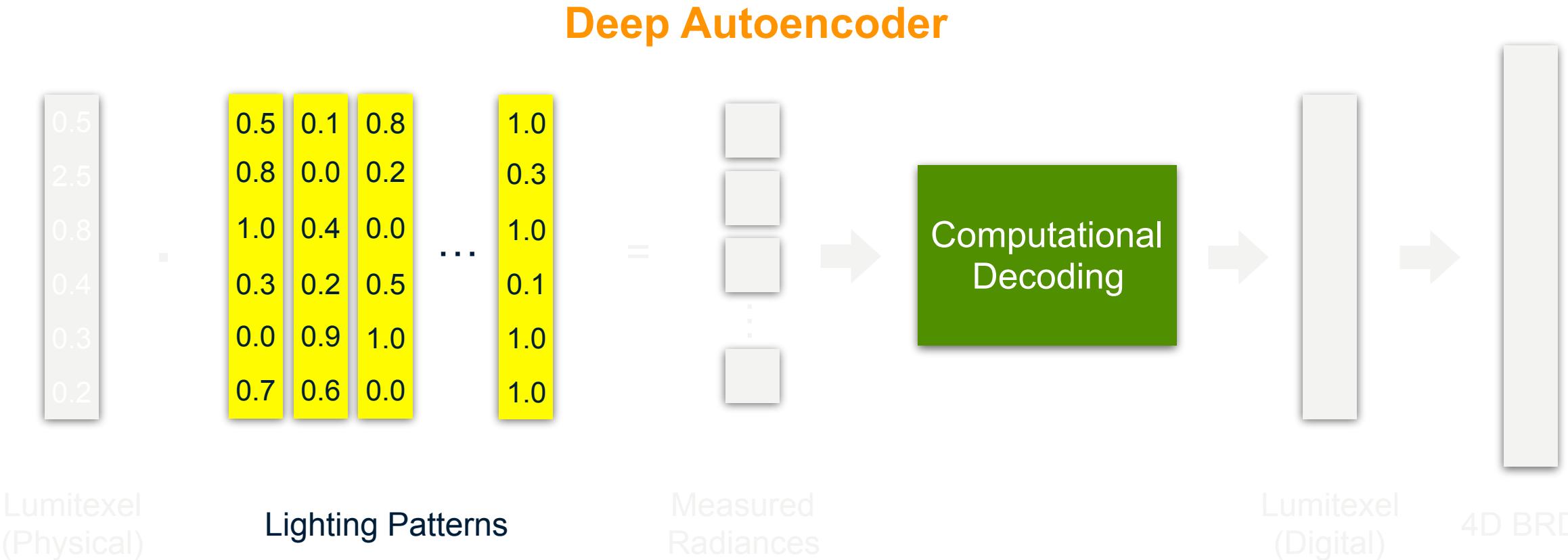
LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

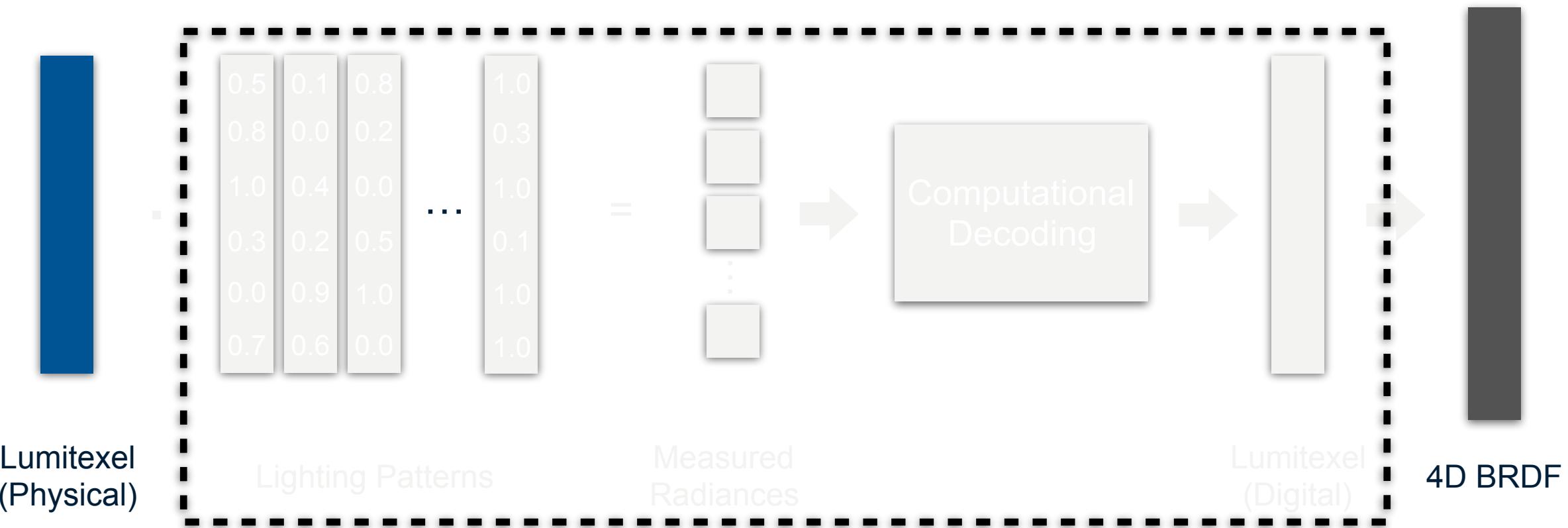


LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



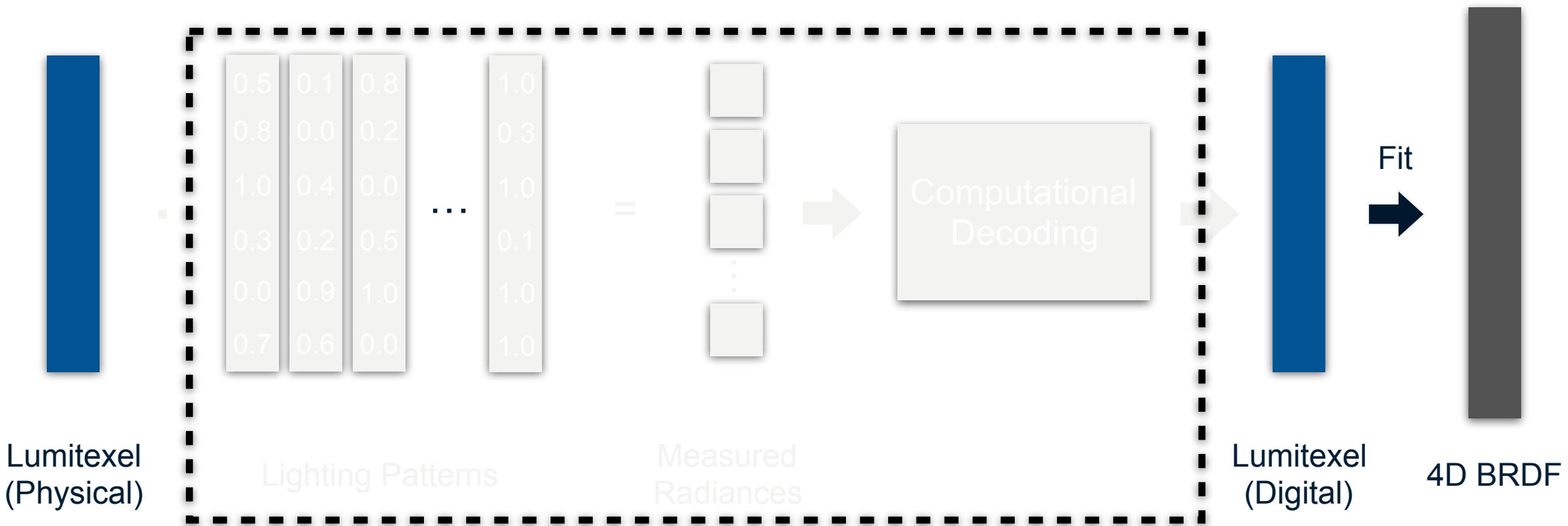
WHAT TO LEARN

- 1. Straightforward / Most End-to-end
- 2. Difficult to Learn the Complex Mapping



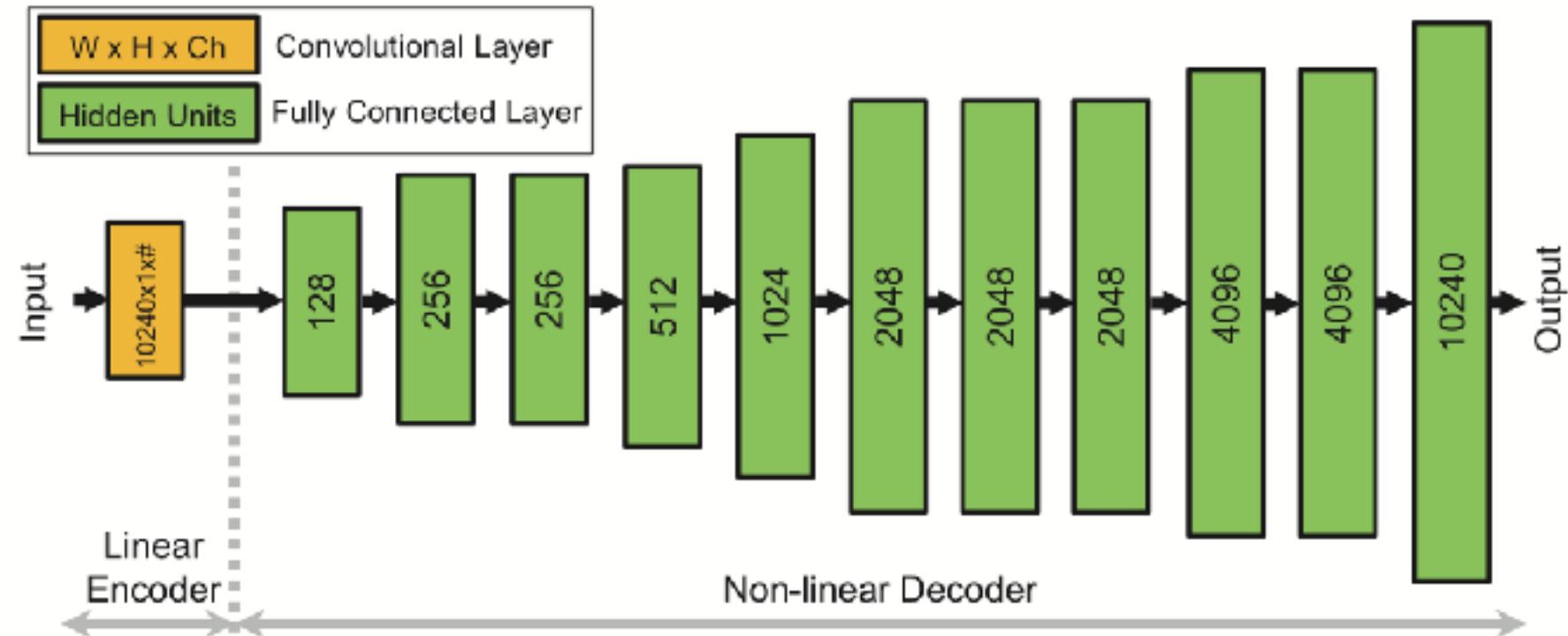
WHAT TO LEARN

1. Easy to Learn the One-to-one Mapping (Autoencoder)
2. Nonlinear Fitting to Obtain the 4D BRDF
3. Shift the Complexity to Fitting to Handle Near-field Lighting



DEEP AUTOENCODER FOR LUMITEXELS (L-DAE)

- Input/Output
 - Lumitexels
- Encoder
 - Bounded, Linear
 - 1 conv Layer
- Decoder
 - Nonlinear
 - 11 fc Layers
- Asymmetric & Mixed Domain



LOSS FUNCTION

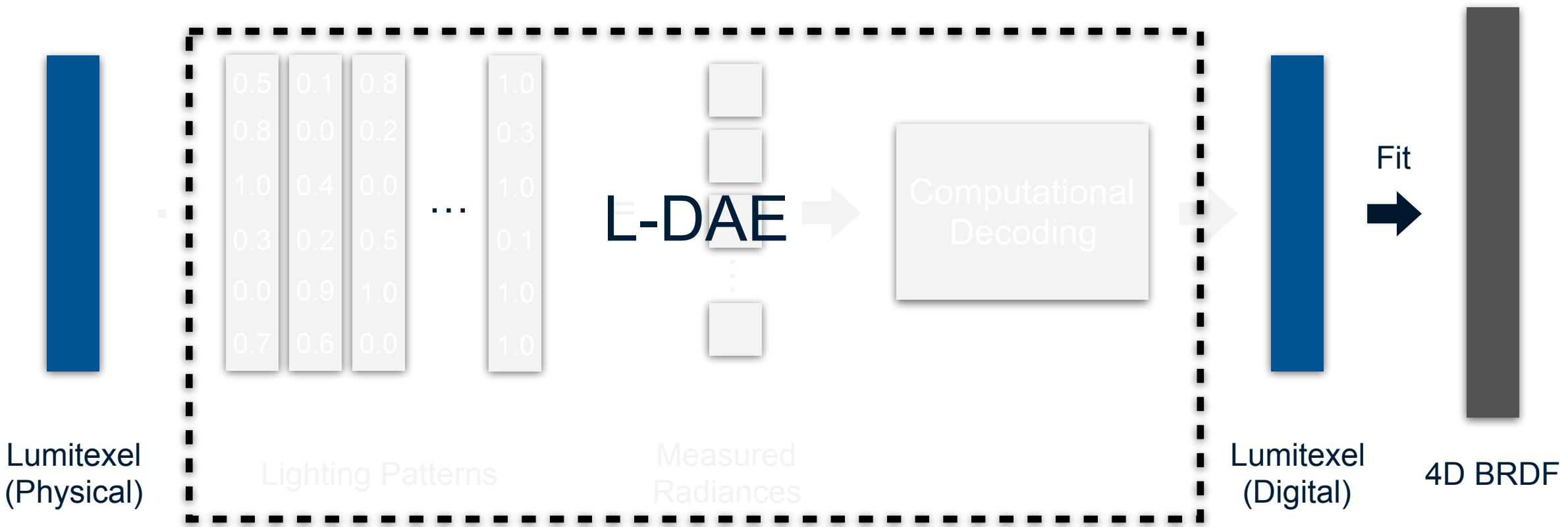
$$L = L_{\text{auto}}(m) + \lambda \sum_{w \in \text{enc.}} L_{\text{barrier}}(w).$$

- Reconstruction Error Term $L_{\text{auto}}(m) = \sum_j [\log(1 + m(j)) - \log(1 + m_{\text{gt}}(j))]^2.$
- Encoder Plausibility Term $L_{\text{barrier}}(w) = \tanh\left(\frac{w - (1 - \epsilon)}{\epsilon}\right) + \tanh\left(\frac{-w + \epsilon}{\epsilon}\right) + 2.$

TRAINING DATA

- 1 Million Synthetic Lumitexels
 - Randomly Sampled Location on the Sample Plane, Local Frame, BRDF Parameters (anisotropic GGX)
 - Based on Calibration Data of the Acquisition Device

BRDF FITTING



BRDF FITTING

- Input: Decoded Lumitexel
- Output: GGX BRDF Parameters
 - Diffuse / Specular Albedo, Roughnesses, Normal, Tangent
- Nonlinear Optimization
 - Levenberg-Marquardt

Results

TIMING

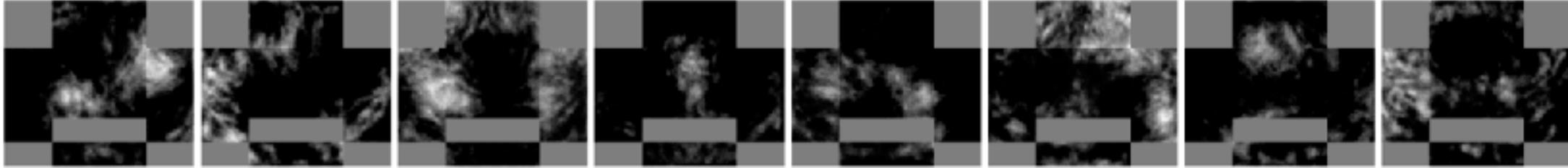
- Training Time: 5h
- Encoding Time: 12~25s
- Decoding Speed: 4m / 1M Lumitexels
- Fitting Speed: 1.6h / 1M Lumitexels

LIGHTING PATTERNS

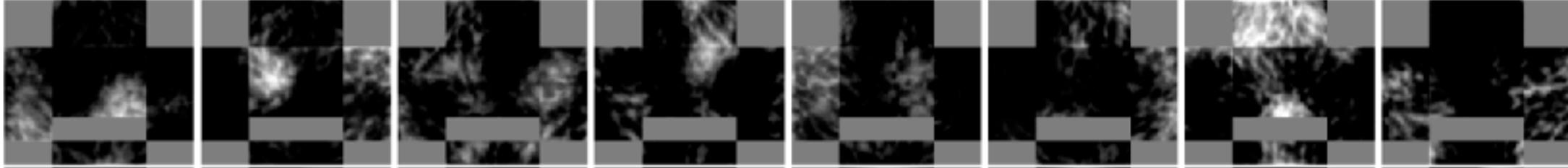
Sample Lit with Our Patterns (Aniso.)



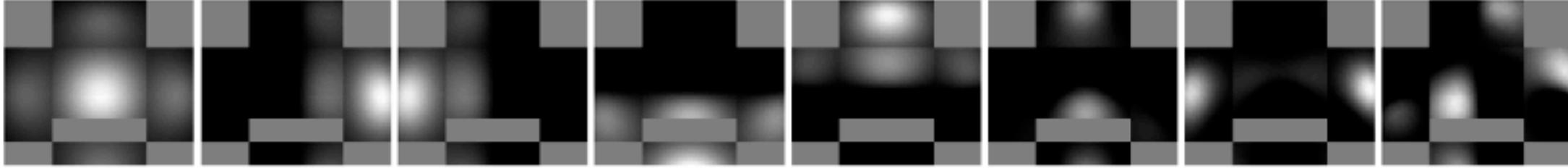
Ours (Aniso.)



Ours (Iso.)



PCA

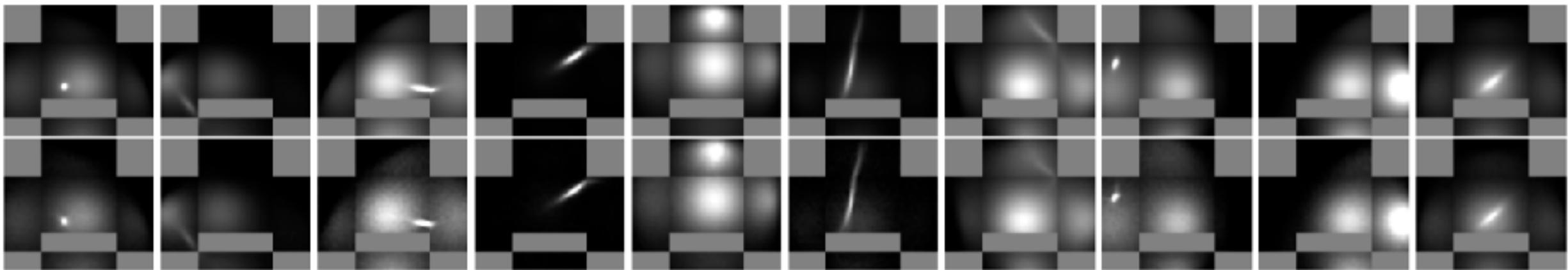


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[256x256]	Toe	[64x32]
[1024x1024]	Rail	[64x32]

SYNTHETIC LUMITEXEL RECONSTRUCTION

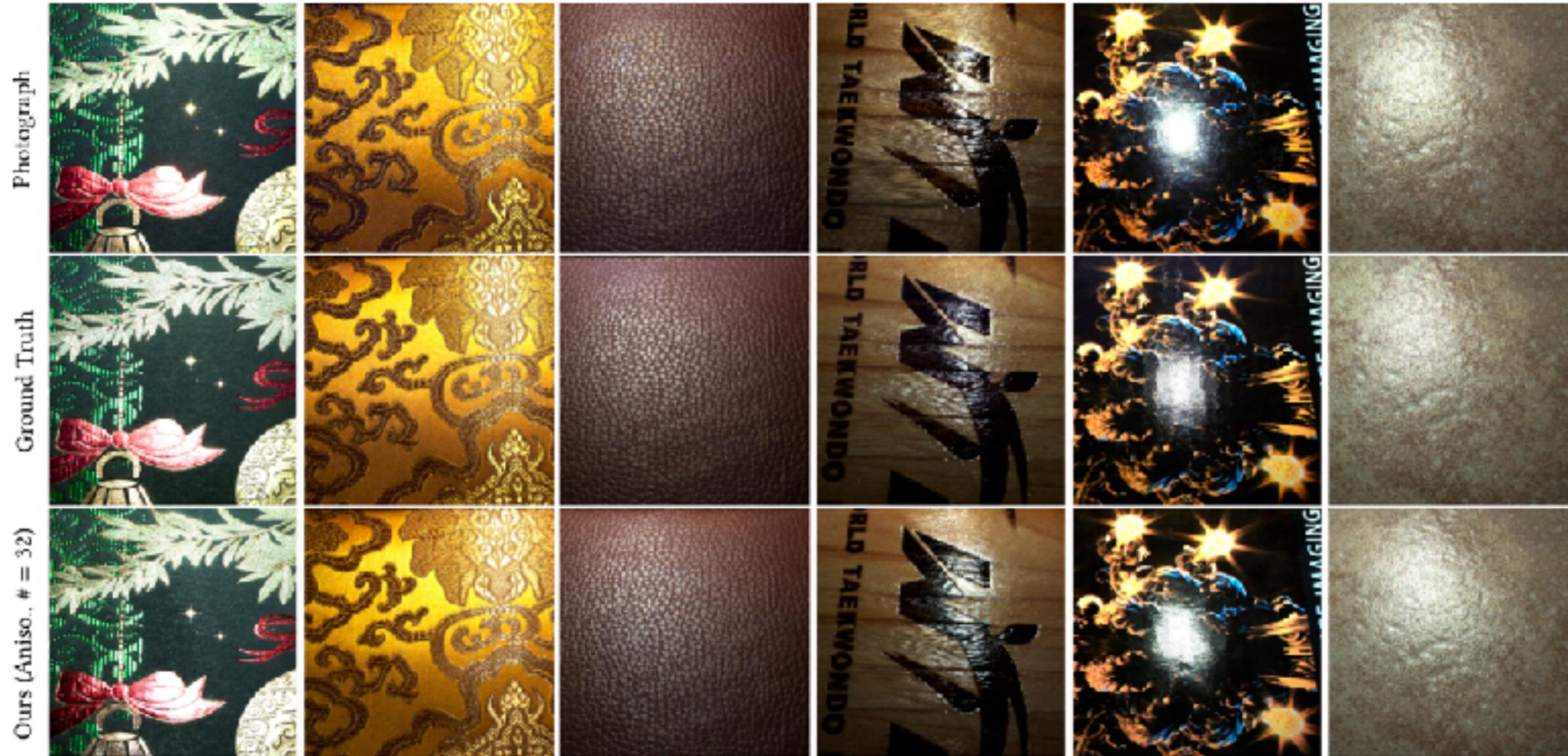
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Input Lumitexels



Reconstructed Lumitexels
with 32 Lighting Patterns

VALIDATION ON PHYSICAL SAMPLES



Complete Acquisition Using
32 Learned Lighting Patterns

DIFFUSE ALBEDO

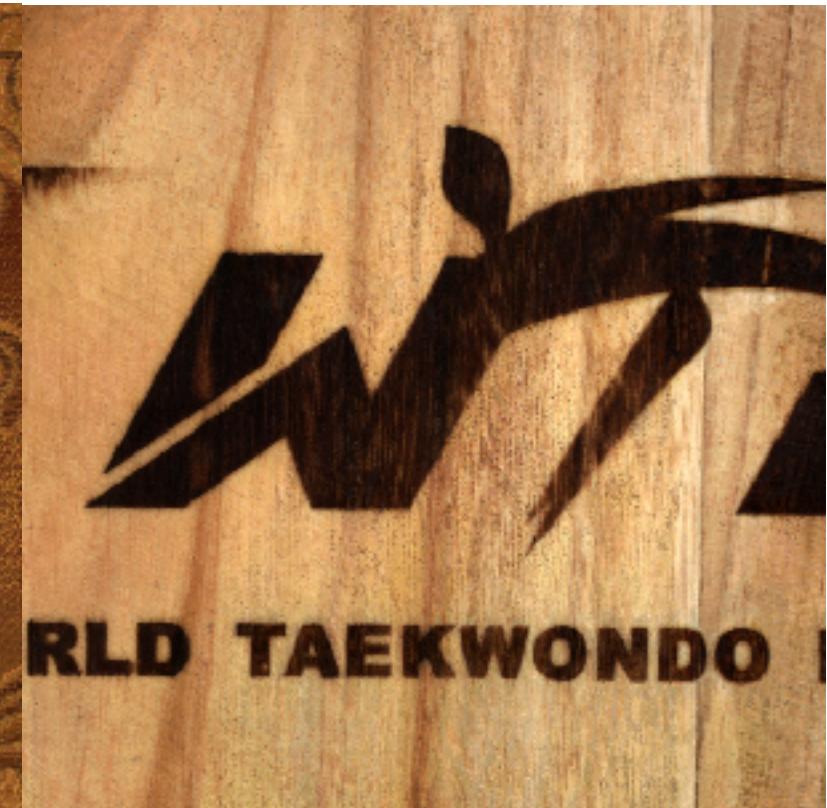
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Xmas-card



Fabric



Wood

SPECULAR ALBEDO



Xmas-card

Fabric

Wood

NORMAL



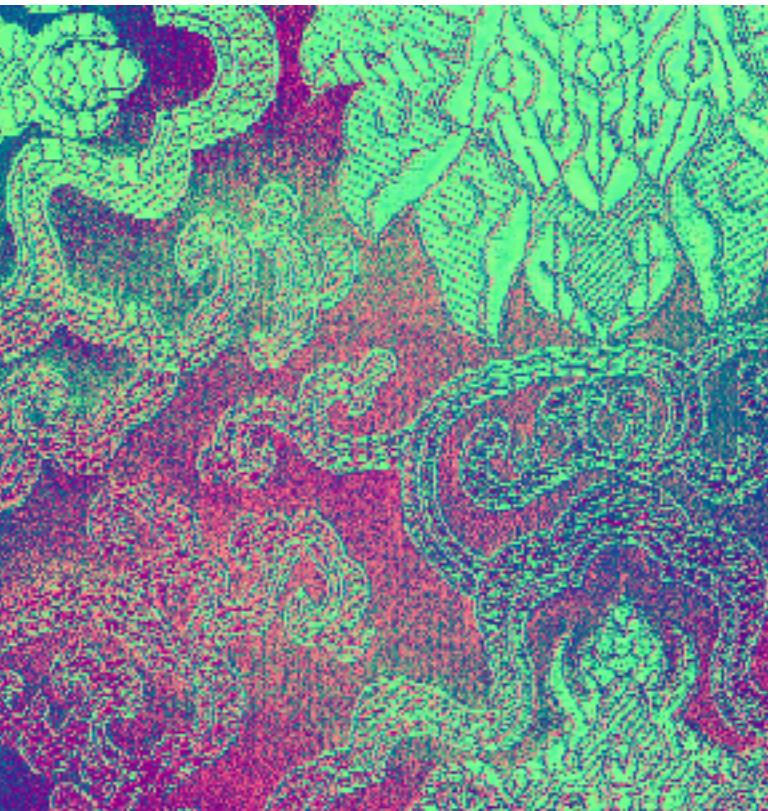
Xmas-card

Fabric

Wood



Xmas-card

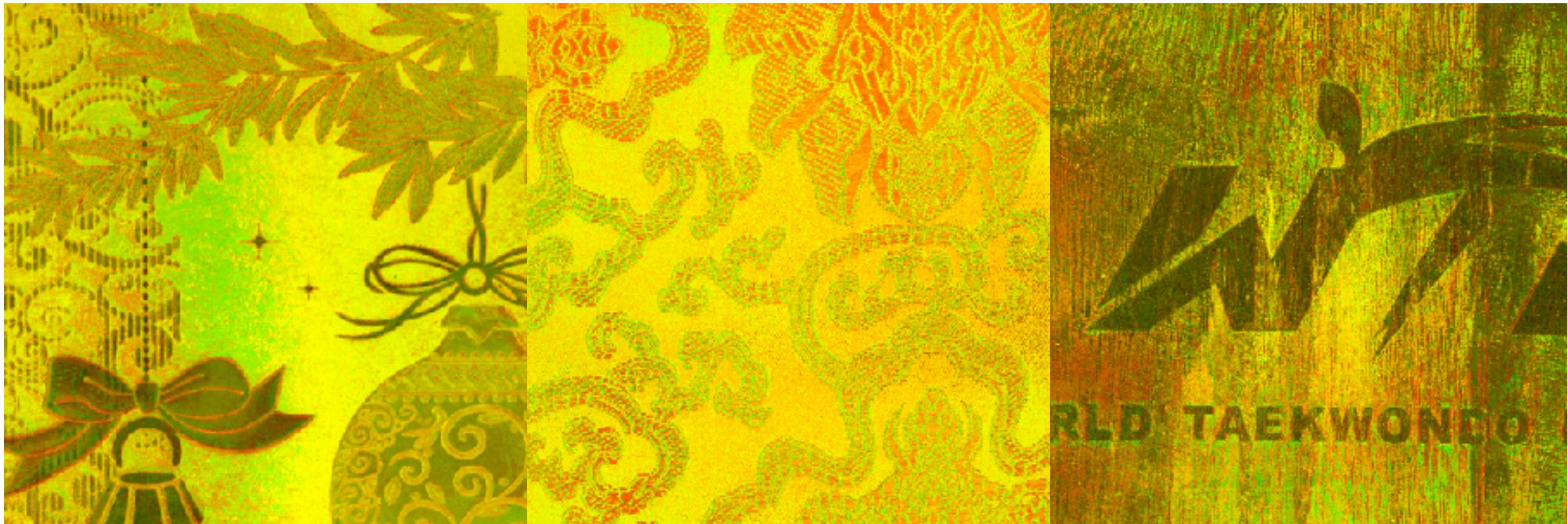


Fabric



Wood

ROUGHNESS

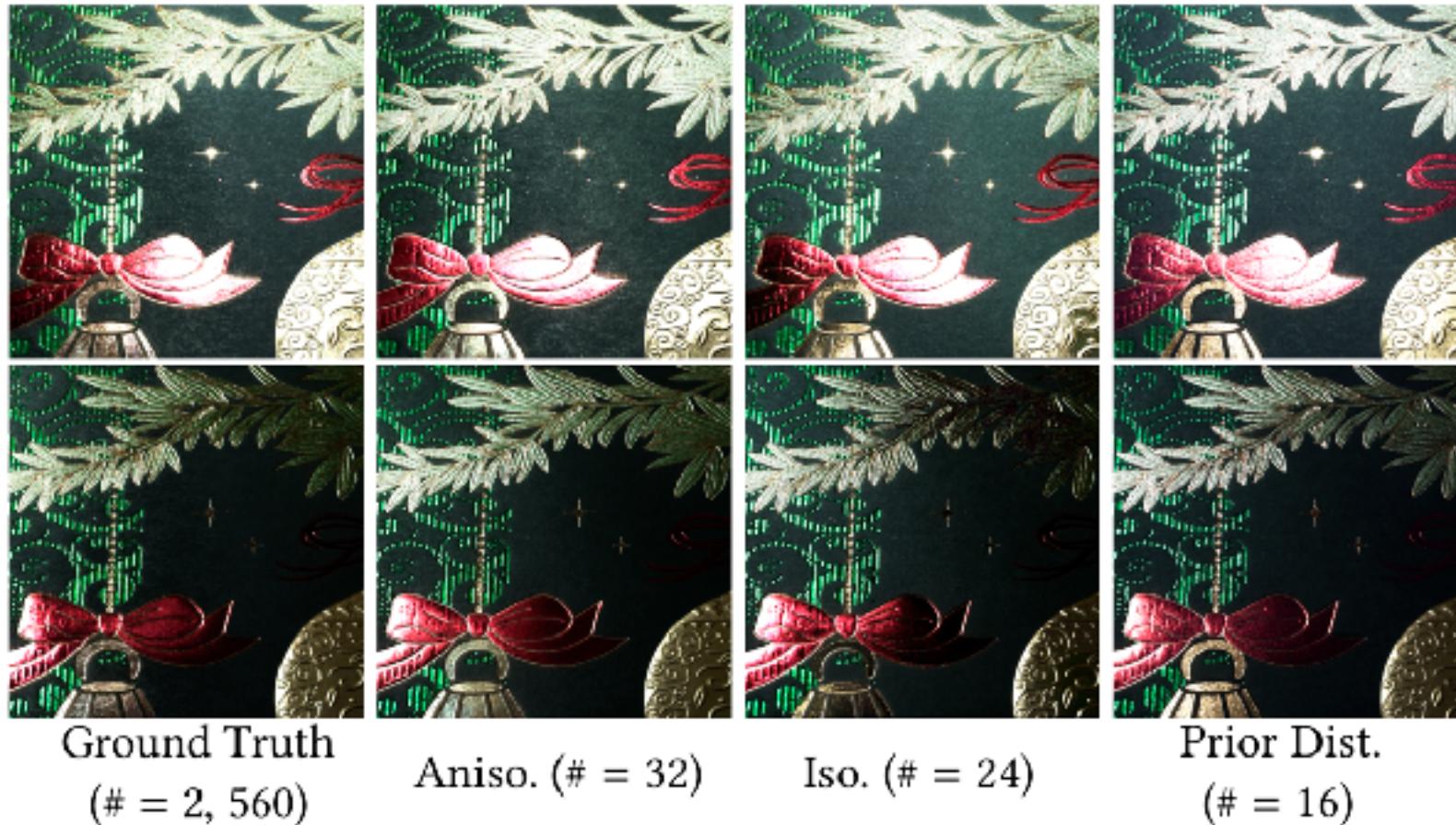


Xmas-card

Fabric

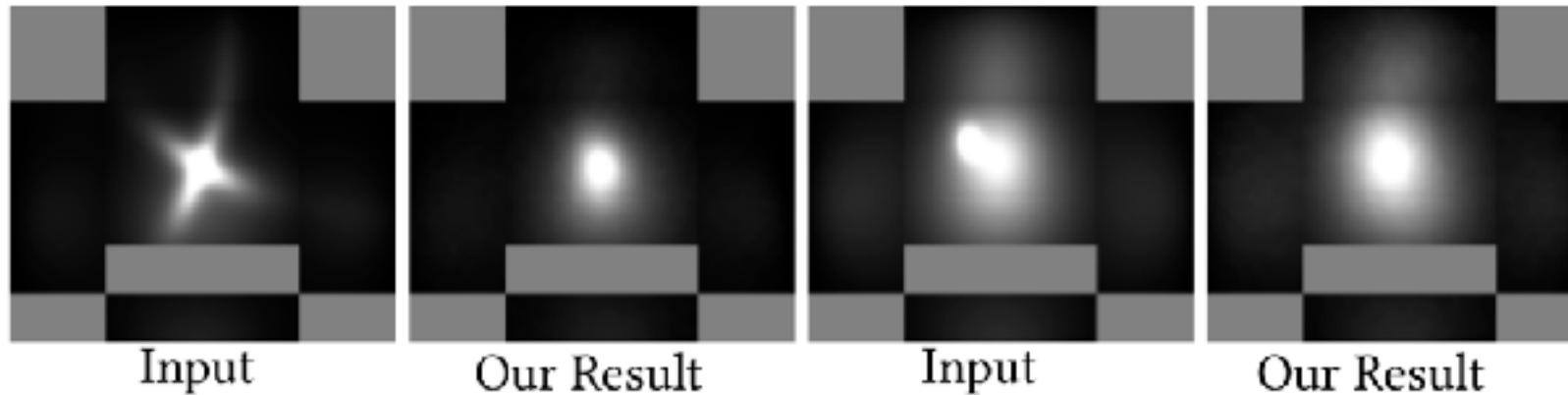
Wood

IMPACT OF TRAINING DATA



LIMITATIONS

- Cannot Reconstruct Lumitexels Substantially Deviated from Training Data



LIMITATIONS

- Cannot Reconstruct Lumitexels Substantially Deviated from Training Data
- Mostly Planar Samples

LIMITATIONS

- Cannot Reconstruct Lumitexels Substantially Deviated from Training Data
- Mostly Planar Samples
- One Fixed View

Conclusions

CONCLUSIONS & FUTURE WORK

- A Framework that Automatically Learns to Capture SVBRDF Efficiently
- Towards **AI-Assisted** Optimal & Joint Design of **Hardware + Software**

CONCLUSIONS & FUTURE WORK

- A Framework that Automatically Learns to Capture SVBRDF Efficiently
- Towards **AI-Assisted** Optimal & Joint Design of **Hardware + Software**
- Future Directions
 - Apply to Existing Setup (e.g., Lightstage, Linear Light Source)
 - View Optimization
 - Beyond Reflectance (e.g., Image Relighting [Xu et al. 2018])

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- Anonymous Reviewers
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- Fundamental Research Funds for the Central Universities (2017XZZX009-03)



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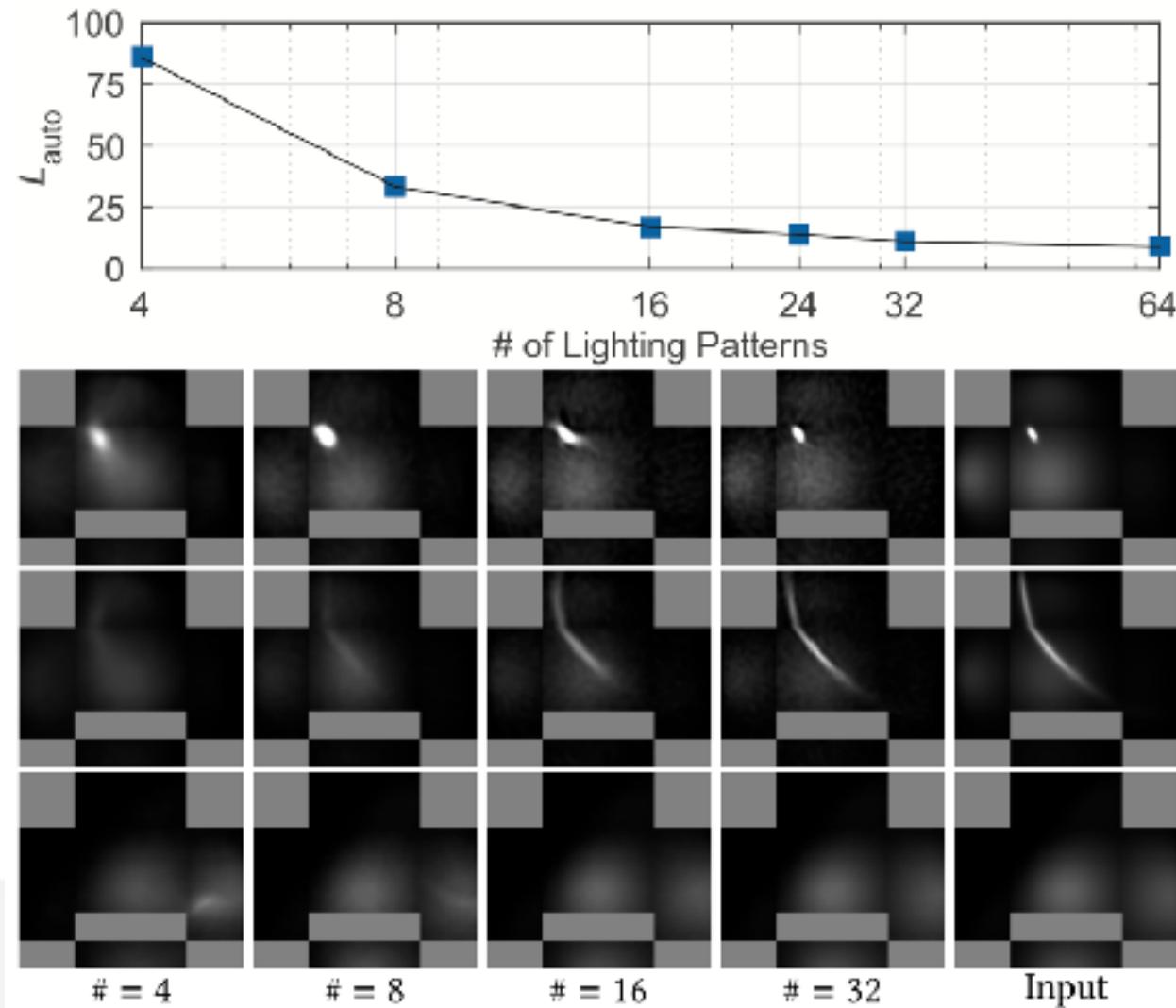
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Email: hwu@acm.org

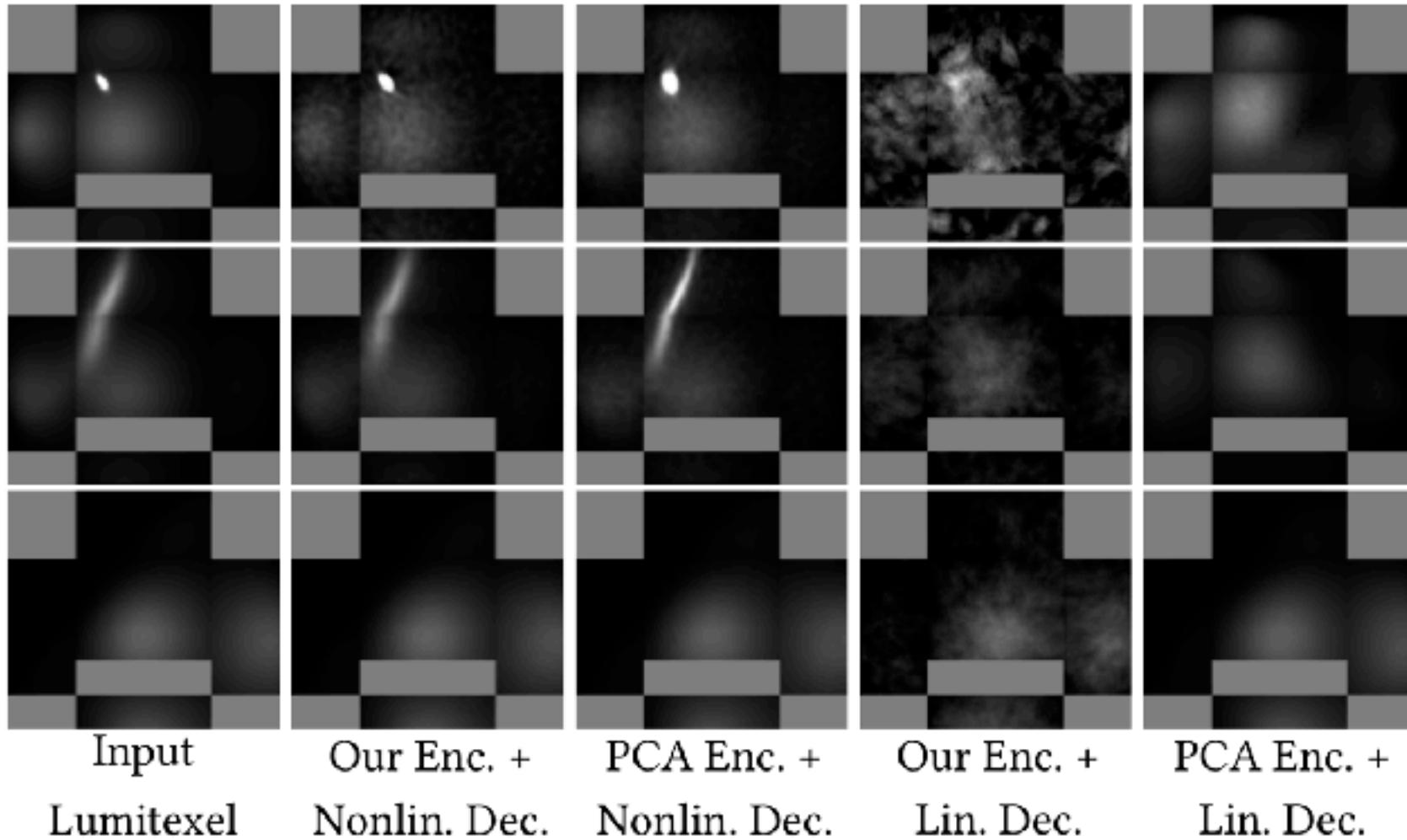
Project Webpage



BACKUP

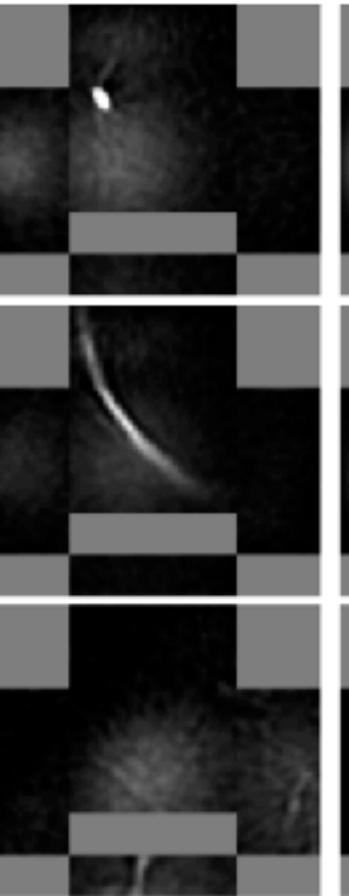
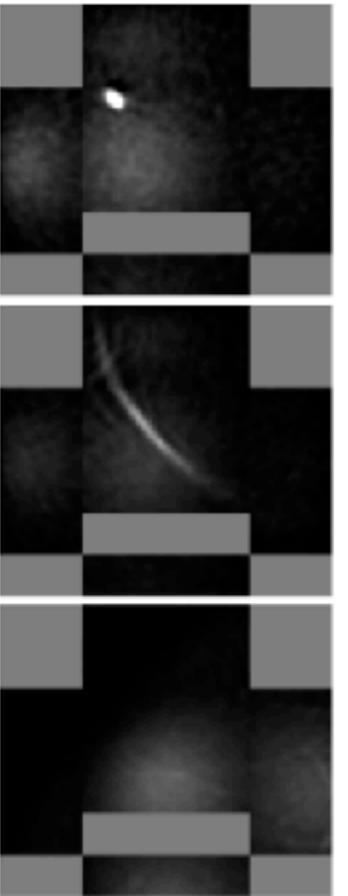
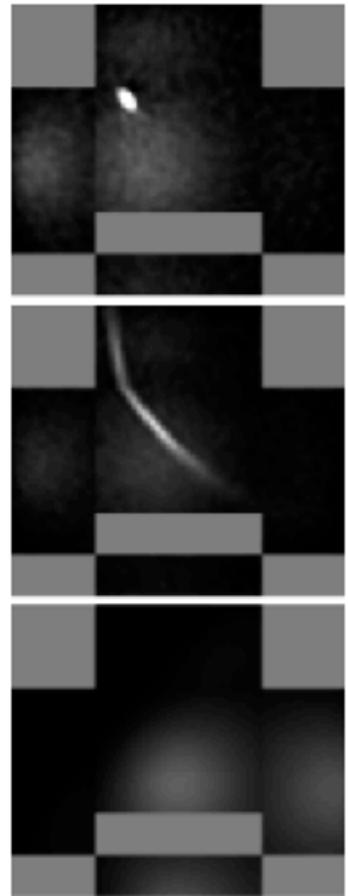


BACKUP



BACKUP

Reconstruction Results

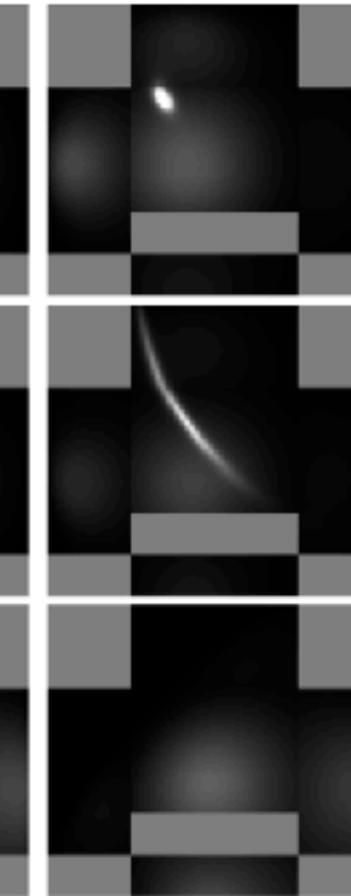
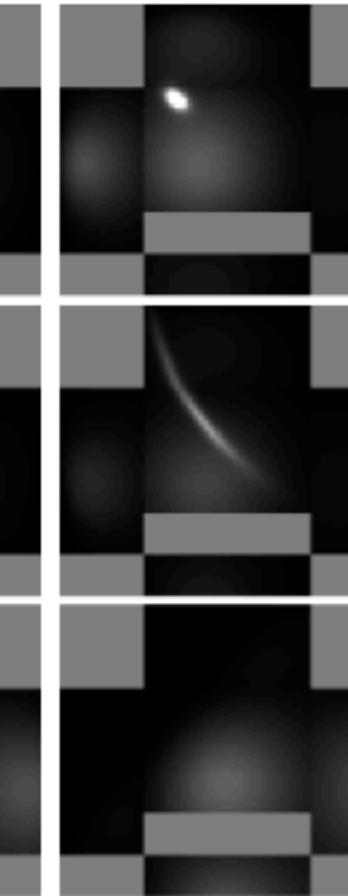


$\kappa = 0\%$

$\kappa = 2\%$

$\kappa = 5\%$

Fitting Results



$\kappa = 5\%$