



GENERATIONS / RECOVERED  
SIGGRAPH2018

# EFFICIENT REFLECTANCE CAPTURE USING AN AUTOENCODER

Kaizhang Kang\*, Zimin Chen\*, Jiaping Wang,  
Kun Zhou & **Hongzhi Wu**

State Key Lab of CAD&CG, Zhejiang University  
Sinovation Ventures

# Introduction

# INTRODUCTION

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- Realistic Material Appearance is **Important**

# INTRODUCTION

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- Realistic Material Appearance is **Important**



Visual Effects

© Paramount Pictures



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

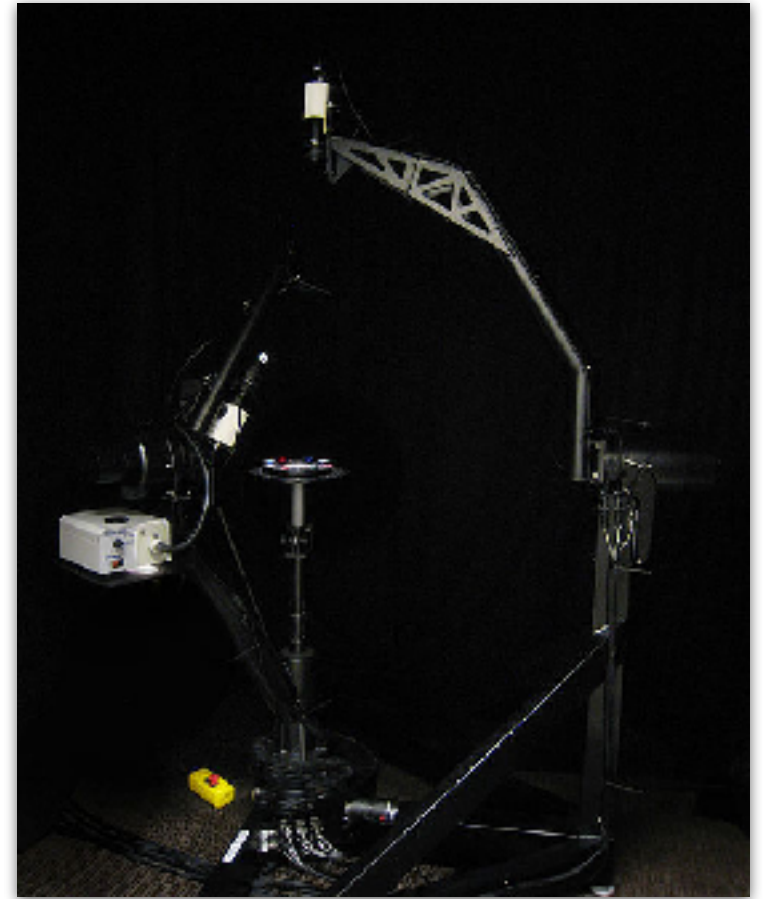
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- Realistic Material Appearance is **Important**
  - **6D** SVBRDF (Location, Lighting & View Directions)

# INTRODUCTION

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- 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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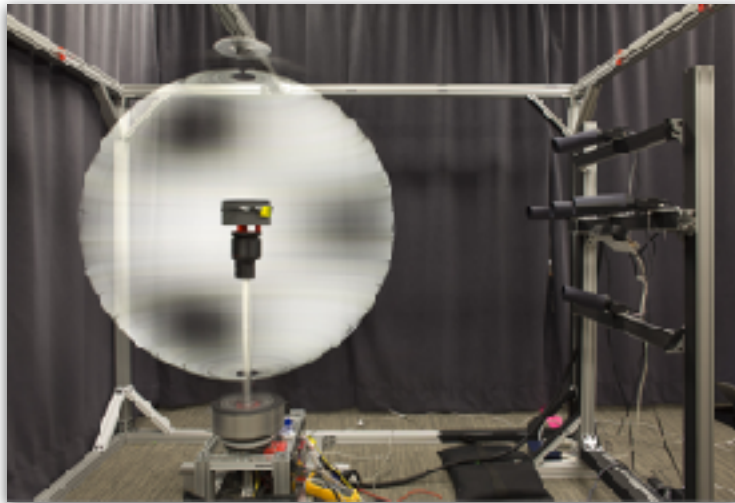
# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

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

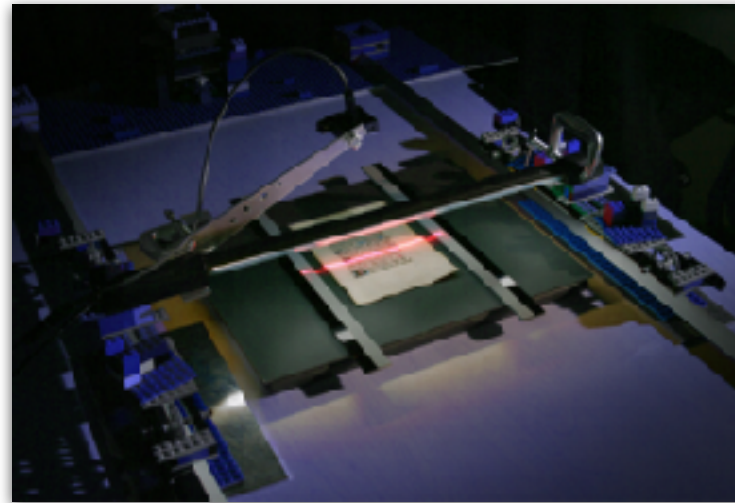
# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

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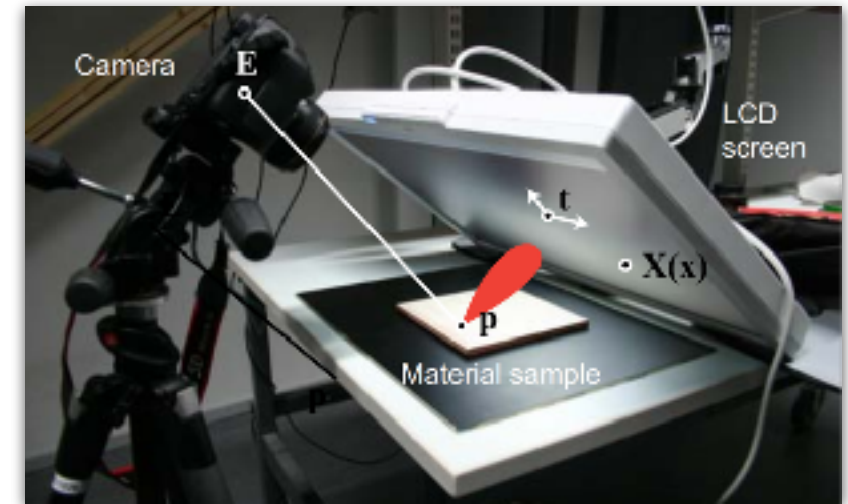
Lightstage

[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]

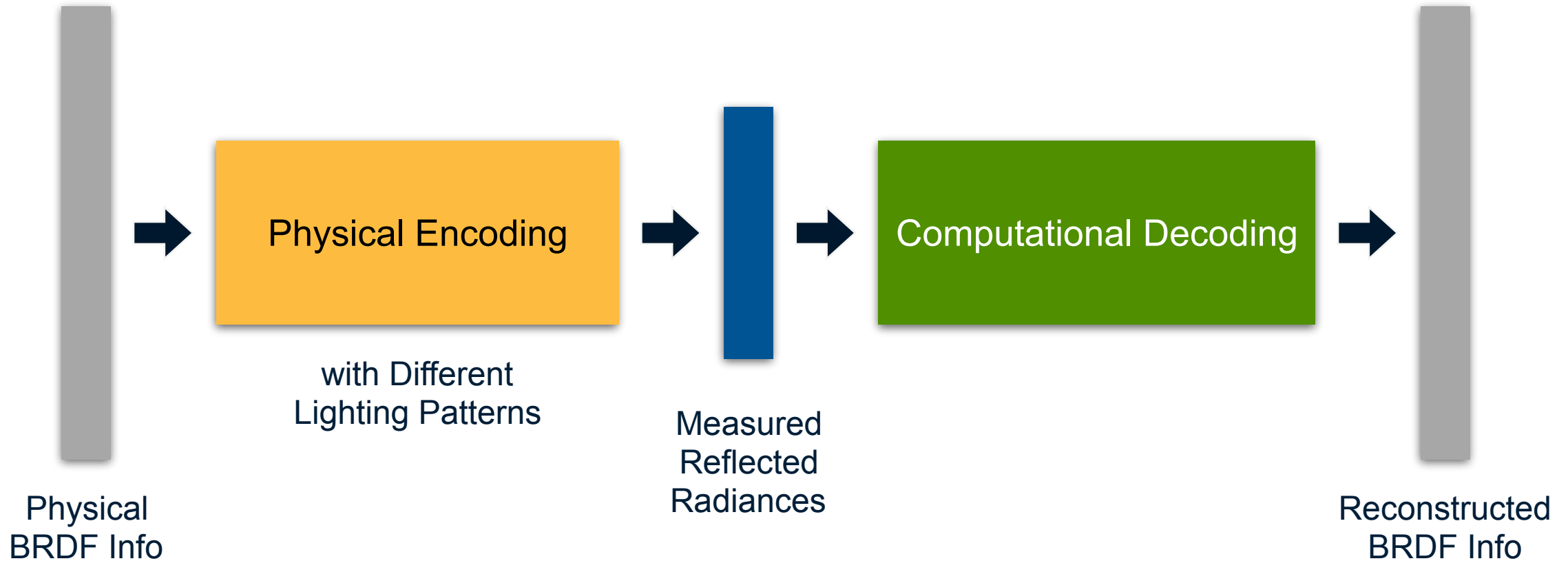


# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

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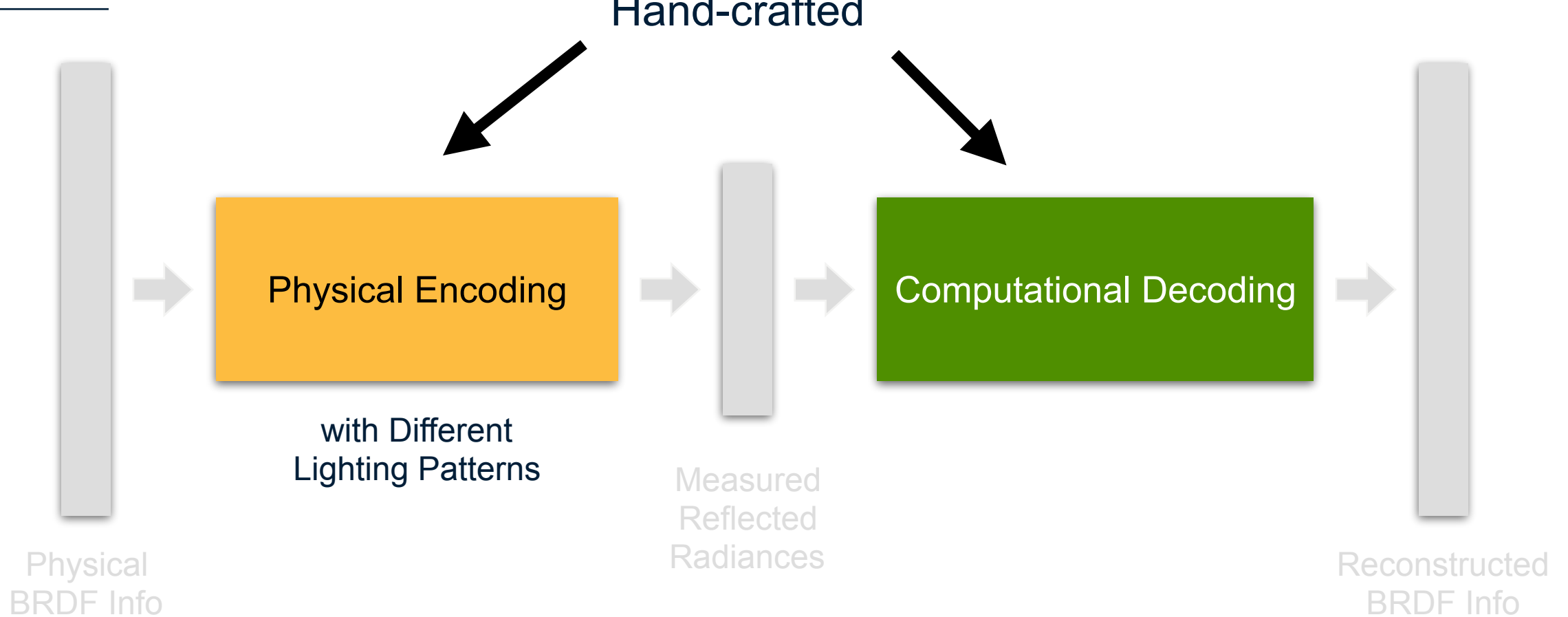
- One Successful Class for High-Quality Capture
- Program Many Lights Simultaneously
  - Forming Different Lighting Patterns
- 240 Lighting Patterns / Photos Needed [Chen et al. 2014]
- Our Goal
  - **To Improve Efficiency for SVBRDF Acquisition**

# REFLECTANCE ACQUISITION = ENCODING + DECODING

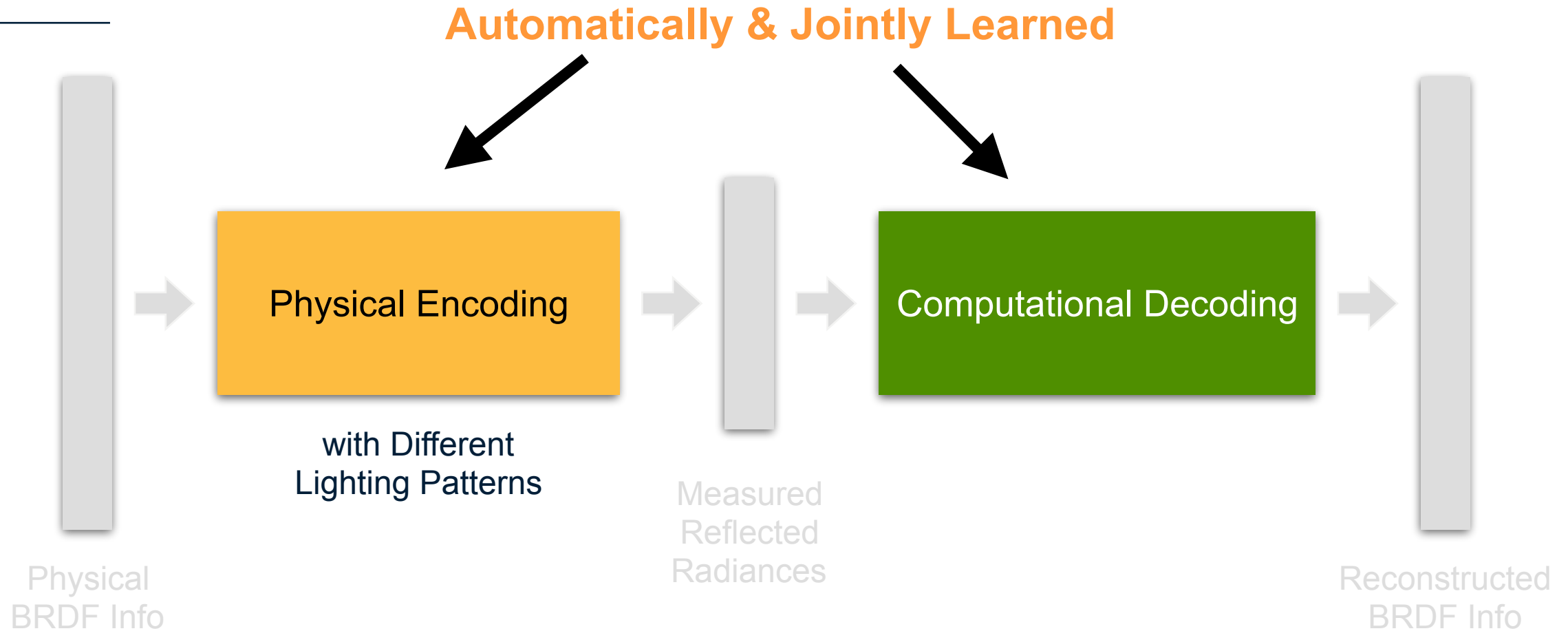


# PREVIOUS WORK

Hand-crafted



# OUR APPROACH



## OUR APPROACH

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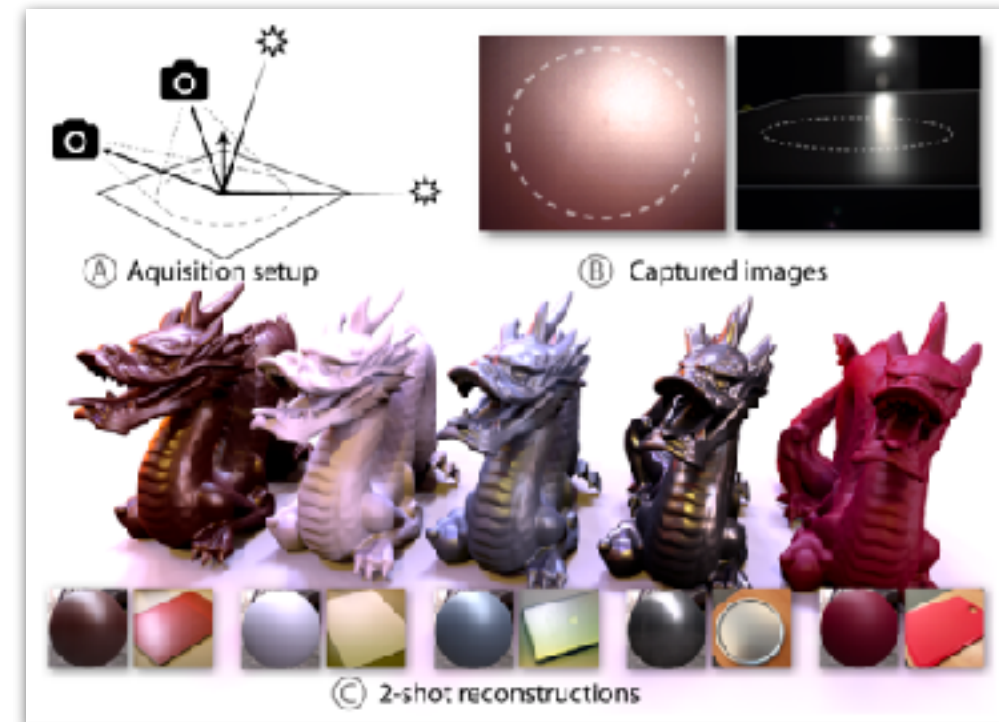
- Lighting Pattern #: 16~32
- Acquisition Time: 12~25s
- Faithfully Recovers **General** Reflectance under **Near-field Lighting**
- Automatically Adapt to Various Factors
  - Setup's Geometry
  - Properties of Appearance

# Related Work

## RELATED WORK

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- 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

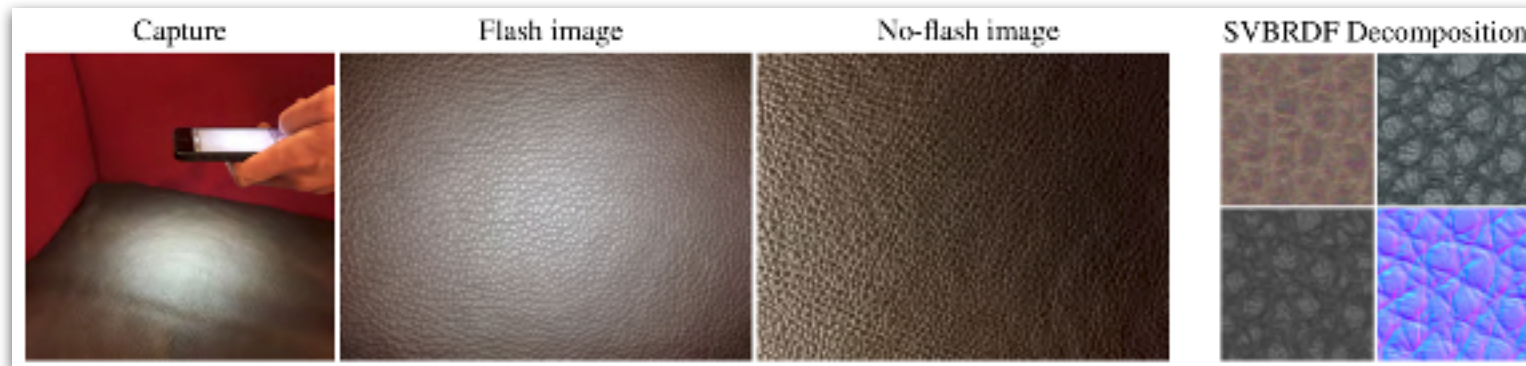


[Xu et al. 2016]

## RELATED WORK

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- 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]



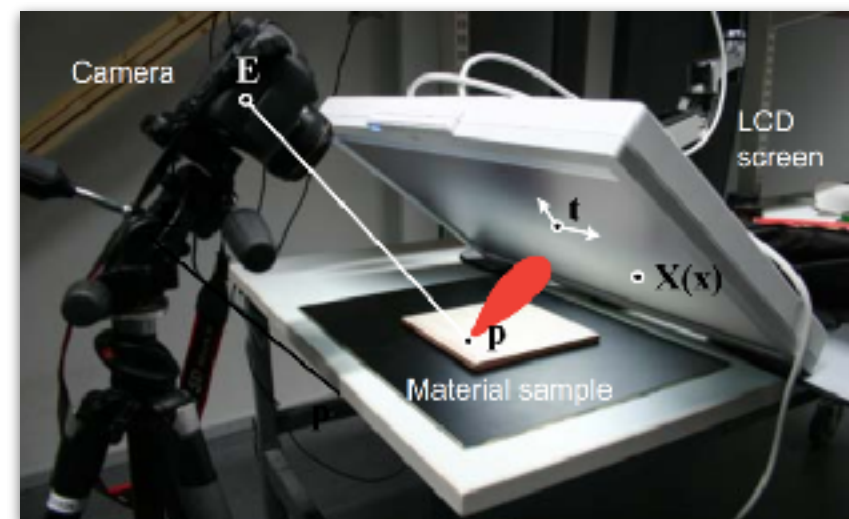
[Aittala et al. 2015]



## RELATED WORK

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- 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

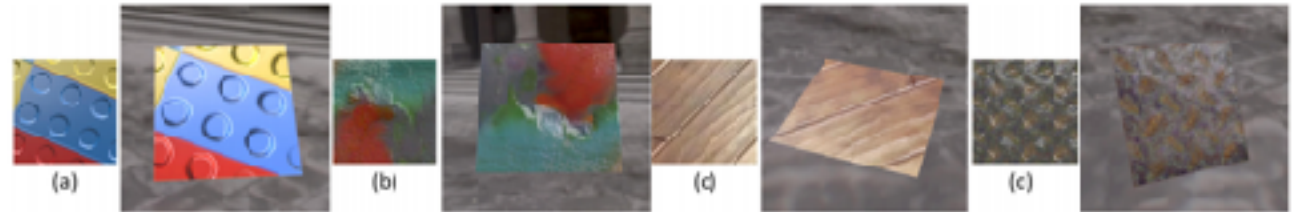


[Aittala et al. 2013]

## RELATED WORK

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- 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

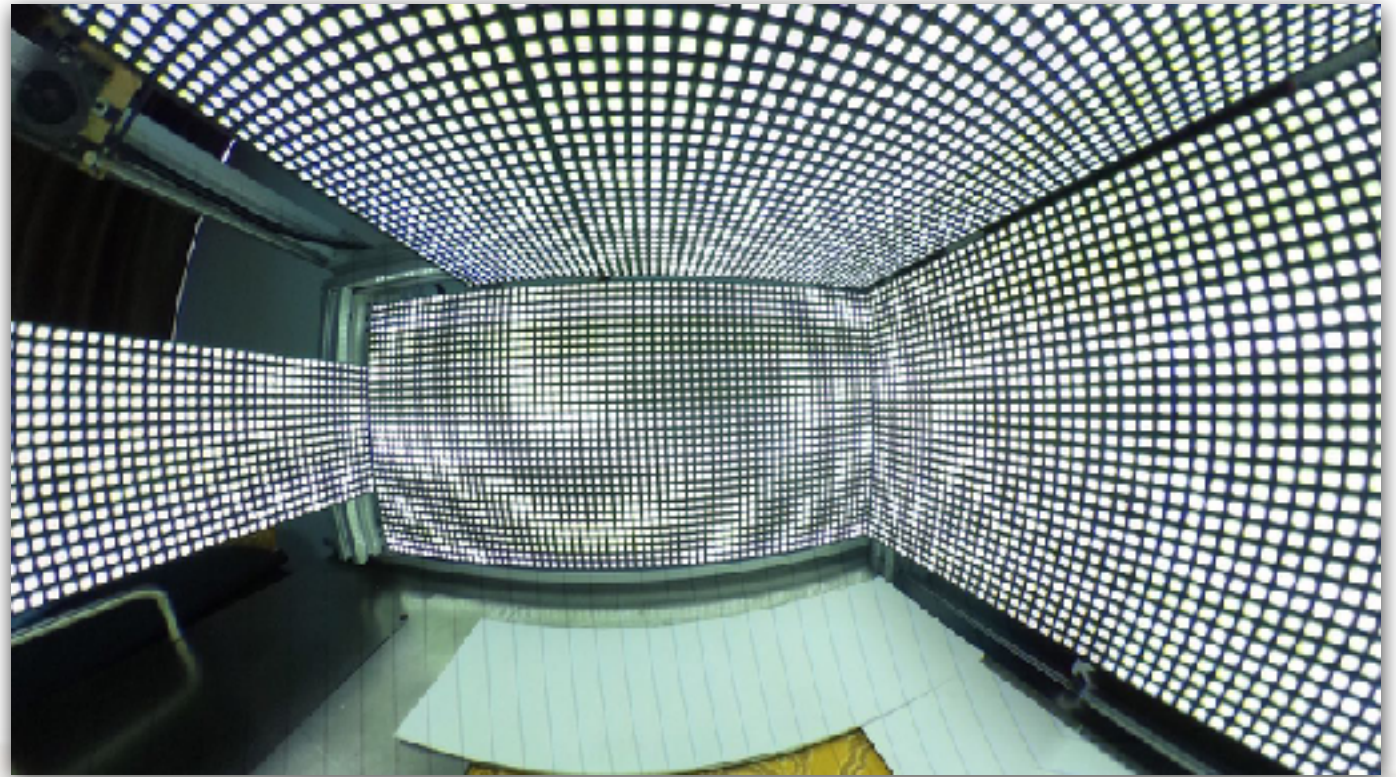
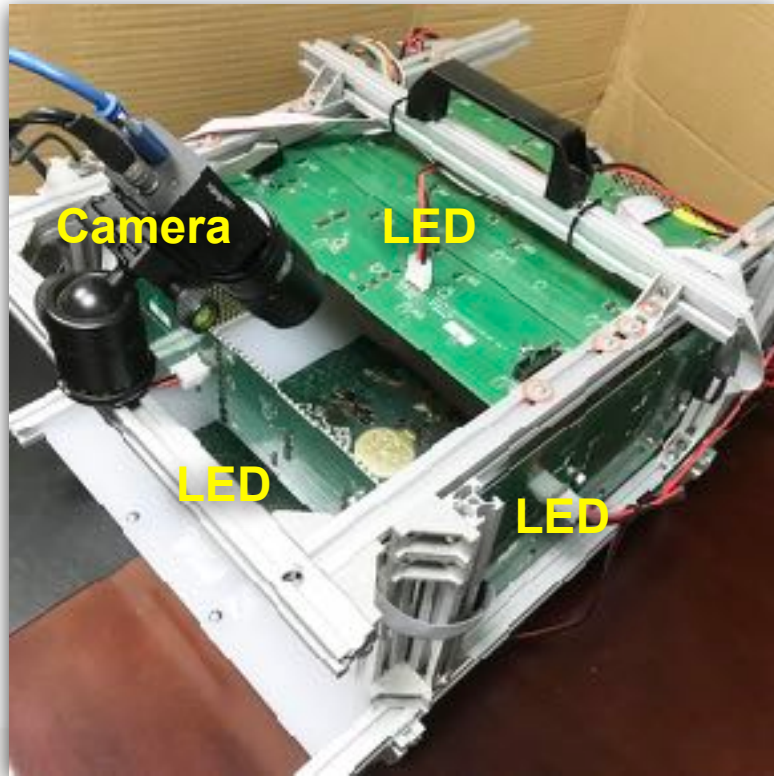
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- A Mostly Planar Sample
- Independent Reflectance Reconstruction on Each Location
- For a Single Channel

# OUR DEVICE

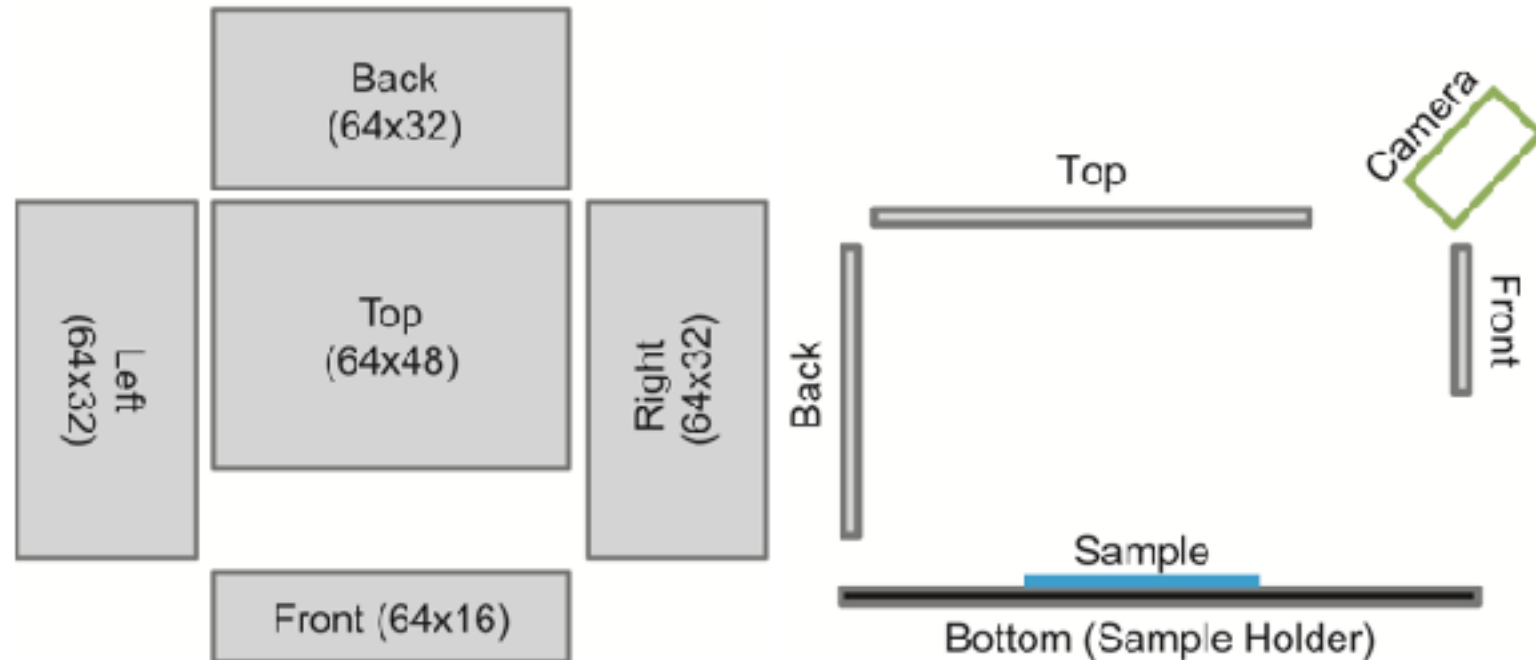
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- Mini, Near-field Lightstage



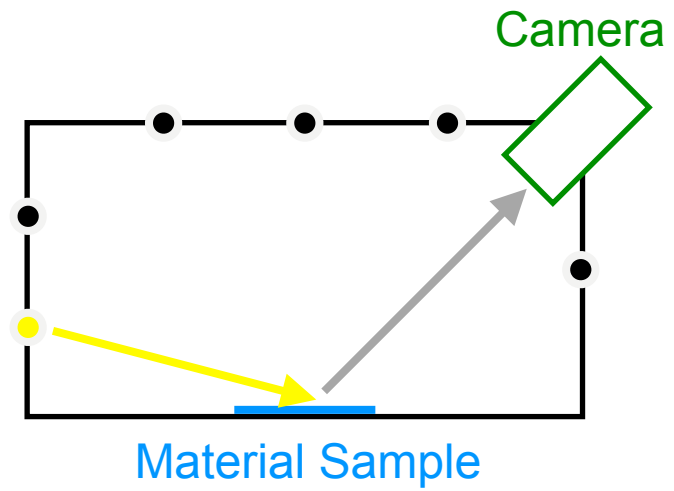
## OUR DEVICE

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



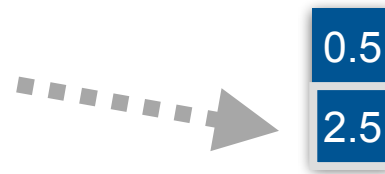
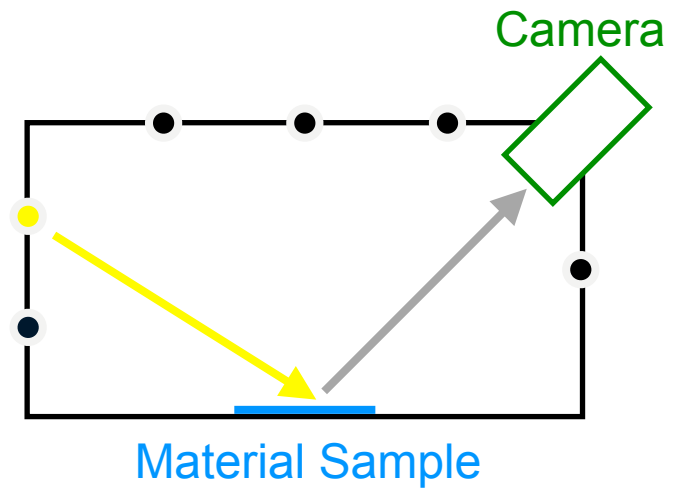
# LUMITEXEL

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Lumitexel

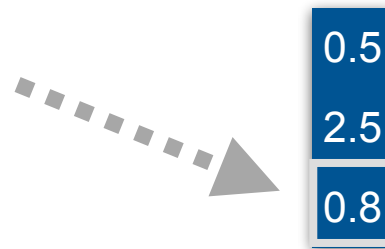
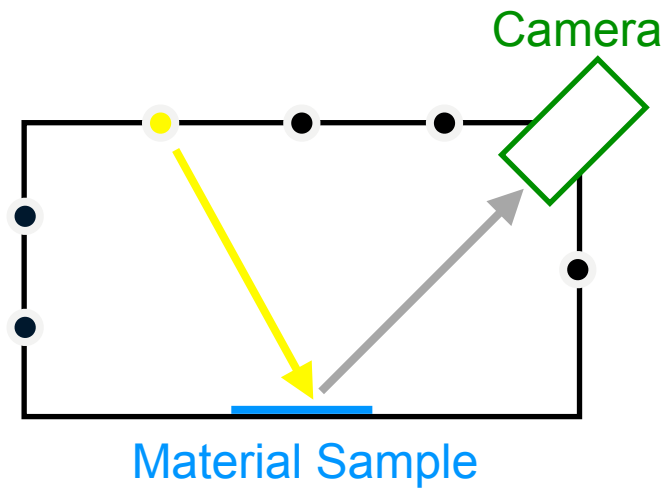
# LUMITEXEL



Lumitexel

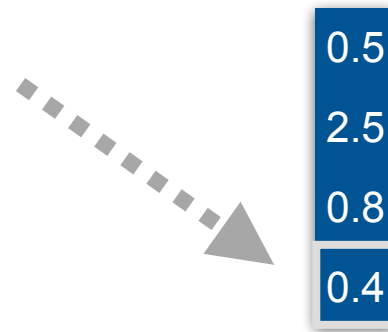
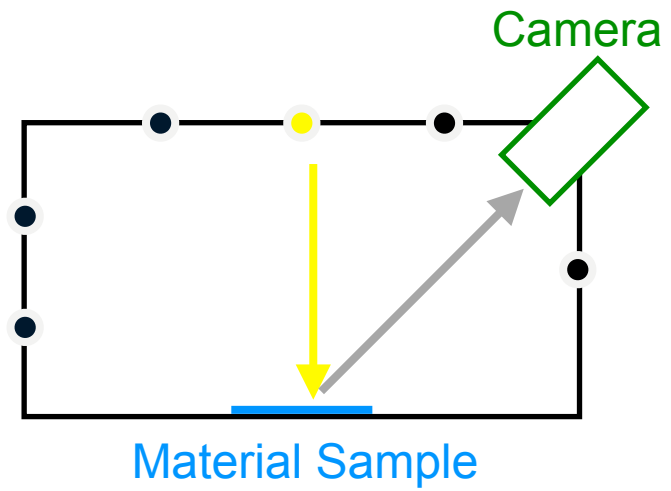


# LUMITEXEL



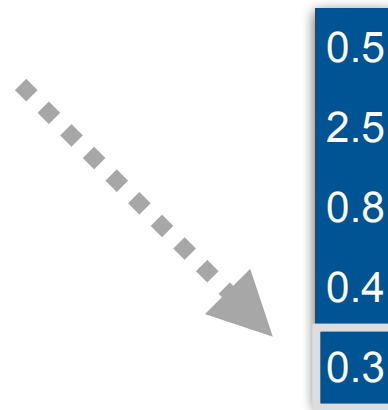
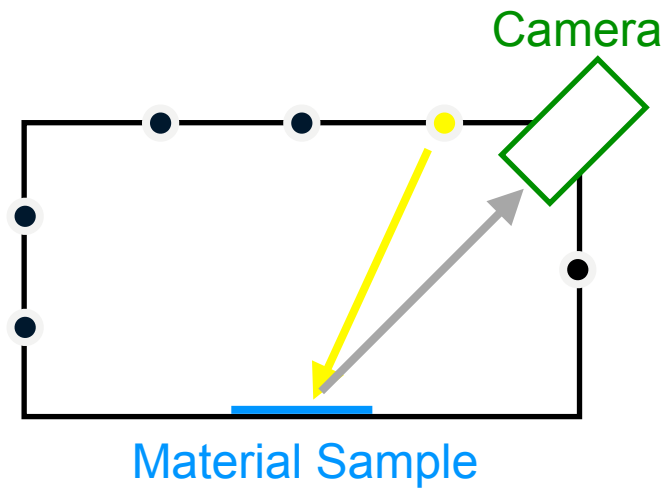
Lumitexel

# LUMITEXEL



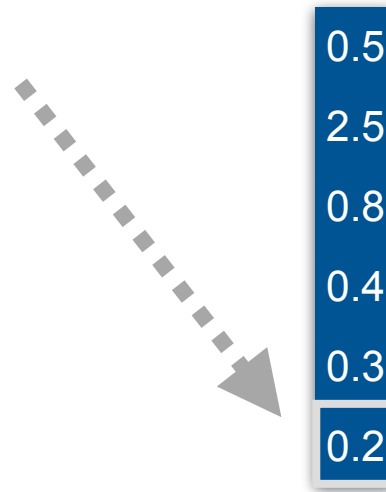
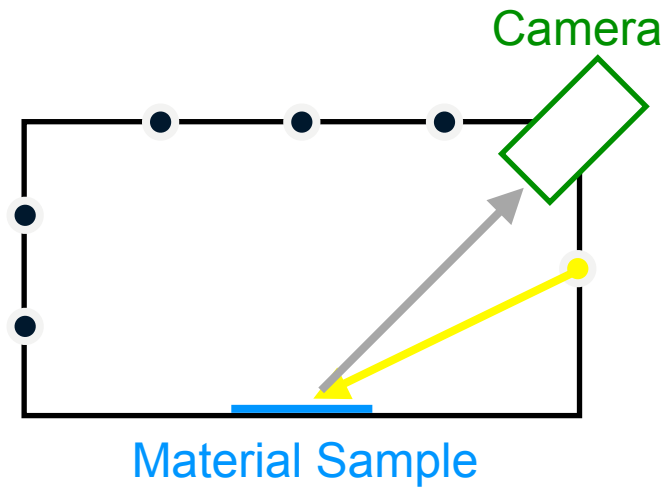
Lumitexel

# LUMITEXEL



Lumitexel

# LUMITEXEL



Lumitexel

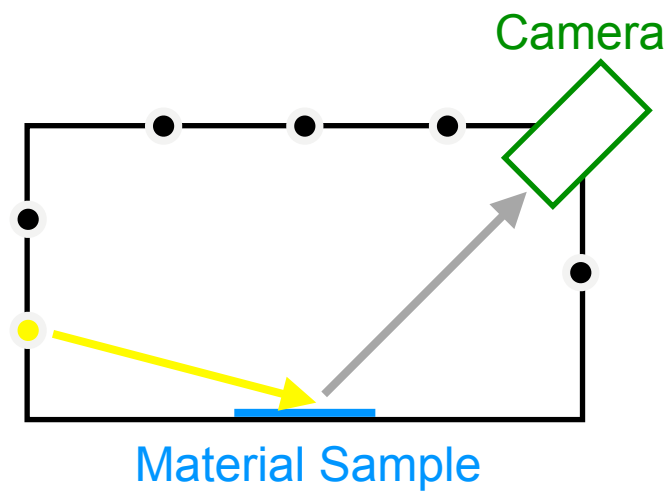
Most Informative

Time Consuming

1. Many Lights

2. Long Exposure

# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

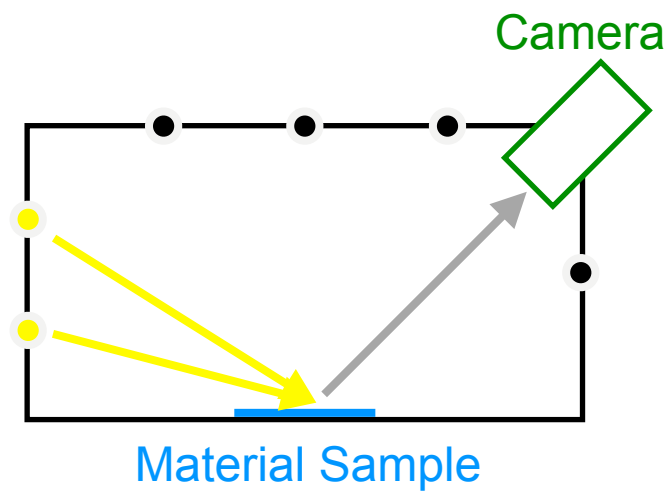


Lumitexel



Lighting  
Pattern

# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

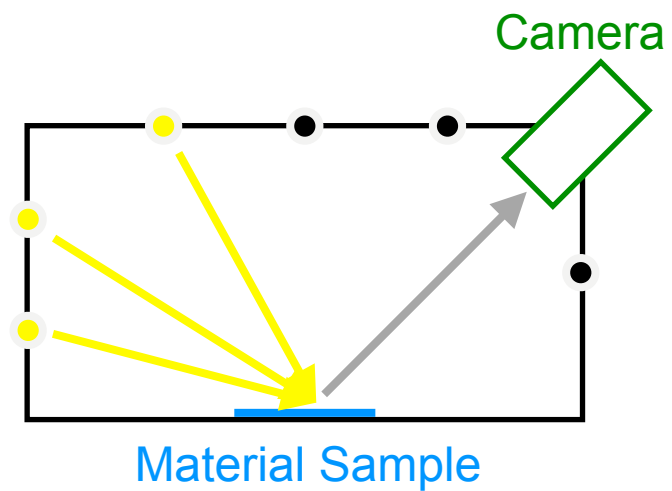


Lumitexel

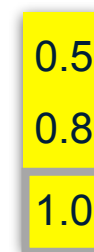


Lighting  
Pattern

# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

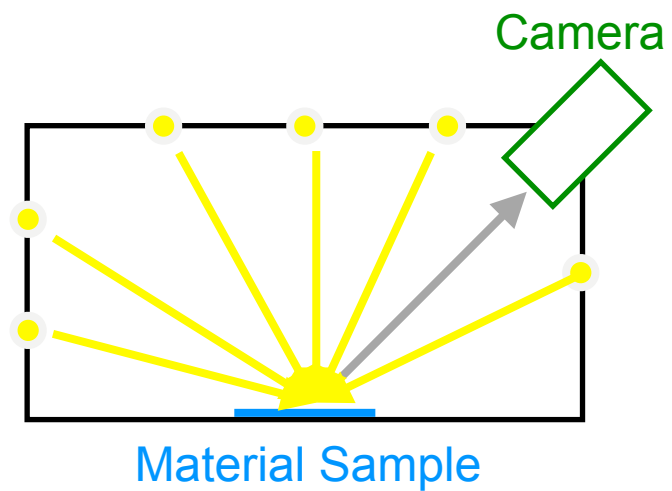


Lumitexel



Lighting  
Pattern

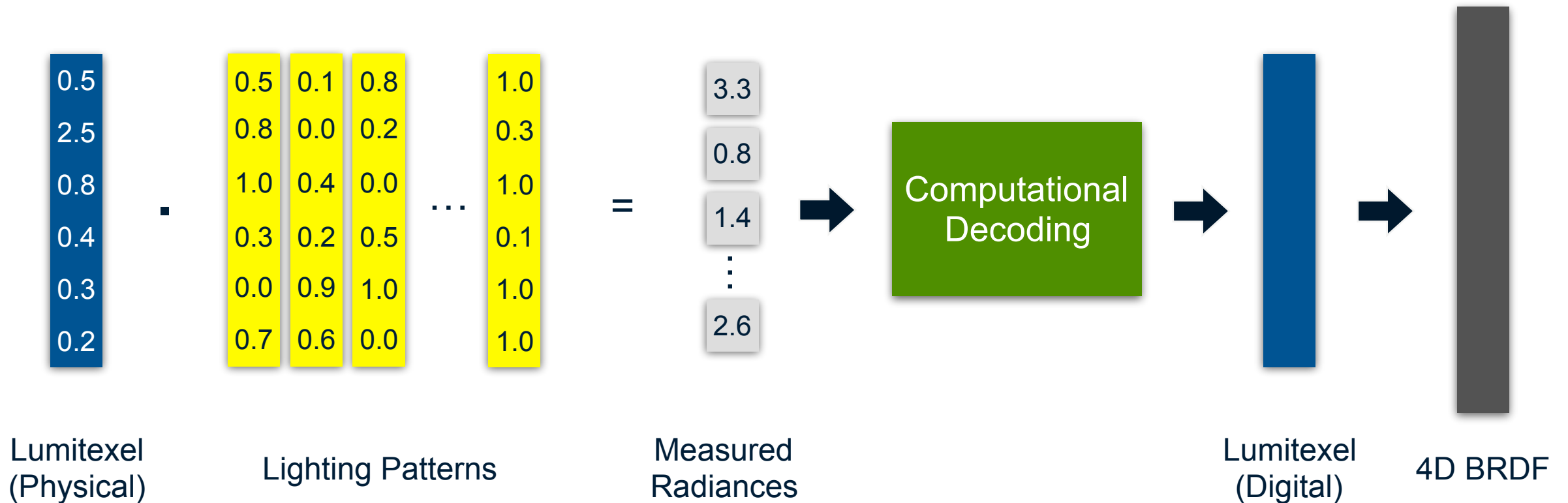
# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



0.5	0.5	=	3.3
2.5	0.8		
0.8	1.0		
0.4	0.3		
0.3	0.0		
0.2	0.7		
Lumitexel	Lighting Pattern		Measured Radiance

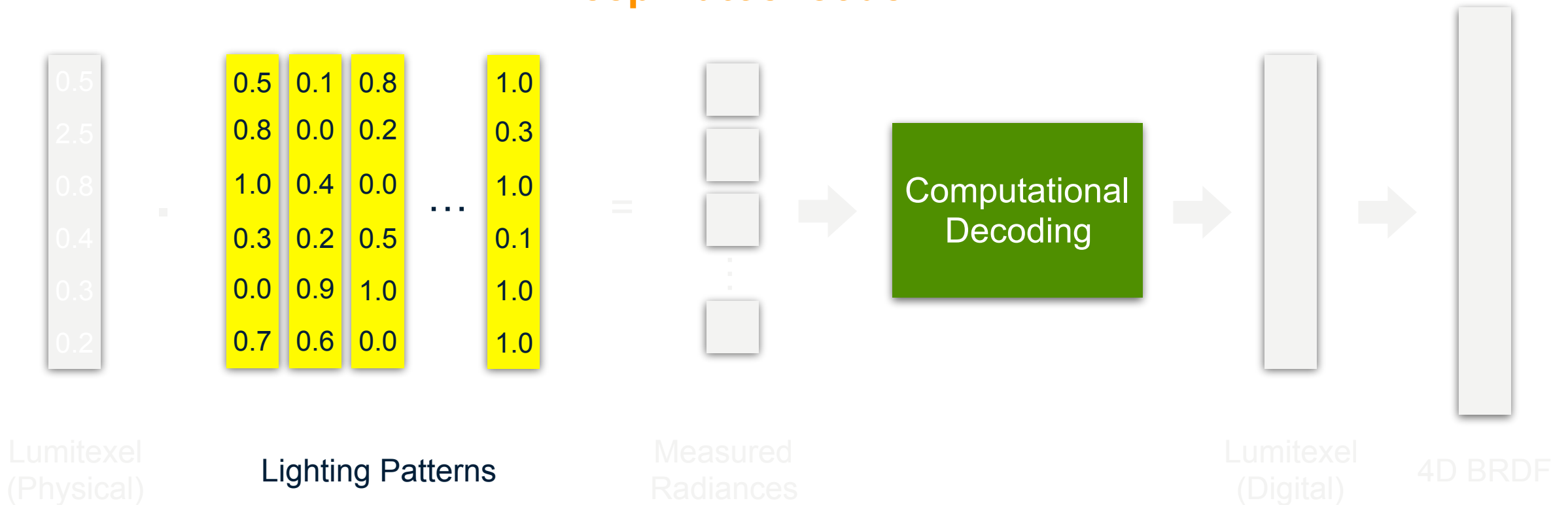


# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION



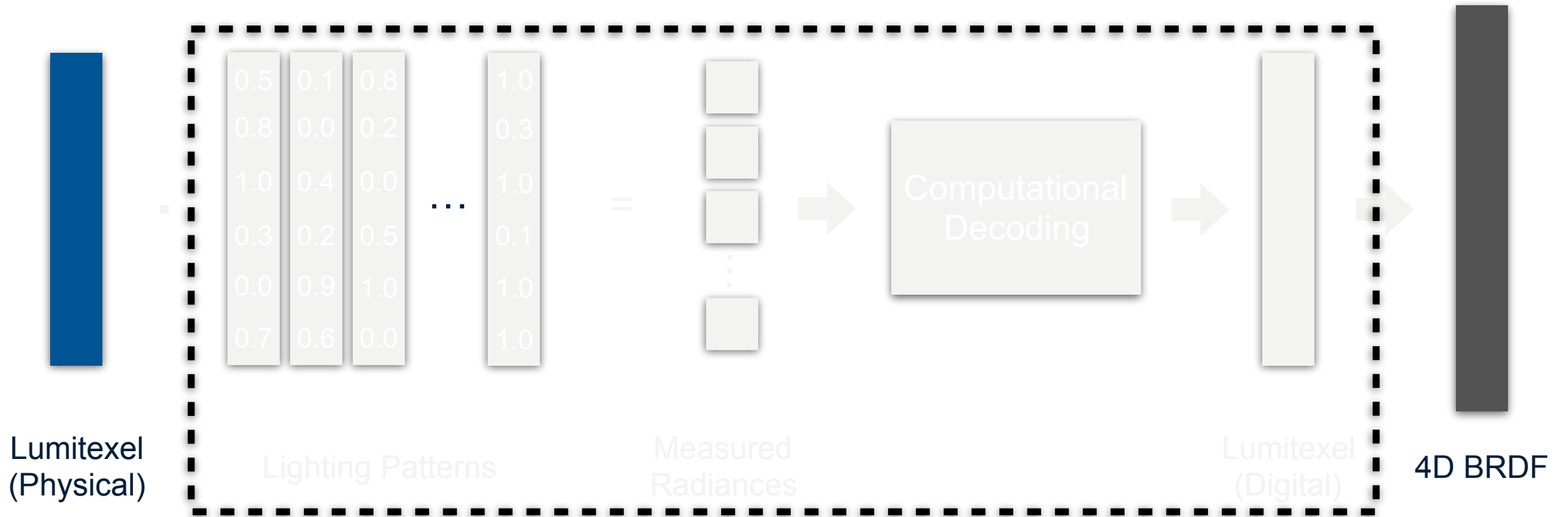
# LIGHTING-PATTERN-BASED REFLECTANCE ACQUISITION

## Deep Autoencoder



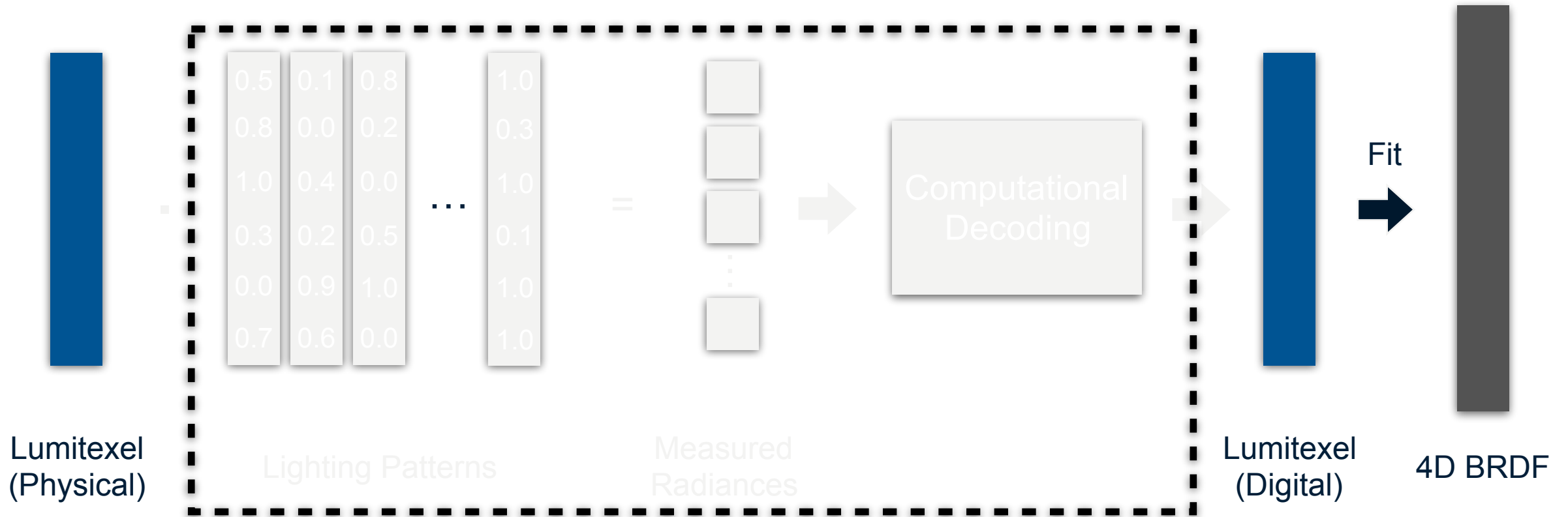
# WHAT TO LEARN

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



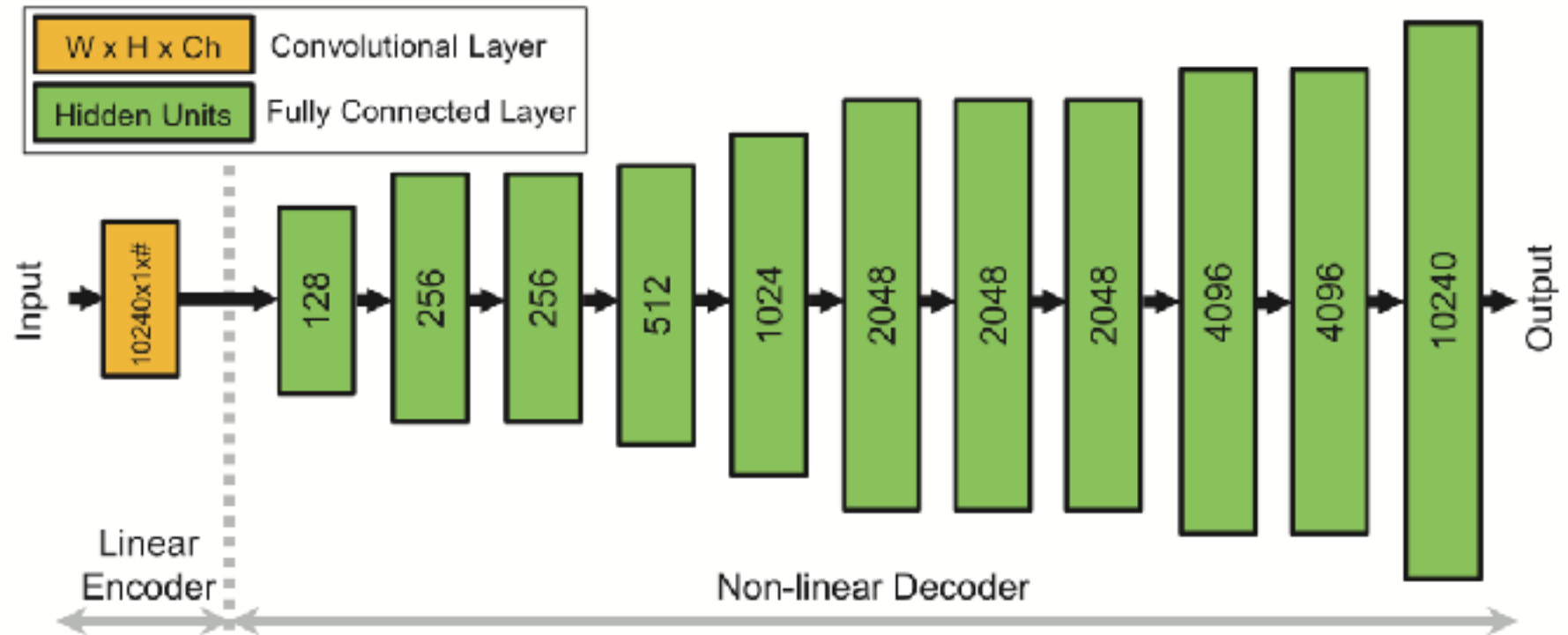
# WHAT TO LEARN

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

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

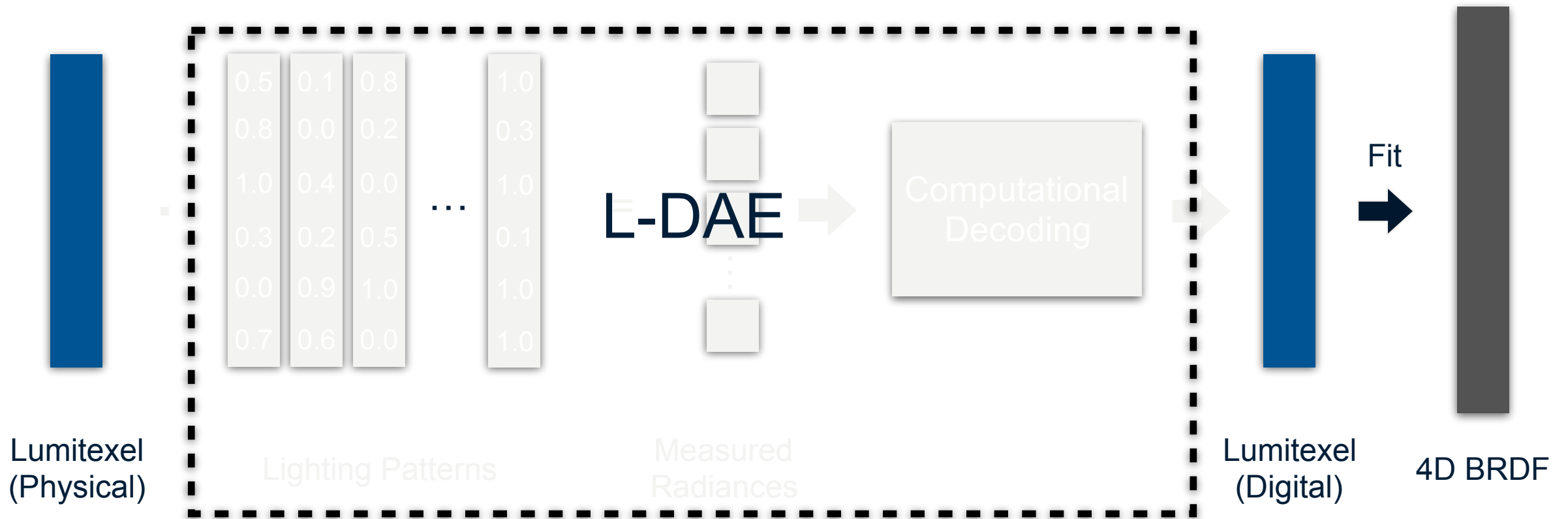
- Reconstruction Error Term  $L_{\text{auto}}(m) = \sum_j [\log(1 + m(j)) - \log(1 + m_{\text{gt}}(j))]^2.$
- Encoder Plausibility Term  $L_{\text{barrier}}(w) = \tanh\left(\frac{w - (1 - \epsilon)}{\epsilon}\right) + \tanh\left(\frac{-w + \epsilon}{\epsilon}\right) + 2.$

# 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

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

# Results

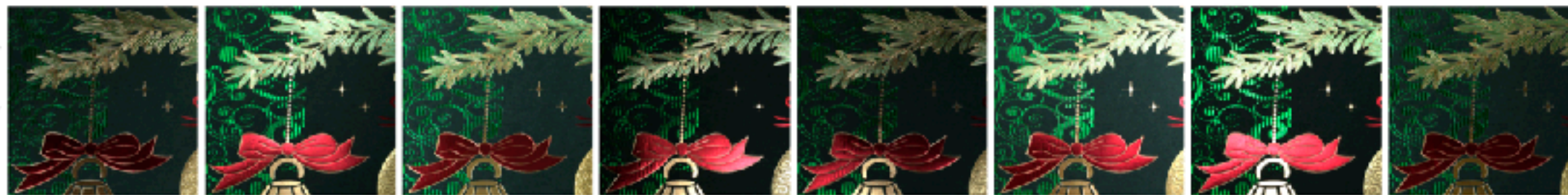
# TIMING

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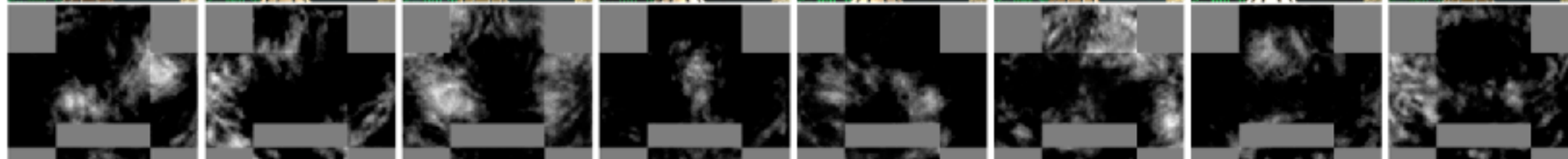
- Training Time: 5h
- Encoding Time: 12~25s
- Decoding Speed: 4m / 1M Lumitexels
- Fitting Speed: 1.6h / 1M Lumitexels

# LIGHTING PATTERNS

Sample Lit with Our  
Patterns (Aniso.)



Ours (Aniso.)



Ours (Iso.)



PCA

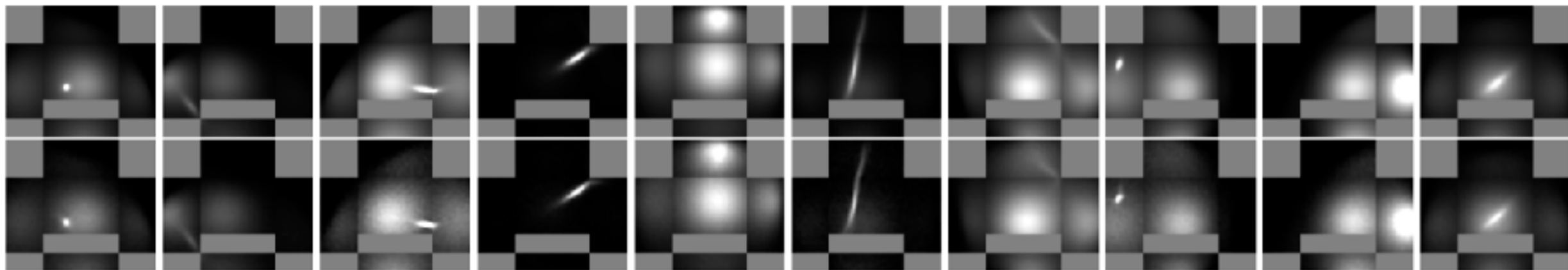


# SYNTHETIC LUMITEXEL RECONSTRUCTION

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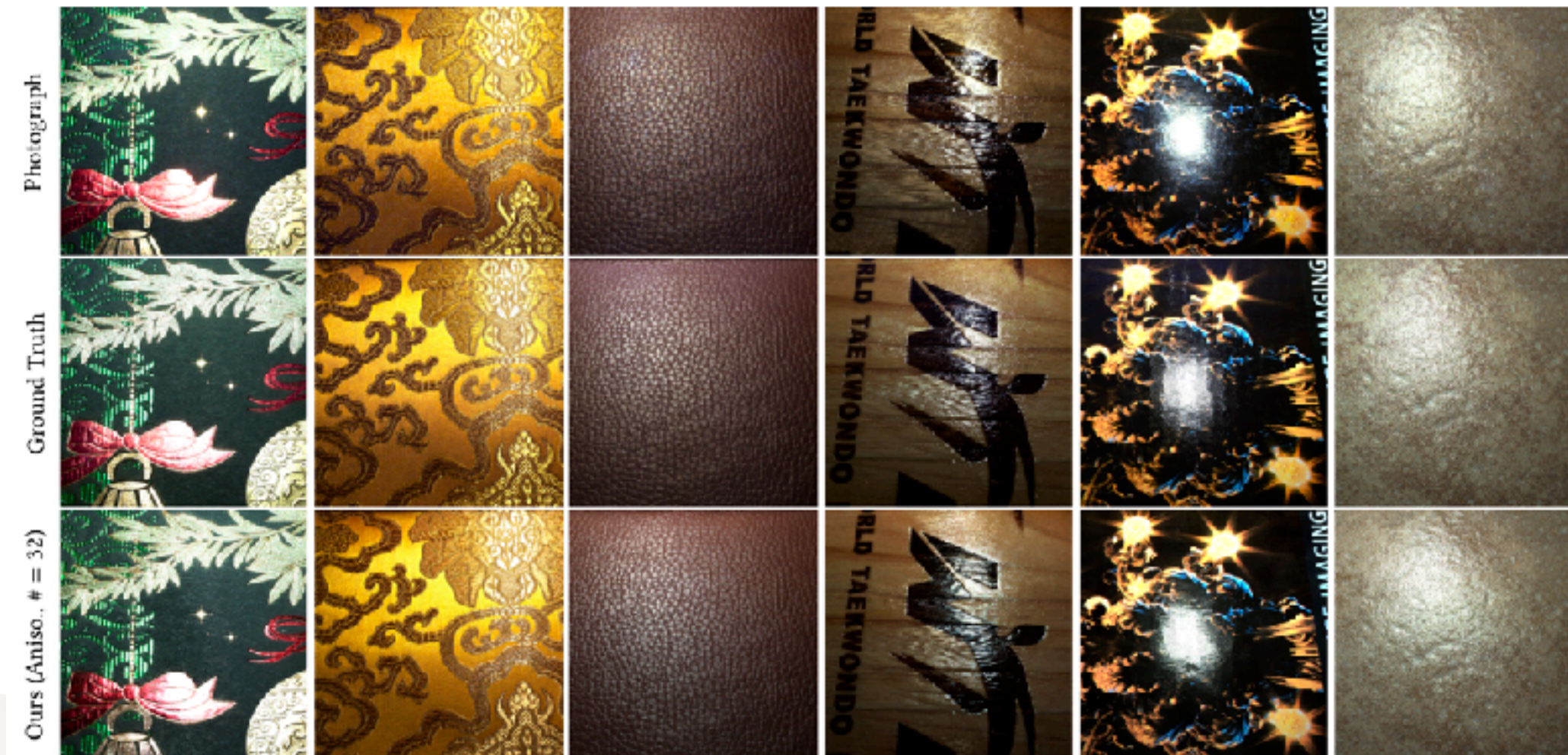
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Input Lumitexels



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

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Xmas-card



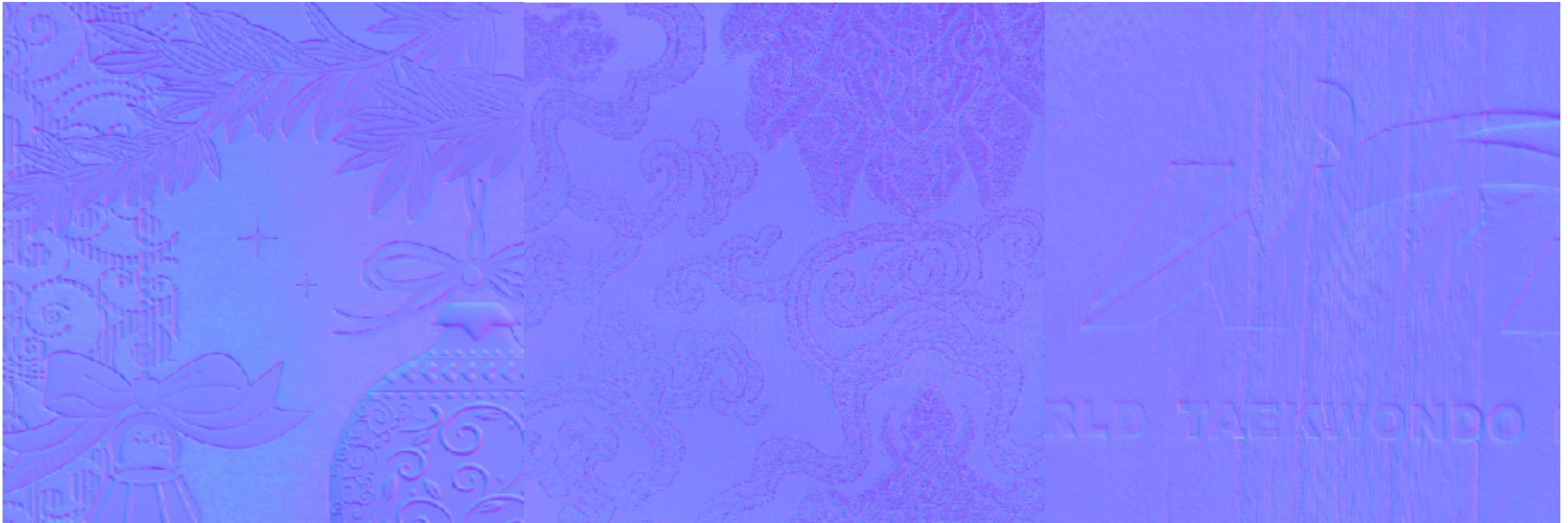
Fabric



Wood

# NORMAL

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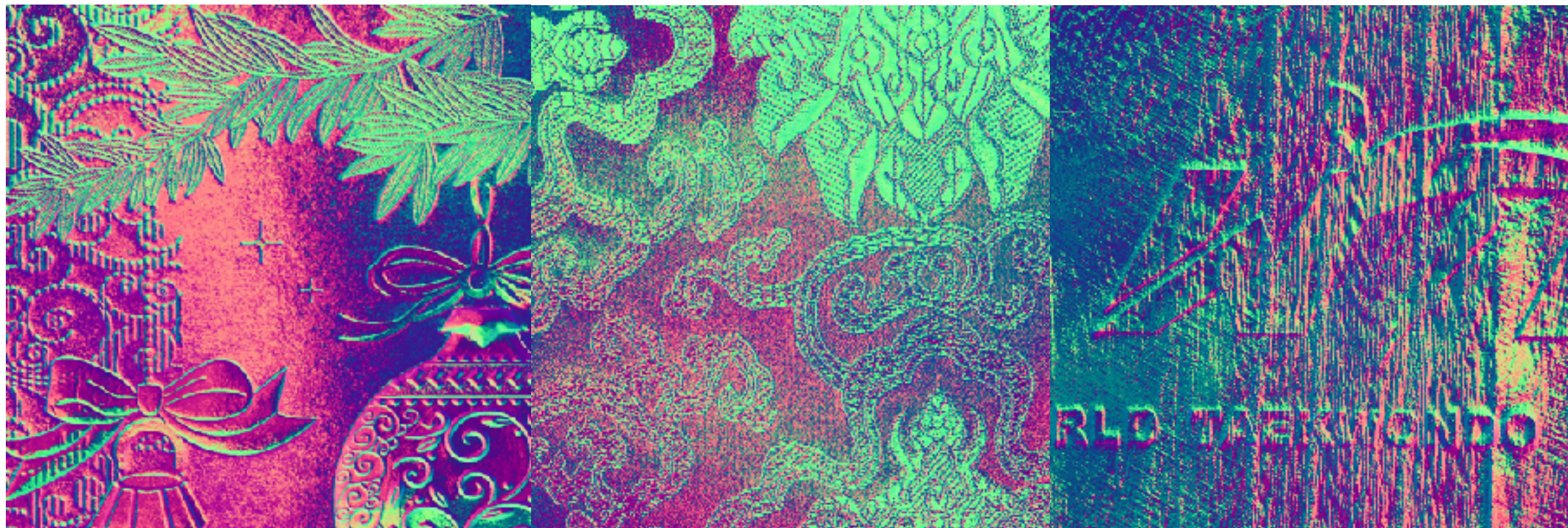
Xmas-card

Fabric

Wood

# TANGENT

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Xmas-card

Fabric

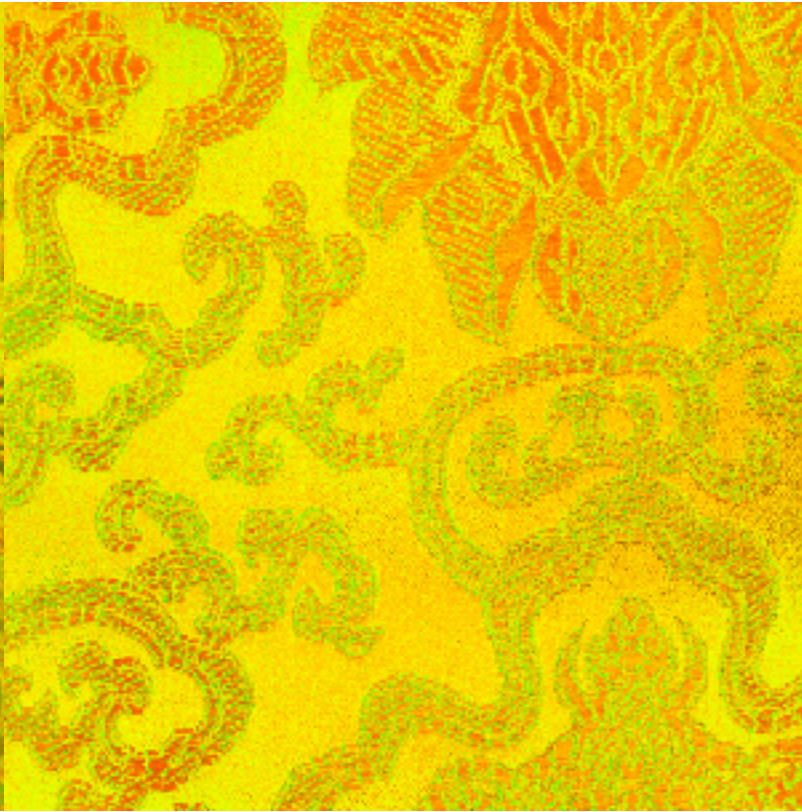
Wood

# ROUGHNESS

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Xmas-card

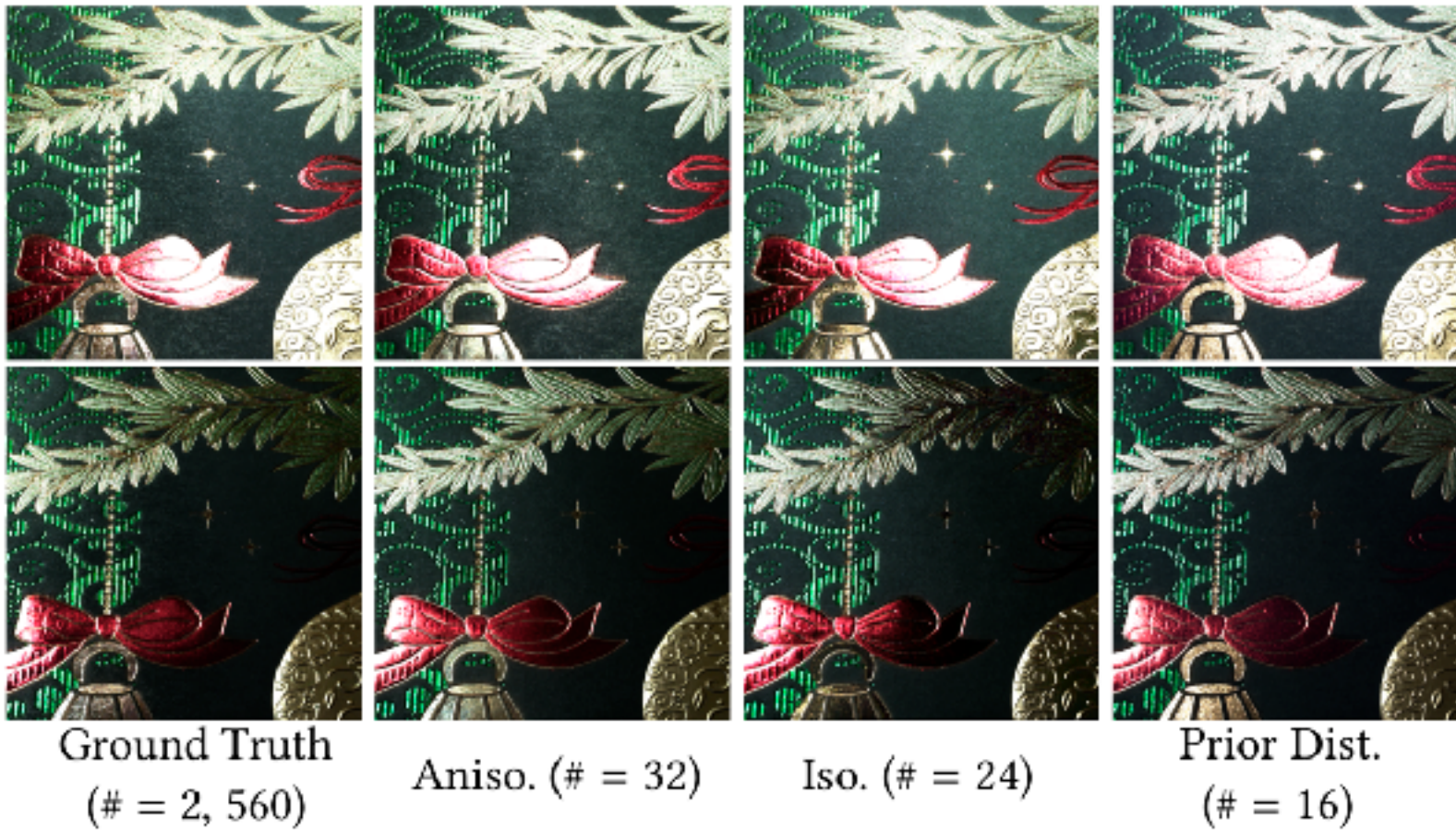


Fabric



Wood

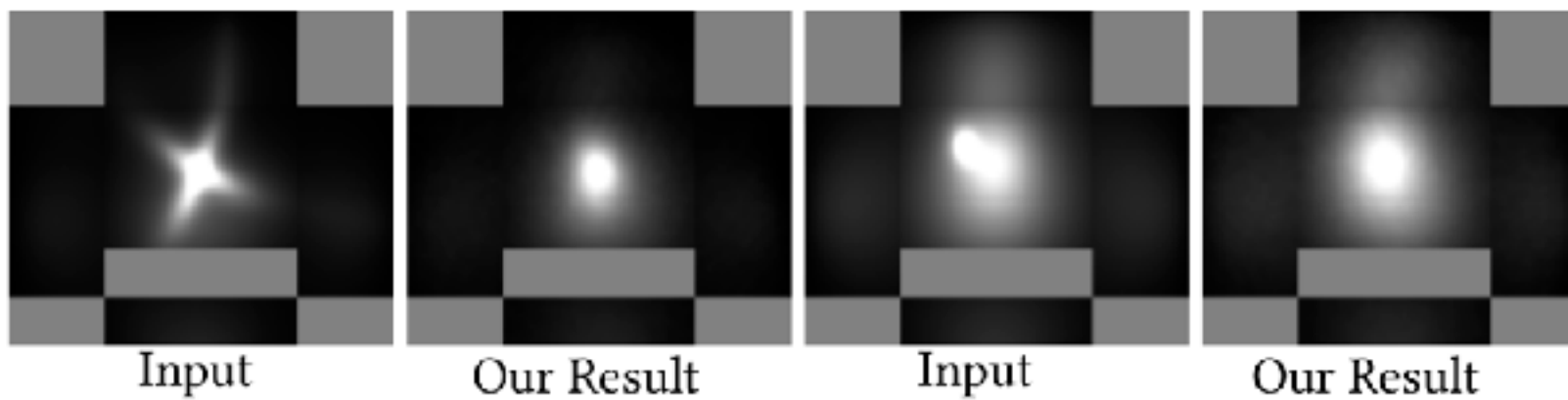
# IMPACT OF TRAINING DATA



# LIMITATIONS

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- Cannot Reconstruct Lumitexels Substantially Deviated from Training Data



# LIMITATIONS

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- 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

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

# CONCLUSIONS & FUTURE WORK

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- 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])

# ACKNOWLEDGEMENTS

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謝謝 / MERCI

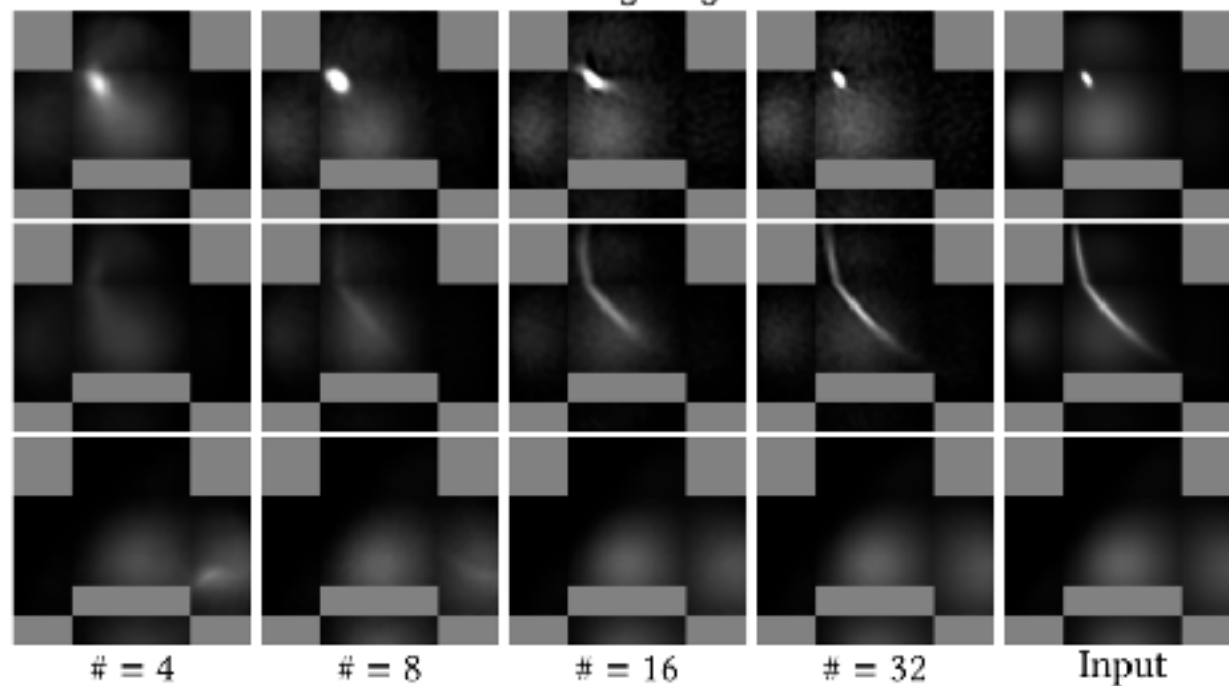
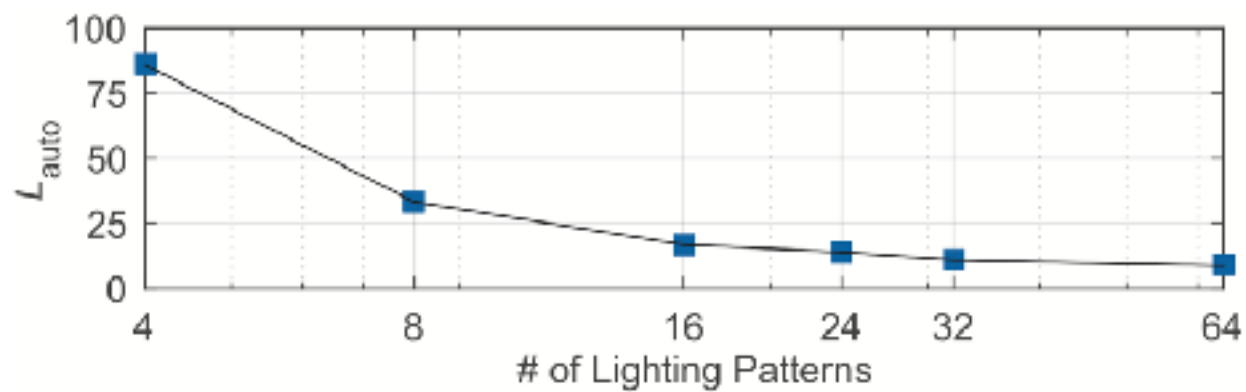
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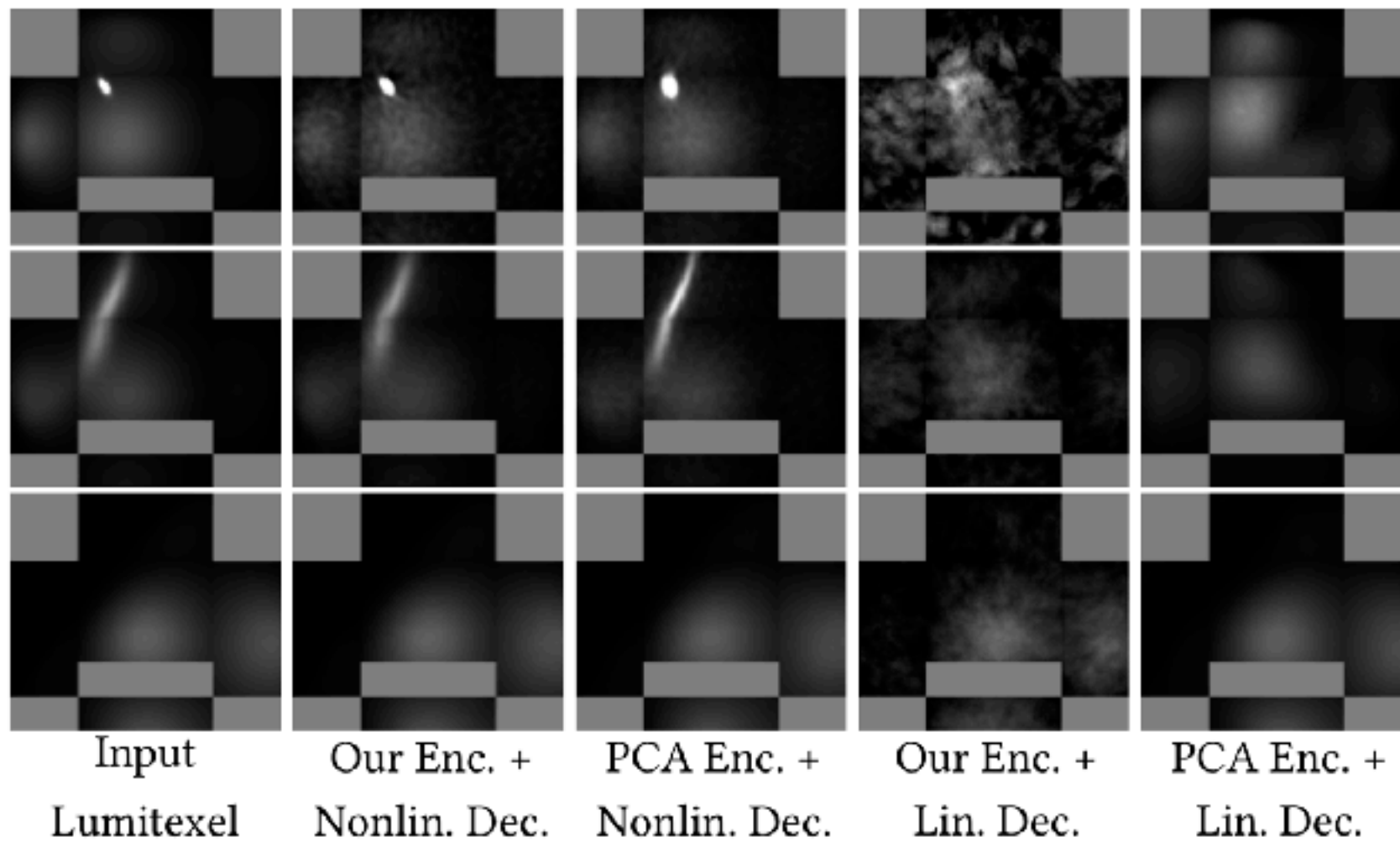
Project Webpage



# BACKUP



# BACKUP



# BACKUP

## Reconstruction Results

## Fitting Results

