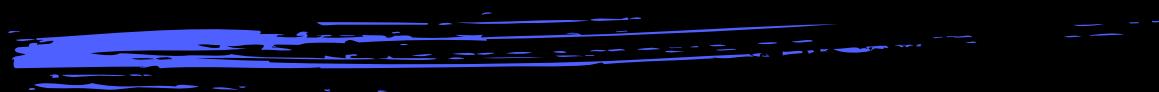


Computational Imaging



Lecture 24: End-to-end Computational Camera Design



点昀
POINT SPREAD



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点昀技术 (Point Spread Technology)



香港中文大學(深圳)
The Chinese University of Hong Kong, Shenzhen



Today's Topic

- Step One: Joint Optimization of Optics and Image Recovery
 - Learned Large Field-of-View Imaging with Thin-Plate Optics
- Step Two: Compressive and Optically Coded Super-resolution SPAD Camera
 - Compressive and Optically Coded Super-resolution SPAD Camera
- Step Three: Enabling Comprimization for Color Channels
 - Learning Rank-1 Diffractive Optics for Single-shot High Dynamic Range Imaging
- Step Four: End-to-end Complex Lens Design
 - End-to-end Complex Lens Design with Differentiable Ray-Tracing

Step One: Joint Optimization of Optics and Image Recovery

Learned Large Field-of-View Imaging
with Thin-Plate Optics

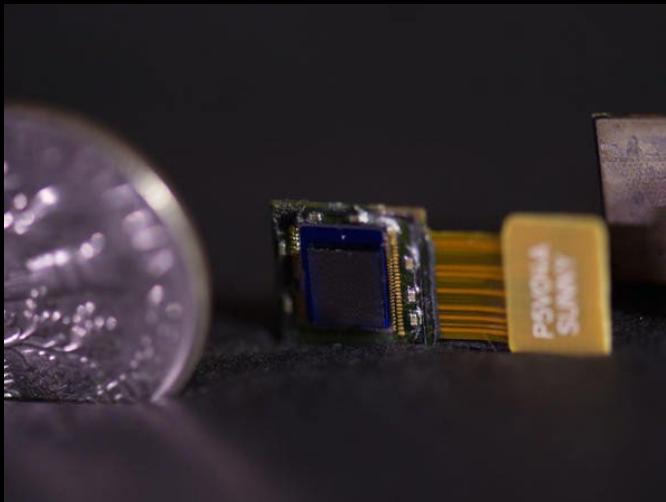
Thin-Plate Optics

- Diffractive optics
- Metasurfaces

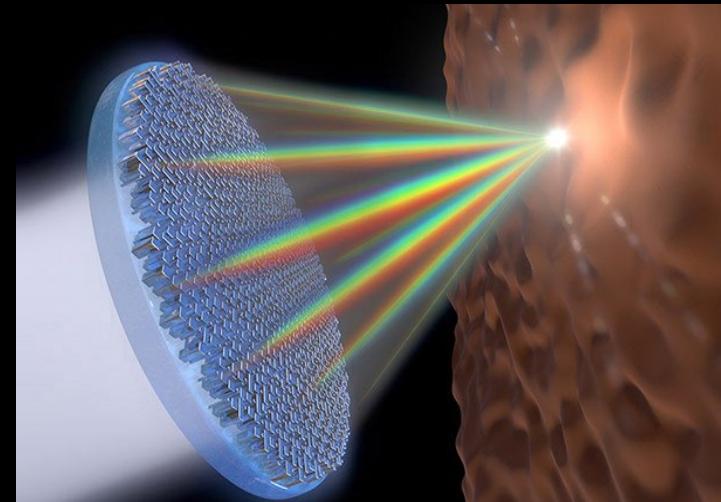
- Thin form factor
- Large design freedom

- Limited field-of-view
- Limited color fidelity
- Limited lens size

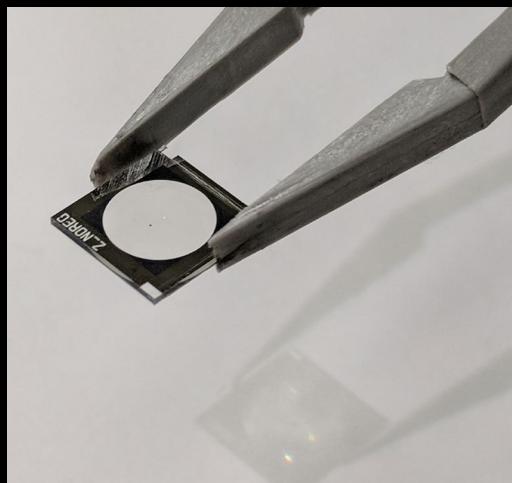
This work



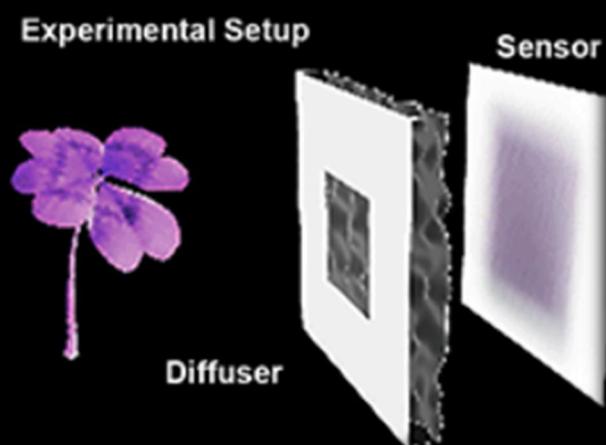
[Ayremlou et al. 2016]



[Capasso et al. 2017]



[Peng et al. 2016]
[Sitzmann et al. 2018]

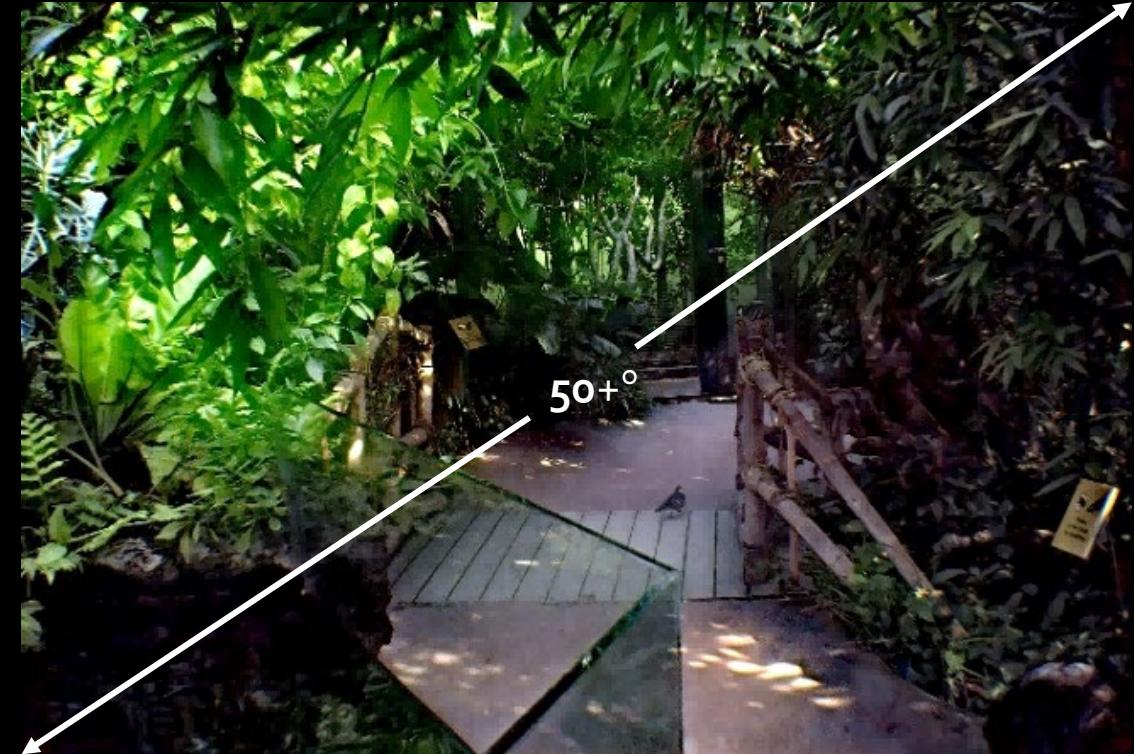


[Antipa et al. 2018]

Wide Field-of-View Imaging

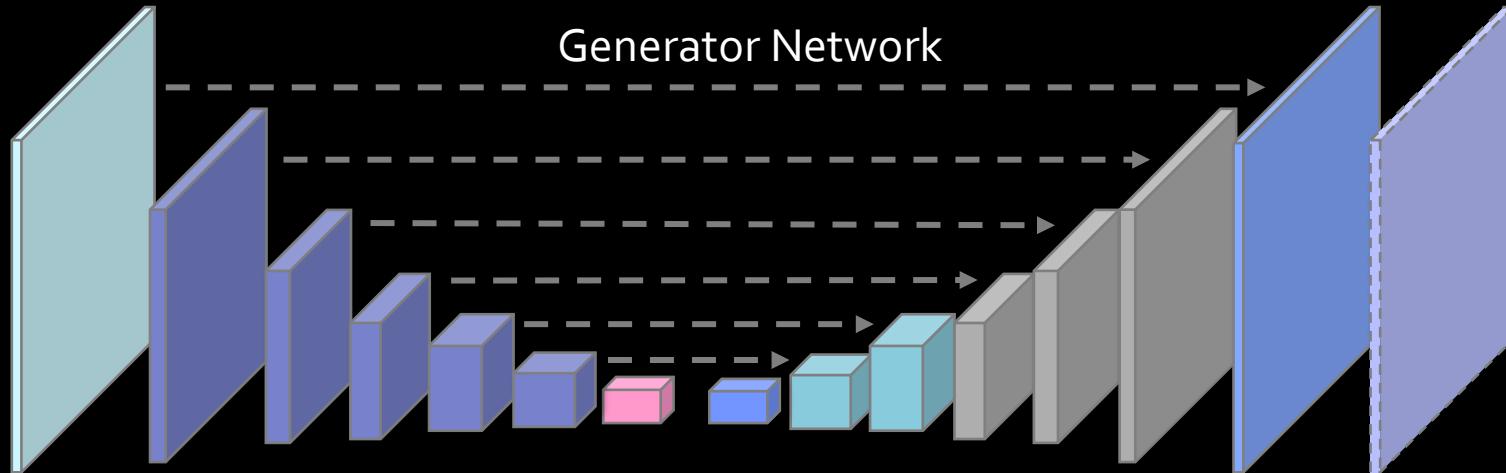
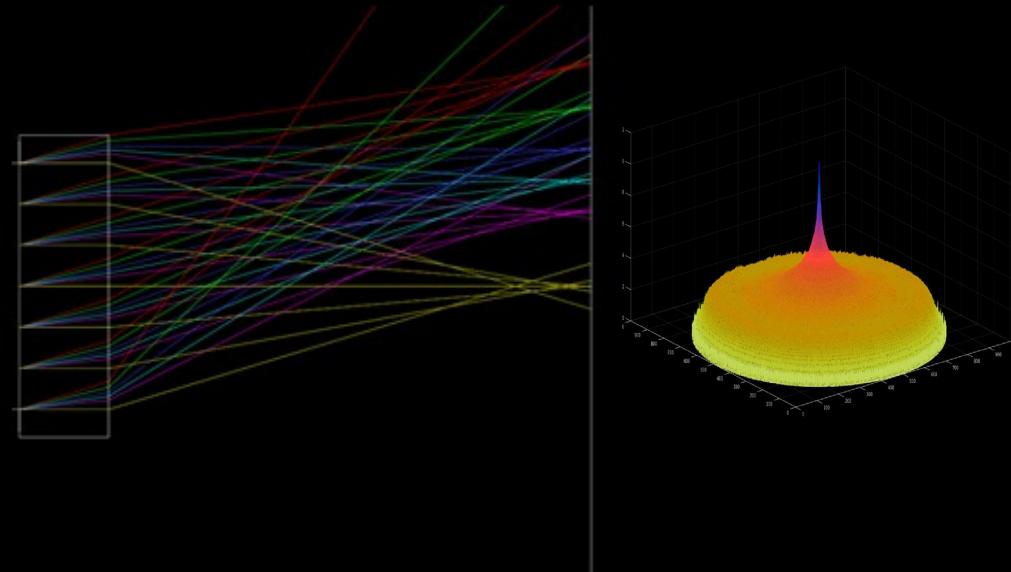


[Peng et al. 2016]
[Heide et al. 2016]
[Ayremlou et al. 2016]
[Sitzmann et al. 2018]

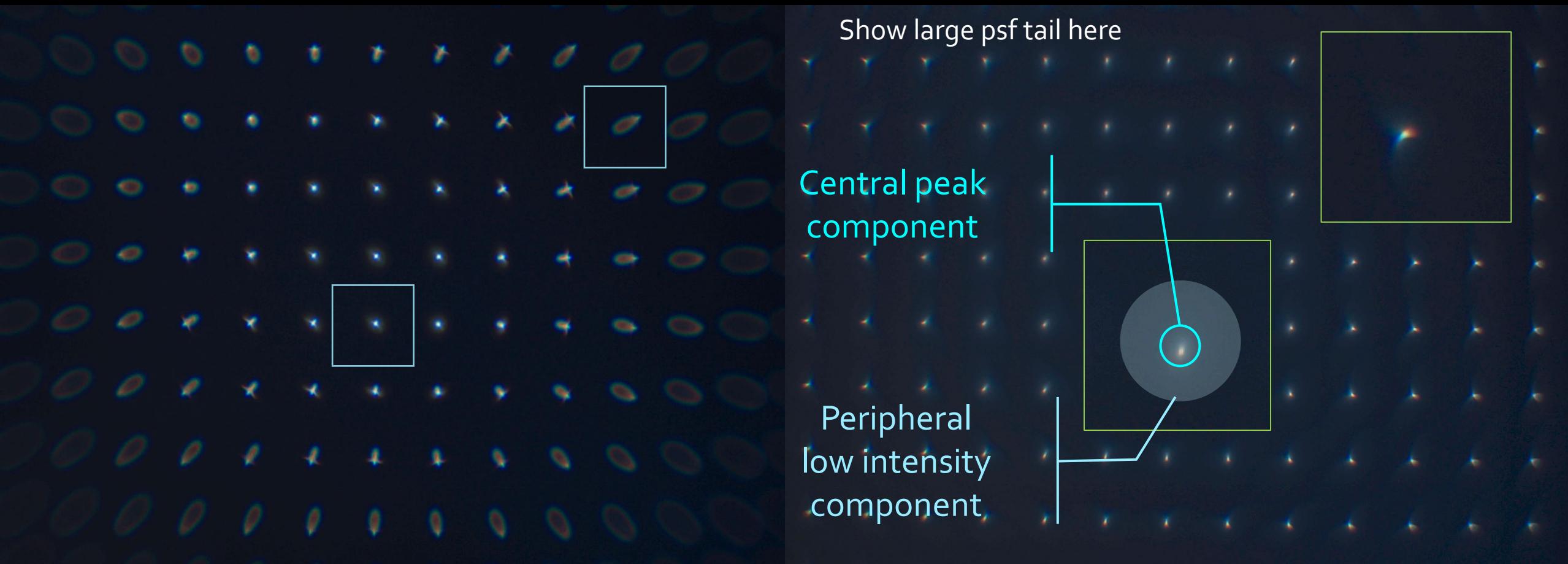


Co-design of Optics and Learned Recovery

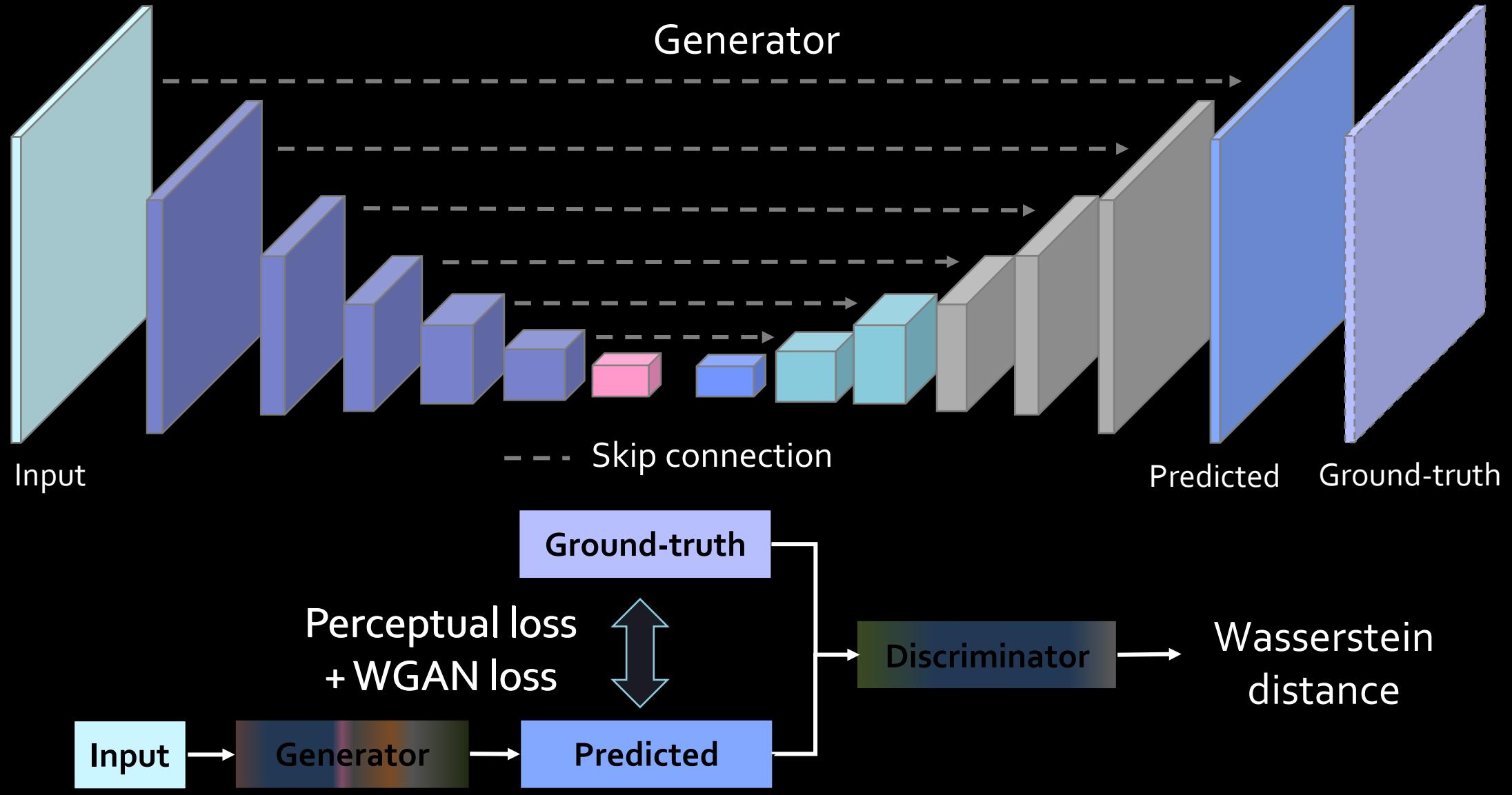
Co-design of Optics and Learned Recovery



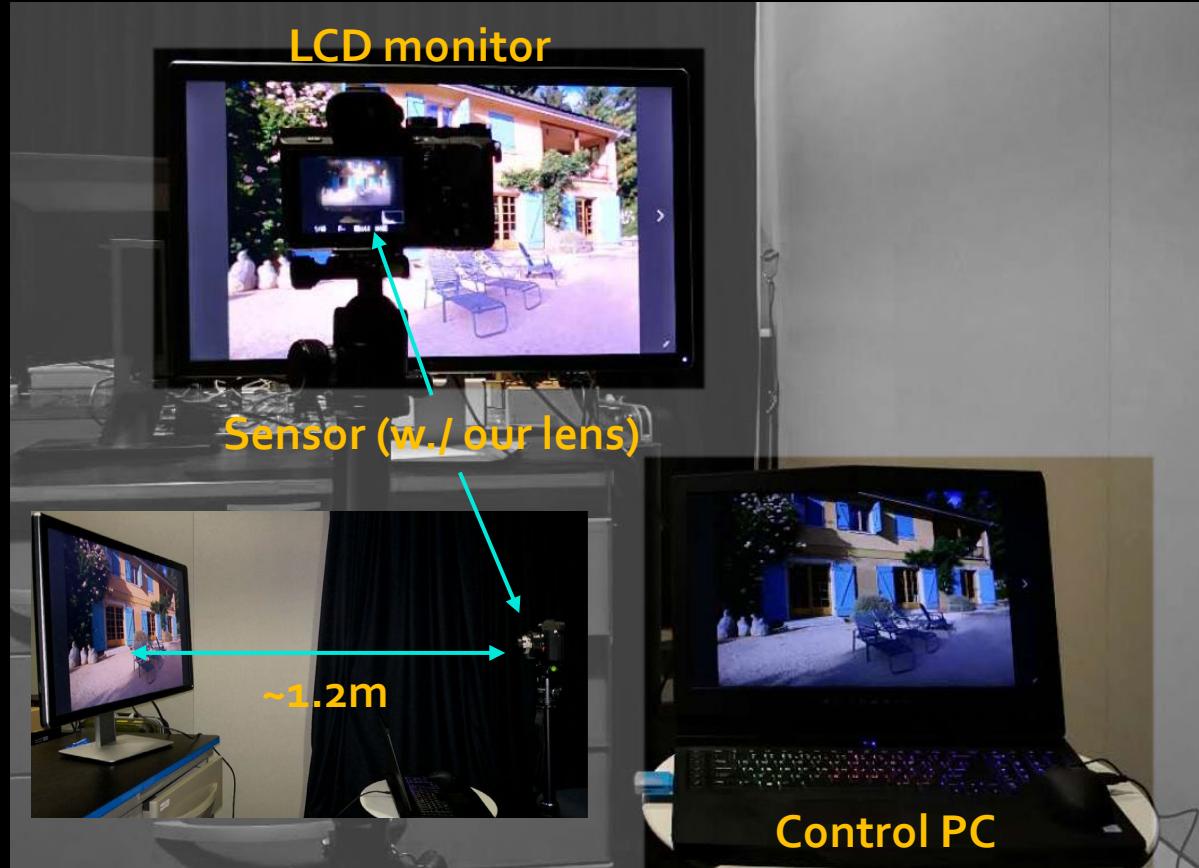
Angularly-Invariant PSF



Learned Reconstruction Framework

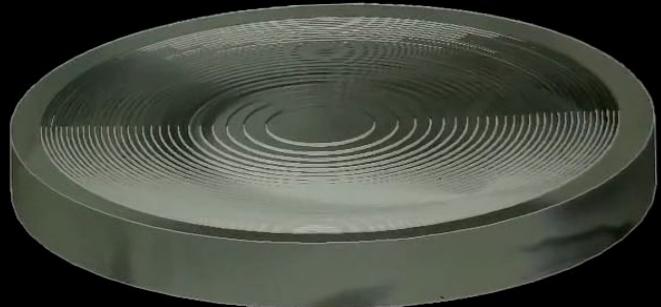


Automated Training Data Acquisition

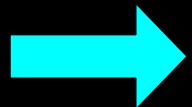
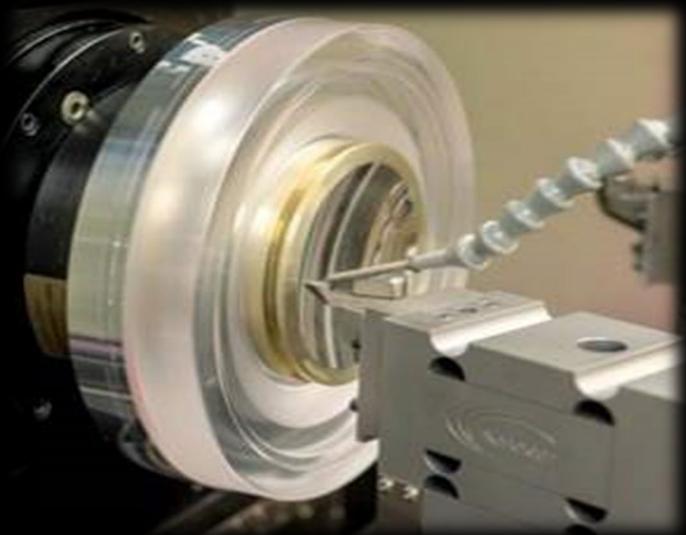


814 image pairs in total

Prototypes



- Clear aperture $D : 23.4 \text{ mm}$
- Focal length $f : 43.0 \text{ mm}$
- Field of view : $50+$ deg.



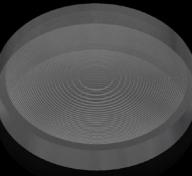
10 mm substrate

3 mm substrate



Remarks

➤ Large field-of-view + Single element



➤ Display-capture lab training



➤ High-quality color imaging



Step Two: The First End-to-end Optics for Single Wavelength

Compressive and Optically Coded Super-resolution SPAD Camera

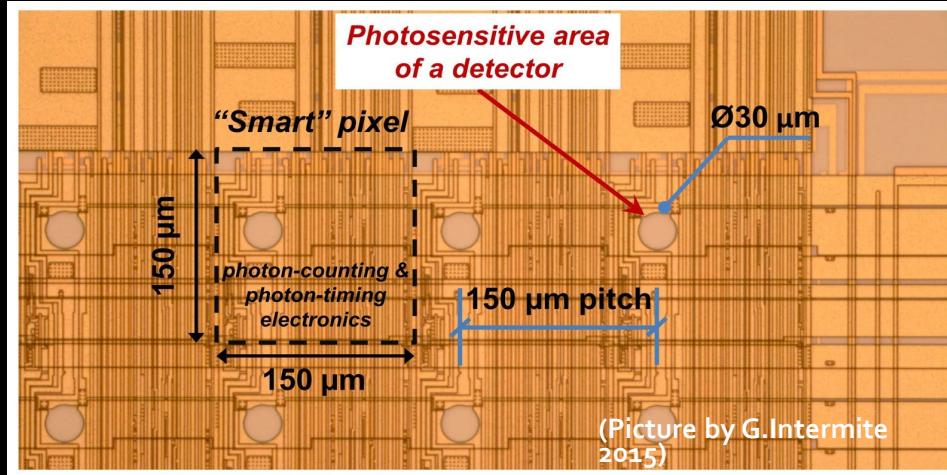
Motivation

SPAD array sensor



(Picture from MPD)

- Single photon level sensitivity
- Pico-second level time resolution
- Low spatial resolution
- Low fill-factor



Low fill-factor and LR image



Sampling HR image with low ratio

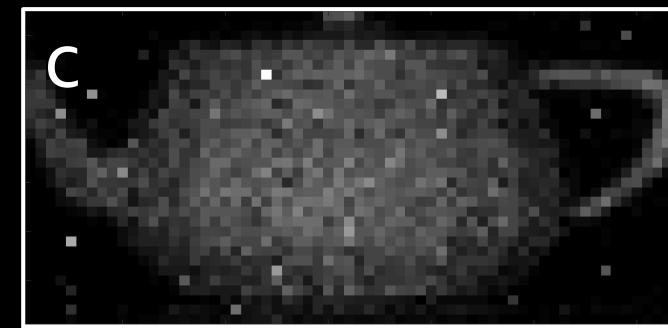
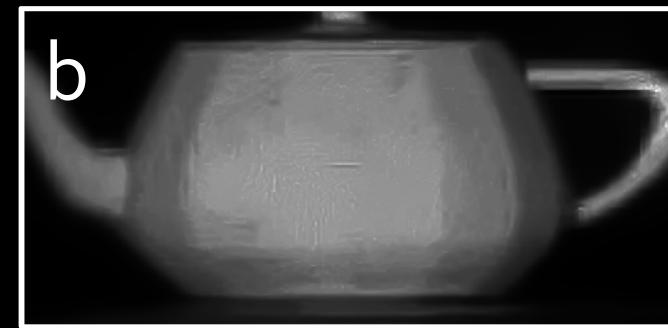
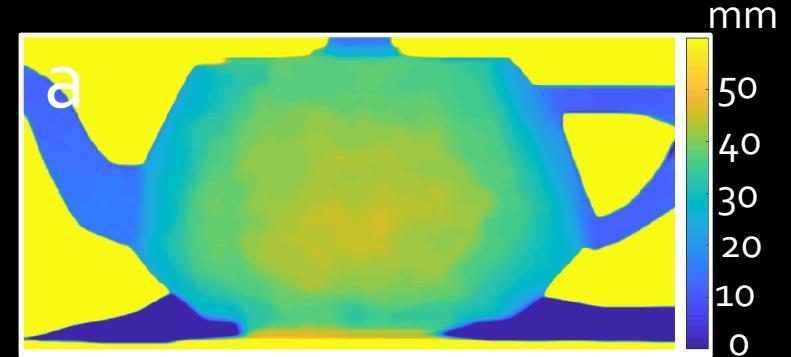
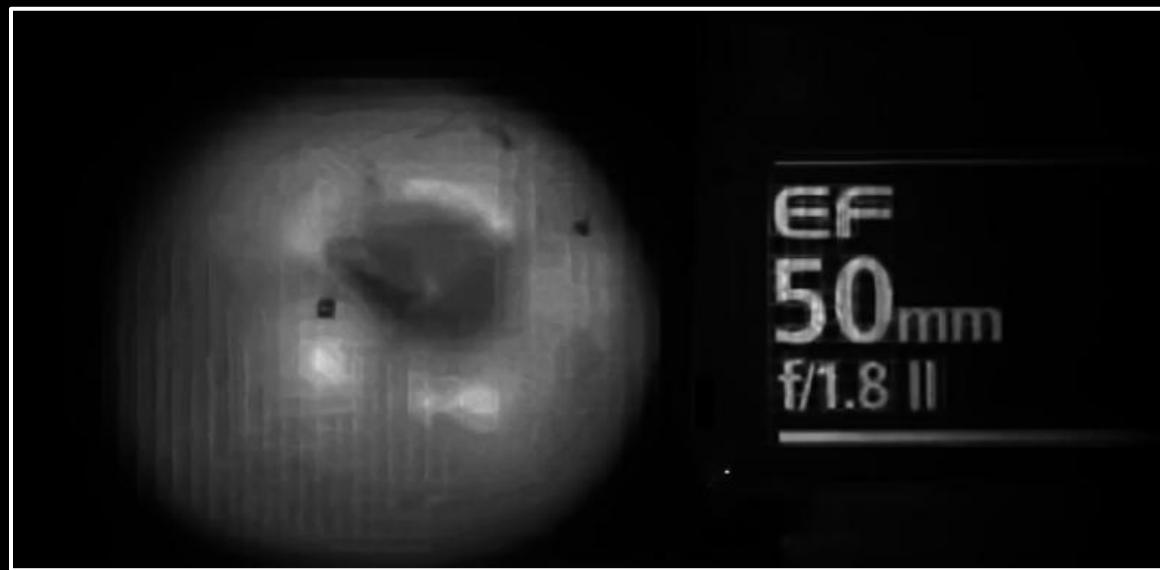
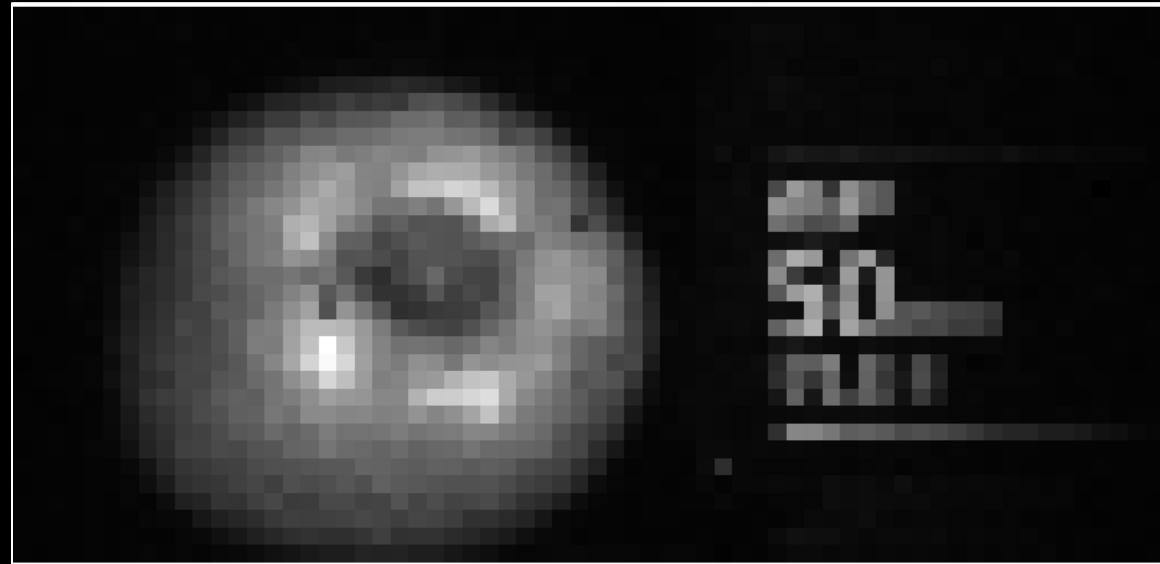
Optimal optics PSF encoding

Deep Network based SR method

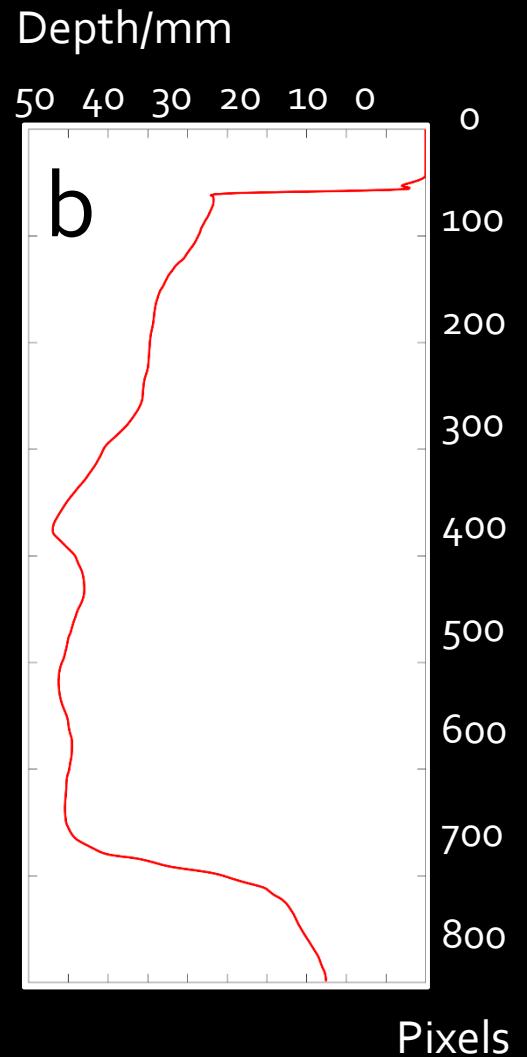
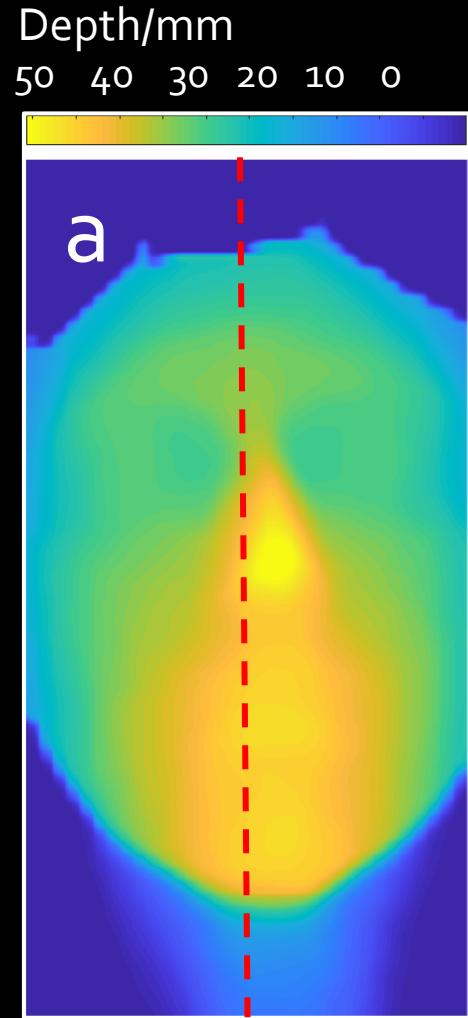


Recover HR
image with
single-shot

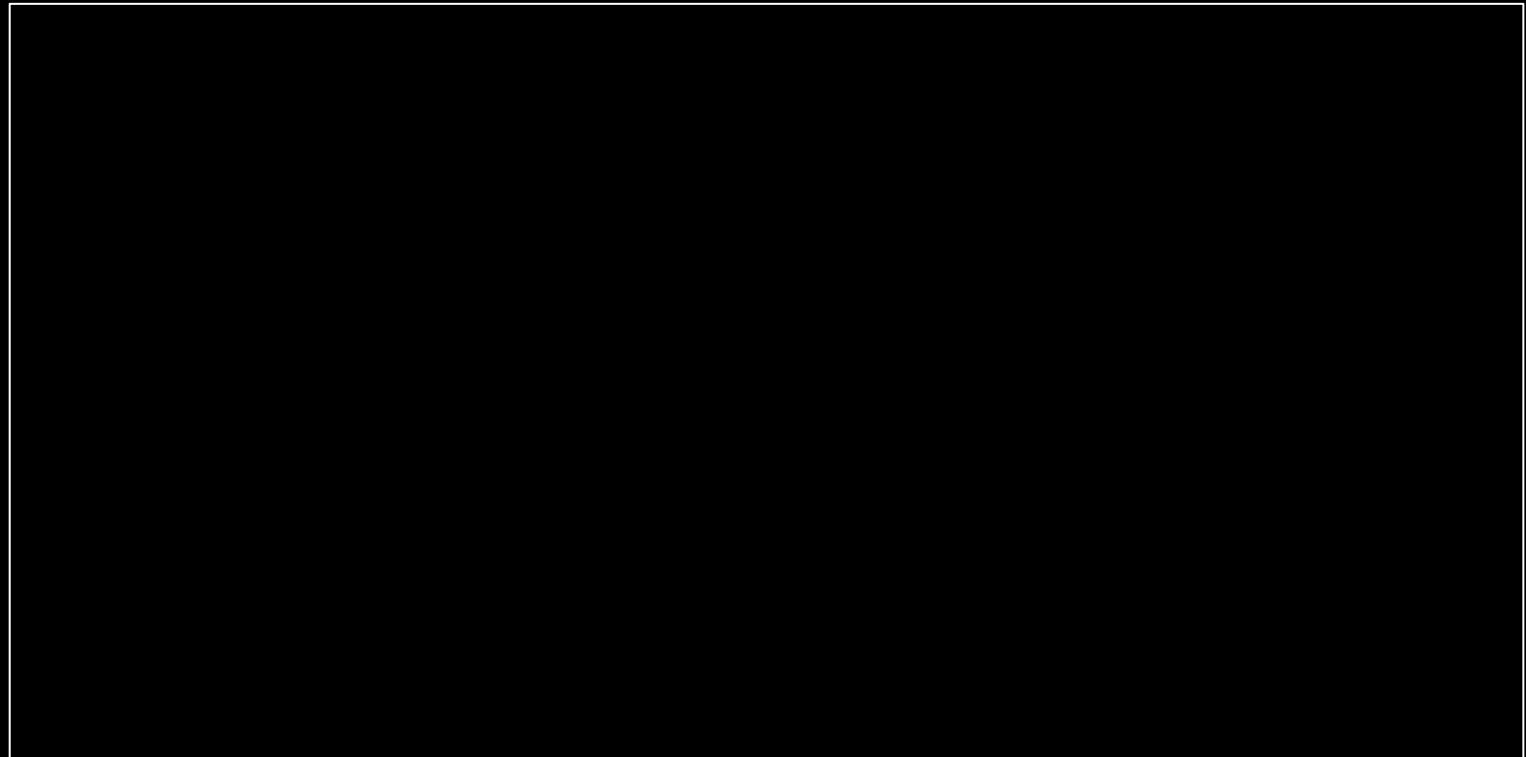
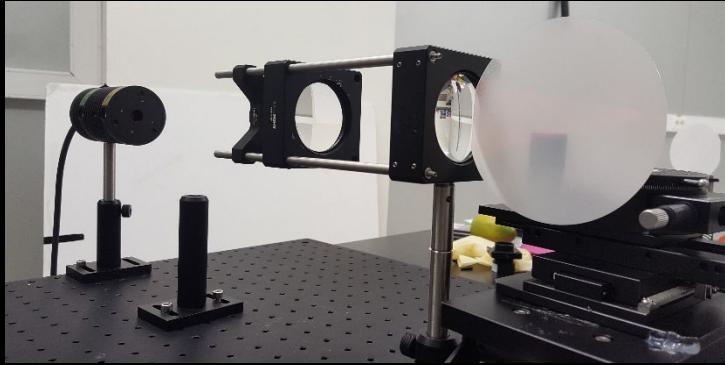
Compressive SPAD Camera—Results



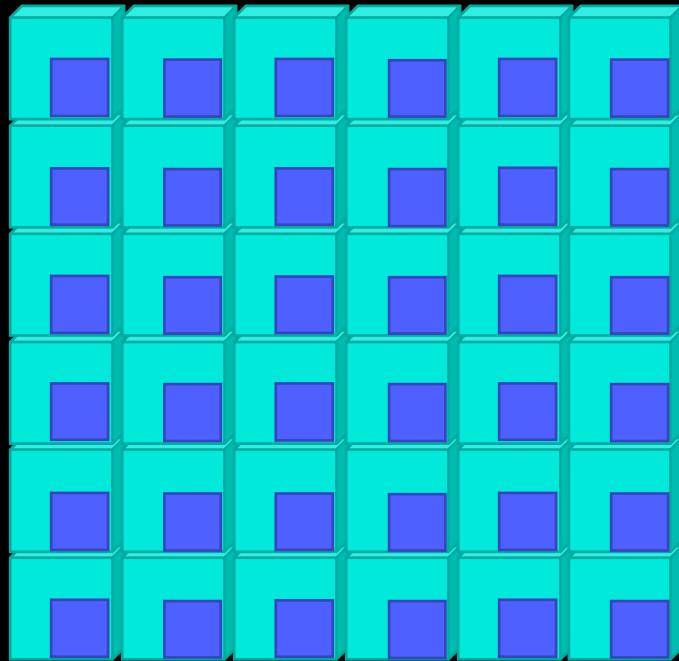
Compressive SPAD Camera—Results



Compressive SPAD Camera—Results



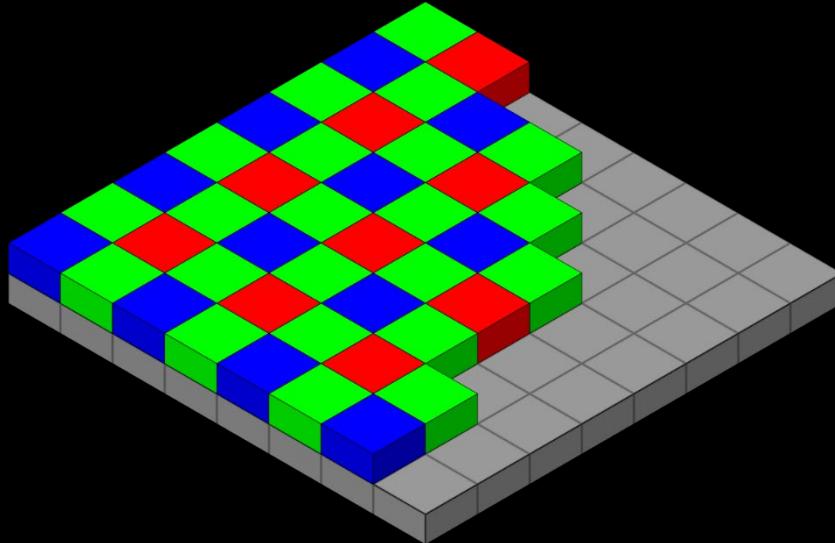
Sampling with a Physical Sensor Structure



Low fill-factor



- Aliasing artifacts
- Low efficient



Bayer filter



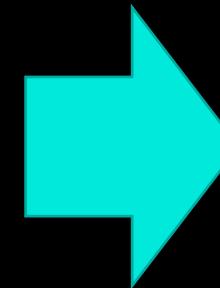
- Moiré pattern
- Mosaicing

Optimal Sampling
for

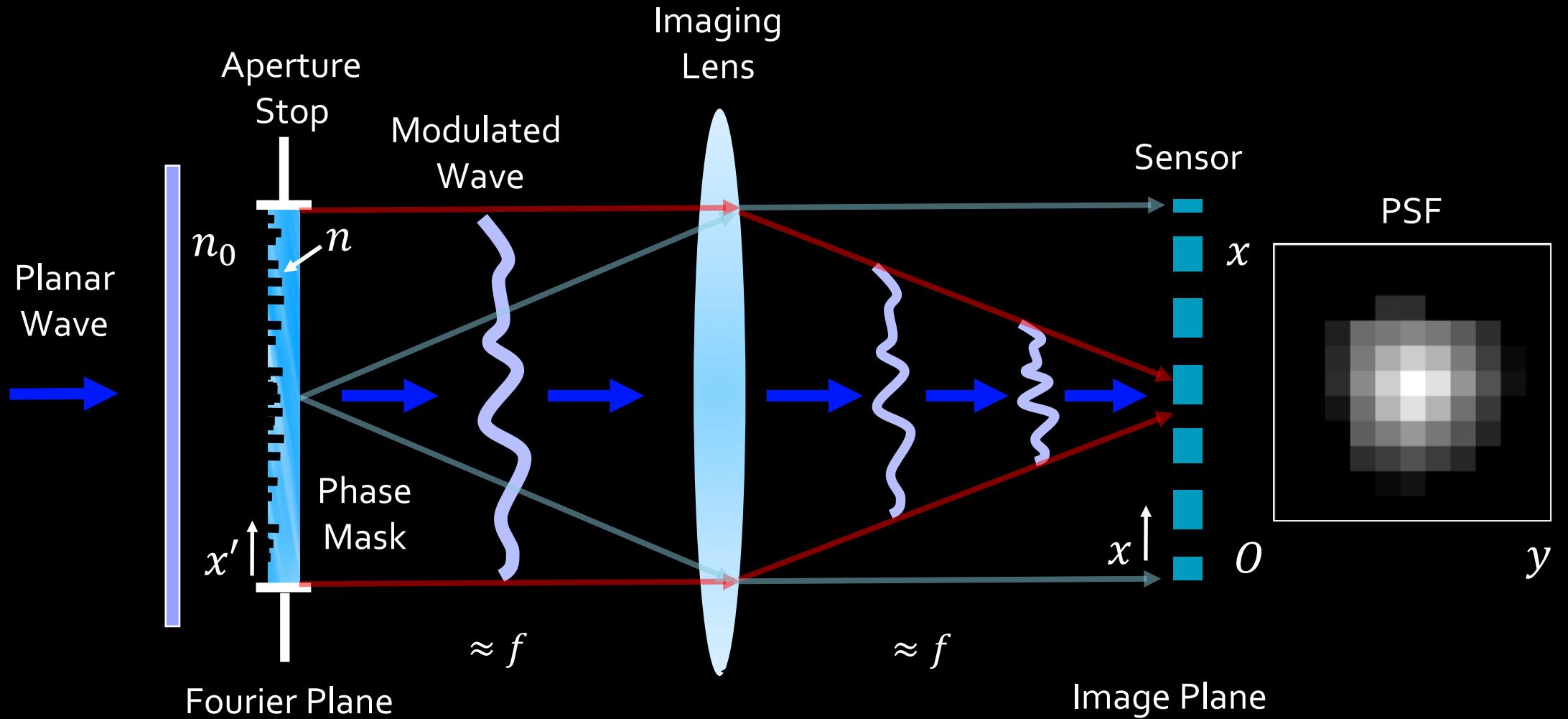
Demosaicing
Anti-aliasing
Super-resolution

⋮

Etc.



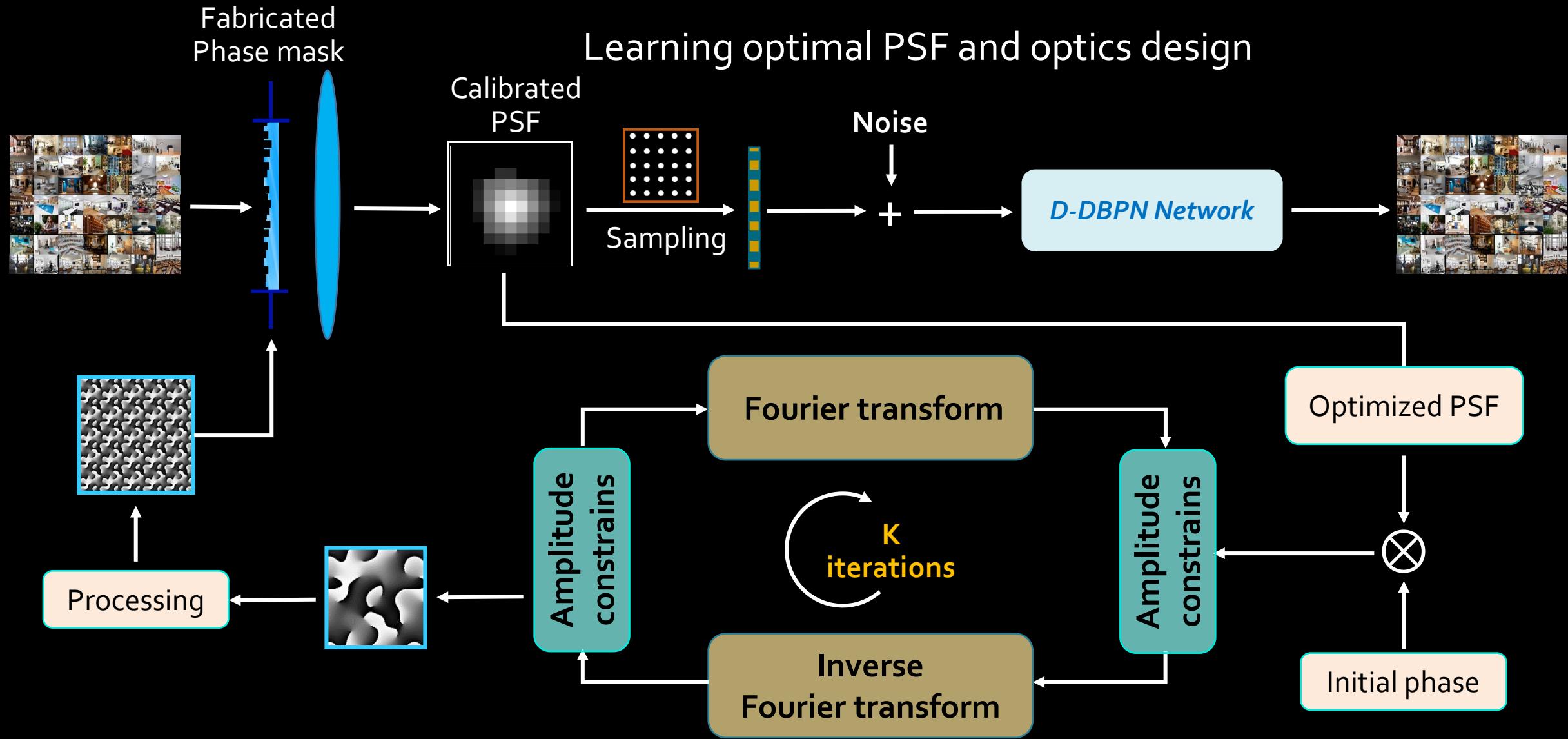
Optical Model



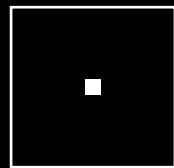
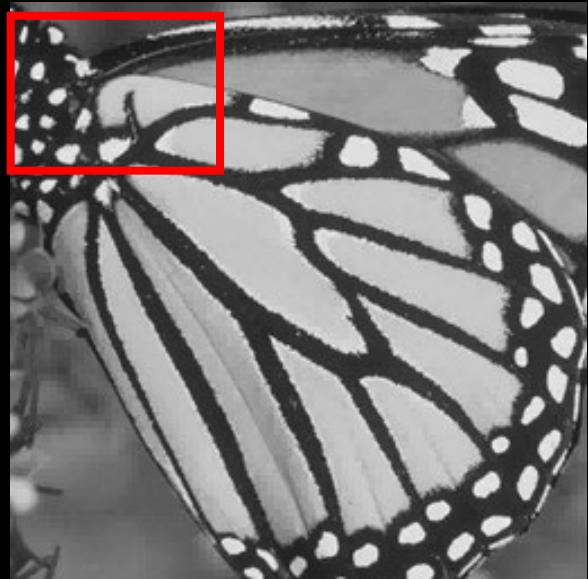
$$\phi(x', y')$$

$$\rho_\lambda(x, y) \propto \|\mathcal{F}\{\phi(x', y')\}\|^2$$

End-to-end Framework of SR SPAD Camera



Effectiveness of the Encoding Phase Mask



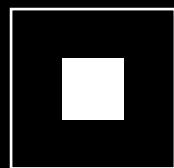
Low fill-factor



Low fill-factor
(no mask)



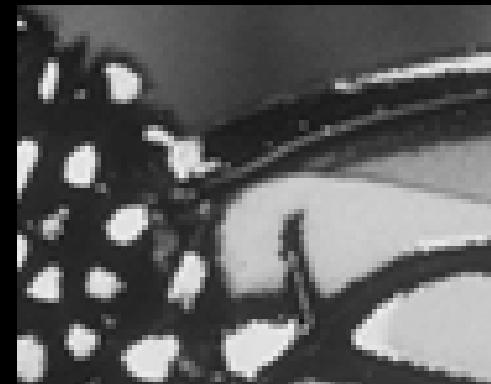
With phase mask



Full fill-factor



Full fill-factor



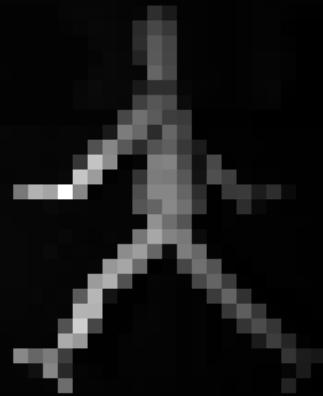
Ground Truth

model	Set5 PSNR/SSIM	Set14 PSNR/SSIM	BSDS100 PSNR/SSIM
Low fill-factor	27.17/ 0.9019	23.97/ 0.8066	23.82/ 0.7691
Full fill-factor	29.77/ 0.9317	26.13/ 0.8442	25.59/ 0.8069
Ours	30.76/ 0.9399	26.91/ 0.8557	26.23/ 0.8198

Effectiveness of the Encoding Phase Mask



Raw image without mask



Result without mask

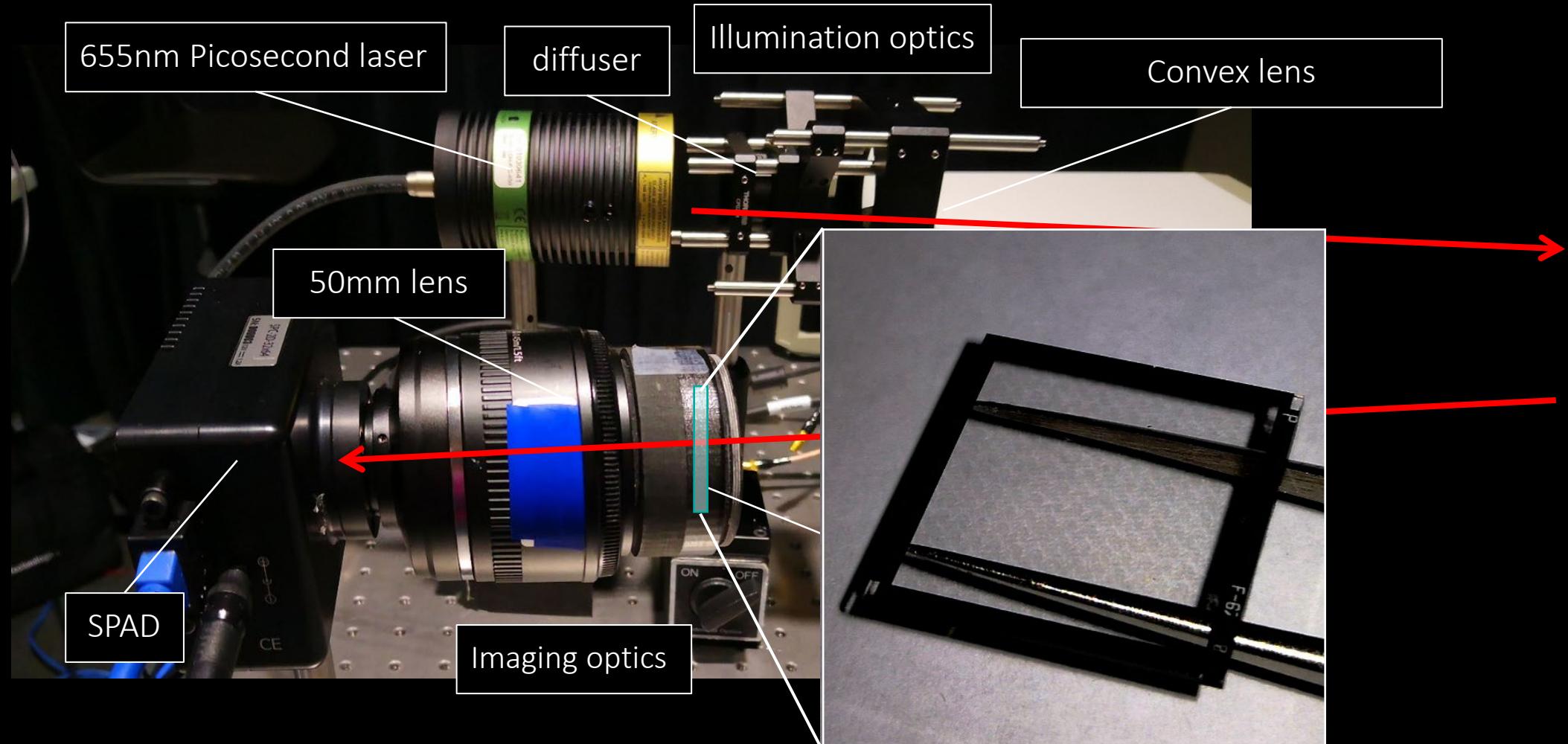


Raw image with mask



Result with mask

Prototype System

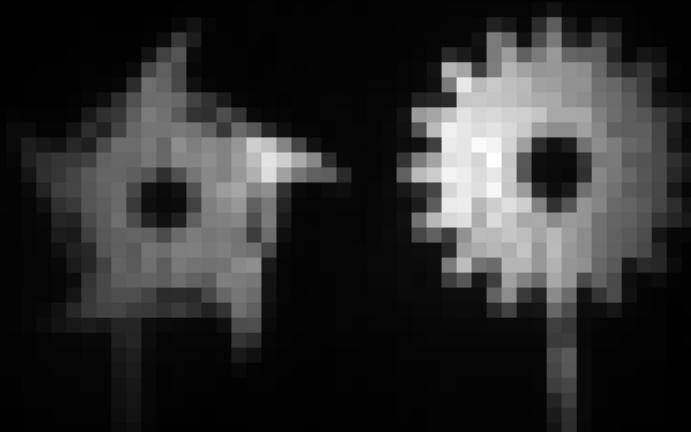


Applications: Intensity Imaging

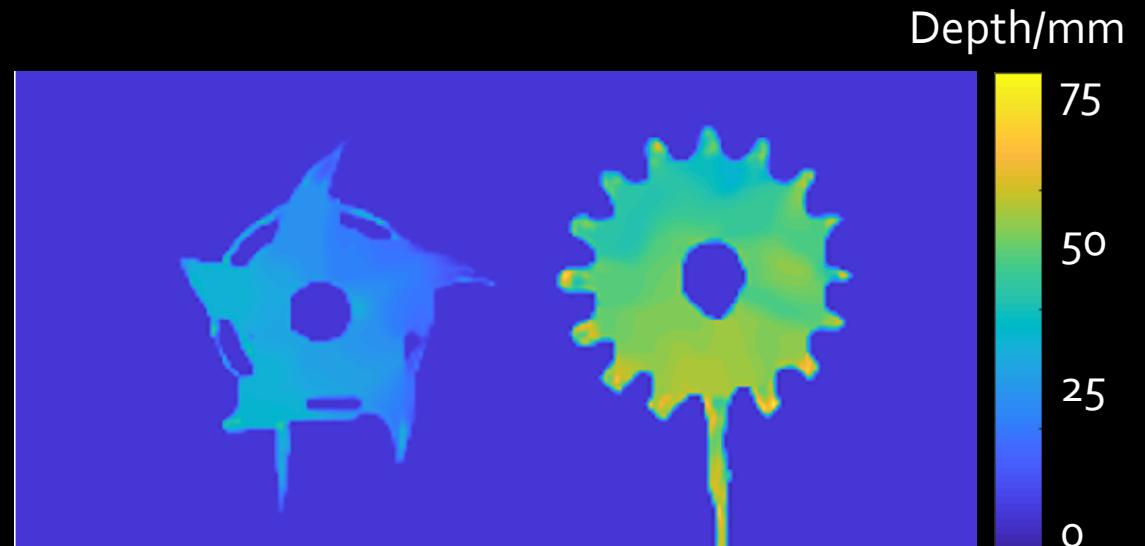


Integration time 5.2ms

Applications: Depth Imaging



SPAD measurement
summed over time axis

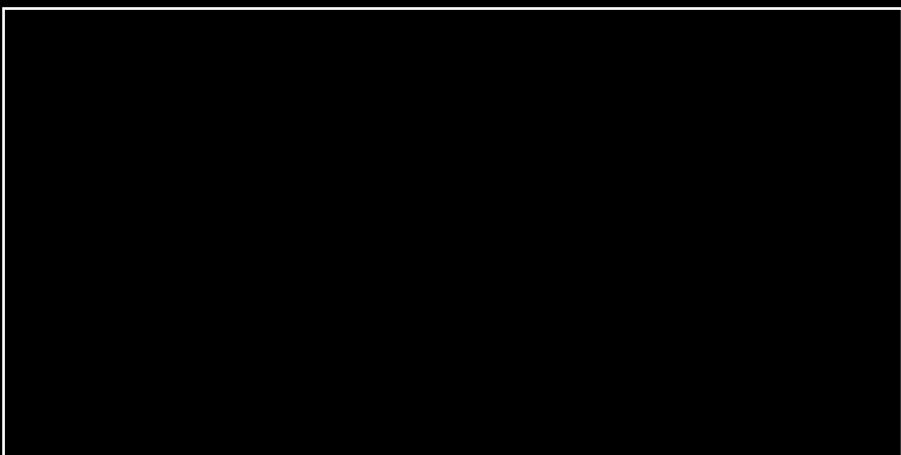


Reconstruction depth map

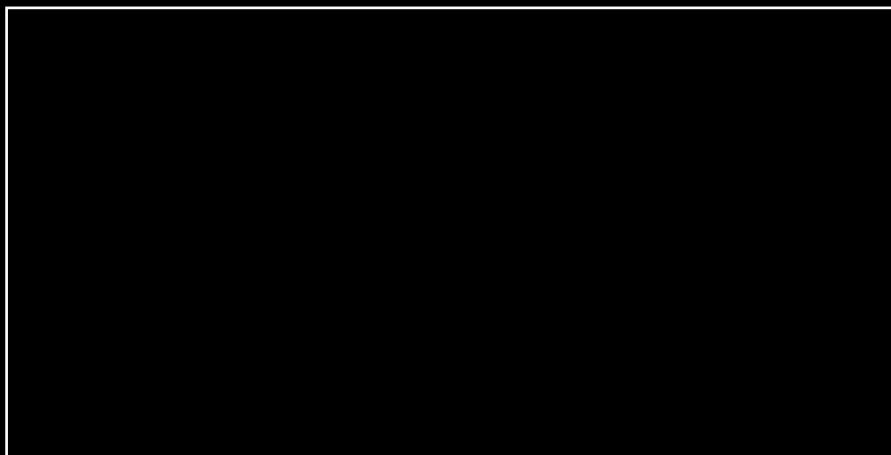
Applications: Transient Imaging



Scene



SPAD measurement
(time resolved 20ps)



Reconstruction result

Step Three: Enabling Comprimization for Color Channels

Learning Rank-1 Diffractive Optics for Single-shot High Dynamic Range Imaging

Limited Sensor Dynamic Range



Over Exposed



Under Exposed

Traditional Method—Burst Imaging



- Motion artifacts
- Long capture and processing time

Open Problem in HDR Imaging



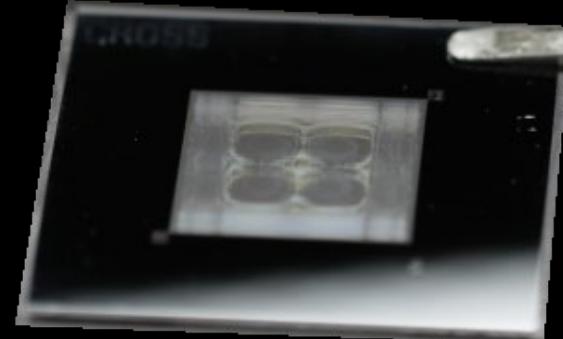
Open Problem in HDR Imaging

Single-shot HDR imaging using jointly learned optic + neural network

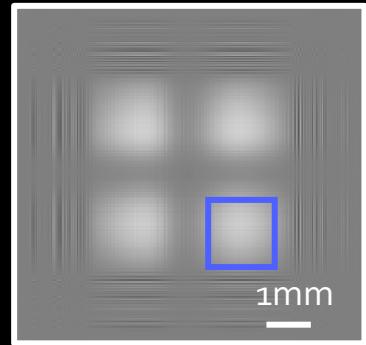
Prototype



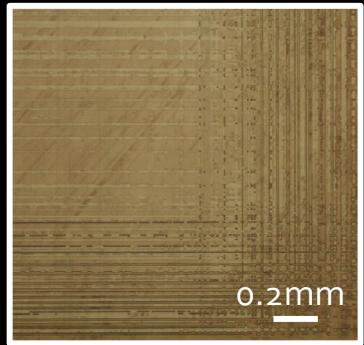
Fabricated learned optic



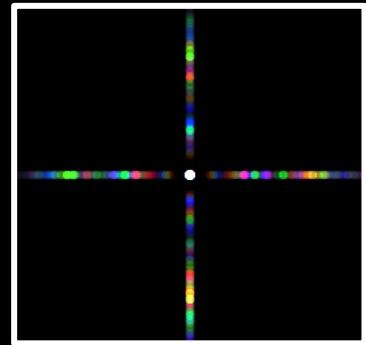
Simulated Height Map



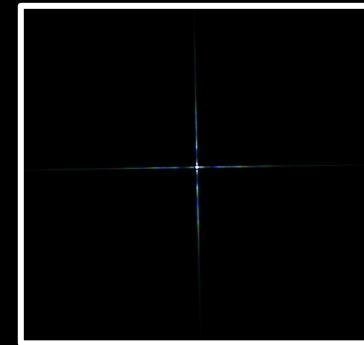
Fabricated Height Map



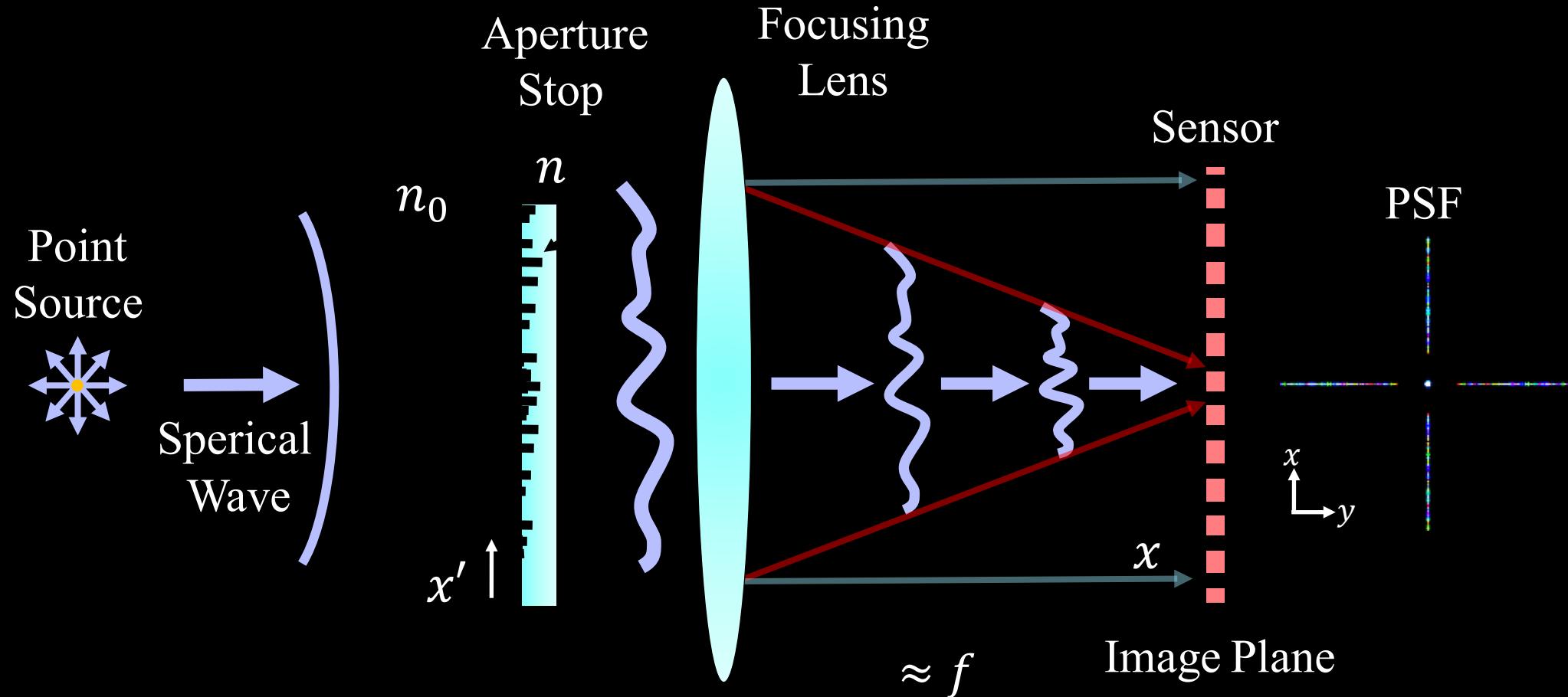
Simulated PSF



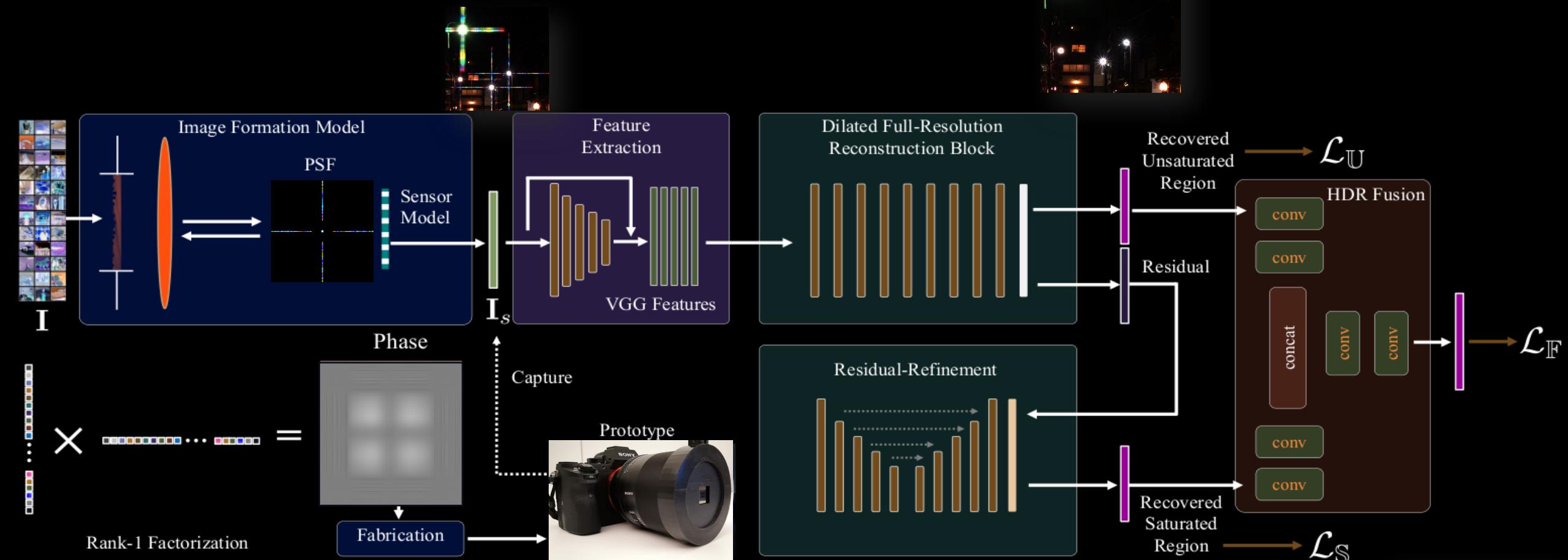
Captured PSF



Color-comprised Optical Model



End-to-End HDRI Pipeline



Learning Rank-1 Diffractive Optics

- Point Light Source

$$\mathbf{u}_- = A_0 e^{jk\sqrt{x'^2+y'^2+z^2}}$$

- DOE Layer and Rank-1 Factorization

$$\mathbf{u}_+ = \mathcal{A}(x', y') \mathbf{u}_- e^{jk(n_\lambda - 1)\mathbf{h}(x', y')}$$

- Heightmap

$$\mathbf{h}(x', y') = h_{\max} \sigma(\mathbf{v}\mathbf{q}^\top)$$

Trainable parameters: $\mathbf{v} \in \mathbb{R}^{m \times 3}$ $\mathbf{q} \in \mathbb{R}^{m \times 3}$

- Fresnel Propagation

$$\mathbf{u}_s = \mathcal{F}^{-1}\{\mathcal{F}\{\mathbf{u}_l\}\mathcal{H}\}$$

$$\mathcal{H}(f_x, f_y) = e^{jkL} e^{-j\pi\lambda L(f_x^2 + f_y^2)}$$

Loss Function

➤ Total Loss

$$\mathcal{L}_{\text{Total}} = \mathcal{L}_{\mathbb{U}} + \mathcal{L}_{\mathbb{S}} + \mathcal{L}_{\mathbb{F}}$$

➤ Loss on Unsaturated Regions

➤ VGG Loss

$$\mathcal{L}_{\text{VGG}}(\hat{x}, x) = \sum \nu_l \|\phi_l(\hat{x}) - \phi_l(x)\|_1$$

➤ Exclusion loss

$$\mathcal{L}_{\text{excl}} = \|\tanh(\eta_{\mathbb{U}} \mid \nabla \hat{\mathbf{I}}_{\mathbb{U}} \mid) \odot \tanh(\eta_r \mid \nabla \hat{\mathbf{I}}_r \mid)\|_F$$

➤ Loss on Saturated Region

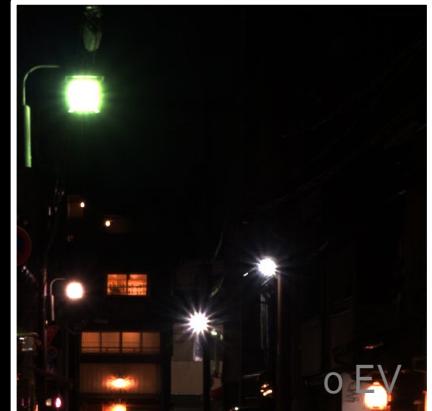
$$\mathcal{L}_{\mathbb{S}} = \beta \mathcal{L}_{\text{VGG}}(\hat{\mathbf{I}}_{\mathbb{S}}, \mathbf{I}_{\mathbb{S}})$$

➤ Loss in Fused Output

$$\mathcal{L}_{\mathbb{F}} = \mathcal{L}_{\text{Huber}} \left((\hat{\mathbf{I}} + \epsilon)^\gamma, (\mathbf{I} + \epsilon)^\gamma \right)$$

Simulated Results

HDR-CNN



0 EV

Glare-HDR



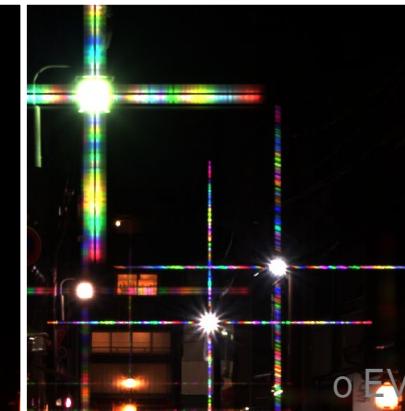
0 EV

Deep Optics



0 EV

Our Measurement



0 EV

Our Reconstruction



0 EV

Reference HDR



0 EV

-8 EV

-8 EV

-8 EV

-8 EV

-8 EV

-8 EV

Simulated Results

HDR-CNN

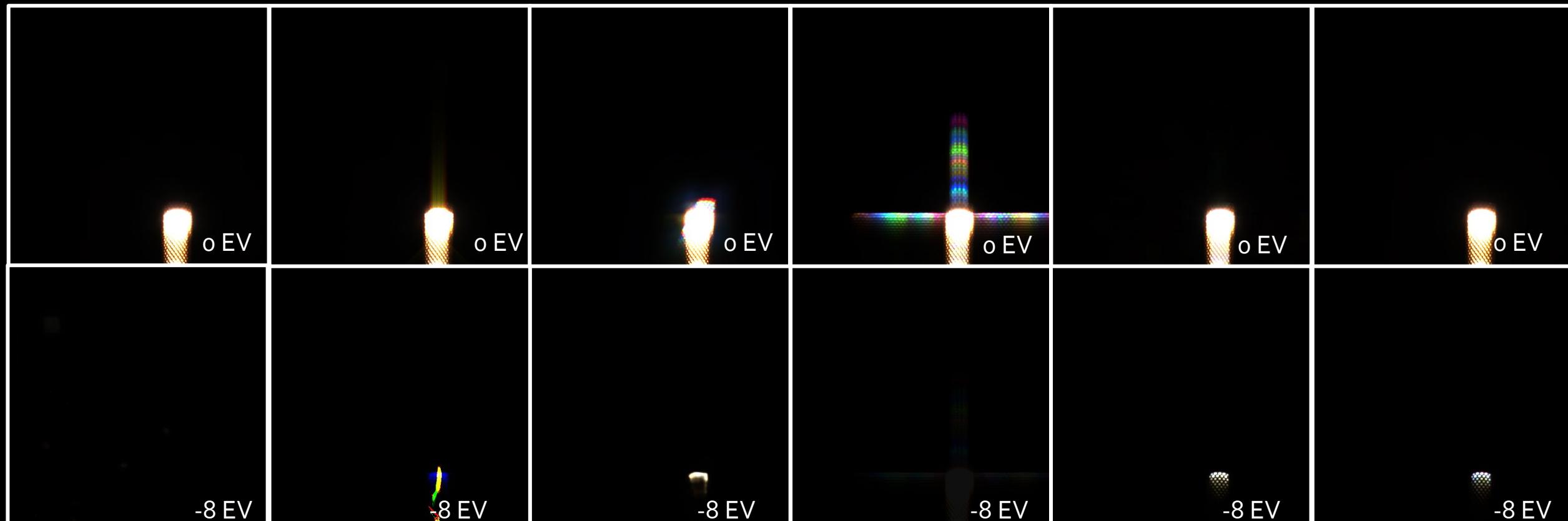
Glare-HDR

Deep Optics

Our Measurement

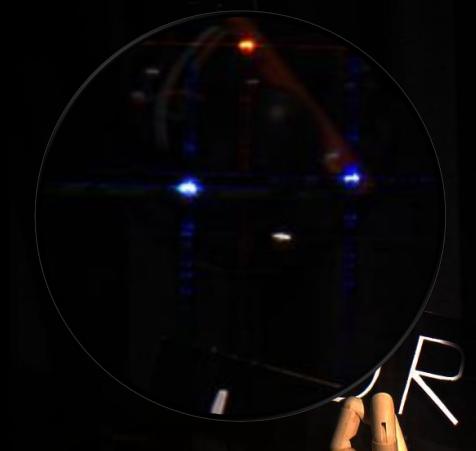
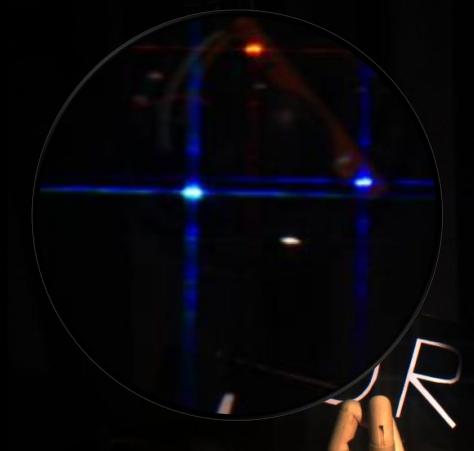
Our Reconstruction

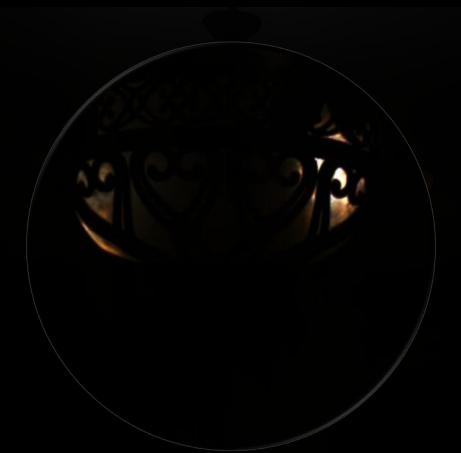
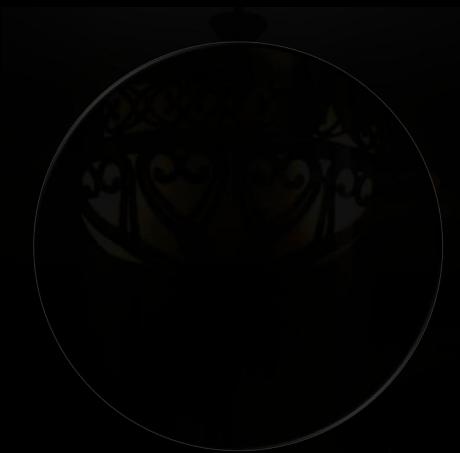
Reference HDR



Comparison in Simulation

Methods	PSNR	HDR-VDP 2
Ours	48.26	74.47
Deep Optics	40.30	67.96
Glare-HDR	32.23	56.76
HDR-CNN	34.06	54.34
LDR	33.57	52.43





Sensor Measurement



香港中文大學(深圳)
The Chinese University of Hong Kong, Shenzhen



Reconstruction



Measurement



-4 EV



-6 EV

Reconstruction



-4 EV



-6 EV

Reference



-4 EV



-6 EV

Measured



Recovered



Reference



Measured



Recovered



Reference



0 EV

0 EV

0 EV

-4 EV

-4 EV

-4 EV

-6 EV

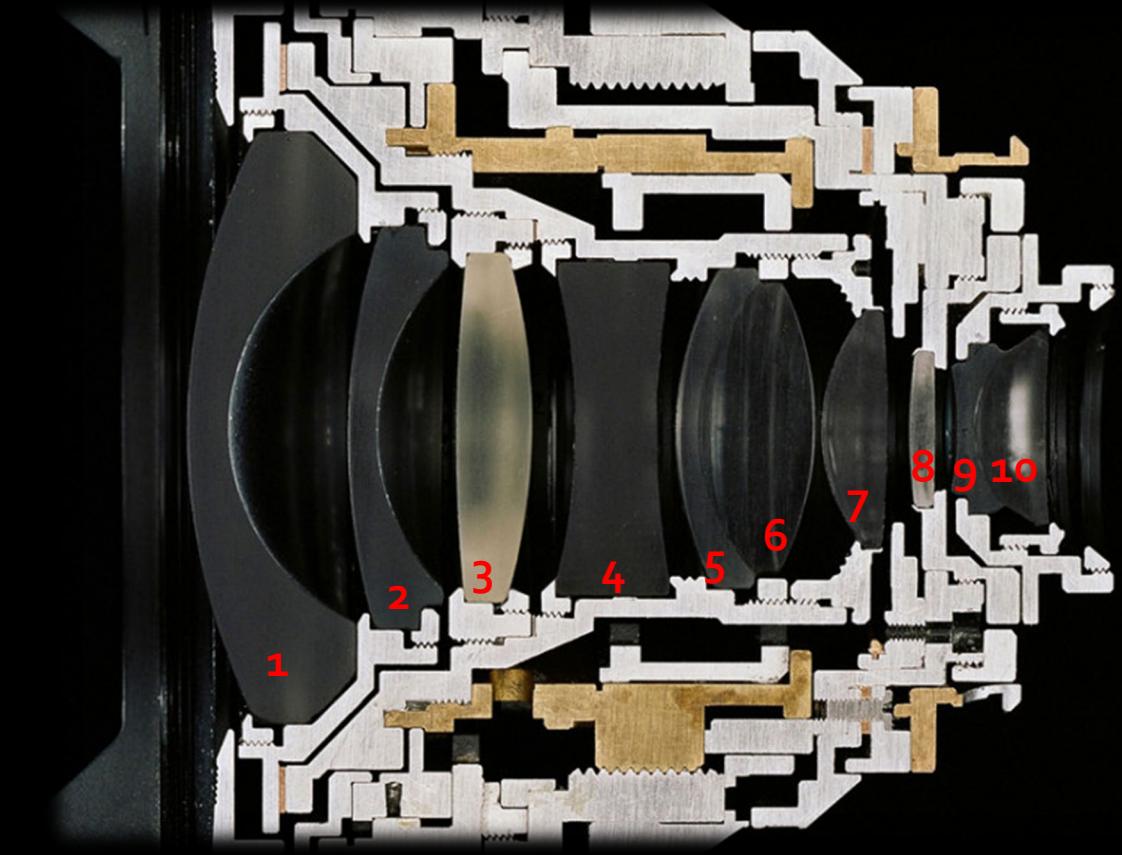
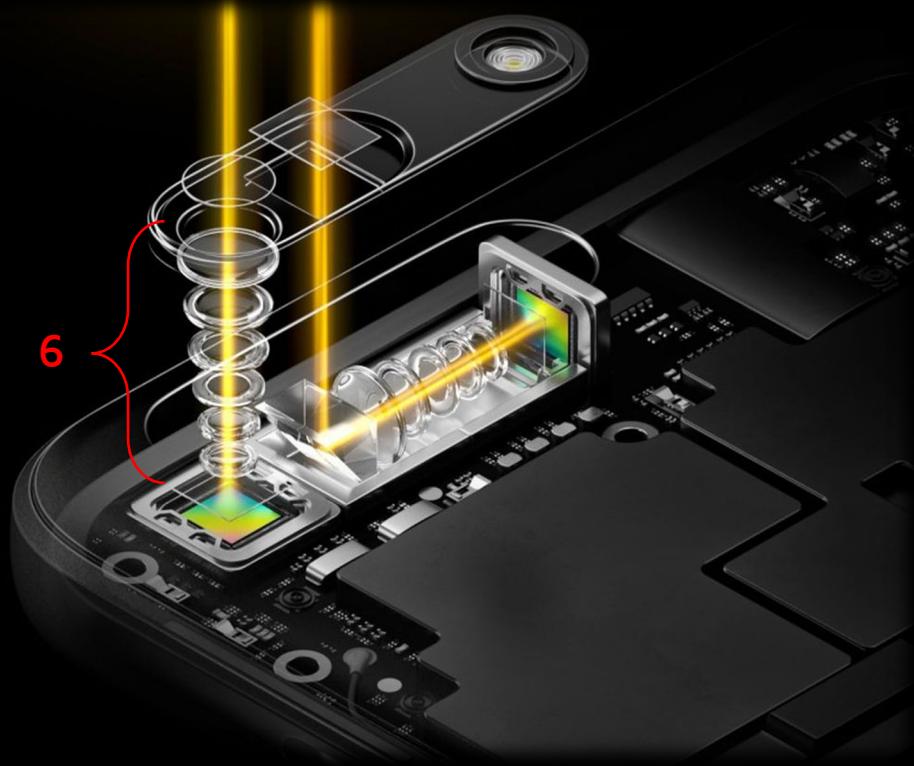
-6 EV

-6 EV

Step Four: End-to-end Complex Lens Design

End-to-End Complex Lens Design with Differentiable Ray Tracing

Complex Lens Systems

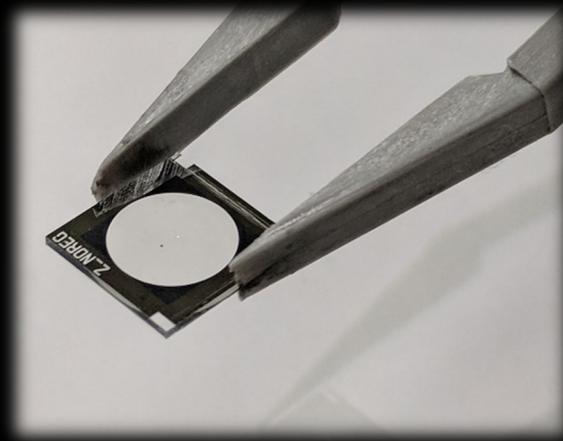


- High quality imaging combat with aberrations of all kinds
- Sometimes not optimal for a given task

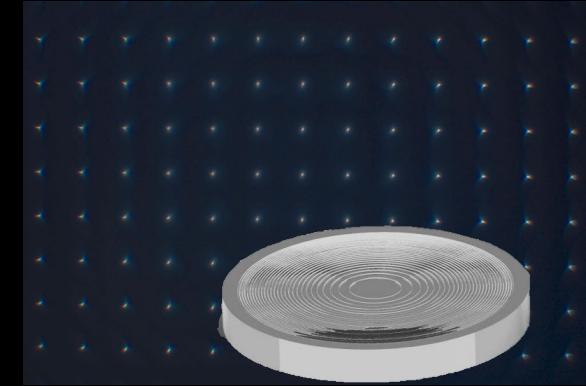
* Images credits @ Oppo, Nikon

End-to-end Optics Designing

- Diffractive optics
- Refractive optics
- Thin form factor



[Sitzmann et al. 2018]

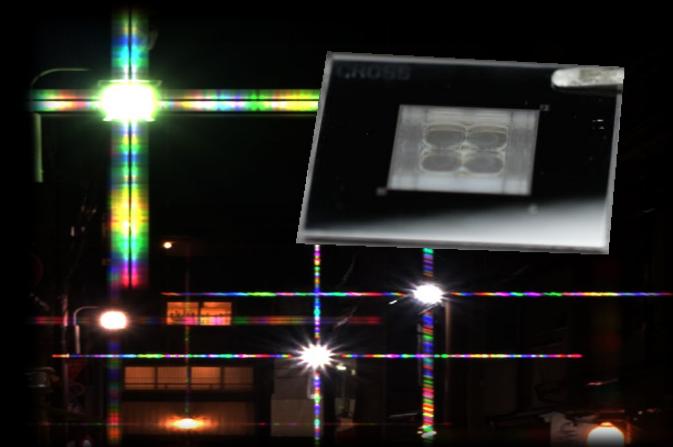


[Peng et al. 2019]

- Single layer differentiable
- Paraxial approximation
- Limited design freedom
- Limited image quality
- Not fully differentiable

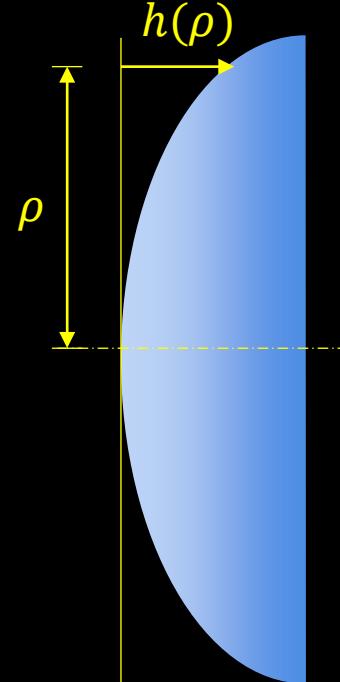


[Sun et al. 2020]



[Sun et al. 2020]

Differentiable Lens Model and Ray Tracer



➤ Aspherical Lenses:

$$h(\rho) = \frac{c\rho}{1 + \sqrt{1 - \alpha\rho}} + \sum_{i=2}^n a_{2i}\rho^i \quad \alpha = (1 + \kappa)c^2$$

➤ Aspherical Lenses's Derivative:

$$h'(\rho) = c \frac{1 + \sqrt{1 - \alpha\rho} - \alpha\rho/2}{\sqrt{1 - \alpha\rho}(1 + \sqrt{1 - \alpha\rho})^2} + \sum_{i=2}^n ia_{2i}\rho^{i-1}$$

➤ Implicit form $f(x, y, z)$ and its derivative:

$$f(x, y, z) = h(\rho) - z$$

$$\nabla f = (2h'(\rho)x, 2h'(\rho)y, -1)$$

Differentiable Lens Model and Ray Tracer

- Ray-surface Intersection by Newton's Method: $\text{Ray}(\mathbf{o}, \mathbf{d})$

- Find the $t > 0$ such that:

$$f(x, y, z) = f(\mathbf{o} + t\mathbf{d}) = 0$$

- We solve the problem numerically using Newton's method

$$\begin{aligned} t^{(k+1)} &\leftarrow t^{(k)} - \frac{f(\mathbf{o} + t^{(k)}\mathbf{d})}{f'(\mathbf{o} + t^{(k)}\mathbf{d})} \\ &\leftarrow t^{(k)} - \frac{f(\mathbf{o} + t^{(k)}\mathbf{d})}{\nabla f \cdot \mathbf{d}} \end{aligned}$$

- Dispersion extention of Mitsuba2 by Cauchy's equation

$$n(\lambda) = A + \frac{B}{\lambda^2} + \frac{C}{\lambda^4} + \dots \quad A = n_D - \frac{B}{\lambda_D^2} \quad B = \frac{n_D - 1}{V(\lambda_F^{-2} - \lambda_C^{-2})} \quad \dots$$

Differentiable Lens Model and Ray Tracer

- Recorded image on sensor:

$$I_c(x', y') = \int Q_c(\lambda) [p(x', y', d, \lambda) * s_c(x', y', d)] d\lambda + n(x', y')$$

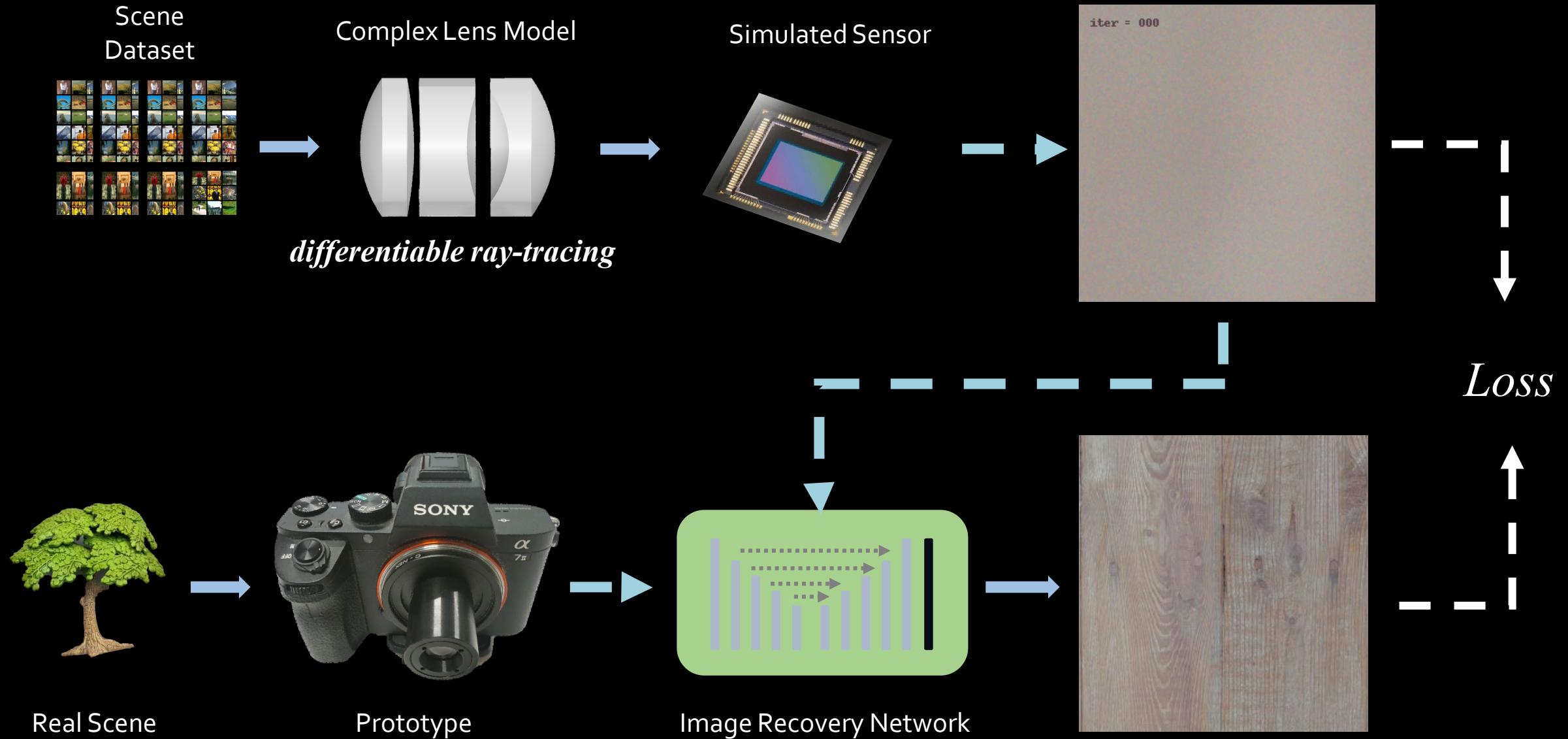
- Monto Carlo Sampling
- Image auto alignment during training:
 - The relationship between traced points on sensor \mathbf{r}_d and ideal points \mathbf{r} can be expressed as:

$$\mathbf{r}_d = \zeta \mathbf{r} (1 + k_1 \mathbf{r}^2 + k_2 \mathbf{r}^4 + k_3 \mathbf{r}^6)$$

- We solve a least square problem to obtain the current distortion coefficients and magnification.

$$\min_K \| [\mathbf{r}, \mathbf{r}^3, \mathbf{r}^5, \mathbf{r}^7] K^T - \mathbf{r}_d \|_2^2$$

Differentiable Complex Lens and Recovery



Large Field-of-View Imaging

AL2550



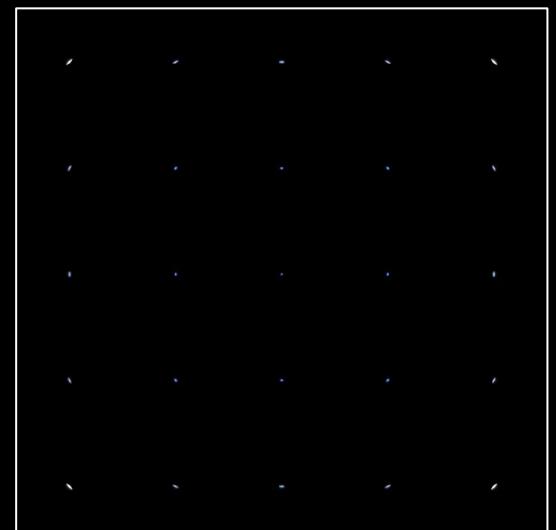
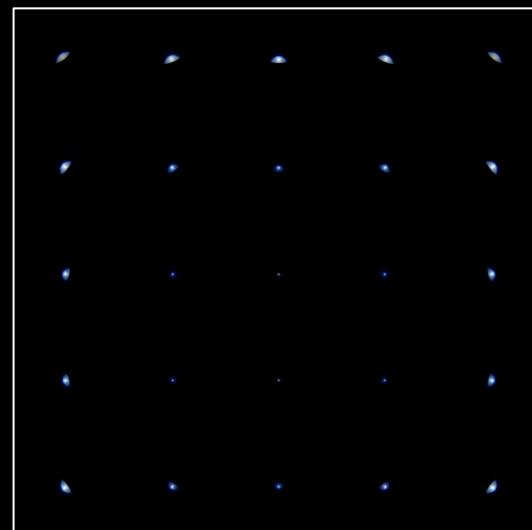
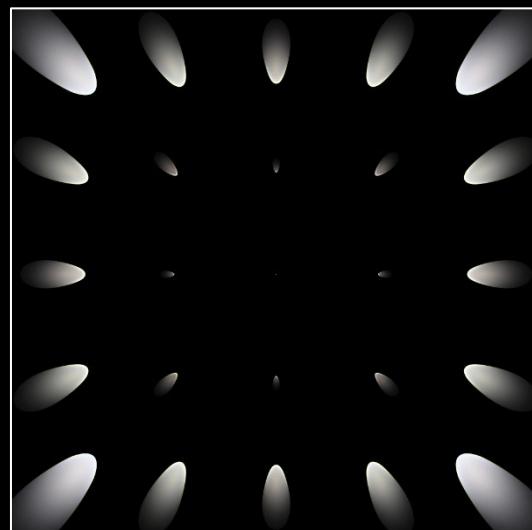
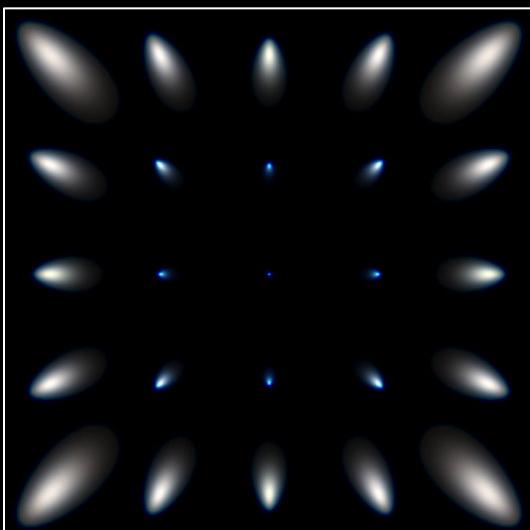
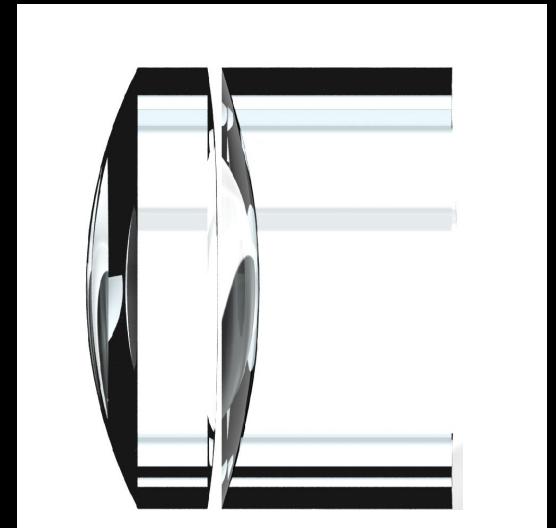
ACA254



Peng et al 2019

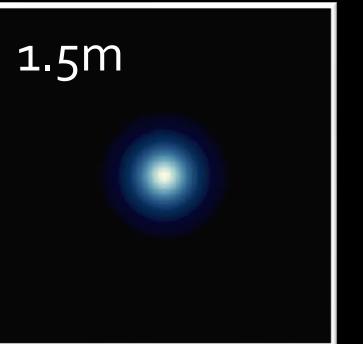
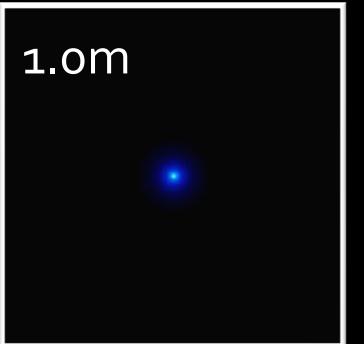
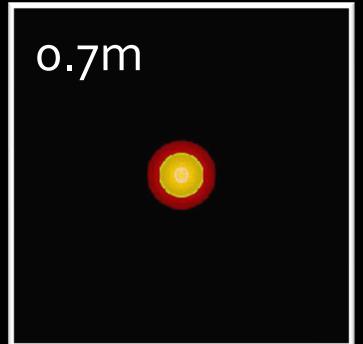
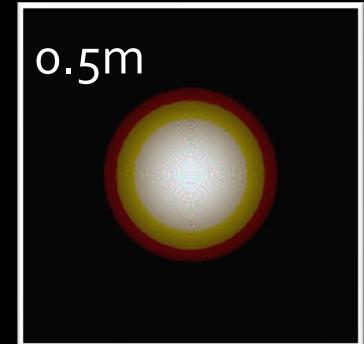


Ours

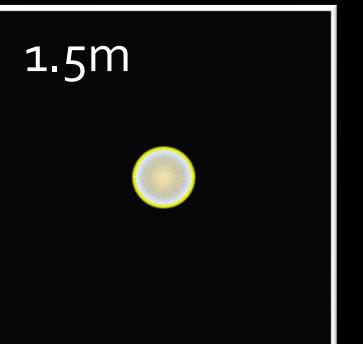
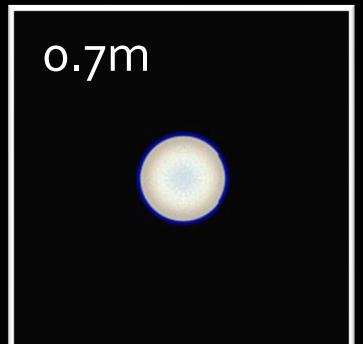
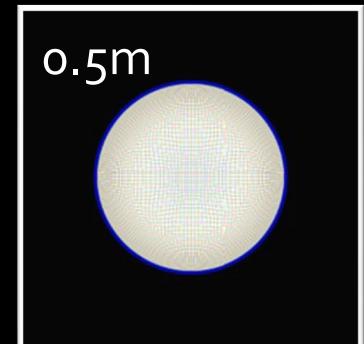


Extended Depth-of-Field

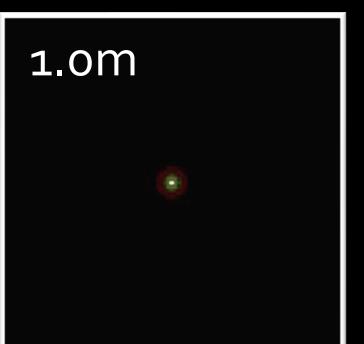
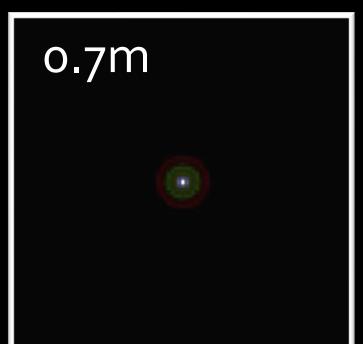
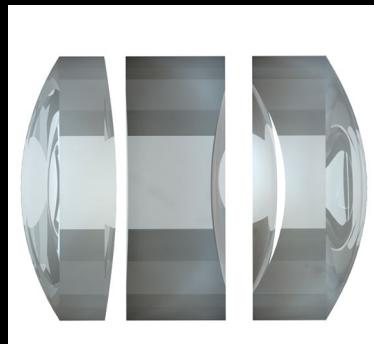
AL250



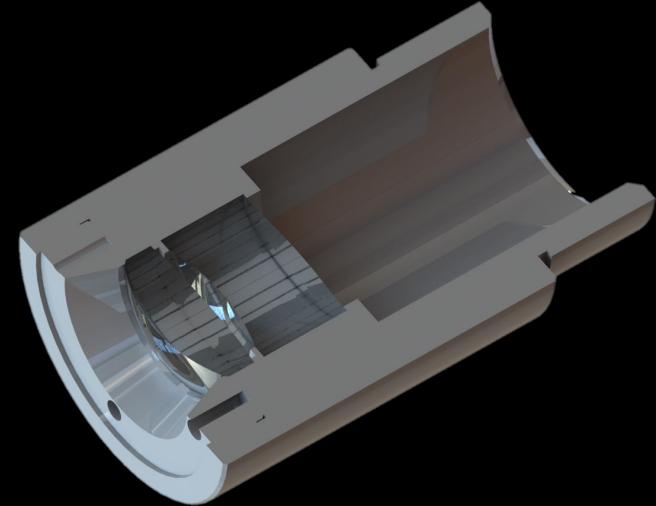
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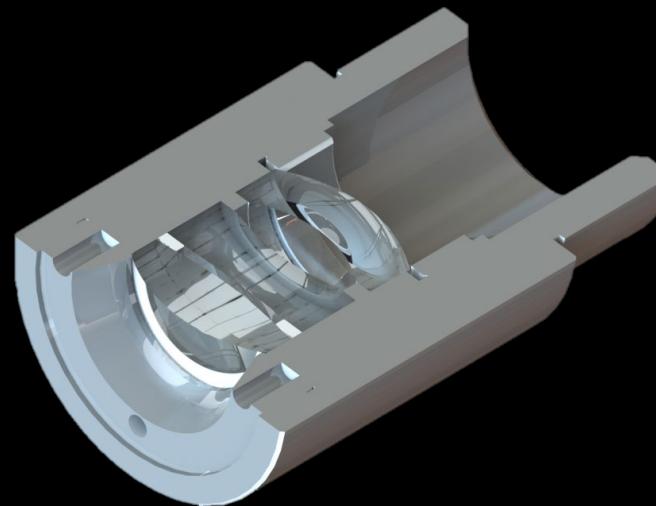
Ours



Prototypes



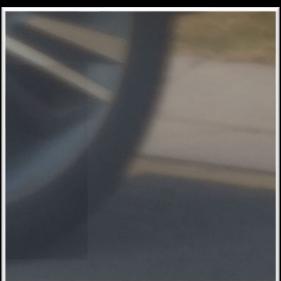
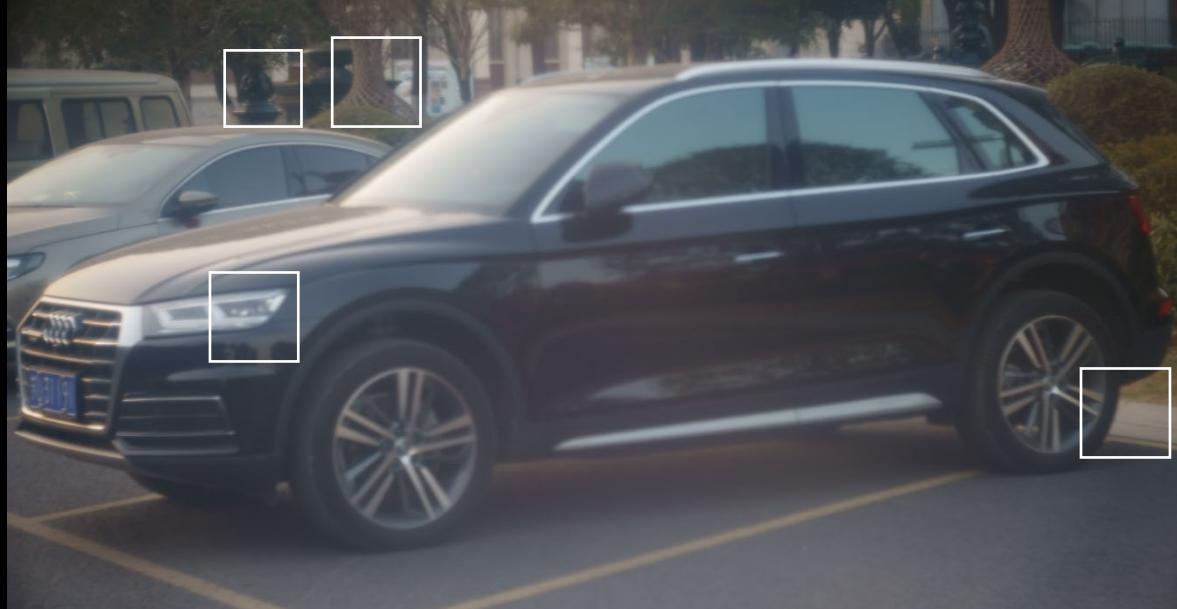
50mm/f4



50mm/f4



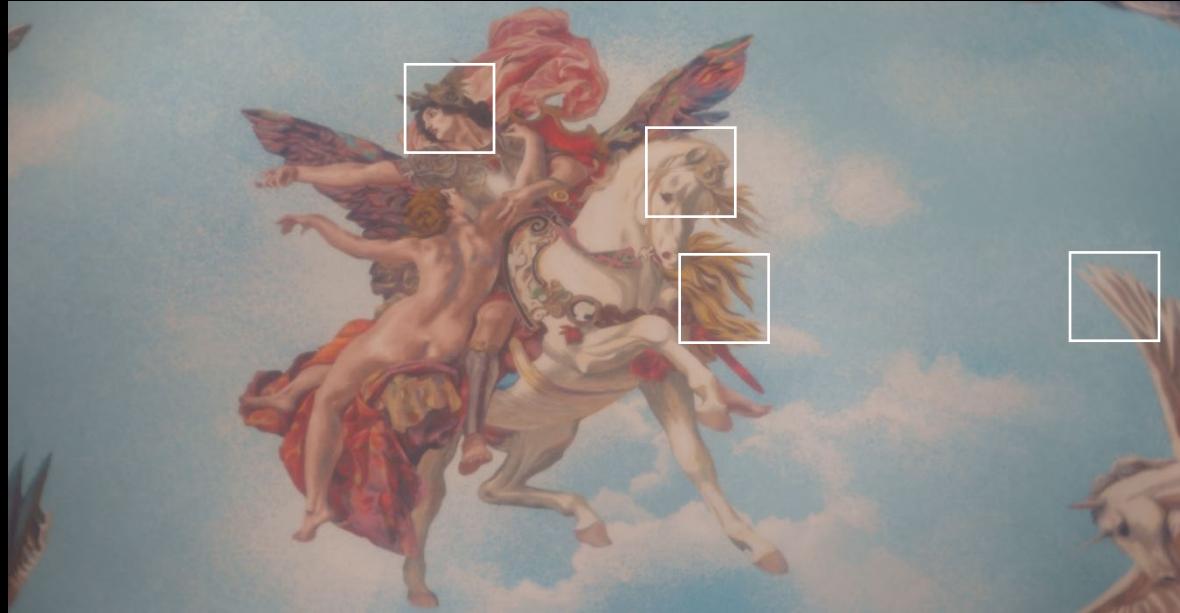
Results of Large Field-of-View Imaging



Captured

Recovered

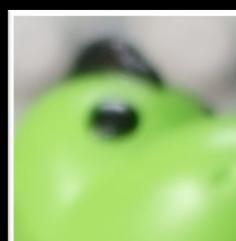
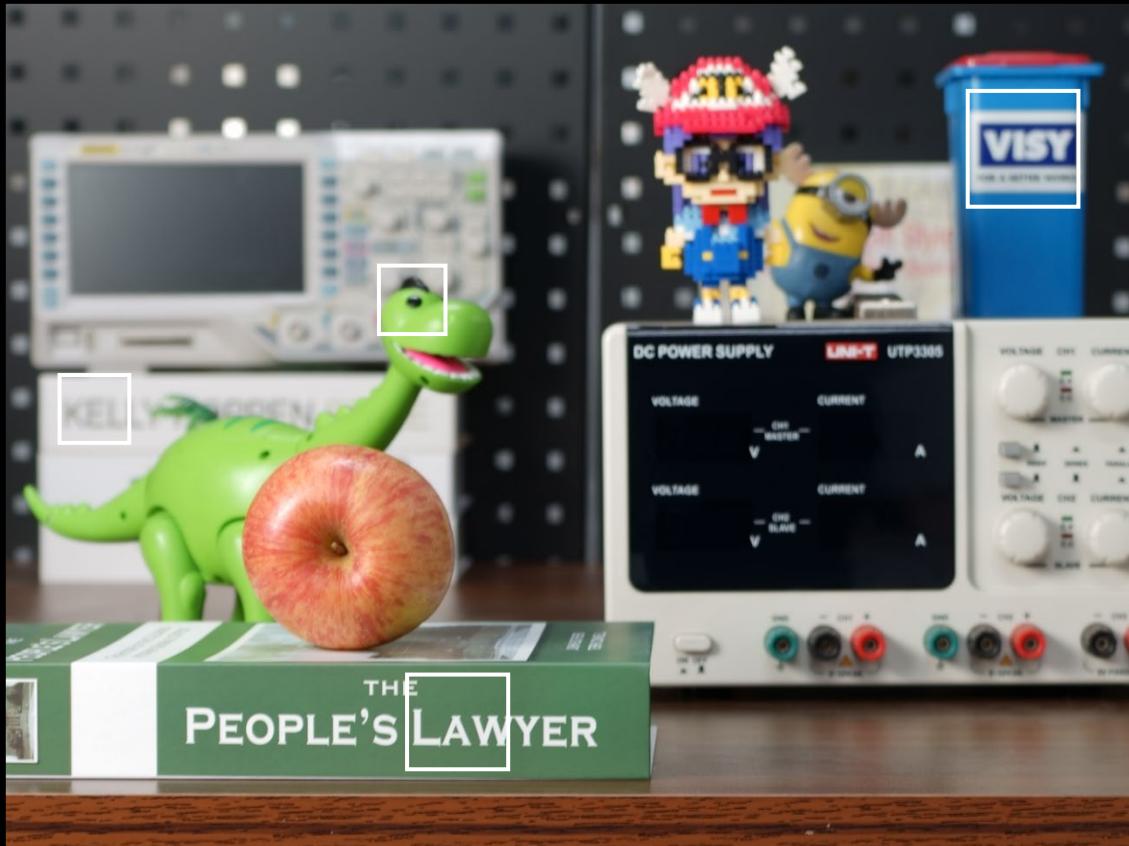
Results of Large Field-of-View Imaging



Captured

Recovered

Results of Extended Depth-of-Field



Sony standard zoom lens at $50mm/f4.5$

Ours at $50mm/f4$

Results of Extended Depth-of-Field



Ours at 50mm/f4
Captured

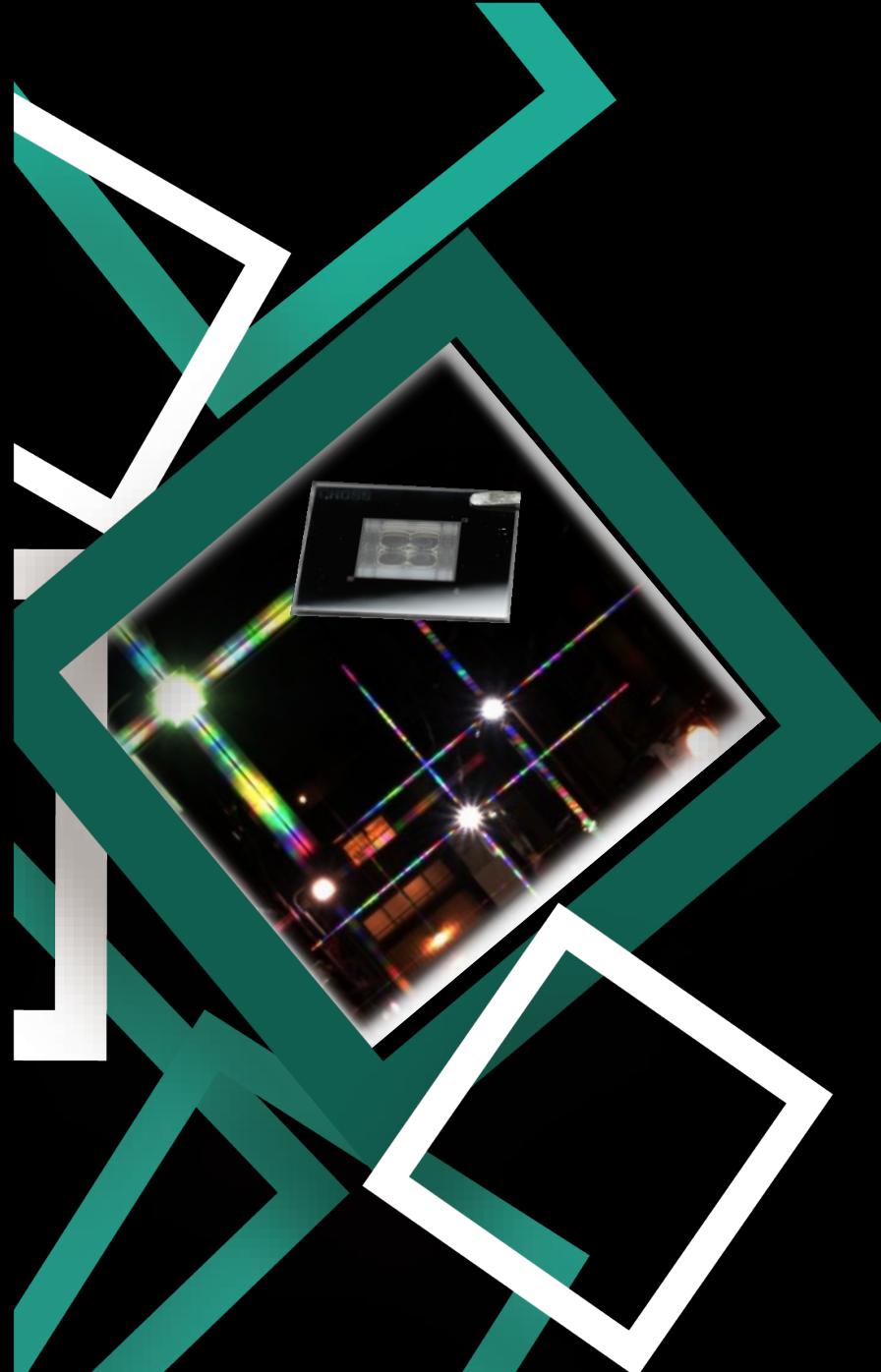
Ours at 50mm/f4
Processed

Sony standard zoom lens
at 50mm/f4.5

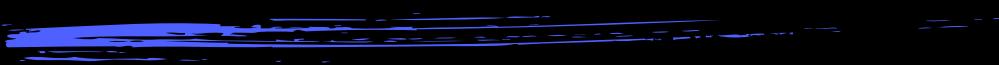


Today's Topic

- Step One: Joint Optimization of Optics and Image Recovery
 - Learned Large Field-of-View Imaging with Thin-Plate Optics
- Step Two: Compressive and Optically Coded Super-resolution SPAD Camera
 - Compressive and Optically Coded Super-resolution SPAD Camera
- Step Three: Enabling Comprimization for Color Channels
 - Learning Rank-1 Diffractive Optics for Single-shot High Dynamic Range Imaging
- Step Four: End-to-end Complex Lens Design
 - End-to-end Complex Lens Design with Differentiable Ray-Tracing



Computational Imaging



Thank You!



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