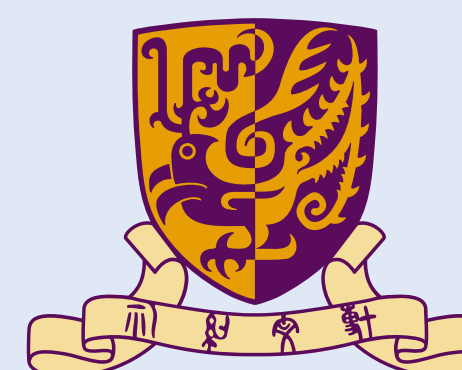


Learning to Remove Wrinkled Transparent Film with Polarized Prior

Jiaqi Tang^{1,2,3} Ruizheng Wu⁴ Xiaogang Xu^{5,6} Sixing Hu⁴ and Ying-Cong Chen^{1,2,3} *

¹HKUST(GZ), ²HKUST, ³HKUST(GZ) – SmartMore Joint Lab, ⁴SmartMore, ⁵CUHK, ⁶ZJU



Project Page

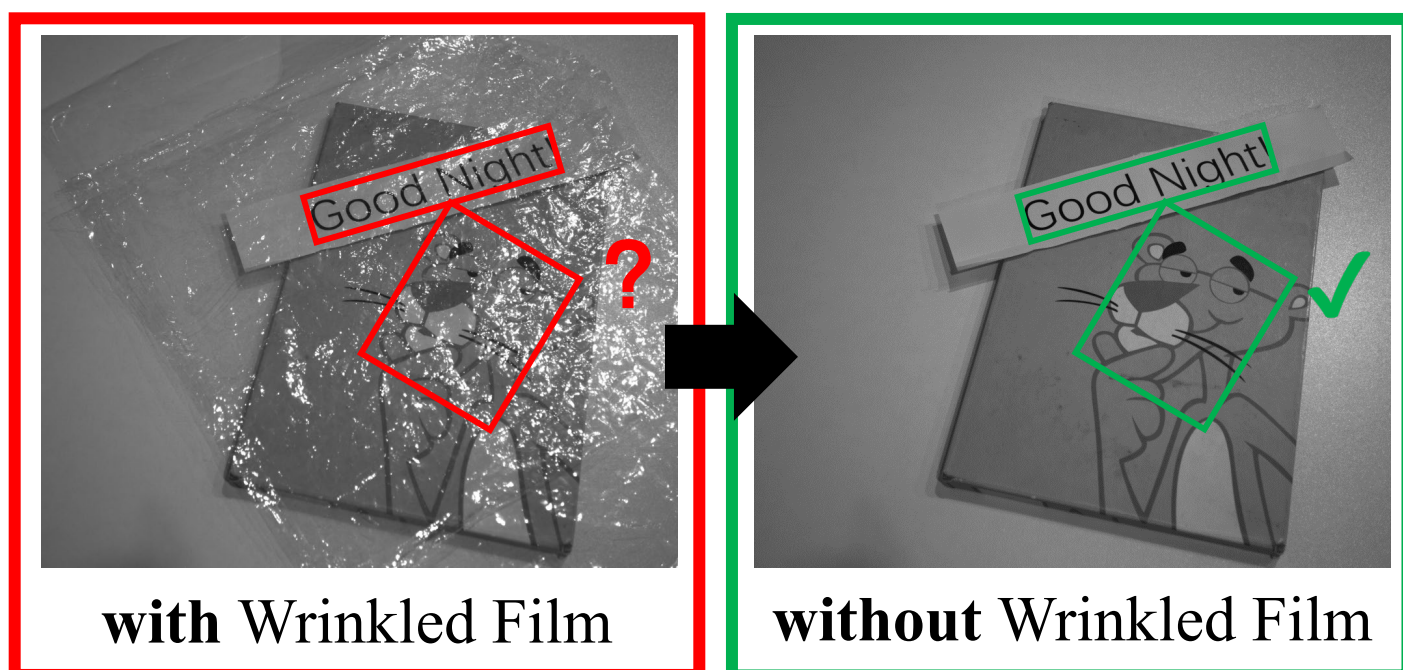


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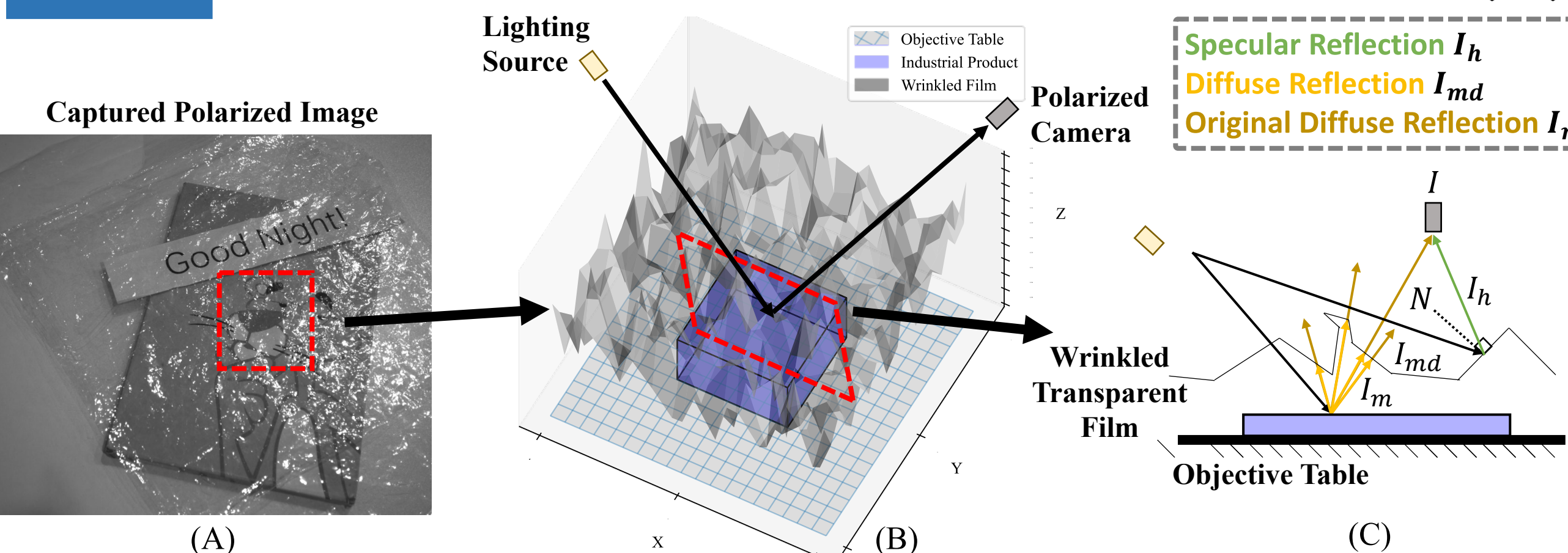
New Problem in Low-Level Vision

Film Removal (FR)

- To remove the interference of wrinkled transparent films.
- To reconstruct the original information under films.
- Application: industrial recognition systems.



Physics Model of Wrinkled Transparent Film



- All compositions are $I = I_{md} + I_h = I_m + I_d + I_h$ + : linear superposition
- Our **final goal** is $I_m = I - I_h - I_d$ -: decoupling operator

Data Engineering

Industrial Optical Photography Pipeline

How to maintain the Data Diversity and Robustness?

- 315 dynamic industrial scenarios.
- Three types: QR codes, text, and products.
- Diverse properties: coverage areas, film thicknesses, levels of wrinkling.
- Fixed: to minimize the influence of errors external.

Methodology

Two Parts

- **[Observation]** Specular Reflection (Highlight) I_h is **Polarized**.
- **[Solution]** Estimating a **Polarized Prior** for Locating I_h
- The Prior is: $P = I_m + I_d + \min I_h$
- The polarized version of the prior, can be acquired with Malus's Law and the elliptical polarization model, therefore:

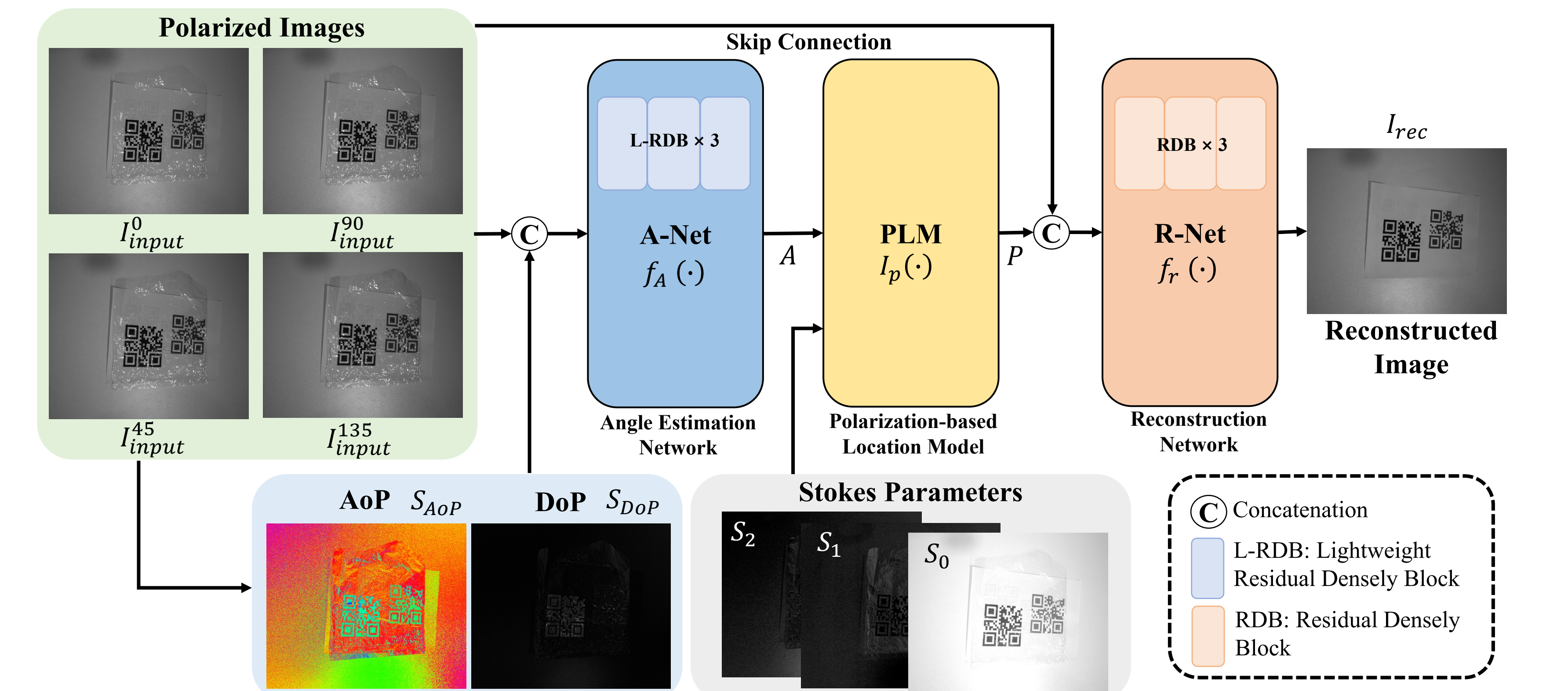
$$I_h = I_p(\theta) = I_{max} \cos^2 \theta + I_{min} \sin^2 \theta$$
- Since I_h is the only polarized component that is determined by θ , P can also be formulated as:

$$P = I_m + I_d + \min I_h$$

$$= I_m + I_d + \min_{\theta} I_p(\theta)$$

$$= I_m + I_d + \min (I_{max} \cos^2 \theta + I_{min} \sin^2 \theta)$$
- Finally, we estimate pixel-wise θ , with a learning-based network (A-Net), to obtain the angle map A :

$$A = f_A(I_{input}^0 \oplus I_{input}^{45} \oplus I_{input}^{90} \oplus I_{input}^{135} \oplus S_{AoP} \oplus S_{DoP})$$



Experiment

➤ [Table 1] Quantitative evaluation in image reconstruction with 10-fold cross-validation.

		K1	K2	K3	K4	K5	K6	K7	K8	K9	K10	$\mu \uparrow$	$\sigma \downarrow$
SHIQ [6]	PSNR	23.47	22.11	21.95	21.69	21.77	21.03	20.86	20.46	21.10	21.31	21.58	0.64
	SSIM	0.7899	0.7640	0.7416	0.7439	0.7459	0.7465	0.7499	0.7412	0.7465	0.7300	0.7499	2.41×10^{-4}
Polar-HR [34]	PSNR	23.31	22.80	22.13	21.58	21.94	22.00	22.03	21.99	22.18	21.95	22.19	0.22
	SSIM	0.7642	0.7421	0.7220	0.7099	0.7064	0.7098	0.7128	0.7017	0.7102	0.6968	0.7176	3.80×10^{-4}
Uformer [33]	PSNR	31.85	31.95	31.39	31.19	31.81	32.04	31.68	31.98	31.85	31.01	31.68	0.11
	SSIM	0.9519	0.9456	0.9371	0.9364	0.9434	0.9421	0.9438	0.9435	0.9457	0.9363	0.9426	2.17×10^{-5}
Restormer [41]	PSNR	34.35	35.02	34.44	33.71	34.88	35.13	34.31	34.33	34.51	32.49	34.32	0.52
	SSIM	0.9771	0.9770	0.9721	0.9678	0.9757	0.9746	0.9742	0.9741	0.9759	0.9633	0.9731	1.75×10^{-5}
Ours	PSNR	36.76	37.29	36.62	35.12	36.93	37.21	36.24	36.67	36.94	35.02	36.48	0.57
	SSIM	0.9852	0.9859	0.9822	0.9767	0.9845	0.9833	0.9836	0.9830	0.9850	0.9749	0.9824	1.23×10^{-5}

Figure 1-3

Qualitative Evaluation.

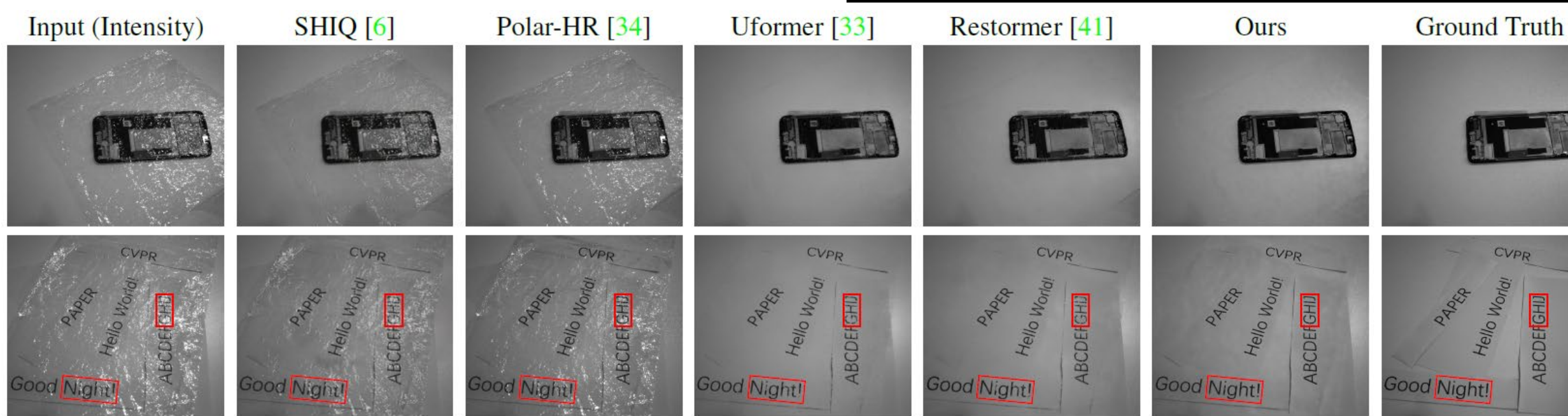


Figure 4-6

Ablation Study.

