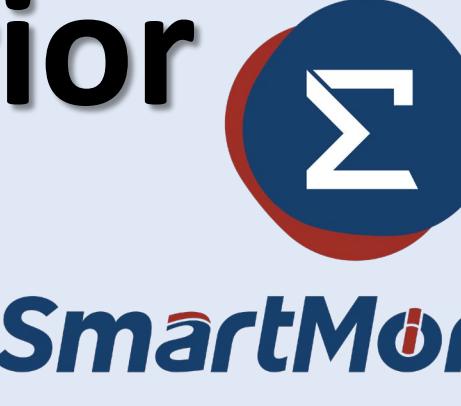


Learning to Remove Wrinkled Transparent Film with Polarized Prior

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SmartMore



Project Page

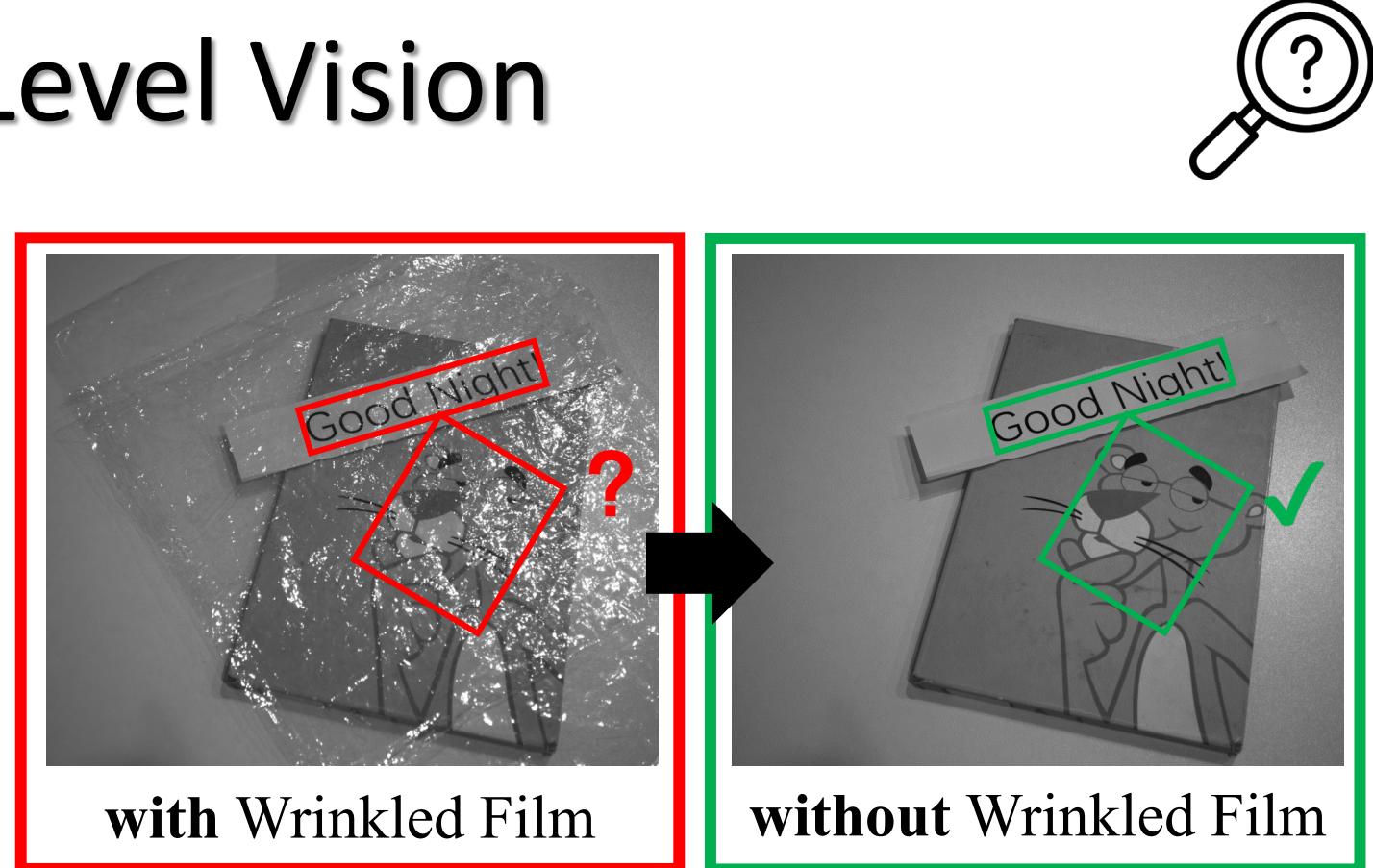
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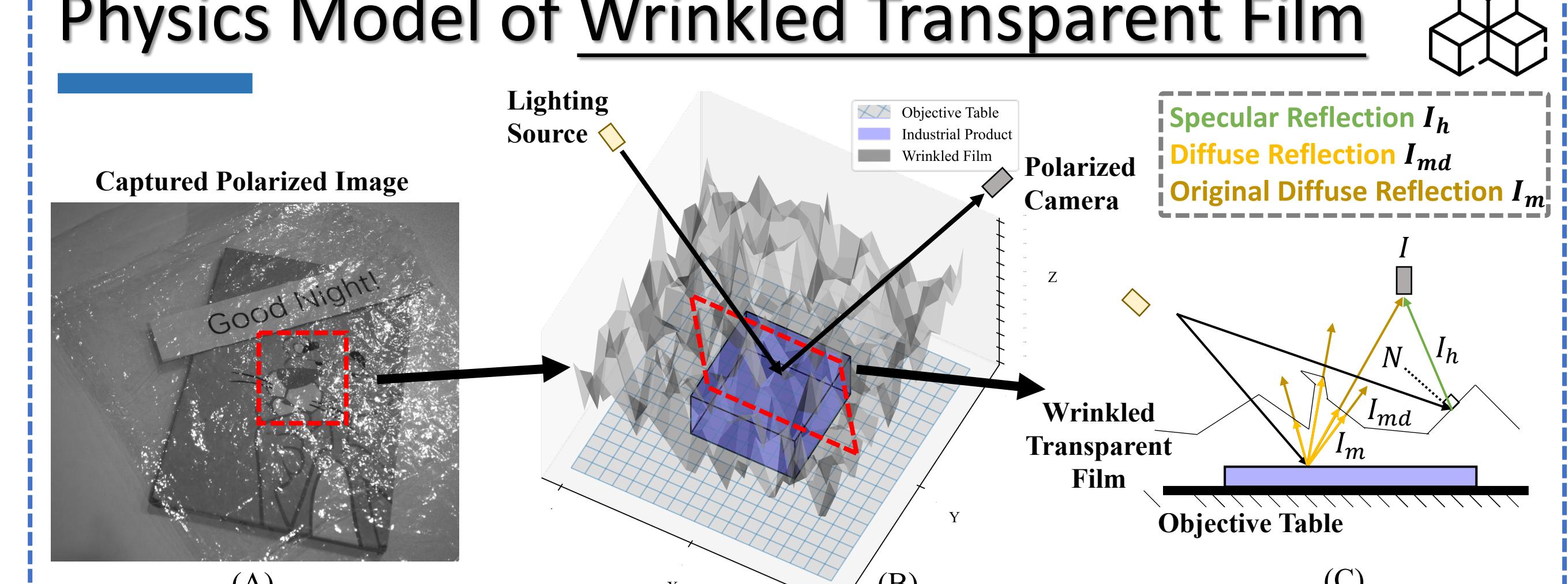
New Problem in Low-Level Vision

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:

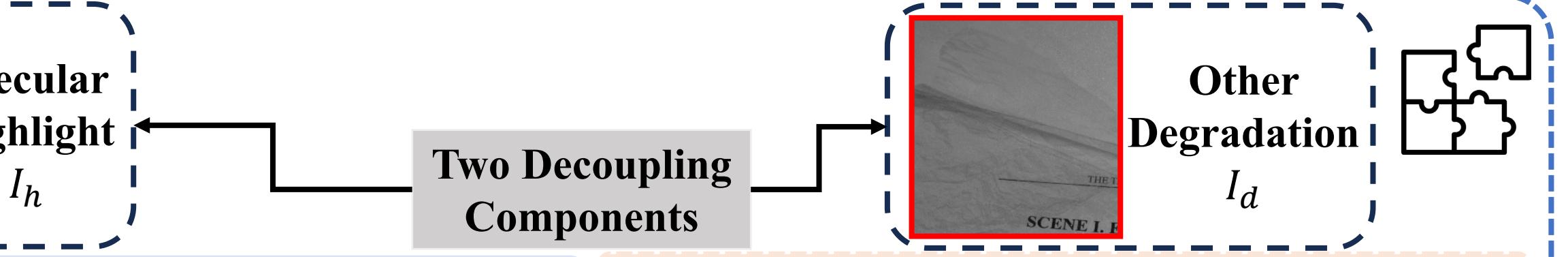
$$I_h = I_p(\theta) = I_{max} \cos^2 \theta + I_{min} \sin^2 \theta$$

- Since I_h is the only polarized component that is determined by θ , P can also be formulated as:

$$\begin{aligned} P &= I_m + I_d + \min I_h \\ &= I_m + I_d + \min I_p(\theta) \\ &= I_m + I_d + \min (I_{max} \cos^2 \theta + I_{min} \sin^2 \theta) \end{aligned}$$

- Finally, we estimate pixel-wise θ , with a learning-based network (A-Net), to obtain the angle map A :

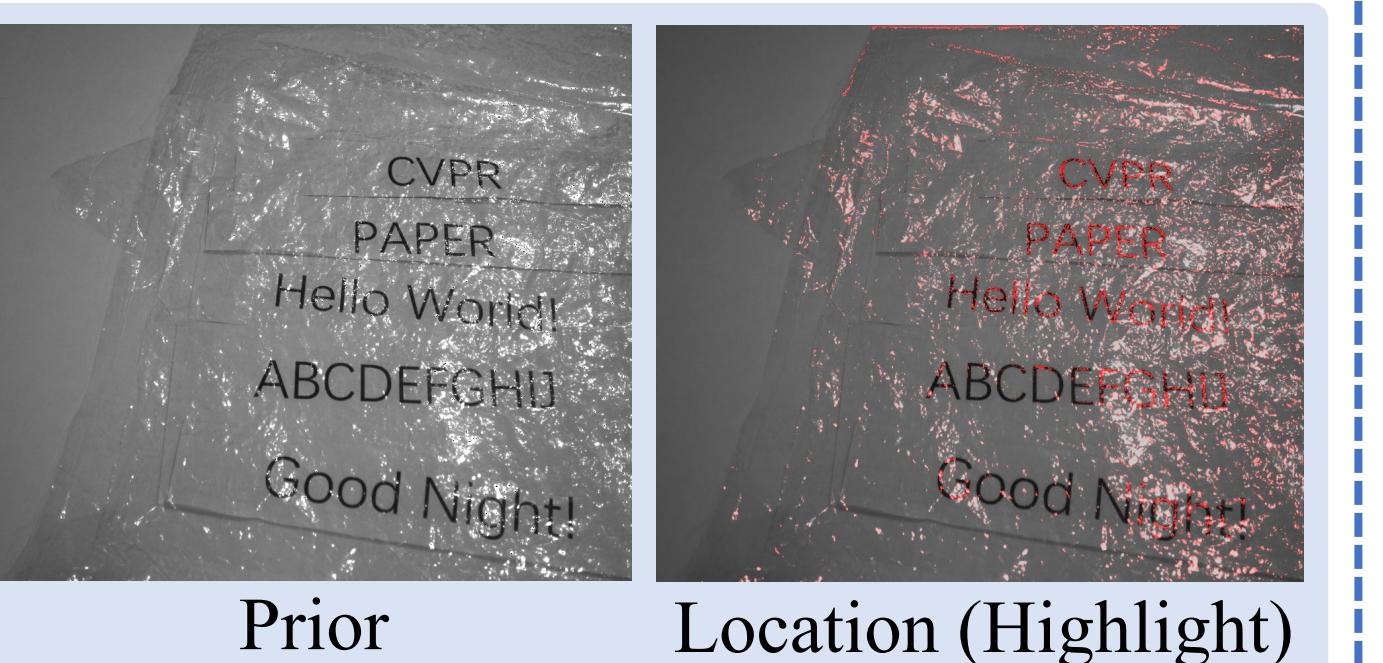
$$A = f_A(I_{input}^0 \oplus I_{input}^{45} \oplus I_{input}^{90} \oplus I_{input}^{135} \oplus S_{AoP} \oplus S_{DoP})$$



[Solution] Reconstructing I_m

- To recover the image, we set a reconstruction network f_r to decouple both I_d and I_h , with P .
- The reconstruction process can be expressed as:

$$I_{rec} = f_r(I_{input}^0 \oplus I_{input}^{45} \oplus I_{input}^{90} \oplus I_{input}^{135} \oplus P)$$



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 SSIM	23.47 0.7899	22.11 0.7640	21.95 0.7416	21.69 0.7439	21.77 0.7459	21.03 0.7465	20.86 0.7499	20.46 0.7412	21.10 0.7465	21.31 0.7300	21.58 $0.7499 \cdot 2.41 \times 10^{-4}$
Polar-HR [34]	PSNR SSIM	23.31 0.7642	22.80 0.7421	22.13 0.7220	21.58 0.7099	21.94 0.7064	22.00 0.7098	22.03 0.7128	21.99 0.7017	22.18 0.7102	21.95 0.6968	22.19 $0.7176 \cdot 3.80 \times 10^{-4}$
Uformer [33]	PSNR SSIM	31.85 0.9519	31.95 0.9456	31.39 0.9371	31.19 0.9364	31.81 0.9434	32.04 0.9421	31.68 0.9438	31.98 0.9435	31.85 0.9457	31.01 0.9363	31.68 $0.9426 \cdot 2.17 \times 10^{-5}$
Restormer [41]	PSNR SSIM	34.35 0.9771	35.02 0.9770	34.44 0.9721	33.71 0.9678	34.88 0.9757	35.13 0.9746	34.31 0.9742	34.33 0.9741	34.51 0.9759	32.49 0.9633	34.32 $0.9731 \cdot 1.75 \times 10^{-5}$
Ours	PSNR SSIM	36.76 0.9852	37.29 0.9859	36.62 0.9822	35.12 0.9867	36.93 0.9845	37.21 0.9833	36.24 0.9830	36.67 0.9836	36.94 0.9830	35.02 0.9749	36.48 $0.9824 \cdot 1.23 \times 10^{-5}$

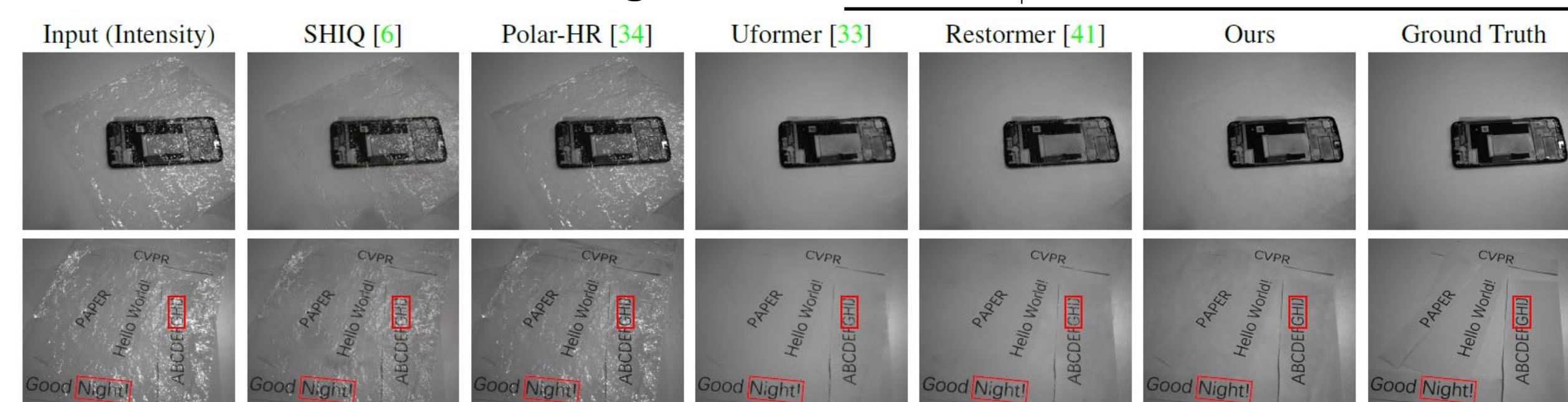
[Table 2]

Quantitative evaluation of QR Reading Rate.

	SHIQ [6]	Polar-HR [34]	Uformer [33]
Read Number	7	17	18
Read Rate	8.05%	19.54%	20.69%
	Restormer [41]	Ours	Ground Truth
Read Number	24	63	87
Read Rate	27.58%	72.41%	100%

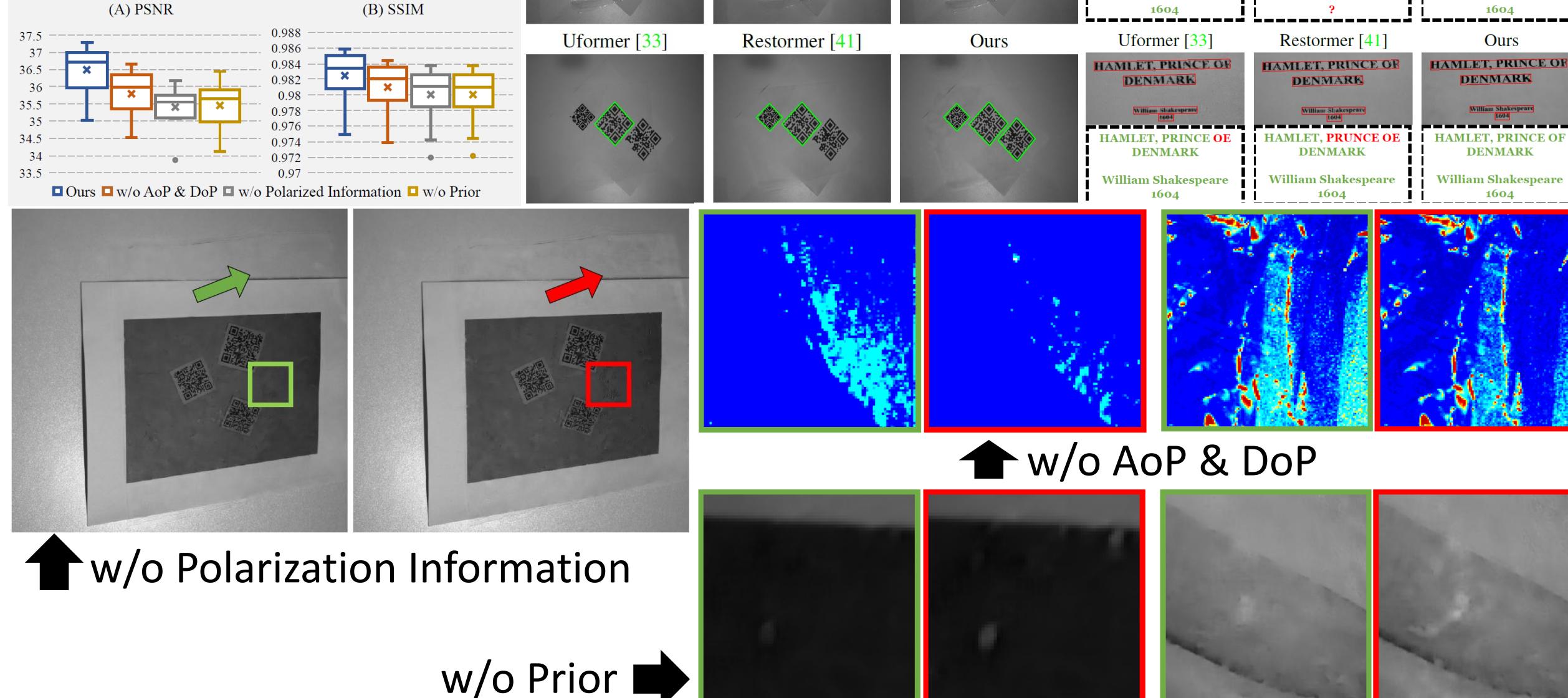
[Figure 1-3]

Qualitative Evaluation.



[Figure 4-6]

Ablation Study.



Acknowledgement

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