

Sticking Information in Plain Sight: Encoding and Detecting Hidden Stickers in the Real World

Christina Shatford and Szymon Rusinkiewicz
Princeton University

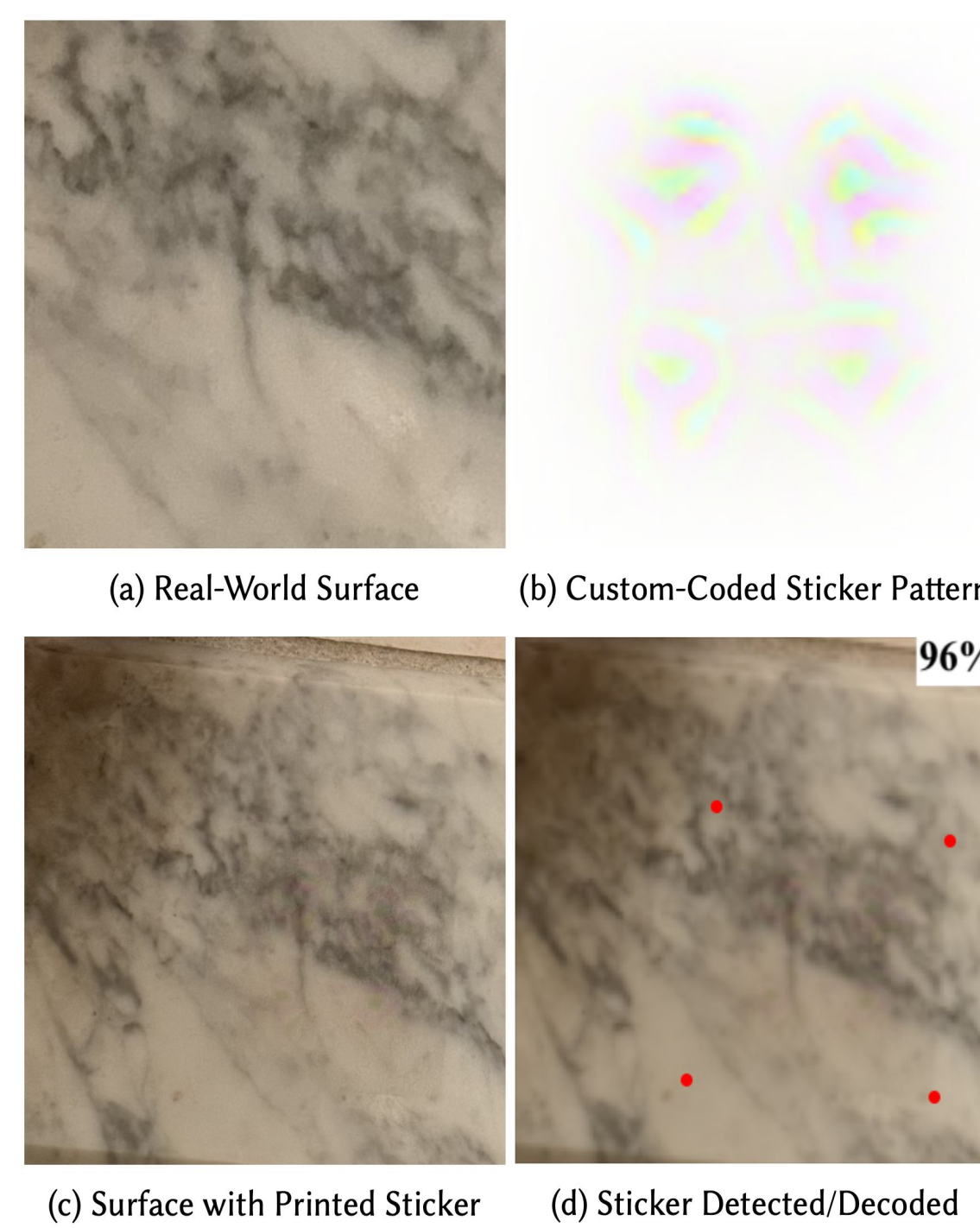


Motivation

- Being able to easily incorporate hidden data into the real world is advantageous in a number of scenarios that range from consumer-based interactions to industrial manufacturing.

Approach

- We optimize patterns, encoding 100 total bits, that are nearly **imperceptible** when **printed** on transparent sticker paper and applied to **real-world surfaces**. When photographed, encoded sections can be robustly localized and accurately decoded.
- Each pattern is generated specifically for a supplied surface location (*cover image*), resulting in a *container image*.
- Optimized to be resistant to **human error in sticker placement**, making it easy to deploy in the real world.



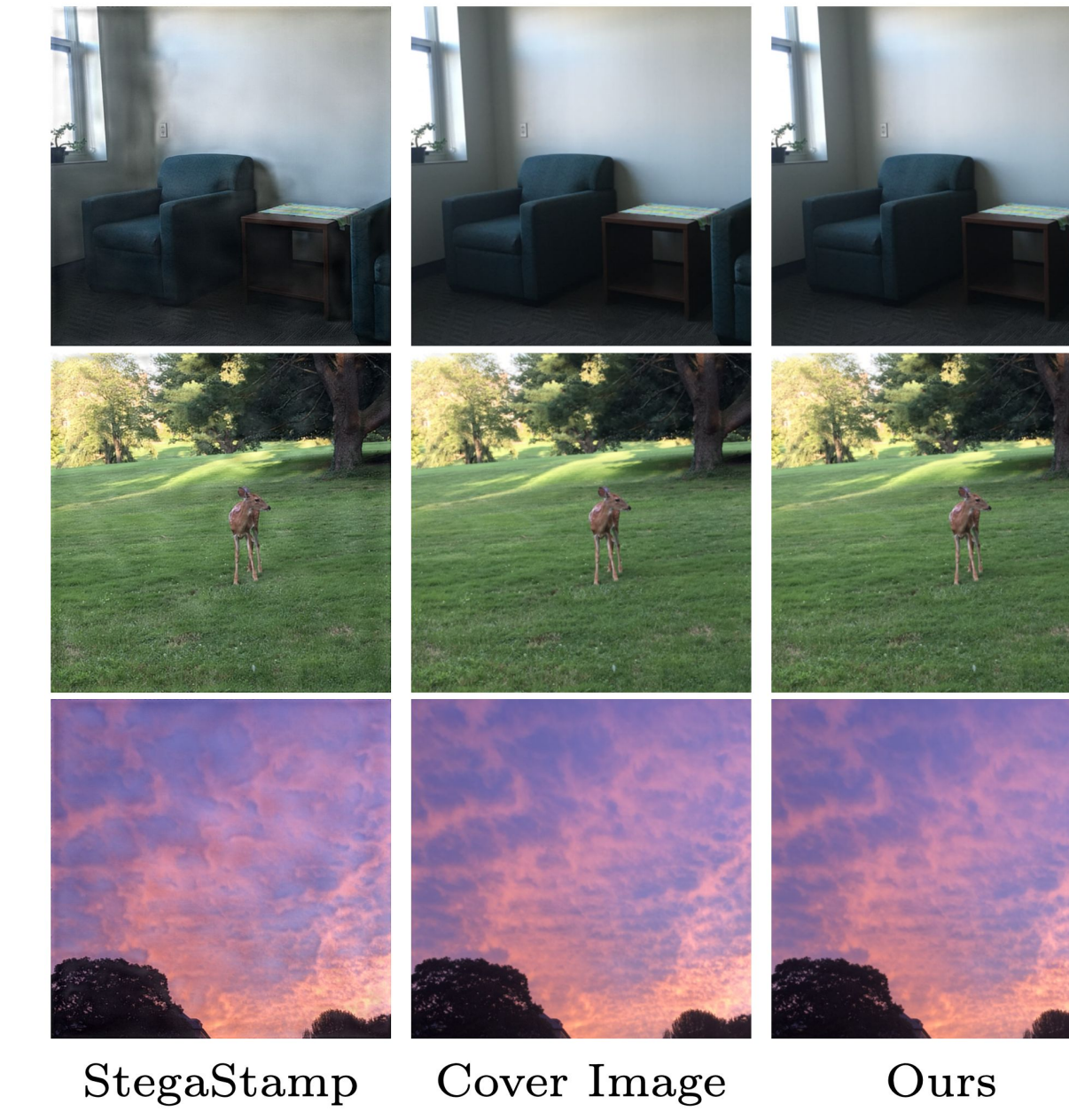
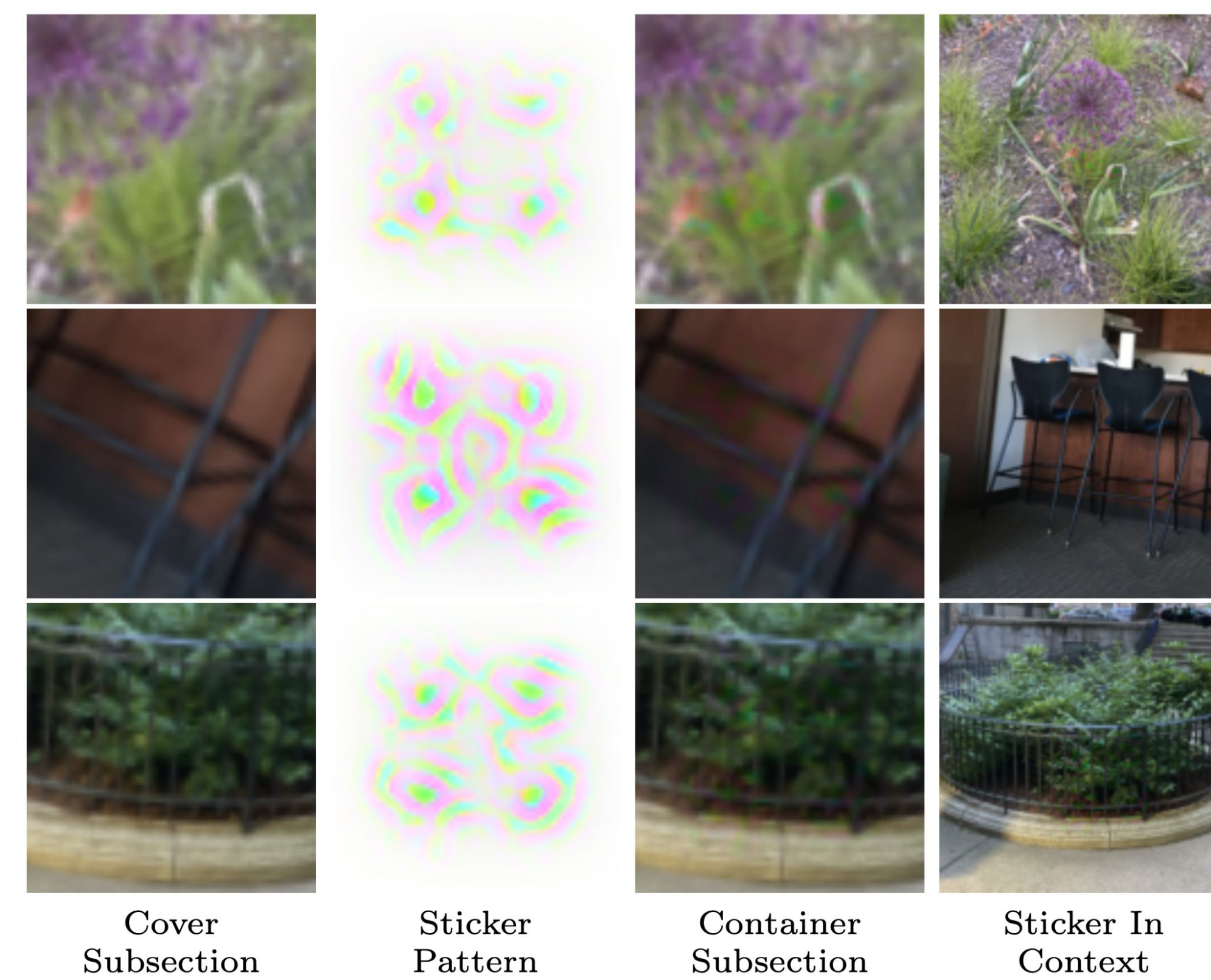
Methodology

- Encoder: UNet [2].
- Decoder: Convolutions followed by linear layers.
- Localizer: A variant of HRNet [4].
- Balance imperceptibility and robustness** by jointly optimizing the encoder, decoder and localizer.
- Simulate human placement error by introducing jitter between the cover subsection a pattern is optimized for, and the actual subsection where a sticker is placed.

Results

Simulation:

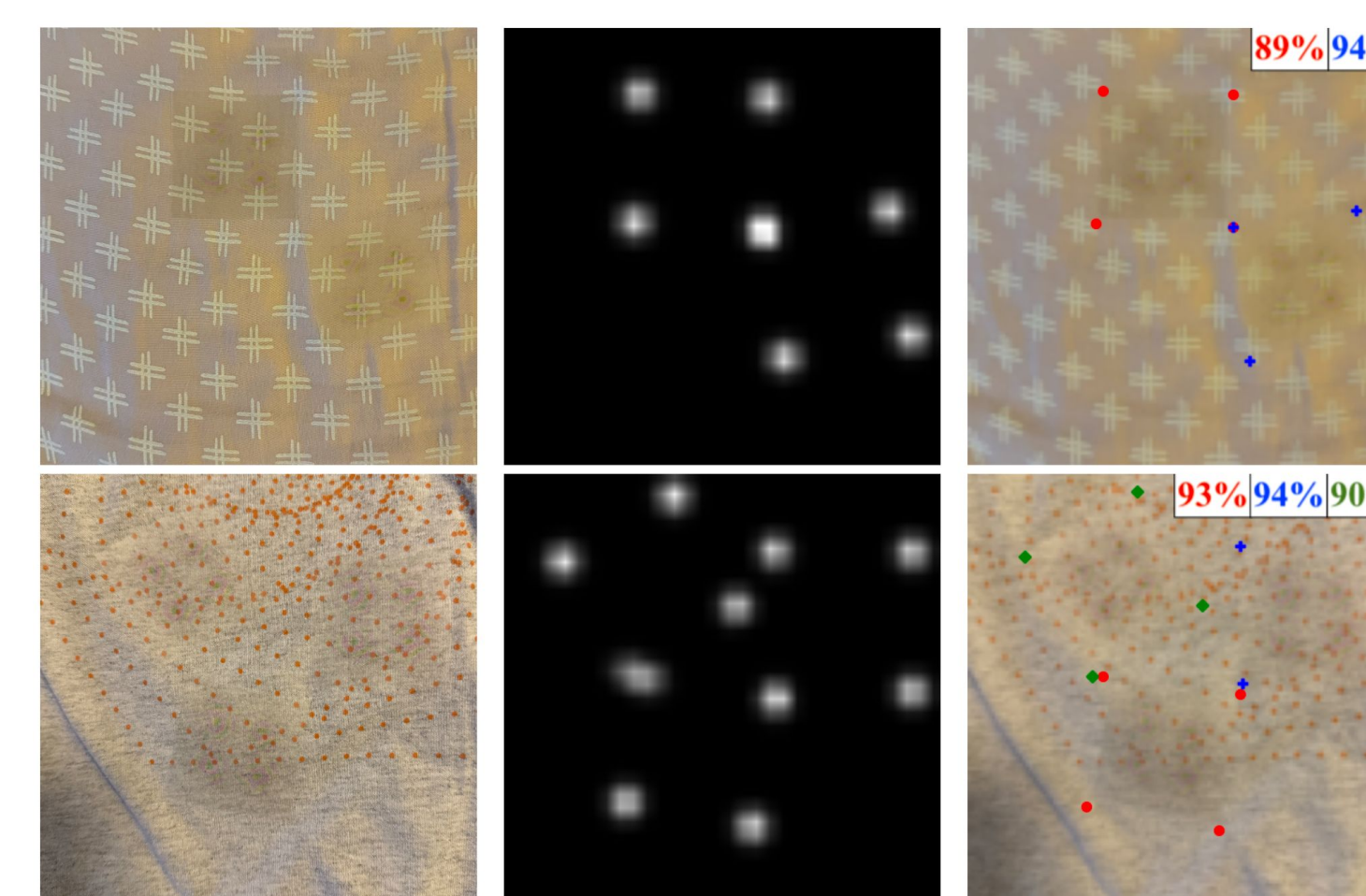
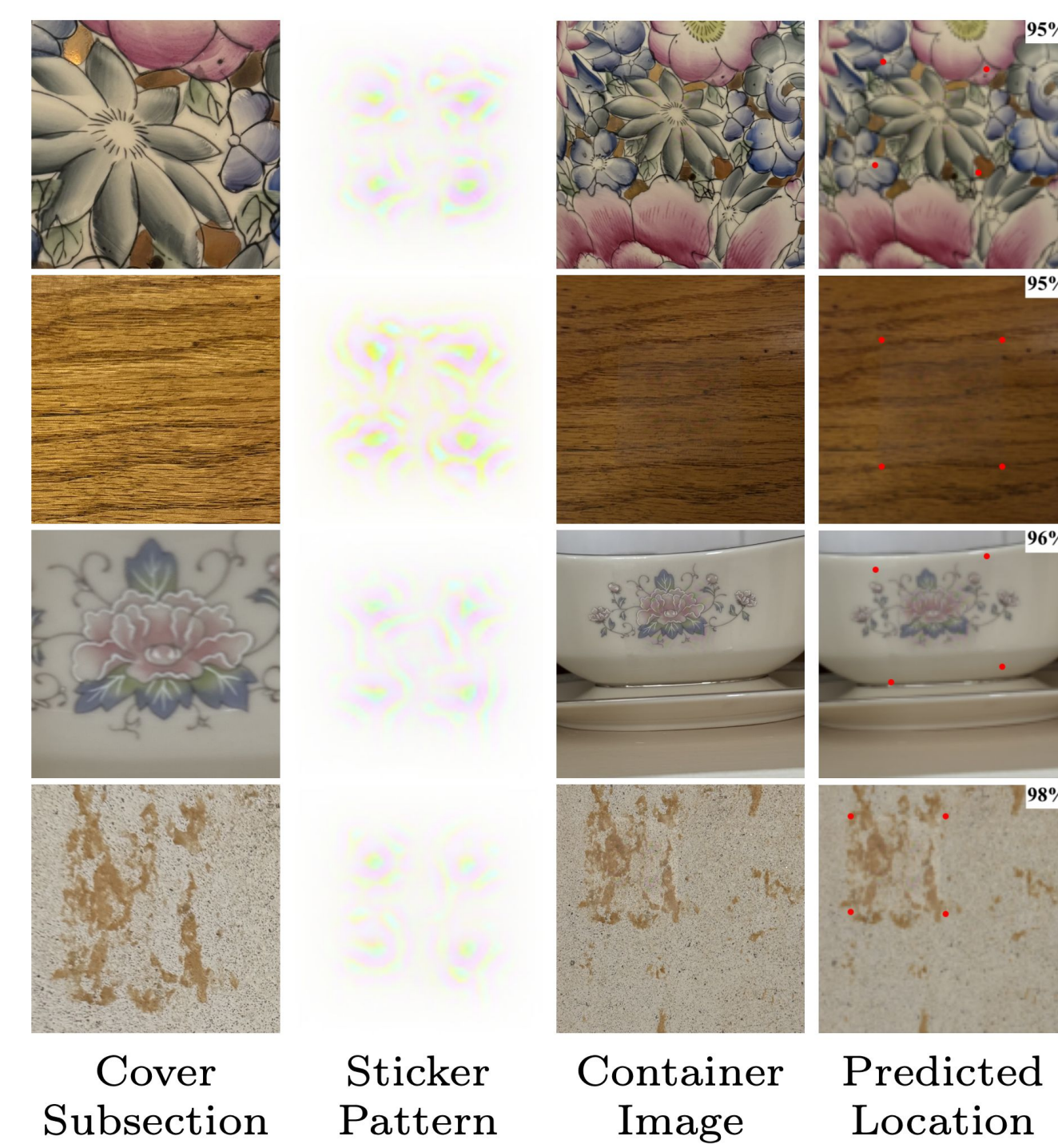
- Our method has average decoding accuracies of 99.14% for clean images and 89.16% with simulated noise, blur, color jitter, and perspective warp.
- A 0.0632 LPIPS score [5] demonstrates good imperceptibility.
- With perfect alignment, StegaStamp [3] achieves 2.40% higher decoding accuracy, but at the cost of greater perceptibility (0.0664 higher LPIPS).
- At 10 pixels of misalignment, StegaStamp achieves 17.32% **lower** decoding accuracy and 0.0999 higher LPIPS.



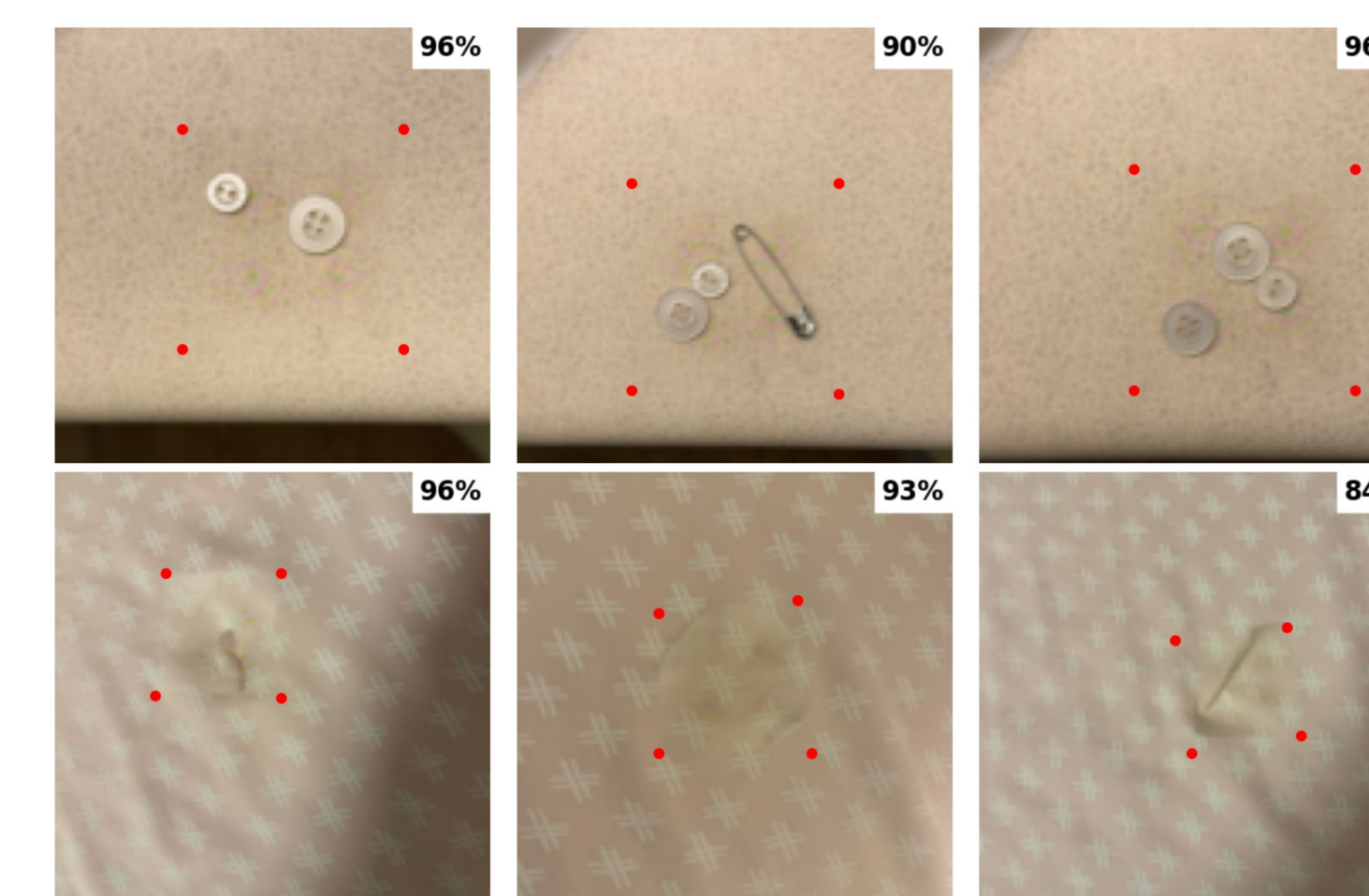
Method	Image Size	Encoding Size	PSNR $\uparrow$	SSIM $\uparrow$
Ours	400 $\times$ 400	400 $\times$ 400	32.27	0.9792
StegaStamp	400 $\times$ 400	400 $\times$ 400	28.60	0.8857
Ours	170 $\times$ 170	64 $\times$ 64	40.62	0.9922
Jia <i>et al.</i> [1]	256 $\times$ 256	96 $\times$ 96	32.95	0.9677

Real-world:

- Dataset of 34 encoded surfaces, 84 images total, with a variety of imaging conditions.
- Average decoding accuracy of 93.10% and median accuracy of 94%.

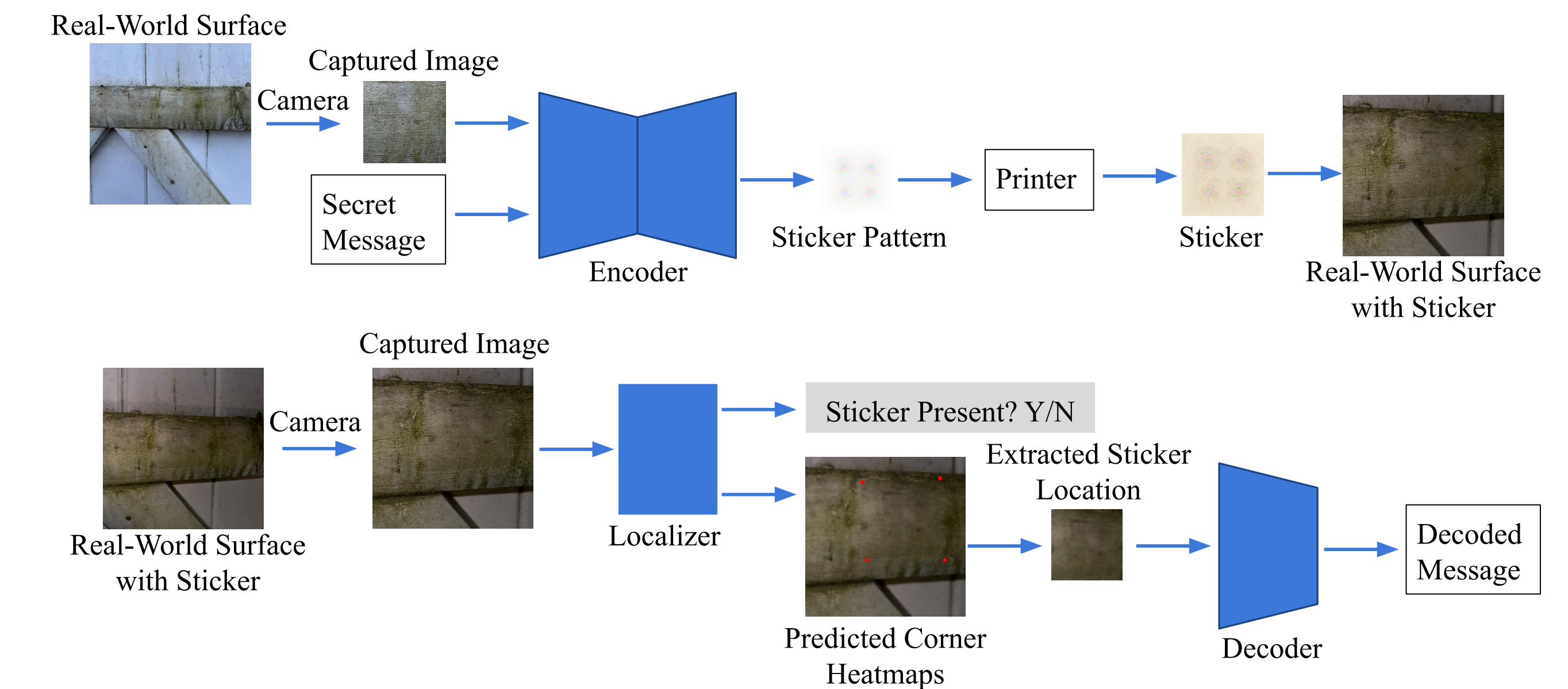


Multi-sticker detection



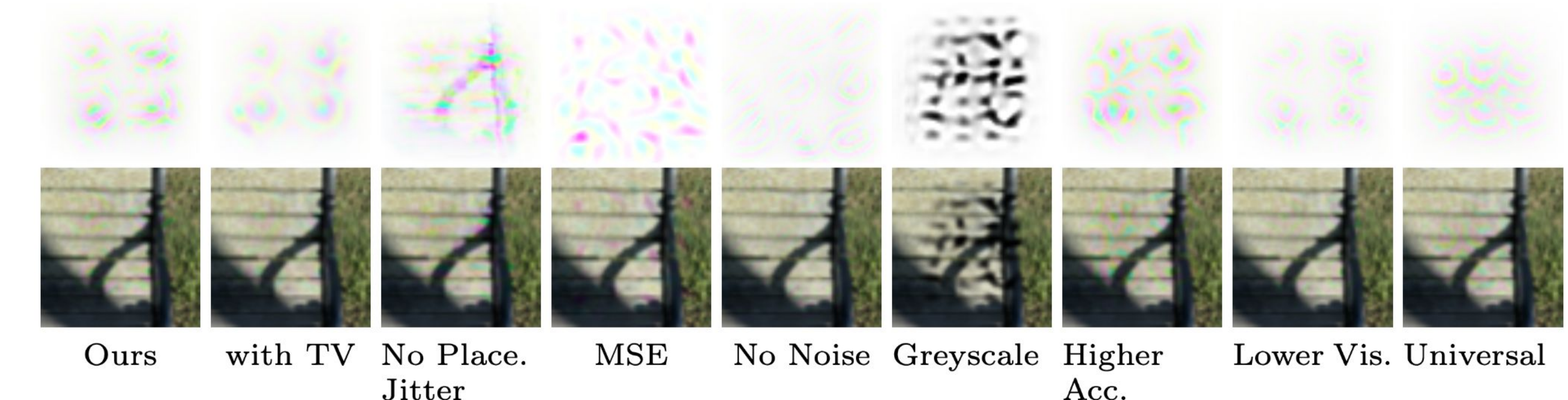
Robust to occlusion and wrinkles

Pipeline



Ablations

- Experimented with how removing or modifying components of the loss function influenced our evaluation metrics.



Method	Decode Acc. $\uparrow$	LPIPS $\downarrow$	Class. Acc. $\uparrow$
Ours	89.16%	0.0632	97.24%
With TV	84.20%	0.0428	94.04%
No Placement Jitter	86.85%	0.0557	96.69%
MSE	89.11%	0.1111	95.76%
No Noise	90.22%	0.0125	46.45%
Greyscale	93.79%	0.2132	99.81%
Higher Accuracy	94.80%	0.1099	98.10%
Lower Visibility.	81.93%	0.0290	89.73%
Universal	87.89%	0.0667	94.36%

References

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