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Colon and Splicing

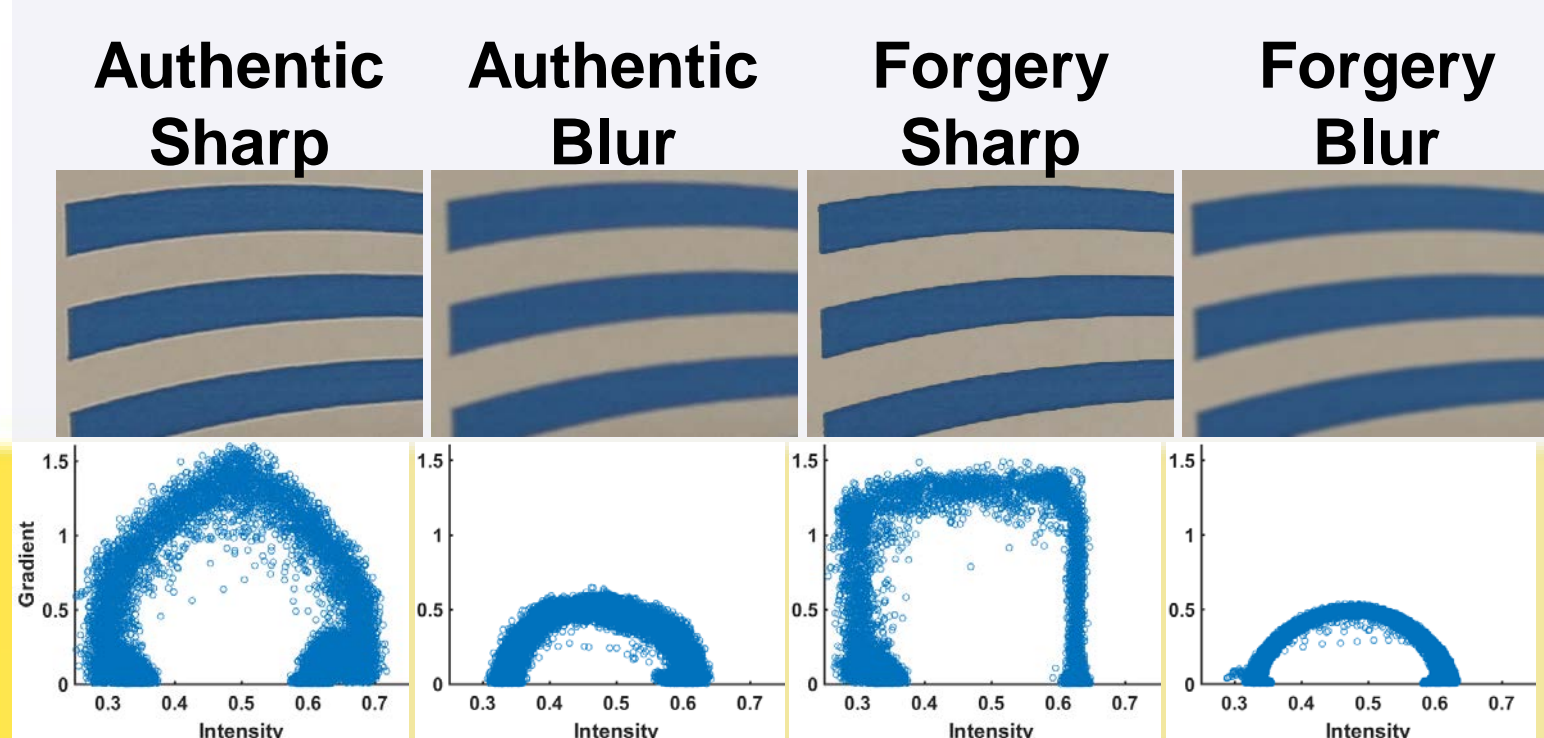
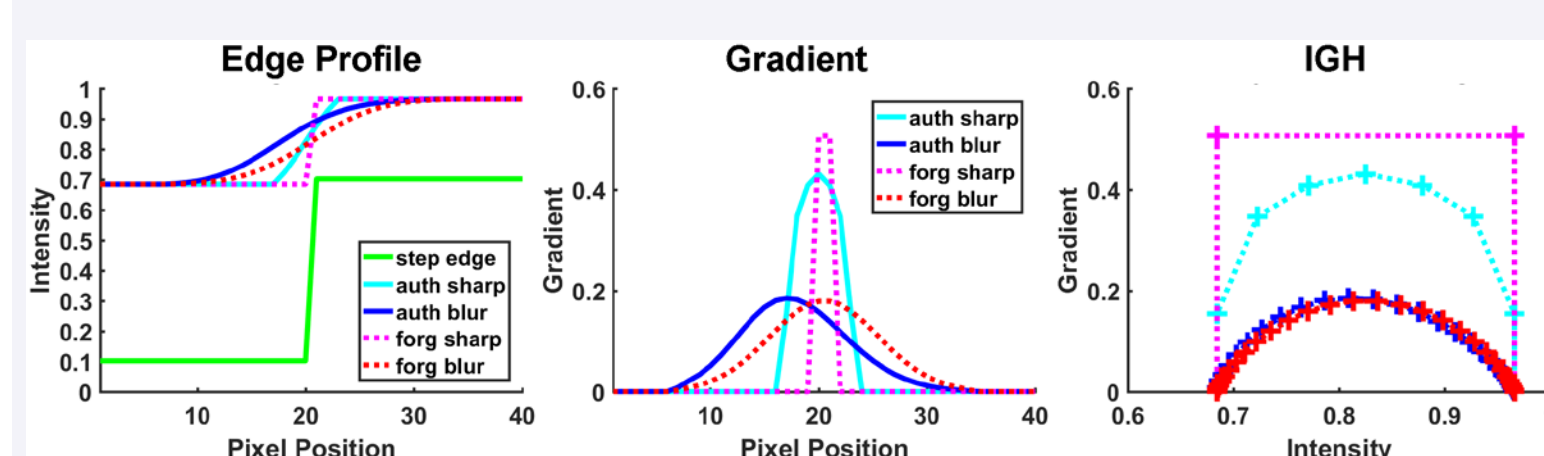
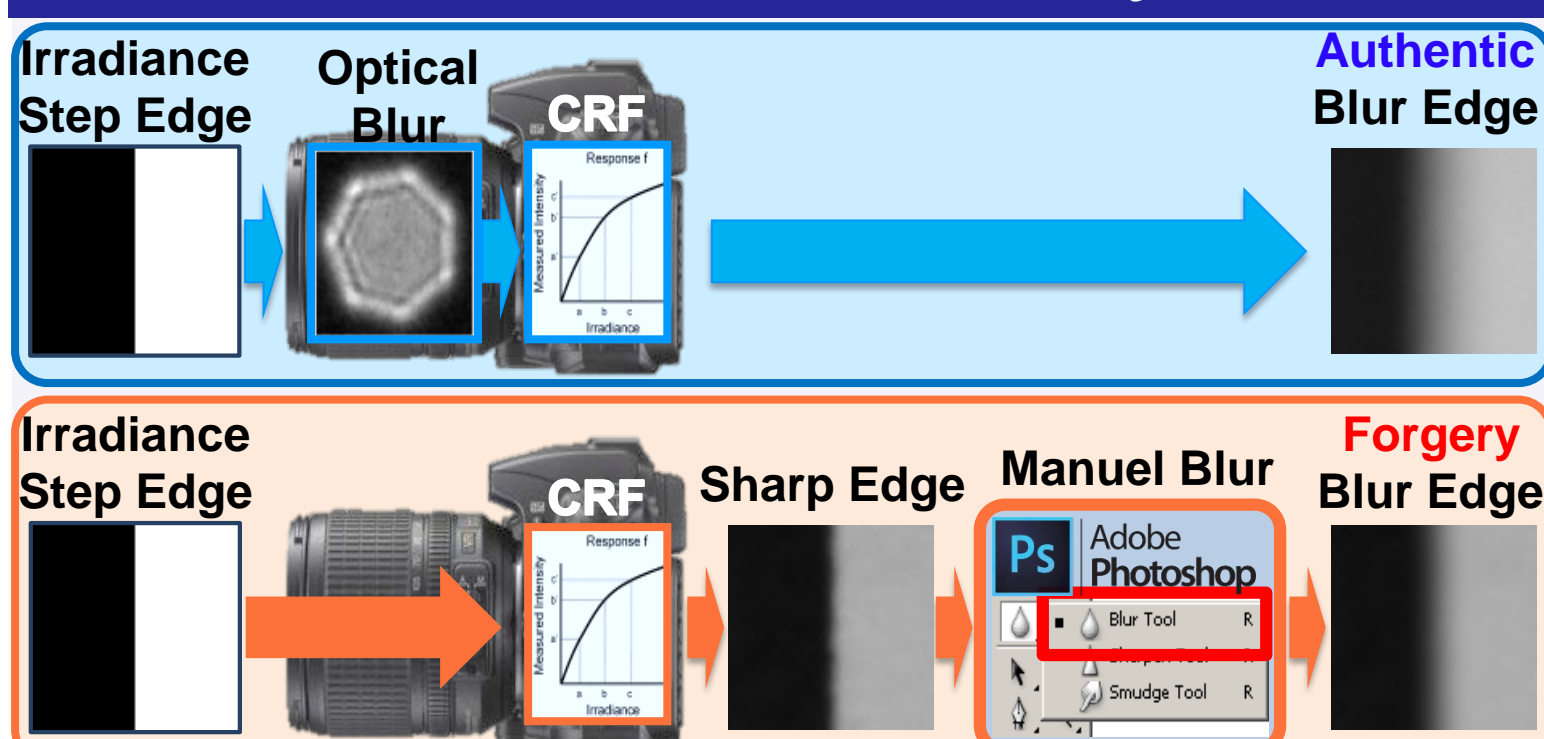
- Splicing: contents are extracted from one image and then copied into a new image.
- Fake blur to hide and smooth harsh boundary.



Previous Splicing Detection Approaches

- CRF Inconsistency**
 - Check the estimated CRFs consistency between the image segments.
 - No guarantee CRF estimation accuracy;
 - Can't detect segments from same camera or blurred splicing boundary.
- Blur Inconsistency**
 - Check blur type (motion or defocus), the extent of blur, or direction of blur.
 - No guarantee blur kernel estimation accuracy;
 - High false alarm rate with spatially-varying blurred authentic images.

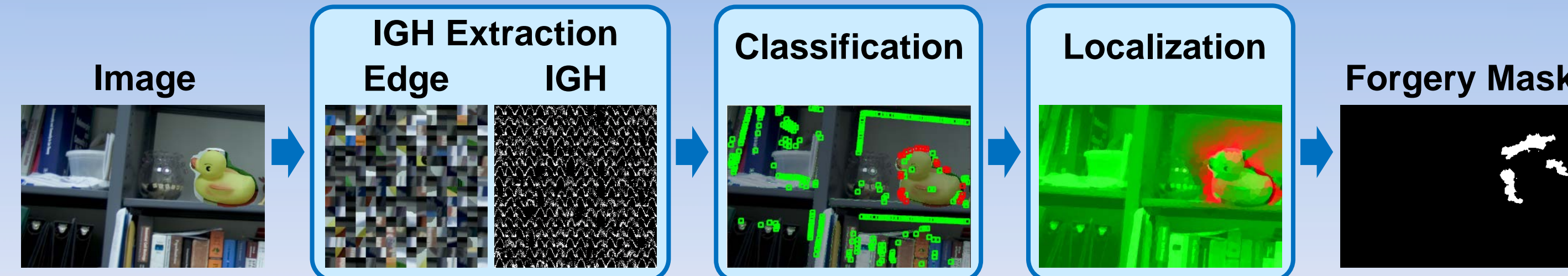
Intensity Gradient Histogram



- CRF is a non-linear operator, it's not commutative.
- $\text{CRF}(\text{Blur}(\text{edge})) \neq \text{Blur}(\text{CRF}(\text{edge}))$

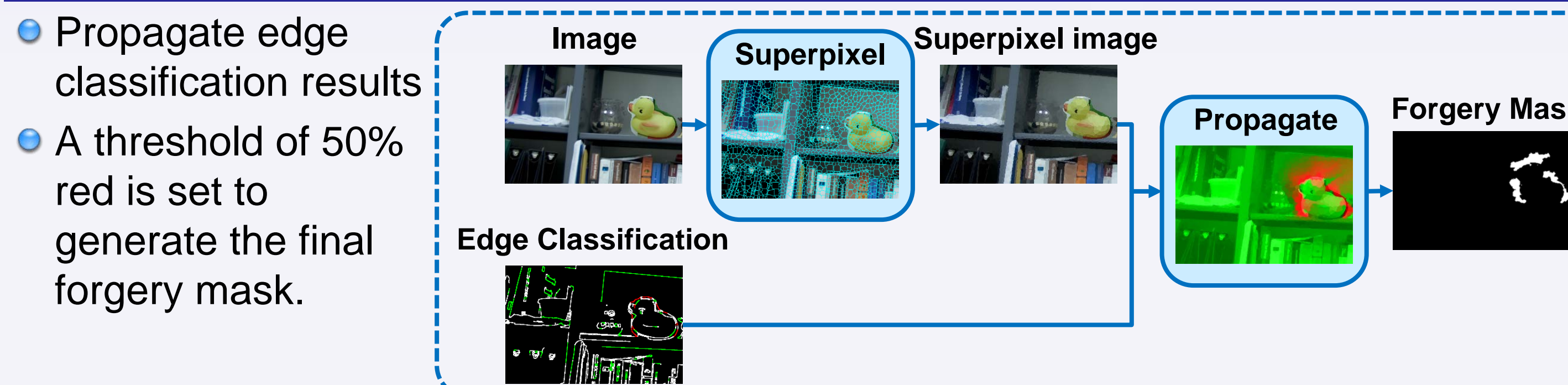
- Authentic Blur**
 $I_{ab} = f(K_g(x) * H_{step}(x))$
 $\nabla I_{ab} = (b - a) f'(f^{-1}(I_{ab})) \cdot K_g(x - c)$
- Authentic Sharp**
 $I_{as} = f(K_s(x) * H_{step}(x))$
 $\nabla I_{as} = (b - a) f'(f^{-1}(I_{ab})) \cdot K_s(x - c)$
- Forgery Blur**
 $I_{fb} = K_g(x) * f(H_{step}(x))$
 $\nabla I_{fb} = (b - a) f'(b) \cdot K_g(x - c)$
- Forgery Sharp**
 $I_{fs} = f(H_{step}(x))$

Our Pipeline



- Color Variance Check**
 - $\text{variance}(C_L) < \text{thresh}_{cv}, \text{variance}(C_H) < \text{thresh}_{cv}$
- Color Difference Check**
 - $|\text{mean}(C_H) - \text{mean}(C_L)| < \text{thresh}_{cd}$
- Edge Range Check**
 - $\max(C_E) < \max(C_H), \min(C_E) < \min(C_L)$
- Area Difference Check**
 - $\|C_H\| - \|C_L\| < 0.5 * (\|C_H\| + \|C_L\|)$

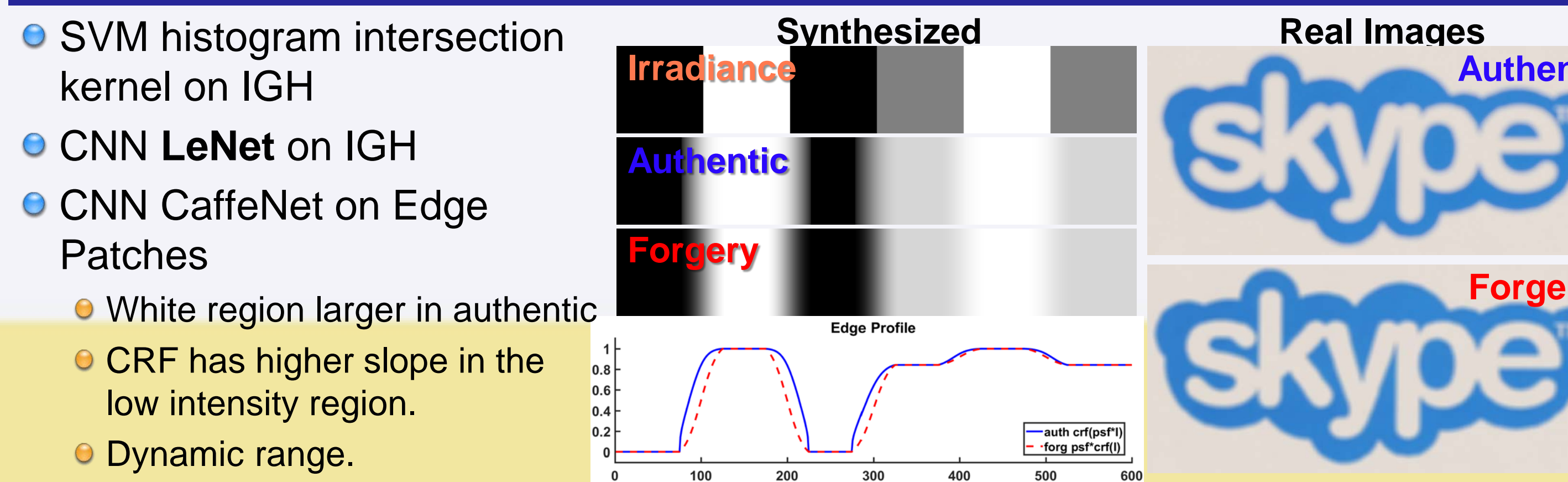
Mask Generation



Handling Large Blur

- Multi-scale scheme in case of large amounts of blur.
 - Kernel size may be larger than some of the structures in the image.
 - Downsample to $\frac{1}{4}$.

Classifying IGH



Experimental Results

- Splicing Logo Dataset (**SpLogo**) containing 1533 authentic images and 1277 forged blur images of logos with different colors and different amounts of blur.

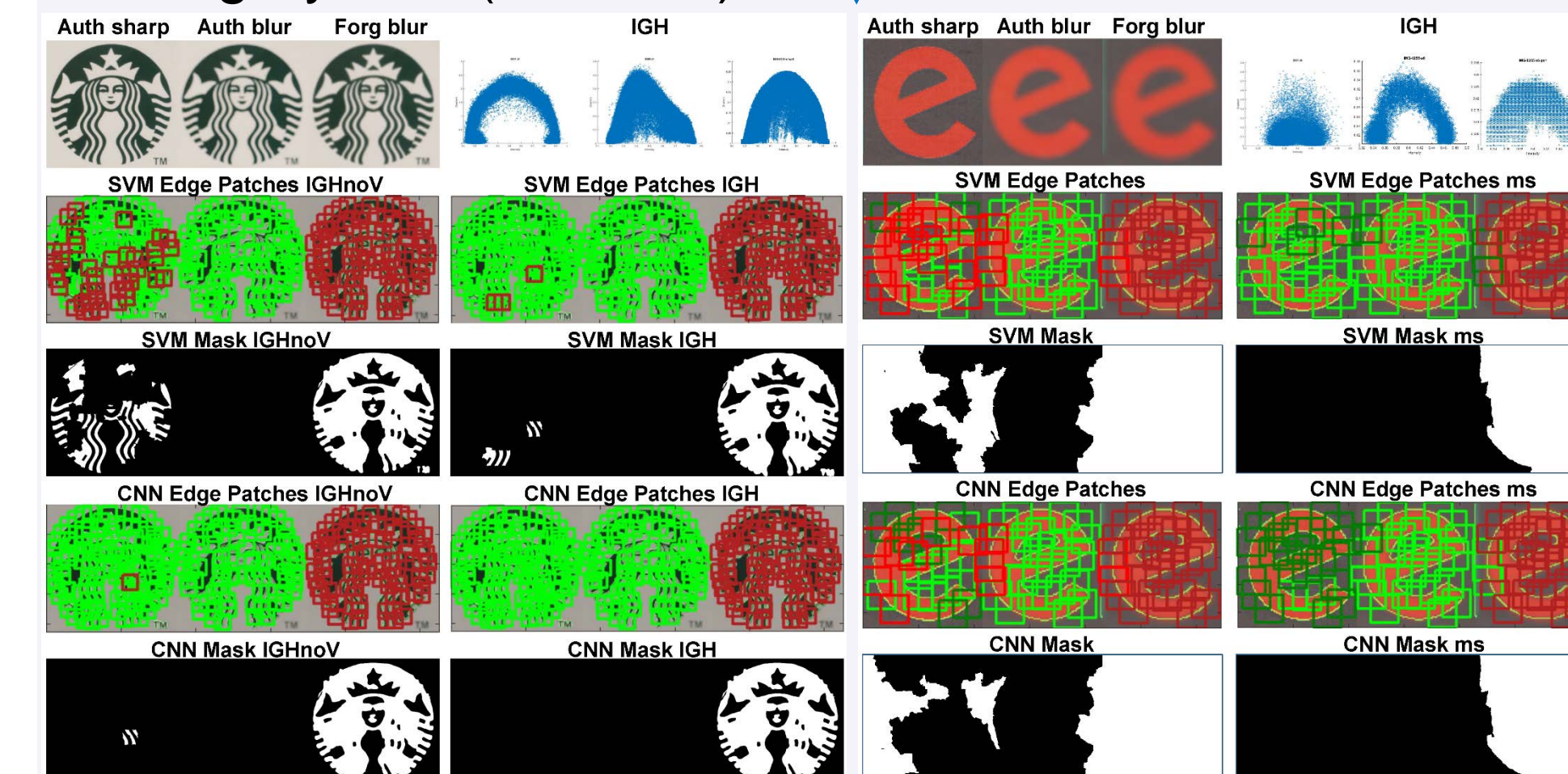
Table 1. Patch Classification Accuracy.

Classifier	Image			IGHnoV			IGH		
	CUISDE	SpLogo	Combine	CUISDE	SpLogo	Combine	CUISDE	SpLogo	Combine
SVM	-	-	-	0.924	0.972	0.937	0.940	0.972	0.951
CNN	0.888	0.896	0.891	0.943	0.972	0.979	0.97	0.99	0.978

Table 2. Patch Accuracy on CUISDE.

Classifier	Precision	Recall	Accuracy
CRF SVM [15, 13]	0.7	0.7	0.87
IGH SVM	0.93	0.94	0.94
IGH CNN	0.95	0.95	0.97

- Authentic Sharp (dark green)
- Authentic Blur (bright green)
- Forgery Sharp (bright red)
- Forgery Blur (dark red)



- Both SVM classification and CNNs applied to the patches give false negatives on forged sharp edges, which are reduced significantly by applying the CNN to our IGH feature.

