

Digital Image Manipulation

Anti-forensics

Previous Works

- [1] H. Farid, "Exposing digital forgeries by detecting traces of **resampling**," IEEE Trans on SP, 2005.
- [2] B. Mahdian, "Blind authentication using periodic properties of interpolation," IEEE TIFS, 08.

- [3] J. Huang, "A novel method for detecting cropped and re**compressed** image block", ICASSP 2007.
- [4] H. Farid, "Exposing digital forgeries from JPEG ghosts," IEEE TIFS, 09.

- [5] K. J. R. Liu, "Blind forensics of **contrast enhancement** in digital images," in ICIP, 2008.
- [6] K. J. R. Liu, "Forensic detection of image tampering using intrinsic statistical fingerprints in histograms," Proc.APSIPA,2009.
- [7] K. J. R. Liu, "Forensic estimation and reconstruction of a contrast enhancement mapping," ICASSP 2010.
- [8] K. J. R. Liu, " Forensic detection of image forgeries using intrinsic fingerprints of pixel value mappings ", IEEE TIFS, 2010.
- [9] Y. Zhao, "Forensic estimation of gamma correction in digital images," ICIP, 2010.

- [10] D. Hsiao and S. Pei, "Detecting digital tampering by **blur** estimation," 1st SADFE, 2005.

- [11] Y. Zhao, "Detection of image **sharpening** based on histogram aberration and ringing artifacts," ICME, 2009.
- [12] Y. Zhao, "Unsharp Masking Sharpening Detection via Overshoot Artifacts Analysis", *IEEE SPL*, 2011.

- [13] M. Kirchner and J. Fridrich, "On detection of **median filtering** in digital images," SPIE, 2010.
- [14] Y. Zhao, "Forensic detection of median filtering in digital images," ICME, 2010.

- [15] N. Memon, "Image manipulation detection," Journal of Electronic Imaging, 2006.
- [16] M. Wu, "**Tampering identification** using empirical frequency response," ICASSP, 2009.

Outline

Anti-forensics of Contrast Enhancement

Attacks on Contrast Enhancement Forensics

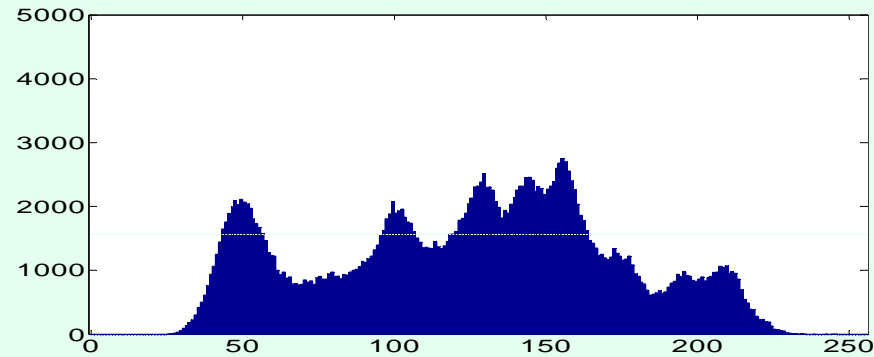
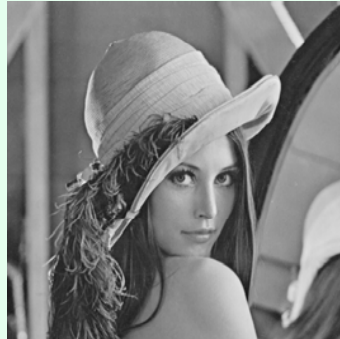
➤ Existing anti-forensic techniques

- ✓ Target: (1) undetectable image *resampling*,
(2) synthesis of *color filter array pattern*,
(3) anti-forensic of JPEG *compression*.
- ✓ Type: manipulation hiding attacks.

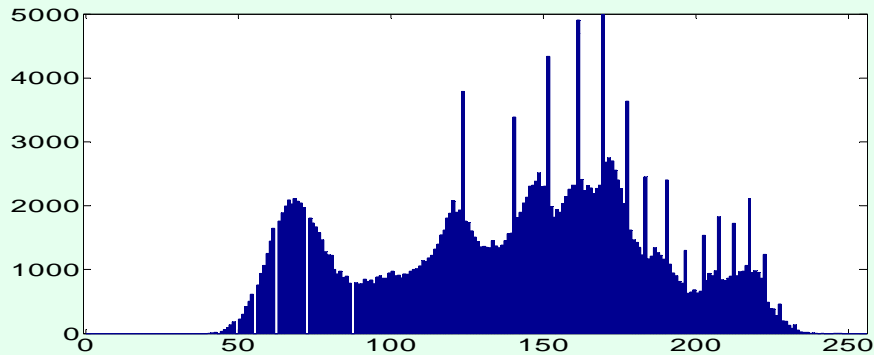
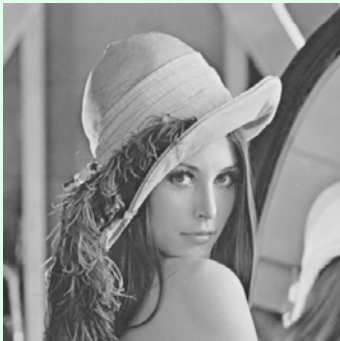
➤ Proposed attacks

- ✓ Target: undetectable *contrast enhancement* (CE)
- ✓ Type: manipulation forging & hiding attacks.

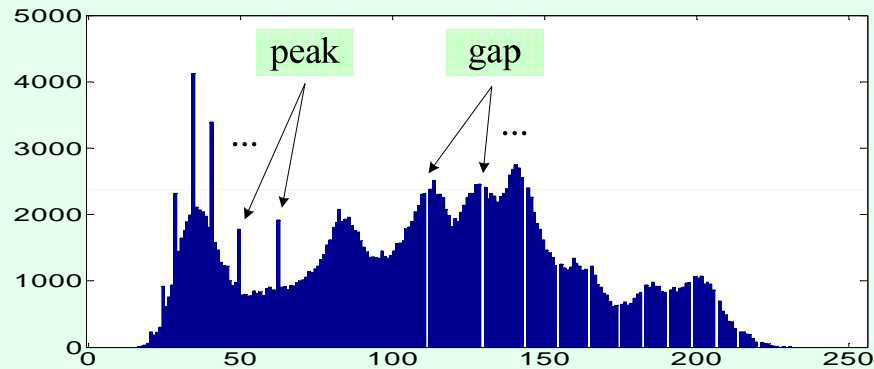
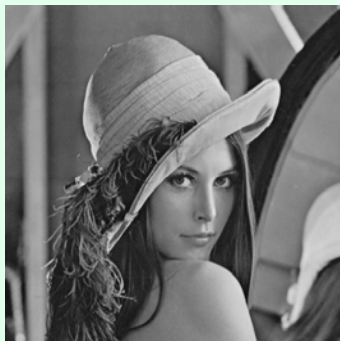
Forensic Detection of CE (1)



$\gamma = 0.8$



$\gamma = 1.2$



Forensic Detection of CE (2)

➤ **CE detection algorithm** (IEEE TIFS, 2010)

- 1). Obtain the test image's histogram $h(x)$.
- 2). Calculate $g(x)$ as follows,

$$g(x) = p(x) \cdot h(x)$$

where $p(x)$ is a pinch off function for combating the false high frequency effect caused by high end and low end saturated images.

- 3). Compute the **high frequency measure** F :

$$F = \frac{1}{N} \sum_{\omega} |\beta(\omega)G(\omega)| \quad G(\omega) = \text{DFT}(g(x)) \quad \beta(\omega) = \begin{cases} 1 & |\omega| \geq c \\ 0 & |\omega| < c \end{cases}$$

where c is a user specified cutoff frequency.

- 4). Thresholding method to determine if contrast enhancement has been performed, F greater than the threshold signifying the detection of contrast enhancement.

Forensic Detection of CE (3)

➤ **Detection of locally applied CE** (IEEE TIFS, 2010)

- ✓ The above forensic technique is extended to detect **locally applied contrast enhancement**.

➤ **Forensic estimation of CE mapping** (ICASSP, 2010)

- ✓ An iterative algorithm to jointly **estimate any arbitrary CE mapping** as well as the pixel value histogram of the image before contrast enhancement.
- ✓ It should be noted that the mapping estimation can be reliable only when the employed CE manipulations have been detected previously. Otherwise, the estimation is **eyeless** and couldn't be alleged as forensic reconstruction of CE.

CE Forging Attack (1)

➤ Locate Peak & Gap Bins

- **Universal CE forging**

- ✓ Position of the peak/gap bins are designated randomly.

- **Targeted CE forging**

- ✓ Given the targeted mapping $y = m(x)$ which is to be falsified, we can compute the position set of peak and gap bins which should appear in the histogram of attacked original image.

$$\Psi_p = \left\{ y \mid \sum_{x \in \Omega} l(m(x) = y) > 1 \right\} \quad x \in \Omega \quad y \in \Omega$$
$$\Psi_g = \left\{ y \mid \sum_{x \in \Omega} l(m(x) = y) = 0 \right\} \quad \Omega = \{0, 1, \dots, 255\}$$

CE Forging Attack (2)

➤ Generate Peak Bin by Moving Histogram

- ✓ Invariance of the pixel number within local histogram:

$$M \cdot 2L + \lambda \cdot M = \sum_{x=x_i-L}^{x_i+L} h_X(x)$$

where $x_i \in \Psi_p$ $M \triangleq \frac{1}{2L} \left(\sum_{y \in \Theta} h_Y(y) \right)$ $\Theta = [x_i - L, x_i) \cup (x_i, x_i + L]$

- ✓ Generated peak bin: $h_Y(x_i) = \frac{\lambda}{2L + \lambda} \sum_{x=x_i-L}^{x_i+L} h_X(x)$

- ✓ Decrease of the neighboring bins: $\Delta_y = w(y - x_i) \cdot (h_Y(x_i) - h_X(x_i)) \quad y \in \Theta$

where $w(x) \sim N(0, \sigma)$

CE Hiding Attack (1)

➤ Mapping Decomposition

- Traditional CE : $m(x) = \text{round} [m_0(x)]$
 - ✓ Step (1): $y_0 = m_0(x)$ (the primary mapping) $m_0 : \mathbb{Z} \rightarrow \mathbb{R}$
 - ✓ Step (2): $y = \text{round} [y_0]$

- Histogram ~ CE:
$$h_Y(y) = \sum_{x \in \Omega} h_X(x) l(y_0 \in [y - \frac{1}{2}, y + \frac{1}{2}))$$
 - ✓ Potential peak & gap:
$$\Psi_p = \left\{ y \mid \sum_{x \in \Omega} l(y_0 \in [y - \frac{1}{2}, y + \frac{1}{2})) > 1 \right\}$$
$$\Psi_g = \left\{ y \mid \sum_{x \in \Omega} l(y_0 \in [y - \frac{1}{2}, y + \frac{1}{2})) = 0 \right\}$$

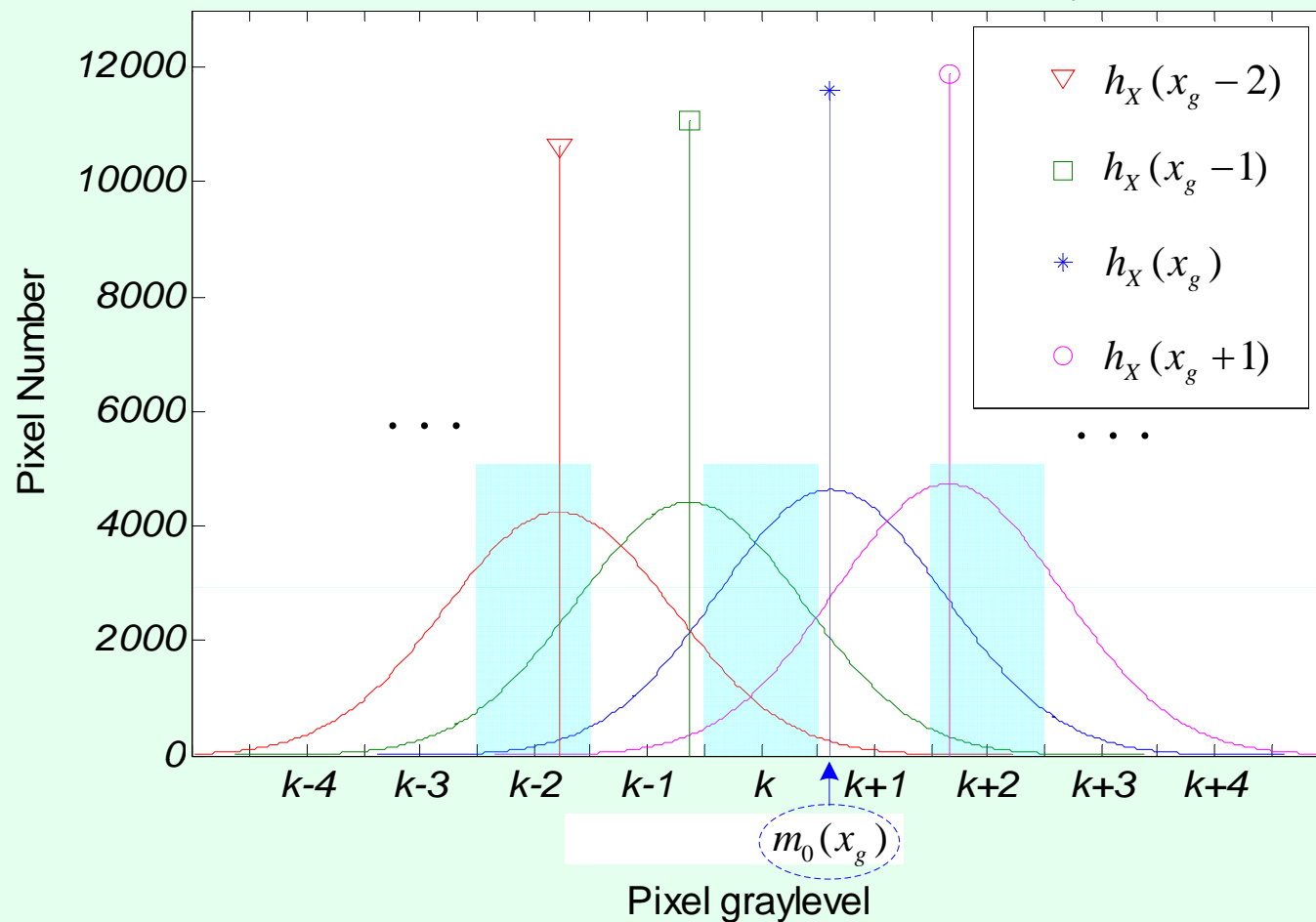
CE Hiding Attack (2)

➤ Local Random Dithering (Integrated Attack)

$$m(x) = \text{round} [m_0(x) + n_x]$$

$$n_x \sim N(0, \sigma_x^2)$$

$$\sigma_x = \begin{cases} \sigma_1 & \text{if } \text{round}[m_0(x)] \in \{\hat{\Psi}_p \cup \hat{\Psi}_g\} \\ \sigma_2 & \text{else} \end{cases}$$



CE Hiding Attack (3)

➤ Analysis for Artifacts Removal

➤ New primary mapped value: $t_{x_g} = m_0(x_g) + n_{x_g}$

$$t_{x_g} \sim N(m_0(x_g), \sigma_x^2)$$

➤ Histogram ~ new CE

$$h_Y(k+i) = \sum_{j=-\infty}^{+\infty} h_X(x_g + j) \int_{k+i-\frac{1}{2}}^{k+i+\frac{1}{2}} G_t \left(m_0(x_g + j), \sigma_{x_g}^2 \right) dt$$

$$\text{where } G_t \left(m_0(x_g + j), \sigma_{x_g}^2 \right) = \frac{1}{\sqrt{2\pi}\sigma_{x_g}} \exp \left[-\frac{(t - m_0(x_g + j))^2}{2\sigma_{x_g}^2} \right]$$

CE Hiding Attack (4)

➤ Adding Noise (Postprocessing Attack)

➤ Formulated as:

$$\begin{aligned} y &= \text{round} [m(x) + n_x] \\ &= \text{round} [\text{round} [m_0(x)] + n_x] \end{aligned}$$

$$\text{where } n_x \sim N(0, \sigma_x^2) \quad \sigma_x = \begin{cases} \sigma_1 & \text{if } \text{round} [m_0(x)] \in \{\hat{\Psi}_p \cup \hat{\Psi}_g\} \\ \sigma_2 & \text{else} \end{cases}$$

- ✓ If unknowing the used mapping function, the potential peak set Ψ_p and the potential gap set Ψ_g have to be searched by applying a simple and strict threshold-based detector.

Evaluation – *Test Data*

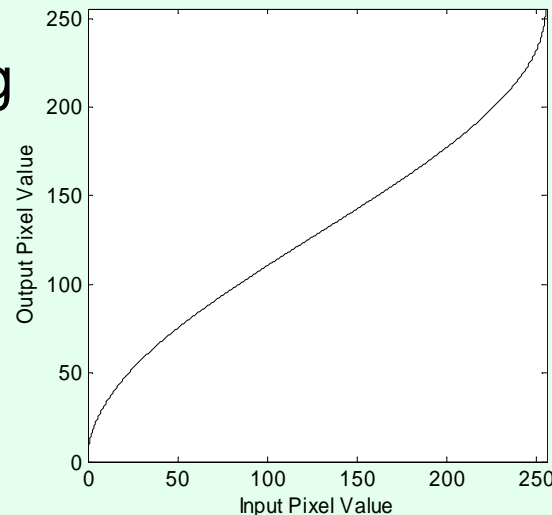
➤ 693 unaltered photograph images

- ✓ Size: $1200 \times 900 \sim 2832 \times 2128$ pixels
- ✓ JPEG
- ✓ Use green channel as grayscale test image

➤ Simulate traditional CE manipulations

- ✓ Gamma correction: $m(x) = \text{round} \left[255 \left(\frac{x}{255} \right)^\gamma \right]$

- ✓ 'S' mapping



- ✓ Parameter Setting:

$$\gamma = 0.5 \sim 2 \quad c = 7\pi / 8$$

$$\sigma_1 = \sigma_2 = 1.0$$

Evaluation – *Performance Metric*

➤ (un)Detectability : Two Sets

- ✓ Training Set : 347 (random chosen)
 - FN_0 、 $FP_0 \rightarrow F$ - threshold
- ✓ Test Set : 346
 - F - threshold \rightarrow TN 、 TP

➤ Image Quality of Attacked Images

- ✓ PSNR
 - Forging attack: attacked image **vs.** unaltered image
 - Hiding attack: attacked image **vs.** contrast-enhanced image

Baseline Detection Results (1)

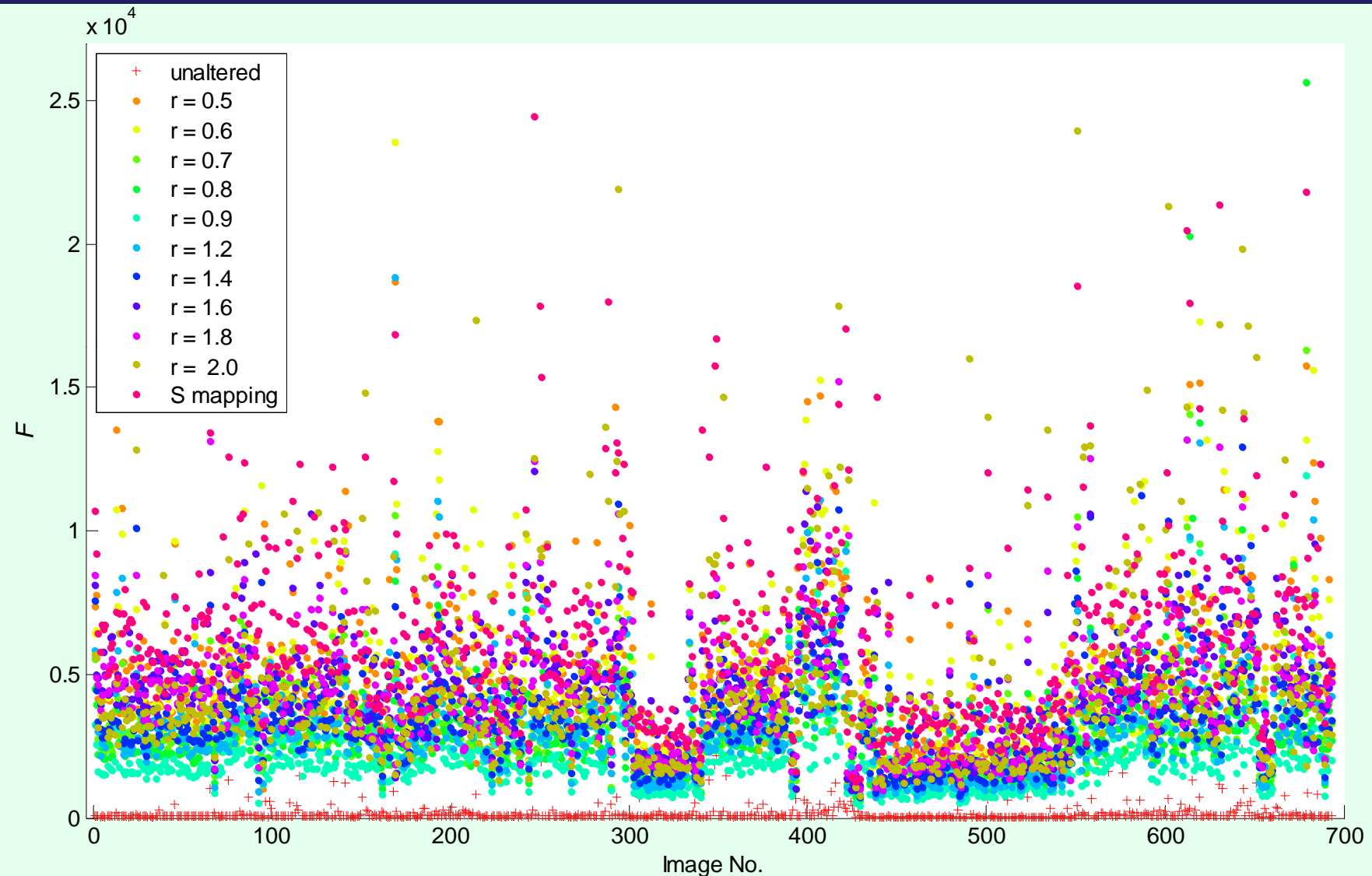


Fig. Feature value distribution of samples.

Baseline Detection Results (2)

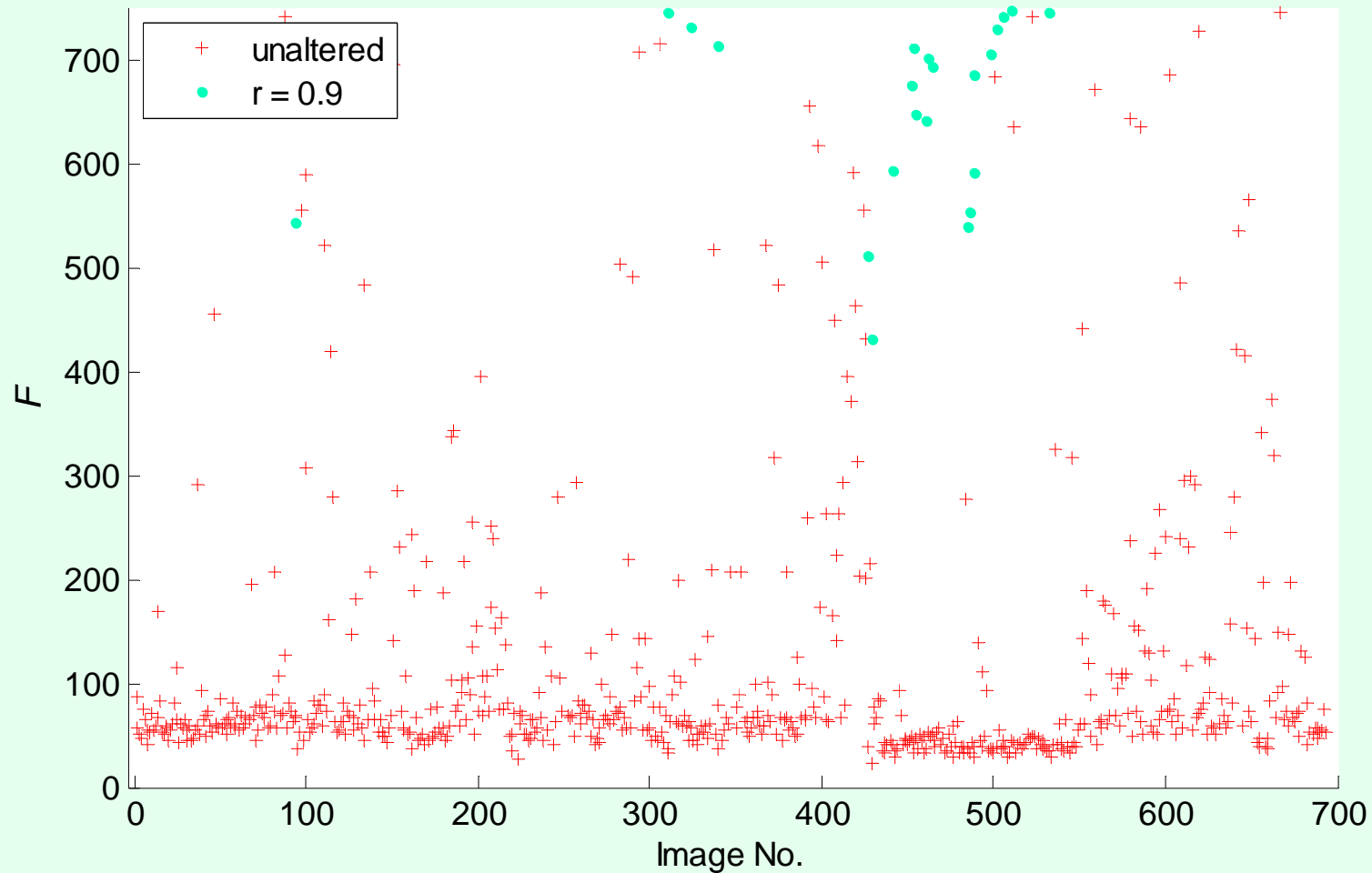


Fig. Feature value distribution of samples (*Locally magnified*).

Baseline Detection Results (3)

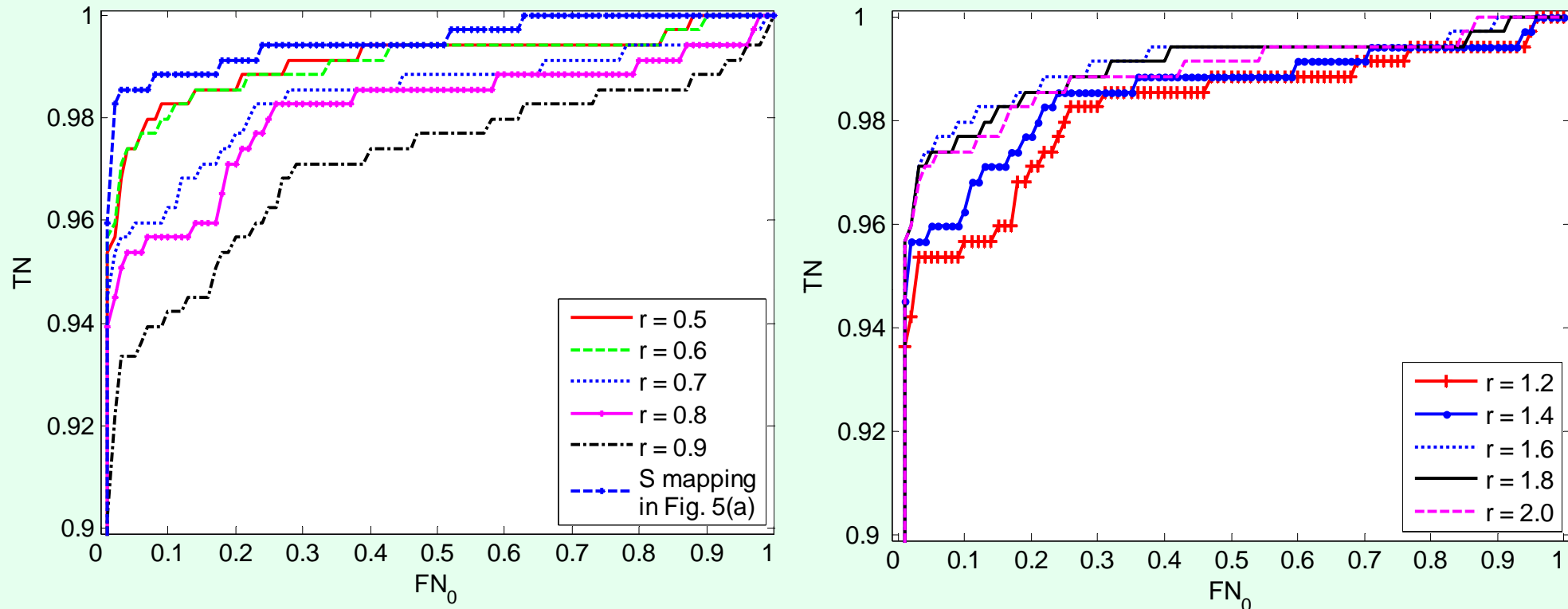


Fig. Detection rate of unaltered images (TN) under varying amounts for false negative rates (FN_0) .

Baseline Detection Results (4)

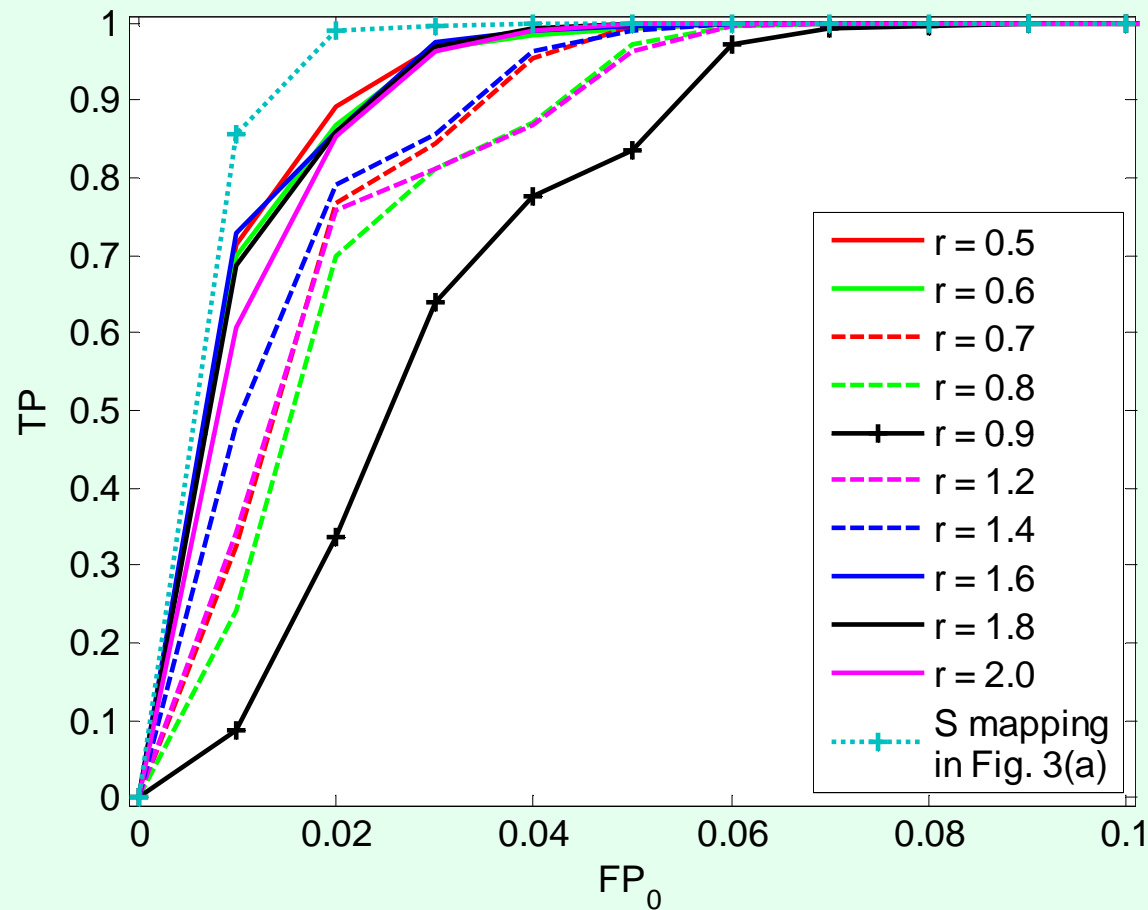


Fig. Detection rate of altered images (TP) under varying amounts for false positive rates (FP_0) .

Evaluate CE Forging Attack (1)

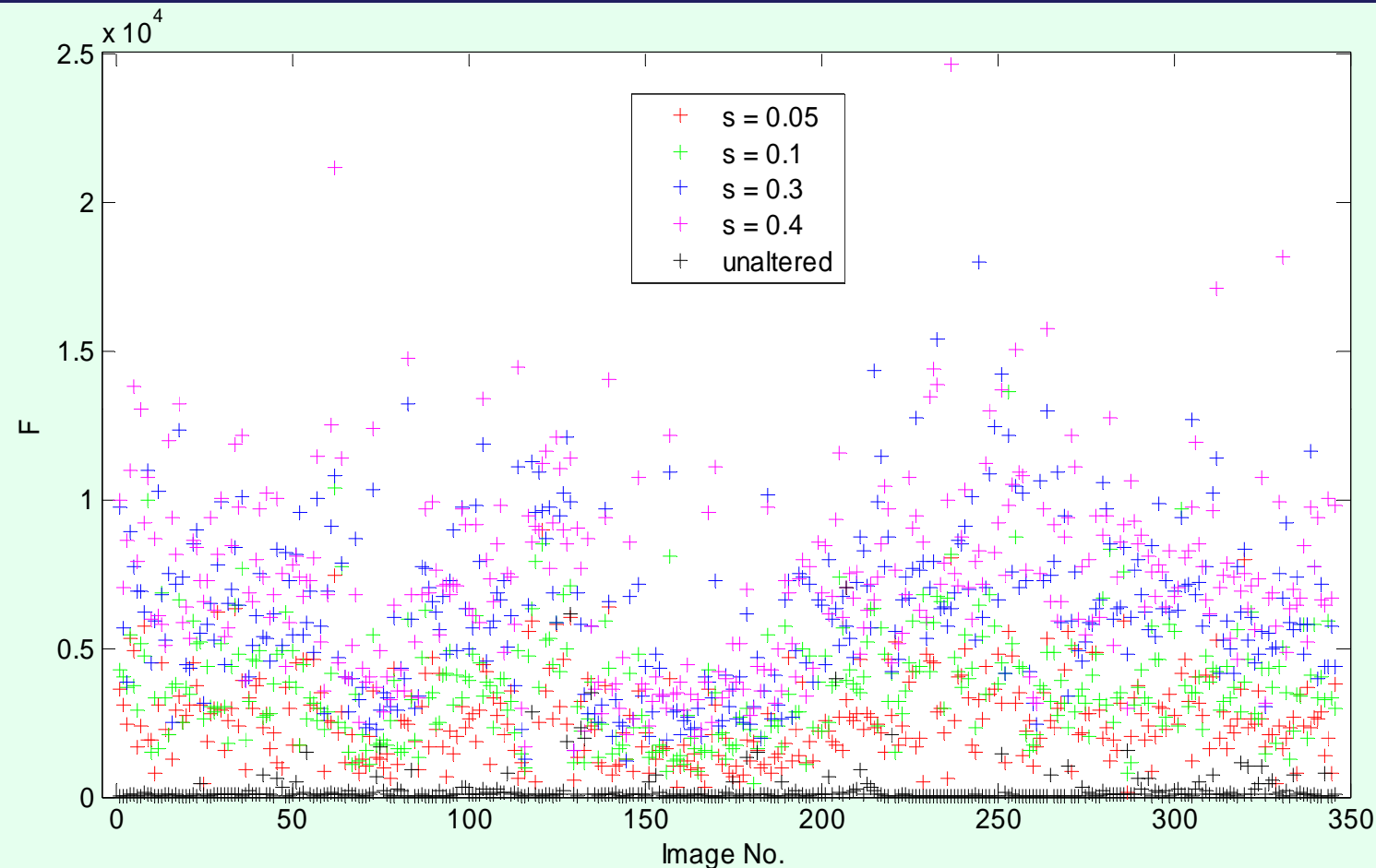


Fig. F – distribution for unaltered images and their *universal* CE forging attacked versions by random-LSB method.

s : attack strength

Evaluate CE Forging Attack (2)

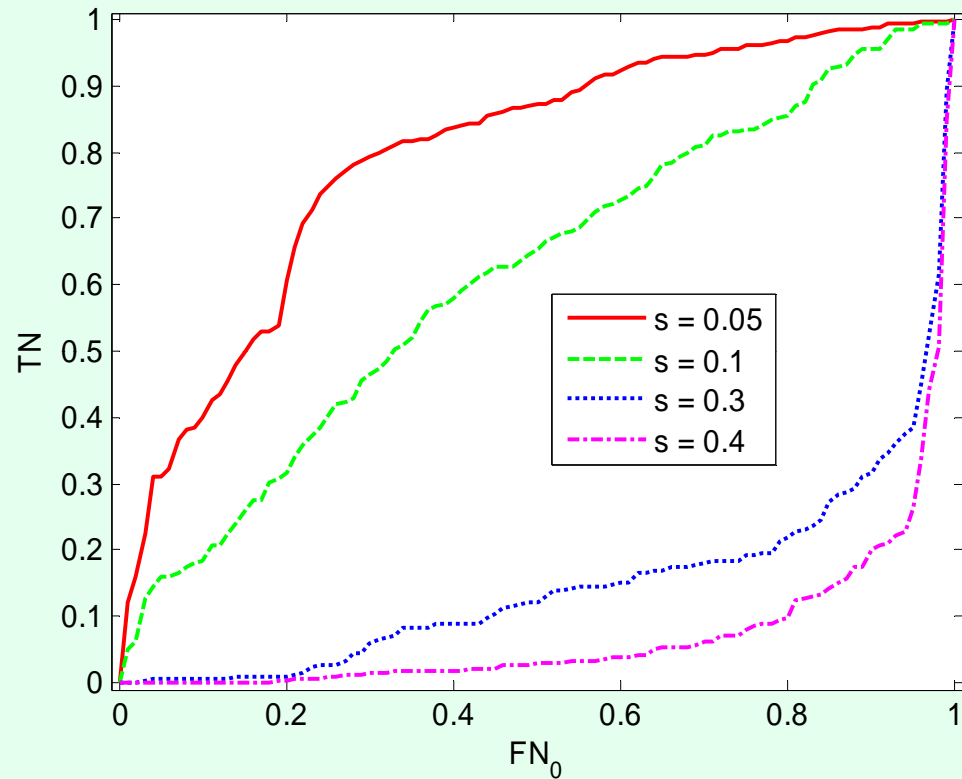


Fig. TN- FN_0 curve

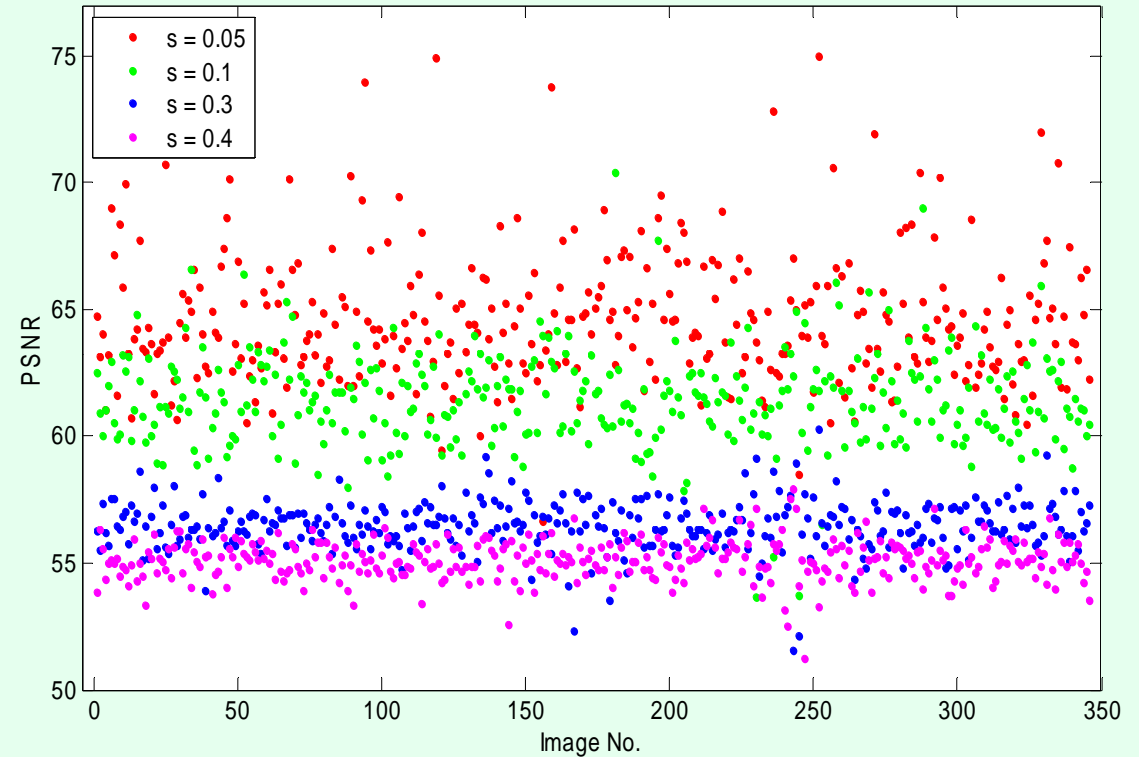


Fig. PSNR distribution

Evaluate CE Forging Attack (3)

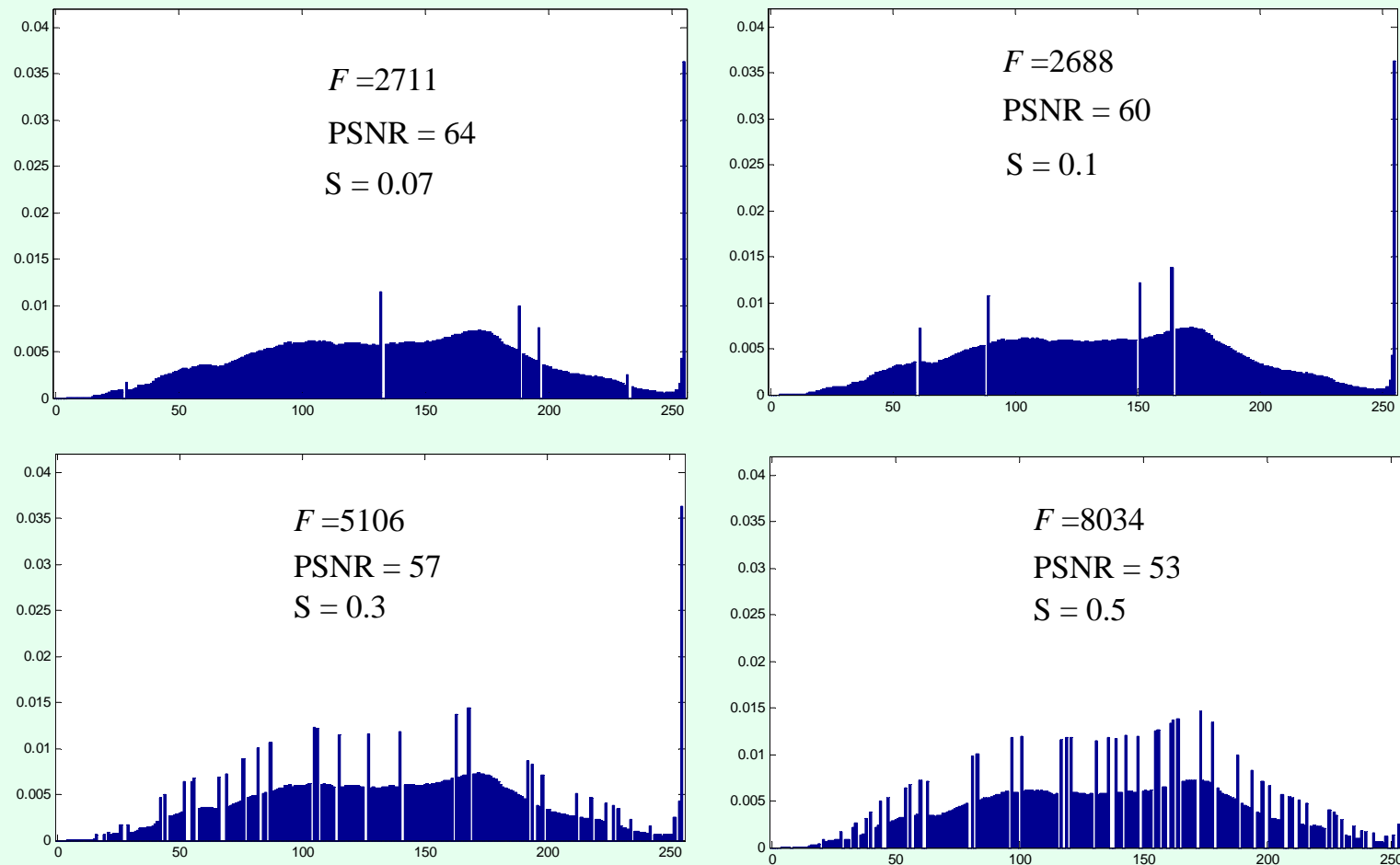


Fig. Histogram of attacked images ([random-LSB method](#)).

Evaluate CE Forging Attack (4)

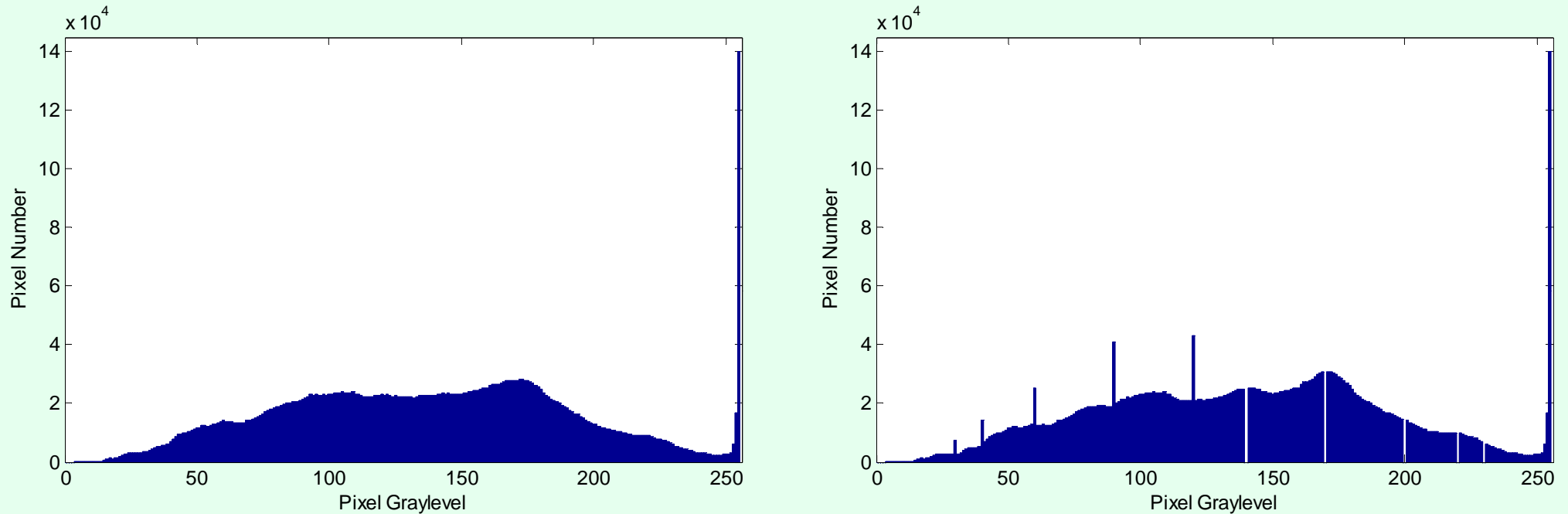


Fig. Histogram of *targeted CE forging attacked* image.

Unaltered image: $F = 63$

Attacked image : $F = 1281.7$, PSNR = 49.9

Evaluate CE Hiding Attack (1)

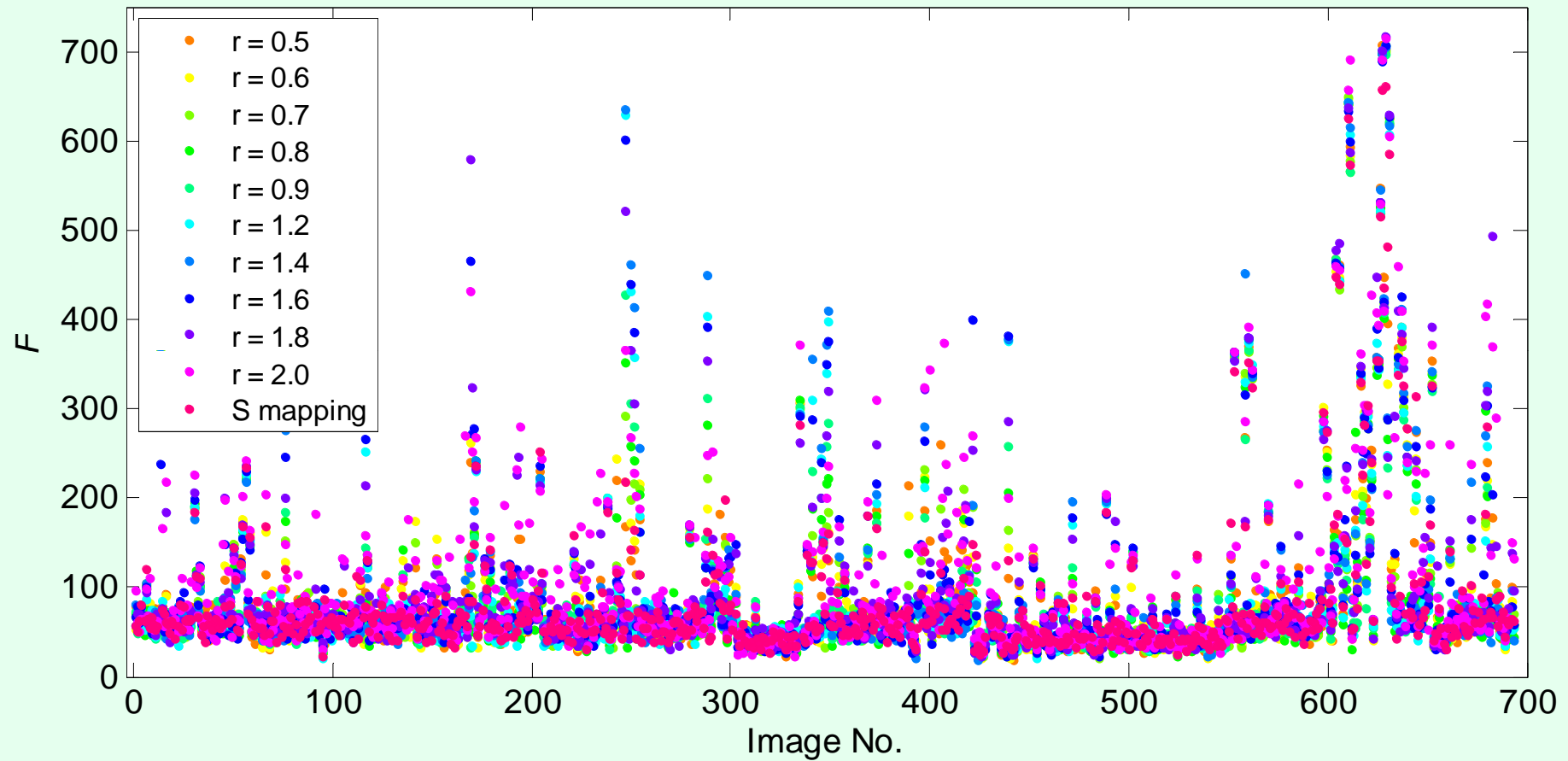


Fig. F distribution for *integrated CE hiding attack*.

Evaluate CE Hiding Attack (2)

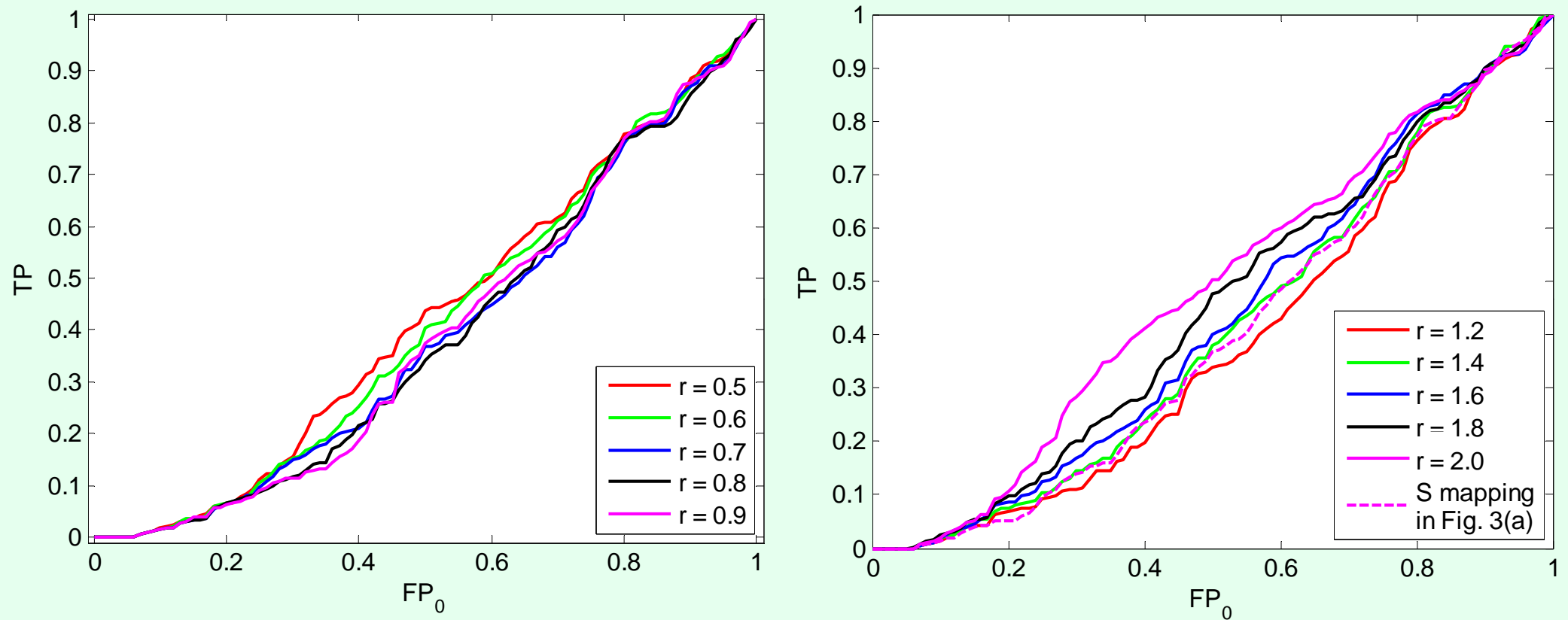


Fig. *Detection results for **integrated** CE hiding attack.*

Evaluate CE Hiding Attack (3)

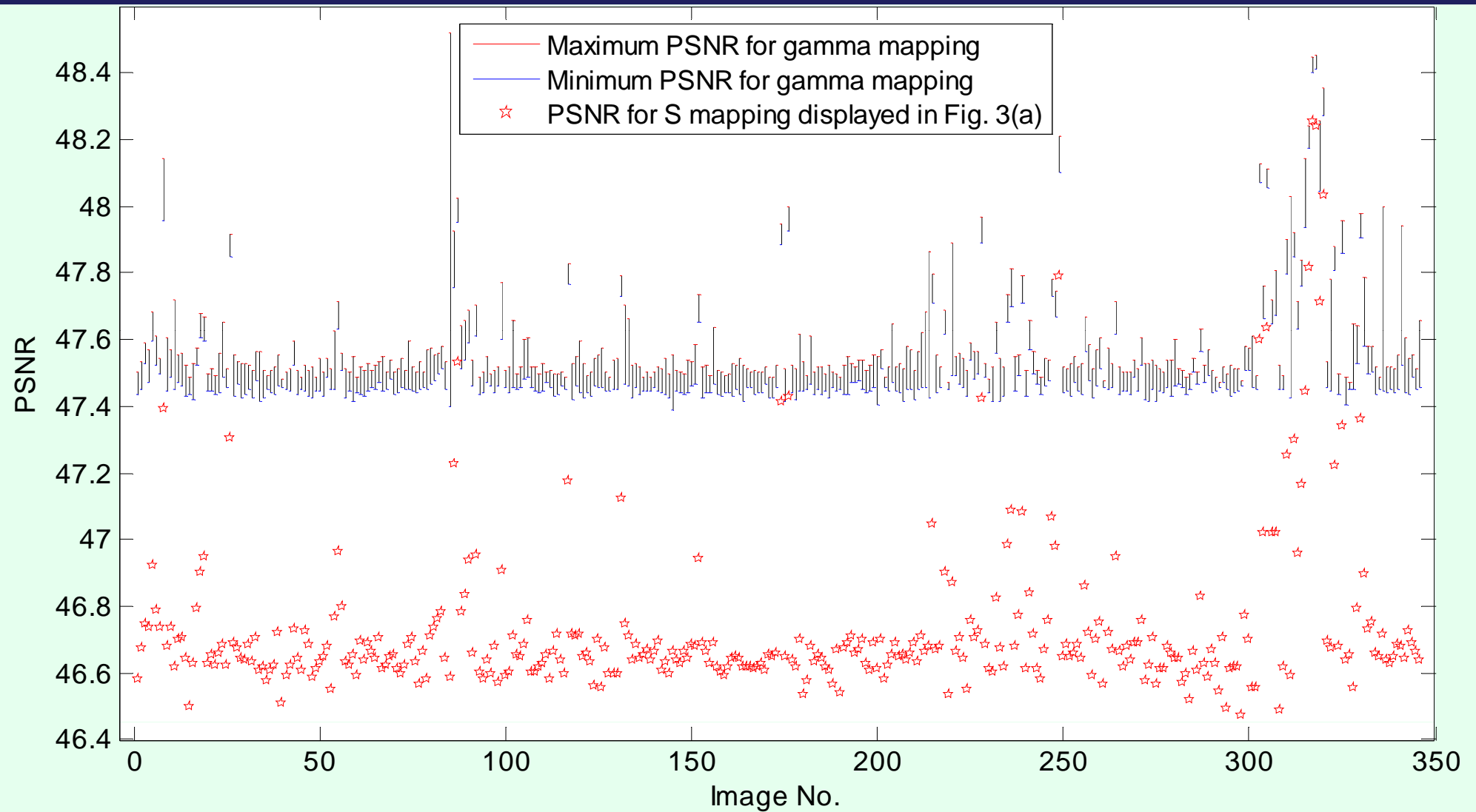


Fig. *PSNR for integrated CE hiding attack.*

Evaluate CE Hiding Attack (4)

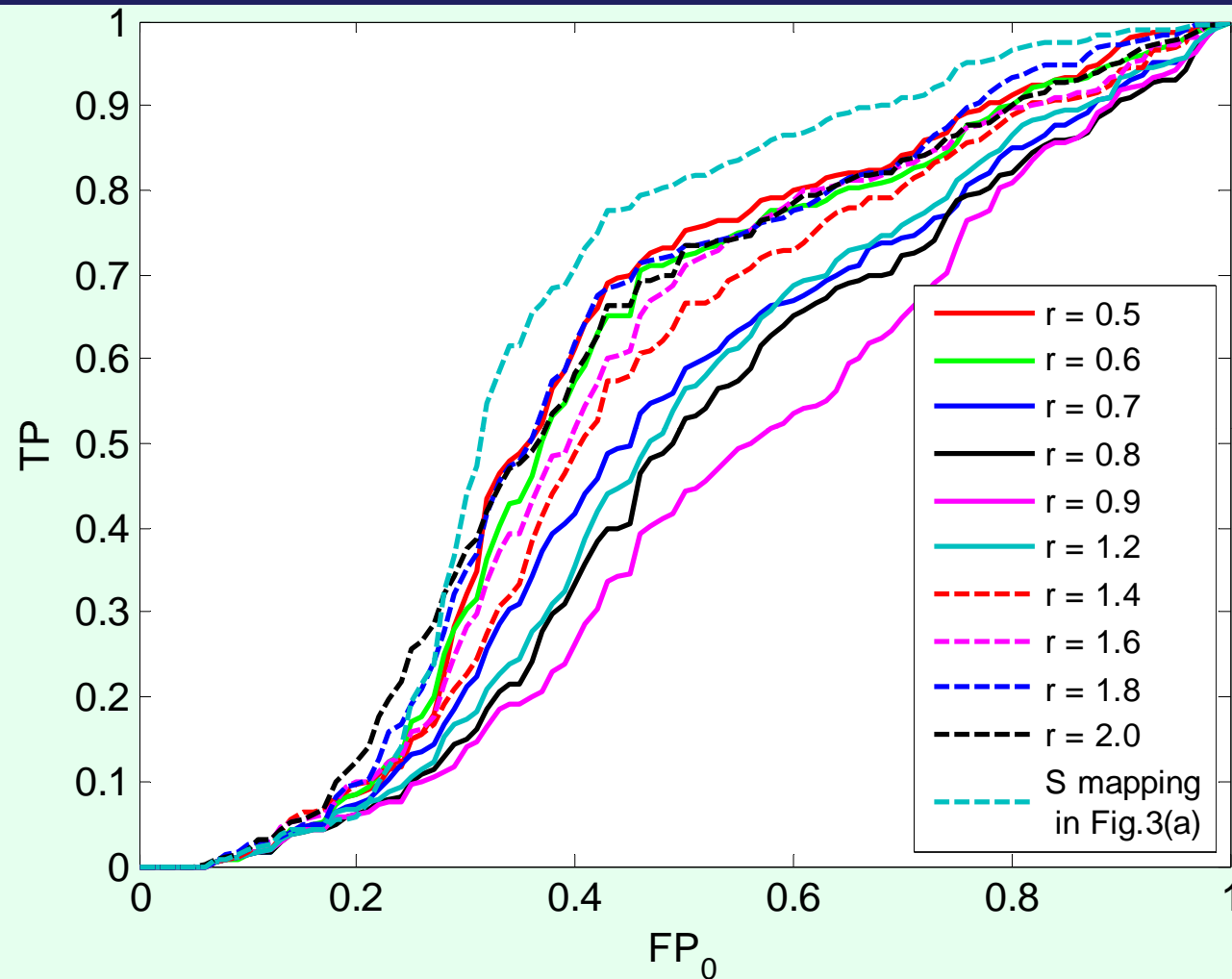


Fig. *Detection results for **postprocessing** CE hiding attack.*
(adding noise method)

Evaluate CE Hiding Attack (5)

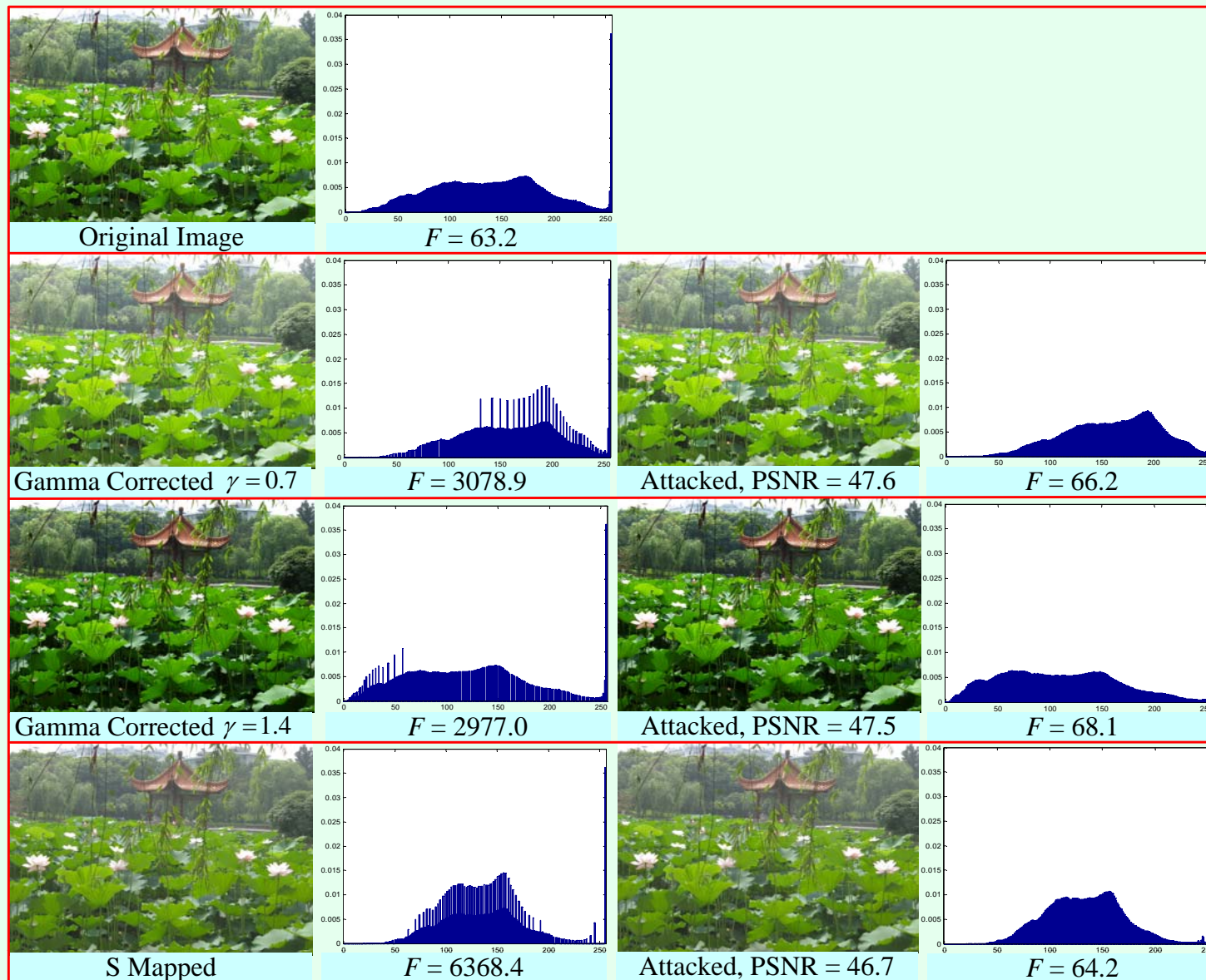


Fig. Test on *an example image* for **integrated** *CE hiding attack*.

Application to Forgery Making (1)

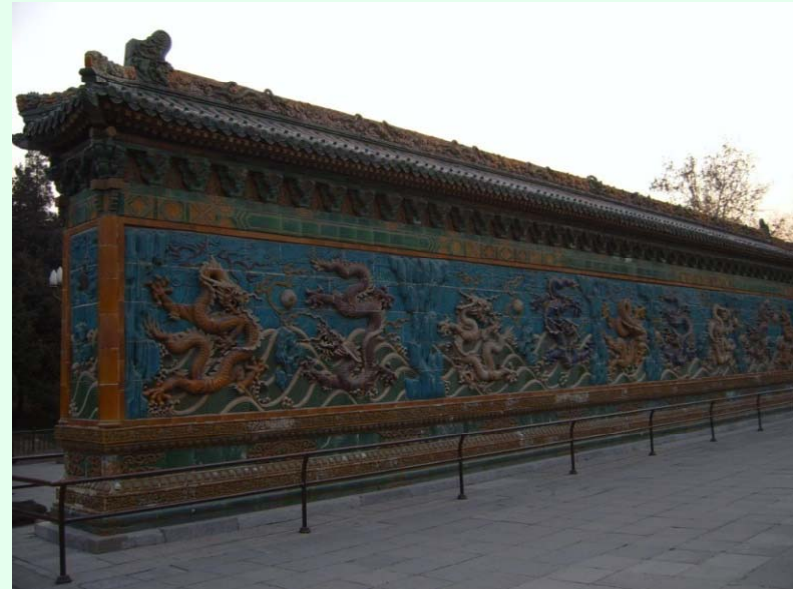


Fig. An image forgery example. without CE → Need CE ?

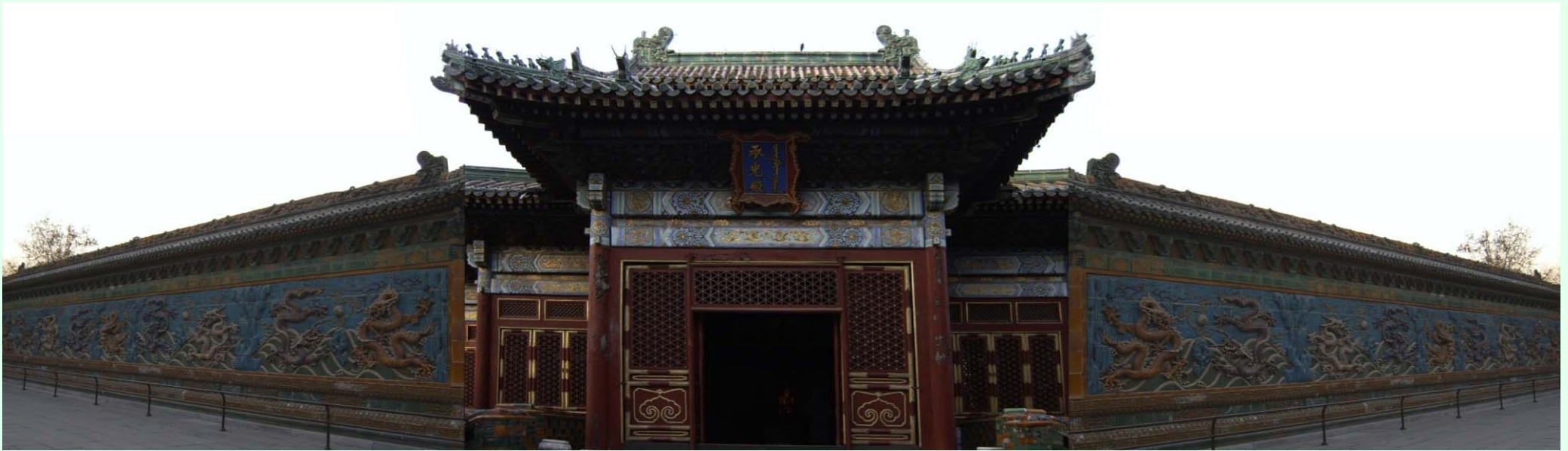
Application to Forgery Making (2)



0	0	0	0	0	0	0	3	2	2	4	4	3	2	3	3	3	3	2	0	0	0	0	0	0	0
0	0	0	0	0	0	0	5	5	6	13	6	7	7	5	7	6	7	3	0	0	0	0	0	0	0
0	0	0	0	0	0	1	4	5	8	4	4	4	13	7	13	4	4	4	0	0	0	0	0	0	0
2	2	1	3	4	4	7	5	4	5	4	7	6	7	6	7	8	9	7	4	4	5	3	1	1	3
5	4	6	6	6	5	4	5	5	5	5	4	6	7	6	5	4	7	4	7	5	4	4	4	5	5
7	7	6	5	5	5	8	7	6	5	7	6	9	11	5	5	5	6	5	8	6	5	5	4	7	6
5	5	7	3	9	4	6	7	7	10	3	7	17	21	5	4	7	6	5	7	3	5	6	4	8	6

Fig. F -map of original forgery. (without CE) block size : 200x200

Application to Forgery Making (3)



0	0	0	0	0	0	0	4	8	7	11	18	11	10	15	10	8	11	2	0	0	0	0	0	0	0
0	0	0	0	0	0	0	8	53	32	33	46	34	33	33	25	40	50	3	0	0	0	0	0	0	0
0	0	0	0	0	0	1	4	42	88	98	101	65	75	117	92	70	21	4	0	0	0	0	0	0	0
2	2	1	3	4	4	7	11	39	35	26	30	35	35	29	32	48	29	7	4	4	5	3	1	1	3
5	4	6	6	6	5	4	8	32	27	47	47	52	53	45	31	40	27	4	7	5	4	4	4	5	5
7	7	6	5	5	5	8	10	41	32	52	61	260	231	56	81	89	113	5	8	6	5	5	4	7	6
5	5	7	3	9	4	6	7	40	44	87	50	222	281	50	51	67	81	5	7	3	5	6	4	8	6

Fig. F -map of gamma-corrected forgery. ($r = 1.5$)

Application to Forgery Making (3)



0	0	0	0	0	0	0	6	8	7	11	18	11	10	15	10	8	11	7	0	0	0	0	0	0	0
0	0	0	0	0	0	0	13	53	32	33	46	34	33	33	25	40	50	8	0	0	0	0	0	0	0
0	0	0	0	0	0	2	23	42	88	98	101	65	75	117	92	70	21	14	1	0	0	0	0	0	0
10	4	5	23	38	45	65	53	39	35	26	30	35	35	29	32	48	29	78	56	38	31	15	3	2	9
46	77	100	126	135	184	203	132	32	27	47	47	52	53	45	31	40	27	36	41	57	86	83	68	64	43
195	174	168	122	162	158	123	97	41	32	52	61	260	231	56	81	89	113	25	49	27	24	31	29	26	38
52	71	108	119	116	99	96	89	40	44	87	50	222	281	50	51	67	81	51	63	54	47	44	35	42	63

Fig. Attacked forgery. (universal forging attack, random LSB, $s: 0.1$) PSNR: 71.0dB

Application to Forgery Making (4)



0	0	0	0	0	0	0	4	14	14	10	9	13	13	11	11	12	12	2	0	0	0	0	0	0	0
0	0	0	0	0	0	0	4	5	3	4	5	6	5	6	5	4	6	3	0	0	0	0	0	0	0
0	0	0	0	0	0	1	5	8	10	7	8	4	7	10	3	12	6	4	0	0	0	0	0	0	0
2	2	1	3	4	4	7	8	4	4	6	6	5	5	8	4	5	5	7	4	4	5	3	1	1	3
5	4	6	6	6	5	4	6	6	4	7	9	7	5	6	5	7	4	4	7	5	4	4	4	5	5
7	7	6	5	5	5	8	6	7	6	8	6	5	11	7	4	5	3	5	8	6	5	5	4	7	6
5	5	7	3	9	4	6	7	8	4	4	8	13	5	7	12	4	6	5	7	3	5	6	4	8	6

Fig. Attacked forgery. (integrated hiding attack, dithering-based CE) PSNR: 56.6dB

Conclusion & Remarks

➤ Main Contribution

- Forensics algorithms design for detecting Median filtering, Gamma correction, Blur/Sharpening, composition.
- Anti-forensics of contrast enhancement

➤ Further Work

- Algorithm Design:
 - ✓ improve **reliability & security of algorithms**
 - ✓ **counter-counter-.... forensics**
(cat-mouse game)
- Theoretical Analysis: **forensics behavior model & analysis**
(game theory).