

The Impact of Exposure Settings in Digital Image Forensics

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Outline

Motivation: StegoAppDB – A Benchmark Image Database for Steganography Apps ¹

Steganalysis under Different Exposure Settings

Camera Device Identification under Different ISO Settings

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- ▶ Academic steganography and steganalysis techniques are very successful in the academic environment using sophisticated embedding and detection methods, and data typically collected from digital still cameras.
- ▶ However, large gaps exist between academic results and applications used by forensic analysts, especially for detecting stego images created from mobile apps.

Steganography and Steganalysis in the wild

Seven selected stego apps from Google Play Store

App Name	# Installs	Output Format	Open Source
PixelKnot	100,000 - 500,000	JPEG	yes
Steganography Master	10,000 - 50,000	PNG	no
Steganography_M	10,000 - 50,000	PNG	no
Da Vinci Secret Image	5,000 - 10,000	PNG	no
PocketStego	1,000 - 5,000	PNG	no
MobiStego	1,000 - 5,000	PNG	yes
Passlok Privacy	1,000 - 5,000	JPEG	yes

Building StegoAppDatabase

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Questions

- ▶ Which exposure settings/scenes shall we use? Auto Exposure?
- ▶ How can we make the image data more representative?

Experiments on BOSSbase

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- ▶ BOSSbase [Bas and Filler, 2011]: The most popular image database for steganography and steganalysis.
- ▶ 10,000 cover images, EXIF data, seven digital still cameras.
- ▶ Images were collected in auto-exposure settings or semi-auto settings, with the intent to produce high visual-quality images.

Experiments on BOSSbase

- ▶ Select images taken by the PENTAX 20D in BOSSbase: 603 images with ISO 200 and 358 images with ISO 100.
- ▶ Cut all those 961 images into 5 pieces.
- ▶ Implement embedding algorithm “Mipod” [Sedighi, 2016] with 0.1 payload size.
- ▶ Use the “Spatial Rich Model” [Fridrich and Kodovsky, 2012] for feature extraction, and the FLD ensemble classifier [Kodovsky, 2012] for classification.

Experiments on BOSSbase

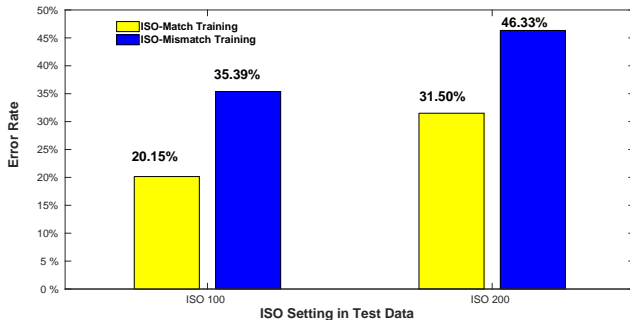


Figure 1: Steg Detect Error rates on different ISO data from BOSSbase.

Discussion and further experiments

- ▶ In the auto-exposure mode, a digital camera is programmed to choose an ISO value and exposure time, such that the digital photograph has a high signal-to-noise ratio (SNR) and low “image noise.”

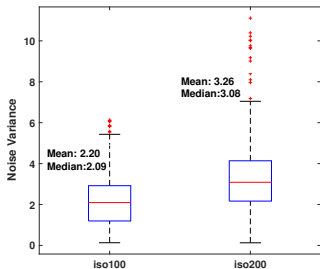
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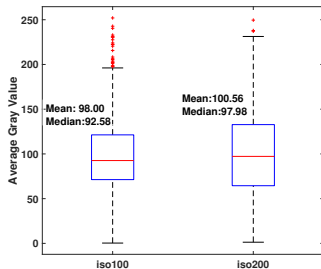
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- ▶ We apply a wavelet denoiser for each image we used in BOSSbase to get a clean image, and then compute the noise level as the mean of the residual image.

Further experiments on BOSSbase.



(a) Boxplots of Noise Levels



(b) Boxplots of Grey Values

Further experiments on BOSSbase.

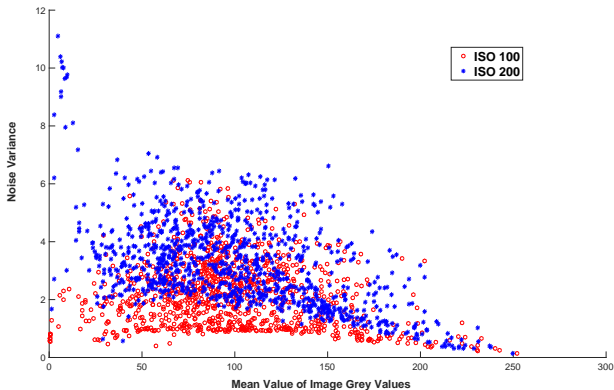


Figure 2: Noise level v.s. average grey value

Experiments on Images from one iPhone 7

Table 1: Misclassification rates on data with different ISO settings

Test data ISO	Training data ISO	Exposure time =1/200s	Exposure time =1/50 s
		Avg. error rate (a)	Avg. error rate (b)
100	100	7.72%	14.92%
	200	28.89%	24.93%
	1000	38.04%	40.70%
200	100	17.89%	26.18%
	200	9.90%	18.67%
	1000	30.39%	46.79%
1000	100	48.50%	42.16%
	200	42.18%	34.04%
	1000	18.08%	12.57%

Experiments on Images from one iPhone 7

Table 2: Misclassification rates on data with different exposure time

Test exposure time	Training exposure time	ISO = 100	ISO = 200
		Avg. error rate (c)	Avg. error rate (d)
1/200 s	1/200 s	7.72%	9.90%
	1/50 s	16.89%	20.52%
	1/10 s	37.88%	40.71%
1/50 s	1/200 s	24.47%	25.13%
	1/50 s	14.92%	18.67%
	1/10 s	27.89%	28.18%
1/10 s	1/200 s	34.31%	24.50 %
	1/50 s	24.09%	19.90 %
	1/10 s	15.95%	12.20%

Summaries of the Impact of Exposure Settings on Steganalysis

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- ▶ Applying a well-trained classifier to target data with different exposure settings brings significantly higher errors.
- ▶ Adapting the exposure parameters for the target images can significantly improve the performance of a forensic analyzer.

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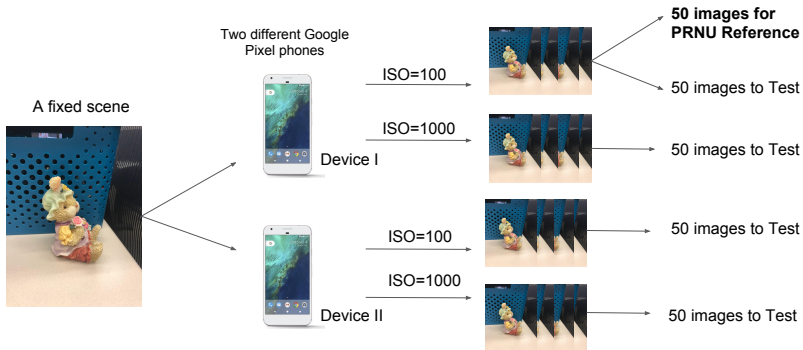
- ▶ Behind the exposure settings are the noise levels, which play an important role in steganalysis.
- ▶ Applying a well-trained classifier to target data with different exposure settings brings significantly higher errors.
- ▶ Adapting the exposure parameters for the target images can significantly improve the performance of a forensic analyzer.
- ▶ In order to build a benchmark image database for detecting stego apps, the diversity of exposure settings for images in such a database must be taken into account.

PRNU-based Camera Device Identification under Different ISO Settings

Photo-Response Non-Uniformity

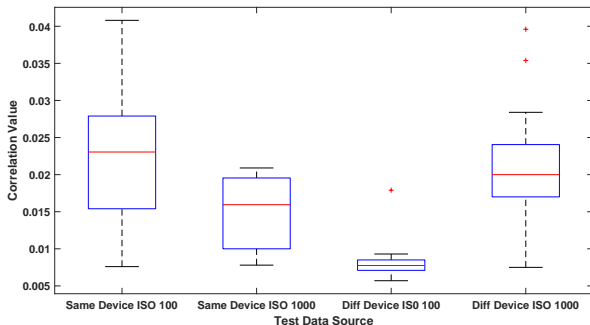
- ▶ Multiplicative noise, known as the camera fingerprint.
- ▶ Independent of temperature and time
- ▶ Inhomogeneous nature of silicon sensor for the digital camera.
- ▶ Similar for all images from the same camera, but different between distinct devices.
- ▶ The correlation between the reference PRNU and the target ones are widely used for device identification.

Experimental Design for the Camera Identification under Different ISO Settings (Exposure Time = 1/50 s)



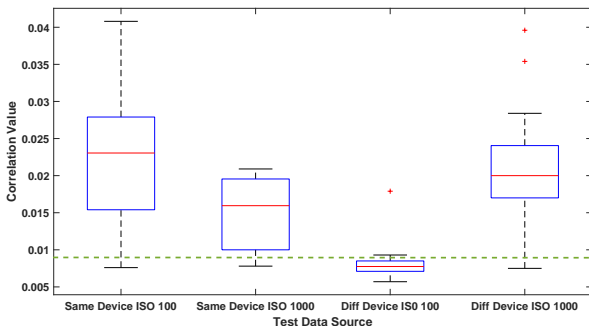
Experimental Result for the Camera Identification under Different ISO Settings

Setting ISO=100, and exposure time=1/50 s, the PRNU reference is generated by 50 images with a fixed scene indoor.



Experimental Result for the Camera Identification under Different ISO Settings

Setting ISO=100, and exposure time=1/50 s, the PRNU reference is generated by 50 images with a fixed scene indoor.



Demo of StegoAppDB

Questions?

The End