Multimedia Security and Privacy

TP5: Watermark performance evaluation

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Submission

Please archive your report and codes in "Name_Surname.zip" (replace "Name" and "Surname" with your real name), and upload to "Assignments/TP5: Watermark performance evaluation" on https://chamilo.unige.ch before Wednesday, May 23 2019, 23:59 PM. Note, the assessment is mainly based on your report, which should include your answers to all questions and the experimental results.

In this TP work you will assess the performance of the watermark detection model that was build in the previous TP.

1 Non-blind watermark detection

1.1 Exercise

Read the image cameraman.tif in. It will serve as host image x. For given hypothesis:

$$\begin{cases} H_0: & \mathbf{v} = \mathbf{x} + \mathbf{z} \\ H_1: & \mathbf{v} = \mathbf{x} + \mathbf{w} + \mathbf{z} \end{cases}$$

where \mathbf{x} is the host image, \mathbf{v} is the marked image, \mathbf{w} is the watermark and \mathbf{z} is additive white Gaussian noise, i.e. $\mathbf{Z} \sim \mathcal{N}(\mathbf{0}, \sigma_{\text{noise}}^2 \mathbf{I})$

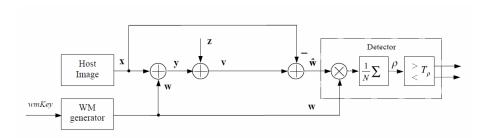


Figure 1 – Non-blind watermark detection

1.2 Exercise 2

1.2 Exercise

1. The detection threshold is denoted with $T_{\rho \text{ non-blind}}$, evaluate the following numerically for the non-blind detection shown in Figure 1:

- p_f , the probability of false alarm
- p_m , the probability of miss
- p_d , the probability of correct detection, defined as $1-p_m$

where $\mu_{\rho|H_0}$, $\sigma_{\rho|H_0}$, $\mu_{\rho|H_1}$ and $\sigma_{\rho|H_1}$ are the means and variances of the linear correlation ρ under hypothesis H_0 and H_1 for J realizations each. They can be ascertained as follows:

$$\mu_{\rho|H_0} = \frac{1}{J} \sum_{k=1}^{J} \rho_k^{H_0} \tag{1}$$

$$\mu_{\rho|H_1} = \frac{1}{J} \sum_{k=1}^{J} \rho_k^{H_1} \tag{2}$$

$$\sigma_{\rho|H_0}^2 = \frac{1}{J} \sum_{k=1}^{J} \left(\rho_k^{H_0} - \mu_{\rho|H_0} \right)^2 \tag{3}$$

$$\sigma_{\rho|H_1}^2 = \frac{1}{J} \sum_{k=1}^{J} \left(\rho_k^{H_1} - \mu_{\rho|H_1} \right)^2 \tag{4}$$

Obviously, ρ^{H_0} and ρ^{H_1} need to be experimentally obtained.

- For hypothesis H_0 , ρ^{H_0} is determined J = 100 times, $k = \{1...J\}$ for noise realizations **z** with a fixed $\sigma_{\text{noise}} = 50$.
- For hypothesis H_1 and ρ^{H_1} the watermark **w** is generated J times with a fixed strength $\gamma = \pm 1$ and a fixed density $\theta_N = 0.1$. The noise realization **z** is again fixed with $\sigma_{\text{noise}} = 50$.
- 2. Calculate and display the Receiver Operating Characteristic (ROC) curve for the binary threshold test following the above mentioned experiment set up. The detection threshold is denoted with $T_{\rho \text{ non-blind}}$.
- 3. Fill out Table 1 with all results. Note that obviously only the noise and not the watermark has influence on hypothesis H_0 , so the relevant cells have been grayed out.
- 4. What can you conclude about *non-blind* watermark detection given the strength of the watermark and the noise variance?

		$\sigma_{ m noise}$	= 50		$\sigma_{ m noise} = 100$			
	$\theta_N = 0.1$		$\theta_N = 0.3$		$\theta_N = 0.1$		$\theta_N = 0.3$	
	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$
$\mu_{ ho H_0}$								
$\sigma_{ ho H_0}^2$								
$\mu_{ ho H_1}$								
$\sigma^2_{ ho H_1}$								

Table 1 – Data for non-blind watermark detection

2 Blind watermark detection using the Maximum Likelihood estimate

This exercise will follow the same structure and tests as the previous one for non-blind watermark detection, except that this time you will blindly detect the watermark using the Maximum Likelihood estimate of \mathbf{x} used in previous TP.

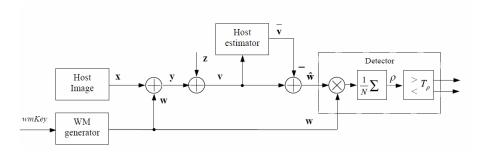


Figure 2 – Blind watermark detection

2.1 Exercise

For given hypothesis:

$$\begin{cases} H_0: & \mathbf{v} = \mathbf{x} + \mathbf{z} \\ H_1: & \mathbf{v} = \mathbf{x} + \mathbf{w} + \mathbf{z} \end{cases}$$

where \mathbf{x} is the host image, \mathbf{v} is the marked image, \mathbf{w} is the watermark and \mathbf{z} is additive white Gaussian noise, i.e. $\mathbf{Z} \sim \mathcal{N}(\mathbf{0}, \sigma_{\text{noise}}^2 \mathbf{I})$,

- 1. Evaluate numerically p_f , p_m and p_d using the same conditions as for the previous task.
- 2. Calculate and display the Receiver Operating Characteristic (ROC) curve for the binary threshold test.
- 3. Fill out Table 2 with all results. With all results. Again note that obviously only the noise and not the watermark has influence on hypothesis H_0 , so the relevant cells have been grayed out.
- 4. What can you conclude about *blind* watermark detection given the strength of the watermark and the noise variance?

		$\sigma_{ m noise}$	= 50		$\sigma_{\rm noise} = 100$			
	$\theta_N = 0.1$		$\theta_N = 0.3$		$\theta_N = 0.1$		$\theta_N = 0.3$	
	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$	$\gamma = \pm 1$	$\gamma = \pm 5$
$\mu_{\rho H_0}$								
$\frac{\mu_{\rho H_0}}{\sigma_{\rho H_0}^2}$								
$\mu_{\rho H_1}$								
$\sigma^2_{ ho H_1}$								

Table 2 – Data for blind watermark detection

2.2 Exercise

Compare the ROC curves from the *non-blind* and *blind* watermark detection schemes. What can you conclude about their comparative performance?