SEMANTIC MEDIA ANALYSIS FOR PARALLEL HIDING OF DATA IN VIDEO AND AUDIO TRACK

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Keywords: Audio steganography, video steganography, DWT, data hiding, cash machine.

Abstract: This paper is dealing with the role of steganography in system of intelligent cash machines. Hiding methods dedicated especially for this application are presented. In the system typical content protection was extended by hiding multistream secret additional information. Video files with concealed information are stored in the system by monitoring subsystem (Controll AV recording and storage) and third part has no possibility to delete, add or change video archives. Two different steganography methods were mixed to prepare more advanced approach.

1 INTRODUCTION

Data hiding has a various applications in multimedia area. In this paper are proposed data hiding methods for a specific implementation. They are used as a part of intelligent cash machine system. The base concept is assuming that the cash machine is interactive, user-friendly and safe. It is obvious that the last requirement is very important. There exist some methods ensuring high security for cash machines and message exchanging with central computer system. These are usually cryptography and public key infrastructure based methods. Our goal is to add extra security tools based on data hiding which increase safety and facilitate investigation in case of protection violation.

The intelligent cash machine system includes a surveillance module registering an interactive session between user and cash machine and other events generated by accidental person or intruders. The registered audiovisual material is taken by cameras and microphones at the cash machine. In next step an additional information (metadata) is hidden into material in such a way that the data can not be removed without destroying host media. As metadata are used time and date information, cash machine id, and personal data of a client. In case of destruction or replacement the original material with fabricate one, it is impossible to retrieve hidden metadata, what detects a violation case.

The following sections contain details of the algorithm. Section 2 describes general data hiding concept. Section 3 provides the methods to ensure data robustness. In section 4.5 algorithms of audio and video steganography are described in details. Last two sections include experiments, their results, and final conclusions.

2 GENERAL DESCRIPTION OF THE SYSTEM

In Fig.1 general idea of metadata hiding in multimedia stream is presented. Original data \( D \) is duplicate three times \( (D_1, D_2, D_3) \) and send to blocks which are transfered to inputs of three different subsystems to conceal data:

1. subsystem using only audio information as host medium (Block 1),
2. subsystem using only video information as host medium (Block 2),
3. subsystem hiding data from input in audio and video stream (Blocks 1, 2 and 3).

To ensure indexed data integrity, safety and robustness for intentional attacks of metadata hidden in multimedia material. Block 1 and Block 3 include special subblocks which prepare data for hiding and have special methods to increase robustness, which are described in section 3.

In Block 2 the raw data \( D_2 \) and secured data from Block 3 output \( S(DV) \) is hidden in video content and...
MD5 function result \( s \) is generated using bits of video after hiding process. Next data \( D_1 \) and \( s \) are secured and embedded in audio stream.

The fact that algorithm of data hiding in video is very fragile causes that intruder has no possibility to attack the multimedia stream. Modification will be detected by:

1. finding the differences between data \( D_1 \) with \( D_2 \) or \( D_3 \),
2. MD5 results with \( s \) comparison.

3 ADDITIONAL METHODS TO PROVIDE DATA ROBUSTNESS

To ensure robustness we use two techincs for the data hidden in audio \( (D_1) \) and for part of the data hidden in video \( (D_V) \):

1. pseudorandom permutation of bits with seed unknow for third part persons,
2. pseudorandom partition of bits in block Random Partition Block with seed unknow for third part persons,
3. Viterbi’s error correction codes (Fig.2) ((Inge-mar J.Cox, 2001)).

4 AUDIO STEGANOGRAPHY ALGORITHM

The main concept is based on DWT (Discrete Wavelet Transform) and general idea is illustrated in Fig.3. An audio record is divided into blocks of samples. For each audio block H-level DWT is performed. In next step the embedded position is selected.

In order to take the advantage of frequency masking effect of human auditory system and develop the robustness against variety signal attacks such as noising, low pass filtering, subsampling, requantization and mainly against lossy compression. In the scheme one bit of secret message is embedded into selected DWT coefficient. The embedding process is controlled by a quantization step parameter. Finally to form stegoaudio Inverse Discrete Wavelet Transform is performed. This simple method has some disadvantages. Using the same embedded threshold for all audio blocks causes an audibility distortion or low robustness against signal attacks. In addition it strongly depends on characteristic of sound i.e. the kind of music, singer, speaker etc. We used the extended method. The quantization step is adaptively selected to characteristic of a given sound block and its surroundings. According to the requirements we have applied fuzzy clustering algorithm. A local audio features have to be calculated before the clustering analysis is performed. The membership value of all clusters is used to choose appropriate quantization step.

It was a short introduction to an idea of how to improve the basic steganography algorithm based on DWT. The scheme of hidden data extraction is pre-
The quantization step has an important role in transparency and robustness in the audio steganography algorithm. The larger the quantization step is, the more robust becomes hidden information against signal attacks, but more perceptible the hidden information is. However, the smaller size of quantization step will influence the robustness. If the same quantization step is adopted for the whole host audio, probably it will cause one or both of the problems in some parts of the host audio.

The quantization should be selected according to the local audio correlation, human auditory masking and possible signal attack.

Digital audio itself always has various auditory features. It is very important to choose these, which represent different parameters of the digital audio and allow to selected appropriate embedded threshold. A single bit is hidden in a short audio block. So in this case an audio features should be computed for audio samples from given audio block or from its neighboring if the block is very short. Too short block might cause distortion of the computed features value. In addition the values should be independent from length of the audio block for a given range. To meet above requirements, based on reviews and experience reported in papers (Peeters, 2004; E.Schubert, 2004; S.H.Srinivasan, 2003; Wang Xiang-yang, 2004; Changsheng Xu, 2002), we have selected audio features such as fundamental frequency, short time mean energy, harmonic concentration, spectral centroid, harmonic energy distribution, max energy, zero-crossing rate.

Values of the features for a single digital audio block form a feature vector \( X_i = (x_{i1}, x_{i2}, \ldots, x_{i9}) \) which determine a point in 9-dimensional space. Such a vector is computed for each audio block from all training audio records. All vectors construct a data set \( X \). Similarity between some vectors gives a way to find a common quantization step for close located points. To group them we have used the fuzzy-C-mean clustering (J. C. Bezdek, 1987) algorithm. As a result we get the centers of groups. For each center we adopt separate quantization step \( T_k \).

### 4.2 Fuzzy Clustering Algorithm

Let \( X \) be a data set, and \( x_i \) denotes one sample \( (i = 1, 2, \ldots, N) \). The goal of clustering is to partition \( X \) into \( K (2 \leq K \leq N) \) subsets or representatives clusters, and make most similar samples be in the same cluster if possible. The typical clustering algorithm gain is to strictly classify each sample to a certain cluster. However, as a matter of the fact, there is often no explicit characteristic with which the sample can be grouped. Many samples belong to several clusters. Under that situation, the fuzzy clustering algorithm could provide a better performance. The fuzzy clustering algorithm divides data set \( X \) into \( K \) clusters according to fuzzy membership \( u_{ik} \) \( (0 \leq u_{ik} \leq 1) \) that represents the degree by which the sample \( x_i \) belongs to \( k \)-th \( (k = 1, 2, \ldots, K) \) cluster. So the results of the clustering can be described as a \( N \times K \) matrix, that is composed by each \( u_{ik} \), and

\[
\sum_{k=1}^{K} u_{ik} = 1; 0 < \sum_{i=1}^{N} u_{ik} < N. \quad (1)
\]

Proposed by J.C. Bezdek (J. C. Bezdek, 1987), the fuzzy C-mean clustering algorithm seeks to find fuzzy division by minimizing the following objective function:

\[
J_q(U, V) = \frac{1}{K} \sum_{k=1}^{K} \left( \sum_{i=1}^{N} (u_{ik})^q \right)^{\frac{1}{q}} \sum_{j=1}^{N} (X_j - V_k)^2, \quad (2)
\]

where:
- \( N \) - the number of feature vectors,
- \( K \) - the number of clusters (partitions),
- \( q \) - weighting exponent (fuzzifier; \( q > 1 \)),
- \( u_{ik} \) - the i-th membership function on the k-th cluster,
- \( V_k \) - the center of k-th cluster,
- \( X_j \) - the i-th feature vector,
distance between the feature vector $X_i$ and the center of cluster $V_k$. Larger membership values indicate higher confidence in the assignment of the feature vector to the cluster. A process of fuzzy clustering can be described by following steps:

1. Choose primary cluster prototypes $V_i$ for the values of the memberships;
2. Compute the degree of membership for all feature vectors in each cluster:
   
   \[
   u_{ik} = \left( \frac{1}{\sum_{k=1}^{K} \frac{1}{d^2(X_i - V_k)}} \right)^p, \tag{3}
   \]

   where:
   
   \[ p = \frac{1}{q - 1}; \tag{4} \]

3. Compute new cluster centers $V_k$
   
   \[
   V_k = \frac{\sum_{i=1}^{N} \left[ u_{ik} \right]^q X_i}{\sum_{i=1}^{N} \left[ u_{ik} \right]^q}; \tag{5}
   \]

4. Iterate back and forth between (2) and (3) until the memberships for successive iteration differ by more than some prescribed value of termination criterion $\epsilon$.

As an example and illustration of clustering results of chosen three audio features sets is shown in Fig.5. On the picture example each feature vector is strictly partitioned to a certain cluster.

![Figure 5: Example of clustering results.](image)

**4.3 DWT**

For each audio segment $A(i)$, H-level DWT is performed [11], and we get the wavelet coefficients of $C(i)^H, D(i)^H, D(i)^{H-1}, \ldots, D(i)^1$ where $C(i)^H$ is the coarse signal and the detail signals are $D(i)^H, D(i)^{H-1}, \ldots, D(i)^1$. The detailed coefficients correspond to the high frequency components and the coarse coefficients correspond to the low ones. In order to balance the transparency and robustness, the coefficient $d^H(i)$ from the detail components $D(i)^H$ is selected for embedding a bit of information. There are several advantages of applying DWT to data hiding into digital audio:

- DWT has the time-frequency localization capability,
- variable decomposition levels are available,
- DWT itself needs a lower computation load compared with DCT and DFT.

**4.4 Bit Embedding Scheme**

In order to embed single bit in a single audio segment a feature vector $x_i$ is computed. For the feature vector $x_i$ a fuzzy membership $u_{ik}$ is found. Finally, if we have adopted quantization steps $T_k$ and the fuzzy membership $u_{ik}$, we calculate adaptive quantization step $T_i$:

\[
T_i = \frac{\sum_{k=1}^{K} u_{ik} T_k}{\sum_{k=1}^{K} u_{ik}}. \tag{6}
\]

The embedding scheme is based on quantization process and the idea is illustrated in Fig.6. The selected DWT coefficient $C_b = d^H(i)$ is quantized with a given quantization step $T$, so the coefficient value $C_b$ is rounded to multiple value of $T$. Depends on hidden bit value $b$ the coefficient $C_b$ is rounded in different way:

- if $b = 1$, the coefficient $C_b$ is rounded to the nearest odd multiple value of $T$,
- otherwise if $b = 0$, the coefficient $C_b$ is rounded to the nearest even multiple value of $T$.

![Figure 6: Single bit embedding scheme.](image)

After a bit of information is embedded, the modified DWT coefficient is placed back into structure of the detail components $D(i)^H$ and H-level Inverse DWT is performed. Retrieved audio segments form audiostream like host audio track.
5 VIDEO STEGANOGRAPHY ALGORITHM

Many watermarking algorithms for video were proposed (Adnan M. Alattar and Celik, 2003; Yulin Wang, 2002; Hartung, 99; Changyong Xu, 2006). The majority of watermarking algorithms operates directly in video compressed bit streams, changing DCT coefficients (Adnan M. Alattar and Celik, 2003; Changyong Xu, 2006).

Method proposed by Wang (Yulin Wang, 2002) used only I-frames for information hiding. other methods (Changyong Xu, 2006; Adnan M. Alattar and Celik, 2003) take also motion vectors in P-frames and B-frames into consideration. Despite of these extensions for information hiding in video stream, high-capacity steganography should use not only one type of elementary stream.

Block diagram (Fig. 7) presents general idea of hiding in video content.

![Block diagram of the method of hiding data in video stream.](image)

Proposed video steganography method is based on changing DCT coefficients. In MPEG-2 standard the quantized DCT coefficients are encoded using run/level encoding and subsequent variable length coding (VLC) (ISO and IEC, 2000). Thus, simple DCT modulation could dramatically change the amount of bits in video stream. This fact could cause increase of probability that stegocommunication will be noticed by third person. To prevent against this situation some methods were previously developed: we can find the quantized coefficient with the same length of bit representation or resign from hiding bit in this coefficient (Hartung, 99).

In our system we use different algorithm. In MPEG-2 Program Stream the bit rate is calculated using time stamps, so we ensure constant bit rate on higher level then single VLC-codes. The method does not change the bitrate for each block of bits between timestamps Fig. (8). This modification gives higher capacity. This task was gained by Algorithm 1.

Algorithm 1.

1. Classification of VLC (run, level,length) coded coefficients in three sets:

(a) pairs of coefficients sequence coded using escape method with constant length code (set A),

(b) pairs of coefficients sequence coded with the same length of VLC code and run but with difference between level values equals to 1 (set B),

(c) pairs of coefficients sequence coded with the same run value but difference between level values lower than 2, different lengths of VLC code, but levels \(\geq T\) (for example (1,5,8) and (1,6,13)) (set C),

(d) codes useless for hiding (set D);

2. Hiding the bits by replacement the VLC code by another code from pair:

(a) if level after change is odd, 0-bit was hidden,

(b) if level after change is even, 1-bit was hidden.

In step 2. pairs are chosen from sets A, B or C. One VLC code can be in the pair in set B and C, only in B or only in one set (A,C,D). Algorithm prefers more VLC codes from pairs in set B than C. Set C is used by Rate Control Block to provide the same bitrate of MPEG video stream. Rate Control Block stops process of data hiding if replacement of VLC codes from C) set does not guarantee the same amount of bits between time stamps.

To ensure hidden information invisibility only DCT AC coefficients with value higher then threshold are changed in each macroblock. The algorithm is adaptive, because thresholds are determined after observation of blocks activity. In proposed method audio stegochannel contains information useful to restore hidden bits from video stream (information for stego-decoder, how many last coefficients before next time stamp were skipped in hiding process by rate control block (Fig.7).
6 EXPERIMENTS

6.1 Experiments for Audio

In order to illustrate the performance of presented here steganography algorithm, robustness subjective test were carried out.

6.1.1 Conditions for Audio Stagnography Tests

1. Training stage. A training procedure was conducted to determine the centers of groups of audio segments. Different group will use different quantization step $T_k$. In our method the fuzzy K-mean clustering algorithm is used to group audio segments into four groups. In order to make training results statistically significant, training data should be sufficient and cover various genre of audio. Two audio examples were selected: speech track, tracks with music in the background. All data were 44.1 kHz sample rate, one channel and 16 bits per sample. Each audio track was divided into overlapping segments, and the length of each segment was 1024 sample points.

2. Quality measure. The audibility is estimated by a simple subjective listening test. A listener compares the representation of the reference audio track with that of the audio track under test. Finally the listener affirms if audio quality is degraded, which is equivalent to audibility of hidden image in the audio track. Only inaudible cases are further taken under robustness test. As a robustness test we used the Normalized Cross-correlation (NCC) which is adopted to appraise the similarity between the extracted binary image and the original one:

$$NCC = \frac{\sum_{m,n} I_{m,n} I^*_{m,n}}{\sqrt{\sum_{m,n} I_{m,n}^2 \sum_{m,n} I^*_{m,n}^2}}, \quad (7)$$

where:
- $I_{m,n}$ - original bit at position $(m,n)$,
- $I^*_{m,n}$ - extracted bit at position $(m,n)$.

3. Robustness test. We evaluated the robustness of data hiding in speech recording to audio compression, requantization, subsampling and additive noise. In the test, a 64x64 bit binary image was hidden in a audio track (one channel, 16 bits per sample and 44.1 kHz sample rates). The audio track is divided into 128 samples length segments, but the audio feature vector is computed for longer block (1024 samples length) including the segment in the middle position and its adjacent audio samples. The Daubechies-1 wavelet basis is used, and 7-level DWT is performed.

The data was hidden without protection codes to show robustness of audio algorithm in visible way. Error correction code increases NCC up to near 100%.

6.1.2 Audio Steganography Results

The results of tests are presented in Table 1 and 2 for an audio track with speech and track with pop music in the background:

Table 1: Results for data hiding in speech track.

<table>
<thead>
<tr>
<th>Signal attacks</th>
<th>NCC</th>
<th>original image</th>
</tr>
</thead>
<tbody>
<tr>
<td>subampling 22050 Hz</td>
<td>100,00%</td>
<td>The hidden toast</td>
</tr>
<tr>
<td>requantization 8 bits</td>
<td>98,22%</td>
<td>The hidden toast</td>
</tr>
<tr>
<td>mp3 compression 128 kbps</td>
<td>100,00%</td>
<td>The hidden toast</td>
</tr>
<tr>
<td>mp3 compression 96 kbps</td>
<td>99,77%</td>
<td>The hidden toast</td>
</tr>
</tbody>
</table>

6.2 Experiments for Video

The used video material (Fig. 9) was stored with typical for SD quality and resolution. The video stream capacity for each I-frame could be estimated as 3-4 KB, with PSNR=34 dB. Some techniques change not only VLC DCT AC coefficients but moving vectors too, what gives the possibility to hide information in P-frames and B-frames. In normal use this solution increases stegocapacity, but when static scenery
Table 2: Results for data hiding track with pop music in the background (Norah Jones).

<table>
<thead>
<tr>
<th>Signal attacks</th>
<th>NCC</th>
<th>original image</th>
</tr>
</thead>
<tbody>
<tr>
<td>subsampling 220/00 Hz</td>
<td>100/00%</td>
<td>The hidden text</td>
</tr>
<tr>
<td>requantization 8 bits</td>
<td>92/68%</td>
<td>The hidden text</td>
</tr>
<tr>
<td>mp3 compression 138 kbps</td>
<td>100/00%</td>
<td>The hidden text</td>
</tr>
<tr>
<td>mp3 compression 96 kbps</td>
<td>99/55%</td>
<td>The hidden text</td>
</tr>
</tbody>
</table>

(typical for registered by camera built in typical cash machine) is monitored, small amount of motion vectors, with relatively big length are produced in coded video. Modulation of short vectors give low capacity, low PSNR values and higher complexity of stego-codec, as the result of this observation, we decide to use only Intra-blocks.

Figure 9: I-frame of video content used for tests.

7 CONCLUSION

Proposed method of steganography in more than one stream is advanced and effective approach, which increases system security. Using fragile and robust techniques of hiding in one system is interesting extension. Increase of capacity of host streams is not so important in point of view hiding small amounts of bits in intelligent cash machine system but gives possibility to use effective error correcting codes.

The initial research and obtained promising results, confirm that established assumption are right. Further work on data hiding allows to design the high quality tools for improving security of the intelligent cash machine.

ACKNOWLEDGEMENTS

The work presented was developed within VISNET 2, a European Network of Excellence (http://www.visnet-noe.org), funded under the European Commission IST FP6 Programme.

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