

Audio steganography

High frequency message encoding
exploiting human hearing

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Abstract

Steganography is the practice of concealing a file, message, image, or video within another file, message, image, or video. This paper will focus on the domain of digital signal processing and audio concealment of messages. Apart from other common environments such as images or plain text, sound also represents a good way of transmitting hidden messages. The techniques present in this thesis can be applied not only to the target range of human hearing but can also be implemented with some degree of resemblance to other domains of higher frequencies (radio waves or elctro-magnetic waves).

Contents

1	Introduction	2
1.1	Motivation	2
1.2	Context	2
1.3	Objective	2
1.4	Structure	3
2	Theoretical background	4
2.1	Overview of steganography	4
2.2	Overview of digital signal processing	5
2.2.1	Sound	5
2.2.2	DSP(Digital sound processing)	6
2.3	Just a little bit of anatomy	7
3	The application	9
3.1	Chosen method	9
3.2	Environment	9
3.2.1	File encoding formats	9
3.2.2	Python programming language	11
3.2.3	Matlab and matplotlib integrtrion in python	11
3.2.4	Numpy	11
3.2.5	Tkinter UI development in Python	11
3.3	Limitations	11
3.3.1	Hardware limitations	11
3.3.2	Software limitations	11
3.4	Further improvements	11
4	Comparison with existent methods	12
4.1	Other techniques	12
4.1.1	Structural methods	12
4.1.2	LSB (<i>least significant byte</i>) method	12
4.1.3	Phase coding	12
4.2	Advantages/disadvantages	12
	Bibliography	13

Chapter 1

Introduction

1.1 Motivation

Apart being a chance to delve deeper into a domain of computer science where i had previously less experience, this work was also developed with the idea of improving existent security systems in mind. The sounds and how computers "recognize" them present on one had an interesting challenge to make effective use of them to hide messages, and on the other hand a good way of combining the robustness of mathematical methods with the huge processing power of today's computers.

1.2 Context

Today we live in a world where data confidentiality is of utmost importance. While the techniques of cryptography offer far more security potential than plain steganography, the existence of encrypted messages is obvious for an attacker, so attacks are inevitable. A good practice would be to hide this message, being them encrypted or not. This is where steganography could have an impact. As symmetric key cryptography the strength of a technique lies in a secret known to both parties that take part in a communication. Of course some steganographic techniques have weaker hiding systems (Least significant byte method) than others (phase coding or high frequency encoding). By hiding the message it wouldn't present an immediate chance for an attacker as files such as images or audio are sent over the internet with millions each hour.

1.3 Objective

The objective of this work is to study in more detail an area of the art of steganography which is not as frequently used as it should be, and to create an effective way of hiding messages in sound in such a manner that only the sender and the receiver of the message could know to decode. This should consist in a method which has different variable parameters that can be sent in advance between the participants

of the conversation using similar techniques like the know key exchange methods used in everyday cryptography systems such as AES or DES/3DES.

1.4 Structure

This project consists of a theoretical overview of the current steganographic techniques and the description of an algorithm which makes use of high frequency noise to hide certain messages, together with a practical application developed in Python to support the claims made in this report.

Chapter 2

Theoretical background

The commonly stated range of human hearing is 20 Hz to 20 kHz. Under ideal laboratory conditions, humans can hear sound as low as 12 Hz and as high as 28 kHz, though the threshold increases sharply at 15 kHz in adults, corresponding to the last auditory channel of the cochlea. Humans are most sensitive to (i.e. able to discern at lowest intensity) frequencies between 2,000 and 5,000 Hz. Based on this a good way of hiding information is exploiting this human weakness and encoding certain messages in audio files disguised as short sequences of high frequency signals.

2.1 Overview of steganography

The word steganography comes from the greek language being derived from the two greek words *steganos*(covered) and *graphein*(to write) referring to the art of enabling communication that uses methods of hiding information in plain sight. In the field of computer science steganography is the practice of concealing a file, message, image, or video within another file, message, image, or video. The advantage of steganography over cryptography alone is that the intended secret message does not attract attention to itself as an object of scrutiny. Plainly visible encrypted messages, no matter how unbreakable they are, arouse interest and may in themselves be incriminating in countries in which encryption is illegal.[4]

The process of detecting certain messages hidden inside a stego-object (image, sound, video, etc.) is called steganalysis. The goal of steganalysis is to identify suspected packages, determine whether or not they have a payload encoded into them, and, if possible, recover that payload. Unlike cryptanalysis, in which intercepted data contains a message (though that message is encrypted), steganalysis generally starts with a pile of suspect data files, but little information about which of the files, if any, contain a payload. The steganalyst is usually something of a forensic statistician, and must start by reducing this set of data files (which is often quite large; in many cases, it may be the entire set of files on a computer) to the subset most likely to have been altered.

Usual methods of steganalysis consist in structural detection - difference in file properties(size, checksum, headers) or in statistical detection - changes in patterns

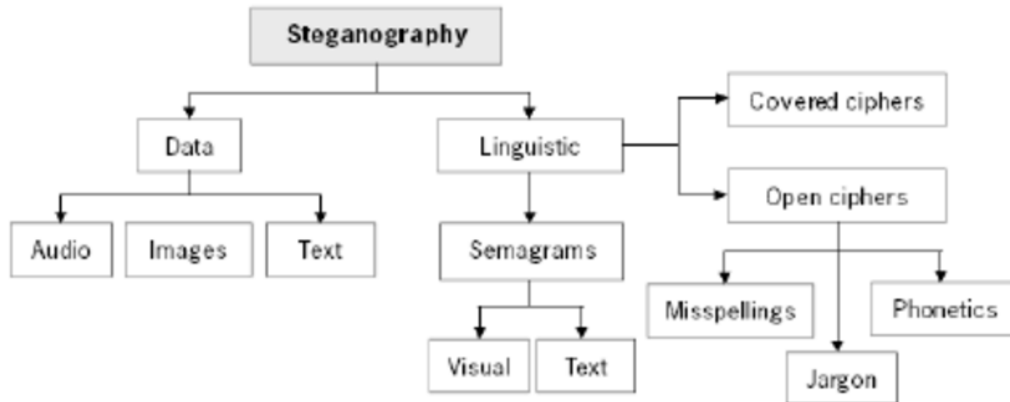


Figure 2.1: General classification of steganographic methods

of bits, LSB changes or histogram analysis.

The algorithm proposed in this paper was designed in order to reduce the efficiency of such methods and to limit the possibility of detection caused by noticeable changes in file structure or composition, or by the appearance of anomalies in the stego-object.

2.2 Overview of digital signal processing

2.2.1 Sound

In physics, sound is a vibration transmitted through different environments: solid, gas or liquids. Sound can be divided mostly in two parts: pressure and time. These two are the characteristic of every wave and because of this a sound can be considered as a continuous wave. Human ear, through its mechanisms, "catches" this stimulus and converts it in electrical signals which are sent to the brain to be analysed. This process is not perfect and this is the key fact on which this method relies.

Although there are many complexities regarding sound transmission, most of the time sound can be characterized by the following properties:

- Frequency
- Amplitude
- Speed of sound
- Direction

In this method we are interested just in the first two properties: frequency and amplitude.

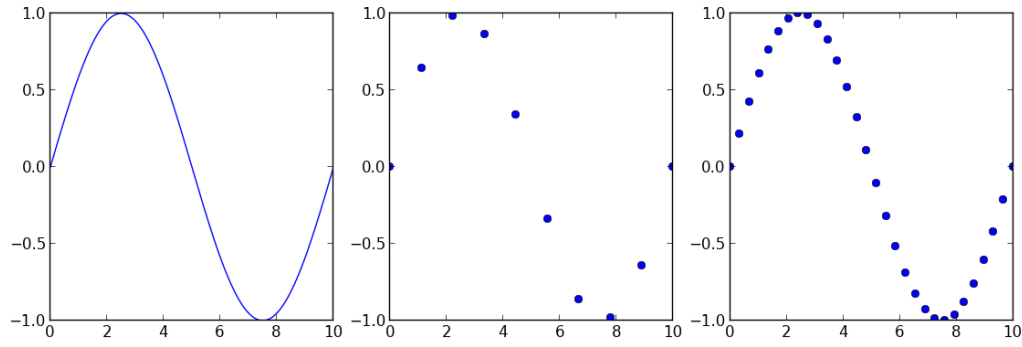


Figure 2.2: Continuous signal splitted in discrete samples

2.2.2 DSP(Digital sound processing)

To be able to analyze an analog(continuous) signal in a digital environment, it must firstly be converted using an analog-to-digital converter. This process is composed of 2 major stages: discretization and quantization. This would transform a continuous signal into a set of consecutive frames equidistant in time, and characterized by a certain amplitude. This is also known as the time domain. It is used to better visualize the overall image of a sound described as a wave. In (fig 2.2) we can observe the discretization of a sinusoidal wave.

Another domain that is of significant importance to us is the frequency domain. Usually to change the time domain into frequency domain, a Fourier transform is applied over the discrete samples of the sound. This process converts the time domain samples into frequencies and amplitudes (fig 2.3).

$$F(\zeta) = \int_{-\infty}^{\infty} f(x) e^{-2\pi x \zeta} dx$$

When the independent variable x represents time, the transform variable represents frequency (e.g. if time is measured in seconds, then frequency is in hertz). Under suitable conditions, f is determined via the inverse transform:

$$f(x) = \int_{-\infty}^{\infty} F(\zeta) e^{2\pi x \zeta} d\zeta$$

This way a sound can be reconstructed exactly by reversing the transformation, mainly using the Fourier inverse transform, to change a set of frequencies to the original sound from which they appeared. Also any change performed over the frequency domain will carry on to the time domain. Using several techniques such as low pass filter or highpass filter, certain ranges of frequencies could be attenuated, this being used mainly to smooth the sound and to clear any noise that could appear during transmission.

The *Nyquist-Shannon* sampling theorem states that a signal can be exactly reconstructed from its samples if the sampling frequency is greater than twice the highest frequency component in the signal. In practice, the sampling frequency is often significantly higher than twice the Nyquist frequency[5]. In this case, where

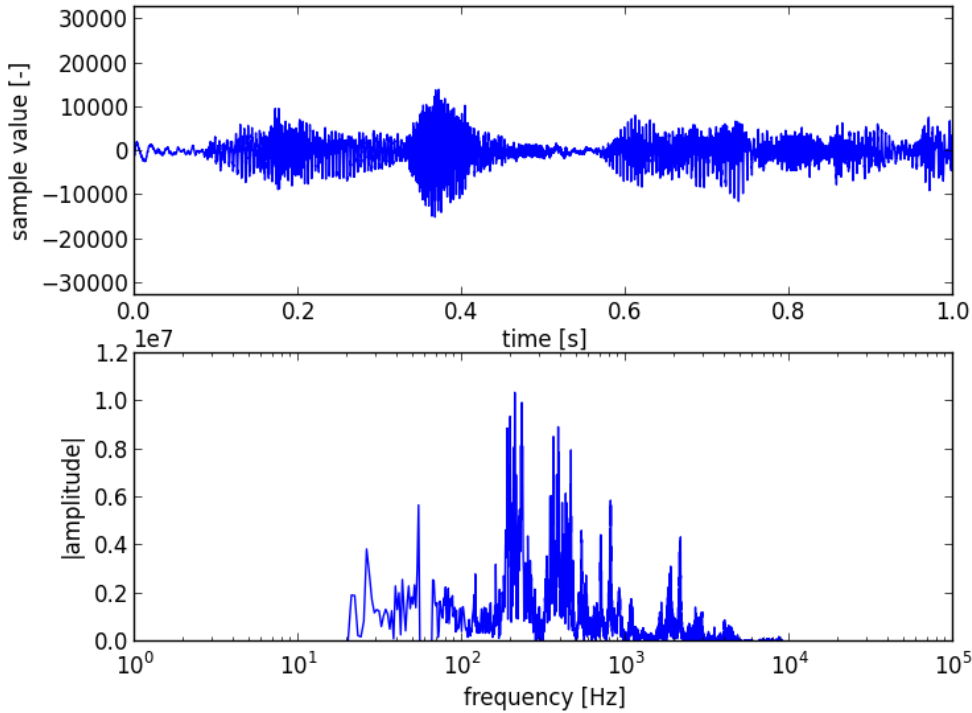


Figure 2.3: Fourier transform applied over a sound

the target range is about 20-26 kHz, a sampling rate of 48000 samples/second is almost at the limit, meaning that some higher frequencies might be omitted and loss of accuracy, so a higher sampling rate was chosen: 96000 samples/second.

2.3 Just a little bit of anatomy

A basic measure of hearing is afforded by an audiogram (fig 2.4), a graph of the absolute threshold of hearing (minimum discernible sound level) at various frequencies throughout an organism's nominal hearing range. The commonly stated range of human hearing is 20 Hz to 20 kHz. Under ideal laboratory conditions, humans can hear sound as low as 12 Hz and as high as 28 kHz, though the threshold increases sharply at 15 kHz in adults. Humans are most sensitive to (i.e. able to discern at lowest intensity) frequencies between 2,000 and 5,000 Hz. Individual hearing range varies according to the general condition of a human's ears and nervous system. The range shrinks during life, usually beginning at around age of eight with the upper frequency limit being reduced. Women typically experience a lesser degree of hearing loss than men, with a later onset. Men have approximately 5 to 10 dB greater loss in the upper frequencies by age 40[6].

This practical application of this work will focus on the range of 15-24 kHz with variable sound amplitude as a program doesn't need a very high signal amplitude

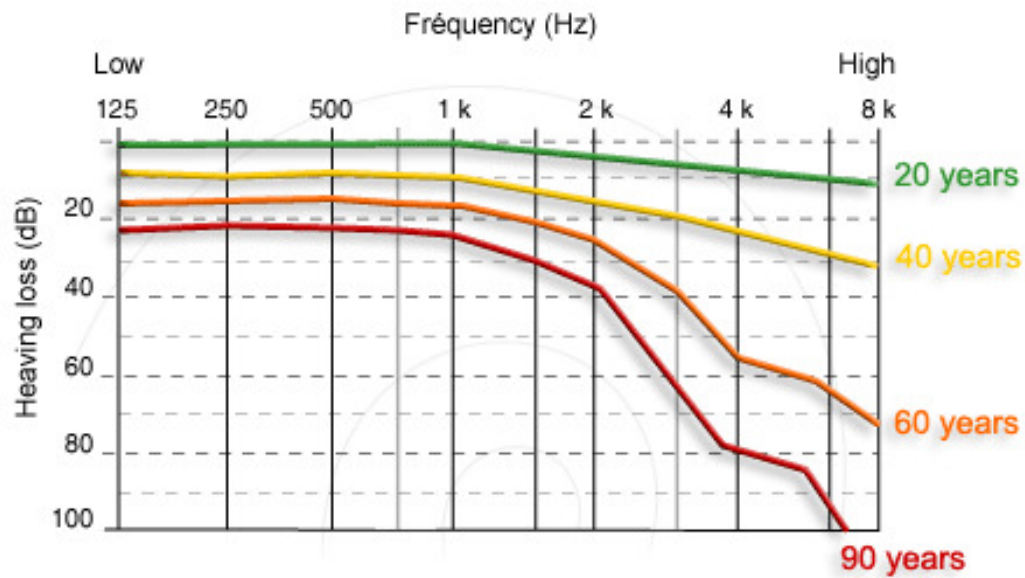


Figure 2.4: Human hearing audiograms based on age[7] MUST CHANGE BAD!!

to recognize it. Preferably the lower end of the encoding range should be as high as possible to exclude the possibility of certain humans hearing the noise added, but due to hardware limitations this threshold shouldn't pass 24-25 kHz (to be further discussed in this paper).

As the range of human hearing is variable from one person to another, there exists the possibility of some persons hearing certain anomalies at the lower end of the frequency specter used in this method, but the application features a mechanism to prevent such detections by raisign the overall height of the frequencies used.

Chapter 3

The application

For proving the concept of hiding information using high frequency signals, an python application was implemented

3.1 Chosen method

The general method consists in several modules each specialized in certain tasks. The aim of the application is to generate sinusoidal wave noise and to insert it inside a carrier file, and after that to be able to extract the information using the fourier transform applied on well defined segments of the sound to get the data from the time domain to the frequency domain where it can be analysed to find if it contains hidden data. By repeating this process on the entire file the message could be recreated with accuracy as long as the sound hasnt been corrupted in the meantime with other high frequency noises.

3.2 Environment

The application consists in a desktop application developed in Python, some additional support scripts developed in Matlab to test the accuracy of the results. For testing purposes an application called Audacity. As for graphical user interface, it was chosen a simple approach using Tkinter.

3.2.1 File encoding formats

The idea behind this method works for any kind of audio format as long as new frequencies can be artificially injected into them. Several audio encoding formats were taken into consideration for this implementation: WAV, MP3 or

WAV format

Due to the simplicity of this format and its representation, WAV was chosen to be the type of the carrier sound for the messages encoded using this method. The WAV

format, also known as WAV (fig 3.1), was developed by Microsoft and IBM and it represented the standard for storing audio data on PCs. It is the main way of storing on Windows data in raw format and typically uncompressed audio files. As a result of this is the rather large size of files stored this way.

This type of format stores chunked data, a file consisting in several different sized chunks. Each chunk has a fixed sized header which describes the properties and the format of the data stored in the corresponding chunk such as number of channels, byte rate or sample rate. The rest of the chunk represents actual information regarding the amplitude of the stored samples. This allows for an easy way of "inserting" artificial noise inside a sound file on a certain channel as long as the previous properties are respected.

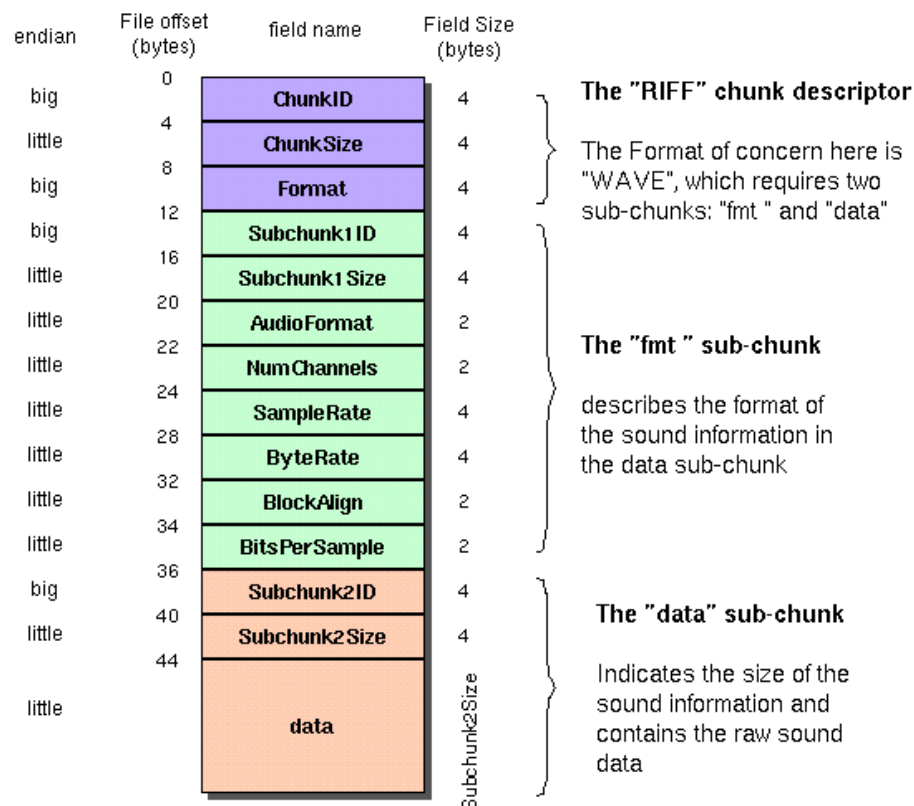


Figure 3.1: WAV format

- 3.2.2 Python programming language**
- 3.2.3 Matlab and matplotlib integrtion in python**
- 3.2.4 Numpy**
- 3.2.5 Tkinter UI development in Python**
- 3.3 Limitations**
 - 3.3.1 Hardware limitations**
 - 3.3.2 Software limitations**
- 3.4 Further improvements**

Chapter 4

Comparison with existent methods

4.1 Other techniques

The number of other methods used for hiding messages has seen a continuous increase over the years. We will present only some of the most frequent used in audio frequency, but this list is far from being exhaustive. Each of these methods comes with its advantages or disadvantages which will be discussed later.

4.1.1 Structural methods

4.1.2 LSB (*least significant byte*) method

4.1.3 Phase coding

4.2 Advantages/disadvantages

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