

Replace the contents of this file with official assignment.
Místo tohoto souboru sem patří list se zadáním závěrečné práce.

Bachelor's thesis

NÁZEV PŘÍKLADNÉ ZÁVĚREČNÉ PRÁCE

Nikita Mortuzaiev

Faculty of Information Technology
Department of Applied Mathematics
Supervisor: Ing. Mgr. Ladislava Smítková Janků, Ph.D.
March 30, 2022

Czech Technical University in Prague
Faculty of Information Technology

© 2022 Nikita Mortuzaiev. Citation of this thesis.

This thesis is school work as defined by Copyright Act of the Czech Republic. It has been submitted at Czech Technical University in Prague, Faculty of Information Technology. The thesis is protected by the Copyright Act and its usage without author's permission is prohibited (with exceptions defined by the Copyright Act).

Citation of this thesis: Mortuzaiev Nikita. *Název příkladné závěrečné práce.* Bachelor's thesis. Czech Technical University in Prague, Faculty of Information Technology, 2022.

Contents

Acknowledgments	v
Declaration	vi
Abstrakt	vii
Acronyms	viii
1 Introduction	1
2 Physical Background	3
2.1 What a Sound Is	3
2.2 Harmonicity and Pitch	5
3 Biological Background	7
3.1 Outer and Middle Ear	7
3.2 Inner Ear	8
3.3 Auditory Scene Analysis	9
4 Mathematical Background	11
4.1 The Basics of Digital Signal Processing	11
5 Computational ASA	13
5.1 Principles, Goals and Applications	13
5.2 Typical Architecture	14
5.2.1 Peripheral Analysis	14
5.2.2 Feature Extraction	14
5.2.3 Grouping	14
5.2.4 Resynthesis	15
5.3 Major Works	15
6 Implementation	17
7 Experiments	19

List of Figures

2.1	A simple vertical mass-spring system	4
2.2	Harmonics of a sound wave	5
3.1	Anatomy of the human ear	8

Chtěl bych poděkovat především sit amet, consectetur adipiscing elit. Curabitur sagittis hendrerit ante. Class aptent taciti sociosqu ad litora torquent per conubia nostra, per inceptos hymenaeos. Cras pede libero, dapibus nec, pretium sit amet, tempor quis. Sed vel lectus. Donec odio tempus molestie, porttitor ut, iaculis quis, sem. Suspendisse sagittis ultrices augue.

Declaration

FILL IN ACCORDING TO THE INSTRUCTIONS. VYPLŇTE V SOULADU S POKYNY.
Lorem ipsum dolor sit amet, consectetur adipiscing elit. Curabitur sagittis hendrerit ante. Class aptent taciti sociosqu ad litora torquent per conubia nostra, per inceptos hymenaeos. Cras pede libero, dapibus nec, pretium sit amet, tempor quis. Sed vel lectus. Donec odio tempus molestie, porttitor ut, iaculis quis, sem. Suspendisse sagittis ultrices augue. Donec ipsum massa, ullamcorper in, auctor et, scelerisque sed, est. In sem justo, commodo ut, suscipit at, pharetra vitae, orci. Pellentesque pretium lectus id turpis.

Lorem ipsum dolor sit amet, consectetur adipiscing elit. Curabitur sagittis hendrerit ante. Class aptent taciti sociosqu ad litora torquent per conubia nostra, per inceptos hymenaeos. Cras pede libero, dapibus nec, pretium sit amet, tempor quis. Sed vel lectus. Donec odio tempus molestie, porttitor ut, iaculis quis, sem. Suspendisse sagittis ultrices augue. Donec ipsum massa, ullamcorper in, auctor et, scelerisque sed, est. In sem justo, commodo ut, suscipit at, pharetra vitae, orci. Pellentesque pretium lectus id turpis.

In Prague on March 30, 2022

.....

Abstrakt

Fill in abstract of this thesis in Czech language. Class aptent taciti sociosqu ad litora torquent per conubia nostra, per inceptos hymenaeos. Cras pede libero, dapibus nec, pretium sit amet, tempor quis. Sed vel lectus. Donec odio tempus molestie, porttitor ut, iaculis quis, sem. Suspendisse sagittis ultrices augue.

Klíčová slova enter, comma, separated, list, of, keywords, in, CZECH

Abstract

Fill in abstract of this thesis in English language. Class aptent taciti sociosqu ad litora torquent per conubia nostra, per inceptos hymenaeos. Cras pede libero, dapibus nec, pretium sit amet, tempor quis. Sed vel lectus. Donec odio tempus molestie, porttitor ut, iaculis quis, sem. Suspendisse sagittis ultrices augue.

Keywords enter, comma, separated, list, of, keywords, in, ENGLISH

Acronyms

DFA	Deterministic Finite Automaton
FA	Finite Automaton
LPS	Labelled Prüfer Sequence
NFA	Nondeterministic Finite Automaton
NPS	Numbered Prüfer Sequence
XML	Extensible Markup Language
XPath	XML Path Language
XSLT	eXtensible Stylesheet Language Transformations
W3C	World Wide Web Consortium

[illegible]

Introduction

Imagine a party. You can hear a variety of sounds: music in the background, conversations between people, noises of somebody coughing, maybe even a dog barking outside. . . . These sounds merge into a single stream that approaches your ears by vibrations in the air and then goes through different physical, biological and psychoacoustical processes to finally come in a form of electrical impulses to the brain. Despite all these sounds from different sources are mixed on the way to your ears, the brain can segregate one (or several) of them. You can focus your hearing on these “target” sounds and separate them from the complex mixture, leaving other sounds in the background. This phenomenon has been described as a “cocktail party effect”, and the process of integrating separate sounds into meaningful streams, or “auditory objects” – auditory scene analysis, or ASA.

In machine perception — specifically in machine hearing — a related concept is referred to as Computational ASA (CASA) and is tightly connected to the fields of sound recognition and digital signal processing. CASA systems indeed aim to separate sounds from mixtures, but they differ from BSS (blind source separation) systems in that they try to do this in a way a human ear does. Being based on and trying to combine works from different fields of science, CASA systems can bring new solutions and insights to the complex problem of signal separation.

The main objective for this thesis is to describe the principles and goals of CASA, existing applications and approaches. Another objective is to practically apply the theoretical knowledge and implement a simple CASA system to separate monophonic music from noise. But before all of this, since this thesis is made for an IT-oriented audience, it is needed to make a brief introduction to the underlying physics and biology.

Thus, the thesis is structured as follows:

Firstly, physical background theory will be provided, including an introduction to what a sound is. Since the implemented system from the practical part aims to segregate music from noise, a special focus in this part will be made on describing harmonic sounds and pitch perception.

Secondly, having in mind that CASA tries to mimic the human auditory system, a brief introduction to the biological structure of the human ear will be made. Here, auditory scene analysis according to Bregman will be introduced too.

Next, to cover the math in the implementation part, the basics of digital sound processing will be described. The related mathematical principles and functions used during the implementation will also be given some attention.

In the following chapter, having all the related theory in mind, an introduction to the main principles and goals of CASA will be made, along with an overview of its applications and selected models.

Then, in the practical part, the focus will be made on describing the implementation of specific parts of the CASA system built for this thesis (see attached medium).

Finally, an overview of the experiments made to test the implemented system will be provided.

Physical Background

Before starting to ponder the structures of the human ear, it is necessary to understand the basics of how sounds work in the real world. It is safe to say that many people don't ask this question – they just make sounds or react to them, unconsciously knowing the outcomes. Human mind has already developed a deep understanding of which sounds are produced under different circumstances – you can easily say what to expect when somebody scratches a blackboard or rings a bell. Some could say that sounds are just “pressure waves that propagate through the air”, but in reality, there is a lot of interesting and complex things beyond this definition to pay attention to. This chapter will introduce the reader to the underlying physics of sound and some interesting related concepts. A special focus will be made on describing harmonic sounds, which are essential to understand to be able to work with music and pitch.

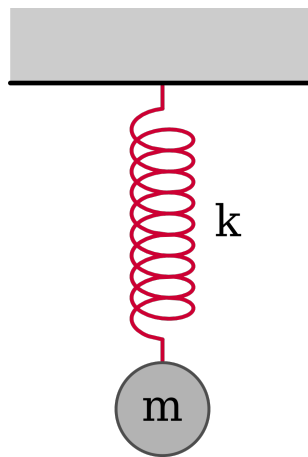
2.1 What a Sound Is

The definition of sound above, saying that it is just vibrations in the air, is hard to be called incorrect from the scientific point of view. Of course, there are improvements to be made: for example, that sound can propagate not only through the air, but through any medium that has inert mass and is “elastic”, or stiff, meaning that it will respond to forces applied to it. Making those corrections, it is also important to note that the definition above relates to sound as a physical phenomenon, but there is another definition that people use mostly in psychology and physiology, saying that the sound is a perception in the brain, or auditory sensation of the concept described above, or “an object of hearing”. It is possible to argue about the question of “What Is Sound?” for a long time, as people still haven't come to a single definition and tend to mix the concepts [1], but in this thesis, the term “sound” will be used primarily in the first, physical sense, unless specified differently.

For better understanding of how physical sound works, keep in mind the mass and elasticity of the air mentioned above. Overall, mass and elasticity (not only of the air, but of any medium) play a very important role in the related studies: mass-spring systems are a highly discussed topic, along with the type of oscillations they tend to have. Any object that can produce sounds may be considered a mass-spring system: a bell, a guitar string, or even air or water, which can be thought of as many small masses connected by invisible springs... This knowledge is quite staggering – in most cases, it is hard to imagine such system, because there could be no obvious mass nor elasticity. Consider an example for explaining resonant cavities: why a can of soda makes that clicking sound when it is being opened? The air is the answer. When you open the can, some parts of the air near its top act as a mass, and other parts near the bottom as a spring.

The pressure in the can drops, and the “spring” at the bottom tries to suck the “mass” back in, producing the expected sound [2].

Now, if you imagine the simplest of such systems, like the one on Figure 2.1, you can notice that when a particular force is applied to it, it tends to oscillate in a sinusoidal manner (due to some famous laws of physics, which will not be further discussed here). In fact, this can be applied to all mass-spring systems: they naturally “want” to vibrate in a sinusoidal fashion with a preferred frequency, called resonance frequency. Sinusoidal vibrations will be given more attention in Chapter 4



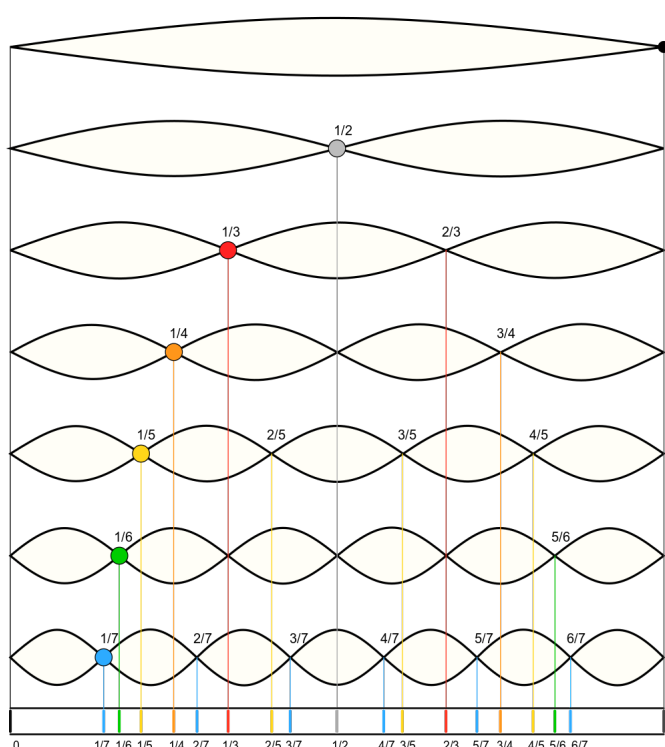
■ **Figure 2.1** A simple vertical mass-spring system. Taken from <https://commons.wikimedia.org/>

When someone talks about applying forces to objects, they can probably say that an impulse is delivered. In classical mechanics, impulse is a widely used concept, but for the purposes of this thesis, it is rather important to note how objects respond to impulses. When an impulse is delivered, the object starts to vibrate at all possible frequencies, but having in mind an understanding of resonance frequencies, it is safe to say that not all applied frequencies have equal amplitudes. Thus, some tones in the resulting sound tend to be louder, and others, if not completely silent, highly attenuated. In signal processing, the related concept is called an impulse response and can find some use when discussing digital filters in Chapter 4. The above-mentioned “chosen” frequencies will be described more in the next section.

Another essential topic to mention here is why sounds fade in time. This is again connected to the concept of mass-spring systems and the amplitudes of their vibrations. Usually, the greater these amplitudes are, the louder the resulting sound is, so if the amplitudes didn’t become smaller, we would live in constant unbearable noise. In brief, the fading is caused by the resistance of the medium, in which the sound propagates, and the manner of this propagating. If you imagine air as it was described above — as many masses connected by invisible springs — the mechanics of the propagation becomes clear: the sound source pushes the closest mass near it, which due to elasticity pushes its neighbors and returns to its starting location. Then its neighbors, in turn, push their neighbors and return, and so on, until these vibrations come to your ears. The air masses must be pushed again and again for the sound to spread, so it tends to lose its strength along the way, and the further from its source it travels, the smaller the amplitudes of the vibrations become.

2.2 Harmonicity and Pitch

The conversation about how harmonics (or overtones) appear was already started in the previous section. In simple words, not all frequencies of the vibrations caused by delivering an impulse to an object keep their amplitudes for long. The ones that benefit the most from this phenomenon are harmonics, which are the periodic waves with frequencies that are positive integer multiplications of a specific frequency called fundamental (Figure 2.2). For example, if the fundamental frequency is 200 Hz, the corresponding harmonics are 400 Hz, 600 Hz, 800 Hz, 1 kHz and so on. Each harmonic can be labeled with a number – the fundamental frequency one is also called the 1st, so the wave with frequency of 1 kHz from the example above would be the fifth. However, the scientific notation for harmonics might be confusing – some authors refer to the fundamental frequency as f_0 (and the fifth harmonic would be f_4 in that case), others as f_1 (and f_5 respectively). In this thesis, fundamental frequency will be notated as f_0 .



■ **Figure 2.2** First seven harmonics of a sound wave (or first seven modes of vibration of a string). Taken from <https://commons.wikimedia.org/>

Another explanation of how harmonics work might be found in [2]. When you pluck a guitar string, it doesn't vibrate only as a whole. The same string might be thought of as two halves, or three thirds, or even one hundred one hundreds, and that each part of it vibrates separately. So, when the string is plucked, all its harmonics are excited, and the resulting sound is not a pure tone, but a complex one. This behavior is often called "modes of vibration" and might be observed not only in strings, but also, for example, in sheets of metal.

The most interesting property of harmonics is that they are all periodic at their fundamental frequency. If you sum up any number of adjacent harmonics of a wave, the period of the resulting wave would be equal to the period of the fundamental. This property plays an impor-

tant role in the perception of pitch and is often used for its estimation in machine hearing systems.

Now, what is pitch exactly? To start using this term in the thesis, it is important to provide a clear definition, but in fact, there is none that is considered a standard. Two most widely used ones were given in [3]. The first one was provided by the American Standards Association (ASA) in 1960 with a reference to music – they defined pitch as *“that attribute of auditory sensation in terms of which sounds may be ordered on a musical scale”* ([3], p. 1). The second one was given by the American National Standards Institute (ANSI) in 1994 without a reference to music, saying that *“Pitch [is] that attribute of auditory sensation in terms of which sounds may be ordered on a scale extending from low to high. Pitch depends primarily on the frequency content of the sound stimulus, but it also depends on the sound pressure and the waveform of the stimulus”* ([3], p. 1). For this thesis, it is enough to consider pitch as the auditory sensation mentioned in both definitions that can be ordered on a scale.

It is important to note that pitch is not a physical property of sound, but perceptual. When someone says “low pitch” or “high pitch”, it is not certainly clear where this “low” or “high” is – low pitch for some people might be high for others. In the related studies of pitch perception in psychophysics, a term “just noticeable difference” (JND) is used, and there are references that humans can distinguish about 1 400 points on the pitch scale.

Pitch is often associated with fundamental frequency, though it is not fully equivalent to it. Experiments have shown that for some short periodic sounds the perception of pitch might not appear at all, though it was clear that they had a fundamental frequency. On the other hand, there are reports saying that sounds with a missing fundamental (the ones made only from higher harmonics) could evoke the perception of pitch associated with the missing frequency, thus giving the illusion of what was not present in reality [2]. Either way, pitch is a major attribute used while describing tones in Western music, as well as is loudness, duration and timbre. Pitch also plays an important role in auditory grouping, given the fact that same sound sources tend to produce sounds of the same pitch. Auditory grouping in humans will be given more attention in Chapter 3.3.

[illegible][illegible]



[illegible][illegible][illegible]

A horizontal row consisting of twenty small blue squares.

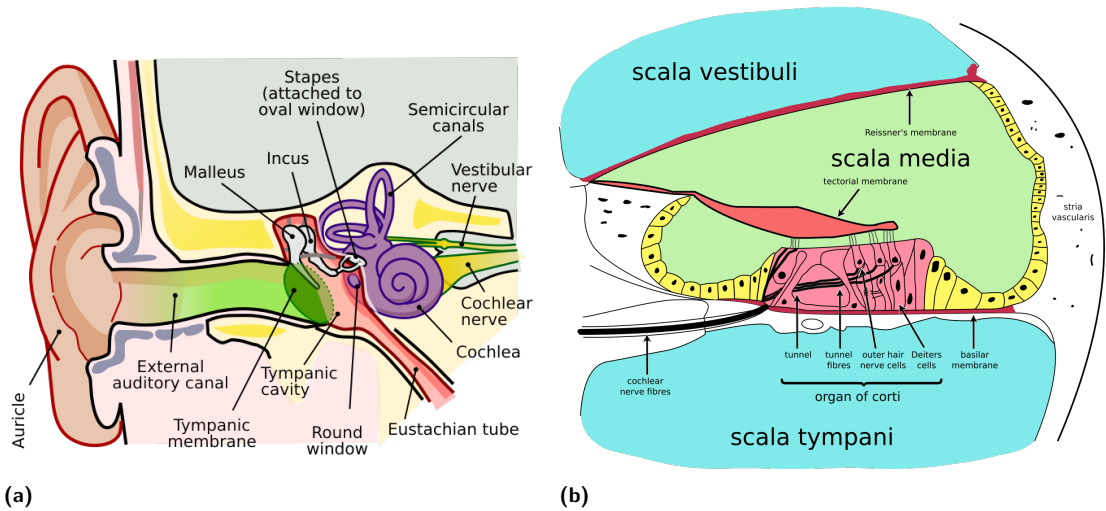


Figure 3.1 (a) Anatomy of the human ear. The ossicles of the middle ear are shown in white. The inner ear is shown in purple. (b) Cross-section of the cochlea showing the organ of Corti and three chambers filled with cochlear fluids. Both pictures were taken from <https://commons.wikimedia.org/>

at least 18.1 times [4].

To regulate the middle ear and protect it from damage due to very loud sounds, two muscles are located inside of it: the stapedius muscle and the tensor tympani muscle. These muscles are controlled by unconscious reflexes and hold the ossicles when the vibrations become too intense. To provide ventilation and drainage of the middle ear and to equalize pressures in this isolated environment, the middle ear is connected to the back of the throat by the eustachian tube [2].

3.2 Inner Ear

The inner ear starts with the above-mentioned oval window, which is connected to the stapes of the middle ear. The oval window is a part of the cochlea — a structure of the inner ear dedicated to hearing. Along with the cochlea, the inner ear also contains the vestibular system, which is responsible for the sense of balance and spatial orientation and uses the same kinds of fluids and cells as the cochlea does. The vestibular system will not be covered in this thesis, but the fluids and cells will be described in more detail later in the section.

The cochlea itself is a spiral-shaped cavity made of bony tissue, which makes about 2.75 turns around its axis and is about 3 cm long. The core component of it is the basilar membrane, which runs along almost its entire length and separates two of the three chambers of the cochlea filled with different fluids [2]: the tympanic duct (scala tympani) filled with perilymph, and the cochlear duct (scala media) filled with endolymph. The third chamber, the vestibular duct (scala vestibuli), is separated from the cochlear duct by the Reissner's membrane and is filled with perilymph (Figure 3.1b). When the footplate of the stapes of the middle ear introduces movements to the cochlear fluids, the basilar membrane is affected too, and the endolymph in the cochlear duct moves along.

The most interesting property of the basilar membrane is that its stiffness and width is different throughout its length – the membrane is narrow and stiff at the basal end of the cochlea, and wide and floppy at the apical end. And here sound waves have two possible routes to take while propagating through the basilar membrane: a shorter path, which includes going through the

stiffer parts of it, or a longer path, which means travelling along the membrane until it becomes easier to pass through, but pushing more fluid on the way. In fact, high-frequency waves tend to choose the shorter path, and low-frequency waves – the longer one. The scale between frequencies passing through the basilar membrane is not linear, but close to logarithmic. In machine hearing systems, equivalent rectangular bandwidth (ERB) scale is usually used.

Thus, the basilar membrane moves in different places depending on the frequencies of the vibrations. The organ of Corti, which sits on top of it and runs along its entire length, contains displacement cells able to respond to movements of the fluid nearby and send electrical impulses when this happens. Such cells are packed with a bunch of stereocilia (hair) that stick out of its top, and thus are called hair cells. These cells can be of two types: inner hair cells that are located closer to the center of the cochlea, and outer hair cells that sit closer to its outer side. Inner hair cells are less numerous than outer hair cells and form a single row along the organ of Corti, while outer hair cells usually form three rows [2].

Now, it is important to mention that the endolymph in the cochlear duct contains high amounts of positively charged ions (primarily potassium and calcium). When it moves in response to the sound pressure, the stereocilia of the inner hair cells are deflected, and tiny ion channels open in them. This allows the charged ions from the endolymph to enter the stereocilia. The cell becomes depolarized, and a receptor potential is produced. This results in releasing the neurotransmitters at the basal end of the cell and then triggering action potentials in the nerve nearby. In this way, inner hair cells detect movements around them and convert mechanical sound waves to electrical nerve signals.

Outer hair cells, in turn, serve as amplifiers of the quiet sounds. Their receptor potentials are converted to cell body movements, thus increasing the sound pressure [5].

3.3 Auditory Scene Analysis

To close up the chapter, it was decided to make an introduction to auditory scene analysis according to Bregman [6]. His book named "Auditory Scene Analysis: The Perceptual Organization of Sound" (1990) made a big influence on further researches, as it attempted to bring together the theoretical knowledge in the field, which did not have any clear base to build on. Bregman's book is now widely recognized as this base, so it is necessary to list at least the primary concepts of ASA described there. This section could have been put to either of the chapters in the theoretical part of the thesis, because it is connected to every field being discussed, but it resides in the biological part, because most of the addressed experiments were testing human auditory perception, which is highly connected to the related studies in Gestalt psychology.

To start off, Bregman brings to the world a new term related to auditory perception. If you recall the two definitions of sound from chapter 2.1, you may remember that there were two of them: one related to sound as a physical phenomenon, and another related to perception in the brain. Bregman introduced a term "auditory stream", or "auditory object", to address the second definition. He made an analogy with vision and how humans tend to group separate surfaces of the same object on their eye retina to see this object as a whole, and referred to "auditory streams" as to the same kind of objects, but for audition. He said, that the term "sound" is not really well suitable in this case, because, for example, a melody in a recording of music consists of different sounds (notes), but people often perceive this melody as a whole and group the sounds into something greater in their perception. Bregman's definition of auditory streams became very popular, so it will be used throughout this thesis too.

Bregman defines ASA as the process of separating these auditory streams from mixtures and refers to it as to a two-stage process. The first stage, segmentation, is said to be splitting the auditory input to so-called "segments", just as the visible object is split to surfaces in the human eye. The second stage is grouping and refers to integrating these segments together based on the grouping cues. With references to experiments from his lab, he describes two possible approaches to grouping and searching for cues: simultaneous (which is also called vertical, or spectral) and sequential (or horizontal). While simultaneous grouping takes into account the segments that appear at the same time, but relate to different frequencies (are spread in space), sequential grouping works with segments that share the frequency component, but are located at different points in time. As an example of a cue for simultaneous grouping one could take common onset and offset, because it is usual for sounds (or different frequency components of the same sound) from the same source to start at the same time. For sequential grouping, probably the most common cue is pitch. If two sounds have pitches associated with fundamental frequencies that are close to each other, the sounds will likely be grouped to the same stream. When the frequencies are further away, the sounds will most certainly appear in different streams.

Mathematical Background

4.1 The Basics of Digital Signal Processing

Computational ASA

Now, having described all the underlying concepts from different fields of science in previous chapters, it is time to finally focus on computational auditory scene analysis. CASA is said to be the study that groups practical, programmable solutions for auditory scene analysis problems, or the study of ASA by computational means. CASA systems are used primarily for source separation, meaning that they are machine listening systems that aim to separate "target" sounds from mixtures, just like people do when try to focus on a specific sound and not to be distracted by others. In that CASA systems differ from systems for blind signal separation – they try to mimic (at least to some extent) the mechanisms inside the human ear, which were discussed in chapter 3. In this chapter, main principles of CASA systems will be described, along with a typical architecture, goals and applications. In the second part, major works that use computational auditory scene analysis for source separation will be reviewed and compared.

5.1 Principles, Goals and Applications

Having the definition of CASA above, to be able to limit the requirements to the models it is necessary to describe the principles of CASA and common concepts across different systems. As the most major one, one could pick the restriction of number of microphones used in the input. Being based on the mechanisms of the human auditory system, CASA models only use recordings from one or two microphones, thus being split to monaural and binaural. Monaural models are researched better, but can't be used for extracting features based on the location of the sound, which is possible to some extent in binaural models, when time differences between the two recordings might be used.

To discuss the goals of CASA, it is useful to refer to the goals of ASA. According to Bregman [6], the primary goal of auditory scene analysis is to produce separate streams from the auditory input. Here, the term "stream" refers to a representation of a distinct sound source in the acoustic environment, but, for example in [7], the authors also use it when talking about these representations in computer memory.

For CASA, Wang [7] proposed that the goal should be to find an ideal binary mask (IBM) for the time-frequency (T-F) representation of the input. If the input is split into T-F units, where time is on horizontal axis and frequency on vertical, an IBM is a binary matrix that has ones in places where the target sound is stronger, and zeroes elsewhere for background units. Ideal binary masks will be discussed in more detail in the following section.

The research of CASA systems and their applications in science [8] have been quite diverse recently. Some of the models are inspired by various biological experiments [9][10], while others are trying to address the cocktail party problem in natural environments [11]. Some models try to explicitly simulate perceptual data, but others may refer to perception only very slightly. The expected output for the system implemented in this thesis is to find an IBM to be able to mask noisy background in monophonic piano music.

Aside from pure scientific interest, CASA systems find useful applications in everyday problems [7]. Some of them are listed below.

Speech recognition. Apparently, the most popular field, where CASA systems have been used. Many speech recognition systems have performance losses in acoustic environments, where multiple sources of sound are present. The development is often put in contrast with computer vision systems that basically fulfill the same purpose, but for another human sense.

Automatic music transcription. A complex problem on its own (even human experts can come up with different solutions) becomes more complicated when multiple musical instruments are involved and need to be transcribed separately. CASA can bring new solutions to these problems.

Hearing prostheses. Modern hearing aids made for people suffering from hearing loss don't separate speech in noisy environments, amplifying the noisy background too. CASA could address this problem to filter the noise out at least to some extent.

Audio information retrieval. Recordings on the Internet usually contain mixtures of sounds from different sources, thus it is necessary to separate them to be able to search efficiently.

5.2 Typical Architecture

The architecture described in the next subsections is based on [7], [12] and [13], though it is impossible to say that it is used in all systems – in different sources the authors use different approaches and methods, and thus different structures of the models.

5.2.1 Peripheral Analysis

Usually, a model for computational auditory scene analysis begins with peripheral analysis of the input sound. Here, based on the knowledge that ASA is a two-stage process, the first, segmentation stage begins. The expected result of this stage usually includes a time-frequency representation of the input sound – a set of so-called T-F units. Since the models try to mimic the human cochlea, the researchers have given it the most attention, and the most common outcome here is a cochleagram.

Cochleagram is produced by a filterbank of N gammatone filters. Gammatone filters were picked as the most close ones to mimic points on the basilar membrane, so different filters from the filterbank represent different points on it. The center frequencies of the filters are not spread linearly in equal intervals, but usually on the ERB (equivalent rectangular bandwidth) scale, which tries to address human hearing.

5.2.2 Feature Extraction

5.2.3 Grouping

5.2.4 Resynthesis

5.3 Major Works

[illegible]

Implementation

..... Chapter 7

Experiments

Bibliography

1. PASNAU, Robert. What Is Sound? *The Philosophical quarterly*. 1999, vol. 49, no. 196, pp. 309–324. ISSN 0031-8094.
2. SCHNUPP, Jan; NELKEN, Israel; KING, Andrew. *Auditory Neuroscience: Making Sense of Sound*. Cambridge, Mass: MIT Press, 2011; 2010; 2012; ISBN 026228975X; 9780262518024; 9780262289757; 0262518023; 026211318X; 9780262113182.
3. PLACK, Christopher J.; OXENHAM, Andrew J.; FAY, Richard R.; POPPER, Arthur N. *Pitch: Neural Coding and Perception*. New York, NY: Springer New York, 2005. ISBN 0387234721; 9780387234724.
4. STANDRING, S.; BORLEY, N.R. *Gray's Anatomy: The Anatomical Basis of Clinical Practice*. Churchill Livingstone/Elsevier, 2008. ClinicalKey 2012. ISBN 9780443066849.
5. HUDSPETH, A.J. Making an Effort to Listen: Mechanical Amplification in the Ear. *Neuron*. 2008, vol. 59, pp. 530–45. Available from DOI: 10.1016/j.neuron.2008.07.012.
6. BREGMAN, Albert S. *Auditory Scene Analysis: The Perceptual Organization of Sound*. Bradford Books, 1990. A Bradford book. ISBN 9780262022972; 9780262521956; 0262022974; 0262521954.
7. WANG, DeLiang; BROWN, Guy J. *Computational Auditory Scene Analysis: Principles, Algorithms, and Applications*. Wiley, 2006. ISBN 9780471741091; 0471741094.
8. SZABÓ, Beáta T.; DENHAM, Susan L.; WINKLER, István. Computational Models of Auditory Scene Analysis: A Review. *Frontiers in Neuroscience*. 2016. ISSN 16624548.
9. WANG, Deliang; CHANG, Peter S. An oscillatory correlation model of auditory streaming. *Cognitive Neurodynamics*. 2008, vol. 2, pp. 7–19.
10. BOES, Michiel; DE COENSEL, Bert; OLDONI, Damiano; BOTTELDOOREN, Dick. A biologically inspired model adding binaural aspects to soundscape analysis. In: Institute of Noise Control Engineering Japan, 2011.
11. ELHILALI, Mounya; SHAMMA, Shihab A. A cocktail party with a cortical twist: how cortical mechanisms contribute to sound segregation. *The Journal of the Acoustical Society of America*. 2008, vol. 124 6, pp. 3751–71.
12. JASTI, Venkata. *Computational Auditory Scene Analysis (CASA) Technique Inspired by Human Auditory System (HAS) - A Review*. 2020. Available from DOI: 10.13140/RG.2.2.18406.45125.
13. VIRTANEN, Tuomas; RAJ, Bhiksha; SINGH, Rita. *Techniques for Noise Robustness in Automatic Speech Recognition*. 1. Aufl.; New York: Wiley, 2012. ISBN 9781119970880; 1119970881; 9781118392669; 1118392663;