# **Studies in Computational Intelligence**

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From Content-based Music Emotion Recognition to Emotion Maps of Musical Pieces



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### **Foreword**

More and more multimedia systems use emotions as a factor in their practical applications. They try to recognize emotions in all forms of content in which humans are emotionally involved—in other words, in texts, music, or films. An example of such solutions is giant music libraries available via the Internet with music search systems using various criteria. Emotions turned out to be one of the more attractive and novel criteria for searching them. Just searching for compositions by title, composer and/or genre was not enough, and adding emotions as an option improved the attractiveness and usefulness of the search systems.

To be able to index music files in terms of emotions, they first have to be identified. Music emotion recognition (MER) is an interdisciplinary problem that affects such fields as signal processing, machine learning, psychology, music theory, and practice. MER requires researchers to be familiar with various distant fields, which on one hand is a complex task and on the other tremendously exciting.

This book presents particular issues that pertain to the creation of a music emotion detection system, such as: deciding on an emotion model, creating ground truth data, extracting and selecting features, and constructing a prediction model. The author focused on content-based MER, i.e., on analyzing the structures contained in music files and finding a relationship with emotions. He focused on examining audio as well as MIDI files and for each presented the relevant feature sets that describe them.

In this work, MER is presented as a classification and regression problem, which is closely connected with the selected emotion model: categorical and dimensional, respectively. Taking into consideration the way humans perceive emotions, the author decided to focus on perceived emotions in music and not felt emotions. During training data construction, he takes advantage of musicians' knowledge who expressed their opinions on the emotions they observed within musical excerpts. The author takes the reader through the specific stages of the emotion recognition system construction and presents the path from music files to the emotion maps created from them, which visualize emotion distribution over time. The book also puts forth practical applications of automatic emotion detection that were tested experimentally.

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Emotions in music are very much connected with humans; we could say that they have a very human nature. They accompany man while composing, performing, or listening to music. Let us hope that computer systems evaluating emotions will not replace humans in creating them, but will only aid us in better understanding their relationship with music.

This book can be seen as a source of information for scientists, academicians, teachers, and students involved in constructing automatic emotion detection systems, as well as musicologists studying and analyzing musical compositions. Additionally, it can serve as an inspiration for other similar experiments conducted with the aim of analyzing emotion distribution in music.

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### **Preface**

This book presents the most important issues with automated systems for music emotion recognition. These problems include emotion representation, annotation of music excerpts, feature extraction, and machine learning. The book concentrates on presenting content-based analysis of music files, which automatically analyzes the structures of a music file and annotates this file with the perceived emotions. Emotion detection in MIDI and audio files is presented.

In the experiments, the categorical and dimensional approaches were used, while for music file annotation, the knowledge and expertise of music experts with a university music education. The built automatic emotion detection systems enable the indexing and subsequent searching of music databases according to emotion. The obtained emotion maps of musical compositions provide new knowledge about the distribution of emotions in music and can be used to compare the distribution of emotions in different compositions as well as for emotional comparison of different interpretations of one composition.

Białystok, Poland May 2017 Jacek Grekow

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