

Cardiac Response to Live Music Performance: Computing Techniques for Feature Extraction and Analysis

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Abstract

Strong emotions and mental stress have been linked to potentially deadly arrhythmias. Music evokes strong emotion through the regulation of tension and release and the modulation of changes and transitions. We exploit this in a novel study involving patients with implanted cardiac defibrillators to study the impact of live music performance on cardiac electrophysiology. The patients' heart rates are artificially fixed with pacing at the higher of 80 beats per minute or 10 above the heart's intrinsic rate. We make continuous recordings directly from the heart muscle whilst the patients are listening to a short classical music concert, which is concurrently recorded in a separate stream. The participants provide annotations of perceived boundaries/transitions and felt tension. The recorded cardiac and music information is further processed to extract relevant features. Here, we describe the experiment design, and the mathematical and computing techniques used to represent and abstract the features from the recorded data. Cardiac reaction is measured by the action potential duration (APD), approximated using the action recovery interval (ARI). The expressive parameters extracted from the music include the time varying loudness, tempo, and harmonic tension. The synchronized information layers allow for detailed analysis of immediate cardiac response to dynamically varying expressive nuances in performed music.

1. Introduction

Strong emotions and mental stress have been linked to abnormal heart rhythms that in some people can be dangerous. However, the mechanisms by which emotions destabilize heart electrical activity and cause these abnormal heart rhythms are not well understood. Music induces strong emotions especially at moments of change or transition through violating or fulfilling expectations. These strong emotions can sometimes be associated with tension

(stress) or pleasure. Here, we monitor cardiac changes during live music performance, comparing them to moments of musical change or transitions, to better understand the interactions between emotion responses and heart rhythm.

In the current study, patients fitted with intracardiac devices—biventricular pacemakers or biventricular ICDs—are invited to a short classical concert featuring music covering a range of moods and containing many changes. Data about participants' musical sophistication is collected. Participants' heart rates are artificially fixed with pacing at 80 beats per minute (bpm) or 10 above intrinsic. While participants listen to the music, electrical signals detected by their pacemaker or ICD from the surface of their heart is downloaded for analysis. Participants also rate their tension (stress) levels through recall, as well as perceived moments of change or transition in the music. We collect the signals and extract salient features so as to analyze changes to the patients' electrical heart signals and compare them to changes and transitions in the music.

This article is based, in part, on our IRAS¹ application no. 242471, approved by the South Central – Oxford C Research Ethics Committee of the UK Health Research Authorities, with preliminary data demonstrations and results.

1.1. Background

Strong emotions and mental stress have been linked to deadly ventricular arrhythmias [1,2]. However, the mechanisms by which emotions destabilize cardiac electrophysiology and cause ventricular arrhythmias are not well understood. Prior studies by the co-authors have demonstrated direct electrophysiological effects of mental stress on the heart [3,4]. Using movie clips, statistically significant decreases were observed to the mean action potential duration (APD) in both the left and right ventricles between high stress mid-movie and end-movie sequences, and the control (matched breathing without movie). There

¹www.hra.nhs.uk/about-us/committees-and-services/integrated-research-application-system

was also statistically significant shortening of the APD between the low-stress movie-start sequence and the high-stress mid-movie and end-movie sequences. While related studies using music exist [5–8], none have explored APD response, nor considered ecological music sources.

Our aim here is to study the impact of music, in particular music in ecological live performance situations and the emotions and stresses it induces, on cardiac electrophysiology. The goal is to correlate analytical measurements of expressive parameters such as tempo, loudness, and harmonic tension with listeners’ physiological responses to identify musical signatures that provoke anxiety as well as promote pleasure, and chart their effects on cardiac electrophysiology so as to increase the understanding of the pathways that affect mood and heart rhythm.

2. Study Design

This section gives the rationale for the study design.

2.1. The Concert

For ecological validity, our study was designed to take place in a live concert setting. Each of the three concerts was attended by two-to-three patients to ensure device programmer and cardiologist coverage. Prior studies mentioned above focussed only on recorded music stimuli, either short duration audio of acoustic music or synthesized MIDI files. Studies have shown that emotion responses to music are heightened in situations of live performance [9], our choice for maximal cardiac response.

The music stimuli have also been chosen to exhibit sharp (and gradual) changes in musical properties, and to range widely in tension. The first concert featured Chopin’s *Balade No.2*, a romantic piano work, and music based on rhythms transcribed from ECG recordings of transitions in/out of Ventricular Tachycardia (VT) during an electrophysiology (EP) procedure. The latter were collaged from the Chopin Ballade and Mars from Holst’s *The Planets*, which had similar rhythms, and are potentially stressful to the patients for non-musical reasons. Jonathan Berger’s “Intermezzo,” a modern piece, and Wilhelm Kempff’s piano arrangement of the “Siciliano” from Bach’s flute sonata (BWV 1031) were added to subsequent concerts.

2.2. Heart Rhythm Acquisition

Prior to each performance, participants’ pacemakers or ICDs are programmed from CRT to dual chamber pacing and fixed at the higher of 80 bpm or 10 above the heart’s intrinsic rate. Pacing slightly above resting heart rates is essential because the APD is strongly affected by heart rate, decreasing with increasing heart rate and vice versa.

Patients are given ten minutes to adjust to the new setting while comfortably seated. Pacing, audio, and video are synchronized with claps. Pacing is maintained for the duration of the performance, then reverted to the original settings. Intracardiac electrogram (EGM) signals at 512Hz resolution are downloaded from the pacemaker or ICD lead connected to the left ventricle (LV Distal tip 1) whilst participants are listening to the music.

2.3. Music Annotations

After each concert, participants provide information on the moments of change—abrupt alterations of acoustic properties or music structure—and transition—changes that take place over a period of time—that they perceived in the music. Audio and video recordings of the concert they just witnessed are played back to the participants during the annotation session to assist in recall. Boundary annotation is done with only audio feedback. Next, participants rate the tension felt during the performance—for the first concert, emotion annotations were also collected. The annotations were collected via existing (ELAN²) and bespoke software. Participants also complete a reduced Goldsmiths Musical Sophistication Index (Gold-MSI³) survey to assess their musical background.

3. Computational Methods

This section describes how the APD and musical parameters are extracted from the EGM and music audio signals.

3.1. EGM: Action Potential Duration

Cardiac reaction is measured by the APD, an electrical parameter of major importance in the development of serious and fatal rhythm disturbances. As in [4], the APD is approximated using the ARI computed from the EGM signal, p . The start of each segment is located using the

²tla.mpi.nl/tools/tla-tools/elan

³www.gold.ac.uk/music-mind-brain/gold-msi

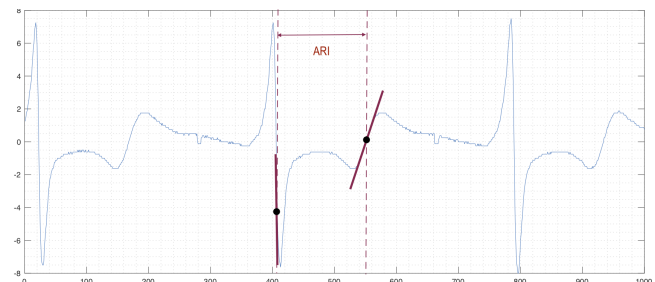


Figure 1. Activation recovery interval: plot shows the EGM signal, p , vs. frames (512 Hz)

findpeaks function in MATLAB with suitable thresholds. The EGM signal from start to end of the altered pacing is then smoothed using loess regression (second-degree polynomial, span 20) to give \tilde{p} . A difference function then estimates the derivative, dp/dt , from the smoothed EGM signal. Inside a given segment, the ARI is given by the time between $\arg \max_t dp/dt$ and $\arg \min_t dp/dt$ as shown in Figure 1. The second plot in Figure 2 shows ARI values with a three-point weighted moving average smoothing. Normal probability plots (not displayed) show that the ARI values follow an approximate normal distribution. Highlighted in the plot are data points more than 1.96 standard deviations from the mean. The graph shows that, for this patient, there are many more statistically significant long ARIs before the concert; and, more statistically significant shorter ARIs (associated with stress) after the concert.

3.2. Audio: Loudness and Tempo

Loudness is one of the most primal features in communication through music and sound. It is a perceptual quality that changes with frequency (same intensity sounds of different pitch have different loudness), duration (sensing loudness takes time, and transient loudness can be modeled by the peak while sustained loudness by the average) and masking by sounds in the same frequency band. We extract loudness values (in sones) from the recorded audio using the method outlined in [10], which uses the MATLAB music analysis (MA) toolbox⁴. An example output can be seen in the third plot in Figure 2. Note that the cluster of statistically significant low ARI values around 840s and 860s coincides with the first loud section in the Chopin.

Tempo tells us the rate at which the music is progressing from beat to beat, where beats are pulses one might tap to while listening to the music. Even state-of-the-art music beat trackers frequently fail when applied to music with highly variable beats, as is the case here, so the performer provides the beats (in seconds), $[b_i]$, by marking them manually on the recorded audio signal using SonicVisualiser⁵. The instantaneous tempo (in beats per minute), $[v_i]$, is then $60/(b_{i+1} - b_i)$ at time $t_i = (b_{i+1} + b_i)/2$. The bottom plot in Figure 2 gives the performed tempo for Chopin, and the before/during/after VT rhythms of the Arrhythmia Suite, approximately the RR intervals of the ECGs from which the pieces were derived. Note that the arousing loud section with the low ARIs in the Chopin is also a fast one.

3.3. Score: Harmonic Tension

Musical tension is an essential part of music. Musicians choreograph expectation to evoke strong emotions. Expectation leads to anticipation; anticipation induces stress

⁴www.pampalk.at/ma

⁵sonicvisualiser.org

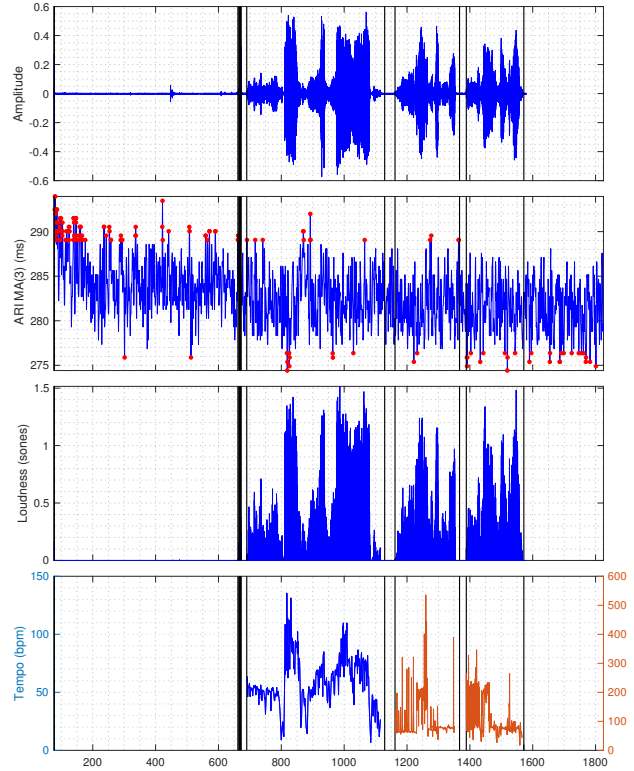


Figure 2. Action recovery intervals (second from top) computed from the EGM, aligned with the music audio signal (top) and audio loudness and music tempo/rhythm information; vertical lines delineate boundaries between the Chopin Ballade and the two Arrhythmia Suite pieces

and dopamine release [11]. The realization of the expected outcomes can trigger tension release, activating pleasure channels associated with food, drugs, and sex [12].

While tension can stem from a number of sources, including silence, the focus here is on harmonic tension. We use the method described in [13, 14] based on the spiral array [15] to compute three measures of harmonic tension: (1) cloud diameter (degree of dissonance or clashing sounds); (2) cloud momentum (how much/quickly the harmony changes); and, (3) tensile strain (how far the context has veered from the global key). Figure 3 shows the harmonic tension parameters aligned with the audio and ARI.

The lowest ARI value, 274.4 ms at 818.6 s, marking the greatest stress response, occurs in the first instance of the stormy passage in the Chopin. It is not only loud and fast, it is also the moment of great harmonic instability, marked by high and rising cloud diameter/momentum and tensile strain. The highest ARI value during the concert, 292.0 ms at 892.2 s, occurs in the languid, slow section after the storm, at the long-awaited harmonic resolution (release) back to the global key, with low diameter and tensile strain.

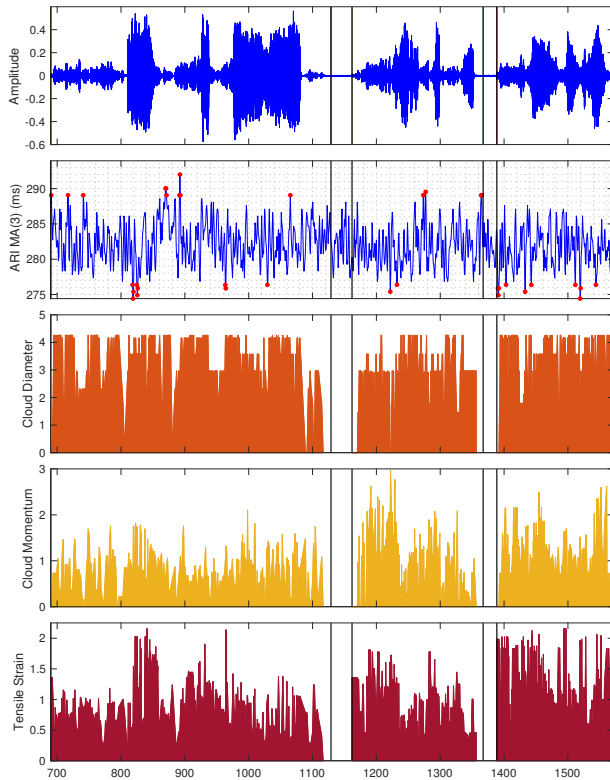


Figure 3. Action recovery intervals, aligned with the music audio (top), and three harmonic tension measures (bottom): dissonance (cloud diameter), chord change (cloud momentum), distance from global key (tensile strain)

4. Future Work

We have described the design and methodology of a research project that explores the connections between music and electric activity of the heart with the aim to understand the pathways that affect mood and heart rhythm. The initial prognosis is promising. By synchronizing and comparing the evolving APD to time-varying expressive parameters (loudness, tempo, and harmonic tension) in the music, we can detect instantaneous stress (and relaxation) responses and link them to developments in the music. Future analyses will reveal individual and group response patterns to specific structures in the live music performance.

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