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# Automated Musical Rhythm Transcription of ECG RR Interval Time Series as a Tool for Representing Rhythm Variations and Annotation Anomalies in Arrhythmia Heartbeat Classifications

Gonzalo Romero-García<sup>1</sup>, Paul Lascabettes<sup>1</sup>, Elaine Chew<sup>2</sup>

<sup>1</sup> STMS, Sorbonne Université, IRCAM, CNRS, Ministère de la Culture, Paris, France

<sup>2</sup> Department of Engineering and School of Biomedical Engineering & Imaging Sciences, King's College London, London, United Kingdom Contact: romero@ircam.fr

#### Introduction

What we present: We introduce a linear time algorithm for transcribing RR intervals into musical rhythms [1, 2] using approximate common divisors (ACDs) [6].

What we obtain: We transcribe long term ECGs records to have a

### **Rhythm & label patterns**

We compare the rhythms with labels of the Long-Term AF Database [4, 5].

For example, N-A-N is most often transcribed rhythmically as 2-4-3.

rhythmic representation of them.

What is this for: The transcription of ECG signals may be useful for improving labelling of heartbeats in ECG signals.

### **Approximate Common divisors**

Let  $T = (t_0, t_1, ..., t_n)$  be a series of onset times. Every *a* is associated with an error measure,

$$\varepsilon_T(a) = \max_n \min_{m \in \mathbb{Z}} |t_n - am|.$$
(1)

We say that a > 0 is an approximate common divisor (ACD) of T with threshold  $\tau > 0$  if

1.  $\epsilon_T(a) \leq \tau$  2.  $\epsilon_T$  has a local minimum at a



Labels	Nb. of 2-4-3	Percentage	Total count
A-A-N	87	3.93~%	309
N-A-N	2104	95.03~%	3869
N-N-N	21	0.95~%	91839
N-V-N	2	0.09~%	3

Table 1: Beat category labels for the rhythm pattern 2-4-3 in LTAFDBRecord 06.

## Visualising transcriptions of triplets

We visualise the rhythm transcriptions of three-interval rhythms using the 2-dimensional *rhythm simplex* [3]

For any point (x, y, z) in the rhythm simplex, where x + y + z = 1, the *arrhythmic norm* is

$$||(x, y, z)||_{\mathbf{R}} = ||(x, y, z) - (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})||_{2}, \qquad (2)$$

which measures the irregularity of the RR intervals. Sinus rhythm will result in norm values at or close to zero; e.g. the rhythm  $(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$  has a norm of 0.

Figure 1: The  $\epsilon_T$  function of T = (0, 0.98, 1.52) for the range  $a \in [0.2, 1]$ . The ACDs are 0.5, 0.312 and 0.25.

# ACD graph

Step 1: We split the sequence in frames of length n
Step 2: For each frame we compute the possible ACDs
Step 3: We create a graph where each ACD is a node
Step 4: We link ACDs of consecutive frames if they share a common duration

[1 1] [2 1] [2 2] Rhythms with big norms are likely to stem from highly contrastive durations.

**Example:** the rhythm (2, 2, 8) corresponds to  $(1/_6, 1/_6, 2/_3)$  and has a norm of 0.41



### **References & Acknowledgements**

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- 10<sup>3</sup>

- 10<sup>2</sup>

 $10^{1}$