

Research Protocol

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1 Identification of the project

1.1 Title of the protocol (in Portuguese and English)

“Detecting life-threatening patterns in Point-of-care ECG using efficient memory and processor power.”

“Detectando padrões de risco de vida no ECG no local de atendimento usando memória e potência do processador eficientes.”

1.2 Author

Francisco Bischoff

1.3 Key-words

anomaly detection, ECG, fading factors, matrix profile, time series, point-of-care

1.4 Project start date and duration

From the 3rd quarter of 2020 to the fourth quarter of 2022.

2 Abstract or Summary (in Portuguese and English)

Atualmente, o monitoramento de ECG junto ao doente (point-of-care) funciona com dispositivos que apenas mostram informação ou disparam alarmes para ritmos cardíacos anormais, usando intervalos normais predefinidos como gatilho. Por outro lado, máquinas completas de ECG de 12 derivações são muito complexas para serem usadas como monitores simples e são usadas com técnicas estritas para o diagnóstico formal de patologias da condução elétrica cardíaca, e o diagnóstico automático é derivado de uma análise completa dos dados das 12 dimensões após totalmente coletados. Ambos os sistemas não funcionam com linhas desconectadas e movimentos do paciente, sendo estritamente necessário ter um sinal bom e estável para permitir o diagnóstico adequado.

Esta investigação tem como objetivo identificar padrões elétricos cardíacos anormais usando dados em tempo real, especificamente aqueles padrões que implicam risco de vida, e fornecer um sinal confiável para que as Unidades de Terapia Intensiva respondam rapidamente a essas situações.

O desenho do estudo é comparável a um estudo de diagnóstico, onde a alta precisão é essencial. Serão utilizados os conjuntos de dados da Physionet e o algoritmo tentará minimizar os falsos negativos e positivos.

O resultado esperado é a concretização de um novo método que, além de preciso, realiza essa tarefa usando tecnologia de ponta de análise de séries temporais que requeira o mínimo de memória e processamento para resolver esse problema. Além disso, espera-se que a utilização de “fading factors” possam contribuir para o estado da arte desta tecnologia.

A equipe de pesquisa é experiente em séries temporais e estuda o “Matrix Profile” desde suas raízes, sendo fundadores da “Matrix Profile Foundation,” cujo objetivo é ter uma API multi-linguagem concisa e estável para desenvolvimento com a tecnologia “Matrix Profile.”^{1,2}

Currently, Point-of-Care (POC) ECG monitoring works either as plot devices or alarms for abnormal cardiac rhythms using predefined normal trigger ranges. On the other hand, full 12-derivation ECG machines are complex to use as simple monitors and are used with strict techniques for formal diagnostics of hearth electric conduction pathologies, and the automatic diagnostics are derived from a full analysis of the 12-dimension data after it is fully collected. Both systems do not handle disconnected leads and patient’s motions, being strictly necessary to have a good and stable signal to allow proper diagnosis.

This research aims to identify abnormal hearth electric patterns using streaming data, specifically those who are life-threatening, being a reliable signal for Intensive Care Units to respond quickly to those situations.

The study design is comparable to a Diagnostic study, where high accuracy is essential. It will use the Physionet datasets, and the algorithm will try to minimize the false negatives and false positives.

The expected result is the concretization of a new method that, besides being accurate, accomplishes this task using state of the art technology for time series analysis that allows minimum space and processor

power to solve this problem. Also, we expect that fading factors can contribute to the state of the art of this technology.

The research team is well experienced in time-series and has studied the Matrix Profile since its beginning, being founders of the Matrix Profile Foundation whose goal is to have a concise and stable cross-language API for developing with the Matrix Profile technology.^{1,2}

3 Rationale

Currently, Point-of-Care (POC) ECG monitoring works either as plot devices and/or alarms for abnormal cardiac rhythms using predefined normal trigger ranges. On the other hand, full 12-derivation ECG machines are complex to use as simple monitors and are used with strict techniques for formal diagnostics of hearth electric conduction pathologies, and the automatic diagnostics are derived from a full analysis of the 12-dimension data after it is fully collected. In CinC/Physionet Challenge 2015, it has been reported that up to 86% resulting of the alarms are false and this can lead to decreased staff attention and increase in patients delirium.³⁻⁵

4 Research question and aims

This research aims to identify, on streaming data, abnormal hearth electric patterns, specifically those who are life-threatening, in order to be a reliable signal for Intensive Care Units to respond quickly to those situations. It also may be able to continuously analyze new data and correct itself shutting off false alarms. Primarily an experiment will be conducted using 2 main algorithms that use Matrix Profile in detecting context changes: SDTD and FLOSS. One uses whole data training and testing, and the other uses a streaming approach that is our main interest. The goal will be detecting the transition from normal to flutter/FA to normal condition with special attention to not rely on rhythm changes. Being this successful, a more generalistic approach will be attempted: to detect changes from normal to abnormal to normal conditions, with special attention to handle with disconnected leads or patient movements. Finally, this research can prove to be a good addition to the Matrix Profile method, using fading factors in order to reduce memory and space consumption, lowering the processor power needed, allowing this algorithm to be used in almost any device.

5 Background / Literature review

In 2015 the PhysioNet/Computing in Cardiology has launched a challenge to address the problem of high false alarm rates by encouraging the development of new algorithms to improve the specificity of ICU alarms⁶. This challenge comprised of minimizing the false alarms for five life-threatening arrhythmia: asystole, extreme bradycardia, extreme tachycardia, ventricular tachycardia and ventricular fibrillation or flutter.

There are other arrhythmias that this challenge didn't assessed, like atrial standstill (hyperkalemia), third-degree atrioventricular block and others that may be life-threatening in some settings like atrial fibrillation (AF), a, atrialflutter and paroxysmal supraventricular tachycardia. Pulseless electrical activity is a frequent condition in cardiac arrest but cannot be identified without blood pressure information.

They used as score the following formula:

The five-best scores (for real-time) in this challenge were:

Table 1: Challenge Results

Score	Authors
81.39	Filip Plesinger, Petr Klimes, Josef Halamek, Pavel Jurak
79.44	Vignesh Kalidas
79.02	Paula Couto, Ruben Ramalho, Rui Rodrigues
76.11	Sibylle Fallet, Sasan Yazdani, Jean-Marc Vesin

Score	Authors
75.55	Christoph Hoog Antink, Steffen Leonhardt

A literature review will be conducted to assess the state of the art for ECG automatic processing:

- The memory and space used to perform the main goal of the algorithm (sound an alarm for ex.) will be collected if available.
- The type of algorithms used to identify ECG anomalies
- The type of algorithms used to identify specific diagnosis (like a flutter, hyperkalemia, etc.)
- Their performance (accuracy, ROC, etc.)

A broad search will be conducted on Pubmed, Scopus, Google Scholar, device manuals, and other specific sources.

Keywords:

- ECG AND monitoring AND ICU
- ECG AND [time series]
- ECG AND automatic AND interpretation

Articles published after “The PhysioNet/Computing in Cardiology Challenge 2015: Reducing False Arrhythmia Alarms in the ICU,” will also be analyzed.

6 Research plan and methods

6.1 Type of study

This will be a diagnostic study as the algorithm must classify the change in pattern as positive or negative for life-threatening.

6.2 Selection of data

Initially, the data used for exploring the properties of the algorithm will be publicly available data on Physionet^{6,7}.

It will be asked for Physionet’s permission to use more sensitive data if needed.

It is desirable that real data extracted from Portuguese ICU could be used in the final stage to assess in real settings the validity of the model.

6.3 Sample size

There is no upper size limitation for the sample size. At least one hundred cases may be reasonable to start with.

6.4 Variables

The first available dataset contains either 549 conventional 12-lead resting ECGs or the corresponding measured Frank Lead System ECGs. The ECGs are digitized at a sampling rate of 1000Hz (0.5 μ V/LSB; 16 Bit ADC). On special request, this database may be available at sampling rates up to 10,000Hz.

Every patient is supplied with an information string containing age, gender, diagnosis, and where applicable, data on the medical history, medications and interventions, coronary artery pathology, ventriculography, echocardiography, and hemodynamics.

These variables may or may not be useful for increasing the sensitivity or specificity of the algorithm. It is planned to use the minimum set of derivations from the 12-lead ECG to classify at first a common Atrial Fibrillation.

6.5 Statistical analysis

The Statistical analysis will be performed using R language v3.6.0 or greater, and it will be computed the ROC curve for the algorithm.

7 Research Team

- Thesis Author: Francisco Bischoff
- Supervisor: Professor Pedro Pereira Rodrigues
- Co-supervisor: Professor Eamonn Keogh (UCR, Riverside)

8 Tasks, milestones and timeline

8.1 Tasks

The timeline is composed of larger tasks I call Epics. They contain multiple subtasks that are expected to change frequently.

- **Elaboration of Research Protocol**
 1. Duration: 1 months 12 days;
 2. Elaboration of this protocol in order to facilitate the management and overview of the project;
 3. This task was developed by the author with input suggestions from other experts.
- **Literature Review**
 1. Duration: 2 months 12 days;
 2. This task aims to survey the literature about what is currently done to tackle the current problem and what th limitations are; Aim and outputs for the task (and relation with the next task);
 3. This task will be done with three independent reviewers using the PRISMA guidelines in the Covidence framework.
- **Obtaining Access to Physionet full data**
 1. Duration: 1 months 4 days
 2. All datasets in Physionet are supposed to be Open Access. However, there is a chance that some datasets may nee permissions.
 3. If any dataset needs permission, it will be first evaluated the real need and asked the proper way to access it.
- **First Experimentation with Public data**
 1. Duration: 2 months 9 days
 2. The Physionet Challenge from 2015 will be the first dataset to be analyzed and will be *a study in scarlet* fo the problems we may face in this kind of dataset;
 3. The datasets will be studied in the case of data preparation for the modeling process.
- **Development of the First Algorithm**
 1. Duration: 4 months 13 days;
 2. In this task, the first model will be constructed: the Atrial Fibrillation start/end detection;
 3. The state of the art methods will be used to detect such changes, with maximum precision and lowest memory an processor usage;
 4. This task depends on the knowledge about the dataset we have from the previous task.

- **Dissertation First Draft**

1. Duration: 2 months 3 days;
2. This task aims to, at the same time, create a draft for the final dissertation, and the content for an actual article to be published;
3. This task depends on the concretization of the previous task.

- **Publication of the First Algorithm**

1. Duration: 3 months 29 days
2. This task aims to refine the text, review, and submit it for publication.
3. The length of this task depends on several variables, including the journal review time;
4. This task depends on the previous task;
5. Financial needs: Publication fees.

- **Development of the Second Algorithm**

1. Duration: 2 months 14 days
2. In this task, the second model will be constructed: an attempt to generalize it for any life-threatening ECG change;
3. The state of the art methods will be used to detect such changes, with maximum precision and lowest memory and processor usage;

- **Dissertation Second Draft**

1. Duration: 3 months 25 days
2. This task aims to, at the same time, create a second draft for the final dissertation, and the content for an actual article to be published;
3. This task depends on the concretization of the previous task.

- **Publication of the Second Algorithm**

1. Duration: 3 months 27 days
2. This task aims to refine the text, review, and submit it for publication.
3. The length of this task depends on several variables, including the journal review time;
4. This task depends on the previous task;
5. Financial needs: Publication fees.

- **Dissertation Review**

1. Duration: 1 months 15 days
2. This task will be a time to review all the work done and prepare it for final presentation;
3. Ideally, two or more independent expert shall read the thesis and give feedback for improvement;

- **Proof Reading**

1. Duration: 0 months 25 days
2. This task comprises in careful reading, ideally by a professional in the English language;
3. It depends on the previous tasks;
4. Financial needs: Proofreading fees

- **Presentation**

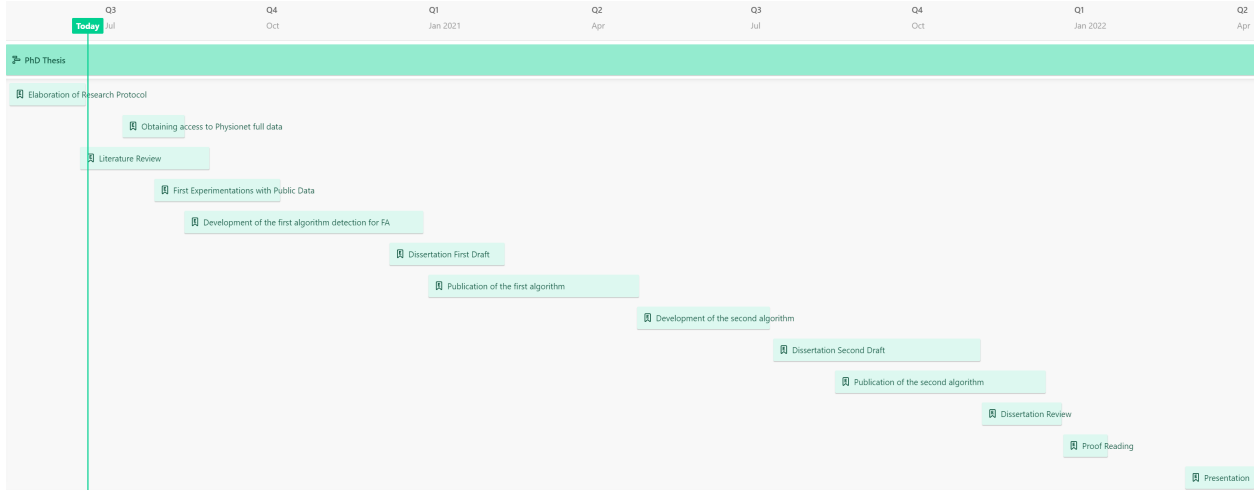
1. Duration: 1 months 11 days
2. This task comprises in preparation for public presentation;
3. It includes the formulation of the slides or any multimedia support that shall be needed;
4. This task depends on having the dissertation done.

8.2 Milestones

Table 2: Milestones

Milestone	Date	Name	Description
M1	Jul, 2020	Protocol	Finish and Deliver Protocol
M2	Oct, 2020	Literature Review	Finish Literature Review
M3	Apr, 2021	Paper 1	Finish and Submit for Publication Paper 1
M4	Jan, 2022	Paper 2	Finish and Submit for Publication Paper 2
M5	Dec, 2022	Thesis	Finish and Deliver Ph.D. Thesis

8.3 Timeline



Click on the image to open an interactive Gantt webpage

9 Budget

Table 3: Expected expenses

Items	Budget	Justification	Obtained
Travel expenses	5000	International Conferences	No
Conferences	3000	Registration Fees	No
Tuition Fees	11000	Researcher Maintenance	No

10 Expected results and outcomes

10.1 Expected results

It is expected that a novel algorithm to detect life-threatening ECG changes can be achieved using lower memory and processor power than the existing ones, maintaining the overall performance level.

10.2 Outcomes

This research will yield at least two publications in indexed journals as well as the final thesis will be available in the university repository.

References

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