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1 Mini Review

Heart Failure: Diagnosis, Severity Estimation and Prediction of Adverse Events Through Machine Learning Techniques

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36 37 ABSTRACT

Heart failure is a serious condition with high prevalence (about 2% in the adult population in developed countries, 22 and more than 8% in patients older than 75 years). About 3–5% of hospital admissions are linked with heart failure 23 incidents. Heart failure is the first cause of admission by healthcare professionals in their clinical practice. The 24 costs are very high, reaching up to 2% of the total health costs in the developed countries. Building an effective 25 disease management strategy requires analysis of large amount of data, early detection of the disease, assessment 26 of the severity and early prediction of adverse events. This will inhibit the progression of the disease, will improve 27 the quality of life of the patients and will reduce the associated medical costs. Toward this direction machine 28 learning methodologies applied for the assessment of heart failure. More specifically, models predicting the presence, estimating the subtype, assessing the severity of heart failure and predicting the presence of adverse 31 events, such as destabilizations, re-hospitalizations, and mortality are presented. According to the authors' 32 knowledge, it is the first time that such a comprehensive review, focusing on all aspects of the management of afailure, is presented. 34

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46 47Contents $\frac{59}{49}$ 511. Introduction . Detection of HF 522. 0 3. HF Subtypes Classification . . . 530 544. Severity Estimation of HF . . 0 Prediction of Adverse Events 555 0 565.1. Destabilizations . . 575.2. **Re-Hospitalizations** 0 5.3. Mortality 0 58596. Summary and Outlook . 0 60 Acknowledgment 0 61 References 0

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1. Introduction

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Heart failure (HF) is a complex clinical syndrome and not a disease. It 64 prevents the heart from fulfilling the circulatory demands of the body, 65

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since it impairs the ability of the ventricle to fill or eject blood. It is 66 67 characterized by symptoms, such as breathlessness, ankle swelling and fatigue that may be accompanied by signs, for example elevated 68 69 jugular venous pressure, pulmonary crackles, and peripheral edema, caused by structural and/or functional cardiac or non-cardiac abnormal-70 71ities. HF is a serious condition associated with high morbidity and mor-72tality rates. According to the European Society of Cardiology (ESC), 26 73million adults globally are diagnosed with HF, while 3.6 million are 74newly diagnosed every year. 17-45% of the patients suffering from HF 75die within the first year and the remaining die within 5 years. The related to HF management costs are approximately 1-2% of all healthcare 76expenditure, with most of them linked with recurrent hospital admis-77 sions [1-3]. 78

The increased prevalence, the escalated healthcare costs, the repeat-79 ed hospitalizations, the reduced quality of life (QoL) and the early 80 81 mortality have transformed HF to an epidemic in Europe and worldwide and highlight the need for early diagnosis (detection of the presence of 82 83 HF and estimation of its severity) and effective treatment. In clinical practice, medical diagnosis, including carefully history and physical 84 examination, is supported by ancillary tests, such as blood tests, chest 85 86 radiography, electrocardiography and echocardiography [4]. The com-87 bination of data produced by the above procedure of diagnosis resulted 88 in the formulation of several criteria (e.g. Framingham, Boston, the Gothenburg and the ESC criteria) determining the presence of HF [5]. 89 Once the diagnosis of HF is established, the experts classify the severity 90 of HF using either the New York Heart Association (NYHA) or the 91American College of Cardiology/American Heart Association (ACC/AHA) 9293 Guidelines classification systems, since this classification allows them to 94determine the most appropriate treatment (medication treatment, guidelines regarding nutrition and physical activity exercising) to be 9596 followed [6].

Although there is a significant progress in understanding the 97 98 complex pathophysiology of HF, the quantity and complexity of data and information to be analyzed and managed convert the accurate 99 and efficient diagnosis of HF and the assessment of therapeutic regi-100 mens to quite challenging and complicated tasks. Those factors, in com-101 102 bination with the positive effects of early diagnosis of HF (which allows experts to design an effective and possibly successful treatment plan, 103 prevents condition worsening, affects positively the patient's health, 104 improves patient's QoL and contributes to decrease of medical costs) 105 are the reasons behind the enormous increase of the application of 106 107 machine learning techniques to analyze, predict and classify medical

data. Classification methods are among the data mining techniques108that have gained the interest of research groups. Accurate classification109of disease stage or etiology or subtypes allows treatments and interven-110tions to be delivered in an efficient and targeted way and permits111assessment of the patient's progress.112

Focusing on HF, different data mining techniques have been 113 employed to differentiate the patients with HF from controls, to recog- 114 nize the different HF subtypes (e.g. HF with reduced ejection fraction, 115 HF with preserved ejection fraction) and to estimate the severity of 116 HF (NYHA class) (Fig. 1). Additionally, data mining techniques can be 117 advantageous even if HF is being diagnosed at a late stage, where the 118 therapeutic benefits of interventions and the prospect of survival are 119 limited, since they allow the timely prediction of mortality, morbidity 120 and risk of readmission. Data recorded in the subjects' health record, 121 expressing demographic information, clinical history information, 122 presenting symptoms, physical examination results, laboratory data, 123 electrocardiogram (ECG) analysis results, are employed. An extended 124 review of the studies reported in the literature addressing the above 125 mentioned issues (HF detection, severity estimation, prediction of 126 adverse events) through the utilization of machine learning techniques 127 is presented in this paper. 128

The systematic literature review was based on sources like 129 i) PubMeD, ii) Scopus, iii) ScienceDirect, iv) Google Scholar, v) Web 130 of Science (WoS) using as keywords the phrases "detection of HF", 131 "severity estimation of HF", "HF subtypes classification", "prediction 132 of HF destabilizations", "prediction of HF relapses", "prediction of HF 133 mortality", "prediction of HF re-hospitalizations". 134

The studies reported in the literature were selected based on the 135 following criteria: i) focus on heart failure and no any other heart 136 disease, ii) are written in English language, iii) are published from 137 2000 (inclusive) until present, iv) cover different geographical loca-138 tions, v) are employing machine learning techniques, vi) employ 139 Electronic Health Records, published databases, observational, trial, 140 *etc.* for the development and validation, vii) provide information 141 regarding the evaluation measures and the validation method that 142 was followed and, viii) the response feature is either differentiation 143 of subjects to normal and HF or differentiation of subjects to different 144 HF subtypes or estimation of the severity of HF or estimation of 145 the destabilization or estimation of re-admission or estimation of 146 mortality. There is no restriction regarding the time frame of the 147 prediction. Furthermore, studies addressing both aspects of HF man- 148 agement (*e.g.* detection and severity estimation of HF) were also 149

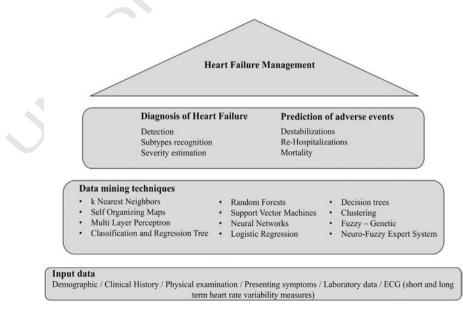


Fig. 1. Overview of studies on heart failure management.

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