

Towards an Evolutionary Open Pediatric Intensive Care Dataset in the ELISE Project

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Abstract Background: To embrace the need for freely accessible training data sets originating from the real world, in the ELISE project, we integrate source data from a pediatric intensive care unit and provide it to researchers. **Objective:** We present our vision, initial results and steps on a trail towards an evolutionary open pediatric intensive care data set. **Methods:** Our evolution plan for the data set comprises three steps. The final data set will include raw clinical data and labels on critical outcomes such as organ dysfunction and sepsis, generated automatically by computerized and well-evaluated methods. **Results:** First step resulted in an initial version data set available in a central repository. **Conclusions:** Our approach has great potential to provide a comprehensive open intensive care data set labeled for critical pediatric outcomes and, thus, contributing to overcome the current lack of real-world pediatric intensive care data usable for training data-driven algorithms.

Keywords. Intensive Care Units, Pediatric; Dataset; Data Science

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1. Introduction

In medical data science, there is a recent shift from knowledge-based approaches towards the application of data-driven methods [1]. Taking the enormous growth of data and the numerous potentials of data-driven learning into account, this progression is advantageous, especially if clinically relevant findings can be derived. Thus, such growing databases could represent an enormous leap in the development towards evidence-based medicine. However, it is not possible for every data scientist to connect to real-world medical centers to access data. Furthermore, ‘*multiple use of routine data*’ in the aforementioned sense is not straightforward since it requires time-consuming assembling and harmonization of heterogeneous and multi-source data [2]. If data are available and integrated, it would be fair to make it accessible for other researchers. The MIMIC database is an outstanding example for such an endeavor as it contains de-identified health-related data associated with over forty thousand patients from a critical care unit in Israel [3]. Another related database is eICU [4], also containing data of adults or neonates, but no data on critically ill children. Thus, in 2020, Zeng et al. published PIC, a pediatric-specific intensive care database comprising children’s data [5].

In our project ELISE (a learning, interoperable and smart expert system for pediatric intensive care), we explore options on computerized, data-driven detection of pediatric systemic inflammatory response syndrome (SIRS), sepsis and organ dysfunction [6]. We strive for setting up processes to continuously assemble and standardize routine data from a patient data management system of the Pediatric Cardiology and Intensive Care Unit of the Hannover Medical School. Afterwards, we develop a knowledge-based approach using international diagnostic criteria as a basis (see Wulff et al. [7, 8]), evaluate its diagnostic accuracy and use this approach as a labeling mechanism for assigning outcomes labels to all patients retrospectively. Based on this training data, data-driven approaches for an early detection of the aforementioned diseases will be developed and integrated into an open demonstrator of a clinical decision-support system. For this task, it is not possible to use the PIC database [5] since it only contains diagnosis codes without exact time mappings of those critical events. Our ELISE dataset contains this highly important time resolution, creating an optimal structure for training data for machine learning. Our vision is to complement available databases by providing a pediatric intensive care data set with outcome labels, generated by well-evaluated computerized models. The data set will develop in an evolutionary way twofold: (1) in terms of scope, since we plan to add new parameters over time, (2) in terms of size, since we set up automatic local processes to continuously add new patient data for already integrated parameters. With this article, we communicate our vision as well as first results and steps on our trail towards an open pediatric intensive care data set.

2. Methods

The evolution of the data set is planned in three major steps (see Figure 1). *First*, a basic data set containing general patient and intensive care data such as vital signs series, laboratory values, device-related data, diagnoses, and procedure codes will be released. *Second*, outcome labels for SIRS and sepsis as well as hepatic, hematologic, renal, cardiovascular, and respiratory organ dysfunctions will be added. *Third*, raw data from electrocardiograms and related devices will be included. Each major step will be developed evolutionary itself in minor steps, e.g., for the basic data set, we first include

only data from one ward, prospectively enhancing with data from other wards. This can provide a more complete picture of the patient’s entire hospital stay. We set up local extraction, load and transform processes (ETL) using Microsoft SQL Server Integration Services to gather data from the primary source systems. ETL processes can be carried out continuously to be able to add new patient data, and data is automatically de-identified. Transformed data are stored in a Microsoft SQL staging database and, later, by using the self-developed HaMSTR tool [9] in a standardized semantically enriched data platform using openEHR [10] or HL7 FHIR [11]. Add-ons will be delivered consecutively, comprising (a) data pre-processing packages, (b) labeling algorithms used, (c) standard-based representation of our data, (d) our open demonstrator app for data-driven detection of the above-mentioned conditions.

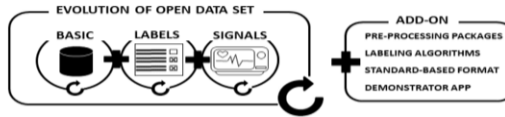


Figure 1. The evolution of the open pediatric intensive care data set and its add-ons.

3. Results

The basic dataset (see Table 1) with a data description is available in csv format in a central data repository, accessible upon request [12]. For the second major release, labels for hematologic and hepatic dysfunctions are already prepared and available with all other labels in the second version of the data set by the end of the year. To expand raw data towards the third release, we currently work on integrating data from the central monitoring system. The aim is to include vector-based data and curves, e.g., for arterial and central venous pressures and the plethysmographic function for pulse oximetry.

Table 1. Overview of the scope of the first basic dataset (seven years of MHH PICU data)

Parameter	Approx. Entries	Parameter	Approx. Entries
Patients	5.500	Medications	4.000.000
Vital signs	90.000.000	Procedures	400.000
Device-related data	20.000.000	Diagnoses	200.000
Laboratory parameter	8.000.000	Movement data	100.000

4. Discussion

We experienced the extraction of primary source data as a time-consuming task requiring enormous initial efforts. Due to heterogeneous data representations, interpretation of data required close collaboration with clinicians, manufacturers, and system administrators. Compared to related studies, such as the MIMIC [3] or PIC [5] studies, our current data set is limited in matters of size and variety. However, we are confident that our evolution plan allows delivering a similar comprehensive data set within the time span of ELISE. Additionally, we strongly concentrate on contributing outcome labels as discussed, which are not yet available in the above-mentioned related initiatives and databases. In contrast to the other studies, we strongly focus on international data standardization and classification (e.g. by using openEHR [10], FHIR [11], SNOMED [13], LOINC [14]) to exculpate data scientists from data preprocessing. This also will facilitate multi-center-

approaches, thus, reaching a new level of quality in data science. In all matters, data security is a risk and needs to be considered by both data providers and data users. Our anonymization approach comprises de-identification. Currently, we work on gathering requirements on new data to be added in future major versions, e.g., from Anesthesia. With this publication, we communicate our approach in an early manner to be able to collect further requirements from interested readers, researchers and data scientists.

5. Conclusions

By developing a comprehensive open intensive care data set, labeled for critical outcomes, our work contributes to overcome the current lack of available pediatric intensive care data reusable to train data-driven algorithms. Further input and requirements from the international community are much appreciated.

Acknowledgments

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