Breakout Session 4: Track A

PREcision Care In Cardiac ArrEst - ICECAP (PRECICECAP)

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Critically III Patients Generate LOTS of Data



Sanchez-Pinto et al. Big Data and Data Science in Critical Care. 2018

Critically III Patients Generate LOTS of Data



Generally available in the electronic health record (EHR)

Sanchez-Pinto et al. Big Data and Data Science in Critical Care. 2018

Current Treatments Lack Precision

- Parent clinical trial: ICECAP aims to find optimal cooling duration for all patients
- Most cardiac arrest trials of effective interventions are neutral
- Little effort to target interventions to likely responders
- What are we trying to predict/improve?
 - Survival (likeliness to be discharged alive)
 - 90-day function/prognosis
- Variables specific to cardiac arrest
 - Patient and arrest characteristics
 - Cardiopulmonary physiology
 - Neurophysiology (EEG)
 - Imaging
 - Response to treatment









Stratified treatments







Available Data

Two types of variables

I. Time-invariant variables from baseline data collection (categorical & quantitative)



Charlson Comorbidity Index (CCI) Aggregating categorical variables in a score used in ER



Distribution of Clinically Relevant Continuous Features

Time-invariant variables are the most critical variables for the initial prediction of survival and optimal hypothermia duration prior to waveform data recording.

Available Data

Two types of variables



These can be used to update the initial prediction on an hourly or six-hour basis.

Available Data

• Time-varying variables heterogeneous availability across participants:







And artifacts





Data Acquisition pipeline



Artifact Annotation Pipeline

Moberg visualization platform





Algorithmic guided and manual annotation







Artifact Annotation Pipeline

Example of artifacts on the Arterial Blood Pressure signal





Expected Arterial Blood Pressure signal





Long Short-Term Memory - RNN model

Predictions: Predict patterns in the next time intervals (used for imputation and prognosis). Classification: Classify and predict the participants group.





Long Short-Term Memory - RNN model

Example of imputation on Arterial Blood Pressure mean, subsampled on 5-min windows



Summary to make variables AI/ML ready

- 100 participants enrolled and collected as part of PREICECAP
 - target 300+ participants

In progress:

- i. Harmonization and quality check of baseline variables across sites
- ii. Waveform data artifacts annotation and imputation
- iii. Waveform data featurization
- iv. Strategies to co-register raw waveforms across participants

TO DO:

Harmonization of medications/treatments across sites



Reflecting on potential predictions

- 100 participants enrolled and collected as part of PREICECAP
 - target 300+ participants

Note that body cooling duration and outcomes (survival and 90-day prognosis) are currently blinded by the parent clinical trial (ICECAP).

At baseline, using time-invariant variables, predict:

Participants most likely to survive, optimal cooling duration, and tailored treatment strategy.

During treatment, using time-varying variables, predict:

- > Whether continuing body cooling is beneficial or not;
- > Whether additional treatments could improve outcomes.



Thank you! Karen Hirsch – <u>khirsch@stanford.edu</u> Jonathan Elmer – <u>elmerjp@upmc.edu</u> Yann Le Guen – <u>yleguen@Stanford.edu</u>

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