

Introduction

Background: Time-series biomedical feedback data is data gathered from medical sensors that have a time dimension. E.g. EEG (electroencephalogram), EKG (electrocardiogram), Accelerometer, etc...

Motivation: Make it easier for biomedical researchers to visualize and create deep learning models from timeseries bio-medical data.

Objective: Create an application for visualizing, labeling, and creating models of biomedical data using deep learning.



Self-sense: Semi-supervised Labeling of Sensor-Generated Time-Series Data

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Add Label

Sitting Standing

Functionality

- 4. Train Model on Labeled Data & Save Model:
- Use specially created Numpy Array class to keep track of subject and time segment for each sample
- Train model on labeled data
- Name the model and save it

5. Load Model:

Select which model to load and which data to label



- Models: 🧿 model.h5
 - tristan.h5
 - tyler.h5
 - O vangelis.h5

6. Upload & Label New Data:

- Use trained model to automatically label the new unlabeled data
- Use time from Numpy Array class to label each corresponding time segment on dataframe

7. Visualize Labels & Flag Labels That Need Review:

Show the predicted activity labels in a scatter plot and highlight areas with yellow that need review





Conclusion

Results:

- Created a class for performing preprocessing on Empatica files
- Created application interface for project pipeline
- Used Jupyter Notebooks that are deployable through Voila (can be used to run, convert, and serve a Jupyter notebook as a standalone web app)
- First interface includes functionality for visualizing and labeling data, as well as creating and saving models
- Second interface includes functionality for uploading a model and using it to label new time series data

Future Work:

- Expand preprocessing library to more devices than Empatica (e.g. Bioradio, I-Phone Accelerometer)
- Expand Model library to include a more diverse
- Selection (e.g. LSTM, Transformers, RNN)
- Expand visualization suite (e.g. UMAP)
- Port project over to a desktop platform



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