

## Preliminary Documentation

### EARIN Project – Predicting ECG Diagnosis

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Dataset: CPSC-2018 ECG Dataset

## 1. Dataset Description

We are using the CPSC-2018 ECG dataset, which contains multilead ECG recordings collected for classification purposes. Each recording may contain one or more diagnosis labels related to cardiac abnormalities.

Key Facts:

- Signals are sampled at 500Hz
- Variable-length recordings (6s to 144s)
- 12-lead ECG signals
- 9 diagnosis classes (e.g., AF, ST, PVC, etc.)

Initial Observations:

- Some recordings are noisy or have missing leads
- The dataset is imbalanced (e.g., many more Normal cases than rare pathologies)

Visualization:

We plotted several sample ECG signals using the WFDB and matplotlib libraries. A clear difference can be seen between normal and abnormal signals.

## 2. Problem Solving Plan

### a. Data Splitting Strategy

We plan to split the dataset as follows:

70% training, 15% validation, 15% testing

We will use stratified sampling to preserve class distribution across subsets.

### b. Algorithms to Be Used

We will compare the following approaches:

1. CNN on spectrogram images (Mel-Spectrogram + 2D CNN)
2. LSTM on raw signals (time series classification)
3. Random Forest on extracted features (peak rate, RR intervals, etc.)

### c. Libraries and Tools

We will use the following tools:

- wfdb for reading ECG data
- pywavelets for denoising
- scipy.signal and librosa for spectral feature extraction
- scikit-learn, keras, and PyTorch for model training and evaluation

#### d. Evaluation Methods

We will evaluate each model using the following metrics:

Accuracy, F1 Score, ROC Curve, Sensitivity, Specificity, and Confusion Matrix

We aim to test at least 5–6 hyperparameter configurations per model.

### 3. Planned Experiments

- Spectrogram conversion versus raw signal input
- Comparison between deep learning and classical machine learning
- Hyperparameter tuning (learning rate, layers, window size, etc.)
- Impact of denoising using wavelets

### 4. Result Visualization

We plan to include:

- Sample spectrograms
- Metric comparison tables
- ROC curve plots
- Confusion matrices

### 5. Conclusion

The preliminary setup is ready. Our dataset is loaded and partially visualized. Our next step will be implementing the pipeline for signal preprocessing and classification using the outlined methods.

### 6. References

1. WFDB Library – <https://www.physionet.org/content/wfdb/>
2. PyWavelets – <https://pywavelets.readthedocs.io>
3. SciPy Spectrogram – <https://docs.scipy.org/doc/scipy/reference/generated/scipy.signal.spectrogram.html>
4. Librosa Mel-spectrogram – <https://librosa.org/doc/main/generated/librosa.feature.melspectrogram.html>
5. Scikit-learn – <https://scikit-learn.org>
6. Papers-with-code-<https://paperswithcode.com/dataset/the-china-physiological-signal-challenge-2018>
7. ICBEBC(China)-<http://2018.icbeb.org/Challenge.html>

