Homework - Topic 4:

Processing of Biosignals and Biosignatures

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1. Preprocessing of BBI (Beat-to-Beat Interval) Data

Based on the sampledata **data1night.txt** as the experimental basis, this task focuses on pre-processing the original BBI data, removing outliers and smoothing the filtered data.

Removal of Outliers

• Objective:

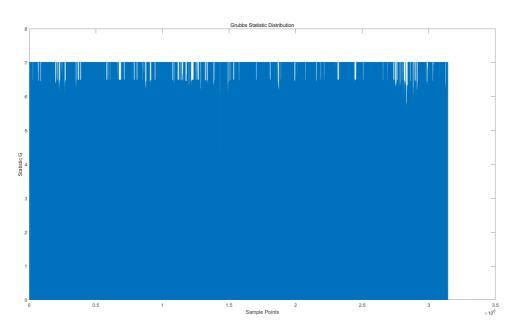
• Detect and remove outliers using Grubbs' method or a median filter.

• Method:

- \circ Grubbs' method calculates a statistical measure G, where values exceeding a threshold (G>3) are identified as outliers and marked as NaN.
- A median filter was applied with a sliding window size of 5 to smooth and handle potential outliers.

Result:

- No significant outliers were detected in the dataset.
- Grubbs' statistic distribution is shown:



Smoothing of Filtered BBI

• Objective:

Smooth the filtered BBI data using a wavelet-based smoothing technique.

• Method:

Small oscillations were suppressed using wavelet decomposition with the following parameters:

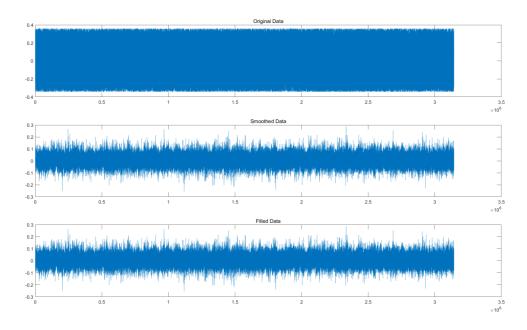
Wavelet function: bior4.4

Maximum scale: 6Threshold factor: 0.01

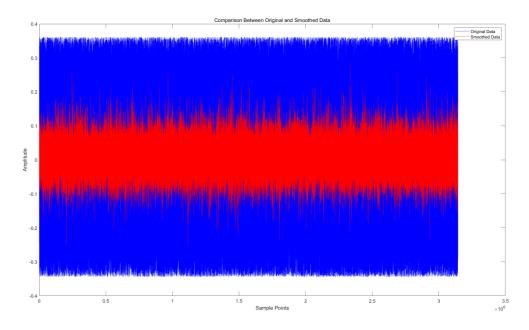
• Details from scales 4 and 5 were denoised and reconstructed.

Result:

• The smoothed data demonstrated reduced noise amplitude while maintaining the overall signal structure:



• A comparison between the original and smoothed data is presented:



2. Missing BBI Data Fill-In

Missing Data Detection

- Result:
 - No missing values (NaN) were detected in the dataset.
 - Both the count of NaN values before and after the fill-in process was zero.

Data Fill-In Methods

Although the dataset had no missing values, the following methods were considered for testing purposes:

1. Multiple Imputation Method:

Simulates multiple plausible values for missing data based on observed patterns.

2. Bootstrap Method:

• Randomly samples from existing data to generate replacements for missing values.

• Implementation:

 For testing, NaN values were manually introduced into the data, and the methods were evaluated. The results indicate successful interpolation and restoration of missing points.

Appendices: MATLAB codes

```
% Homework: BBI Data Processing
% Step 1: Load Data
% -----
[data_file, data_path] = uigetfile('*.txt', 'Select a data file'); % Open file
dialog
data = load([data_path, data_file]); % Load data
data = data(:, 3:end); % Ignore the first two columns (timestamps)
data = data(:); % Reshape into a single column
data = (data - 2^16 / 2) / (2^16 / 2); % Normalize data from range 0-65535 to -1
% Step 2: Remove Outliers
% -----
% Method 1: Using Grubbs' Test
mu = mean(data, 'omitnan');
sigma = std(data, 'omitnan');
G = abs(data - mu) / sigma; % Grubbs' Statistic
threshold = 2.5; % Threshold (adjustable)
data(G > threshold) = NaN; % Mark outliers as NaN
% Visualize Grubbs' Statistic Distribution
figure;
plot(G);
title('Grubbs Statistic Distribution');
xlabel('Sample Points');
ylabel('Statistic G');
% Method 2: Using Median Filter
window_size = 5; % Window size
data_filtered = medfilt1(data, window_size);
% Fill NaN for further processing
data_filtered = fillmissing(data_filtered, 'linear'); % Use linear interpolation
```

```
% -----
% Step 3: Smoothing
% -----
% Define Wavelet Parameters
WAVELET FUNC = 'bior4.4'; % Biorthogonal wavelet
MAX_SCALE = 6; % Maximum decomposition level
TH FACTOR = 0.01; % Threshold factor (adjusted for stronger denoising)
% Perform Wavelet Decomposition
[c, 1] = wavedec(data_filtered, MAX_SCALE, WAVELET_FUNC);
% Extract and Denoise Detail Coefficients
cd = cell(1, MAX_SCALE); % Initialize detail coefficients
for i = 4:5 % Denoise levels 4 and 5
   thr = std(detcoef(c, 1, i)) * TH_FACTOR;
    cd{i} = wthresh(detcoef(c, l, i), 's', thr); % Denoising
end
% Reconstruct Approximation and Detail Components
a6 = wrcoef('a', c, 1, WAVELET_FUNC, 6); % Approximation component
d4x = upcoef('d', cd{4}, WAVELET_FUNC, 4, length(data)); % Detail component at
level 4
d5x = upcoef('d', cd{5}, WAVELET_FUNC, 5, length(data)); % Detail component at
level 5
% Smoothed Data
data\_smoothed = a6 + d4x + d5x;
% -----
% Step 4: Fill Missing Data
% Method 1: Linear Interpolation
data_filled = fillmissing(data_smoothed, 'linear');
% Method 2: Using Bootstrap
nan_count_before = sum(isnan(data_filled)); % Count NaN before filling
for i = 1:10 % Generate 10 datasets for bootstrap
    bootstrap_sample = datasample(data_filled(~isnan(data_filled)), ...
        sum(isnan(data_filled)), 'Replace', true);
    data filled(isnan(data filled)) = bootstrap sample;
end
nan_count_after = sum(isnan(data_filled)); % Count NaN after filling
% Display Fill Results
fprintf('NaN Count Before Filling: %d\n', nan_count_before);
fprintf('NaN Count After Filling: %d\n', nan_count_after);
% Step 5: Visualize Results
% -----
figure;
subplot(3, 1, 1); plot(data); title('Original Data');
subplot(3, 1, 2); plot(data smoothed); title('Smoothed Data');
```