DIGITAL SIGNAL PROCESSING PROJECT

WAVELET-BASED DENOISING OF ECG SIGNAL

PROJECT MENTOR

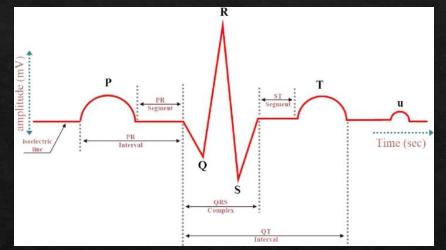
Utkarsh

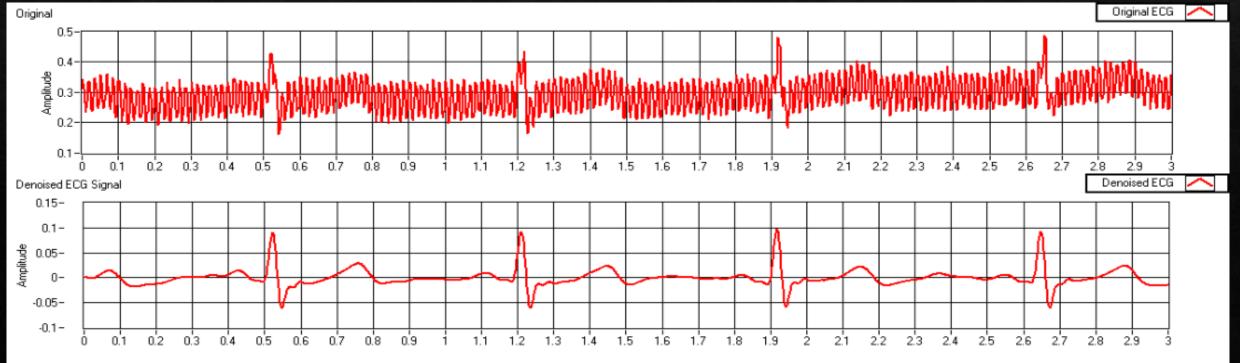
PROJECT TEAM

Sai Krishna Charan Dara Sasi Kiran Dharmala

OBJECTIVE OF PROJECT

- ☐ Mainly ECG(Electrocardiogram) signals consists of PR segment, QRS complex, ST segment.
- ☐ Normally, raw ECG signals consists various types of noises.
- ☐ Our main aim of project is to remove those noises and detect QRS complex of ECG signal.





PROJECT OVERVIEW

- ☐ Design and implementation of an automatic ECG beat detection system.
- □ This project is a modified version of Pan-Tompkins algorithm which detects QRS complexes in ECG signals.
- □ During recording process of ECG signals noise effect the signal heavily which include baseline wandering ,EMG noise ,motion artifact, power line interference and electrode pop or contact noise.
- ☐ The denoising part is done using wavelet-based tools.
- □ This method performs much better in case of non-stationary ECG signal and would result in SNR gain when using wavelet denoising method.
- □ Probability of detection of fiducial points which are QRS complexes is more here compared to Pan-Tompkins algorithm.
- □ In Fourier Transform we usual get a range of values for the instantaneous frequency at a given point of time. This can be overcome by approximating the original signal to wavelets. This increases the resolution of the signal.

RAW ECG SIGNAL

Wavelet denoising using WA
Detrend Virtual Instrument(VI)
using sym5 wavelet

WORKING

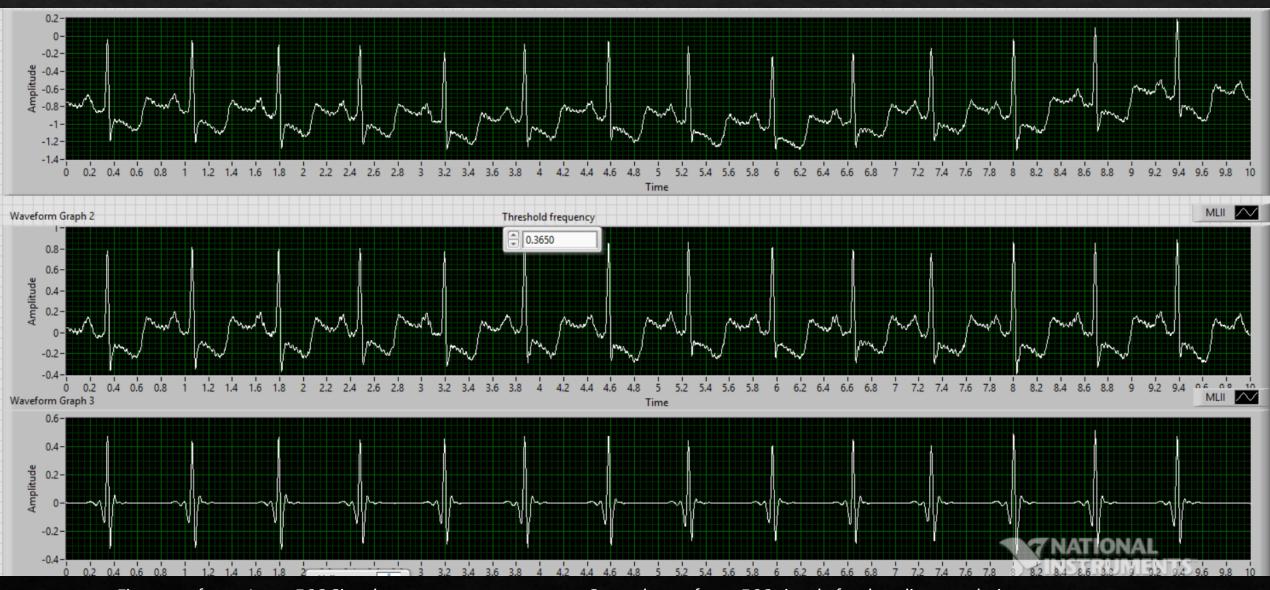
ECG SIGNAL WITH
REMOVED LOW
FREQUENCY BASELINE
WANDERING

Wavelet denoising using undecimated wavelet transform(UWT) sym5 with single level and soft thresholding

SIGNAL GETS SMOOTHENED AND WIDEBAND NOISE REMOVED

Adaptive Thresholding PEAK = Maximum(Peaks)NPK = Minimum(Peaks) $SPK = 0.125 \times PEAK + 0.875 \times SPK$ THR = NPK + 0.25(SPK - NPK)Yes No Signa1 Peak detected Search Back Signal Peak larger Algorithm than THR considered as QRS complex THR=THR/8

The VI first decomposes the ECG signal into several sub-bands by applying wavelet transform and then modifies each wavelet coefficient by applying a threshold or shrinkage function and finally reconstructs the denoised signal. $C(scale, coefficients) = \int_{-\infty}^{\infty} f(t) \varphi(scale, position, t) dt$



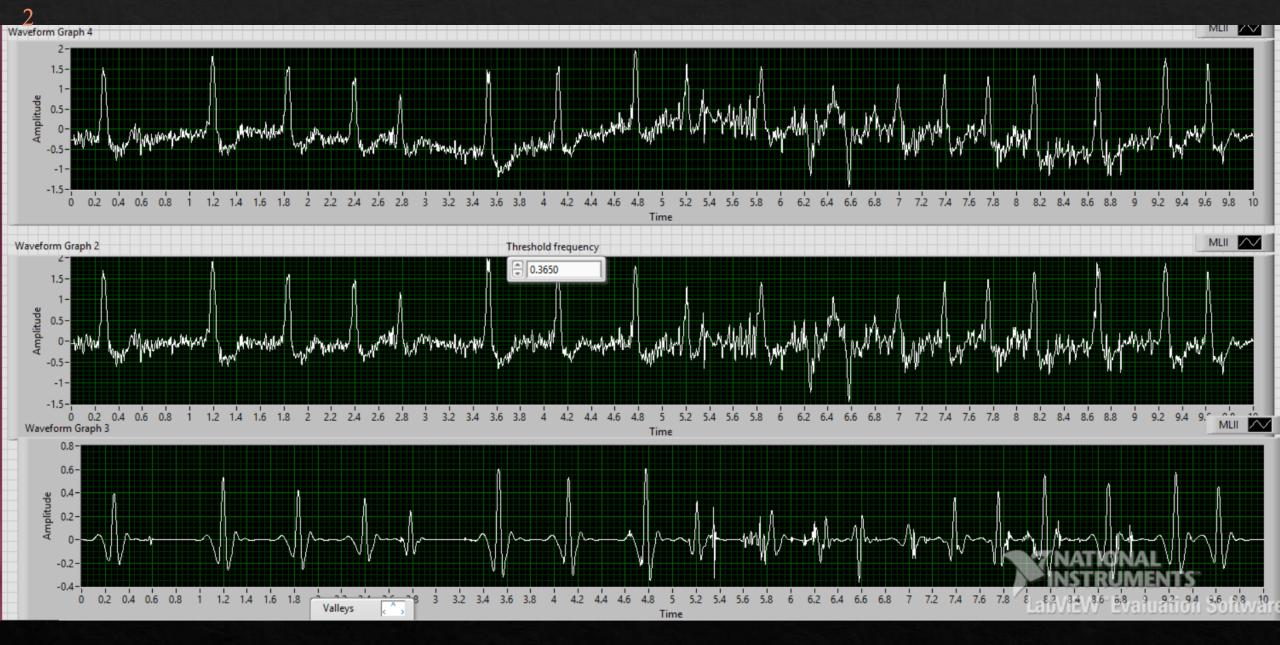
First waveform: Input ECG Signal Second waveform: ECG signal after baseline wandering
Third waveform: ECG signal after smoothening and removal of wide band noise

Threshold Frequency=

Threshold Frequency= $log_2(2*10)/log_2(3600)$



Peaks /Valleys plot shows us the detection of peaks and valleys from the denoised ECG signal and number of peaks and valleys depends on peak/valley width.

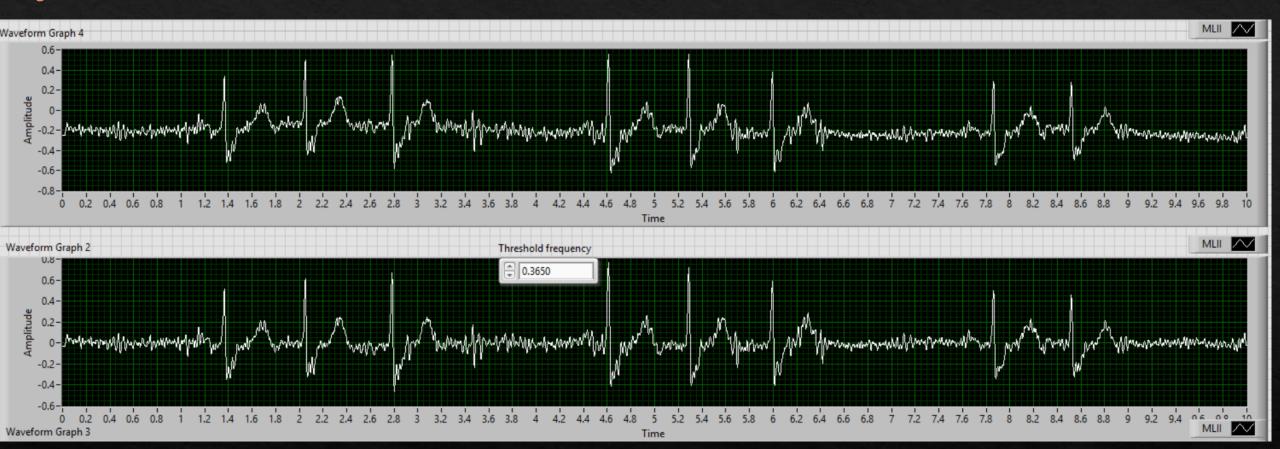


First waveform: Input ECG Signal
Third waveform: ECG signal after smoothening and removal of wide band noise

Second waveform: ECG signal after baseline wandering Threshold Frequency= $log_2(2*10)/log_2(3600)$



Peaks /Valleys plot shows us the detection of peaks and valleys from the denoised ECG signal and number of peaks and valleys depends on peak/valley width.



First waveform: Input ECG Signal

Second waveform: ECG signal after baseline wandering

Threshold Frequency= $\log_2(2*10)/\log_2(3600)$



First Waveform: After Wide band noise removal from ECG signal.

Peaks /Valleys plot shows us the detection of peaks and valleys from the denoised ECG signal and number of peaks and valleys depends on peak/valley width