# A New Method for Acquisition and Analysis of ECG Signal using Virtual Environment

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#### Abstract

Electrocardiogram is used to measure the rate and regularity of heartbeats and to detect any heart arrhythmia. Different ways are submitted and used for cardiogram feature extraction with a reasonable percentage of right detection. Although the problem stays open especially with respect to superior detection accuracy in ECGs. The ECG signal is very sensitive in nature having voltage level as low as 0.5 to 5mv and frequency components fall into the range of 0.05-100Hz and most of the information contained in the range of 0.05- 45Hz. The recorded ECG signal contains different type of noises such as baseline wander, channel noise which becomes very essential for us to remove for the better clinical result which helps in the treatment of the patient. For the

andextraction classification we'll be using discrete wavelet transform (DWT) as wavelet transform could be a two dimensional process technique, fore it's appropriate for the nonstationary ECG signals(due to adequate scale values and shifting in time) in LabVIEW. The flexibility, standard nature and simplicity to use programming possible with LabVIEW, makes it less complex. The proposed algorithm is executed in two steps. First step, it pre-processes de-noises the signal to get rid of the noise from the cardiogram signal, Then it detects pulse, Our extracted parameters are Heart rate, P wave amplitude, T wave amplitude, S value, Q value, R-value, P offset location, P onset location, T onset location, T offset location and the location of P, Q, R, S and T wave.

**Keywords** — Electrocardiogram, discrete wavelet transform, heart arrhythmia, LabVIEW.

# I. INTRODUCTION

An ECG (electrocardiogram) represents cardiac signals generated by cardiac muscles. A typical ECG cycle contains wave segments P, QRS and T which represents periodic depolarization and repolarization of atria and ventricles in a sequential manner. QRS, being the most striking segment of the waveform assumes special significance for the cardiac interpretation of ECG signal. With the semiconductor technology advancement, embedded systems are adopted to implement an ambulatory ECG monitor as a primary signal-processing device for detecting irregular

heart conditions by evaluating ECG signals [2]. The ECG detection that shows the data of the heart and heart condition is important to port the patient living quality and applicable treatment. It is valuable and a very important tool within the identification and the condition of the heart diseases. In recent year, numerous research and algorithm have been developed for the work of analyzing and classifying the ECG signal. The ECG features can be extracted in a time domain or in a frequency domain. Manual beat-by-beat measurements of all characteristic points in each lead are impractical in routine clinical observe. Especially for long term ECGs. For this reason, automatic ECG feature extraction methods are more relevance. Beat QRS advanced detection is the most significant part which is Associate in Nursing ECG feature system. Therefore peak detection extraction Wavelet Transforms Algorithms are needed. will present a time versus frequency illustration of the signal and work well on the non-stationary signal. Wavelets also overcome the present resolution problem of the short time Fourier transforms by using a variable length window. The large range of various wavelet functions provides an exclusive area to look for wavelet with efficiency represent a symbol of interest. Although there are some methods available in order to select the best wavelet for an application. The orthogonal Daubechies wavelet family, specifically Db6 is used here. We have enforced here the DWT to extract ECG signal features. In ECG signals, instrumentation plays a major role, since signals generated by the human body are very low in amplitude. High gain must be obtained with a high common-mode rejection ratio (CMRR). Two electrical circuits were studied in the present work, using common electronic parts and application-specific parts. LABVIEW is a software application from National Instruments that is specially designed for easy and powerful data acquisition purpose. Thus, LABVIEW software was used for data recording and visualization, due to its known capabilities [1]. Finally, LABVIEW were used again to implement real-time filtering of the signal.

# II. METHODOLOGY

The ECG signal is obtained using three lead system i.e., Einthoven Triangle. The solution for the abnormality detection problem contain 4

vital parts First phase is that the acquiring of the signal, here we have a tendency to use the ECG signal within the system and provides to our program for the analysis. The second phase is the filtering of the raw ECG signal to remove unwanted noises. A third is the core phase to extracting the features from the signal i.e. ECG signal in terms of its parameters by the actual analysis of it. The last is the detection of different types of abnormalities on the basis of different values of parameters obtained [3]. The parameters are obtained in LabVIEW software where its features are known.

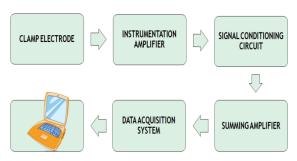


Fig. 1 Flowchart of the proposed system

## A. Signal Amplification

The combinations of differential amplifiers are used to construct to achieve the obtained output from the noised input signal which is collected from bio-potentials that is called an instrumentation amplifier. In this proposed project we have used INA126 instrumentation amplifier. An instrumentation amplifier is typically the very first point in an instrumentation system. This is as for the very small voltages generally received from the probes have to be amplified. The ECG signal is too small and contains a lot of added noise. Also, the signal extracted from the heart has an amplitude of approximately 0.5mV. Since it is essential to amplify the signal and remove the noise, and then extract the ORS complex. The INA126 is used for the low-level amplification. The gain of the amplifier is 13v/v.

# B. Signal Conditioning

The Filtering circuit includes Bandpass filter with the non-inverting amplifier. In that, the frequency range of the Bandpass filter is 0.5-30 Hz and the Gain of the non-inverting amplifier is 100v/v. Thus, after the filtering stage noise will be removed and get the better signal of ECG. After that with the use of a summing amplifier, the signal has been summed in the range of 0-5v. There are many types of noises which are removed using this filtering technique such as power line interference, Electrode contact noise, muscle artifact, motion artifact and baseline wander.

# C. Summing Amplifier

The signal must be in a range of 0 to 5 volts so that the signal will be converted from analog to digital form to perform analysis on it.

$$Vout = \left(1 + \frac{Rf}{R1}\right) \left(\frac{V1 + V2}{2}\right)$$

#### D. Data Acquisition

Analog to digital conversion is one of the most important things in the data acquisition system. It is basically performed using ADC0804. Which is generally done in two ways: First, parallel mode where data is transferred at a faster rate with more number of lines which is used for short-range data transfer and Secondly in Serial mode where it uses one or two data lines which are used for longer range data transfer.

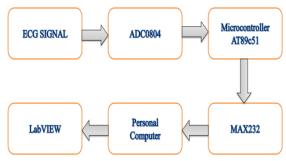


Fig. 2 flowchart of DAQ

# E. LabVIEW (Laptop)

A Virtual ECG Instrumentation system is a new Instrumentation system, which is much better and flexible with the aid of computer intelligent resource.

# 1. Continuous Wavelet Transform

A Continuous wavelet transform is used to divide a continuous time function into wavelets. It is a convolution of the input data sequence with a set of functions generated by the mother wavelet. It is represented as:

$$W_{s}(a, b) = \frac{1}{\sqrt{a}} \int_{-\infty}^{\infty} f(t) \bar{h} \left(\frac{t-b}{a}\right) dt$$

# 2. Discrete wavelet transform

A discrete wavelet transform is used for functional analysis and numerical analysis. The DWT of signal x is calculated by using a series of filters. The sample is passed through a low pass filter with the impulse response g resulting in a convolution of both parameters.

$$y[n] = (x * g)[n] = \sum_{k=-\infty}^{\infty} x[k]g[n-k]$$

#### 3. Feature Extraction

There are so many features can be extracted from the ECG signal and all are specific to heart function and conditions. Before feature extraction, we have to remove baseline wander and filter the noise present in the raw ECG signal. For this used VI is as follow:

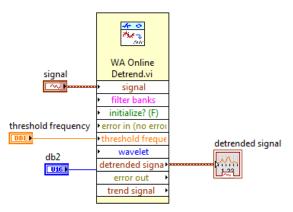


Fig. 3 Wavelet Detrend VI

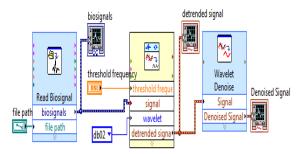


Fig. 4: VI diagram to remove Baseline Wandering and Wideband Noise

# III. RESULT AND CONCLUSION

This evaluation system was tested and validated on a few healthy subjects. Our extracted parameters are Heart rate, P wave amplitude, T wave amplitude, S value, Q value, R-value. The display window of our VI also gives the location of P, Q, R, S and T wave. It also gives output value of P offset location, P onset location, T onset location and T offset location. Heart measurement is done by detecting the R-pecks which is detected by providing threshold value to the ECG signal. If the signal is exceeded from the threshold value it will be counted as an R peak. We have put 0.8 mV threshold values according to the ideal peak value of R. In our analysis first, Wavelet detrend VI is used which removes baseline wander of ECG signal. So, the output of this VI will remove the trend of ECG signal. The Output of Wavelet detrend VI applies to the Wavelet denoise. Transform type of Wavelet denoise is UWT (Undecimated Wavelet Transform) with db02 and level 5, which denoise the signal and wideband type noise. The output shows the smoother ECG signal and it displays sharp peaks. After filtering of ECG signal, Each and every individual ECG superimposed on single ECG cycle and after overlapping of ECG signal we can get averaged ECG signal.

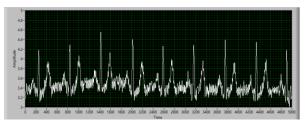


Fig. 5(a): Row ECG signal

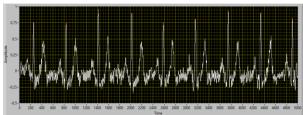


Fig. 5(b): R-peak detection

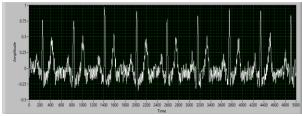


Fig. 5(c): Detrended ECG signal

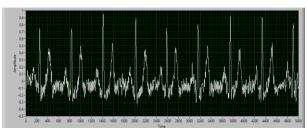


Fig. 5(d): denoised ECG signal

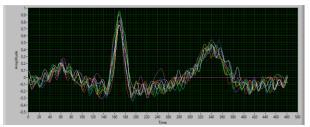


Fig. 5(e): Overlapping of ECG waveforms

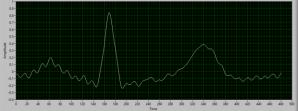


Fig. 5(f): ECG waveform after averaging

Table 1: ECG Analysis and feature extraction components

Sr No.	P Wave ampli- tude	Q valu e	R value	S valu e	T Wave ampli- tude	P Lo- ca- tion	Q Lo- catio n	R Lo- ca- tion	S Lo- catio n	T Lo- catio n	P onset loca- tion	P offset loca- tion	T onset loca- tion	T offset loca- tion	HR
1	0.206	0.19 2	1.220	0.23 2	0.267	85	146	166	221	361	62	105	310	393	95
2	0.198	0.22	0.844	0.23 6	0.387	63	145	169	193	340	40	83	291	378	96
3	0.076	0.12 2	0.568	0.13 0	0.146	35	56	96	136	251	31	39	245	257	56
4	0.113	0.15 9	0.892	0.32 6	0.339	72	112	149	171	303	53	82	265	350	85
5	0.278	0.18 7	1.345	0.26 8	0.244	37	93	143	184	322	13	57	292	358	82
6	0.010	0.18	1.193	0.28 0	0.560	29	68	108	149	252	29	29	214	292	61
7	0.047	0.32 4	1.340	- 0.31 9	0.714	43	112	136	181	296	32	51	248	337	78
8	0.109	- 0.19 1	0.850	0.18 7	0.399	18	70	123	191	295	8	23	242	328	70
9	0.077	- 0.16 1	1.147	- 0.55 3	0.284	42	96	122	140	290	27	52	242	322	70
10	0.289	0.25 1	1.173	- 0.44 2	0.186	58	116	141	189	401	37	66	394	403	82

Above table shows the features extraction of acquired ECG signal from 10 healthy people.

Analysis of heart rate variability of various subjects is shown in below figure 6.

#### 120 100 80 Heart Rate (bpm 60 40 20 0 Subject 1 Subject 2 Subject 5 Subject 6 Subject 8 Subject 3 Subject 4 Subject 7 Trial 1 106 ■ Trial 2 74 86 104 93 93 ■ Trial 3 93 81 108 101 98 97 101 91

### **Heart Rate Variation**

104 Fig. 6: Heart rate variability (HRV chart)

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### IV. CONCLUSION

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■ Trial 4

Trial 5

In the proposed work, After the complete analysis of all ECG signal, we were able to determine and calculate the parameters with high precision and use these parameters to detect and confirm heart abnormalities. We additionally checked for baseline wandering because of motion artifact, so that we do not come out with wrong measurements due to movement of electrodes. Hence the analysis could be a very economical technique and much quicker than the present technology. This is less expensive, very less time consuming and can be done without any expert. Therefore we are able to term this methodology as a life-saving system. We have with efficiency calculated Various ECG signal parameters however they a lot might be used effectively out more heart abnormalities with higher accuracies exploitation a lot of advanced case structures. We can conjointly study and analyze the ECG patterns of various diseases using parameters found higher than can help in deciding higher algorithms for heart diseases. We have targeted solely on software System design however the work may be extended portable hardware design that a user continually and instantly sees the will wear heart condition.

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