A Review: Biometric Signal Information and Techniques Available for Feature Extraction

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Abstract— Heart Rate signal analysis is a trait that is widely used in medical science to monitor the heart activities of the person. To monitor the activities, the heart beat or signals are captured by using a electronic device named as Electrocardiogram i.e. ECG. This device captures the heart signals by using the electrodes that are placed on the limbs of the person. The captured signals are categorized into different waves known as P, QRS, T waves. Then these wave signals are used to extract the features and to perform classification on them in order to reach to a specific decision regarding the disease. The feature extraction and classification is the major and tedious task to perform. Therefore a large number of techniques have been developed to perform it efficiently.

The objective of this work is to represents a review to the ECG signals along with its wave type. This work also focuses to the past research work that had been done to develop the feature extraction and classification techniques for processing the ECG signals.

Keywords— *Heart Rate, Heart Waves, PQRST waves, Feature Extraction, Classification.*

I. INTRODUCTION

Electrocardiogram (ECG) is an electronic device that is used in biomedical field in order to monitor the different activities of human heart. By using ECG, huge data can be collected related to pathological physiology of heart. Although, the Electrocardiograph signals are varying in nature and this is not an easy task to directly examine these [1]. Hence, it is required to implement computer based techniques to analyze the electrocardiogram signal. To scrutinize the ECG signals is a time consuming process as the signals varies constantly. The disparity in signals did not occur at a fixed interval. The analysis of visual based ECG signals are not considers as the reliable one [2]. Thus to overcome the inefficiency of visual based ECG signals, the computer base ECG techniques has been developed to explore the ECG signals. When the heart beats, it starts depolarizing to initiate contraction process. The contraction and relaxation of heart produce the electrical signal that has been forwarded to the whole body and can be measured at skin [3]. This is the major theory, on which the working criteria of ECG machine are based upon. In ECG machine, the electrodes are deployed to record the rhythmic activities of the heart.

The electrodes are attached with the skin and show the recorded data on graph. To monitor the heart activity through ECG, ten types of electrical cables are attached with the human body. Out of these ten cables, 6 cables are attached to the chest and rest of the 4 is attached to the each limb. The signals recorded by the ECG delineates that there are 4 types of activities that are performed by the heart i.e. atria depolarization, atria re-polarization, ventricular depolarization and ventricular re-polarization. In ECG signal, it was observed that there is repetition of complex pattern of waveform and that have frequency of near about 1Hz [4]. The one cycle of ECG waveform contain PQRST waves. This technique is utilize to determine the pace and periodicity of heartbeats so that if there is any kind of irregularity in the heartbeat can be easily detected. The electrocardiogram may be described as a biomedical device that plots the electrical activity of heart on the screen graph. The rise and falls in the graph are represented as P-QRS-T wave. This wave has been depicted in figure number 1. Heart can be described as pumping system made up of muscles and it consist of four types of chambers [5]. The upper 2 chambers of heart are known as atria whereas lower 2 chambers are known as ventricles. Due to rhythmic contraction and relaxation of heart muscles the blood enters into the heart then move toward lungs for purification and then pumped out to whole body [6]. In the PORST wave in ECG graph, the P wave represents the electrical activity in upper heart chamber that is atria and the QRS complex defines the electrical activities performed by the heart in lower heart chamber. Then after this there is one ST segment in the waveform which represents the contraction of ventricles but no electrical current flow while this operation executed. ST part in the waveform generally forms the straight line in between the T wave and QRs wave. And in this case the T wave represents the time period, when lower chambers of heart that are ventricular are at relaxed position and preparing for upcoming contraction of muscles. On the basis of various waves that has explained above in ECG graph it is possible to diagnose cardiovascular disease with the help of ECG machine. When the heart beat is normal then the PQRST wave form made the usual pattern but in case if there is any variation in PQRST wave then it means heart operation is not common. On the basis of the rhythmic contraction and relaxation in heart muscles creates the heartbeat [7]. The distance among different waves helps in determining the rhythm of heart and on the other hand the amplitude of pulses depicts the strength of pumping. By comparing the whole ECG waveform of patient with the standard ECG waveform of healthy person, the disease can be diagnosed. Various improvements in the technological field like had resulted in enhanced version of ECG devices. Technological advancement has revolutionized the whole world in terms of early diagnosis, cure, etc. Due to introduction of wireless transmission system, the medical healthcare system has also improved and provides large number of benefits to the society [8].

The standard waveform of ECG containing the following waves: first one is a P wave, second one is QRS complex, and third one is T wave. Because of production of electrical signal generated due to depolarization of upper chamber of heart that is atria and it should be generated before the contraction of atria muscles, then the P wave obtained. After this the QRS complex wave is produced due to the electrical signals generated during the depolarization process in lower chamber of heart and it should be before the contraction of their muscles and this takes place while depolarization in the muscles of ventricular. In the QRS complex waveform, there are further three types of waves that are as follow: Q wave, R wave, and S waves [9]. It can be clearly observe that in the PORST waveform that QRS complex comprises three types of waveform. When the electrical signal produced while the ventricles are in recovery state from the present state of depolarization, then the T wave is obtained. This operation occur in two ventricle chambers of heart and for duration of 0.25sec to 0.35secand should takes place when the depolarization is over.

Table 1. Descript	ion and Duration of each wave in ECG w	vaveform [24]

Feature	Description	Duration
RR interval	It is the time interval between consecutive peaks of R wave and it is inverse of R wave. Generally the heart beat of healthy person lies in between 50 -100 beats per minute.	0.6 to 1.2 sec
P wave	Whenever there is depolarization in atrial then the electrical impulse originated from the sino- atrial node and start propagation towards the Atrioventricle node and after that moves right atrium to left atrium. This yields the P wave.	80 ms
PR interval	This intervals starts from the P wave and terminates at QRS complex. This interval represents the time period taken by the electrical signal to moves from sinus node to Atrioventricle node and finally moves inside the ventricles.	120 to 200 ms
PR segment	This segment is used to create a link among P wave	50 -120 ms

	and QRS complex. The	
	electrical signal generated at	
	Atrioventricle node and	
	start moving towards the	
	bundle of His then bundle	
	of branches and finally to	
	Purkinje fibers. While	
	executing this process no	
	contraction takes place and	
	just signals travels down	
	towards the ventricles and	
	therefore flat	
	Electrocardiogram signal	
	obtained at this interval.	
	This interval is highly	
	clinically relevant.	
	This represents the rapid	
	depolarization of both	
	ventricles left as well as	
QRS complex	right ventricle. Generally	80-120 ms
- 1	the amplitude of QRS	
	complex is much greater	
	than the P wave.	
	This is the point at which	
	ORS terminates and ST	
	segment initiates. It is	
J point	implemented to analyze the	N/A
	1 2	
	degree of elevation and	
	depression.	
	This segment is used to link	
	the QRS complex with the	
	T wave. This segment	
ST segment	depicts the time interval	80 – 100 ms
51 segment	when the depolarization of	30 - 100 IIIs
	ventricles takes place. The	
	ST segment is iso-electric in	
	nature.	
	This wave shows the re-	
	polarization of lower	
	chambers of heat that are	
	ventricles. The absolute	
	refractory period is the time	
T wave	interval when QRS complex	160 ms
	starts to summit the T wave.	
	In T wave the last segment	
	is called as relative	
	refractory time period.	
	This time interval initiates	220
ST interval	at J point and terminates at	320 ms
	T wave.	
	This time interval initiates	
	at QRS complex and	
	terminates at T wave. The	
QT interval	increased OT time interval	300-430 ms
~ ~ ~ ~	may results in ventricular	
	tachyarrhythmia. It varies as	
	rate of heart beat varies.	
	Mostly U wave is easily	
U wave	observed as the amplitude	
	of its peak is very low. The	
	U wave follows the T wave.	

II. FEATURE EXTRACTION

The heart rate recognition system on the basis of the heart rhythms is performed on the basis of the feature vectors that depict the original ECG segments. The ideal recognition system relies upon the features that are extracted from the ECG signals to pretend the difference among various waveforms related to the various heartbeats [1]. Thus the accurate feature selection is the major task to attain the useful information. In order to extract the features in an effective manner a large number researches has been conducted till now. This section represents an overview to the work that had been done in this domain.

Zhao et al [1] proposed a new method which was based on the wavelet transform and support vector machines. This new approach was used to extract the features and to monitor the rhythm of the heart. This technique worked upon three components such as preprocessing, feature extraction and classification of ECG signals. Basically two techniques were involved in the paper which was used to achieve the feature vector of ECG data. One of the two techniques were wavelet transform which extracted the coefficients of the transform and another one is AR i.e. Auto regressive modeling which was used to attain the temporal structure of the ECG. And for the classification purpose, SVM classifier with Gaussian kernel was used. The simulations were performed to identify the accuracy of the proposed technique which was 99.68%.

Castro et al. In [2] proposed another approach for the feature extraction in ECG signals which was also based on the wavelet transform. The proposed algorithm first extract features from the signals and then recognized any abnormal heartbeats. The wavelet transforms can be recognized into both in frequency as well as in time domains. This study introduced a technique by using optimal mother wavelet. Thus in selection of this process; firstly denoising method has applied over the ECG signal. In the denoising process, a soft and hard threshold having a limitation of 99.99 with reconstructs ability and after that with the help of optimal wavelet function PQRST cycle has decomposed into number of coefficients vector. After doing all the processing the ECG signals were examined on the basis of the coefficients and then the approximation of the signals on the behalf of last scale level were done. The coefficient used by each cycle is categorized to three unique segments named as P-Wave, QRS complex and T waves. At last the values of the coefficients were combined to attain the feature vectors.

Mahmoodabadi et al. In [3] introduced a technique to extract the features from ECG signals by using Daubechies wavelet transformation technique that works upon the basis of the multi resolution wavelet transform. The signal processing was done by using the Modified Lead II. Along with this the wavelet filters and scaling function was applied to filter and scale the shape of the signals respectively. Initially, the noise was removed from the signal by deducting the wavelet coefficient which was at the higher scales. Lastly, the QRS complexes were detected where an individual complex was treated as a method to trace the peaks of each wave. The complex QRS wave was utilized to evaluate the P and T wave in the cardiac cycle. The experimental analysis has performed and the results acquired for the proposed technique are 99.18 % sensitivity and 98% prediction accuracy.

Jen et al. in [19] described an approach to determine the features of the ECG signal which was based upon the neural networks. This paper proposed an integrated system for the

diagnosis of ECG. This system consisted with the cepstrum coefficient method which has used the long-term ECG signals in order to extract the features from the signals. Moreover, for the classification purpose, it has used artificial neural network. This proposed method was capable of identifying the hiding characteristics from the ECG signal, classification of signal and diagnoses of abnormalities. The experimental analysis has performed to evaluate the performance of the proposed method. For such purpose, ECG data was taken from MIT/BIH database for the verification. The results acquired showed that the accuracy rate in identifying the cardiac disease was 97.5% and it was capable in extracting the feature vectors, distinguished the difference and ECG signals classification.

Chouhan and Mehta in [20] proposed an algorithm which was used to detect QRS complexities. In this algorithm a filtering procedure based on the moving averages was used that resultant into smooth spike free ECG signal appropriate for the extraction of slope feature of ECG. Initially, slope feature was extracted from the drift corrected as well as filtered ECG signal. The extraction of slope feature has done in a way where the extracted feature signal enhanced in the area of QRS and suppressed in the non-QRS region. Consequently, the proposed method's detection rate and 99.18% productivity accuracy was 98.56% and correspondingly.

Xu et al. in [21] introduced a mechanism to investigate the QRS complex and RR interval by using the Slope Vector Waveform (SVW). The slope vector was evaluated by using the variable stage differentiation technique. Similarly, the feature extraction and non-linear amplifications process was used for better SNR ratio. The proposed method provided a fast as well as accurate search for the R location, RR interval and for the duration of QRS complex.

Alexakis et al. [22] described a new method which was used for the extraction of features and time interval from the ECG signal for the classification into normal and arrhythmic. The proposed method combined ANN i.e. artificial neural network and Linear Discriminant Analysis such as LDA for the extraction of feature. The ECG features extracted were RR, RTc, T wave skewness, amplitude and kurtosis. The above defined features were acquired through the assistance of the automatic algorithms.

Alan et al. in [23] described that chaos theory can be applied to ECG feature extraction procedure. Furthermore, the various chaos techniques like central tendency measure, phase space, correlation dimension, attractors etc. In this paper, a new ECG chaos extractor technique has proposed which applied above mentioned methods. Thus, a new semi-automatic program has been implemented in this paper which was used to extract the features of ECG. Moreover, the paper specified a graphical interface that employed several ECG files for the extraction and selection of method and saving of results. This program extracted several features from these ECG files.

III. HEART RATE PREDICTION USING VARIOUS TRAITS

Detecting heart rate on the basis of the ECG signal is a quite old method of heart rate prediction. On the basis of the above defined work in the domain of ECG signals it is concluded that the heart rate prediction using ECG signals is quite complex and time consuming process because first of all it is required to have ECG report of a patient and then the features are extracted by using feature extraction mechanism. Thus this process can also sometimes leads to the less qualitative results due to wrong choice of feature extraction mechanism. Thus after then, the heart rate prediction is done on the basis of video of the patient by observing the facial expressions, physical activities or speech of the patients. Some contributions in this domain are discussed as follows:

3.1 Heart rate prediction using physical activities

In **[5], Feng Xiao et al** depicted that the heart rate is highly affected by the physical activities thus it provided heart rate prediction model by using neural network on the basis of physical movement of one's body. The simulation was done by using the real life signals received from a healthy male person. The obtained Mean Absolute Error rate of the present work was relatively low which showed the proficiency of the proposed neural model of heart rate prediction.

Ming Yuchi et al [7] developed a physical activities based heart rate prediction scheme which had the capability to be employed in cardiopathy R&D, warning indicator for heart rate and mental activities evaluation etc. The feed forward neural network was adopted for the purpose of proposal. The simulation was done on the real life dataset of 90 minutes and was simulated continually with the interval of 30 seconds.

3.2 Heart rate prediction on the basis of the audio signals Andreas Tsiartas et al [11] developed a method to examine the heart rate on the basis of the variations in speech of a person. The objective of this study was to analyze the way desirable system reacts with physiology of a person and to which extent the heart rate prediction can be done on the basis of the speech signals only. For this purpose, the speech and high qualitative physiology recordings were used. The concept of trigger was used to create a frustration for the user during the recording so that the user can changes his emotions as well. Then the comparison of the results was done on the basis of the pre-triggered and post-triggered regions. And on the basis of the results it was concluded that the both spectral and temporal variations have the capability to predict the heart rate with an accuracy of ~70%.

3.3 Heart rate prediction on the basis of facial expressions Ming Zher Poh et al in **[12]** presented a novel low cost approach for evaluating the multiple physiological activities such as heart rate, pulse rate on the basis of the video by observing the color variation in the video. In this work the facial region was used for extracting the rate of pulse. And the heart rate, Respiratory rate and heart variability was sensed by using the sensors that were placed on the body of the patients. This research had some important significance in the domain of telemedicine and health prediction.

Table 2. Review to the Related Work								
Domain	Authors	Approac h	Simplicit y	Accurac y	Predictio n status			
ECG signal extractio n	Zhao et al.[1]	Wavelet Transfor m and Auto Regressio n	High	High	Moderate			
	Castro et al.[2]	Wavelet Transfor m	Low	High	Low			
	Mahmoodaba di et al.[3]	Multi- resolution wavelet transform	High	High	High			
Various Traits	Vasu Jindal [4]	Deep Learning Mechanis m	Moderate		Moderate			
	Feng Xiao [5]	Neural Network	High	High	High			
	Xiao et al [10]	Neural Network	High	Moderat e	Moderate			
	Aibek Ryskliyev et al [9]	Linear Model	High	High	High			

IV. CONCLUSION

ECG signal processing has gained the attention of various research scholars due to various reasons such as to develop the more enhanced and effective feature extraction mechanism, to develop a tactic to capture the ECG signals such as via video, physical activities, facial expressions etc. This study concludes a brief overview to the ECG signals and PQRST waves. The techniques that are used to extract the features and to perform classification are also covered under this study. After having a review to these traditional techniques, it is observed that the heart rate prediction on the basis of ECG signals can be done by using various traits such as human facial expressions, physical activities performed by a person, by using a video etc. Thus it is concluded that the heart rate prediction on the basis of the video is the most effective way to predict the accurate results.

Therefore in near future, more amendments could be done on frame extracting strategies from the video so that the video with more enhanced quality can be obtained for further processing. Along with this it has been seen that various optimization techniques were used by authors to achieve the optimized results. Accordingly, the advanced optimization techniques can be used for future work.

REFERENCES

- G. Eason, B. Noble, and I.N. Sneddon, "On certain integrals of Lipschitz-Hankel type involving products of Bessel functions," Phil. Trans. Roy. Soc. London, vol. A247, pp. 529-551, April 1955. (*references*)
 - [1]. Q. Zhao, L. Zhan, "ECG Feature Extraction and Classification Using Wavelet Transform and Support Vector Machines," International Conference on Neural Networks and Brain, ICNN&B '05, vol. 2, pp. 1089-1092, 2005.
 - [2]. B. Castro, D. Kogan, A. B. Geva, "ECG feature extraction using optimal mother wavelet," The 21st IEEE Convention of the Electrical and Electronic Engineers in Israel, pp. 346-350, 2000.
 - [3]. S. Z. Mahmoodabadi, A. Ahmadian, M. D. Abolhasani, "ECG Feature Extraction using Daubechies Wavelets," Proceedings of the fifth IASTED International conference on Visualization, Imaging and Image Processing, pp. 343-348, 2005.
 - [4]. V. Jindal, "Mobile SOFT: U: A Deep Learning Framework to Monitor Heart Rate During Intensive Physical Exercise", ACM
 - [5]. F. Xiao, M. Yuchi, J. Jo, M. Y. Ding, W. G. Hou, "A Research of Physical Activity's Influence on Heart Rate Using Feedforward Neural Network", Springer, International Symposium on Neural Networks, 2009
 - [6]. S. Roy, J. Mccrory, "Original Research Validation of Maximal Heart Rate Prediction Equations Based on Sex and Physical Activity Status", International Journal of Exercise Science, vol 8, pp 318-330, 2015
 - [7]. M. Yuchi, J. Jo, "Heart Rate Prediction on Physical Activity using Feed forward Neural Network", International Conference on convergence and hybrid information technology, pp 344-350, 2008
 - [8]. M. Ludwig, H. G. Grohganz, A. Asteroth, "A Convolution Model for Heart Rate Prediction in Physical Exercise", 4th International Congress on Sport Sciences Research and Technology Support, pp 157-164, 2016
 - [9]. A. Ryskaliyev, S. Askaruly, L. P. James, "Speech Signal Analysis for the Estimation of Heart Rates Under Different Emotional States", IEEE, Advances in Computing, Communications and Informatics (ICACCI), 2016.
 - [10]. F. Xiao, Y. Chen, M. Yuchi, "Heart Rate Prediction Model based on Physical Activities using Evolutionary Neural Network", IEEE, pp 198-201, 2010.
 - [11]. A. Tsiartas, A. Kathol, E. Shriberg, M. de Zambotti, A. Willoughby, "Prediction of Heart Rate Changes from Speech Features During Interaction with a Misbehaving Dialog System", ISCA, pp 3715-3719, 2015
 - [12]. M. Z. Poh, D. J. McDuff, R. W. Picard, "Advancements in Noncontact, Multiparameter Physiological Measurements Using a Webcam", IEEE, Transactions on Biomedical Engineering, vol 58, 2011

- [13]. S. Mishra, U. S. Tiwary, "Heart Rate Measurement Using Video in Different User States for Online HCI Applications", ELSEVIER, vol 39, pp 20-27, 2014
- [14]. X. Li, J. Chen, G. Zhao, M. P. ainen ,"Remote Heart Rate Measurement From Face Videos Under Realistic Situations", IEEE, pp 4321-4328, 2014
- [15]. A. Naaz, S. Singh, "Feature Extraction and Analysis of ECG signal for Cardiac Abnormalities- A Review", International journal of Engineering Research and Technology, vol 3, 2014
- [16]. H. Hermansky, N. H. Morgan, P. D. Kohn, "Auditory model for parametrization of speech," Sep. 12 1995, uS Patent, 450-455, 1995.
- [17]. H. Hermansky, N. Morgan, A. Bayya, and P. Kohn, "Rastaplp speech analysis technique," in icassp. IEEE, 1992, pp. 121–124.
- [18]. N. A. Meseguer, "Speech analysis for automatic speech recognition," Norwegian University of Science and Technology, Masters Thesis, vol.109, 2009.
- [19]. K. K. Jen, Y. R. Hwang, "ECG Feature Extraction and Classification Using Cepstrum and Neural Networks," Journal of Medical and Biological Engineering, vol. 28, 2008.
- [20]. V. S. Chouhan, S. S. Mehta, "Detection of QRS Complexes in 12- lead ECG using Adaptive Quantized Threshold," IJCSNS International Journal of Computer Science and Network Security, vol. 8, no. 1, 2008.
- [21]. X. Xu, Y. Liu, "ECG QRS Complex Detection Using Slope Vector Waveform (SVW) Algorithm," Proceedings of the 26th Annual International Conference of the IEEE EMBS, pp. 3597-3600, 2004.
- [22]. C. Alexakis, H. O. Nyongesa, R. Saatchi, N. D. Harris, C. Davies, C.Emery, R. H. Ireland, and S. R. Heller, "Feature Extraction and Classification of Electrocardiogram (ECG) Signals Related to Hypoglycaemia," Conference on computers in Cardiology, pp. 537-540, IEEE, 2003.
- [23]. A. Jovic, N. Bogunovic, "Feature Extraction for ECG Time- Series Mining based on Chaos Theory," Proceedings of 29th International Conference on Information Technology Interfaces, 2007.
- [24]. https://ecgwaves.com/ecg-normal-p-wave-qrs-complex-st-segment-t-wave-j-point/