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An Adaptive Morphologic Filter Applied to ECG De-noising and Extraction of R Peak at Real-time

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Abstract

Electrocardiogram (ECG) is an electrical signal detected by the particular device and method. It is very important to analyze and measure ECG for diagnosis of various heart diseases. However, it is often disturbed by different types of noise. In this paper, an adaptive morphological filter is presented to remove the power line interference from ECG signals, by which the drawbacks of previous morphological method are overcome. Moreover, the extraction of R peak in ECG is also feasible by this filter for further study on ECG. Due to the high speed of our method, the ECG de-noising and R Peak extracting could be both realized at real time, which is an effective method to monitoring patients undergoing surgeries.

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Keywords: Electrocardiogram; Mathematical Morphology; LMS algorithm; Noise suppression

1. Introduction

The electrocardiogram (ECG) is a recording of the heart's electrical potentials which has been always applied to the diagnosis of kinds of heart diseases. The noises disturbing ECG involve generally the power-line interference, baseline drift, electrode polarization noise and the amplifier noise[1]. The most useful

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clinical information in ECG is to find the intervals and amplitudes, for example, the R-R interval detection is a major factor in determining the heart rate, and several related arrhythmias such as Tachycardia, Bradycardia[2]. Many approaches have been applied to the enhancement of filtering ECG, such as FIR filter[3], EMD method[4], Morphology[5]. The primary drawbacks of FIR filter is that it can only filter the interference of fixed frequency, while the spectra of the ECG signal and the interference overlaps. However, EMD method, which is able to effectively remove noise from ECG signal, but requires considerable computational time which is not suitable for real-time processing. In recent years, mathematical morphology has become one of the effective approaches for the processing and analysis of the ECG signals. Although this approach is promising in terms of robustness and computational efficiency, it introduces distortions to the QRS waves [6], which causes the difficulty in extraction of ECG features. In this paper, we modify the morphologic filter to combine it with the adaptive filter algorithm and extracting R peak in ECG. The previous morphologic filter, the adaptive filter, and our method are illustrated in next section. And in the third section, experimental results of clinical ECG recording collected during kinds of surgeries are given. Finally, the discussion and conclusion are in the last part.

2. Methods

2.1. Mathematical morphology

Mathematical morphology(MM) is a branch of mathematics which is based on set theory, topology and random functions. Structuring elements are the basic concept of mathematical morphology, which has a certain form (such as point, line, circle, etc). Erosion and dilation are two basic morphological operators. Open and close can be defined through the transform of them. The morphology operations are defined as follows:

The discrete signal f is defined on $f(n), n=0,1 \dots N-1$, the discrete structure-element g is defined on $g(m), m=0,1 \dots M-1$, and $N > M$, Erosion is a “shrinking” operator while the dilation operation is an “expansion” operation.

$$\text{Erosion operation: } f \ominus g(n) = \min_{m=0,1 \dots M-1} \{f(n+m) - g(m)\} \quad (1)$$

$$\text{Dilation operation: } f \oplus g(n) = \max_{m=0,1 \dots M-1} \{f(n-m) + g(m)\} \quad (2)$$

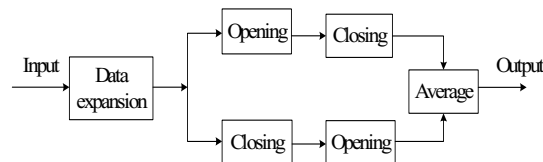
After eroded and dilation operation, the signal is cut down by M length. In order to obtain the whole results, we need to expand the original signal from the beginning and end respectively M length. The open operation and close operation built on eroded operation and dilated operation are defined as follow:

$$\text{Open operation: } f \circ g = (f \ominus g) \oplus g \quad (3)$$

$$\text{Close operation: } f \bullet g = (f \oplus g) \ominus g \quad (4)$$

In most applications, opening is used to suppress peaks, while closing is used to suppress pits. Opening is a shrinking operator while closing is a expansion operator. So the two filters combine in order to suppress one-way migration[7]. The morphology method used in our paper is mainly based on the following equation.[8]. Figure1 illustrates the structure of morphology used in morphology filter.

$$\text{OC_CO}(f, g) = (f \circ g \bullet g + f \bullet g \circ g) / 2 \quad (5)$$



interference in noisy ECG signal f_0 . g_1 is a triangular structure element which width is $0.02F_s$. Signal f_1 is obtained from the noisy ECG signal after OC_CO operation, and the noise has been removed from it, as well as the peaks and pits in high amplitude QRS wave. Signal p contains the important peaks and pits cut down in the operation of morphology, however, there is a lot of noise too.

- (2) Secondly, signal p is processed through adaptive filter to remove the noise. Figure 2 shows the structure of this self-adaptive filter structures.
- (3) Applying the structure element g_2 to filter baseline wander.
- (4) Reconstruct the signals with e to get the de-noised ECG f_3 .
- (5) Lastly, apply the structure element g_3 to extract QRS wave. g_3 is a linear structure element which width is $0.09F_s$.

4. Experimental Results

To prove whether our algorithm is appropriate for the ECG signal, we use the real ECG signal getting from the process of operations of Taiwan hospital. The sample rate is 300Hz, and the noise signal was obtained by adding 60Hz noise into the pure signal. We use the the signal-to-noise ratio (SNR) and root-meansquare error (RMSE) to measure the performance of our methods.

$$SNR = 10 \log_{10} \left(\sum_{i=1}^N \frac{x(i)^2}{[x(i) - y(i)]^2} \right) \quad (6)$$

$$RMSE = \frac{1}{N} \sqrt{\sum_{i=1}^N [x(i) - y(i)]^2} \quad (7)$$

In the above two equations, $x(i)$ is the original ECG signal, $y(i)$ is the signal after filtered. Our experiment is composed of two parts. We chose the record 1 from a 78 years man in experiment1, because it captures normal sinus rhythms and is reasonably free of noise, and record 2 from a 56 years man in experiment 2 is corrupted with synthetic PLI with frequency and baseline wander. Figure 4 show the results of experiment 1.

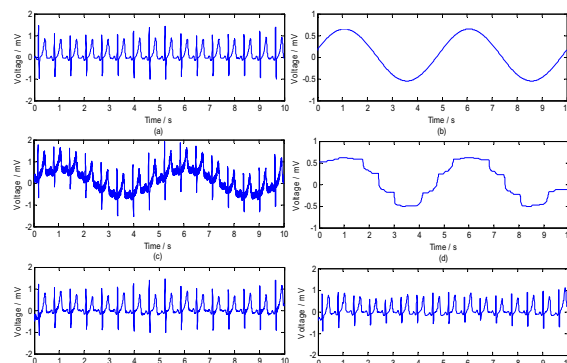


Fig4 .The experimental results by several filtering methods. (a) The pure ECG signal. (b)baseline wander (c) ECG signal corrupted with power line interference and baseline wander. (d)baseline wander extracted from morphology operator (e) The result of adaptive morphologic filter (f)The result of previous morphologic filter.

Table1 The result of SNR and RMSE in the experiment1

Method	SNR/db	RMSE	Time of computation (s)
Adaptive morphologic filter	11.6620	0.0018	0.3606
Previous morphology	10.0172	0.0021	0.3541

From Figure 4 and Table1, our method has the better performance with the improved SNR and RMSE between the pure signal and the filtered ECG was reduced. The morphology method alone can suppress the noise and cut down the peak of R wave at the same time. Although our method will take a little more time to process, it can keep the important features in noisy ECG signal unchanged, which has an important effect on real-time ECG processing.

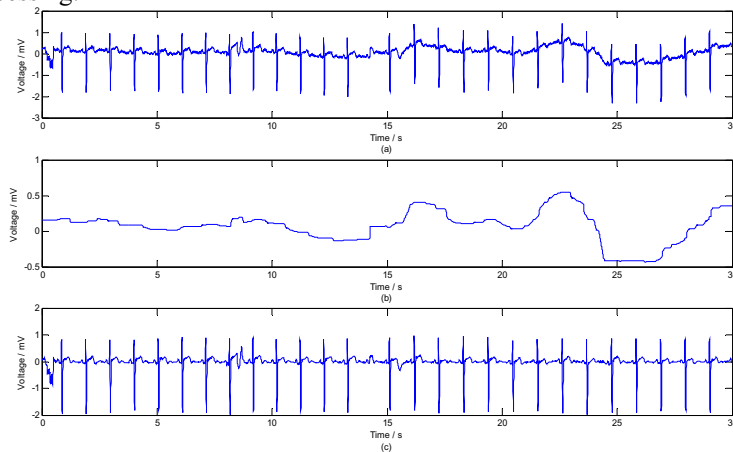


Fig5 .The result of experimental 2(a) The corrupted ECG signal. (b)baseline wander extracted from morphology operator (c) The result of adaptive morphologic filter

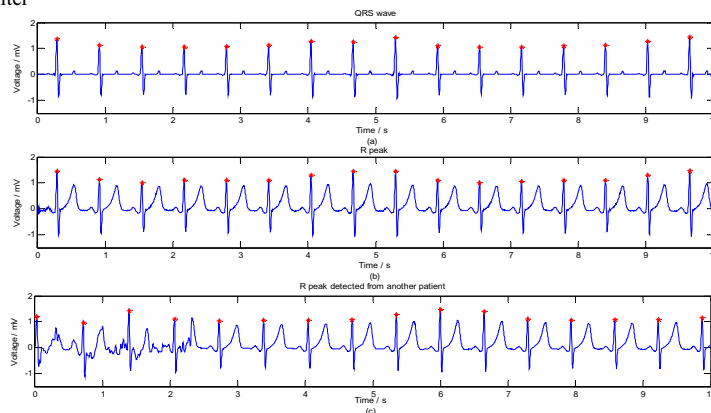


Fig6.QRS wave extracted from morphology and R peak (a) QRS wave obtained from morphology method (b) Detected R peaks (c) R peak detected from another patient

Figure5 proves that the adaptive morphologic filter is very effective in processing real-time ECG signals. Figure6 shows that the R peak calculated from QRS wave which was extracted from morphology filter2, morphology method keeps the ECG signal unchanged and it is very effective in real-time processing in R-R interval.

5. Discussion and Conclusion

In this paper, a new ECG de-noising method which combines morphology with adaptive filter with LMS algorithm is proposed. This new method overcomes the disadvantage that the morphology will cut down the R peaks in ECG signals. It not only suppress the power line interference effectively, but also minimally the losses of the useful information in ECG signal. In addition, our method applies LMS algorithm to reduce the noise in R peaks obtains from morphology. At last, we applied morphology method to extract the QRS wave from the de-noised signal, which can benefit for processing R-R interval in real-time effectively. However, how to select a appropriate structuring elements to adapt to individual differences is the most difficult part in future work.

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