



Removal of Power Line Interference from Electrocardiograph (ECG) Using Proposed Adaptive Filter Algorithm

By Duong Trong Luong, Nguyen Duc Thuan & Dang Huy Hoang

Hanoi University of science and Technology, India

Abstract- ECG signals in measurements are contaminated by noises including power line interference. In recent years, adaptive filters with different approaches have been investigated to remove power line interference in ECG. In this paper, an adaptive filter is proposed to cancel power line interference in ECG signals. The proposed algorithm is experimented with MIT-BIH ECG signals data base. The algorithm's results are compared with the results of other adaptive filter algorithms using Least Mean Square (LMS), Normalized Least Mean Square (NLMS) by Signal to Noise (SNR). These works are performed by LabVIEW software.

Keywords: ECG signal processing; adaptive filters, power line interference, least mean square.

GJCST-C Classification : H.3.5



Strictly as per the compliance and regulations of:



Removal of Power Line Interference from Electrocardiograph (ECG) Using Proposed Adaptive Filter Algorithm

Duong Trong Luong ^α, Nguyen Duc Thuan ^σ & Dang Huy Hoang ^ρ

Abstract- ECG signals in measurements are contaminated by noises including power line interference. In recent years, adaptive filters with different approaches have been investigated to remove power line interference in ECG. In this paper, an adaptive filter is proposed to cancel power line interference in ECG signals. The proposed algorithm is experimented with MIT-BIH ECG signals data base. The algorithm's results are compared with the results of other adaptive filter algorithms using Least Mean Square (LMS), Normalized Least Mean Square (NLMS) by Signal to Noise (SNR). These works are performed by LabVIEW software.

Keywords: ECG signal processing; adaptive filters, power line interference, least mean square.

I. INTRODUCTION

Electrocardiography (ECG) plays an important role in monitoring and diagnosing cardiovascular diseases. The frequency bandwidth for therapy of the ECG signal is from 0.05Hz to 100Hz, and the highest peak is about 1mV [1]. During recording of ECG signals, the ECG signals get contaminated such as power supply harmonic 50Hz or power line interference (Most country including Viet Nam use 50Hz electric system) with the amplitude approaches 50% the highest peak of the ECG signal, artifacts caused by losing the direct contact between electrodes and the skin, or by EMG- this artifact's amplitude is 10% of the highest peak's amplitude of the ECG signal, or by respiratory with noise's amplitude is 15% of the highest peak of the ECG signal at 0.3 Hz frequency [2]. In the listed noises, the power line noise affects P wave and Q wave of the ECG signal. That causes errors in arrhythmia and myocardial infarction diagnosis [3]. In recent years, a few adaptive filters with different approaches are investigated to remove the power line noise 50 Hz in the ECG signal such as design of an adaptive filter with a dynamic structure for ECG signal processing [3], adaptive filtering in ECG denoising: a comparative study [4], denoising ECG signals using adaptive filter algorithm [5], denoising ECG signals with adaptive filtering algorithm & patch based method [6], investigation of adaptive filtering for noise cancellation in ECG signals [7], designing and implementation of algorithms on Matlab for Adaptive noise cancellation from ECG signal [8], performance

comparison of adaptive filter algorithms for ECG signal Enhancement [9], performance evaluation of different adaptive filters for ECG signal processing [10]. Most of the researches use Least Mean Square (LMS) and Normalize Least Mean Square algorithm (NLMS). These algorithms enable to change filter coefficients with given order; the algorithms are quite reliable and effective with small convergent time. However, in case of the power line noise's amplitude equals to 40- 50% amplitude of the highest (QRS peak) of the ECG signal, these algorithms give not effective results, and the filtered signal still contains noise. To overcome this problem, the authors propose an adaptive filter algorithm based on Fast Fourier Transform (FFT). This algorithm is experimented with ECG database such as number of record 117 and aVL lead of patient 279/s0532 from ECG database MIT-BIH [11]. The results of this algorithm are compared with that of LMS, NLMS by SNR criterion. The process and experiments are performed by LabVIEW software.

II. METHODOLOGY

The aim of adaptive filter based on Fast Fourier Transform (FFT) is detecting the power line noise frequency, and determining threshold of this noise's magnitude.

Fourier transform:

If $x(n)$ is a discrete signal satisfying the condition (1),

$$\sum_{n=-\infty}^{+\infty} |x(n)| < \infty \quad (1)$$

So Fourier transform equation for $x(n)$ is given formula (2)

$$X(\omega) = \sum_{n=-\infty}^{\infty} x(n)e^{-i\omega n} \quad (2)$$

where i is the imaging part, $i^2 = -1$

According to the material [12], B. Widrow has shown that the adaptive filter transfer function is described as (3)

$$H(z) = \frac{z^{-2} - (2\cos\omega_0)z^{-1} + 1}{(1 - 2\beta K^2)z^{-2} - \{2(1 - \beta K^2)\cos\omega_0\}z^{-1} + 1} \quad (3)$$

where β is the step size of the adaptive filter; K is the magnitude of power line noise; ω_0 is the angular frequency. In this study, if $S(n)$ is ECG signal contaminated power supply harmonic 50 Hz, so $S(n)$ is expressed in the equation (4).

Author $\alpha \sigma \rho$: Hanoi university of Science and Technology.
e-mail: luong.duongtrong@hust.edu.vn

$$S(n) = X(n) + N(n) \quad (4)$$

Where $X(n)$ is the clean ECG signal, $N(n)$ is the power supply harmonic 50 Hz. Based on the equations (1) and (2), (4) equation can be displayed in another form (5).

$$S(\omega) = X(\omega) + N(\omega) \quad (5)$$

The magnitude-frequency spectrum of $X(\omega)$, $N(\omega)$ and $S(\omega)$ is shown as figure 1, figure 2, and figure 3 respectively.

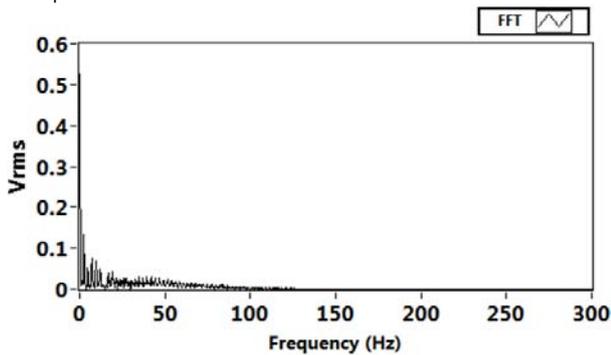


Figure 1 : Magnitude-frequency spectrum of clean ECG signal

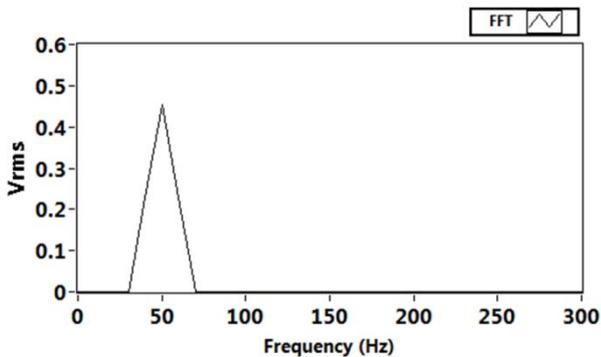


Figure 2 : Magnitude-frequency spectrum of power line noise 50 Hz with magnitude is 0.4mV

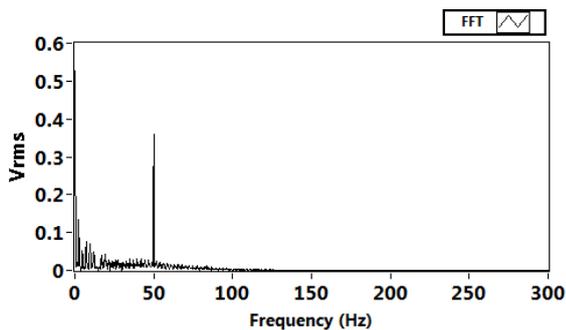


Figure 3 : Magnitude-frequency spectrum of the contaminated ECG signal

That determines the magnitude and frequency of power line noise follows these below steps.

Step 1: Importing input signal $S(n)$

Step 2: Transforming FFT $S(n)$ (for detecting the magnitude and frequency of the 50Hz power supply)

Step 3: Determining the frequency and magnitude of the power line interference: Choosing the frequency bandwidth is from 30Hz to 70Hz; setting the threshold of noise's FFT magnitude is 15 (based on experiments with FFT of the contaminated ECG signal)

Step 4: Initializing for loop, starting from $i = 0$

Step 5: Checking condition: the magnitude of the power line noise in FFT is less than 15 or not? If true:

Step 6: Displaying the filtered signal and ending the processing.

If false:

Step 7: Realizing the transfer function (3)

Step 8: Calculating the output signal:

$$y(n) = H(n) \cdot S(n)$$

Step 9: Transforming FFT $y(n)$ (for checking whether or not existing noise in the frequency bandwidth from 30Hz to 70 Hz?)

Step 10: Increasing the iteration: $i = i + 1$ and returning to the step 5.

To experiment and test these above steps and LMS, NLMS adaptive filter algorithms, the authors used adaptive filter toolkit available in Lab VIEW. The results of the proposed algorithm are compared with that of LMS and NLMS algorithms by Signal to Noise (SNR). This criterion is followed by equation (6).

$$SNR = 20 \log_{10} \left\{ \frac{RMS(y(n))}{RMS(x(n) - y(n))} \right\} \quad (6)$$

where $RMS(y(n))$ is the Root Mean Square of the filtered ECG signal; $RMS(x(n))$ is the Root Mean Square of the original ECG signal.

III. RESULTS AND DISCUSSION

The authors have tested the proposed algorithm by using a few standard ECG database records such as record mitdb117 and a VL lead of record patient 279/s 0532 from ECG database MIT-BIH added with the power supply harmonic 50 Hz having variable magnitude is from 0.4mV to 0.5mV. This signal is generated by using LabVIEW. Figure 4a is the ECG signal of the record mitdb117, and this signal is added with the power line noise 50Hz (shown in Fig.4b). Figure 4c, 4d and 4e display the results of filtering the power line interference with 0.4 mV magnitude using the proposed adaptive filter, LMS and NLMS corresponding adaptive filters. Intuitively, the filtered ECG signal using the proposed algorithm has nearly no appearance of noise and has the morphology similar to the original ECG signal.

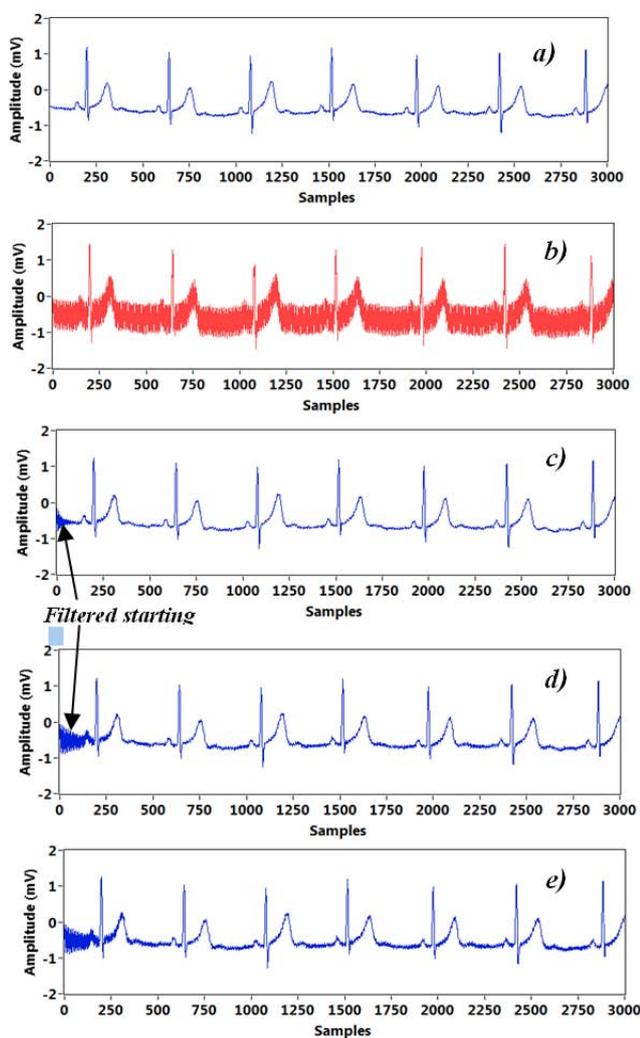


Figure 4 : The results of removing power line interference 50 Hz. a)Clean ECG signal; b) ECG signal contaminated by power line noise with 0.4 mV magnitude; c) recovered ECG signal using proposed Algorithm; d)recovered ECG signal using MLS Algorithm; e) recovered ECG signal using NLMS Algorithm

To prove the efficiency of the proposed algorithm in filtering the power line noise in ECG signal compared with LMS, and NLMS algorithms, the authors continue experimenting these algorithms with ECG data base of aVL lead of recordpatient279/s0532.The results are shown in figures 5c, 5d and 5e.These results indicate that the better efficiency of the proposed adaptive filter algorithm compared with the others. Figure 5c shows the effectiveness.

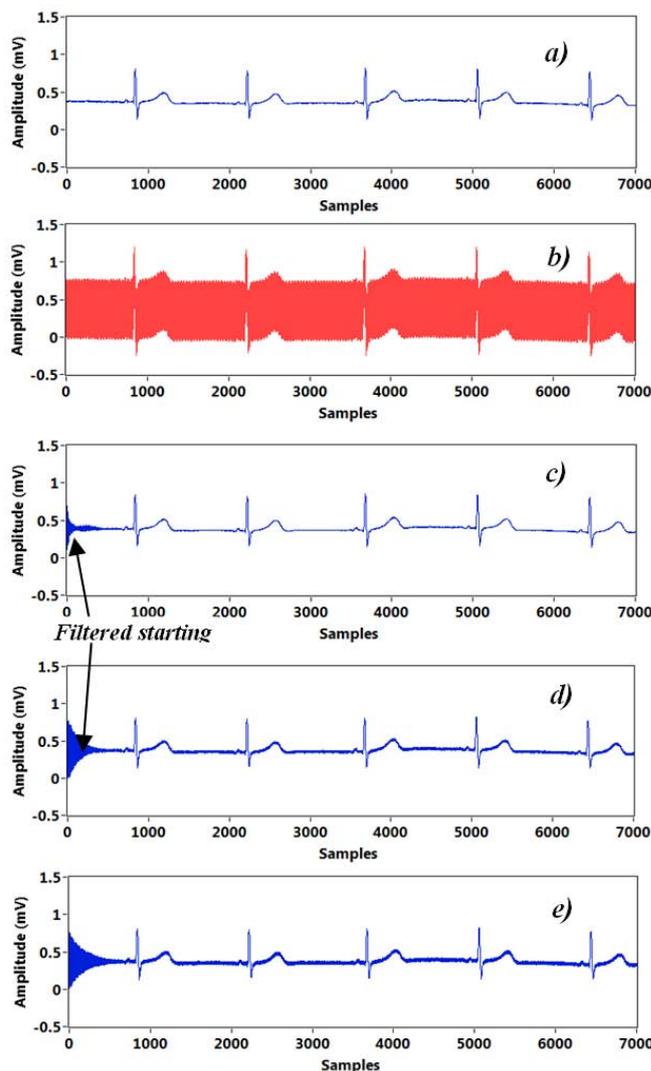


Figure 5 : The results of filtering power supply harmonic 50 Hz. a)Original ECG signal; b) Contaminated ECG signal(noise's magnitude is 0.4 mV); c) recovered ECG signal using proposed Algorithm; d)recovered ECG signal using MLS Algorithm; e) recovered ECG signal using NLMS Algorithm

To affirm the efficiency of the proposed algorithm with LMS, NLMS algorithms, the authors used SNR criterion SNR calculated by equation (6). Table 1 show the comparison among three methods filtering the power line noise with 0.4 mV magnitude, and the noise is added directly to the ECG signal in the record mitdb117.The SNR value of the proposed algorithm is higher than that of LMS and NLMS algorithms when the step size $\beta=0.02$; $\beta= 0.03$ and $\beta=0.04$. However, in case that $\beta=0.01$ the SNR value of the proposed algorithm is just smaller 0.079 than that of NLMS.

Table 1 : Results of comparison among three algorithms filtering the power line noise 50 Hz with 0.4 mV magnitude which is added to the ECG signal of the record mitdb117.

Algorithm	Step size	SNR
LMS	0.01	6.590
NLMS		6.694
Proposed		6.615
LMS	0.02	6.509
NLMS		6.511
Proposed		6.615
LMS	0.03	5.062
NLMS		4.800
Proposed		5.646
LMS	0.04	5.151
NLMS		4.455
Proposed		5.646

Table 2 : Results of comparison among three algorithms which filter the power line noise (with 0.4 mV magnitude) added to ECG signal of aVL lead of record patient279/s0532.

Algorithm	Step size	SNR
LMS	0.01	2.709
NLMS		2.744
Proposed		2.877
LMS	0.02	2.697
NLMS		2.671
Proposed		2.877
LMS	0.03	2.737
NLMS		2.678
Proposed		2.877
LMS	0.04	2.788
NLMS		2.704
Proposed		2.877

From the table 2, the SNR value of the proposed algorithm is higher than that of LMS, NLMS algorithms with step size $\beta=0.01 \div 0.04$.

IV. CONCLUSION

The proposed adaptive filter algorithm for removing the power line interference in ECG signal based on Fast Fourier Transform (FFT) has been investigated with applications in the steps of the algorithm. That detects the power line frequency, and sets the threshold for the noise magnitude in FFT has high efficiency. In addition, that uses many for loops in the proposed algorithm support to filter the power line noise more carefully. The appearance of the noise is insignificant in the filtered ECG signal. Three algorithms are proceeded at the same time: adaptive filter LMS, NLMS (with different step sizes), and the proposed algorithm to remove the power supply harmonic 50 Hz (with 0.4 mV magnitude) added to the ECG signal of the record mitdb117, aVL lead of the record patient 279/s0532. The experimental results show that the proposed adaptive filter algorithm produces good ECG

signal with little noise, similar to the original ECG signal displayed in figure 4 and figure 5. Furthermore, to demonstrate the efficiency of the algorithm, the authors have compared the proposed algorithm with LMS and NLMS adaptive filter algorithms by SNR criterion. From the results in table 1 and table 2, the proposed adaptive filter algorithm for removing the power line noise 50 Hz with 0.4 mV magnitude has higher efficiency. That is asserted by SNR value in the table 1 and table 2.

REFERENCES RÉFÉRENCES REFERENCIAS

1. Rangaraj M.Rangayyan "Biomedical signal analysis". John Wiley & Sons, Inc. Canada, 2002.
2. G.D. Clifford, F. Azuaje, and P. McSharry "Advanced Methods and Tools for ECG Data Analysis". Norwood, MA, USA: Artech House, Inc., 2006.
3. Ju-Won Lee, Gun-Ki Lee " Design of an adaptive filter with a dynamic structure for ECG signal processing". International Journal of control, automation and system, vol 3, no.1, pp.137-142, 2005.
4. I Romero, D Geng, T Berset "adaptive filtering in ECG denoising: a comparative study". Computing in Cardiology, 39:45-48, 2012.
5. Chinmay Chandrakar, M.K. Kowar "Denoising ECG signals using adaptive filter algorithm". International Journal of Soft Computing and Engineering (IJSCE) ISSN: 2231-2307, Volume-2, Issue-1, March 2012
6. A.Deo, M.K.Bandil, DBV Singh, A.K.Wadhvani "denoising ECG signals with adaptive filtering algorithm & patch based method". International Journal of Computer Networks and Wireless Communications (IJCNCW), ISSN: 2250-3501, Vol.3, No3, 2013
7. Soroor Behbahani "Investigation of Adaptive Filtering for Noise Cancellation in ECG signals". IEEE, 2007.
8. H.K.Gupta, R.Vijay, N.Gupta "designing and implementation of algorithms on Matlab for Adaptive noise cancellation from ECG signal". International Journal of Computer Applications, Volume 71– No.5, 2013
9. S.A.Rehman, R.R.Kumar "performance comparison of adaptive filter algorithms for ECG signal Enhancement". International Journal of Advanced Research in Computer and Communication Engineering, Vol. 1, Issue 2, 2012
10. Sachin Singh, K.L.Yadav" performance evaluation of different adaptive filters for ECG signal processing". International Journal on Computer Science and Engineering, Vol. 02, No. 05, 1880-1883, 2010.
11. <http://physionet.org/cgi-bin/atm/ATM>
12. B.Widrow, S.D.Steams, "Adaptive signal processing". Englewood Cliffs, Prentice Hall. 1985