



Performance Analysis of the Recursive Least Squares Algorithm

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ABSTRACT: Adaptive filter algorithm is a widely used method in communication systems, control systems, digital signal processing etc. This method helps to find out the unknown parameters iteratively by adjusting the filter parameters. There are many efficient adaptive filter algorithms. But among them, the basic algorithms are: Least Mean Square (LMS) and Recursive Least Square (RLS) Algorithms. The LMS algorithm is based on gradient optimization and the RLS algorithm is based on direct form FIR and lattice realization. The RLS algorithm is popular because of its fast convergence although the LMS algorithm is very simple to implement. There are modified LMS algorithms and they are: Leaky Least Mean Square (LLMS) Algorithm and Normalized Least Mean Square (NLMS) Algorithm. ‘Step size’ is an important parameter which is used to implement any of these LMS algorithms. In case of RLS algorithm, one term ‘forgetting factor’ plays an important role in times of implementing any system.

KEYWORDS: Recursive Least Square (RLS) Algorithm, Least Mean Square (LMS) Algorithm, Leaky Least Mean Square (LLMS) Algorithm, Normalized Least Mean Square (NLMS) Algorithm, Frequency Response Curve, Coefficient Response Curve, Error Response Curve, Step Size, Forgetting Factor.

I. INTRODUCTION

Adaptive filter is a popular filter which is used in the domain of Digital Signal Processing (DSP). It has received considerable attention from the researchers for last forty years. Naturally good number of efficient adaptive filtering methods has been developed within this span of time. Now, in times of designing an intelligent system, adaptive process is required [7-10]. As a designer, a digital system is always preferable because of its small size, lower cost, fast speed of operations and flexibility in operation. In the field of DSP, adaptive filter algorithm is a popular tool to design any intelligent system. The digital filters may be classified as Infinite Impulse Response (IIR) and Finite Impulse Response (FIR) Filters. FIR filters are preferred for more stable design due to the lack of feedback in the design and such reliable design in will affect the evolution on adaptive process.

An adaptive filter is a filter which is implemented by using error signal with adjustable coefficients. In times of implementing any adaptive process, a cost function is taken for optimization purpose. The optimum solution of the algorithm can be achieved by minimizing the cost function. The adaptive process is time varying and non-linear in nature.

In this paper the basics of RLS and LMS adaptive filters algorithms are surveyed and discussed. The Recursive Least Square (RLS) algorithm is surveyed and discussed. At the same time for doing the comparison, the Least Mean Square (LMS) algorithm is also discussed. The LMS algorithm is the basic algorithm which is used for any adaptive filter and the modified LMS algorithms are Normalized Least Mean Square (NLMS) and Leaky Least Mean Square (LLMS) Algorithms which are also discussed very briefly.

In the next section of the paper, the Adaptive Filtering Algorithms (II), the RLS and different LMS algorithms, the implementation of RLS and different LMS Algorithms (III), the Result Analysis (IV), the Conclusion (V) and the References are given.

II. ADAPTIVE FILTERING ALGORITHMS

To design an adaptive filter, adaptive filter algorithm is used iteratively. The basic block diagram of adaptive filter is given in figure (1). For any adaptive process, the error signal $e(n)$ which is calculated as

$$e[n] = d[n] - y[n] \text{-----(1)}$$

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where the number of iteration is n , the input signal is denoted by $x(n)$, the adaptive-filter output signal is $y(n)$ and $d(n)$ is the desired signal.

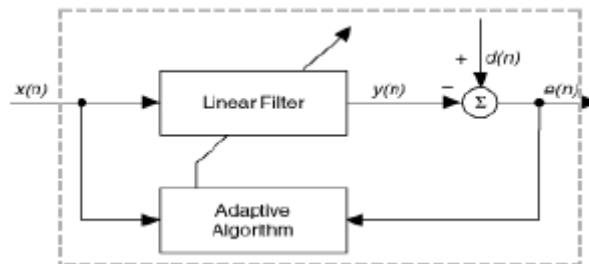


Figure1: Basic Adaptive Filter Algorithm Scheme

To form an objective function, the error signal is used. In order to determine the appropriate filter coefficients, the adaptation algorithm is used. The minimization of objective function will help the adaptive-filter output signal to match with the desired signal. So the error signal is formed and optimized to find the optimized solution. Hence the choice of error signal in an adaptive algorithm is very important. The overall convergence process also influences the adaptive algorithm. A filter consists with many coefficients. There are many methods for performing weight update of an adaptive filter. The different filtering algorithms are: Recursive Least Square (RLS), Least Mean Square (LMS), Normalized Least Mean Square (NLMS) and Leaky Least Mean Square (LLMS)[1,9,10].

Recursive Least Squares Algorithm

The Recursive Least Squares (RLS) algorithm is also used to find out the coefficient of adaptive filter. The algorithm uses information from all past input samples to estimate the autocorrelation matrix of the input vector. To decrease the influence of input samples from the past, a weighting factor for the influence of each sample is used. This weighting factor is introduced in the cost function. The cost function is denoted by

$$J[n] = \sum_{i=1}^n p^{n-i} e^2[i, n] \text{-----(2)}$$

Here, the error signal $e[i, n]$ is computed for $1 \leq i \leq n$ using the current filter coefficients $c[n]$ and the error signal is denoted by

$$e[i, n] = d[i] - c^H[n]x[i] \text{-----(3)}$$

The coefficient update equation is given by in equation (10)

$$c[n] = c[n-1] + k[n] * \{d[n] - x^H[n]c[n-1]\} \text{-----(4)}$$

To update the coefficient, the following informations are required.

$$p[n] = \varphi^{-1}[n] \text{-----(5)}$$

Here $p[n]$ can be represented by using the update equation (12) with the help of matrix inversion lemma. Now the value of $p[n]$ can be founded by

$$p[n] = \lambda p[n-1] + \lambda^{-1}k[n]x[n] \text{-----(6)}$$

And the value of $k[n]$ can be calculated by using equation (13)

$$k[n] = \frac{\lambda^{-1}p[n-1]x[n]}{1 + \lambda^{-1}x^H[n]p[n-1]x[n]} \text{-----(7)}$$

The value λ is also termed as forgetting factor.

The RLS algorithm is computationally more complex than the LMS algorithm. The RLS algorithm typically shows a faster convergence compared to the LMS algorithm.



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Least Mean Square Algorithm

Least Mean Square (LMS) was developed by Widrow & Holf in 1959. The steepest descent approach is used in LMS algorithm for estimating the gradient vector for deriving a cost function [7-10]. The sample values of the tap-input vector and an error signal are used for estimating the gradient. To check the performance of any algorithm, it is necessary to use a reference signal $d[n]$ and this reference signal is taken as the desired filter output. The difference between the reference signal and the actual output is taken as the error signal which is represented in Equation 2.

$$e[n] = d[n] - c^H [n]x[n] \text{-----(8)}$$

A set of filter coefficients is represented by c and this set of coefficients is minimized for the expected value of the quadratic error signal to achieve the least mean squared error. To update the coefficients in LMS algorithm, Equation 3 is used and it depends on every time instant n ,

$$c[n+1] = c[n] + \mu e^* [n].x[n] \text{-----(9)}$$

In this context, the selection of the parameter ‘step-size’ (μ) in equation 3 is important. The movement of coefficients along the error function surface at each updated step is controlled by varying the step size.

Leaky Least Mean Square Algorithm

In Leaky LMS (LLMS) algorithm, the cost function is defined by the following equation:

$$j[n] = e^2[n] + \gamma \sum_{i=0}^{N-1} w_i^2[n] \text{-----(10)}$$

where γ is the leaky factor and the range of γ is 0 to 0.1. The cost function of the LLMS algorithm is different from the standard LMS algorithm. The leaky LMS algorithm updates the coefficients of an adaptive filter by using the following equation:

$$w[n+1] = (1 - \gamma\mu)w[n] + \mu e[n]u[n] \text{-----(11)}$$

Now another term μ is present in leaky LMS. In leaky LMS algorithm, the migration of coefficients may cause an overflow problem.

If $\gamma = 0$, the leaky LMS algorithm becomes the same as the standard LMS algorithm. A steady state error may arise for selecting a large leaky factor is chosen,

Normalized Least Mean Square Algorithm

The simple LMS algorithm is sensitive to the scaling of its input $x(n)$. Due to this reason, it is important to choose a learning rate μ properly. At the same time, the parameter must be chosen in such a way that the stability of the algorithm may not hamper. This problem can be overcome by using the recursion formula of Normalized Least Mean Square (NLMS) algorithm which is stated in equation 12;

$$w[n+1] = w[n] + \mu[n]e[n]u[n] \text{-----(12)}$$

Here, $\mu[n]$ is the step size and it is varying with the time.

$$\mu[n] = \frac{\beta}{x[n]x^T[n] + c} \text{-----(13)}$$

$\mu[n]$ is denoted by Equation 7 and c is a small positive constant to avoid division by zero and β is normalized step-size nearly varying from $0 < \beta < 2$.

III. IMPLEMENTATION OF ADAPTIVE FILTER ALGORITHMS

MATLAB (version 7.8) is used to implement the design and test the system with RLS and all the LMS algorithms. Here a system is used to derive a desired response. A FIR low pass filter is used as the desired system to implement the adaptive filter algorithms and the duration of impulse response is finite. Window method is a popular method to implement this FIR low pass filter. Blackman window is used for test purpose. FIR system is used because it is easy to implement. At the same time the FIR filter is inherently stable.

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So the Blackman window function is expressed as:

$$w[n] = 0.42 - 0.5 \cos\left(\frac{2\pi n}{N-1}\right) + 0.08 \cos\left(\frac{4\pi n}{N-1}\right) \text{-----(14)}$$

where, $0 \leq n \leq N-1$

The input is taken as

$$x[n] = \sin(\omega_c(n - \alpha + \epsilon)) / (\pi * (n - \alpha + \epsilon)) \text{-----(15)}$$

where ω_c is the cut-off frequency, $\epsilon = \text{phase} = 0.001$ and $\alpha = (\text{No. of taps} - 1) / 2$;

Finally the RLS, LMS, LLMS and NLMS algorithms are implemented by using MATLAB and the system is tested by using the above input with Blackman window function. A comparative study of system outputs are carried out by fixing all depending parameters of all the algorithms.

IV. RESULT ANALYSIS

To analyse the RLS algorithm, the test system is a low pass filter (LPF). In times of designing, a finite impulse response (FIR) type LPF is used and that FIR filter is preferred to design by using Blackman window function. Initially the RLS algorithm is applied to find out the frequency response for different values of the forgetting factor (λ). The coefficient response curves and the error response curves are also plotted. Finally the LMS, LLMS and NLMS Algorithms are applied to the system and the frequency response curves for these systems are plotted and compared against the response curve that we have obtained for RLS algorithm. The number of iterations is taken as 1000. The frequency response curves are plotted by varying the number of taps with varying forgetting factors.

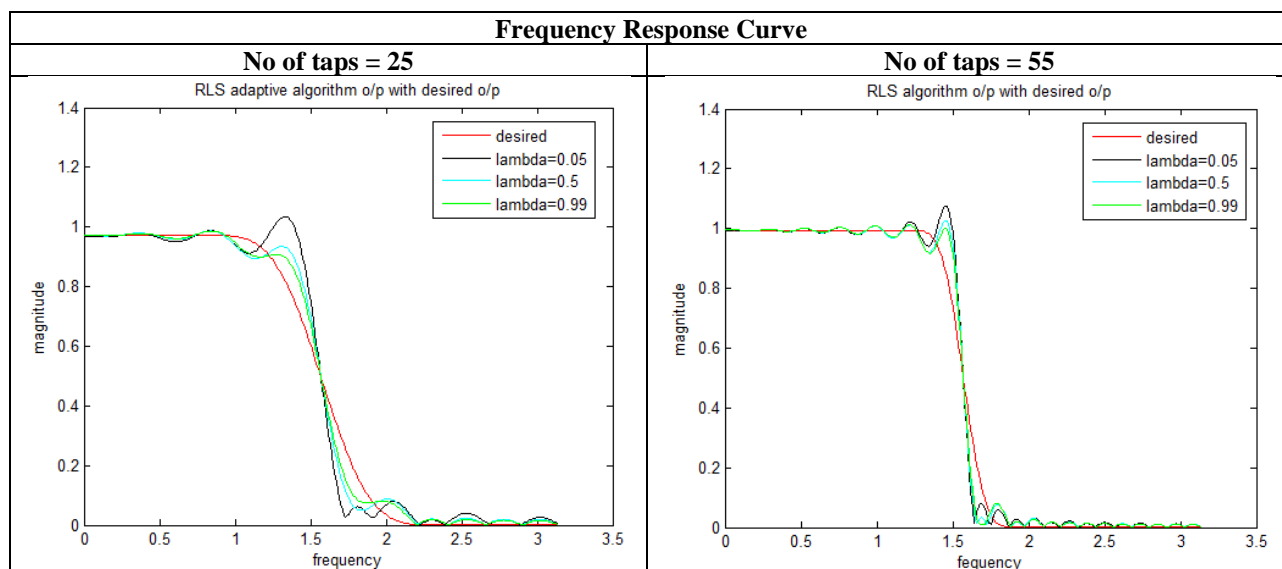


Figure 2: Magnitude vs. Frequency curves of RLS algorithm with different forgetting factors for different taps

From figure 2, we observe that some amount of oscillations is coming in the response curves with the variation λ . But still when $\lambda=0.5$ for 55 number of taps, the amount of oscillation coming in the responses is becoming reduced. Hence, we get the best response here.

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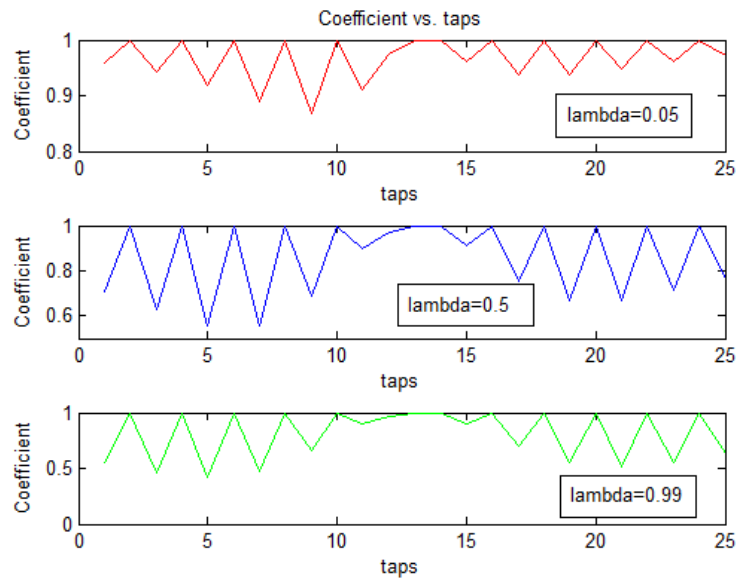


Figure 3: Coefficient response curves using RLS Algorithm with different forgetting factors

In figure 3, we observe that if the value of λ is increased, the response of the coefficient response becomes best.

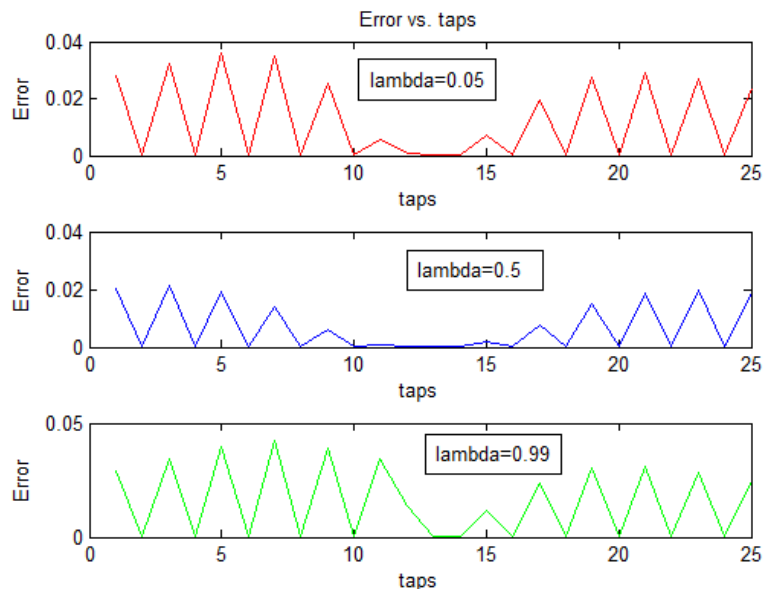


Figure 4: Error response curves using RLS Algorithm with different forgetting factors

In figure 4, we observe that if the value of λ is 0.5, the response of the error response becomes best.

In the following section, the step size is chosen as 0.75 and number of iteration is chosen as 1000. The plots are for 25 taps and for 55.

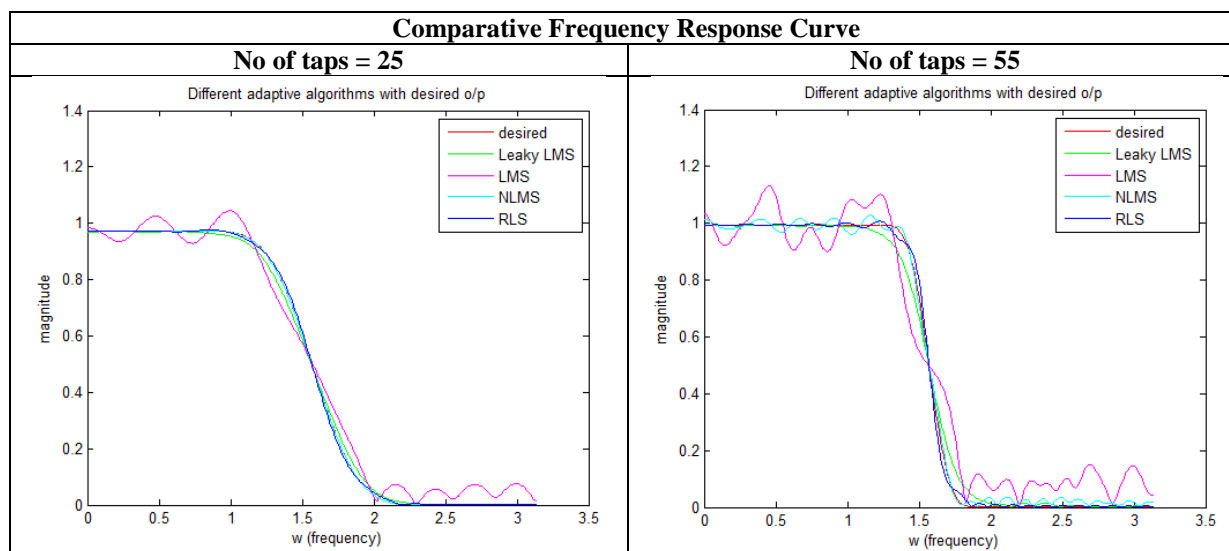


Figure 5: Comparative study on Frequency response curves of adaptive algorithm for different taps

In figure 5, we observe that the frequency responses of LMS, NLMS, LLMS and RLS are plotted together. Now we can notice that we get best response for NLMS when the number of taps is taken as 25. Now considering the number of taps as 55, the responses correspond to LMS, NLMS, LLMS got distorted but the response of RLS got better than response when the number of taps was 25.

So, if the number of tap value is increased, responses of LMS, NLMS, and LLMS get noisy and RLS get better response. Hence, the RLS algorithm shows a better response with the increase of number of taps.

V. CONCLUSION

In case of RLS algorithm, the range of forgetting factor must be chosen in such a way that we get reduced oscillation. In this paper, we kept the range $0 < \lambda < 1$. We get less oscillation when $\lambda = 0.5$ in the frequency response curve. At the same time the amount of error is less when the forgetting factor is chosen as $\lambda = 0.5$ in the error response curve. If the value of λ is increased, the better coefficient response can be obtained.

If the number of iterations and the step size of all adaptive algorithms are kept constant, the responses of LMS, NLMS, and LLMS get noisy but the response of RLS algorithm get better with the increased step size. Hence, the RLS algorithm shows a better response with the increase of number of taps.

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