

HOMEWORK ASSIGNMENT #7

Due Wed. Nov. 23, 2005

1. Recall homework problem 5.2 where we examined LMS in a time-varying system identification application. We will now examine RLS in this application. Referring to the notation and diagram of problem 5.2, assume that

- $\{x(n)\}$  is generated by an AR-2 model with  $[a_1, a_2] = [0.6, -0.1]$  and driving noise  $\sigma_v^2 = 0.5$ .
- $\sigma_z^2 = 0.002$ .
- $\mathbf{h}(0) = [-1, 0, 1]^t$ .
- $\mathbf{h}$  and  $\mathbf{w}$  have equal length.

You will need results from the steady-state RLS analysis handout to answer the following questions.

- (a) Assuming  $\mathbf{Q} = 10^{-5}\mathbf{I}$ , calculate, in Matlab, the optimal RLS forgetting factor  $\lambda_{\text{opt}}$  and the corresponding (theoretical) level of RLS steady-state MSE, as well as the optimal LMS stepsize  $\mu_{\text{opt}}$  and the corresponding (theoretical) level of LMS steady-state MSE.
  - (b) Simulate both LMS and RLS adaptation of  $\mathbf{w}(n)$  using length-2500 sequences, the optimal  $\mu$  &  $\lambda$ ,  $\delta = 1$ , and initialization  $\mathbf{w}(n) = \mathbf{0}$ . Generate a plot that superimposes the true parameters  $\{\mathbf{h}(n)\}$  and adapted estimates  $\{\mathbf{w}(n)\}$  versus  $n$ . Generate a second plot that superimposes  $J(n)$  for LMS and RLS with the  $\lim_{n \rightarrow \infty} J(n)$  values calculated in (a), *all in a dB scale*. The results should look something like Fig. 1 and Fig. 2. Finally, average the experimentally-derived steady-state MSE (after convergence) for comparison with theoretical values. Which algorithm converges faster? Which tracks better?
  - (c) Repeat (a) & (b) for  $\mathbf{Q} = 10^{-5}\mathbf{R}$ . Which algorithm converges faster? Which tracks better?
  - (d) Repeat (a) & (b) for  $\mathbf{Q} = 10^{-5}\mathbf{R}^{-1}$ . Which algorithm converges faster? Which tracks better?
2. Prove the following inequalities (which were seen in the steady-state RLS analysis) assuming  $\mathbf{R}^{-1}$  exists:

- (a)  $\sqrt{\frac{\text{tr}(\mathbf{R})^2}{M \text{tr}(\mathbf{R}^2)}} \leq 1$
- (b)  $\frac{1}{M} \sqrt{\text{tr}(\mathbf{R}) \text{tr}(\mathbf{R}^{-1})} \geq 1$

Hint: Use the Cauchy-Schwarz inequality for vectors,  $|\mathbf{x}^H \mathbf{y}|^2 \leq \|\mathbf{x}\|^2 \|\mathbf{y}\|^2$ , and the fact that the trace of a matrix equals the sum of its eigenvalues.

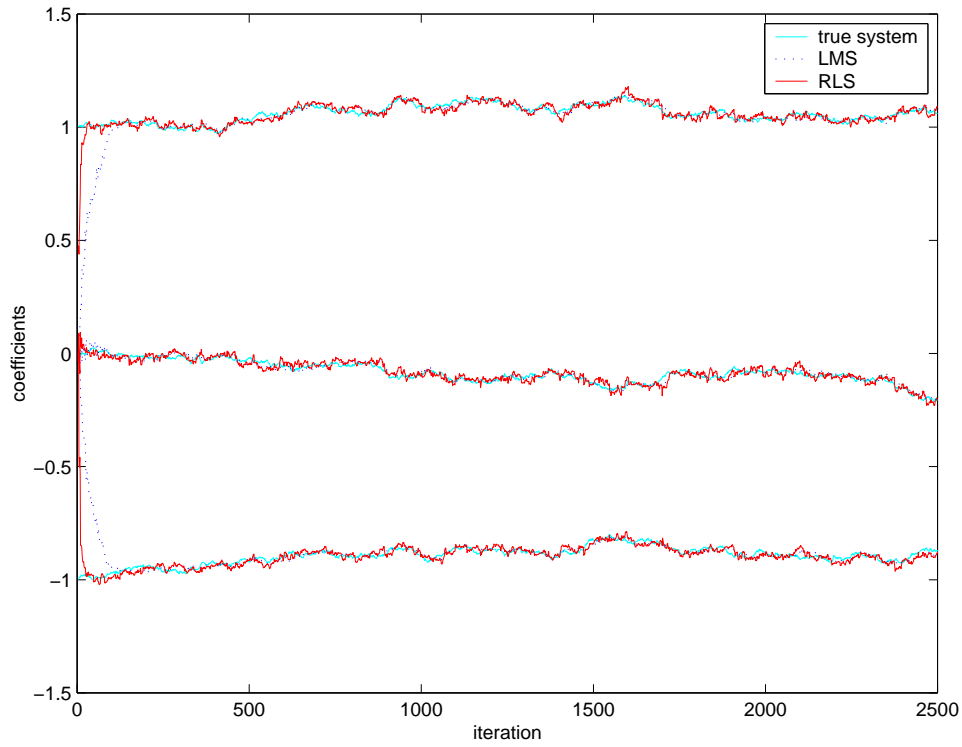


Figure 1: RLS/LMS time-varying system identification.

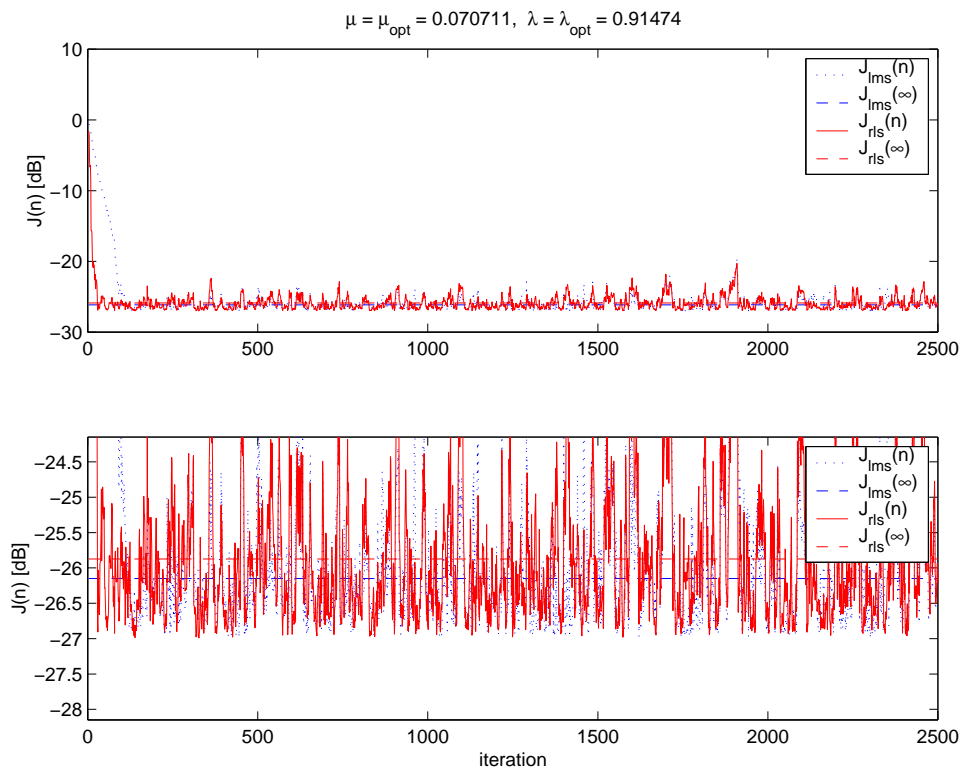


Figure 2: RLS/LMS time-varying system identification.