ECE 652 Adaptive Filtering Computer Simulation Project

In this project two adaptive filtering algorithms will be investigated: the LMS and the fast RLS algorithms. Both algorithms will be analyzed theoretically and by simulation. The topology for the LMS adaptive filter is shown in Figure 1. Note that the buffer numbers and star names are also shown. The topology for the fast RLS algorithm is shown in Figure 2. The *predftf* star is used for the fast RLS algorithm. It implements the Fast Transversal Filter (FTF) algorithm which for our purposes is numerically equivalent to the RLS algorithm (only in infinite precision). In both systems, the impulse response of a system is to be estimated. The input to the system is a white guassian random process. The system is simply a linear convolution of its input with a finite impulse response stored in a file(*imp.dat*). The *convolve* star performs a linear convolution. For your reference the

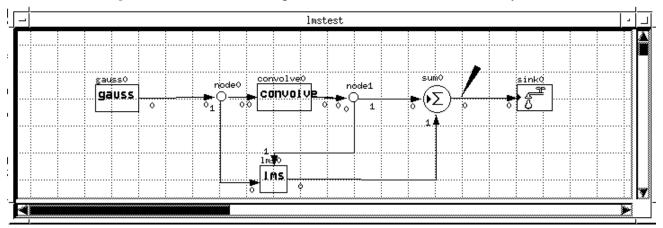


Figure 1. Topology for adaptive LMS filter

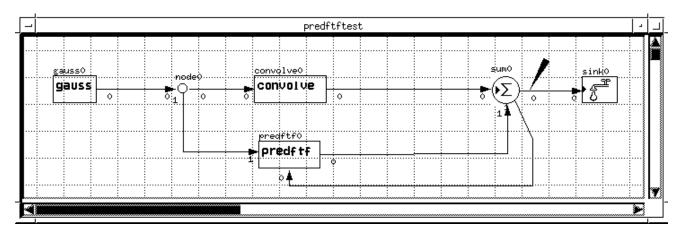


Figure 2. Topology for Fast Transversal Filter

text files of both topologies are included in the following pages. These files contain information about parameters. The default parameters should be okay but some parameters such as the number of points in the *gauss* star must be change. So refer to this information.

The *predftf* star requires a file called *prfile* which contains information about the number of input channels, outputs and the initial values of the filters. This file is included for your reference.

Construct both topologies and test them. The probe in both figures is a *plot* probe. You should observe very rapid convergence of the FTF algorithm.

Part I.

(a) In this part we will investigate the LMS algorithm. Connect a *stats* probe to the input to the *lms* star so we can measure the input power. Increase the number of points generated by the gauss star. We are interested in the convergence of the LMS algorithm with different gains. Try the following gains:

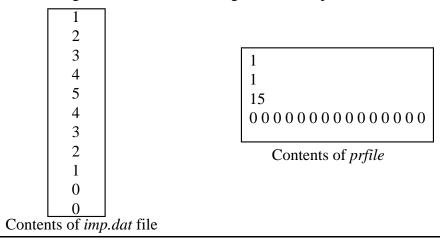
 μ = 1.0, 0.5, 0.1. For each case, plot the learning curve in dB's. Calculate the rate of convergence (dB's per 100 iterations) and compare with the simulation result.

(b) Insert an *addnoise* star between the adaptive filters (desired response) and the *convolve* star output for both systems. In Figure 1 insert it between convolve0 and node1. In Figure 2, insert it between convolve0 and sum0. We are interested in investigating excess noise due to the LMS and FTF algorithms. Set the noise variance in the *addnoise* star to 0.2. Obtain the steady state error for the LMS algorithm for μ =1.0, 0.5, 0.1 and λ =1.0, 0.99, and 0.9 for the FTF algorithm. Calculate the mean square excess error theoretically and compare with the simulation results.

Part II

In this part we will investigate the effect of signal correlation on the performance of the adaptive filters. Replace the *gauss* star with the *arprocess* star. This star produces an autoregressive process. By changing the AR coefficient we can increase signal correlation. Start out with a first order process. Let the AR coefficient be 0.99 initially. Observe the convergence of the LMS algorithm. You may want to insert a *gain* star between the *arprocess* star and the *convolve* and *lms* star (between the *arprocess* star and node0) in order to keep the variance at 1.0. Let the LMS gain be 1.0. Can you explain the slow convergence of the LMS algorithm compared to the FTF algorithm?

Predict the convergence rate of the LMS algorithm. Compare it to simulation results.



	П
# topology file: lmstest.t	# topology file: predftftest.t
#	#
# Title:	# Title:
# Author:	# Author:
	# Addior: # Date:
# Date:	
# Description:	# Description:
#	#
inform title	inform title
inform author	inform author
inform date	inform date
inform descrip	inform descrip
arg -1 (none)	arg -1 (none)
param int 500	param int 1000
param float 1	param float 1
param int 3759	param int 3759
param float 1	param float 1
param int 128	param int 128
star gauss0 gauss	star gauss0 gauss
star node0 node	star node0 node
param file imp.dat	param file imp.dat
param int 8	param int 8
star convolve0 convolve	star convolve0 convolve
star node1 node	param array 2 1 -1 star sum0 sum
param int 10	star sumo sum
param float 0.1	param file prfile
param int 0	param file prfileo
star lms0 lms	param float 1
Star miso mis	param float 1 param float 0.0001
param array 2 1 -1	param int 0
star sum0 sum	param int -1
star sumo sum	star predftf0 predftf
param int 500	
param int 0	param int 256
param file Imserror	param int 0
param file X	param file PLOT
param file Y	param file X
param int 1	param file Y
param int 1	param int 1
star plot0 plot	param int 1
star sink0 sink	star plot0 plot
	star sink0 sink
connect gauss 0 0 node 0 0	0.0 1.00
connect node0 0 convolve0 0	connect gauss 0 0 node 0 0
connect node0 1 lms0 0	connect node0 0 convolve0 0
connect convolve0 0 node1 0	connect node0 1 predftf0 1
connect node1 0 lms0 1	connect convolve0 0 sum0 0
connect node1 1 sum0 0	connect sum0 0 plot0 0
connect lms0 0 sum0 1	connect sum0 1 predftf0 0
connect sum0 0 plot0 0	connect predftf0 0 sum0 1
connect plot0 0 sink0 0	connect plot0 0 sink0 0
•	



