```
% EE7000-3 Adaptive Filter Theory
% Demonstration of Wiener filter, LMS filter, Steepest-descent algorithm
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clear;
clc;
N = 10000;
          % the length of the observation sequence
          % the filter length
M = 2;
               % white process as the AR excitation
v = randn(1, N);
a = poly(sign(randn(1,M))) * rand(1,M)); % coefficients of AR process
% The input sequence
u = filter(1, a, v);
% The desired response
d = v;
rf = xcorr(u,M,biased');
rv = rf(M+1:2*M+1);
% The correlation matrix of the input
R = toeplitz(rv);
pf = xcorr(d,u,M,biased');
% The cross-correlation vector between the input and the desired response
pv = pf(M+1:2*M+1).';
% The optimal tap weight vector for Wiener filter
wopt = inv(R) * pv;
% Selection of a stable step size mu
[V,D]=eig(R);
lambda max = max(diag(D));
mu = 0.9 * 2/lambda max;
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% The steepest descent learning
wsd = randn(M+1, 1);
                    % initial weight vector for steepest descent
total_iteration_number = 100; % total iteration number
for i = 1: total iteration number
 wsd = wsd + mu * (pv - R * wsd);
end;
% The LMS learning
% initial weight vector for LMS
wlms = randn(M+1, 1);
                    % initial input vector
uv = zeros(M+1,1);
mu = 0.1 * 2/lambda max;
                    % step size mu for LMS
for n = 1: N;
 uv(2:M+1) = uv(1:M);
 uv(1) = u(n);
 y = wlms'*uv;
 e = d(n) - y;
 wlms = wlms + mu * uv * conj(e);
end;
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