QAM 16 Viterbi

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Algorithm Initialization

```
clear all; close all; clc
numIter = 5; % The number of iterations of the simulation
nSym = 1000;
             % The number of symbols per packet
SNR_Vec = 0:2:16;
lenSNR = length(SNR_Vec);
M = 16;
               % The M-ary number, 2 corresponds to binary modulation
chan = 1;
                    % No channel
chan = [1 .2 .4]; % Somewhat invertible channel impulse response,
Moderate ISI
%chan = [0.227 0.460 0.688 0.460 0.227]';  % Not so invertible,
 severe ISI
% Create a vector to store the BER computed during each iteration
berVec = zeros(numIter, lenSNR);
eqlBerVec = zeros(numIter, lenSNR);
% Run the simulation numIter amount of times
```

Convolutional

```
%trel = poly2trellis([5 4],[23 35 0;0 5 13]); % Define trellis.
trel = poly2trellis(7,[171 133]); %7 is the constraint length.
171 173 define the system.
traceBack = 32;
codeRate = 1/2;
```

Below is the hard decision approach we tried. It worked worse than soft

iteration

```
tic
for i = 1:numIter
   bits = randi(2,[codeRate*nSym*log2(M), 1])-1;
   % If you increase the M-ary number, as you most likely will,
you'll need to
   % convert the bits to integers. See the BIN2DE function
   % For binary, our MSG signal is simply the bits
   msg = bi2de(reshape(bits,log2(M),[]).');
   encodedMSG = convenc(bits,trel);
                                              %Encoding using
trellis generated above.
   for j = 1:lenSNR % one iteration of the simulation at each SNR
Value
       noise_addition = 10*log10(log2(M)*codeRate);
 %To use soft decision, the variance of the noise is required for
demodulation.
       noise\_var = 10.^{(-(SNR\_Vec(j) + noise\_addition)/10)};
qammod(encodedMSG,M,'UnitAveragePower',true,'InputType','bit');  %
BPSK modulate the signal
qammod(bits,M,'UnitAveragePower',true,'InputType','bit');
       if isequal(chan,1)
           txChan = tx;
           txChan2 = tx2;
       elseif isa(chan, 'channel.rayleigh')
           reset(chan) % Draw a different channel each iteration
           txChan = filter(chan,tx);
       else
           txChan = filter(chan,1,tx); % Apply the channel.
           txChan2 = filter(chan,1,tx2);
       end
       % Convert from EbNo to SNR.
       % Note: Because No = 2*noiseVariance^2, we must add ~3 dB
       % to get SNR (because 10*log10(2) \sim= 3).
       txNoisy =
awgn(txChan,10*log10(log2(M))+SNR_Vec(j), 'measured'); % Add AWGN
       txNoisy2 =
awgn(txChan2,10*log10(log2(M))+SNR_Vec(j),'measured');
       rx = txNoisy;
       rx2 = txNoisy2;
```

```
% Again, if M was a larger number, I'd need to convert my
symbols
       % back to bits here.
       rxMSG =
qamdemod(rx2,M,'UnitAveragePower',true,'OutputType','bit');
      noEqlMSG =
qamdemod(rx,M,'UnitAveragePower',true,'OutputType','bit');
       %rxMSG = reshape(de2bi(rx,log2(M)),[],1);
       %Equilizer
       mu = 0.001; %step size
       trainlen = 200;
       n = 8; %number of weights
       const = qammod((0:1:M-1),M);
       trainSig=tx(1:trainlen);
       %LMS decision-feedback equalizer
      nfwd = 16;
       nfbk = 12;
       dfeLMS = dfe(nfwd,nfbk,lms(mu));
       dfeLMS.SigConst = const; % Set signal constellation.
       dfeLMS.ResetBeforeFiltering = 0;
       %RLS decision-feedback equalizer
       dfeRLS = dfe(nfwd,nfbk,rls(0.99,0.9));
       dfeRLS.SigConst = const; % Set signal constellation.
       dfeRLS.ResetBeforeFiltering = 0;
       %lms,linear
       %trainMSG = reshape(de2bi(tx(1:trainlen),log2(M)),[],1);
       linLMS = lineareq(n, lms(mu)); % Create an equalizer object.
       linLMS.SigConst = const; % Set signal constellation.
       linLMS.ResetBeforeFiltering = 0;
       %rls,linear
       linRLS = lineareq(n, rls(1,0.1)); % Create an equalizer
object.
       linRLS.SigConst = const; % Set signal constellation.
       linRLS.ResetBeforeFiltering = 0;
       %Again, linRLS is picked.
       %delay = (numRefTap-1)/eqobj.nSampPerSym;
       [y,eqlSig] = equalize(linRLS,rx,trainSig); % Equalize.
       eqlMSG =
qamdemod(eqlSig,M,'UnitAveragePower',true,'OutputType','bit');
       eqlMSG_soft = qamdemod(eqlSig,M,'OutputType','approxllr', ...
           'UnitAveragePower', true, 'NoiseVariance', noise var);
       %decodedMSG = dec(eqlMSG soft);
```

```
decodedMSG =
vitdec(eqlMSG_soft,trel,traceBack,'cont', 'unquant');

% Compute and store the BER for this iteration

rxMSG = bi2de(reshape(rxMSG,log2(M),[]).');
%decodedMSG = bi2de(reshape(decodedMSG,log2(M),[]).');
[zzz, berVec(i,j)] = biterr(msg, rxMSG); % We're interested
in the BER, which is the 2nd output of BITERR

[zzz, eqlBerVec(i,j)] = biterr(bits(trainlen+1:end-traceBack),
decodedMSG(trainlen+traceBack+1:end)); %For Conv Code
[zzz, eqlBerVec(i,j)] = biterr(bits(trainlen+1:end-traceBack),
decodedMSG(trainlen+traceBack+1:end));
```

Except for getting rid of the traning part, we also need to deal with delay between input and output caused by convolution, which is equal to traceback in this case because the rate is 1/2.

```
% The best result we got is 8-QAM, the bit rate is improved
while
% keeping the BER below e-6.

Bit_Rate = (log2(M)*nSym - trainlen) / 1000;

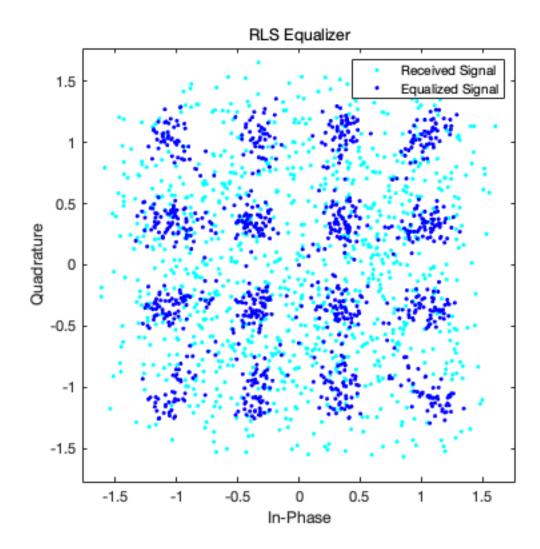
end % End SNR iteration
end % End numIter iteration
time = toc;
time

time =
7.4328
```

Plots

Played with M-ary QAM Scatter plot to see how equilizers work on

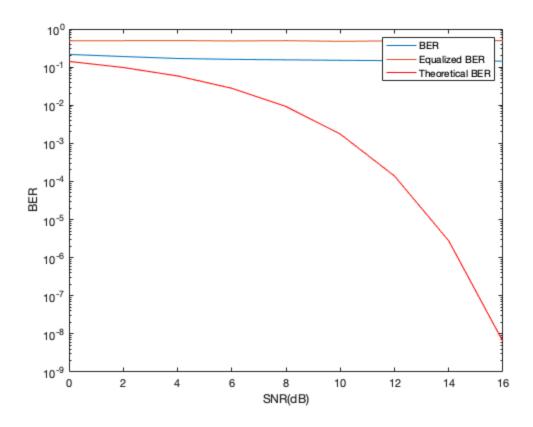
```
h = scatterplot(rx,1,0,'c.');
hold on
scatterplot(y,1,0,'b.',h)
legend('Received Signal','Equalized Signal')
title('RLS Equalizer')
hold off
%
```



Compute and plot the mean BER. The result at 12dB is 0. Which fits the requirement.

```
figure;
ber = mean(berVec,1);
eqlBer = mean(eqlBerVec,1);
semilogy(SNR_Vec, ber);
hold on
semilogy(SNR_Vec, eqlBer);

berTheory = berawgn(SNR_Vec,'qam',M);
hold on
semilogy(SNR_Vec,berTheory,'r')
xlabel("SNR(dB)")
ylabel("BER")
legend('BER', 'Equalized BER','Theoretical BER')
```



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