



Digital Communication Course

Project (2): Matched Filter

Presented to:

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Requirement 1:

- In this requirement we will build the Matched filter and the Correlator in Free Noise environment (before adding the Noise).
- First, we create the pulse signal p(t) with symbol duration Ts = 1 sec, then we sample it 5 times ($T_{sampling} = 200 \text{ ms}$).
- In addition to that we normalize the pulse signal to obtain unity energy ($E_p = 1$).
- We generate 10 random bits and from them we create the impulse train and to apply the convolution between the pulse signal p(t) and the impulse train we want to make each symbol in the impulse train take 5 samples not only one. Therefore, we apply up sampling to insert 4 zeros after each symbol in the impulse train.
- after applying the convolution between the pulse signal and the impulse train \rightarrow The resultant signal will be the output signal from the transmitter (Tx_{out}) and this output signal will pass through the channel and reach to the Receiver (Assuming free noise).

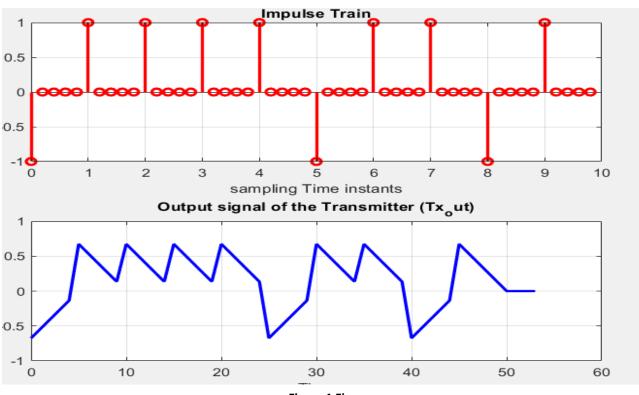


Figure 1 Fig

> Part A:

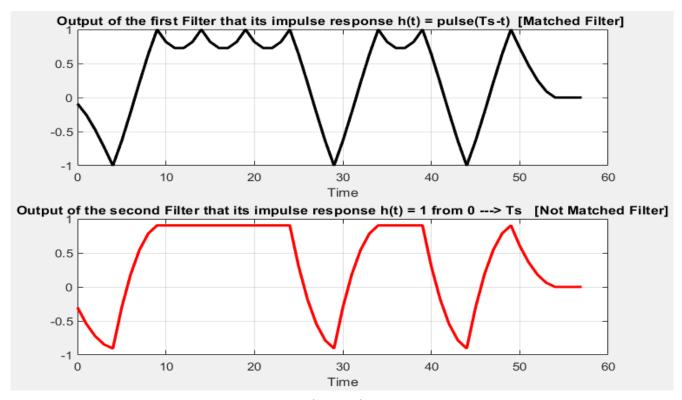


Figure 2 Fig

Comments:

- In the above figure (figure 1), we compare the output that result from the Matched filter with other filter (Not Matched Filter) that it's impulse response h(t) = 1 from (0 → Ts) but we must normalize the impulse response first to make fair comparison between both filters.
- To normalize any discrete signal, we bring its energy first from this equation $\Rightarrow E_S = \sum_{n=-\infty}^{n=\infty} (x[n])^2$, After that we bring $X[n]_{normalized} = \frac{X[n]}{\sqrt{E_S}}$.
- In this simulation the Random bits generated are [0 1 1 1 1 0 1 1 0 1] as shown in figure (1). Therefore, the Matched filter output curve verifies that, as there are **peaks** at 1 that represent **ones** and **minimum values at -1** that represent **zeros**.

> Part B:

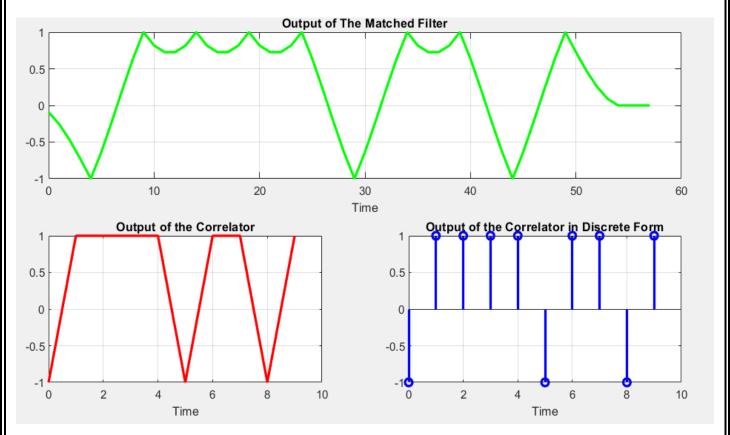


Figure 3 Fig

Comments:

- In the above figure (figure 3), we compare the output that result from the Matched filter with the output of the correlator where the concept of the correlator is that we take every 5 samples of the output signal produced from the transmitter (Tx_{out}) and multiply them by the pulse signal p(t) which have the same length =5 then we sum the product and store it in the correlator_{out} vector which has length equal to the number of bits (symbols) that are generated.
- As shown in figure (3) we draw the output of the correlator two times one continuous and the other is discrete where in discrete domain it is obvious that the output of the correlator represents the values of the random bits.

(1 represent one & -1 represent zero)

Requirement 2:

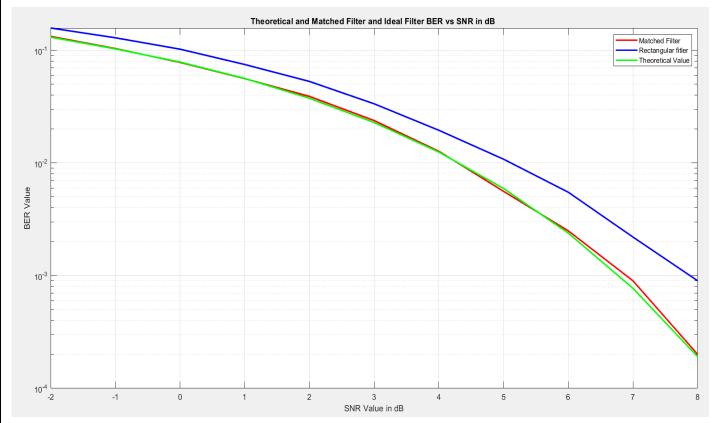


Figure 4 fig

Comments:

- For this requirement White gaussian noise is added to the system with different variances to obtain the required SNR.
- The Noisy output is sampled every five Samples (Ts), the sampled output is compared with the threshold, if it's below the threshold (zero) then the received bit is decoded as zero if its higher it's decoded as one.
- The decoded bits are then compared with the original bits and the bit error rate for each output is calculated.
- For the same energy of sent pulse the matched filter provides a lower bit error rate and better performance than using a rectangular filter as matched filter provides highest SNR value.
- Increasing Number of bits increases the simulation resolution as the generated gaussian noise mean approaches zero more and the variance approaches unity more, so the simulated results become even closer to the theoretical values.

> Requirement 3

- At Point A:

A. Roll-off = 0 & Delay = 2

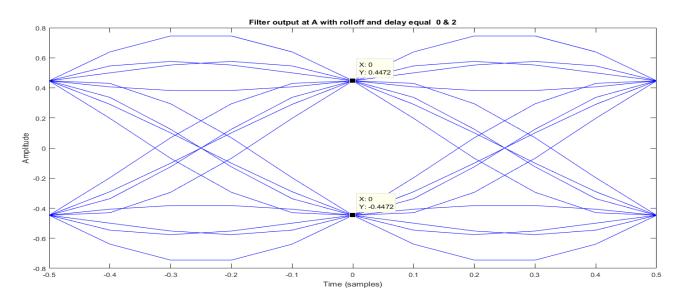


Figure 5: Eye diagram at point A with (r=0, d=2)

B. Roll-off = 0 & Delay = 8

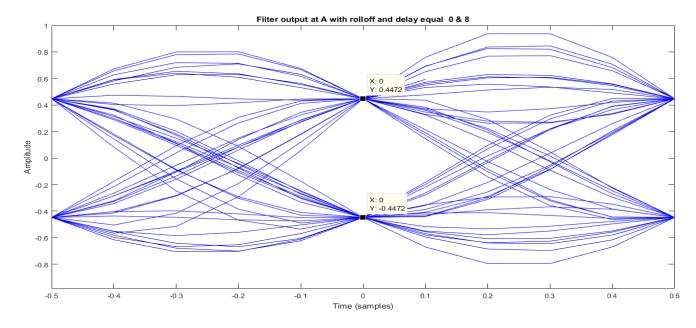


Figure 6: Eye diagram at point A with (r=0, d=8)

C. Roll-off = 1 & Delay = 2

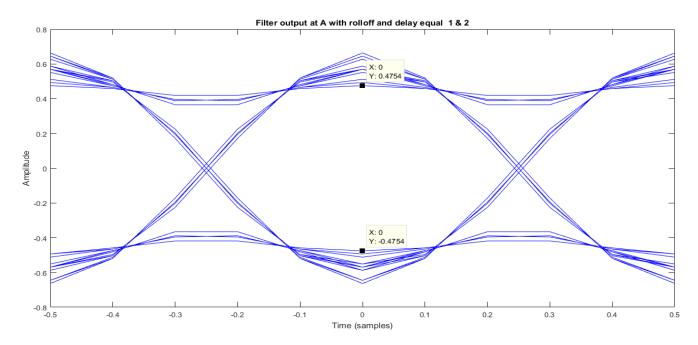


Figure 7: Eye diagram at point A with (r=1, d=2)

D. Roll-off = 1 & Delay = 8

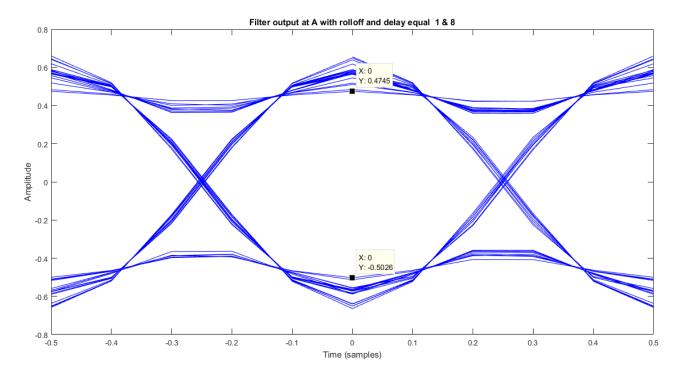


Figure 8: Eye diagram at point A with (r=1, d=8)

- At Point B:

A. Roll-off = 0 & Delay = 2

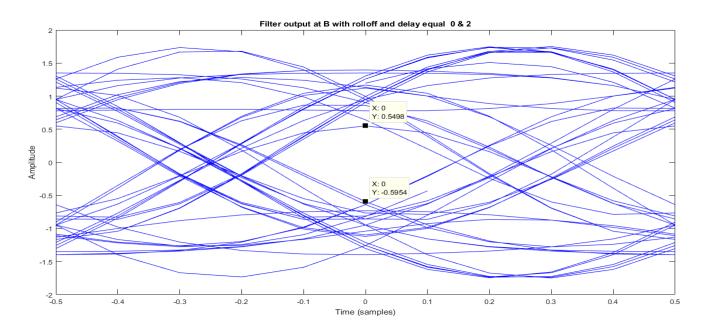


Figure 9: Eye diagram at point B with (r=0, d=2)

B. Roll-off = 0 & Delay = 8

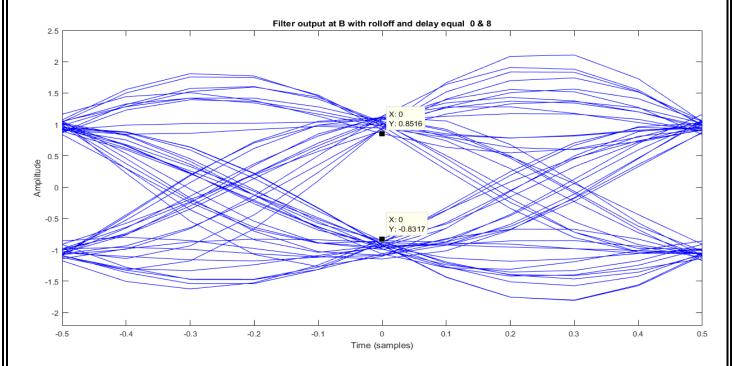


Figure 10: Eye diagram at point B with (r=0, d=8)

C. Roll-off = 1 & Delay = 2

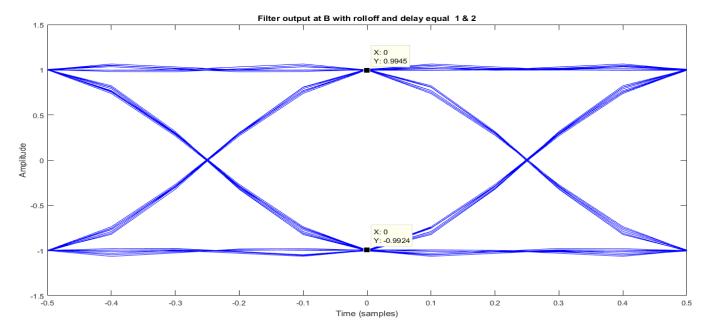


Figure 11: Eye diagram at point B with (r=1, d=2)

D. Roll-off = 1 & Delay = 8

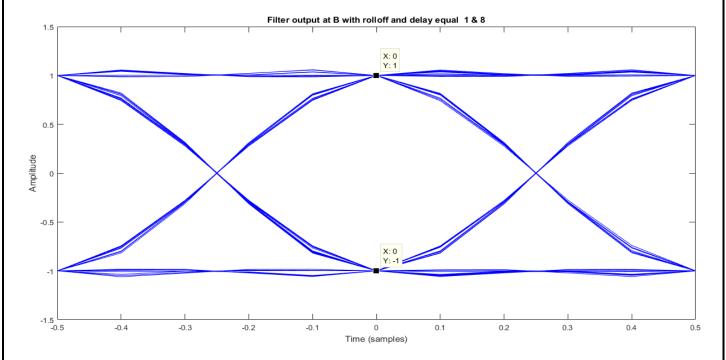


Figure 12: Eye diagram at point B with (r=1, d=8)

Comments:

- This part consists mainly of three parts after up sampling the random data ranging between [-1,1] we generated the SRRC filter coefficients using Rcosine filter giving it the following parameters:
 - 1) Sampling frequency (Fs)
 - 2) Fd = Fs *Sps where represents the no. of samples taken per symbol preferably taken a large number to increase the filter resolution.
 - 3) Shape of desired raised cosine filter: 'sqrt'
 - 4) Roll-off factor
 - 5) delay: represents the number of zeros seen in a single side of the filter describing the length of the finite window taken from the IIR of the filter.
- We generated the SRRC filter coefficients then we convoluted it with the up-sampled data to generate the transmitted data from A and plot its eye diagram then we repeated the convolution between the Tx signal and the same SRRC coefficients to get the received signal at B and plot the eye diagram.
- In the code: when plotting the eye diagram at either point A or B we didn't consider the entire length of the convolution output. Instead, we specifically extracted the primary intended part of the signal, which corresponds to the main lobe. Therefore, we began with the very first main lobe resulting from the convolution with the first bit and ended with the last one, extracting some side lobes from the start and the end of the convolution output.
- About the eye opening and the sampling time relation as the eye remains quite open for a larger period around the optimum sampling time (center of the eye opening). The system becomes more tolerant of sampling timing errors making the system more immune to synchronization errors.
- Observations:
 - 1) When the roll-off factor increases this means the bandwidth used increases so the signal shrinks in time domain decreasing the ISI and leading to larger eye opening (better sampling accuracy).
 - 2) Also, when the delay increases, the filter response expands in time making it more similar to the ideal response so, the eye opening slightly widens leading to a larger noise margin.

3) The effects of the roll-off factor and the delay parameters are much noticeable at the receiver (point B) than at the transmitter (point A) as at Tx data is convolved with the SRRC filter once before transmission while at Rx the Tx signal is received by a matching SRRC filter to maximize SNR resulting to a nearly raised cosine overall effect this could contribute to the difference of the observed results.

► Index:

```
clc
1 -
2 -
      clear
3 -
      close all;
4
5
      %% ------%%
6
7 -
                                 % it refer to the symbol duration Ts = 1 sec
8
      % generation of pulse p[n] but divided over sqrt(55) to normalize it's energy
9 -
     pulse = [5 4 3 2 1]/sqrt(55);
10 -
      Number of bits = 10;
                            % this variable represent the number of the random bits that are generated
      Data = randi([0 1] , 1 , Number of bits); % array contain the random data bits (0 & 1)
11 -
12
13 -
                        % it refer to the amplitude of the impulse train
      14 -
      impulse_train = upsample(PAM_signal,5);
                                             % we upsample the input signal (PAM_signal) by inserting 4 zeros (5-1) in each symbol duration Ts
15 -
16
17 -
    Tx out = conv(impulse train, pulse); % where Tx out refer to the signal at the output of the transmitter
      % where the length(Tx out) = length(impulse train) + length(pulse) - 1 [According to the discrete convolution]
18
19
20
      % For plotting, we will use the stem() function instead of the plot() function as stem() is specified for the discrete plotting
21 -
      figure;
22 -
      subplot(2,1,1);
30 -
      title('Output signal of the Transmitter (Tx out)');
31 -
      xlabel('Time');
      grid on;
32 -
33
34
      %-----% Methods of filtering the Tx_out signal (2 Methods)-----%
35
36
      % First Method is by take Tx out & apply convolution with impulse response of the Matched filter h[n] = Pulse[Ts - n] (Matched Filter)
37
38 -
      impulse response MF1 = fliplr(pulse);
                                            % Matched filter impulse response
39 -
      Rx out1 = conv(Tx out, impulse response MF1); % Rx out is the output signal of the matched filter
40
41
      % second Method is by take Tx out & apply convolution with impulse response of another filter h(t) = 1 from 0 --> Ts (NOT Matched Filter)
42
43 -
      impulse response MF2 = [1 1 1 1 1]/sqrt(5);
                                                % First step we must normalize the impulse response to obtain unity energy
44 -
      Rx out2 = conv(Tx out, impulse response MF2);
45
      % we make subplotting to compare between the two outputs of two different filters (Rx_out1,Rx_out2)
46
47
48 -
      figure;
49 -
      subplot(2,1,1);
50 - plot(0:length(Rx out1)-1, Rx out1, 'k', 'LineWidth', 2);
51 - title('Output of the first Filter that its impulse response h[n] = pulse[Ts-n]');
52 - xlabel('Time');
53 - grid on;
54
55 - subplot (2,1,2);
      plot(0:length(Rx out2)-1, Rx out2, 'r', 'LineWidth', 2);
56 -
57 -
      title('Output of the second Filter that its impulse response h[n] = 1 from 0 ---> Ts');
58 -
    xlabel('Time');
```

```
ELC3070 Project2
59 -
       grid on;
 60
 61 - figure;
 62 - subplot (2,1,1);
 63 - stem(0:length(Rx out1)-1 , Rx out1 , 'k', 'LineWidth', 2);
 64 - title('Output of The First Matched Filter in Discrete Form');
 65 - xlabel('Time');
 66 - grid on;
 67
 68 -
     subplot(2,1,2);
 69 -
      stem(0:length(Rx out2)-1, Rx out2, 'r', 'LineWidth', 2);
 70 - title('Output of the second Filter in Discrete Form');
 71 - xlabel('Time');
 72 - grid on;
 73
74
      % Build the correlator & compare it's output with the output of the Matched Filter
75
 76 -
     correlator_out = zeros(1,length(Number_of_bits)); % the output signal of the correlator is the same length as the number of random bits generated
 77
 78 -
     j = 1;
                     % j variable is used to increment the index of correlator out
 79 - \Box for i = 1 : 5 : length(Tx out)-5
             % Calculate the correlation for each segment (5 samples) of Tx out and update correlator out
81 -
             correlator out(1,j) = sum( pulse .* Tx out(1,i:i+4) );
82 -
             j = j + 1;
83 - end
84
85
      % plot the output of the correlator and compare it with the Matched Filter
86 - figure;
87 - subplot(2,2,[1 2]);
 88 - plot(0:length(Rx_out1)-1 , Rx_out1 , 'g', 'LineWidth', 2);
 89 -
      title('Output of The Matched Filter');
 90 -
     xlabel('Time');
 91 - grid on;
 92
 93 - subplot(2,2,3);
      plot(0:length(correlator out)-1, correlator out , 'r', 'LineWidth', 2);
 94 -
 95 - title('Output of the Correlator');
 96 - xlabel('Time');
 97 - grid on;
 98
 99 -
     subplot (2, 2, 4);
100 -
     stem(0:length(correlator_out)-1, correlator_out , 'b', 'LineWidth', 2);
101 - title('Output of the Correlator in Discrete Form');
102 - xlabel('Time');
103 - grid on;
104
105
       106
107
       % a):
108 - Number of bits = 10000;
                                   % this variable represent the number of the random bits that are generated
109 - pulse = [5 4 3 2 1]/sqrt(55);
110 -
      Data = randi([0 1] , 1 , Number of bits); % array contain the random data bits (0 & 1)
111 - Amplitude = 1;
                        % it refer to the amplitude of the impulse train
112 - PAM_signal = (2 * Data - 1) * Amplitude; % This is the PAM sampling which represent (+Amplitude or -Amplitude) every Ts = 1 sec
113 - impulse train = upsample(PAM signal,5); % we upsample the input signal (PAM signal) by inserting 4 zeros (5-1) in each symbol duration Ts
114 - Tx out = conv(impulse train, pulse); % where Tx out refer to the signal at the output of the transmitter
       115 -
116 -
     impulse_response_MF2 = [5 5 5 5 5]/sqrt(125); % First step we must normalize the impulse response to obtain unity energy
```

```
117
118
        Gaussian Noise = randn( 1,length(Tx_out) ); % generate gaussian white noise of zero mean and unity variance of length equal to that of transmitted signal
119 -
120
121
122 -
        Noise Variance = [1.585 1.25 1 0.794 0.631 0.501 0.3981 0.316 0.2511 0.1995 0.15845] ; % Variable to control gaussian noise variance value
123
124
125 -
        BER Simulated = zeros(2,length(Noise Variance));
126 - BER Theoretical = zeros(1,length(Noise Variance));
127 -
      for j= 1 :length(Noise Variance)
128 -
          Noisy Tx out = Tx out + ( Gaussian Noise * sqrt( Noise Variance(j)/2 )) ;
129 -
                                                            % Noisy Rx out1 is the output signal of the matched filter with noise added
            Rx out1 = conv(Tx out, impulse response MF1);
130 -
                                                                       % Noisy Rx out1 is the output signal of the matched filter with noise added
          Noisy Rx out1 = conv(Noisy Tx out, impulse response MF1);
131 -
          Noisy Rx out2 = conv(Noisy Tx out, impulse response MF2);
132 -
          Sampled Output1 = zeros(1, Number of bits); %Samples Noisy Rx out1
133 -
          Sampled_Output2 = zeros(1, Number_of bits); %Samples Noisy_Rx_out2
134
135 -
          Sampled Output Index = 1;
136 -
          for i= 5:5: (Number of bits*5)
137 -
               Sampled Output1(Sampled Output Index ) = Noisy Rx out1(i);
138 -
                Sampled Output2(Sampled Output Index ) = Noisy Rx out2(i);
139 -
               Sampled Output Index = Sampled Output Index + 1;
          end
140 -
141 -
          Number Of Errors = zeros(1,2);
142
           %Calculation Of BER
143 -
          for i = 1:Number of bits
144 -
              if(PAM signal(i) == 1 && Sampled Output1(i) <=0)
145 -
                   Number Of Errors(1) = Number Of Errors(1) + 1;
146 -
                elseif(PAM signal(i) == -1 && Sampled Output1(i) >0 )
147 -
                    Number Of Errors(1) = Number Of Errors(1) + 1;
148 -
                end
149 -
                if (PAM signal(i) == 1 && Sampled Output2(i) <=0)
                    Number Of Errors(2) = Number Of Errors(2) + 1;
150 -
151 -
                elseif(PAM signal(i) == -1 && Sampled Output2(i) >0 )
                    Number Of Errors(2) = Number Of Errors(2) + 1;
152 -
153 -
154 -
155 -
            BER Simulated(1,j) = Number Of Errors(1) / Number of bits ;
            BER Simulated(2,j) = Number Of Errors(2) / Number of bits ;
156 -
157 -
            BER Theoretical(1,j) = 0.5 * erfc(sqrt(1/Noise Variance(j)));
158 -
       end
159
160
        %plotting BER vs SNR
161 -
        figure;
162 -
        SNR DB=-2:length(Noise Variance)-3;
163 -
        semilogy(SNR DB, BER Simulated(1,:) ,'r', 'linewidth', 2)
164 -
165 -
        semilogy(SNR DB, BER Simulated(2,:),'b', 'linewidth', 2 )
166 -
        hold on;
167 -
        semilogy(SNR DB, BER Theoretical , 'g', 'linewidth', 2)
        legend('Matched Filter', 'Ideal fitler', 'Theoretical Value')
168 -
        grid on
169 -
170 -
        xlabel("SNR Value in dB")
171 - ylabel("BER Value")
172 -
        title("Theoretical and Matched Filter and Ideal Filter BER vs SNR in dB ")
173
```

```
174
        %% %%************ ISI and Raised cosine filter part***********
175
        % Parameters
176 -
        numBits = 100;
                               % Number of bits
177 -
        rolloff = [0,0,1,1];
                                      % Roll-off factor
178 -
                                   %Delay introduced to filter coefficient
        delay=[2,8,2,8];
179 -
                     *Samples taken per symbol preferred large to increase resolution
        sps =5;
180 -
        Fs = 1;
                   %Sampling freq where each bit is represented by a symbol
181
        % Generate random bits
182
        dataTx = randi([0,1],1,numBits)*2-1; % Randomize bits to -1 or 1
183 -
184 -
        dataTx upsampled = upsample(dataTx,sps);
185
186 - | for i=1:4
187
188
            % Transmitter
189
                %Generating the filter coefficients
190 -
                    srrc = rcosine(Fs ,sps*Fs, 'sqrt' , rolloff(i), delay(i) );
191
192
               % Apply SRRC filter to the upsampled data by convolution
193 -
                    txSig = conv(dataTx upsampled, srrc);
194
195
                %Calculating the start of the main intended output
196 -
                start = length(srrc)+1;
197
             % Eyediagram at A(transmitter)
198
199 -
                eyediagram(txSig(start-1:end-start-sps-1),2*sps);
200 -
                xlabel('Time (samples)');
201 -
                 title(['Filter output at A with rolloff and delay equal
202
                     num2str(rolloff(i)) ' & ' num2str(delay(i))]);
203
204
            % Receiver
205
                 %Using the same filter srrc as the transmitter
206 -
                     RxSig = conv(txSig,srrc);
207
208
                 % Eyediagram at B(reciever)
209 -
                    eyediagram(RxSig(start-1:end-start-sps-1),2*sps);
210 -
                    xlabel('Time (samples)');
211 -
                    title(['Filter output at B with rolloff and delay equal
212
                        num2str(rolloff(i)) ' & ' num2str(delay(i))]);
213
214 -
        end
```