
Table of Contents

Exercise 11.3	1
Exercise 11.4	5
Exercise 12.1 a	10
Exercise 12.1 b Testing 2-PAM	11
Exercise 12.1 b Testing 6-PAM	13
Exercise 12.2	15
Exercise 12.3	16
Exercise 13.1	17
Exercise 13.2 a	24
Exercise 13.2b	25
Exercise 13.2 c	36
Exercise 13.2 d	37

Exercise 11.3

```
alphabet = [2, 3, 5];
variance = [1, 5, 35/3];
for iter = 1:3
    figure() % used to plot figure eyediag3
    N=1000; m=pam(N,alphabet(iter),variance(iter)); % random
    +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    ps=ones(1,M); % square pulse width M
    x=filter(ps,1,mup); % convolve pulse shape with mup
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*c+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,1), plot(q)
    title(['Eye diagram for rectangular pulse shape using
    ',num2str(alphabet(iter))])

    N=1000; m=pam(N,alphabet(iter),variance(iter)); % random
    +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    ps=hamming(M); % square pulse width M
    x=filter(ps,1,mup); % convolve pulse shape with mup
    %x=x+0.15*randn(size(x));
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*c+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,2), plot(q)
    title(['Eye diagram for hamming pulse shape using
    ',num2str(alphabet(iter))])
```

```

    N=1000; m=pam(N,alphabet(iter),variance(iter));           % random
    +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    L=10; ps=srrc(L,0,M,50);
    ps=ps/max(ps); % sinc pulse shape L symbols wide
    x=filter(ps,1,mup); % convolve pulse shape with mup
    %x=x+0.15*randn(size(x));
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*(c-3)+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c-3); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,3), plot(q)
    axis([0,100,alphabet(iter) *-1,alphabet(iter)])
    title(['Eye diagram for sinc pulse shape using
    ',num2str(alphabet(iter))])
end

```

```

c =

    200

```

```

c =

    200

```

```

c =

    200

```

```

c =

    200

```

```

c =

    200

```

```

c =

    200

```

```

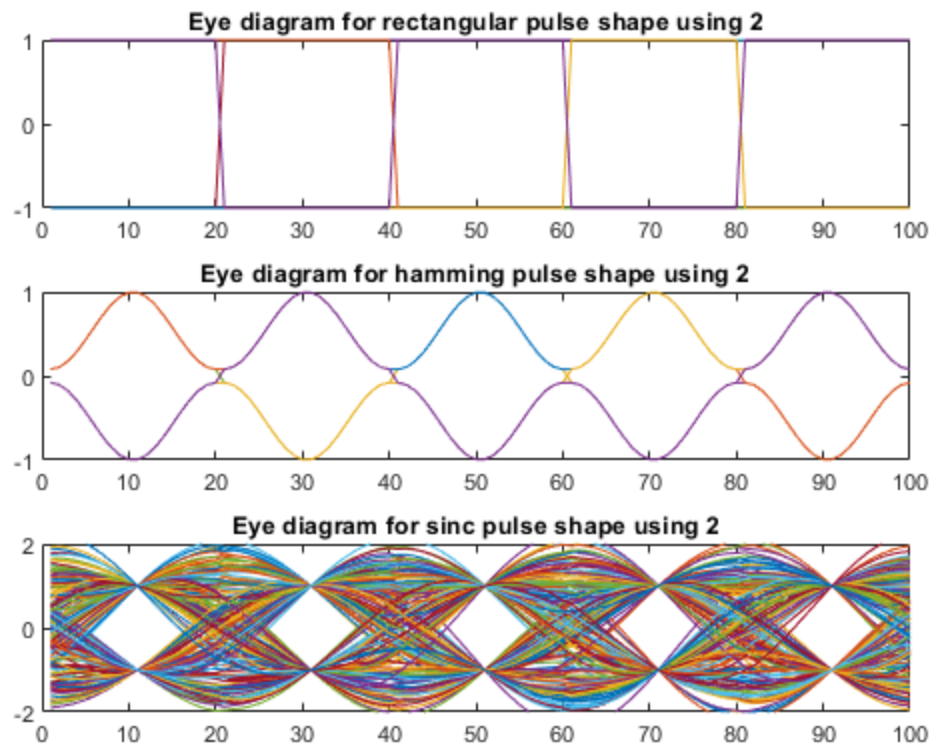
c =

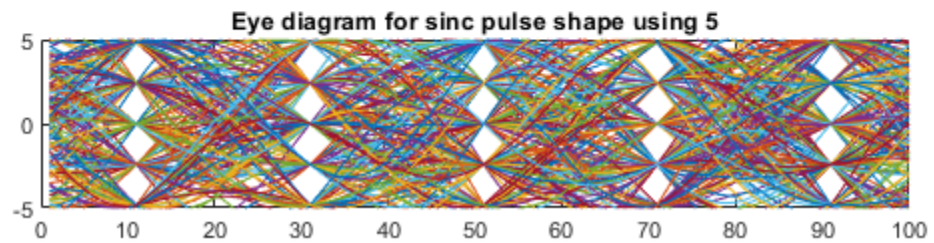
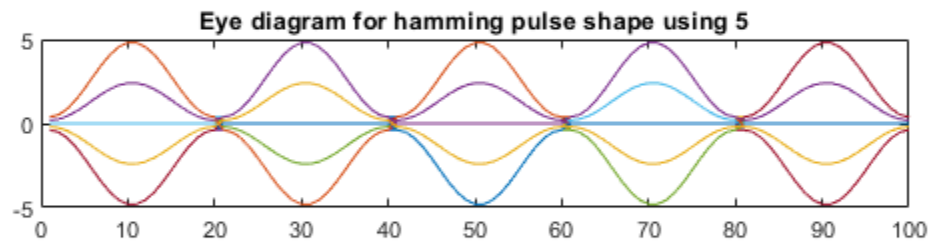
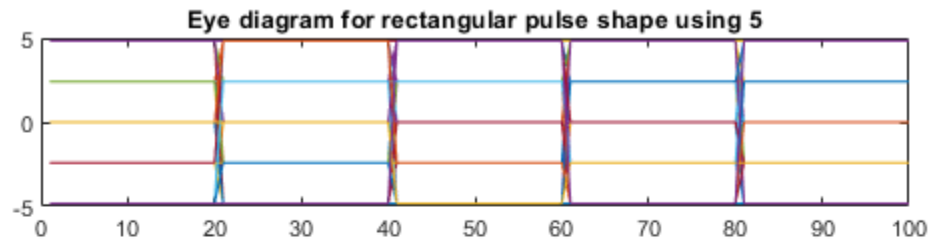
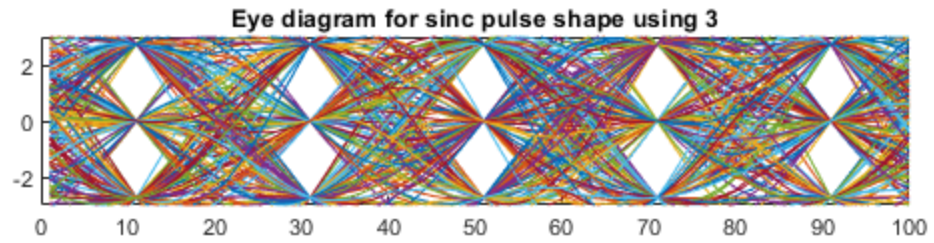
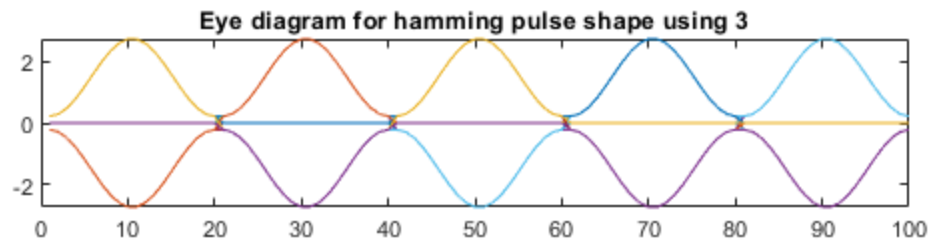
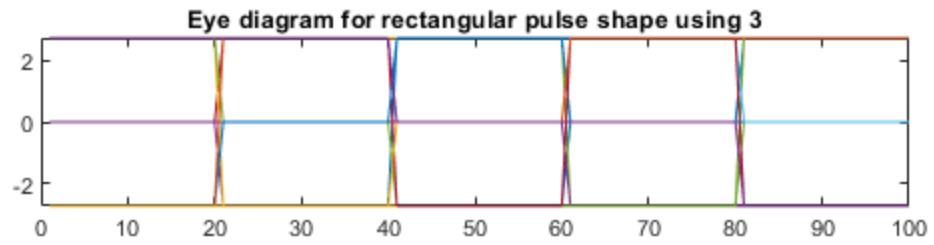
    200

```

$C =$
200

$C =$
200





Exercise 11.4

```
random_values = [0.075,0.15, 0.35 , 0.45, 0.55];
for iter = 1:5
    figure % used to plot figure eyediag3
    N=1000; m=pam(N,2,1); % random +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    ps=ones(1,M); % square pulse width M
    x=filter(ps,1,mup); % convolve pulse shape with mup
    x=x+random_values(iter)*randn(size(x));
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*c+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,1), plot(q)
    title(['Eye diagram for rectangular pulse shape using
    ',num2str(random_values(iter))])

    N=1000; m=pam(N,2,1); % random +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    ps=hamming(M); % square pulse width M
    x=filter(ps,1,mup); % convolve pulse shape with mup
    x=x+random_values(iter)*randn(size(x));
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*c+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,2), plot(q)
    title(['Eye diagram for hamming pulse shape using
    ',num2str(random_values(iter))])

    N=1000; m=pam(N,2,1); % random +/-1 signal of length N
    M=20; mup=zeros(1,N*M); mup(1:M:N*M)=m; % oversampling by factor
    of M
    L=10; ps=srrc(L,0,M,50);
    ps=ps/max(ps); % sinc pulse shape L symbols wide
    x=filter(ps,1,mup); % convolve pulse shape with mup
    x=x+random_values(iter)*randn(size(x));
    neye=5;
    c=floor(length(x)/(neye*M))
    xp=x(N*M-neye*M*(c-3)+1:N*M); % dont plot transients at start
    q=reshape(xp,neye*M,c-3); % plot in clusters of size
    5*Mt=(1:198)/50+1;
    subplot(3,1,3), plot(q)
    axis([0,100,random_values(iter) *-1,random_values(iter)])
    title(['Eye diagram for sinc pulse shape using
    ',num2str(random_values(iter))])
end
```

```
% ANSWER
% It appears that the largest value of v is around 0.5, anything after
  that
% and it appears that the "eye" is no longer visible.
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

```
C =

    200
```

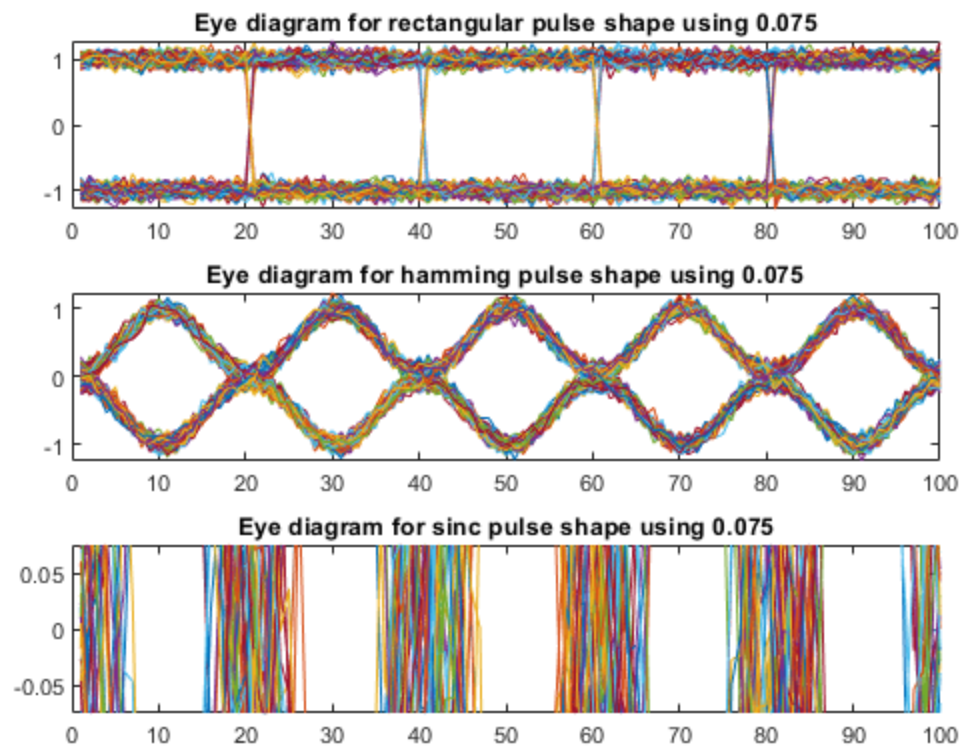
$C =$
200

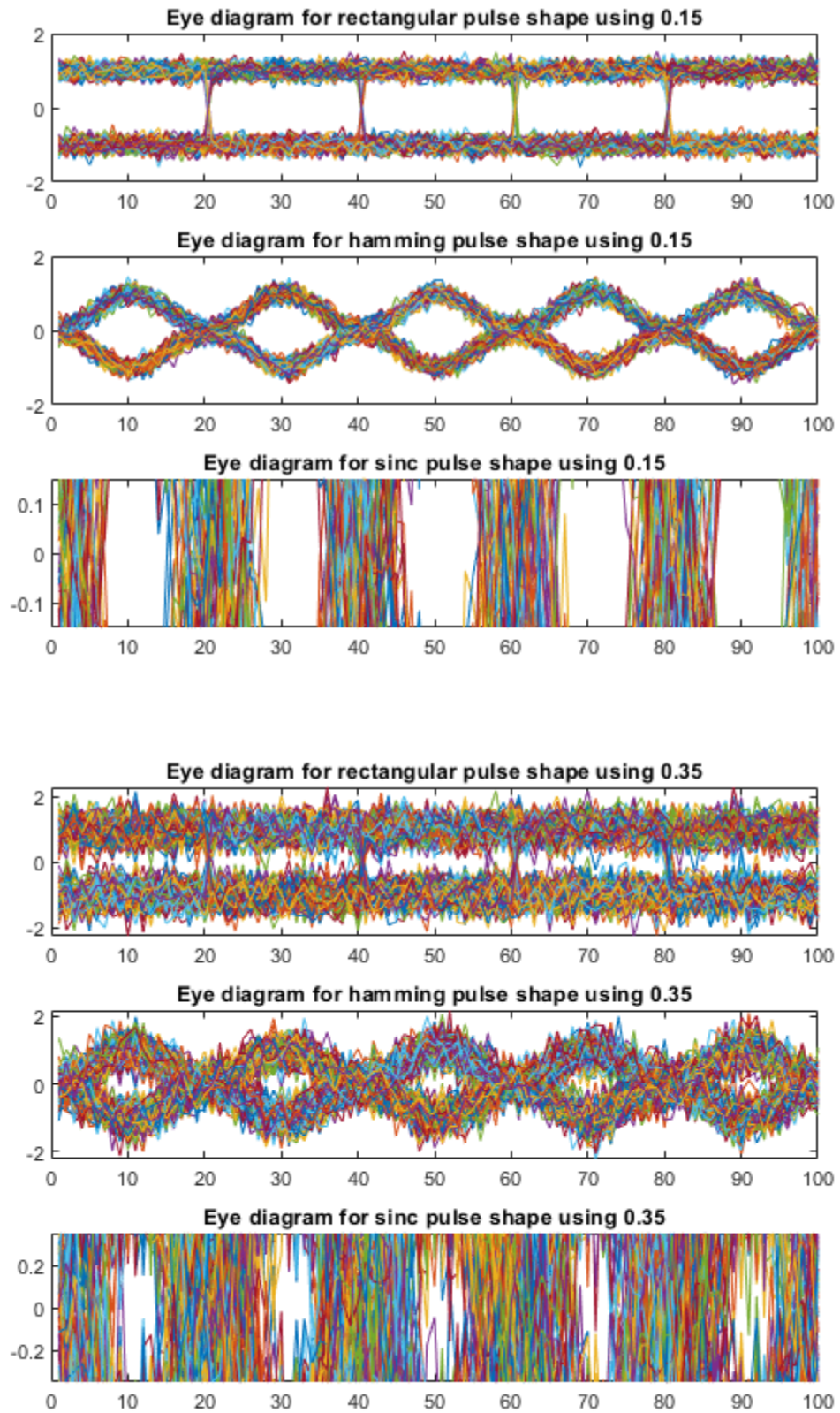
$C =$
200

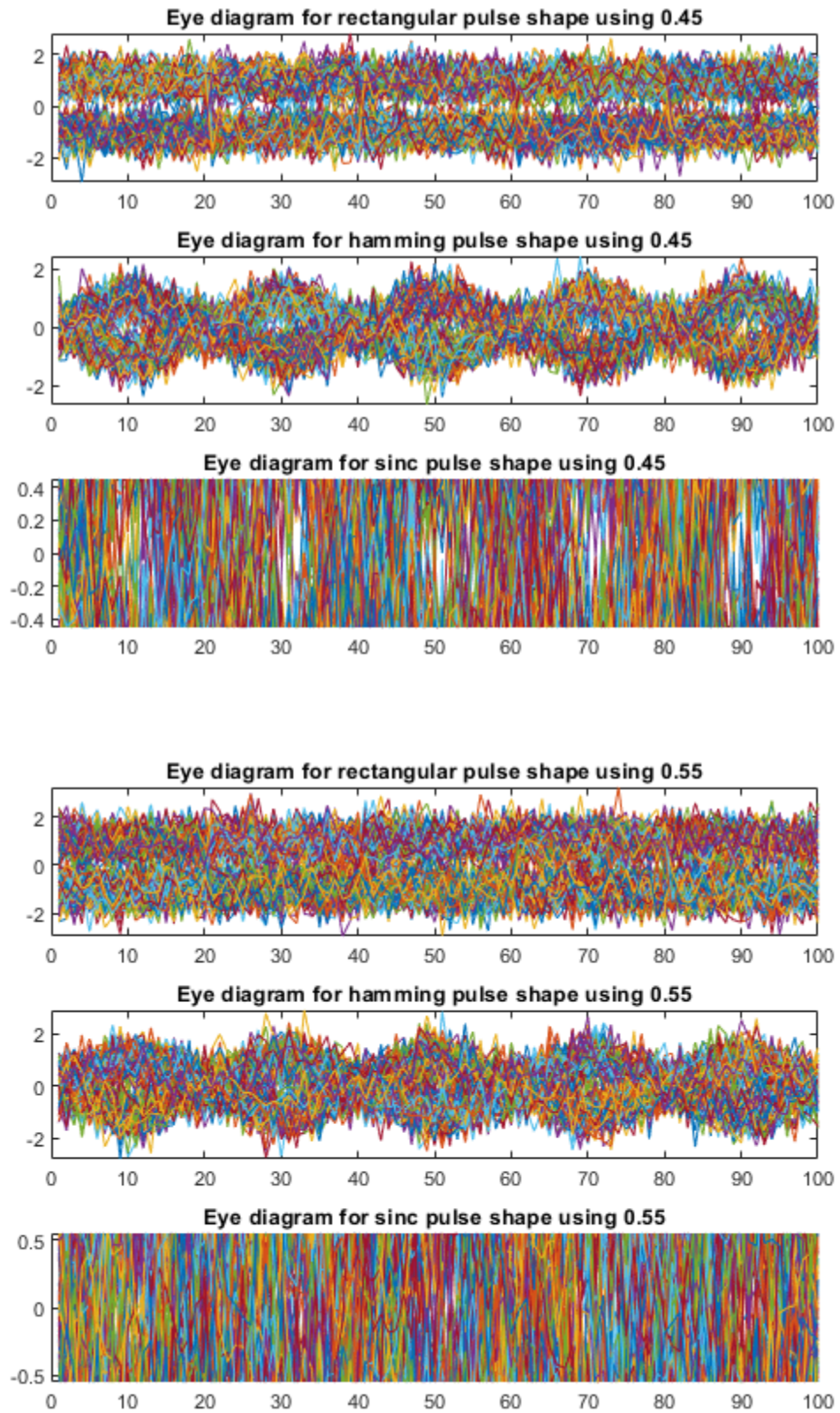
$C =$
200

$C =$
200

$C =$
200







Exercise 12.1 a

clockrecDD.m: clock recovery minimizing 4-PAM cluster variance to minimize $J(\tau) = (Q(x(kT/M+\tau)) - x(kT/M+\tau))^2$

```
% prepare transmitted signal
n=10000; % number of data points
m=2; % oversampling factor
beta=0.3; % rolloff parameter for srcc
l=50; % 1/2 length of pulse shape (in
symbols)
chan=[1]; % T/m "channel"
toffset=-0.3; % initial timing offset
pulshap=srcc(l,beta,m,toffset); % srcc pulse shape with timing offset
s=pam(n,4,5); % random data sequence with var=5
sup=zeros(1,n*m); % upsample the data by placing...
sup(1:m:n*m)=s; % ... m-1 zeros between each data
point
hh=conv(pulshap,chan); % ... and pulse shape
r=conv(hh,sup); % ... to get received signal
matchfilt=srcc(l,beta,m,0); % matched filter = srcc pulse shape
x=conv(r,matchfilt); % convolve signal with matched filter

% clock recovery algorithm
tnow=l*m+1; tau=0; xs=zeros(1,n); % initialize variables
tausave=zeros(1,n); tausave(1)=tau; i=0;
mu_values = [0.005, 0.01, 0.02, 0.05]
mu=2; % algorithm stepsize
delta=0.1; % time for derivative
while tnow<length(x)-2*l*m % run iteration
    i=i+1;
    xs(i)=interp sinc(x,tnow+tau,l); % interp value at tnow+tau
    x_deltap=interp sinc(x,tnow+tau+delta,l); % value to right
    x_deltam=interp sinc(x,tnow+tau-delta,l); % value to left
    dx=x_deltap-x_deltam; % numerical derivative
    qx=quantalph(xs(i),[-3,-1,1,3]); % quantize to alphabet
    tau=tau+mu*dx*(qx-xs(i)); % alg update: DD
    tnow=tnow+m; tausave(i)=tau; % save for plotting
end

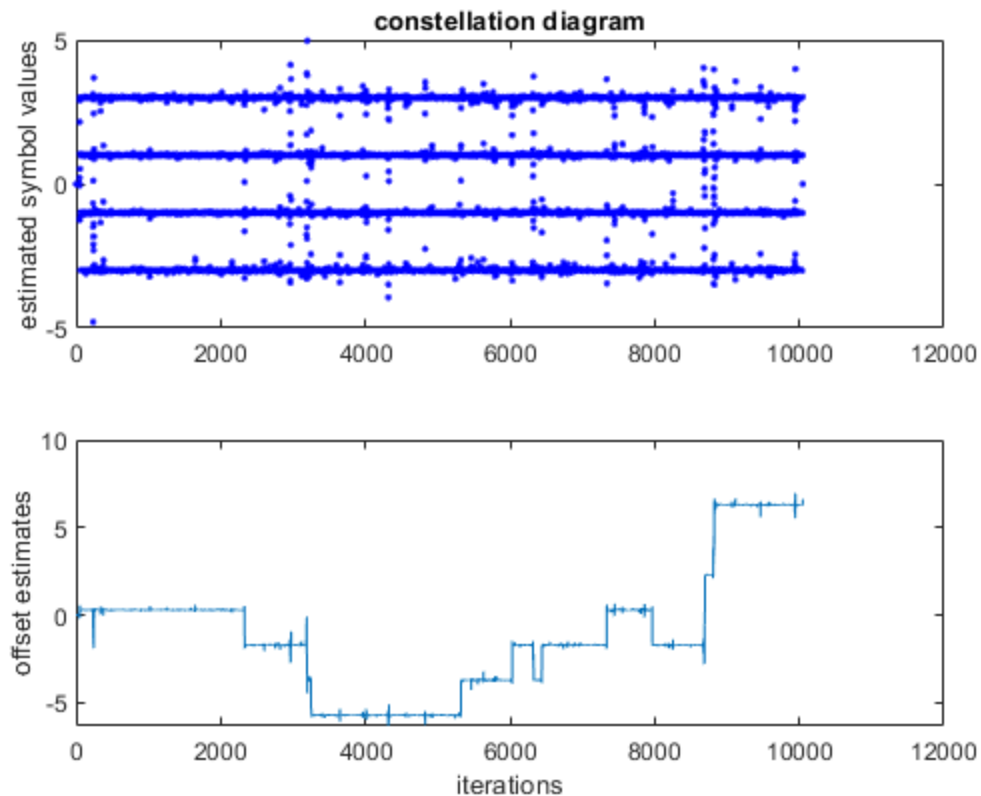
% plot results
figure();
subplot(2,1,1), plot(xs(1:i-2),'b.') % plot constellation
    diagram
    title('constellation diagram');
    ylabel('estimated symbol values')
    subplot(2,1,2), plot(tausave(1:i-2)) % plot trajectory of tau
    ylabel('offset estimates'), xlabel('iterations')

%ANSWER
% It appears that mu seems to affect the amount of noise present in
the
```

```
% signal, around 0.2 noise can start to appear visible. However if mu
goes
% as large as 2 it no longer converges to the correct value of the
offset.
```

```
mu_values =
```

```
0.0050    0.0100    0.0200    0.0500
```



Exercise 12.1 b Testing 2-PAM

prepare transmitted signal

```
n=10000; % number of data points
m=2; % oversampling factor
beta=0.3; % rolloff parameter for srroc
l=50; % 1/2 length of pulse shape (in
symbols)
chan=[1]; % T/m "channel"
toffset=-0.3; % initial timing offset
pulshap=srrc(l,beta,m,toffset); % srroc pulse shape with timing offset
s=pam(n,2,1); % random data sequence with var=5
sup=zeros(1,n*m); % upsample the data by placing...
```

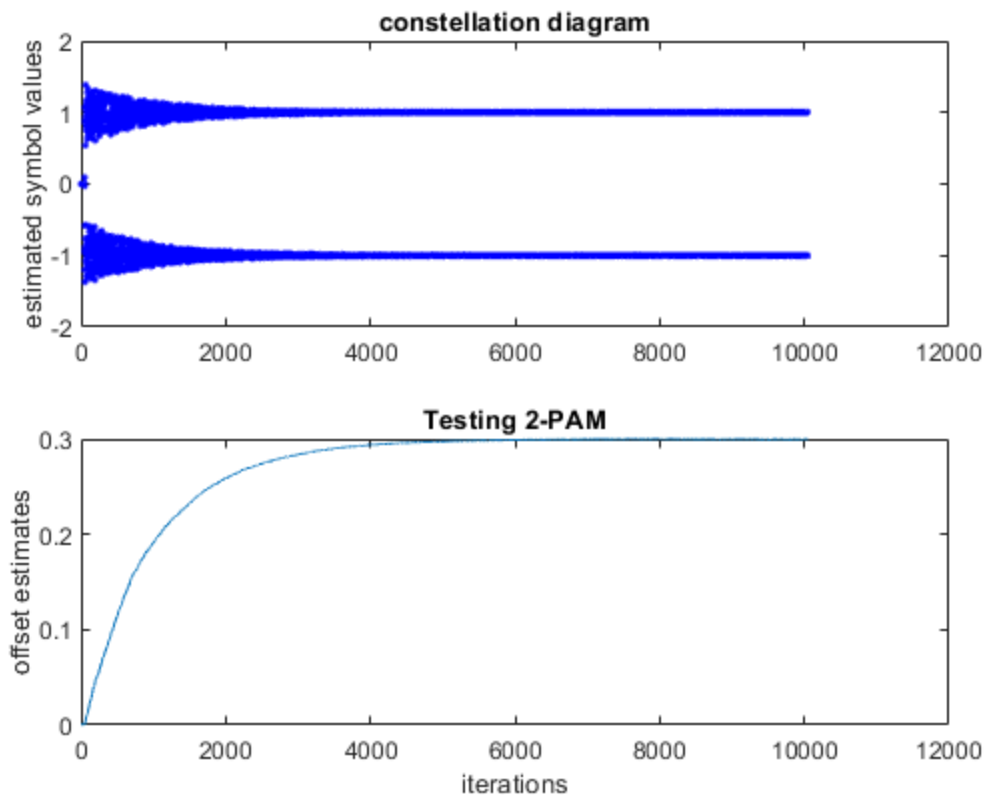
```

sup(1:m:n*m)=s; % ... m-1 zeros between each data
point
hh=conv(pulshap,chan); % ... and pulse shape
r=conv(hh,sup); % ... to get received signal
matchfilt=srrc(1,beta,m,0); % matched filter = srrc pulse shape
x=conv(r,matchfilt); % convolve signal with matched filter

% clock recovery algorithm
tnow=1*m+1; tau=0; xs=zeros(1,n); % initialize variables
tausave=zeros(1,n); tausave(1)=tau; i=0;
mu=0.01; % algorithm stepsize
delta=0.1; % time for derivative
while tnow<length(x)-2*1*m % run iteration
    i=i+1;
    xs(i)=interpinc(x,tnow+tau,1); % interp value at tnow+tau
    x_deltap=interpinc(x,tnow+tau+delta,1); % value to right
    x_deltam=interpinc(x,tnow+tau-delta,1); % value to left
    dx=x_deltap-x_deltam; % numerical derivative
    qx=quantalph(xs(i),[-3,-1,1,3]); % quantize to alphabet
    tau=tau+mu*dx*(qx-xs(i)); % alg update: DD
    tnow=tnow+m; tausave(i)=tau; % save for plotting
end

% plot results
figure();
subplot(2,1,1), plot(xs(1:i-2),'b.') % plot constellation
    diagram
title('constellation diagram');
ylabel('estimated symbol values')
subplot(2,1,2), plot(tausave(1:i-2)) % plot trajectory of tau
ylabel('offset estimates'), xlabel('iterations')
title('Testing 2-PAM');

```



Exercise 12.1 b Testing 6-PAM

prepare transmitted signal

```

n=10000; % number of data points
m=2; % oversampling factor
beta=0.3; % rolloff parameter for srroc
l=50; % 1/2 length of pulse shape (in
symbols)
chan=[1]; % T/m "channel"
toffset=-0.3; % initial timing offset
pulshap=srroc(l,beta,m,toffset); % srroc pulse shape with timing offset
s=pam(n,6,11); % random data sequence with var=5
sup=zeros(1,n*m); % upsample the data by placing...
sup(1:m:n*m)=s; % ... m-1 zeros between each data
point
hh=conv(pulshap,chan); % ... and pulse shape
r=conv(hh,sup); % ... to get received signal
matchfilt=srroc(l,beta,m,0); % matched filter = srroc pulse shape
x=conv(r,matchfilt); % convolve signal with matched filter

% clock recovery algorithm
tnow=l*m+1; tau=0; xs=zeros(1,n); % initialize variables
tausave=zeros(1,n); tausave(1)=tau; i=0;
mu=0.01; % algorithm stepsize
delta=0.1; % time for derivative

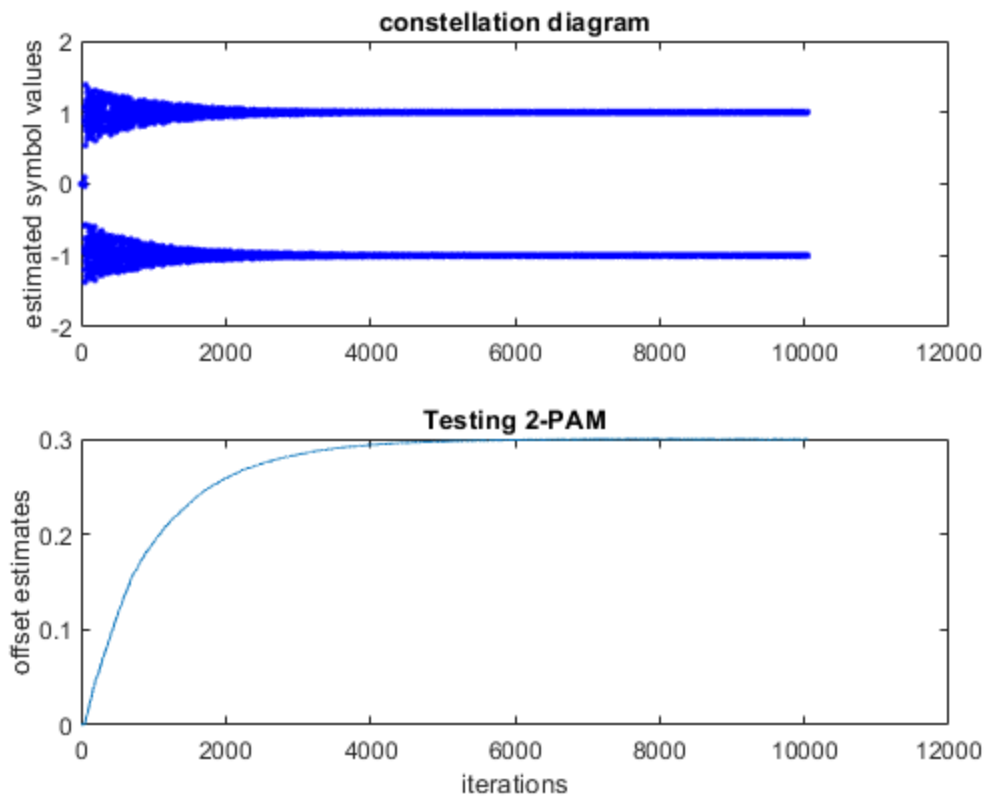
```

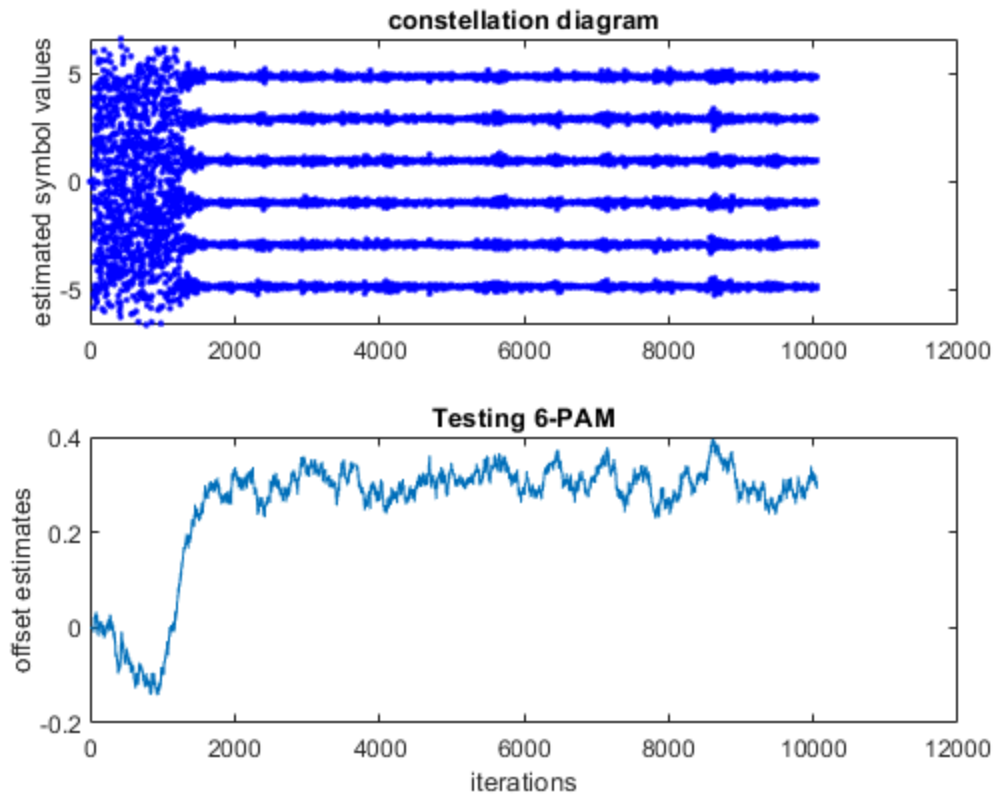
```

while tnow<length(x)-2*1*m           % run iteration
    i=i+1;
    xs(i)=interpsinc(x,tnow+tau,1);    % interp value at tnow+tau
    x_deltap=interpsinc(x,tnow+tau+delta,1); % value to right
    x_deltam=interpsinc(x,tnow+tau-delta,1); % value to left
    dx=x_deltap-x_deltam;              % numerical derivative
    qx=quantalph(xs(i),[-3,-1,1,3]);   % quantize to alphabet
    tau=tau+mu*dx*(qx-xs(i));          % alg update: DD
    tnow=tnow+m; tausave(i)=tau;       % save for plotting
end

% plot results
figure();
subplot(2,1,1), plot(xs(1:i-2),'b.')   % plot constellation
    diagram
title('constellation diagram');
ylabel('estimated symbol values')
subplot(2,1,2), plot(tausave(1:i-2))    % plot trajectory of tau
ylabel('offset estimates'), xlabel('iterations')
title('Testing 6-PAM');
% Answer
% It appears that the more levels are used to represent a signal more
% noise appears in the signal as well.

```





Exercise 12.2

prepare transmitted signal

```

n=10000; % number of data points
m=2; % oversampling factor
beta=0.3; % rolloff parameter for srrc
l=50; % 1/2 length of pulse shape (in
symbols)
chan=[1]; % T/m "channel"
toffset=-0.3; % initial timing offset
x = (m+toffset)/2:beta:(m+toffset)/2;
%pulshap = 1*rectpuls(x, 1);
pulshap = rect_pulse_maker(1, beta, m, toffset);
s=pam(n,4,5); % random data sequence with var=5
sup=zeros(1,n*m); % upsample the data by placing...
sup(1:m:n*m)=s; % ... m-1 zeros between each data
point
hh=conv(pulshap,chan); % ... and pulse shape
r=conv(hh,sup); % ... to get received signal
%matchfilt=srrc(1,beta,m,0); % matched filter = srrc pulse shape
matchfilt = rect_pulse_maker(1, beta, m, 0);
x=conv(r,matchfilt); % convolve signal with matched filter

% clock recovery algorithm
tnow=l*m+1; tau=0; xs=zeros(1,n); % initialize variables

```

```

tausave=zeros(1,n); tausave(1)=tau; i=0;
mu=0.01; % algorithm stepsize
delta=0.1; % time for derivative
while tnow<length(x)-2*1*m % run iteration
    i=i+1;
    xs(i)=interpinc(x,tnow+tau,1); % interp value at tnow+tau
    x_deltap=interpinc(x,tnow+tau+delta,1); % value to right
    x_deltam=interpinc(x,tnow+tau-delta,1); % value to left
    dx=x_deltap-x_deltam; % numerical derivative
    qx=quantalph(xs(i),[-3,-1,1,3]); % quantize to alphabet
    tau=tau+mu*dx*(qx-xs(i)); % alg update: DD
    tnow=tnow+m; tausave(i)=tau; % save for plotting
end

% ANSWER
% It appears that a rectangular pulse has very poor spectral
% properties
% and because it varies so high from the maxima you end with a lot of
% interference. A raised cosine can control the excess bandwidth but
% that
% is hard to do in a rectangular pulse leading to a noisy signal that
% does
% not converge properly.

```

Exercise 12.3

```

% clockrecDD.m: clock recovery minimizing 4-PAM cluster variance
% to minimize  $J(\tau) = (Q(x(kT/M+\tau)) - x(kT/M+\tau))^2$ 

% prepare transmitted signal
n=10000; % number of data points
m=2; % oversampling factor
beta=0.3; % rolloff parameter for srcc
l=50; % 1/2 length of pulse shape (in
symbols)
chan=[1]; % T/m "channel"
toffset=-0.3; % initial timing offset
pulshap=srcc(l,beta,m,toffset); % srcc pulse shape with timing offset
s=pam(n,4,5); % random data sequence with var=5
sup=zeros(1,n*m); % upsample the data by placing...
sup(1:m:n*m)=s; % ... m-1 zeros between each data
point
hh=conv(pulshap,chan); % ... and pulse shape
r=conv(hh,sup); % ... to get received signal
matchfilt=srcc(l,beta,m,0); % matched filter = srcc pulse shape
x=conv(r,matchfilt); % convolve signal with matched filter
x=x+0.15*randn(size(x));
% clock recovery algorithm
tnow=l*m+1; tau=0; xs=zeros(1,n); % initialize variables
tausave=zeros(1,n); tausave(1)=tau; i=0;
mu=0.01; % algorithm stepsize
delta=0.1; % time for derivative
while tnow<length(x)-2*1*m % run iteration

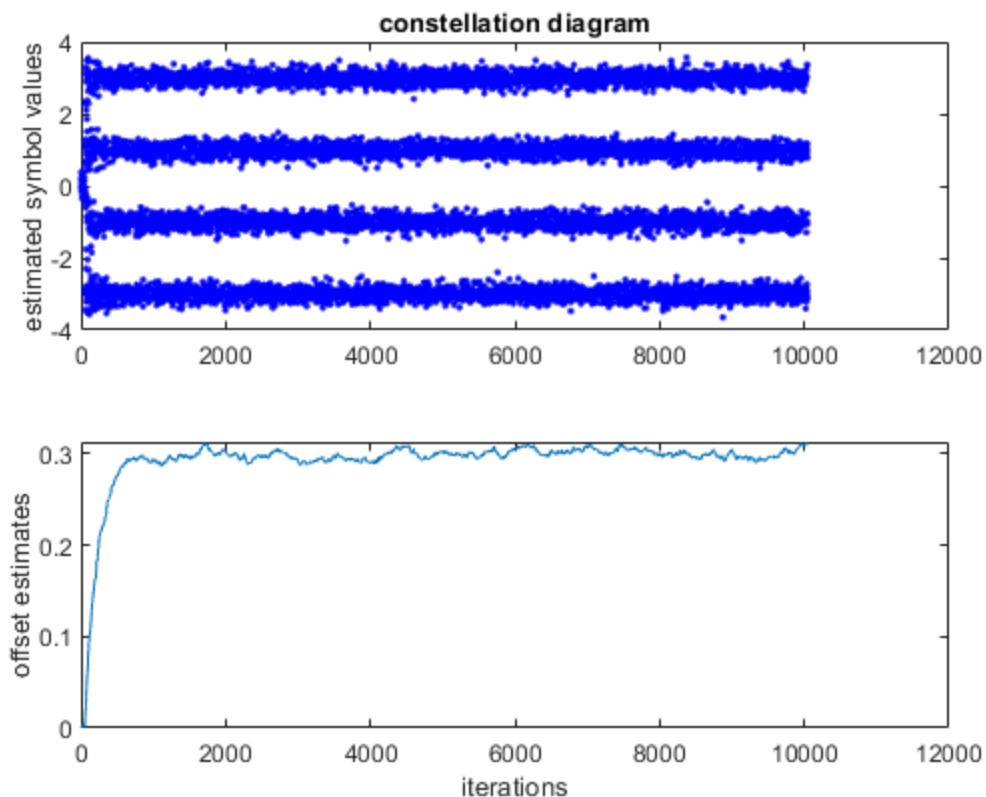
```

```

i=i+1;
xs(i)=interpinc(x,tnow+tau,1); % interp value at tnow+tau
x_deltap=interpinc(x,tnow+tau+delta,1); % value to right
x_deltam=interpinc(x,tnow+tau-delta,1); % value to left
dx=x_deltap-x_deltam; % numerical derivative
qx=quantalph(xs(i),[-3,-1,1,3]); % quantize to alphabet
tau=tau+mu*dx*(qx-xs(i)); % alg update: DD
tnow=tnow+m; tausave(i)=tau; % save for plotting
end

% plot results
subplot(2,1,1), plot(xs(1:i-2),'b.') % plot constellation
    diagram
title('constellation diagram');
ylabel('estimated symbol values')
subplot(2,1,2), plot(tausave(1:i-2)) % plot trajectory of tau
ylabel('offset estimates'), xlabel('iterations')
% ANSWER
% It appears that it stills converges to the correct value however it
% is a
% lot more noisy then before.

```



Exercise 13.1

LSequalizer.m find a LS equalizer f for the channel b

```

b=[0.5 1 -0.6]; % define channel
figure
plot(freqz(b));
title('Plotting channel ');
[hb, wb] = freqz(b);
m=1000; s=sign(randn(1,m)); % binary source of length m
r=filter(b,1,s); % output of channel
n=3; % length of equalizer - 1
for iter = 1:4
    delta=iter-1; % use delay <=n*length(b)
    p=length(r)-delta;
    R=toeplitz(r(n+1:p),r(n+1:-1:1)); % build matrix R
    S=s(n+1-delta:p-delta)'; % and vector S
    f=inv(R'*R)*R'*S; % calculate equalizer f
    Jmin=S'*S-S'*R*inv(R'*R)*R'*S % Jmin for this f and delta
    y=filter(f,1,r); % equalizer is a filter
    dec=sign(y); % quantize and find errors
    err=0.5*sum(abs(dec(delta+1:m)-s(1:m-delta)))
    figure
    plot(freqz(b,f))
    title(['Trying equalizer with delta value ', num2str(iter)])

    [h,w] = freqz(b,f);
    figure
    plot(abs(h.*hb))
    title(['Plotting magnitude of frequency response ',
num2str(iter)]);
end
% ANSWER
% Even though the magnitude of teh channel and the equalizers seems to
be
% opposite they never really cancel out and reach unity.

Jmin =

    816.1689

err =

    395

Jmin =

    134.4687

err =

    0

```

$J_{min} =$

31.0168

$err =$

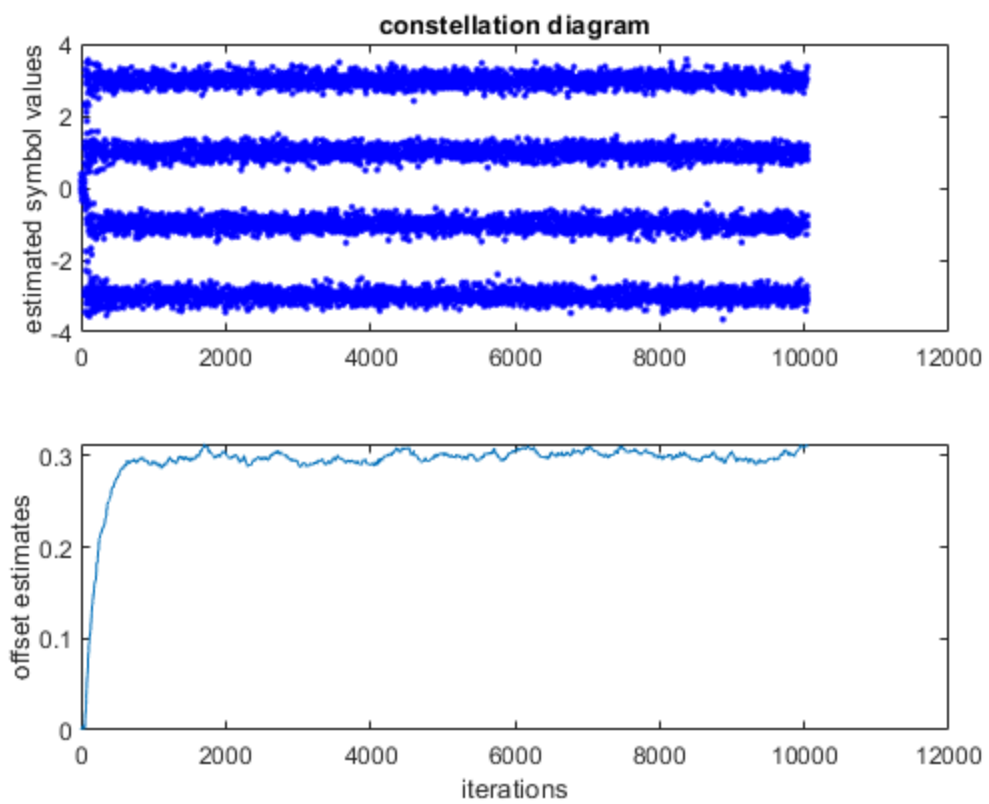
0

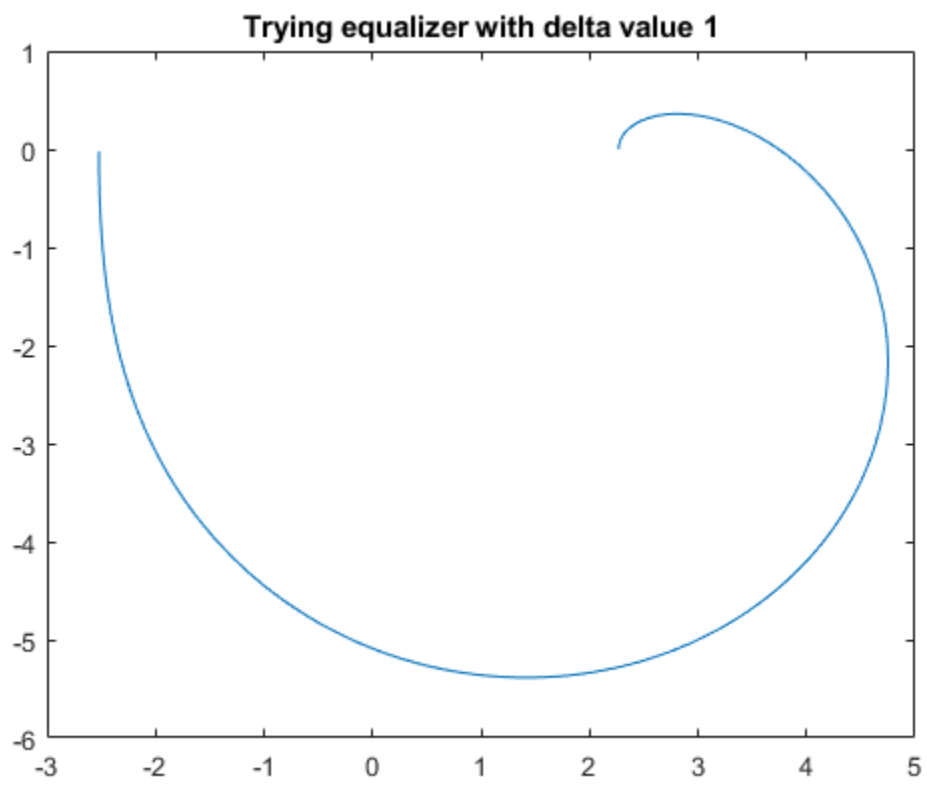
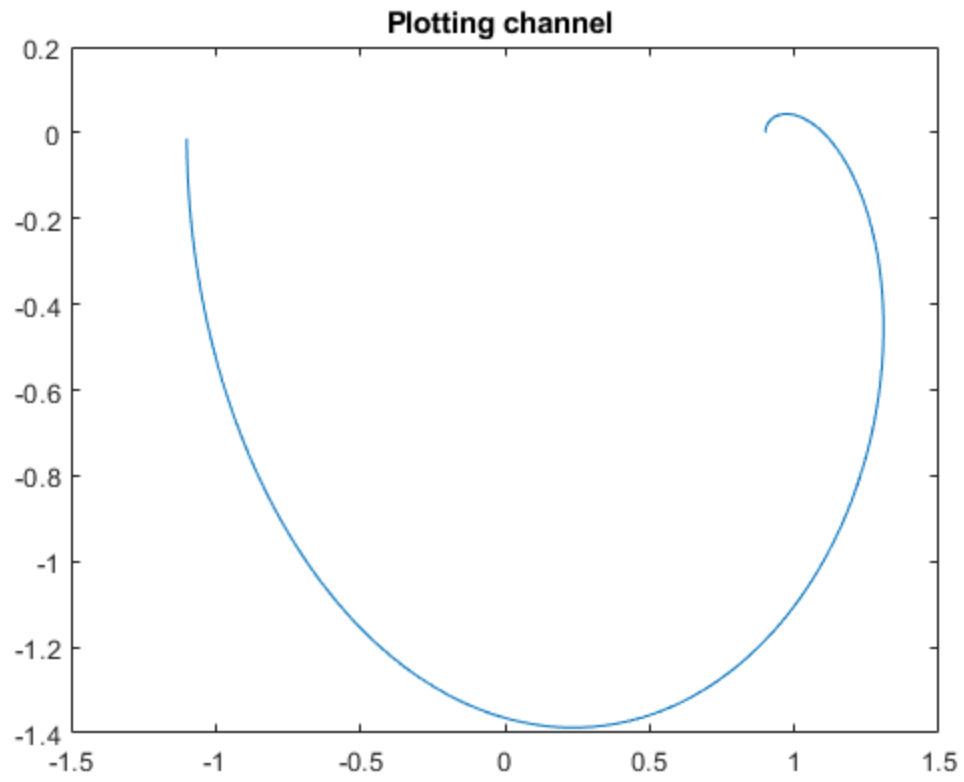
$J_{min} =$

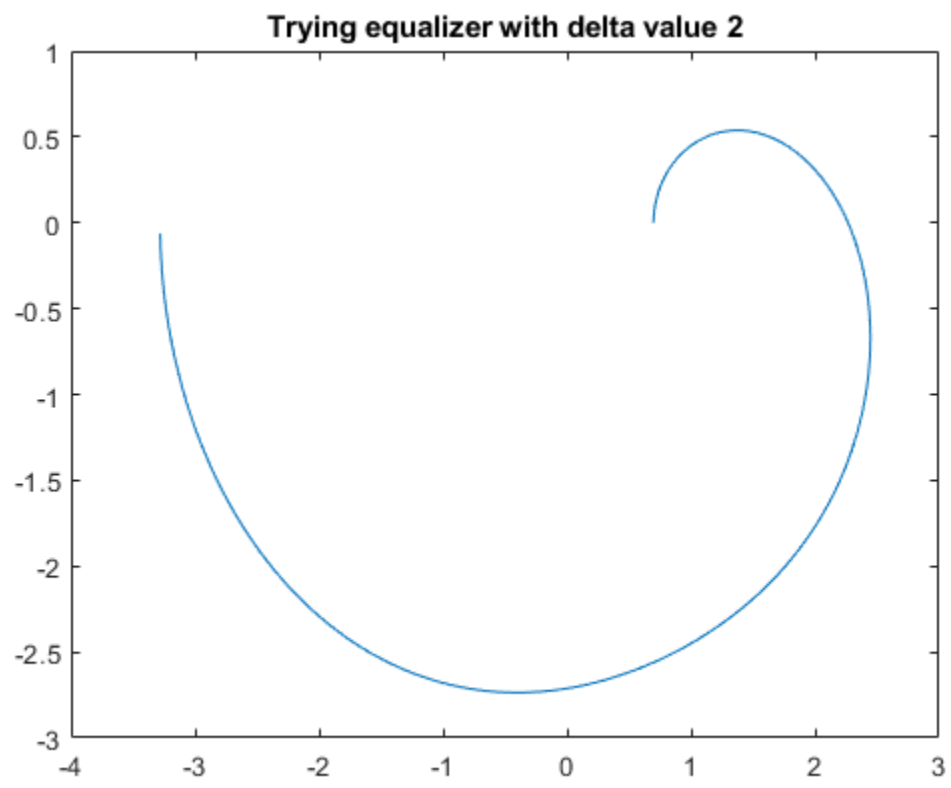
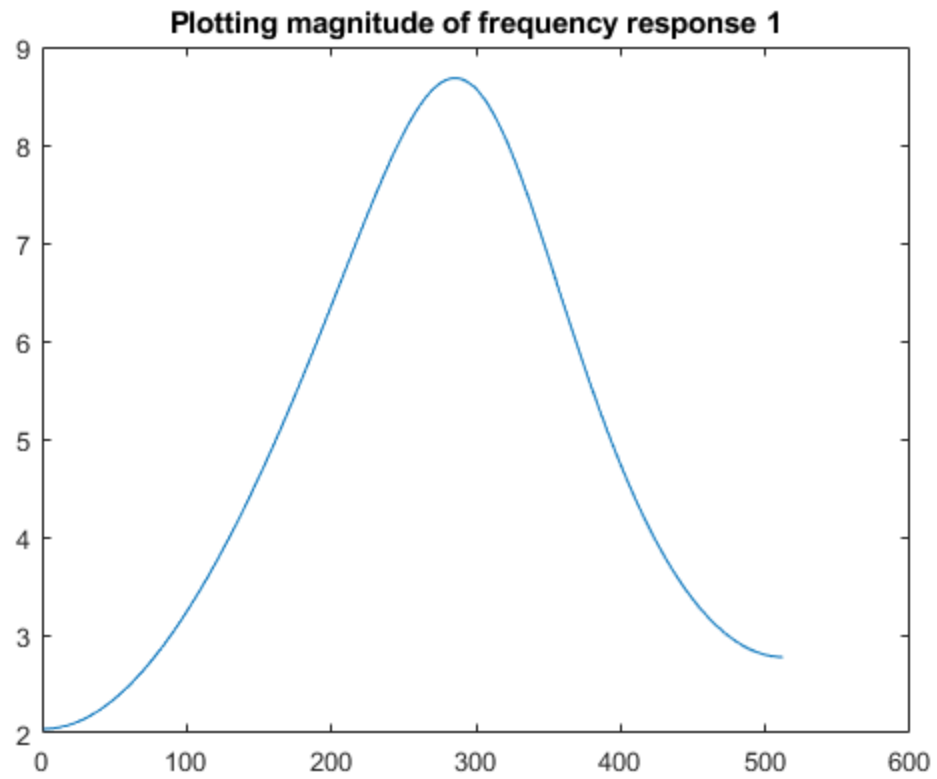
44.8434

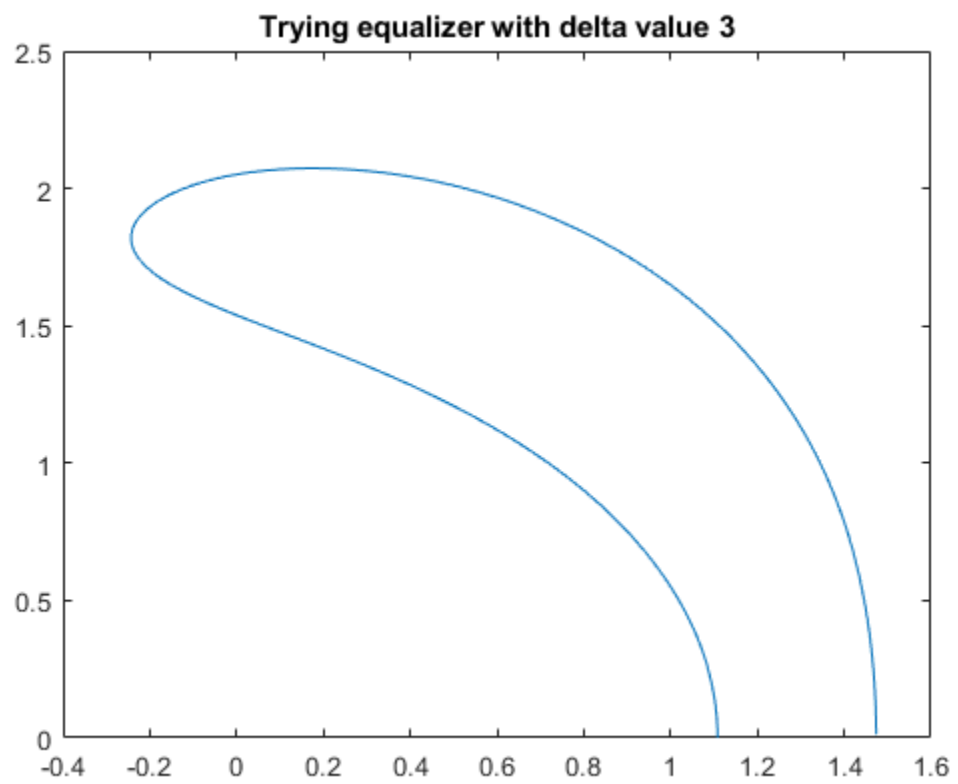
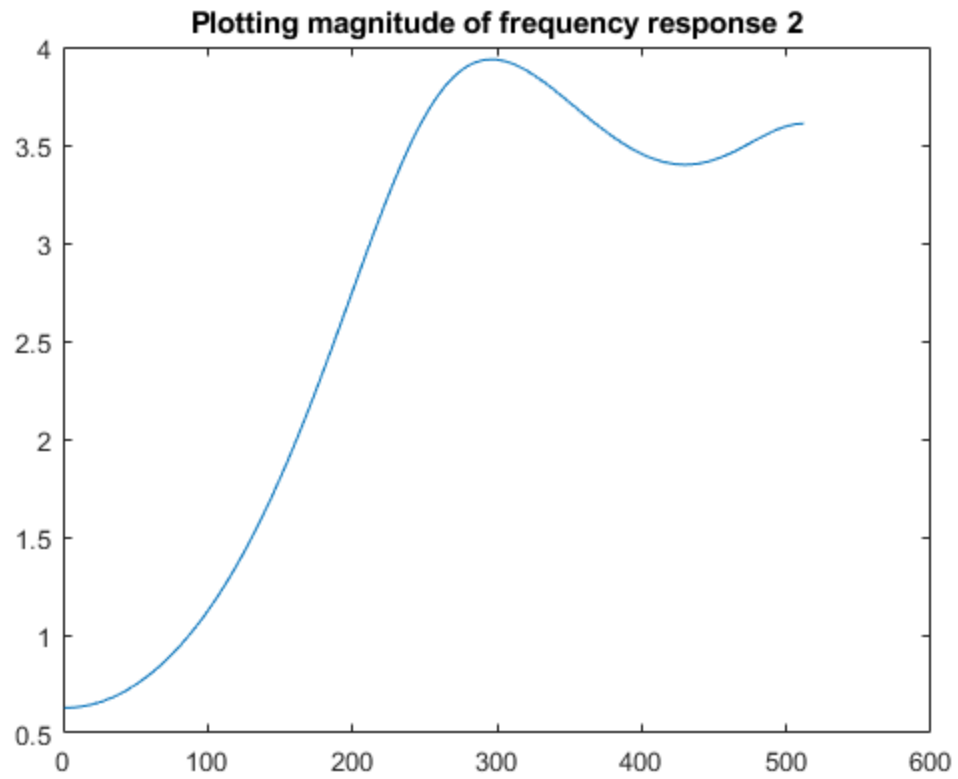
$err =$

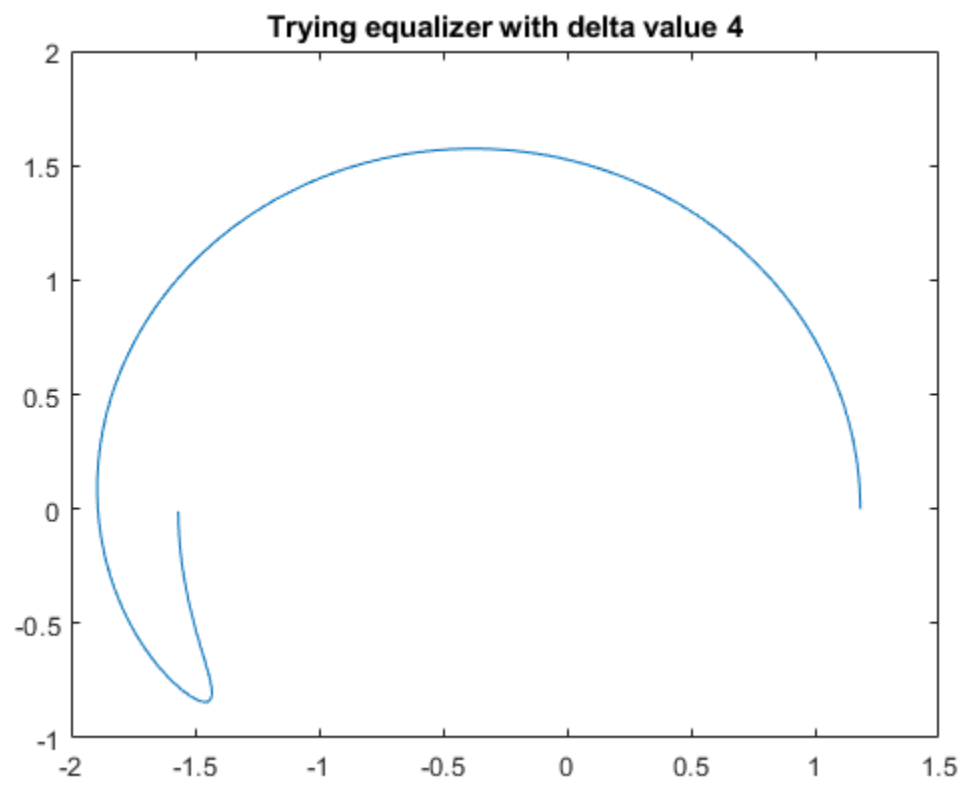
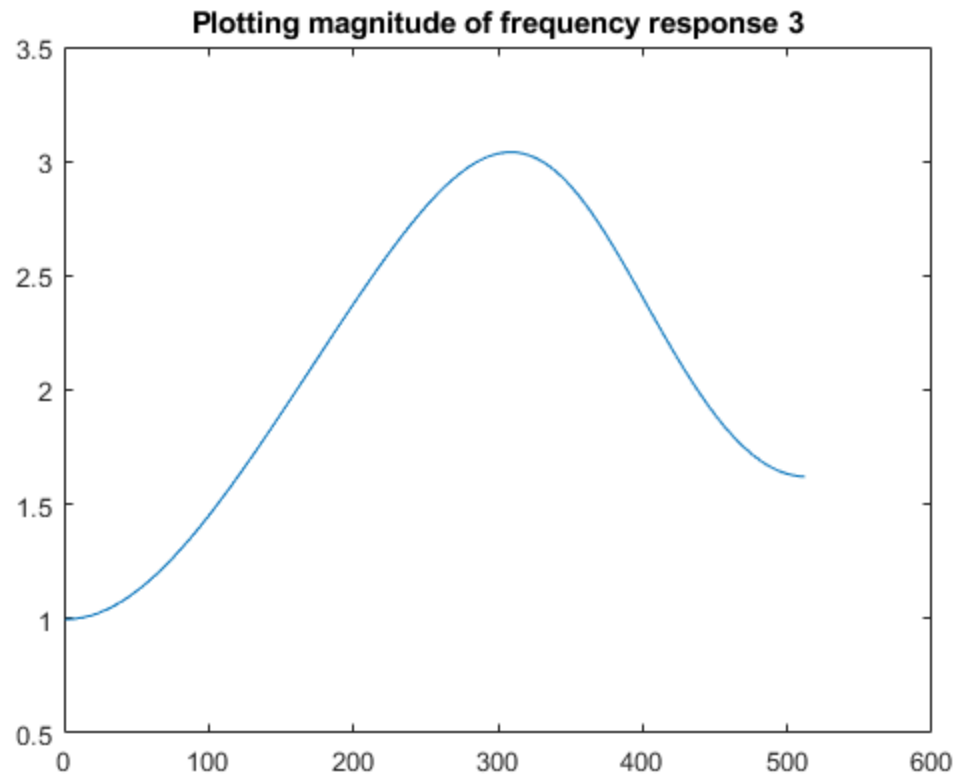
0

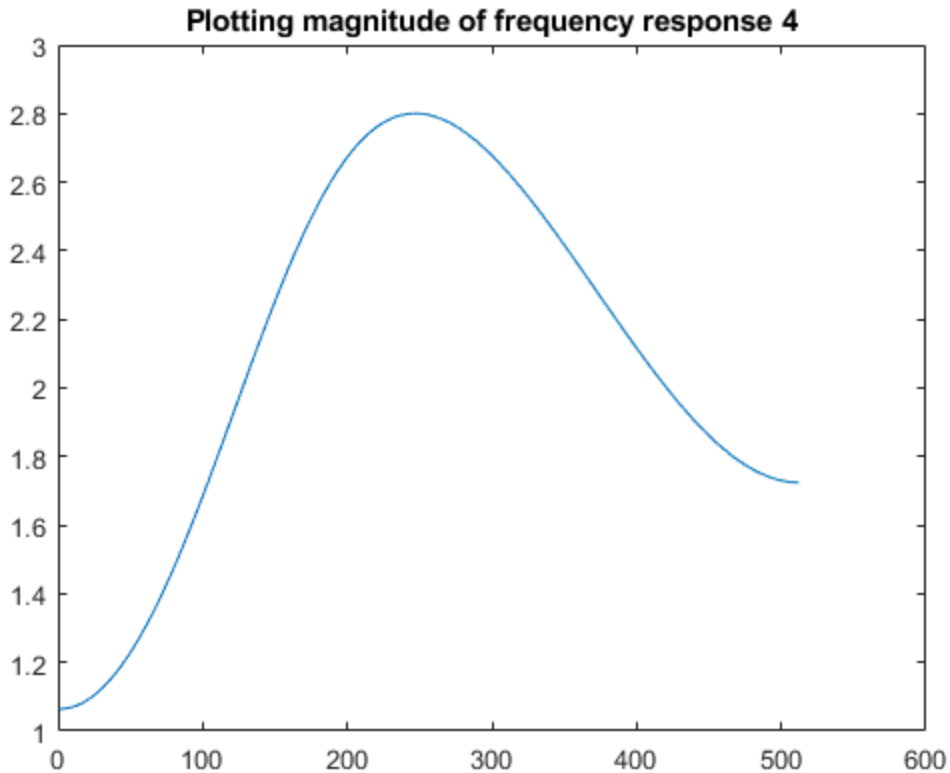












Exercise 13.2 a

LSequalizer.m find a LS equalizer f for the channel b

```

b=[0.5 1 -0.6]; % define channel
m=1000; s=sign(randn(1,m)); % binary source of length m
sd = 0.3;
r=filter(b,1,s)+sd*randn(size(s)); % output of
channel
n=3; % length of equalizer - 1
delta=2; % use delay <=n*length(b)
p=length(r)-delta;
R=toeplitz(r(n+1:p),r(n+1:-1:1)); % build matrix R
S=s(n+1-delta:p-delta)'; % and vector S
f=inv(R'*R)*R'*S % calculate equalizer f
Jmin=S'*S-S'*R*inv(R'*R)*R'*S % Jmin for this f and delta
y=filter(f,1,r); % equalizer is a filter
dec=sign(y); % quantize and find errors
err=0.5*sum(abs(dec(delta+1:m)-s(1:m-delta)))
% ANSWER
% It appears that the value where error first start appearing is
around 0.3

```

$f =$


```

-0.2706
0.6171
0.2828
0.1253

```

```
Jmin =
```

```
79.2882
```

```
err =
```

```
0
```

Exercise 13.2b

```

x = []
for sd = 0.1:0.05:1
    b=[0.5 1 -0.6]; % define channel
    m=1000; s=sign(randn(1,m)); % binary source of length m
    r=filter(b,1,s)+sd*randn(size(s)); % output of
    channel
    n=3; % length of equalizer - 1
    delta=2; % use delay <=n*length(b)
    p=length(r)-delta;
    R=toeplitz(r(n+1:p),r(n+1:-1:1)); % build matrix R
    S=s(n+1-delta:p-delta)'; % and vector S
    f=inv(R'*R)*R'*S % calculate equalizer f
    Jmin=S'*S-S'*R*inv(R'*R)*R'*S % Jmin for this f and delta
    y=filter(f,1,r); % equalizer is a filter
    dec=sign(y); % quantize and find errors
    err=0.5*sum(abs(dec(delta+1:m)-s(1:m-delta)))
    x = [x, Jmin]
end

figure
plot(0.1:0.05:1, x)
title('Jmin vs sd plot')

```

```
x =
```

```
[]
```

```
f =
```

```

-0.2655
0.6455
0.3088

```

```
0.1387

Jmin =

39.1061

err =

0

x =

39.1061

f =

-0.2741
0.6383
0.3082
0.1452

Jmin =

44.6709

err =

0

x =

39.1061    44.6709

f =

-0.2778
0.6249
0.2920
0.1281

Jmin =

55.0269
```

`err =`

`0`

`x =`

`39.1061 44.6709 55.0269`

`f =`

`-0.2672
0.6152
0.2911
0.1331`

`Jmin =`

`64.2515`

`err =`

`0`

`x =`

`39.1061 44.6709 55.0269 64.2515`

`f =`

`-0.2551
0.6033
0.2987
0.1274`

`Jmin =`

`84.2713`

`err =`

`0`

`x =`

`39.1061 44.6709 55.0269 64.2515 84.2713`

$f =$

-0.2744
0.5907
0.2580
0.1113

$J_{min} =$

98.5940

$err =$

0

$x =$

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940

$f =$

-0.2656
0.5859
0.2831
0.1118

$J_{min} =$

117.8857

$err =$

8

$x =$

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

$f =$

-0.2612
0.5644
0.2828
0.0944

Jmin =

142.1911

err =

6

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Column 8

142.1911

f =

-0.2485

0.5480

0.2589

0.1110

Jmin =

169.2462

err =

16

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 9

142.1911 169.2462

f =

```

-0.2448
 0.5221
 0.2601
 0.0855

Jmin =

166.8200

err =

13

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 10

142.1911 169.2462 166.8200

f =

-0.2282
 0.5234
 0.2492
 0.0974

Jmin =

202.2782

err =

21

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 11

142.1911 169.2462 166.8200 202.2782

```

$f =$

-0.2251
0.4995
0.2491
0.0808

$J_{min} =$

237.7480

$err =$

34

$x =$

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 12

142.1911 169.2462 166.8200 202.2782 237.7480

$f =$

-0.2537
0.4999
0.2117
0.0819

$J_{min} =$

248.1171

$err =$

40

$x =$

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 13

142.1911	169.2462	166.8200	202.2782	237.7480	248.1171	
----------	----------	----------	----------	----------	----------	--

$f =$

-0.2199
0.4590
0.2379
0.0784

$J_{min} =$

269.5336

$err =$

55

$x =$

Columns 1 through 7

39.1061	44.6709	55.0269	64.2515	84.2713	98.5940	117.8857
---------	---------	---------	---------	---------	---------	----------

Columns 8 through 14

142.1911	169.2462	166.8200	202.2782	237.7480	248.1171	269.5336
----------	----------	----------	----------	----------	----------	----------

$f =$

-0.2254
0.4366
0.2161
0.0649

$J_{min} =$

302.8387

$err =$

60

$x =$

Columns 1 through 7

39.1061	44.6709	55.0269	64.2515	84.2713	98.5940	117.8857
---------	---------	---------	---------	---------	---------	----------

Columns 8 through 14

142.1911	169.2462	166.8200	202.2782	237.7480	248.1171	269.5336
----------	----------	----------	----------	----------	----------	----------

Column 15

302.8387

$f =$

-0.2080
0.4357
0.2247
0.0714

$J_{min} =$

322.4758

$err =$

77

$x =$

Columns 1 through 7

39.1061	44.6709	55.0269	64.2515	84.2713	98.5940	117.8857
---------	---------	---------	---------	---------	---------	----------

Columns 8 through 14

142.1911	169.2462	166.8200	202.2782	237.7480	248.1171	269.5336
----------	----------	----------	----------	----------	----------	----------

Columns 15 through 16

302.8387	322.4758
----------	----------

$f =$

-0.2076
0.4092
0.2084
0.0780

Jmin =

356.3454

err =

89

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 14

142.1911 169.2462 166.8200 202.2782 237.7480 248.1171 269.5336

Columns 15 through 17

302.8387 322.4758 356.3454

f =

-0.1994

0.4034

0.1798

0.0521

Jmin =

375.0249

err =

97

x =

Columns 1 through 7

39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 14

142.1911 169.2462 166.8200 202.2782 237.7480 248.1171 269.5336

Columns 15 through 18

302.8387 322.4758 356.3454 375.0249

f =

-0.1847
0.3874
0.1822
0.0361

Jmin =

414.4879

err =

111

x =

Columns 1 through 7

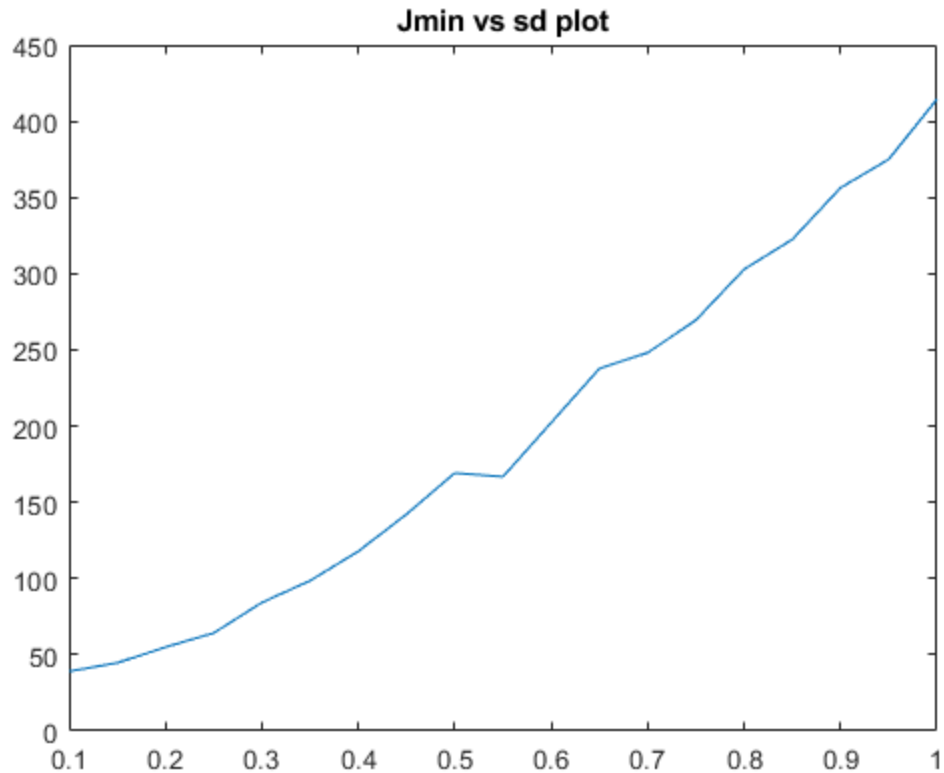
39.1061 44.6709 55.0269 64.2515 84.2713 98.5940 117.8857

Columns 8 through 14

142.1911 169.2462 166.8200 202.2782 237.7480 248.1171 269.5336

Columns 15 through 19

302.8387 322.4758 356.3454 375.0249 414.4879



Exercise 13.2 c

LSequalizer.m find a LS equalizer f for the channel b

```

b=[0.5 1 -0.6]; % define channel
m=1000; s=sign(randn(1,m)); % binary source of length m
sd = 0.19;
r=filter(b,1,s)+sd*randn(size(s)); % output of
channel
n=3; % length of equalizer - 1
delta=1; % use delay <=n*length(b)
p=length(r)-delta;
R=toeplitz(r(n+1:p),r(n+1:-1:1)); % build matrix R
S=s(n+1-delta:p-delta)'; % and vector S
f=inv(R'*R)*R'*S % calculate equalizer f
Jmin=S'*S-S'*R*inv(R'*R)*R'*S % Jmin for this f and delta
y=filter(f,1,r); % equalizer is a filter
dec=sign(y); % quantize and find errors
err=0.5*sum(abs(dec(delta+1:m)-s(1:m-delta)))

```

%ANSWER

% It appears that when sd is around 0.2 error appears

$f =$

```
0.6471
0.3528
0.1593
0.0631
```

```
Jmin =
```

```
163.2112
```

```
err =
```

```
0
```

Exercise 13.2 d

ANSWER The first one with the larger delay seems to perform better, because the value of σ_d can be larger than the one with the smaller delay. This means it can take larger noise values and is therefore a more robust system.

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