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```
clear;
clc;
close all;
```

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definition

d : desired signal N :length of filter M : length of input signal alpha : learning rate e : errors w : weights of filter
m_error : mean squared error p : power of input signal

```
a=1;
b=[1,1.8,0.81];           % impulse response
inputs=randn(1,100);
d=filter(b,a,inputs);
M=length(inputs);
```

part a

```
N=4;
p= inputs*inputs'/M;
alpha_max=2/(3*N*p);
disp('mu max fro N=4 and is :');
disp(alpha_max);
```

```
mu max fro N=4 and is :
    0.1730
```

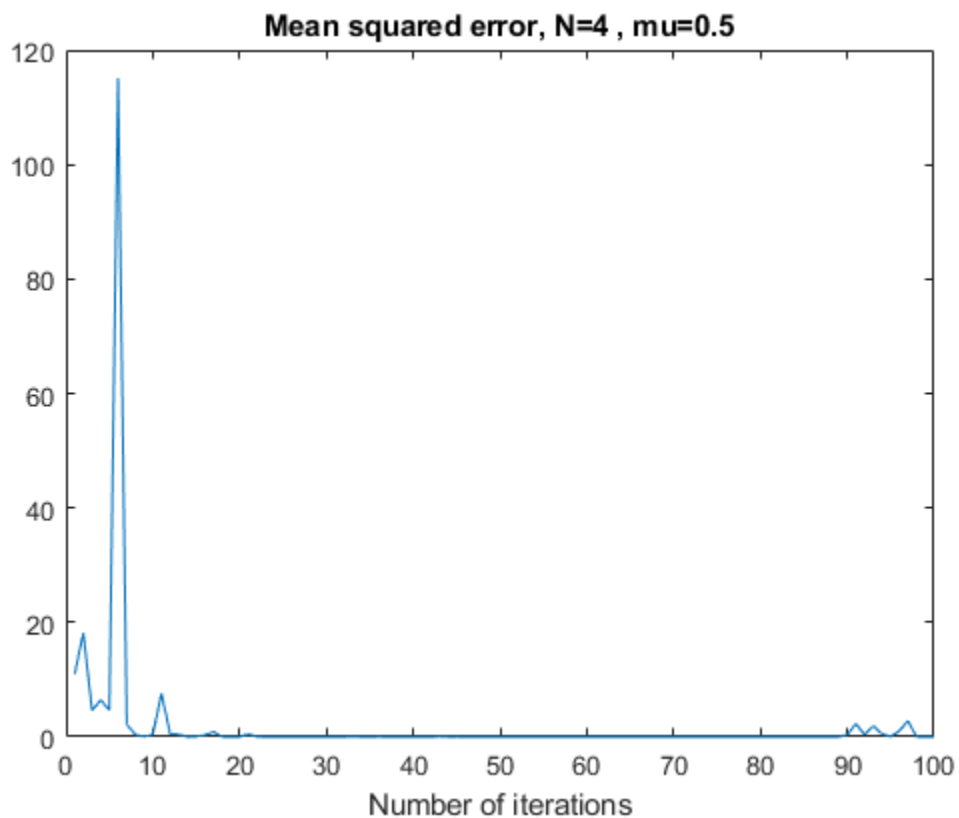
part b

```
alpha=0.5;
[w,~]=LMS(inputs,d,N,alpha,M);
disp("weights for mu=0.5 and N=4 :");
disp(w');
disp('because mu is bigger than u_max may be LMS algorithm not converged')
```

part c

```
k=5;
m_error=zeros(1,M);

for i=1:k
    [~,cost]=LMS(inputs,d,N,alpha,M);
    m_error=m_error+cost;
end
m_error=m_error/5;
figure
plot(m_error);
title('Mean squared error, N=4 , mu=0.5');
xlabel('Number of iterations');
```



part d

```
[~,~,J_min,J_inf]=LMS(inputs,d,N,alpha,M);
J_ex=J_inf - J_min;
disp("excess Mean squared error")
disp(J_ex)
```

```
excess Mean squared error
    0.4650
```

part e

$\mu = 0.1$ and $N=4$

$\alpha=0.1$;

```
for i=1:k
    [w,cost]=LMS(inputs,d,N,alpha,M);
    m_error=m_error+cost;
end
m_error=m_error/5;
disp('weights for mu=0.1 and N=4  :');
disp(w');
figure
plot(m_error);
title('Mean squared error, N=4 , mu=0.1');
xlabel('Number of iterations');
```

$\mu = 0.05$ and $N=4$

$\alpha=0.05$;

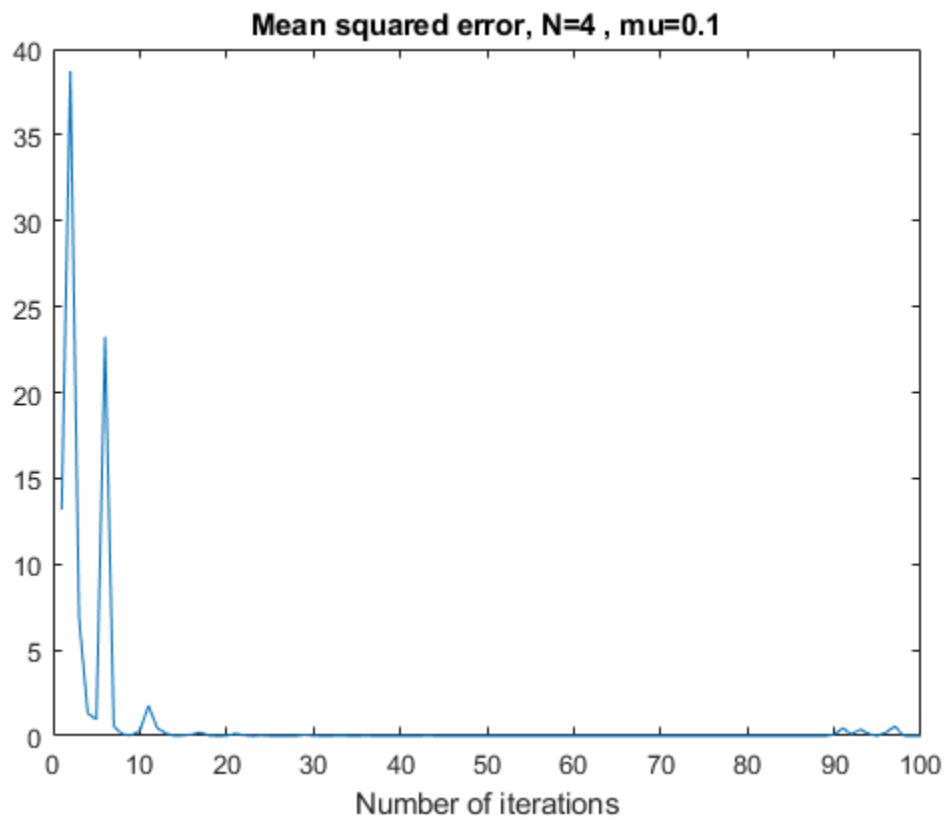
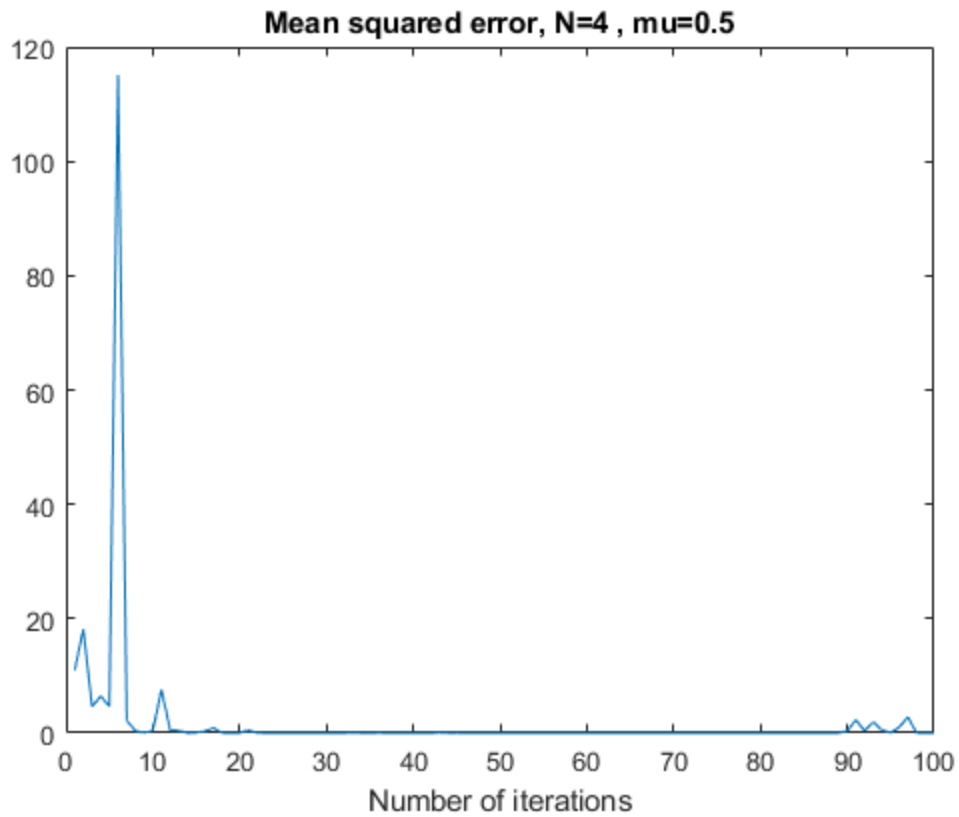
```
for i=1:k
    [w,cost]=LMS(inputs,d,N,alpha,M);
    m_error=m_error+cost;
end
m_error=m_error/5;
disp('weights for mu=0.05 and N=4  :');
disp(w');
figure
plot(m_error);
title('Mean squared error, N=4 , mu=0.05');
xlabel('Number of iterations');
```

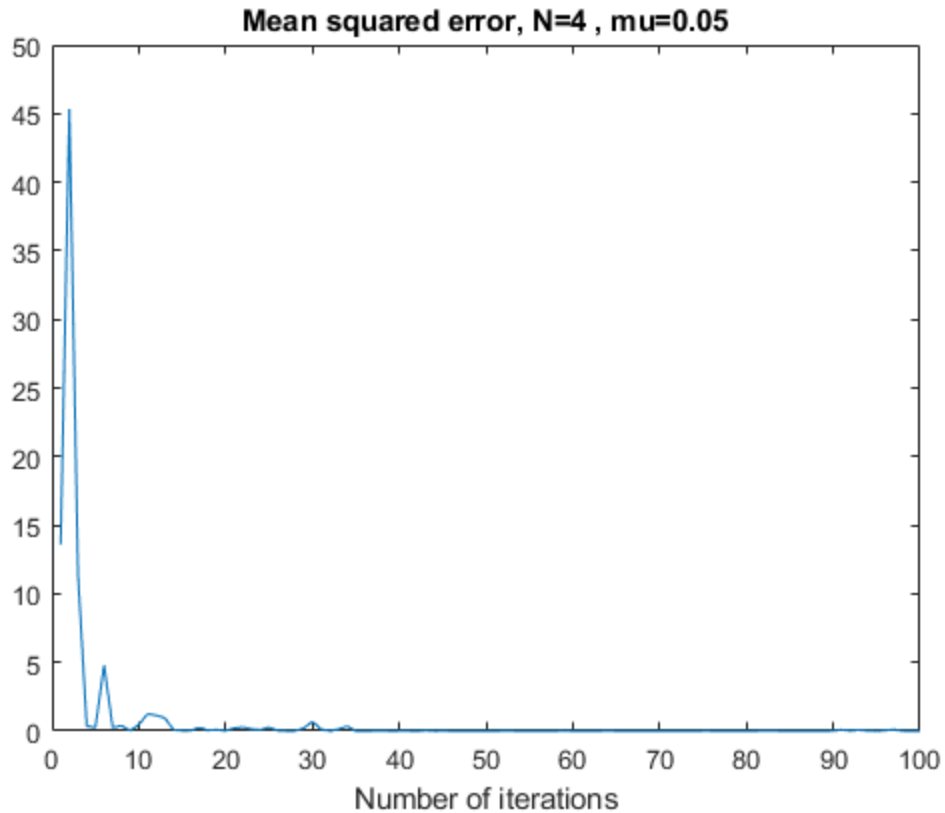
weights for $\mu=0.1$ and $N=4$:

```
1.0000
1.8000
0.8100
-0.0000
```

weights for $\mu=0.05$ and $N=4$:

```
1.0016
1.7968
0.8096
-0.0021
```





part e

mu = 0.5 and N=2

```
alpha=0.5;
N=2;

for i=1:k
    [w,cost]=LMS(inputs,d,N,alpha,M);
    m_error=m_error+cost;
end
m_error=m_error/5;
disp('weights for mu=0.5 and N=2 :');
disp(w');
figure
plot(m_error);
title('Mean squared error, N=2 , mu=0.5');
xlabel('Number of iterations');

% mu = 0.5 and N=3

alpha=0.5;
N=3;

for i=1:k
    [w,cost]=LMS(inputs,d,N,alpha,M);
```

```

        m_error=m_error+cost;
    end
    m_error=m_error/5;
    disp('weights for mu=0.5 and N=3 :');
    disp(w');
    figure
    plot(m_error);
    title('Mean squared error, N=3 , mu=0.5');
    xlabel('Number of iterations');

```

LMS algorithms

```

function[w,cost,J_min,J_inf]=LMS(inputs,d,N,alpha,M)
% e : error
% u_temp : because LMS run when the first sample arrive, we put M-1 zeros in
begining of inputs, if whe don't put this zeros we must wait to m sample arrive
    u_temp=[zeros(1,N-1),inputs];
    e=zeros(1,M);
    w=zeros(1,N);
    for i=N:M
        u=u_temp(i:-1:i-N+1);
        y=dot(w,u);
        e(i-N+1)=d(i-N+1)-y;
        w = w + alpha*e(i-N+1)*u;
    end
    cost=e.^2;
    J_min=min(cost);
    J_inf=sum(cost(M-19:M))/20;

end

weights for mu=0.5 and N=4 :
    0.7950
    2.6811
    0.6021
    0.4111

```

because mu is begger than u_max may be LMS algorithm not converged

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