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EE 5322 Intelligent Control Systems

Assignment no 3

Discrete Time Simulation, Observers, Kalman Filter

1. Discrete-Time System.

Q1.a)

Solution :

```
clc;
clear all;
close all;

X=zeros(2,100);
Xn=zeros(2,100);
A=[0 1
   -0.89 1.8];
B=[0
   1];
u=ones(100,1);
t=1:100;
for i=1:99
X(:,i+1)=A*X(:,i)+B*u(i);
y(i,:) = X(:,i+1) '
hold on
end
figure(1)
plot(t,X(1,:));
hold on;
grid on;
plot(t,X(2,:), 'r');
legend t X
grid on;
title('System States Without Noise')
max_overshoot = max(y(:,1))
SteadyState_value = max(y(i,:))
peak_overshoot = (max_overshoot(1) - SteadyState_value)/SteadyState_value *100
```

Solution :

y =

```
0 1.0000
1.0000 2.8000
2.8000 5.1500
5.1500 7.7780
...
11.1361 11.1243
11.1243 11.1126
11.1126 11.1020
```

max_overshoot =

```
17.2655
```

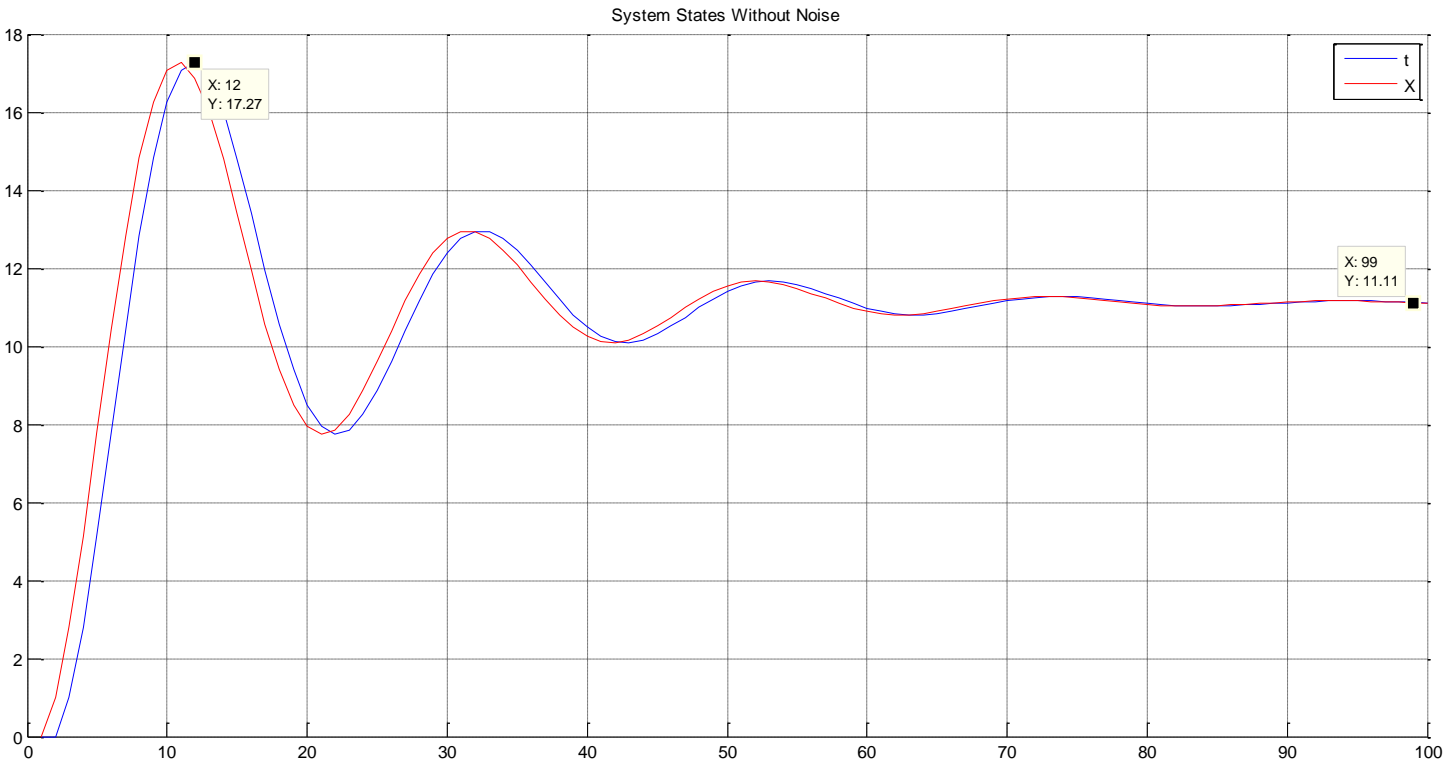
SteadyState_value =

```
11.1126
```

peak_overshoot =
55.3691

Plots

:



Q1.b)

Code :

```
clc;
clear all;
close all;

x=zeros(2,100);
xn=zeros(2,100);
A=[0      1
   -0.89  1.8];
B=[0
   1];
u=ones(100,1);
t=1:100;
w = 0.2*rand(2,100);
for i=1:99
    x(:,i+1)=A*x(:,i)+B*u(i)+w(i);
    y(i,:) = x(:,i+1)'
    hold on
end
figure (1)
plot(t,x(1,:),t,x(2,:));
hold on;
plot(t,x(2,:), 'r');
grid;
title('System States With Noise')
max_overshoot = max(y())
min_overshoot = max(y(i,:))
peak_overshoot = (max_overshoot(1) - min_overshoot)/min_overshoot *100
```

Solution :

y =

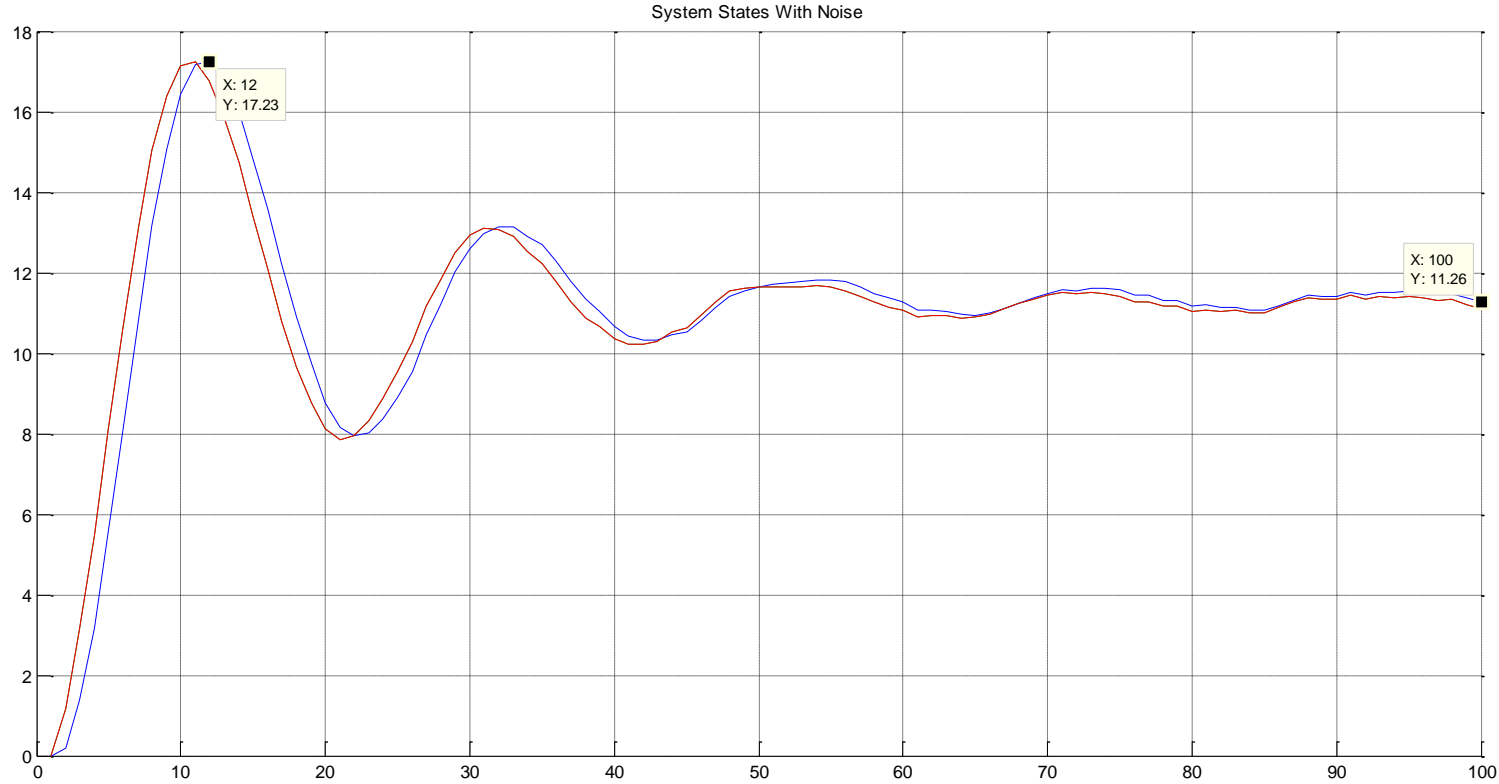
0.1054	1.1054
1.1782	2.9687
2.9827	5.3091
5.3588	7.9514
7.9632	10.5551
...	
11.5183	11.3675
11.4731	11.3158
11.4842	11.3258
11.3617	11.2014
11.2566	11.1058

```
max_overshoot =
    17.2334
```

```
min_overshoot =
    11.2566
```

```
peak_overshoot =
    53.0955
```

Plot:



2.

CODE :

```
clc;
clear all;
close all;

A=[0      1
   -0.89 1.8]; % State space description of system
B=[0
    1];
u=ones(100,1);
H=[2 0];
x(:,1)=[0;0];
X_time_mes(:,1)=[10
                  15]; %initial wrong estimate of states
                        %initial guess of error covariance
P=100*eye(2);
P_cov(:,1)=diag(P);
G=eye(2);
Mean=.1*eye(2);
Var=.1;

%% Implement DT Kalman Filter
for i=1:100
    x(:,i+1) = A * x(:,i) + B * u(i) + 0.1*randn(2,1); % system at time i+1
    z(i+1) = H * x(:,i+1) + Var * randn; %measurement at i+1
    x_copy(:,i+1) = x(:,i+1);

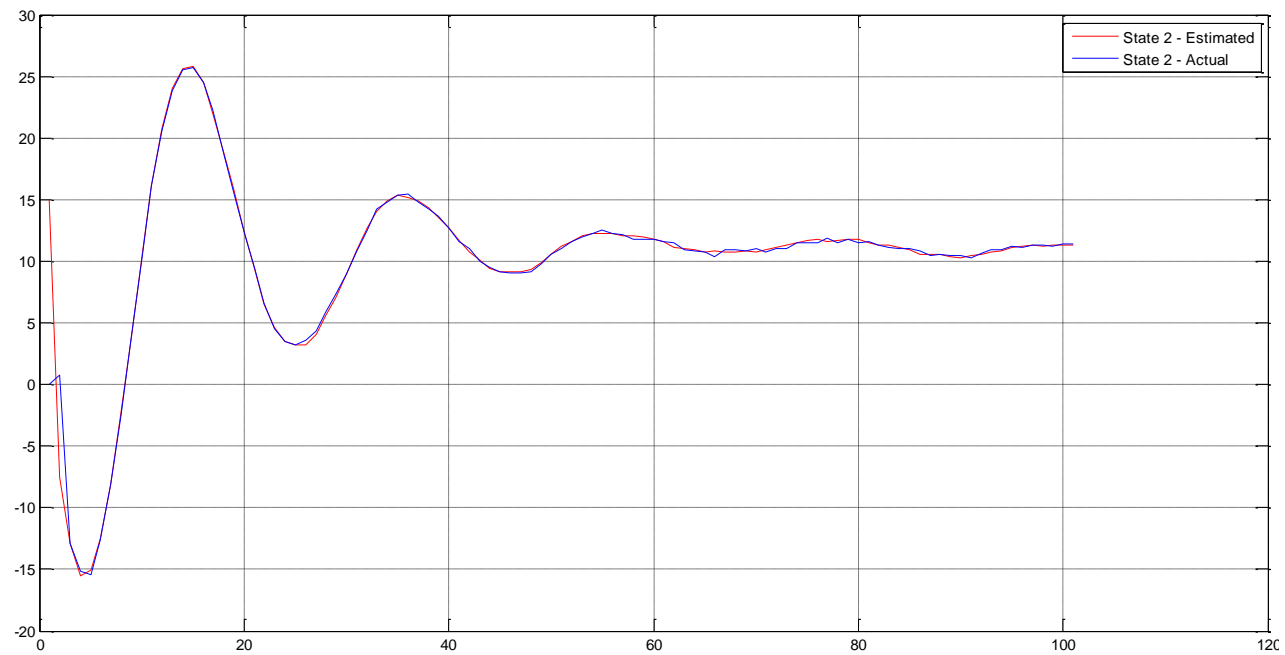
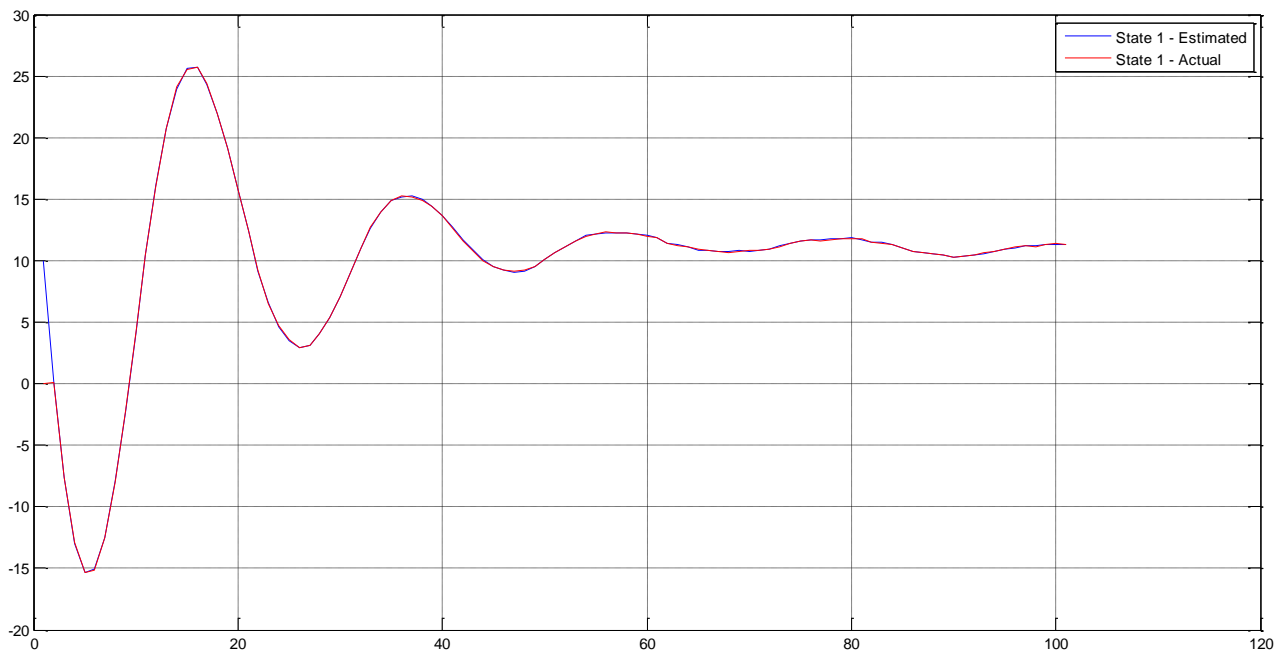
    X_time_update(:,i+1) = A * X_time_mes(:,i) + B * u(i); %Time Update state at i+1
    P = A * P * A' + Mean * (G * G)'; %Time Update Error Covariance
    P_error_cov(:,i+1) = diag(P);

    P = P - P * H' * 1/(H * P * H' + Var) * H * P; %Cov Measurement Update
    P_updated_cov(:,i+1) = diag(P); %track time updated dovariance

    KalmanGain = P * H' * 1/(Var); %Kalman Gain
    Kalman_Gain_update(:,i)=KalmanGain; %Record kalman gain
    X_time_mes(:,i+1) = X_time_update(:,i+1) + KalmanGain * (z(i+1) - H * X_time_update(:,i+1));
    %Measurement Update

    x(:,i+1)=X_time_mes(:,i+1); %Feedback of updated state
end
%% Plot the states Estimates and the Actual states
t=1:101;
figure (1)
plot(t,X_time_mes(1,:), 'b');
hold on;
plot(t,x_copy(1,:), 'r')
hold on;
grid on;
legend('State 1 - Estimated', 'State 1 - Actual')
figure (2)
plot(t,X_time_mes(2,:), 'r');
hold on
plot(t,x_copy(2,:), 'b');
grid on;
hold on;
legend('State 2 - Estimated', 'State 2 - Actual')
```

Plot :



Q3.

Code:

```
clc;
clear all;
close all;

A=[0 1 % State space description of system
   -0.89 1.8];
B=[0
   1];
u=ones(100,1);
H=[2 0];
x(:,1)=[0;0];
X_time_mes(:,1)=[10
   15]; %initial wrong estimate of states
P=100*eye(2); %initial guess of error covariance
P_cov(:,1)=diag(P);
G=eye(2);
Mean=.1*eye(2);
Var=.1;

%% Implement DT Kalman Filter
for i=1:100
    x(:,i+1) = A * x(:,i) + B * u(i) + 0.1*randn(2,1); % system at time i+1
    z(i+1) = H * x(:,i+1) + Var * randn; %measurement at i+1
    x_copy(:,i+1) = x(:,i+1);

    X_time_update(:,i+1) = A * X_time_mes(:,i) + B * u(i); %Time Update state at i+1
    P = A * P * A' + Mean * (G * G)'; %Time Update Error Covariance
    P_error_cov(:,i+1) = diag(P);

    P = P - P * H' * 1/(H * P * H' + Var) * H * P; %Cov Measurement Update
    P_updated_cov(:,i+1) = diag(P); %track time updated dovariance

    KalmanGain = P * H' * 1/(Var); %Kalman Gain
    Kalman_Gain_update(:,i)=KalmanGain; %Record kalman gain
    X_time_mes(:,i+1) = X_time_update(:,i+1) + KalmanGain * (z(i+1) - H * X_time_update(:,i+1)); %Measurement Update

    x(:,i+1)=X_time_mes(:,i+1); %Feedback of updated state
end
%% Plot the states Estimates and the Actual states
t=1:101;
figure (1)
plot(t,X_time_mes(1,:), 'b');
hold on;
plot(t,x_copy(1,:), 'r')
hold on;
grid on;
legend('State 1 - Estimated', 'State 1 - Actual')

figure (2)
plot(t,X_time_mes(2,:), 'r');
hold on
plot(t,x_copy(2,:), 'b');
grid on;
hold on;
legend('State 2 - Estimated', 'State 2 - Actual')

figure(3)
plot(t(1:100),Kalman_Gain_update(1,:), 'r');
hold on;
grid on;
plot(t(1:100),Kalman_Gain_update(2,:), 'g');
grid on;
hold on;
legend('Kalman Gain - State 1', 'Kalman Gain - State 2');
title(' Kalman Gain With Iteration')

[M,P,Z,E] = dlqe(A,G,H,Mean,Var);
```

```

M = 'The value of gain by Steady State Solution is'
M
%Simulation of system
X_time_update(:,i+1) = A * X_time_mes(:,i) + B * u(i);           %Time Update state at i+1
X_time_mes(:,i+1) = X_time_update(:,i+1) + M * (z(i+1) - H * X_time_update(:,i+1)); %Measurement Update end

figure (4)
plot(t,X_time_mes(1,:), 'r', tx(1,:), 'b');
grid on;
hold on;
legend('State 1 - Estimated', 'State 1 - Actual ')
title('Fixed Kalman Gain : State 1');

Figure(5)
plot(t,X_time_mes(2,:), 'r');
grid on;
hold on;
plot(t,x(2,:), 'b');
grid on;
hold on;
legend('State 2 - Estimated', 'State 2 - Actual ')
title('Fixed Kalman Gain : State 2')

clear P
P=1000*eye(2);
for i=1:40 %some arbitrary number of Iterations to get the solution converged
P=A*P*A'+Q*(G*G');
P=P-P*H'*inv(H*P*H'+Var)*H*P;
end
KG_Iteration=P*H'*inv(Var)

```

Kalman Gain =

0.8607

1.1552

ans =

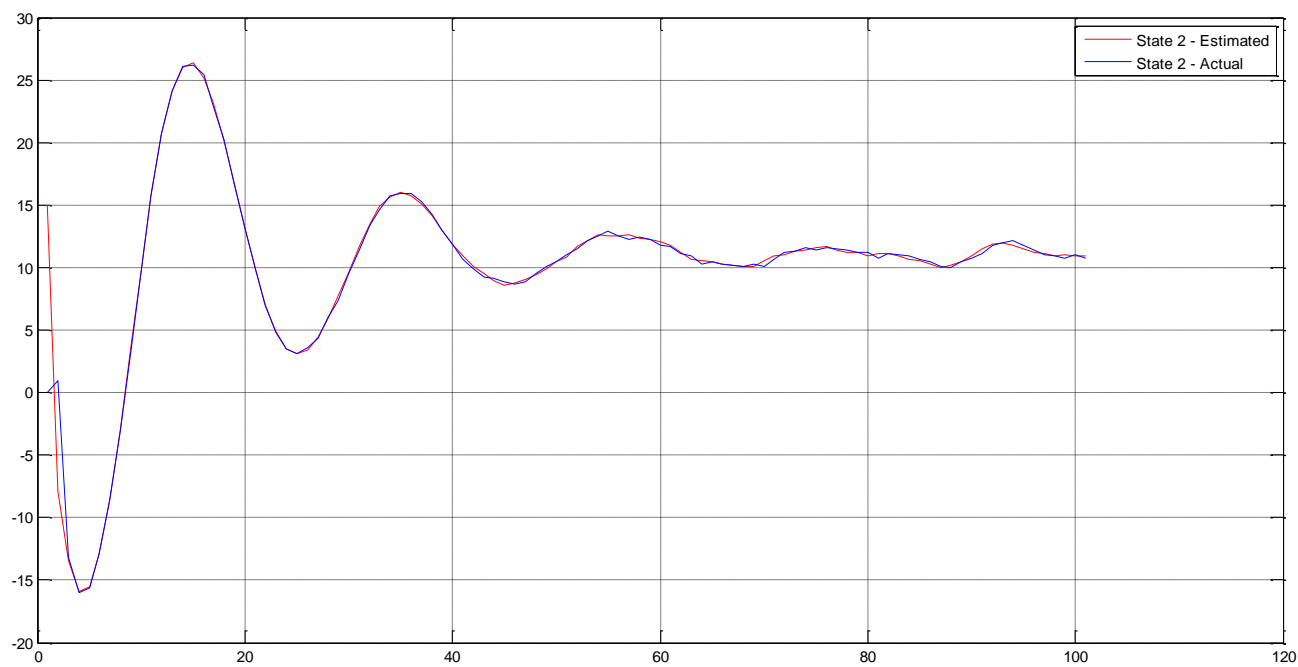
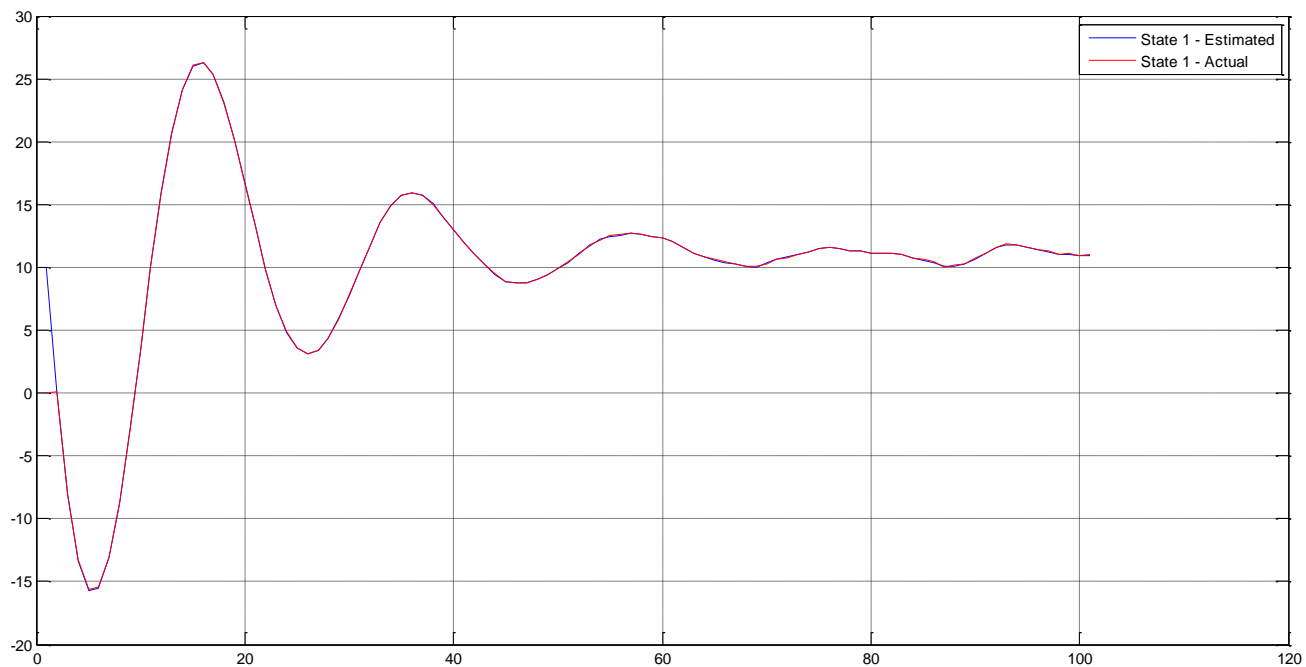
The value of gain by Steady State Solution is

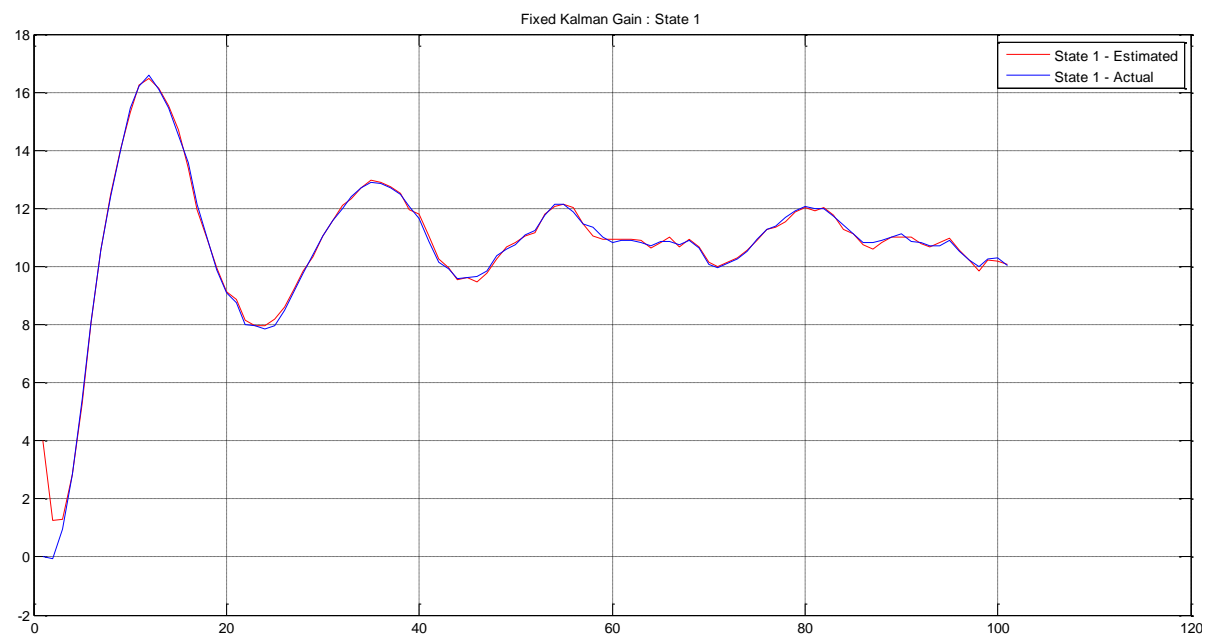
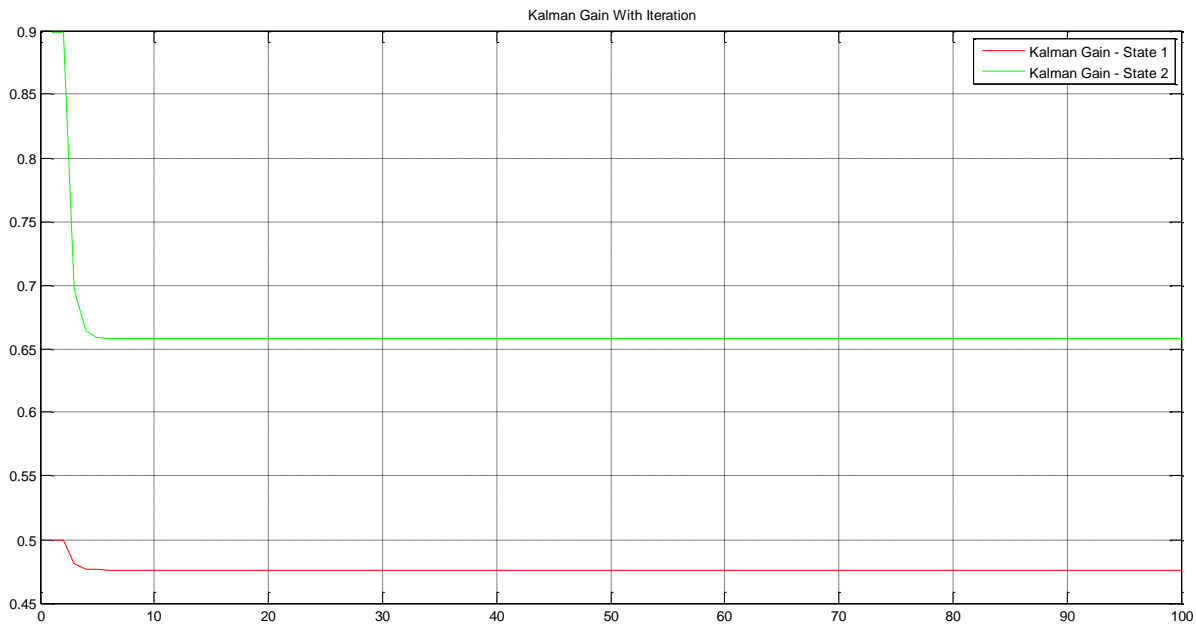
M =

0.8607

1.1552

Plot:





Fixed Kalman Gain : State 2

