
LQG Observers

NDSU ECE 463/663

Lecture #29

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Please visit [Bison Academy](#) for corresponding
lecture notes, homework sets, and solutions

Full-Order Observers

Given a dynamic system

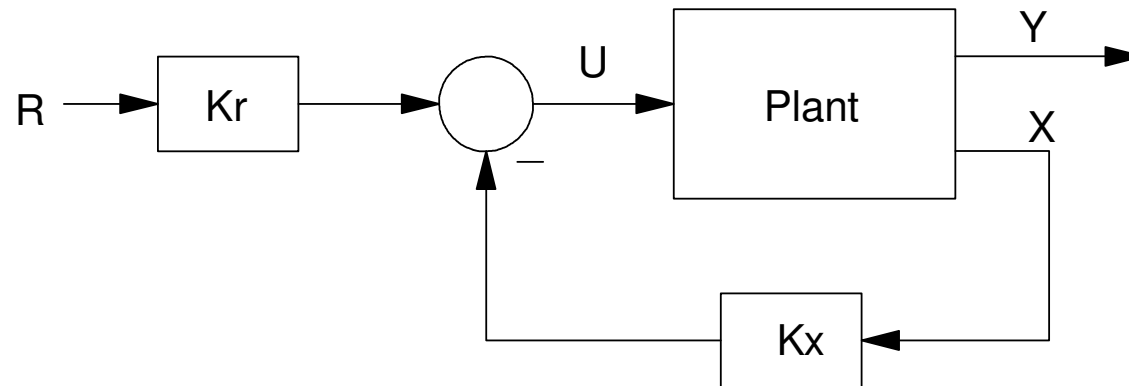
$$sX = AX + BU$$

$$Y = CX$$

you can stabilize the system using full-state feedback

$$U = -K_x X + K_r R$$

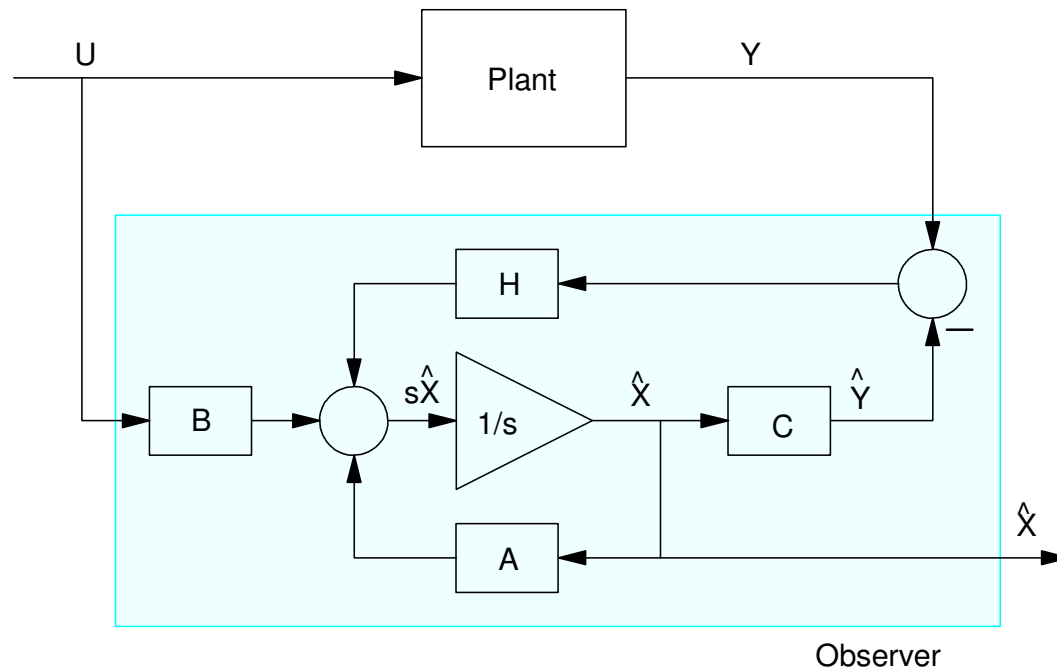
K_x , can be computed using Bass-Gura (pole placement) or LQR methods.



If the states are not measured, they can be estimated

$$sX_e = AX_e + BU + H(Y - Y_e)$$

$$Y_e = CX_e$$



Bass Gura can be used to find H

LQR can also be used to find H

- A is replaced with A^T
- B is replaced with C^T , and
- The gain you compute is H^T

$$s \begin{bmatrix} X \\ X_e \end{bmatrix} = \begin{bmatrix} A & 0 \\ HC & A - HC \end{bmatrix} \begin{bmatrix} X \\ X_e \end{bmatrix} + \begin{bmatrix} B \\ B \end{bmatrix} U$$

Example: Metal Bar (4th order RC filter).

First, pick Q and R. All states matter, so let $Q = I$

- $Q = I$
- $R = 1$

```
Q = diag([1,1,1,1]);  
R = 1;  
H = lqr(A', C', Q, R) '
```

```
0.3713  
0.3116  
0.2826  
0.2705
```

```
eig(A - H*C)
```

```
-3.5606  
-2.4418  
-1.1441  
-0.2248
```

Dominant pole for the observer

The augmented system (plant plus observer) is then

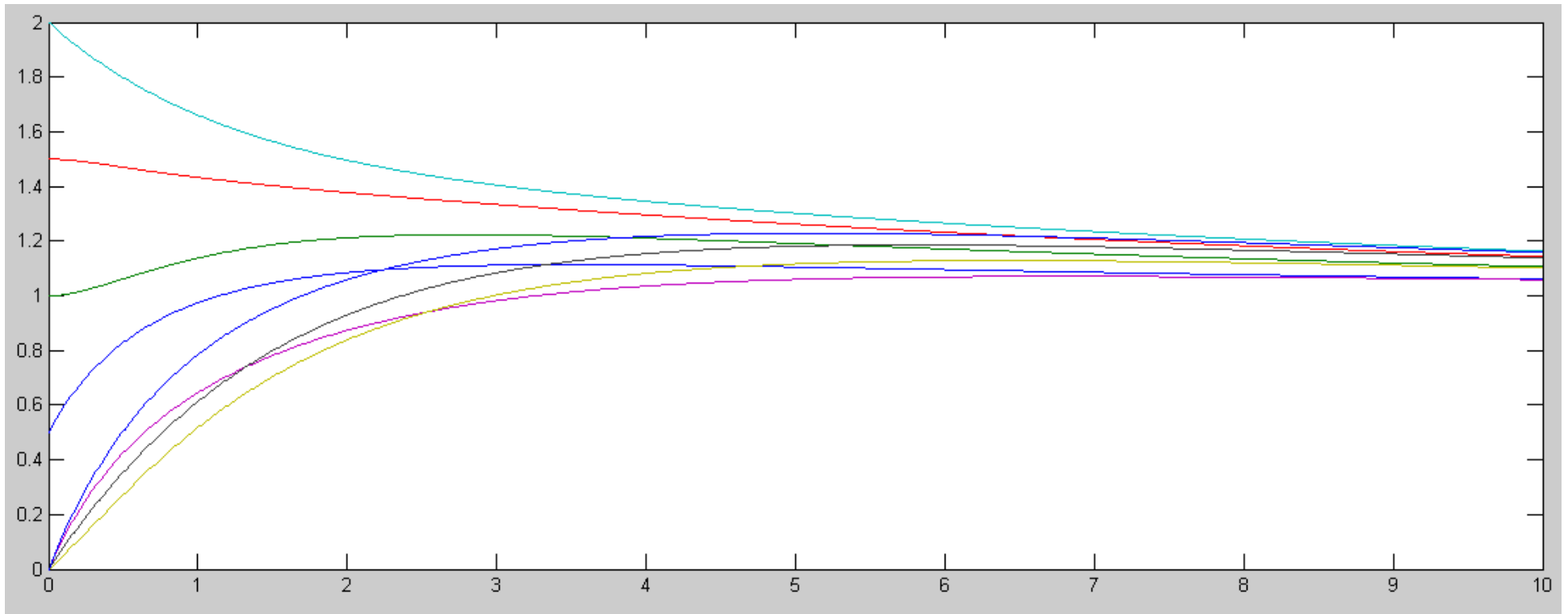
```
A8 = [A, zeros(4,4); H*C, A-H*C]
B8 = [B; B];
C8 = eye(8,8);
D8 = zeros(8,1);
```

Add initial conditions:

```
X0 = [0.5;1;1.5;2; 0;0;0;0]
```

```
0.5000
1.0000  initial conditions of the plant
1.5000
2.0000
- - - - -
      0
      0  initial conditions of the observer
      0
      0
```

```
y8 = step3(A8, B8, C8, D8, t, X0, 0*t+1);  
plot(t,y8)
```



Plant and Observer States for $Q = I$, $R = 1$. Note that the states converge

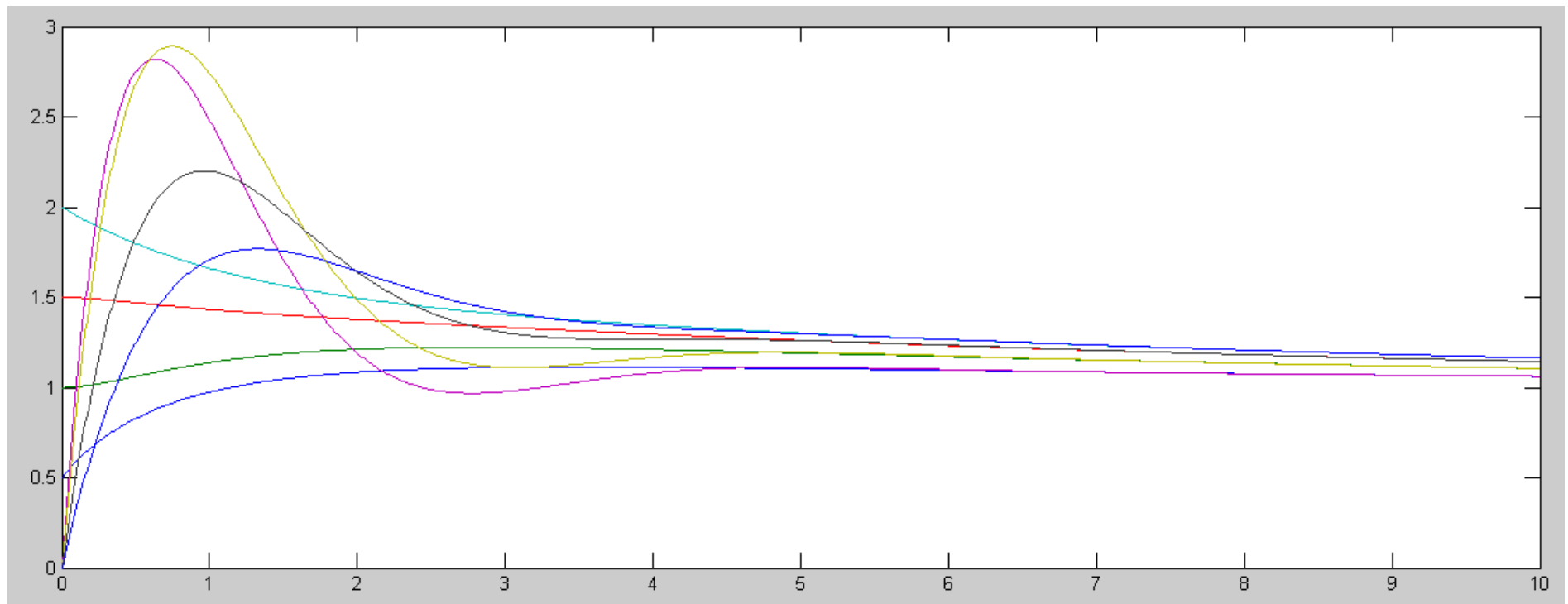
To speed up the observer, increase the weightings.

Weighting X2 works best, so go with that. (Q is somewhat arbitrary)

```
Q = diag([1,1000,1,1]);  
R = 1;  
H = lqr(A', C', Q, R)'
```

```
5.4325  
4.9030  
2.8963  
1.7915
```

```
A8 = [A, zeros(4,4); H*C, A-H*C];  
B8 = [B; B];  
C8 = eye(8,8);  
D8 = zeros(8,1);  
y8 = step3(A8, B8, C8, D8, X0, t);  
plot(t,y8);
```



Step Response with Initial Conditions: $Q = \text{diag}(1 \ 1000 \ 1 \ 1)$ $R = 1$

Example 2: Gantry System: Output = Position

$$A = [0, 0, 1, 0; 0, 0, 0, 1; 0, -19.6, 0, 0; 0, 29.4, 0, 0]$$

$$\begin{array}{cccc} 0 & 0 & 1.0000 & 0 \\ 0 & 0 & 0 & 1.0000 \\ 0 & -19.6000 & 0 & 0 \\ 0 & 29.4000 & 0 & 0 \end{array}$$

$$B = [0; 0; 1; -1]$$

$$\begin{array}{c} 0 \\ 0 \\ 1 \\ -1 \end{array}$$

$$C = [1, 0, 0, 0]$$

$$C = \begin{array}{cccc} 1 & 0 & 0 & 0 \end{array}$$

Q and R are tools you can adjust to get the response you want. Trying different weightings to see which ones speed up the system the most:

Weighting velocity seems to work best, so let $Q = \text{diag}([1, 1, 1000, 1])$

```
Q = diag([1,1,1e3,1]);  
R = 1;  
H = lqr(A', C', Q, R)'
```

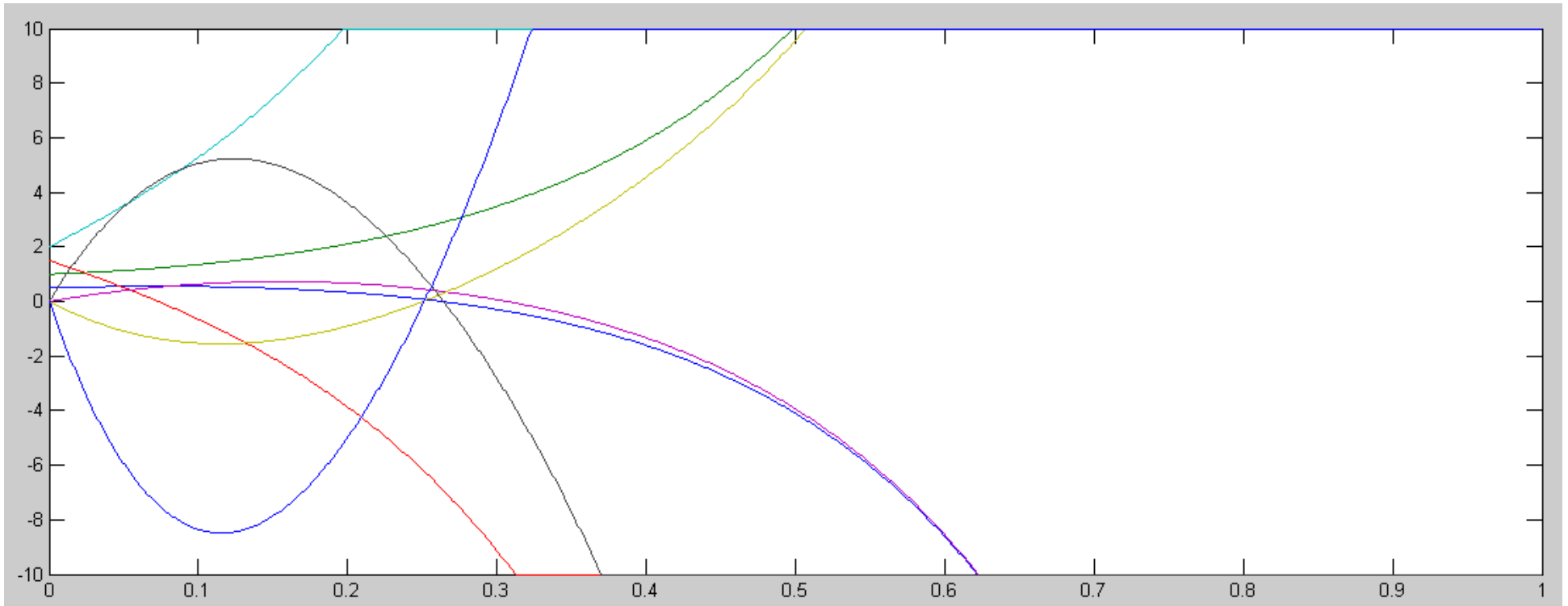
```
18.8640  
-57.8416  
177.4249  
-313.5820
```

```
eig(A - H*C)
```

```
-4.0134 + 3.9521i  
-4.0134 - 3.9521i  
-5.6431  
-5.1940
```

The plant is unstable, so the open-loop response will take off to infinity.
Regardless, the observer states do converge - even if the plant isn't.

```
A8 = [A, zeros(4,4); H*C, A-H*C];  
B8 = [B; B];  
C8 = eye(8,8);  
D8 = zeros(8,1);  
X0 = [0.5;1;1.5;2; 0;0;0;0];  
t = [0:0.001:1]';  
y8 = step2(A8, B8, C8, D8, X0, t);  
plot(t,min(10,max(-10, y8)))
```



Plant and Observer States for an Open-Loop Step Response:

Add full-state feedback to stabilize the system:

```
Q = diag([1000,0,0,0]);  
R = 1;  
Kx = lqr(A, B, Q, R);  
eig(A-B*Kx)
```

```
-5.3273 + 2.4049i  
-5.3273 - 2.4049i  
-2.8173 + 1.0648i  
-2.8173 - 1.0648i
```

```
Kx =    -31.6228  -164.2922   -29.5050   -45.7942
```

```
A8 = [A, -B*Kx ; H*C, A-B*Kx-H*C];  
DC = -C*inv(A - B*Kx)*B  
-0.0316
```

```
Kr = 1/DC  
-31.6228
```

The combined system is then

$$\begin{bmatrix} sX \\ sX_e \end{bmatrix} = \begin{bmatrix} A & -BK_x \\ HC & A - BK_x - HC \end{bmatrix} \begin{bmatrix} X \\ X_e \end{bmatrix} + \begin{bmatrix} BK_r \\ BK_r \end{bmatrix} R$$

```
A8 = [A, -B*Kx; H*C, A-H*C-B*Kx];
```

```
B8 = [B*Kr; B*Kr];
```

```
C8 =
```

```
[1, 0, 0, 0, 0, 0, 0, 0; 0, 1, 0, 0, 0, 0, 0, 0; 0, 0, 0, 0, 1, 0, 0, 0; 0, 0, 0, 0, 0, 1, 0, 0];
```

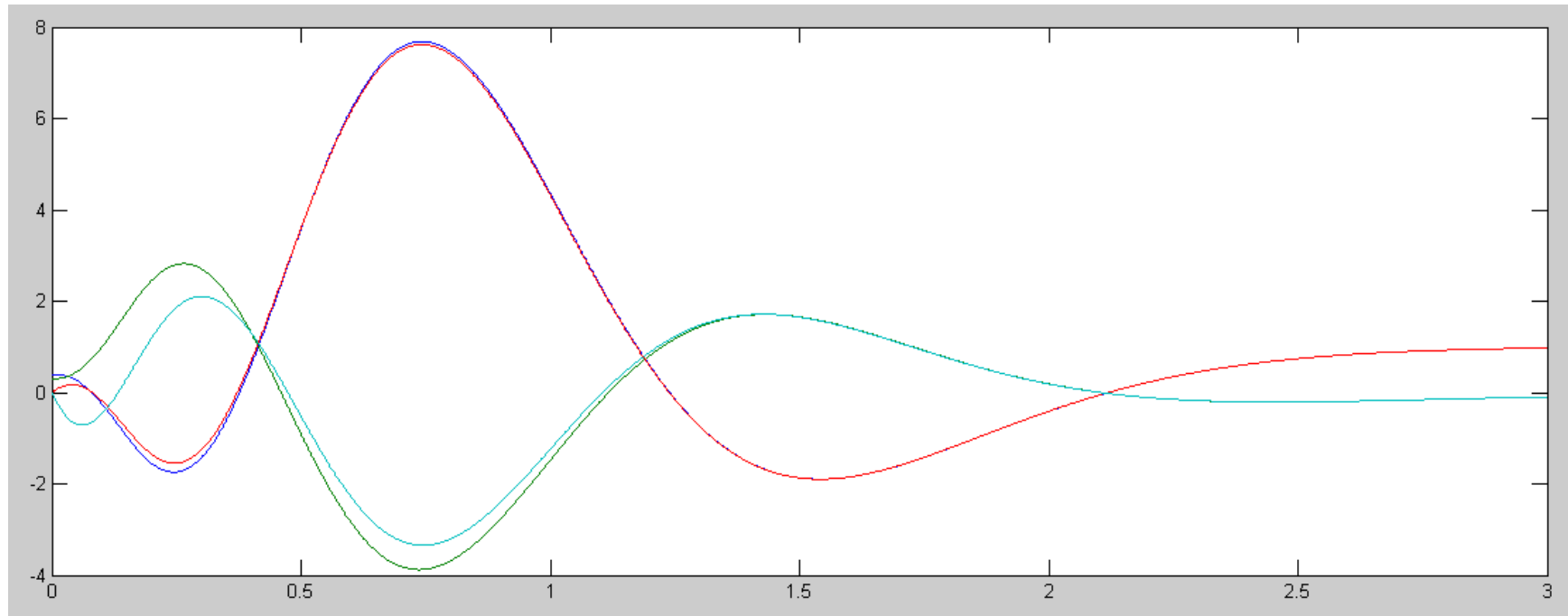
1	0	0	0	0	0	0	0	position
0	1	0	0	0	0	0	0	angle
0	0	0	0	1	0	0	0	position est
0	0	0	0	0	1	0	0	angle estimate

```
D8 = zeros(4,1);
```

```
X0 = [0.4; 0.3; 0.2; 0.1; 0; 0; 0; 0];
```

```
y8 = step2(A8, B8, C8, D8, X0, t);
```

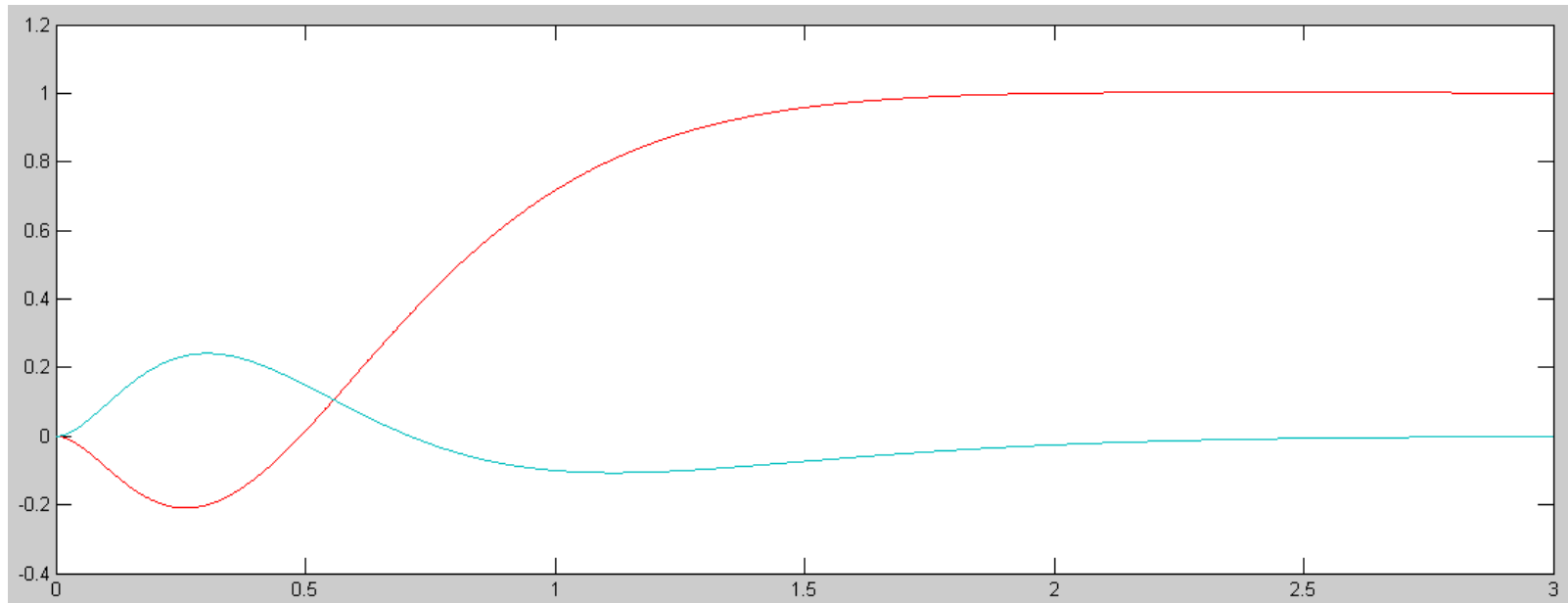
```
plot(t, min(10, max(-10, y8)))>>
```



Step Response for the Plant, Observer, and Full-State Feedback with Errors in the Initial Estimates of the States
Note that the state estimates converge to the actual states.

Once the state estimates converge, (X_0 is zero meaning no error in the state estimate) the system behaves much better:

```
y8 = step2(A8, B8, C8, D8, 0*X0, t);  
plot(t, y8)
```



Step Response for the Plant, Observer, and Full-State Feedback with No Error in the State Estimates

Observers with Multiple Outputs:

With LQR, you can use multiple outputs. For example, assume you can measure both position and angle:

```
C = [1, 0, 0, 0 ; 0 1 0 0]
```

1	0	0	0	position (x)
0	1	0	0	angle (q)

```
Q = diag([1, 1, 1e3, 1e3]);
```

```
R = diag([1, 1]);
```

```
H = lqr(A', C', Q, R)'
```

x	q
8.4497	-1.6848
-1.6848	11.9071
36.6179	-26.1793
-8.1168	71.8085

```
eig(A - H*C)
```

```
-3.9395 + 4.1662i  
-3.9395 - 4.1662i  
-6.2388 + 2.5857i  
-6.2388 - 2.5857i
```

Note that with LQR methods, you can handle two outputs without any problems: the observer gain, H , has two columns - one for position (x) and one for angle (q)

The step response of the plant plus observer plus full-state feedback is then:

```
A8 = [A, -B*Kx; H*C, A-H*C-B*Kx];
```

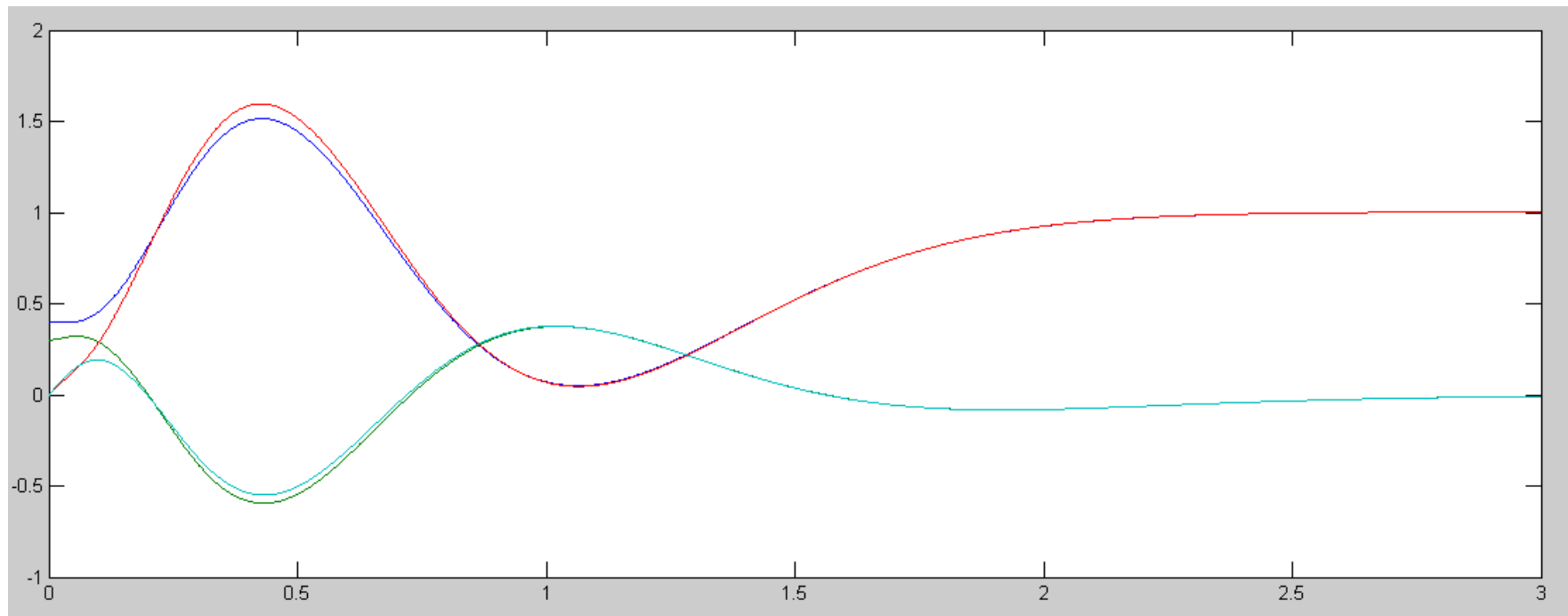
```
B8 = [B*Kr; B*Kr];
```

```
C8 = [1,0,0,0,0,0,0,0;0,1,0,0,0,0,0,0;0,0,0,0,1,0,0,0;0,0,0,0,0,1,0,0]
```

1	0	0	0	0	0	0	0	position, x
0	1	0	0	0	0	0	0	angle, q
0	0	0	0	1	0	0	0	position estimate xm
0	0	0	0	0	1	0	0	angle estimate qm

```
D8 = zeros(4,1);
```

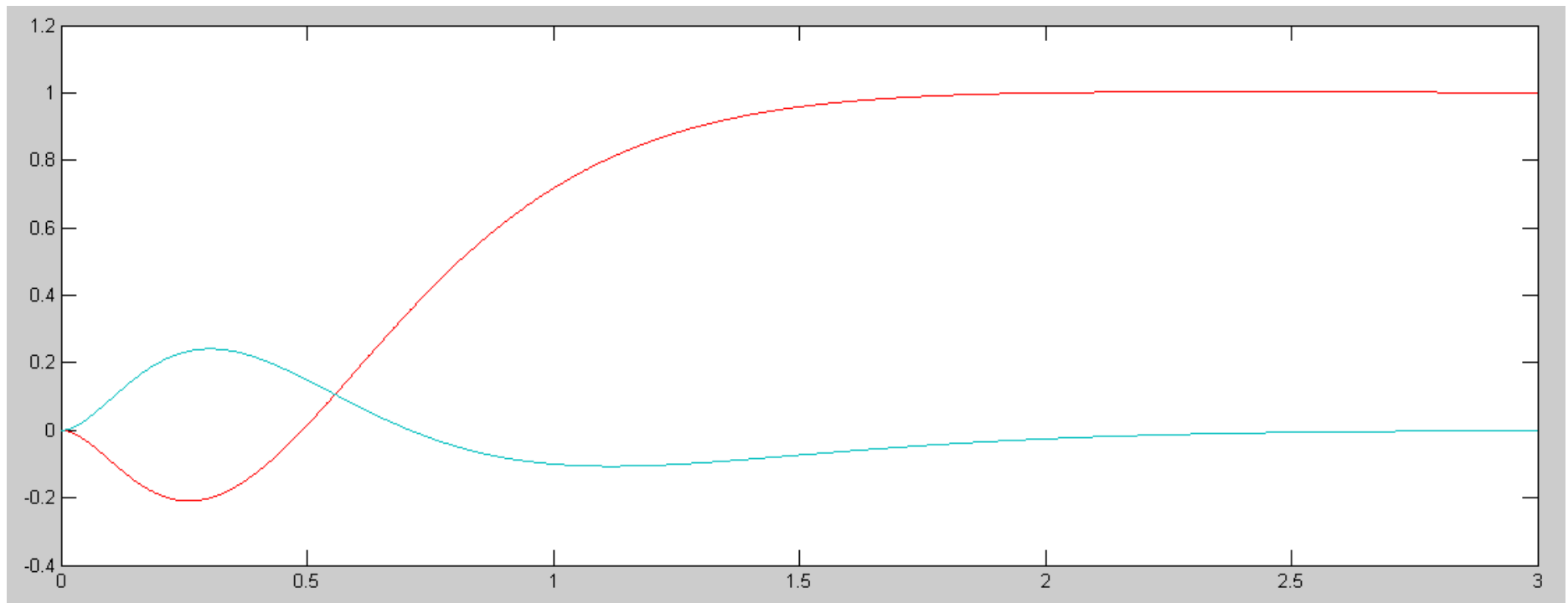
```
y8 = step2(A8, B8, C8, D8, X0, t);  
plot(t, y8)
```



Position (blue and red) and angle (green teal) for the plant and observer with initial error in the state estimates:

If initial conditions match, the tracking is better:

```
>> y8 = step2(A8, B8, C8, D8, 0*X0, t);  
>> plot(t,min(10,max(-10, y8)))
```



Step response with no error in the initial state estimates: Position (red) and angle (green)

Note that H is

x	q	
8.4497	-1.6848	x update
-1.6848	11.9071	q update
36.6179	-26.1793	sx update
-8.1168	71.8085	sq update

As you would expect

- The position sensor mostly updates position and velocity estimates
- The angle sensor mostly updates angle and angular velocity estimates
