

Chapter #1

EEE 8007

Digital Control

- **State Space Representation of Discrete Time Control Systems**
- **State Feedback**
- **Discrete Linear Quadratic Regulator (LQR)**
- **Closed Loop Estimators**

State Space Representation of Discrete Time Control Systems

Most controllers are digital, hence we need to transform the continuous systems to discrete.

Many methods for that

- 1) First Order Hold (exists in Matlab)
- 2) Tustin's Bilinear Method (exists in Matlab)
- 3) Bilinear with pre-warping (exists in Matlab)
- 4) Zero pole matched (exists in Matlab)
- 5) Impulse (exists in Matlab)
- 6) Forward where $\Phi=I+AT$, $\Gamma=BT$
- 7) Backward where $\Phi=(I-AT)^{-1}BT$, $\Gamma=C(I-AT)^{-1}$, $C=C(I-AT)^{-1}$, $D=D+(I-AT)^{-1}BT$

Forward:

$$\dot{\mathbf{X}}(kT) = \frac{\mathbf{X}((k+1)T) - \mathbf{X}(kT)}{(k+1)T - kT} = \frac{\mathbf{X}((k+1)T) - \mathbf{X}(kT)}{T}$$

$$\text{Hence } \frac{\mathbf{X}((k+1)T) - \mathbf{X}(kT)}{T} = f(\mathbf{X}(kT), \mathbf{U}(kT), kT)$$

or $\mathbf{X}((k+1)T) = Tf(\mathbf{X}(kT), \mathbf{U}(kT), kT) + \mathbf{X}(kT)$ or

$$\mathbf{X}((k+1)T) = g(\mathbf{X}(kT), \mathbf{U}(kT), kT)$$

Hence, for time-varying (linear or not) discrete time systems the state space representation is:

$$\mathbf{X}(k+1) = f(\mathbf{X}(k), \mathbf{U}(k), k)$$

$$\mathbf{Y}(k) = g(\mathbf{X}(k), \mathbf{U}(k), k)$$

If the system is linear and there is no direct coupling between the input and the output then:

$$\mathbf{X}(k+1) = \mathbf{\Phi}(k)\mathbf{X}(k) + \mathbf{\Gamma}(k)\mathbf{U}(k)$$

$$\mathbf{Y}(k) = \mathbf{C}(k)\mathbf{X}(k)$$

And for LTI systems:

$$\mathbf{X}(k+1) = \mathbf{\Phi}\mathbf{X}(k) + \mathbf{\Gamma}\mathbf{U}(k)$$

$$\mathbf{Y}(k) = \mathbf{C}\mathbf{X}(k)$$

The relation of the matrices Φ and A , Γ and B can be found in any text book and is:

$$\mathbf{\Phi} = e^{\mathbf{A}T} = \mathbf{I} + \mathbf{A}T + \frac{\mathbf{A}^2T^2}{2!} + \frac{\mathbf{A}^3T^3}{3!} + \dots$$

$$\mathbf{\Gamma} = \int_0^T e^{\mathbf{A}n} dn \mathbf{B}$$

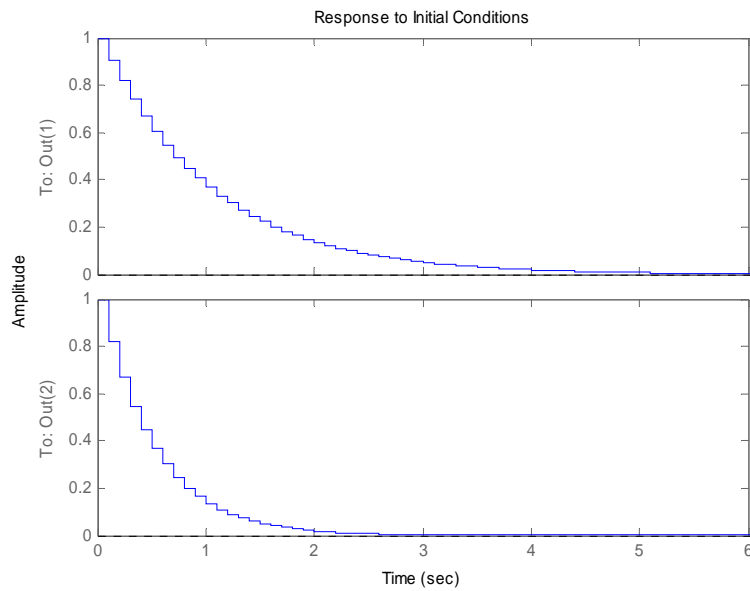
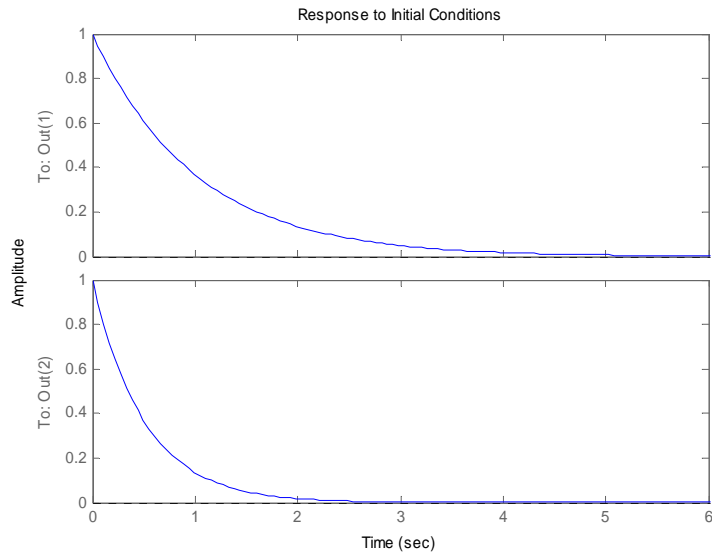
In Matlab use: `c2d`

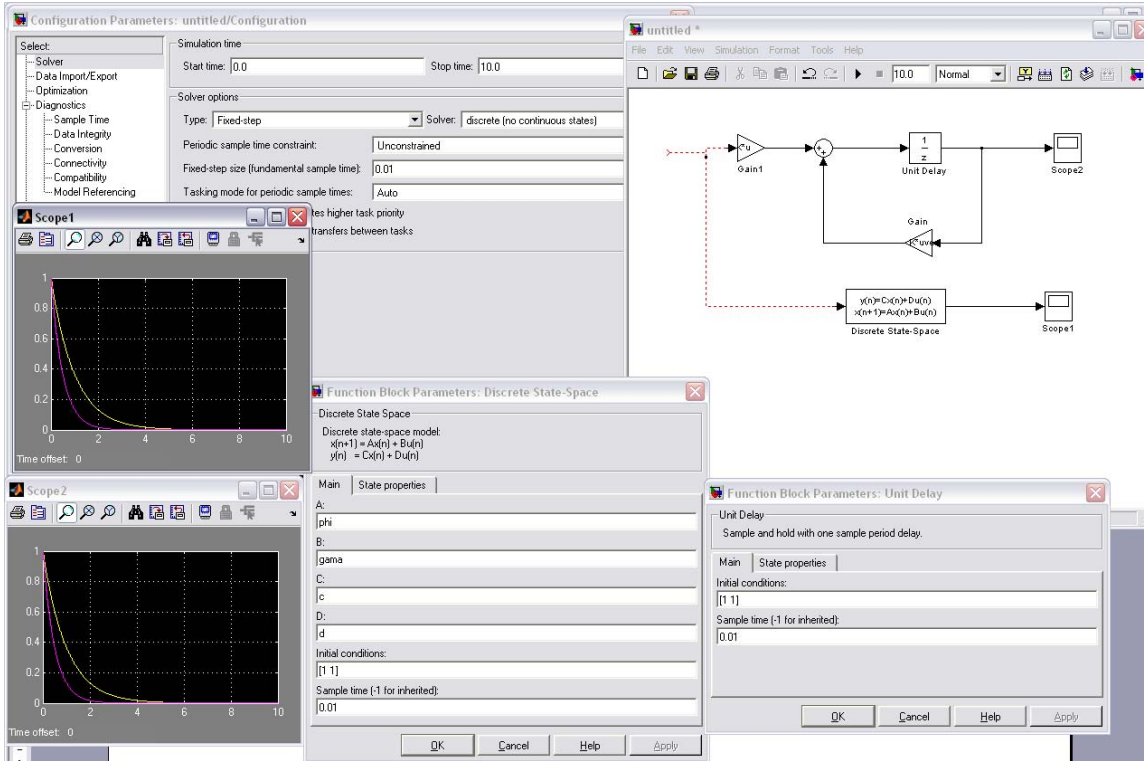
All the other concepts are the same:

TF of digital system: $\mathbf{C}[z\mathbf{I} - \mathbf{\Phi}]^{-1}\mathbf{\Gamma}$ and hence poles of system: $eig(\mathbf{\Phi})$

```
A=[-1 0;0 -2]; B=[1;0]; C=eye(2); D=0;
sys_c=ss(A,B,C,D);
sys_d=c2d(sys_c, 0.1);
initial(sys_c,[1,1]);
figure
```

```
initial(sys_d,[1,1]);
```





State Feedback

By following the same steps, as in continuous time systems:

$\dot{\mathbf{X}}(k + 1) = \Phi_{cl} \mathbf{X}(k) + \Gamma_{cl} \mathbf{U}(k)$ where $\Phi_{cl} = \Phi - \Gamma K$, $\Gamma_{cl} = \Gamma F$, $C_{cl} = C - DK$, $D_{cl} = DF$. Thus we have described the closed loop system into the same form as the open loop system.

Since the poles of the OL can be found by the $|zI - \Phi| = 0$, the close loop poles are: $|zI - (\Phi - \Gamma K)| = 0$.

If we want the close loop poles at $P = [P_1 \ P_2 \ \dots \ P_n]^T$, (n is the order of the system) then:

$|zI - \Phi + \Gamma K| = (z - P_1)(z - P_2) \dots (z - P_n) \Leftrightarrow K = [K_1 \ K_2 \ \dots]^T$. Pole placement

Check **CONTROLLABILITY**

$$S = \begin{bmatrix} \Gamma & \Phi\Gamma & \Phi^2\Gamma & \dots & \Phi^{n-1}\Gamma \end{bmatrix}$$

Example:

```

clc
close all
clear all
A=[0 1 0;0 -1 1;0 0 -5];B=[0 0 5]';C=eye(3);D=0
sys_c=ss(A,B,C,D)
initial(sys_c,[1 1 1])

K=place(A,B,[-10 -5+2*j -5-2*j])

Acl=A-B*K;
sys_c_cl=ss(Acl,[],C,[])
figure
initial(sys_c_cl,[1 1 1])

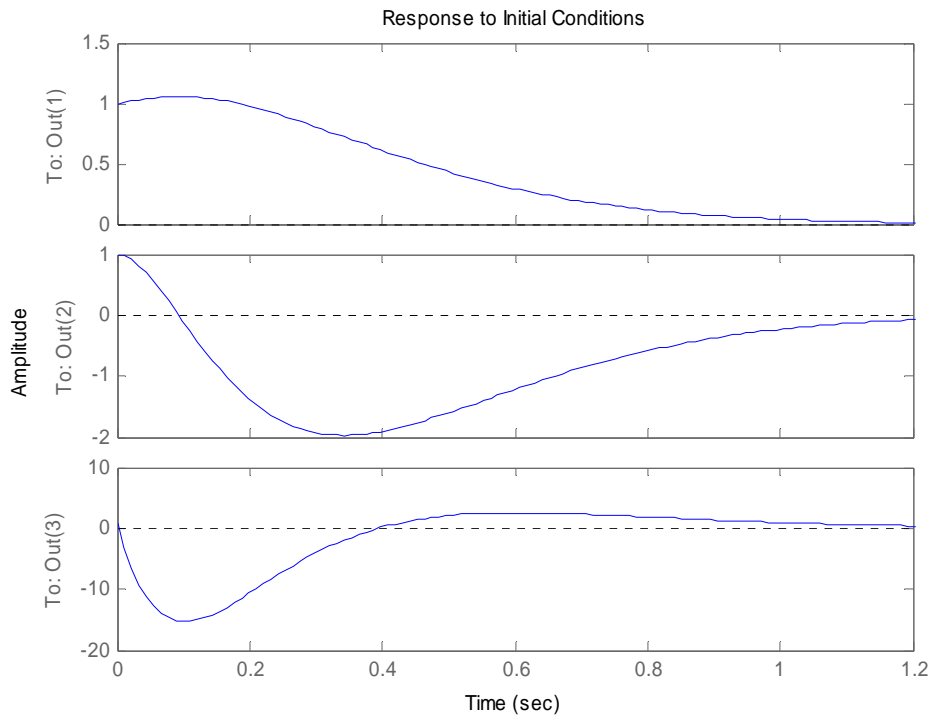
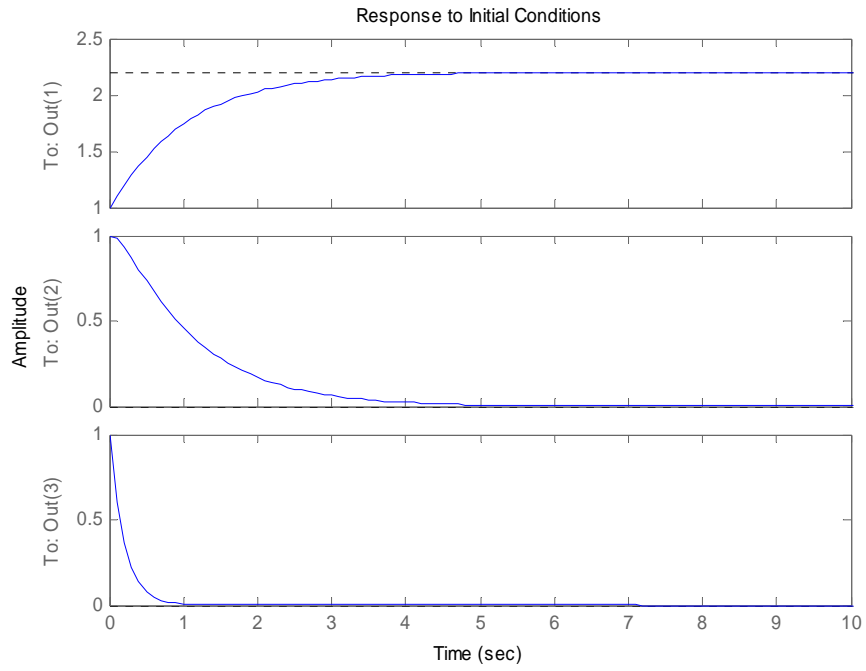
Ts=0.01;
sys_d=c2d(sys_c,Ts)

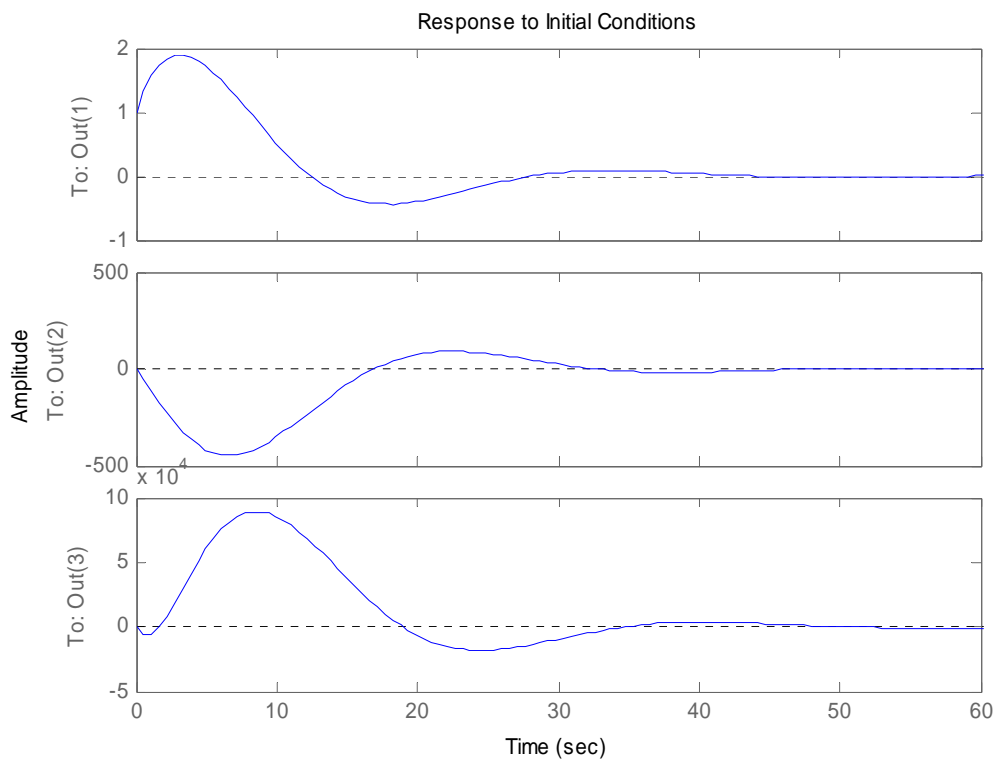
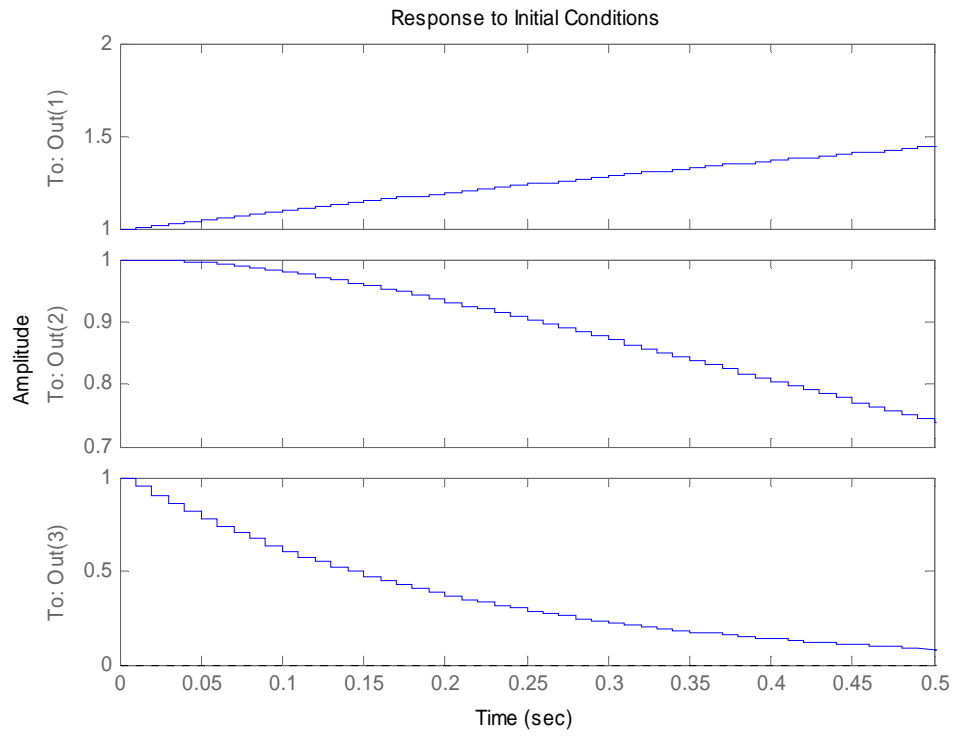
figure
initial(sys_d,[1 1 1])

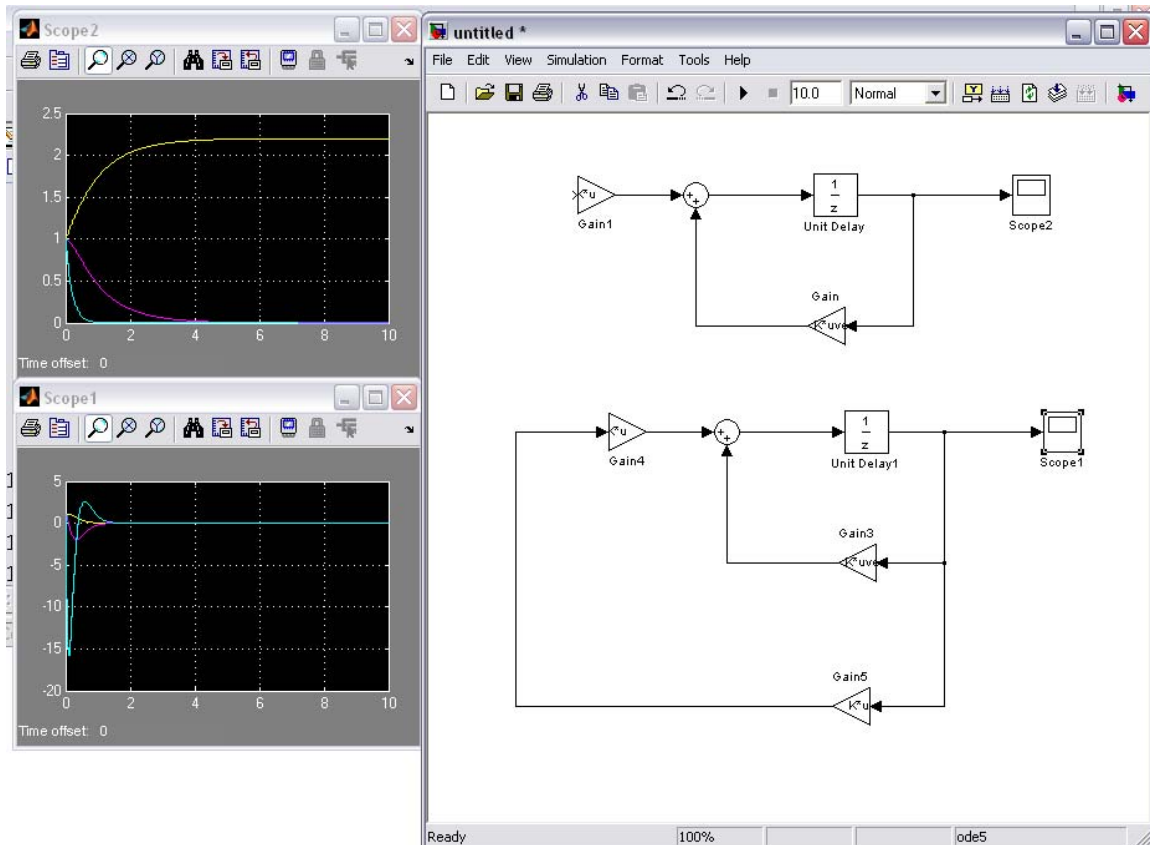
[F,G,C,D,Ts1]=ssdata(sys_d)
rank(ctrb(F,G))
K_d=place(F,G,[-0.5 -0.1+0.2*j -0.1-0.2*j])

Fcl=F-G*K_d;
figure
sys_d_cl=ss(Fcl,[],C,[])
initial(sys_d_cl,[1 1 1])

```







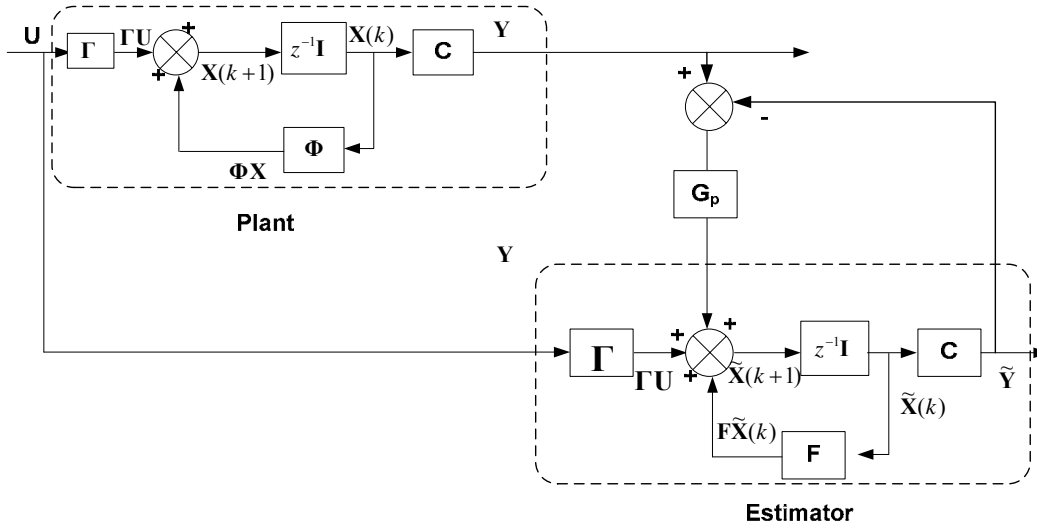
LQR Control

To use an LQR controller use the command:

```
>>dlqr()
```

Discrete Time Estimators

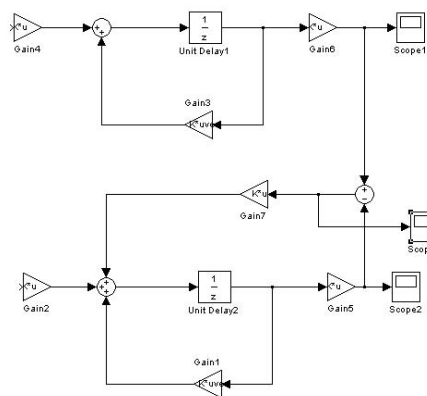
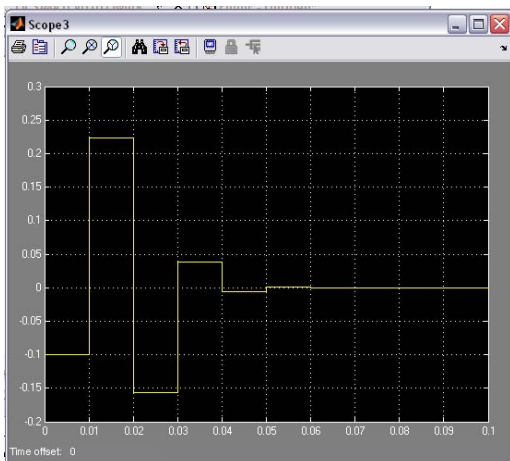
The same method can be applied at the DTE:



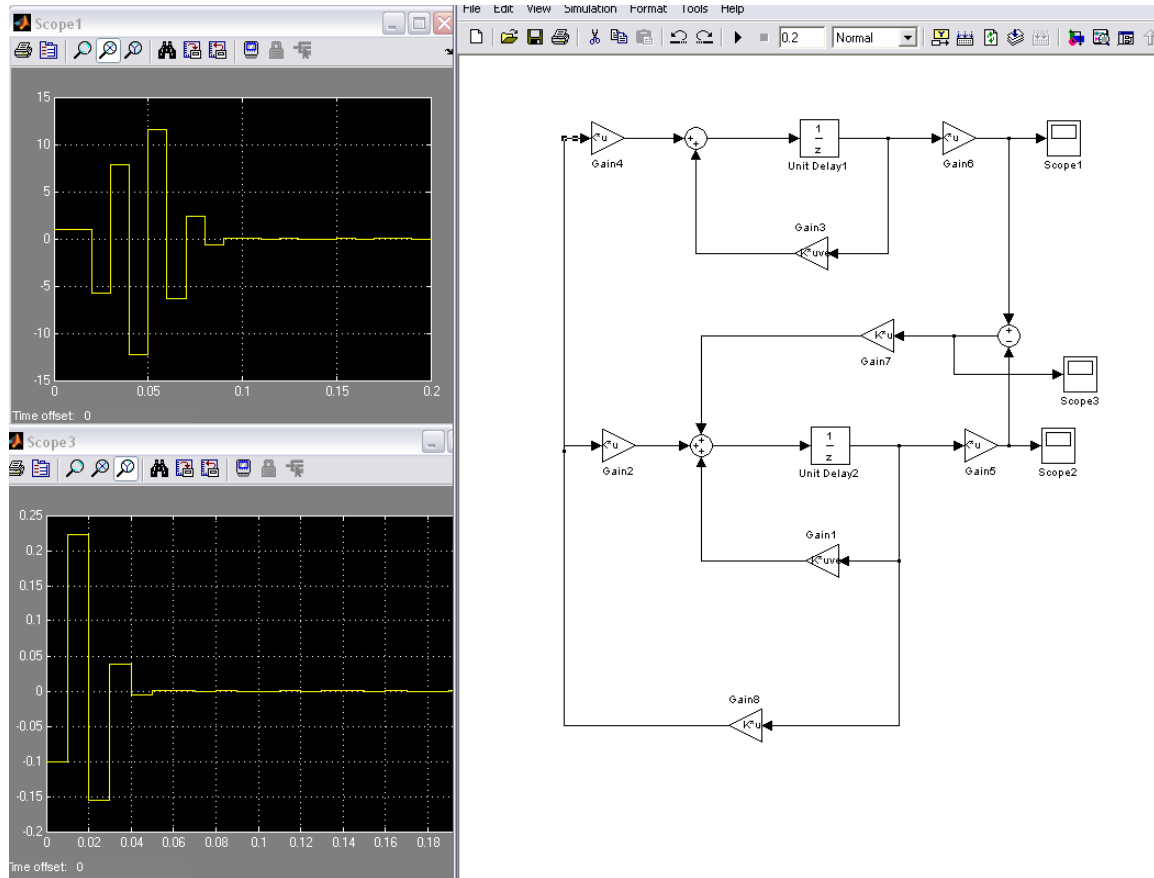
End the error dynamics are: $e(k + 1) = (\Phi - G_p C)e(k)$.

10% error of ICs:

```
F=[1 0.01 0; 0 0.99 0.0097; 0 0 0.9512];G=[0 0.002
0.0488]'; C=eye(1,3);D=0;
rank( obsv(F,C) )
Gp=place(F',C',[-0.1 -0.1+0.1*j -0.1-0.1*j])'
```



```
K_d=place(F,G,[-0.2 -0.2+0.2*j -0.2-0.2*j])
```



As it can be seen we estimate the $\mathbf{X}(k)$ from the estimation of $\mathbf{X}(k+1)$ after a delay. To calculate the estimation of $\mathbf{X}(k+1)$ we use $\mathbf{Y}(k)$ and not $\mathbf{Y}(k+1)$. So this estimator predicts $\mathbf{X}(k+1)$ from $\mathbf{Y}(k)$. For this reason is called prediction Estimator (or priory estimator).

Current Estimators:

Until now the estimated state vector $\mathbf{X}(k)$ was calculated by the one sample delay of $\mathbf{X}(k+1)$. The $\mathbf{X}(k+1)$ was estimated by the measurements of $\mathbf{Y}(k)$. What about if we use the sample $\mathbf{Y}(k+1)$ to calculate the $\mathbf{X}(k+1)$ and hence the $\mathbf{X}(k)$. Then this estimator is called current estimator – or “posteriori” estimator (used in **Kalman Filter**).

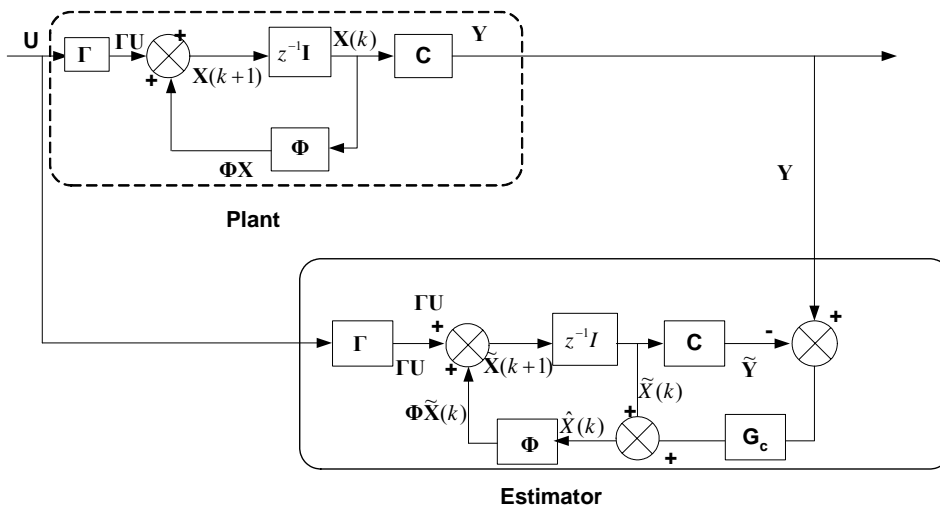
The current estimator will be symbolised by: $\hat{\mathbf{X}}(k)$ and its relation with the $\tilde{\mathbf{X}}(k)$ is: $\hat{\mathbf{X}}(k) = \tilde{\mathbf{X}}(k) + G_c(\mathbf{Y}(k) - \mathbf{C}\tilde{\mathbf{X}}(k))$

Hence it is the same as the predictor plus some extra information. But it must be noticed that for the calculation of the predictor the current is used:

$$\tilde{\mathbf{X}}(k) = \Phi\hat{\mathbf{X}}(k-1) + \Gamma\mathbf{U}(k-1) \text{ (OL estimation of current value)}$$

The estimator is predicting the states and then it is using the current information of $\mathbf{Y}(k)$ to find the correct the estimating value

Hence:



And the error dynamics are: $\mathbf{e}(k+1) = (\Phi - \mathbf{G}_c \mathbf{C} \Phi) \mathbf{e}(k)$

The previous example with a current estimator:

```
>> Gc=place(F',F'*C',[-0.1 -0.1+0.1*j -0.1-0.1*j])'
```

