ELEC4410

Control System Design

Lecture 19: Feedback from Estimated States and Discrete-Time Control Design

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Outline

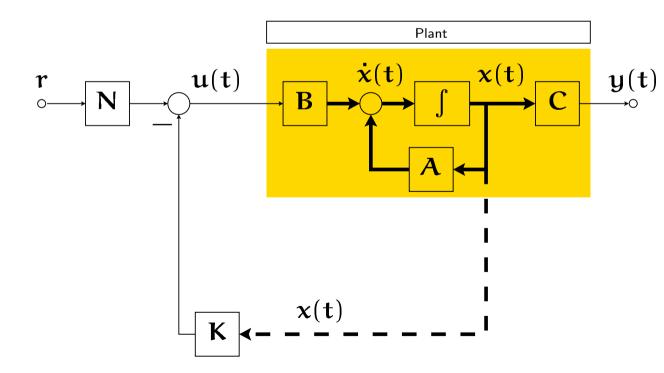
- Feedback from Estimated States
- Discrete-Time Control Design
- Dead-Beat Control



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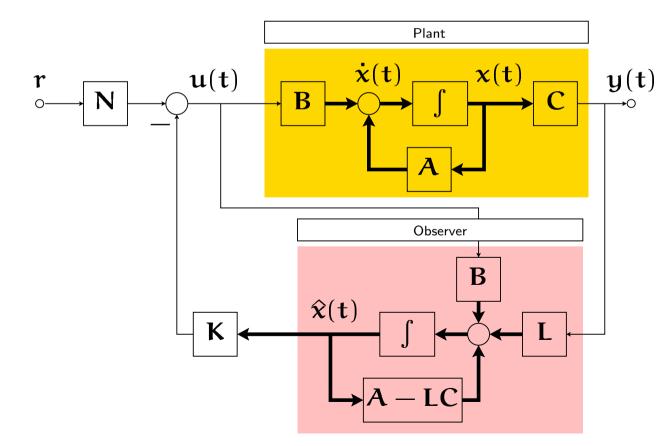
▶ a state feedback gain K





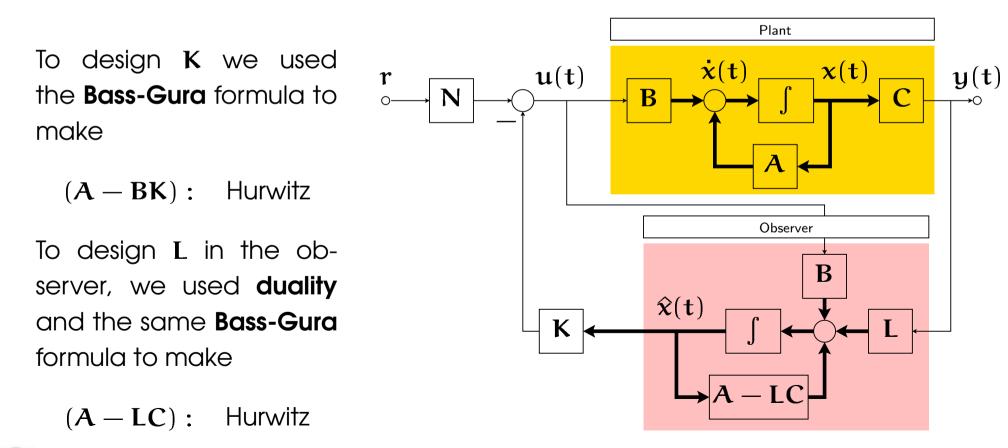
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- By designing L such that (A LC) is Hurwitz, we guarantee that the observer will be asymptotically stable, and the estimate of the states $\hat{x}(t)$ will converge to the real states x(t) as $t \to \infty$.
- But K and the observer are designed independently...Will they work the same when we put them together in a feedback from estimated states scheme?



Three basic questions arise regarding **feedback from estimated states:**

The closed-loop eigenvalues were set as those of (A - BK) by using state feedback

$$u = -Kx.$$

Would we still have the same eigenvalues if we do **feedback from estimated states**

$$\mathbf{u} = -\mathbf{K}\hat{\mathbf{x}}?$$



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- Would the interconnection affect the **observer** eigenvalues, those of (A LC)?
- What would be the effect of the observer in the closed-loop transfer function?

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To answer these questions we take a look at the state equations of the full system, putting together plant and observer, that is,

$$\dot{\mathbf{x}} = \mathbf{A}\mathbf{x} + \mathbf{B}\mathbf{u}$$
 plant
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after replacing $\mathbf{u} = \mathbf{N}\mathbf{r} - \mathbf{K}\hat{\mathbf{x}}$. Packaging these equations in a more compact form we have

$$\begin{bmatrix} \dot{\mathbf{x}} \\ \dot{\hat{\mathbf{x}}} \end{bmatrix} = \begin{bmatrix} \mathbf{A} & -\mathbf{B}\mathbf{K} \\ \mathbf{L}\mathbf{C} & (\mathbf{A} - \mathbf{L}\mathbf{C} - \mathbf{B}\mathbf{K}) \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix} + \begin{bmatrix} \mathbf{B}\mathbf{N} \\ \mathbf{B}\mathbf{N} \end{bmatrix} \mathbf{r}$$
$$\mathbf{y} = \begin{bmatrix} \mathbf{C} & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix}$$



Let's make a change of coordinates, so that the new coordinates are the plant state x and the estimation error $\varepsilon = x - \hat{x}$,

$$\begin{bmatrix} \mathbf{x} \\ \mathbf{\epsilon} \end{bmatrix} = \begin{bmatrix} \mathbf{x} \\ \mathbf{x} - \hat{\mathbf{x}} \end{bmatrix} = \underbrace{\begin{bmatrix} \mathbf{I} & \mathbf{0} \\ \mathbf{I} & -\mathbf{I} \end{bmatrix}}_{\mathbf{P}} \begin{bmatrix} \mathbf{x} \\ \hat{\mathbf{x}} \end{bmatrix}$$

Note that $P^{-1} = \begin{bmatrix} I & 0 \\ I & -I \end{bmatrix}^{-1} = \begin{bmatrix} I & 0 \\ I & -I \end{bmatrix} = P$.



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Note that $P^{-1} = \begin{bmatrix} I & 0 \\ I & -I \end{bmatrix}^{-1} = \begin{bmatrix} I & 0 \\ I & -I \end{bmatrix} = P$. With this equivalence transformation we get the matrices in the new coordinates as

$$\bar{A}_{KL} = PA_{KL}P^{-1} = \begin{bmatrix} (A - BK) & BK \\ 0 & (A - LC) \end{bmatrix}, \quad \bar{B}_{KL} = PB_{KL} = \begin{bmatrix} BN \\ 0 \end{bmatrix}$$
$$\bar{C}_{KL} = C_{KL}P^{-1} = \begin{bmatrix} C & 0 \end{bmatrix}$$



The full system in the new coordinates is thus represented as

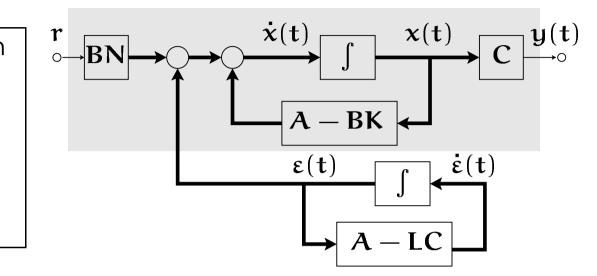
$$\begin{bmatrix} \dot{x} \\ \dot{\epsilon} \end{bmatrix} = \begin{bmatrix} (A - BK) & BK \\ 0 & (A - LC) \end{bmatrix} \begin{bmatrix} x \\ \epsilon \end{bmatrix} + \begin{bmatrix} BN \\ 0 \end{bmatrix} r$$
$$y = \begin{bmatrix} C & 0 \end{bmatrix} \begin{bmatrix} x \\ \epsilon \end{bmatrix}$$

Because \bar{A}_{KL} is **triangular**, its eigenvalues are the union of those of (A - BK) and (A - LC).

Controller and observer **do not affect each other** in the interconnection.



Note that the estimation error is **uncontrollable**, hence the observer eigenvalues will not appear in the closed loop transfer function.





The property of independence between control and state estimation is called the **Separation Principle**

Separation Principle: The design of the state feedback and the design of the state estimator can be carried out independently.

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Separation Principle: The design of the state feedback and the design of the state estimator can be carried out independently.

- The eigenvalues of the closed-loop system are as designed by the state feedback law, unaffected by the use of a state estimator.
- The eigenvalues of the observer are unaffected by the state feedback law.

The closed-loop transfer function will only have the eigenvalues arising from (A - BK), since the estimation error is **uncontrollable**,

 $G_{cl}(s) = C(sI - A + BK)^{-1}BN.$

Transients in state estimation, however, will be seen at the output, since the estimation error is **observable**.



Outline

Feedback from Estimated States

- Discrete-Time Control Design
- Dead-Beat Control



Discrete-Time Control Design

For discrete-time state equations

$$x[k+1] = Ax[k] + Bu[k], \quad y[k] = Cx[k],$$

the design procedure for a state feedback law u[k] = -Kx[k] is the same as for continuous-time systems.

The same goes for a discrete-time state observer,

$$\hat{\mathbf{x}}[\mathbf{k}+\mathbf{1}] = (\mathbf{A} - \mathbf{L}\mathbf{C})\hat{\mathbf{x}}[\mathbf{k}] + \mathbf{B}\mathbf{u}[\mathbf{k}] + \mathbf{L}\mathbf{y}[\mathbf{k}].$$

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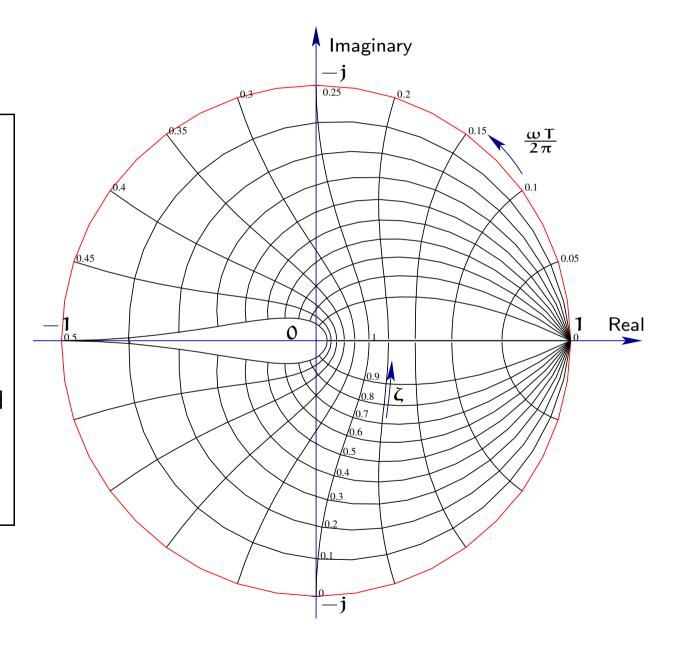
A practical rule to choose desired discrete-time eigenvalues is

1. choose a desired location in **continuous-time**, say p_i

2. translate it to discrete-time using the relation $\lambda_i = e^{p_i T}$

Discrete-Time Control Design

Loci with constant damping ζ and constant frequencies ω in the discrete complex plane. In MATLAB this grid may be obtained with the zgrid command.



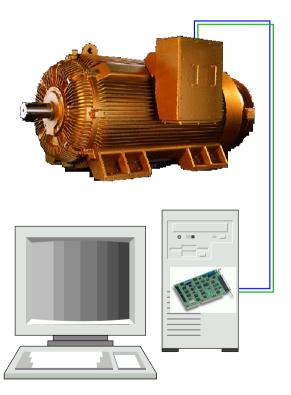
Example (Discrete-time speed control of a DC motor). We return to the DC motor we considered in the examples of the last lecture. We will suppose that the motor is to be controlled with a PC. Hence, the controller has to be **discrete-time**.

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This time, with a view to design a discrete-time control law, we first **discretise** the continuoustime model

$$\frac{d}{dt} \begin{bmatrix} \omega(t) \\ i(t) \end{bmatrix} = \begin{bmatrix} -10 & 1 \\ -0.02 & -2 \end{bmatrix} \begin{bmatrix} \omega(t) \\ i(t) \end{bmatrix} + \begin{bmatrix} 0 \\ 2 \end{bmatrix} V(t)$$
$$y = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} \omega(t) \\ i(t) \end{bmatrix}$$

The first **design parameter** to define is the **sam**-**pling period T**.



Example (Discrete design, step 1: choice of sampling period). From Shannon's Sampling Theorem, the sampling frequency $\omega_s = 2\pi/T$ should be at least twice the bandwidth of the closed-loop system (because we will change the system bandwidth with the control action).

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The specification that we had for the previous continuous-time design was a settling time t_s of about 1s. The rule based on Shannon's Theorem would then give a sampling time of less than T = 0.5s. In practice T is chosen at least 10 to 20 times faster than the desired closed-loop settling time. Here we choose

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Note that the maximum sampling speed is limited by the available computer clock frequency, and thus the time required for computations and signal processing operations.

Example (Discrete design, step 2: discretisation of the plant). Having defined the sampling-time **T**, we compute the discrete-time state matrices

$$A_d = e^{At}$$
, and $B_d = \int_0^T e^{A\tau} B d\tau$

From MATLAB, we do [Ad,Bd]=c2d(A,B,0.1) and obtain

$$A_{d} = \begin{bmatrix} 0.3678 & 0.0563 \\ -0.0011 & 0.8186 \end{bmatrix}, \quad B_{d} = \begin{bmatrix} 0.0068 \\ 0.1812 \end{bmatrix}$$

As we can verify, the open loop **discrete-time** eigenvalues are

$$0.3679 = e^{(-9.9975 \times 0.1)}$$
 and $0.8185 = e^{(-2.0025 \times 0.1)}$



Example (Discrete design, step 3: design of discrete feedback gain). The discrete-time characteristic polynomial is

$$\Delta(z) = (z - e^{-9.9975T})(z - e^{-2.0025T}) = z^2 - 1.1864z + 0.3011$$

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For the desired discrete characteristic polynomial, we first obtain the **discrete mapping** $p \rightarrow e^{pT}$ of the eigenvalues $p_{1,2} = -5 \pm j$ specified for the continuous-time system, i.e.,

$$\delta \pm j\gamma = e^{(-5.j).0.1} = 0.6035 \pm j0.0605$$

which yield the desired discrete characteristic polynomial

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$$\Delta_{\rm K}(z) = z^2 - 1.2070z + 0.3678.$$

From the coefficients of $\Delta_K(z)$ and $\Delta(z)$ we get the discrete state feedback gain

$$\mathbf{\bar{K}}_{d} = \begin{bmatrix} (-1.2070 + 1.1864) & (0.3678 - 0.3011) \end{bmatrix} = \begin{bmatrix} -0.0206 & 0.0667 \end{bmatrix}$$

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Example (Discrete design, step 3, continuation). We then compute, just following the same procedure used in continuous-time, from the **discrete** matrices A_d and B_d , the **discrete controllability matrices** C_d and \overline{C}_d ,

$$\mathcal{C}_{d} = \begin{bmatrix} 0.0068555 & 0.0127368 \\ 0.1812645 & 0.1483875 \end{bmatrix}, \text{ and } \bar{\mathcal{C}}_{d} = \begin{bmatrix} 1 & -1.1864 \\ 0 & 1 \end{bmatrix}^{-1}$$

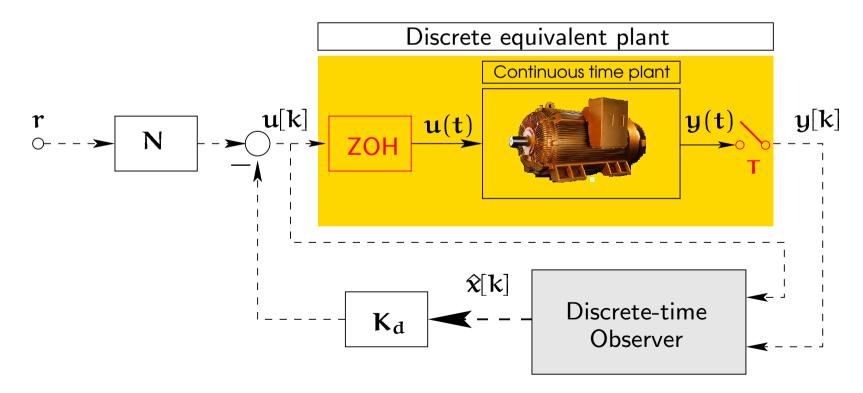
We thus obtain, in the original coordinates, the **discrete-time** feedback gain

$$K_{d} = \bar{K}_{d}\bar{\mathcal{C}}_{d}\mathcal{C}_{d}^{-1} = \begin{bmatrix} 8.3011164 & -0.4270676 \end{bmatrix}$$

As can be verified with MATLAB, $(A_d - B_d K_d)$ will have the desired discrete eigenvalues.

Discrete-Time Control Design: Example

Example (Discrete design: continuation). In a similar fashion, we can carry out the design of the discrete-time observer, based on the discrete-time model of the plant. The output feedback design is finally implemented on the **continuous-time** plant through a Zero Order Hold and a Sampler.



Discrete-Time Control Design: Example

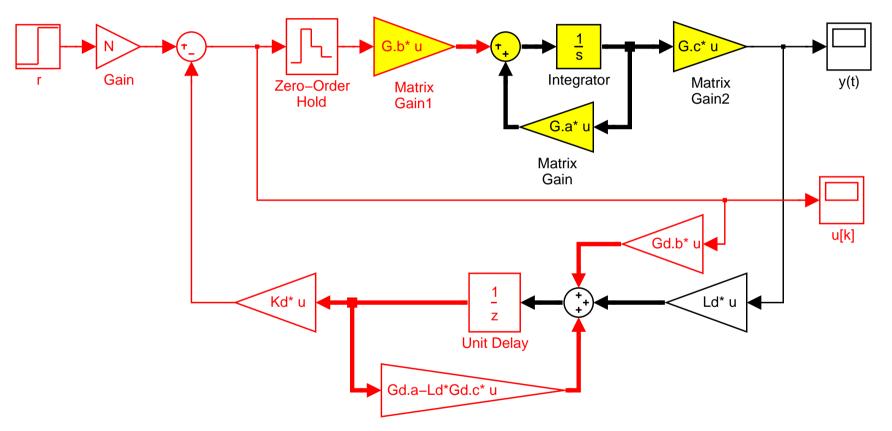
Example (Discrete design: MATLAB script). We used the following MATLAB script to compute the gains and run the simulations.

```
% Continuous-time matrices
A = [-10 \ 1; -0.02 \ -2]; B = [0; 2]; C = [1 \ 0]; D = 0;
G=ss(a,B,C,D); % state space system definition
T=0.1; % Sampling time
Gd=c2d(G,T,'zoh') % discretisation
% Discrete-time feedback gain
Kd=place(Gd.a,Gd.b,exp([-5-i,-5+i]*T))
% Discrete-time observer gain
Ld=place(Gd.a',Gd.c',exp([-6-i,-6+i]*T))'
% steady state error compensation
N=inv(Gd.c*inv(eye(2)-Gd.a+Gd.b*Kd)*Gd.b)
% Run simulink diagram
sim('dmotorOFBK')
% Plots (after simulations have been run)
subplot(211),plot(y(:,1),y(:,2));grid
subplot(212),stairs(u(:,1),u(:,2));grid
```

Discrete-Time Control Design

Example (Discrete design: SIMULINK diagram). We used the

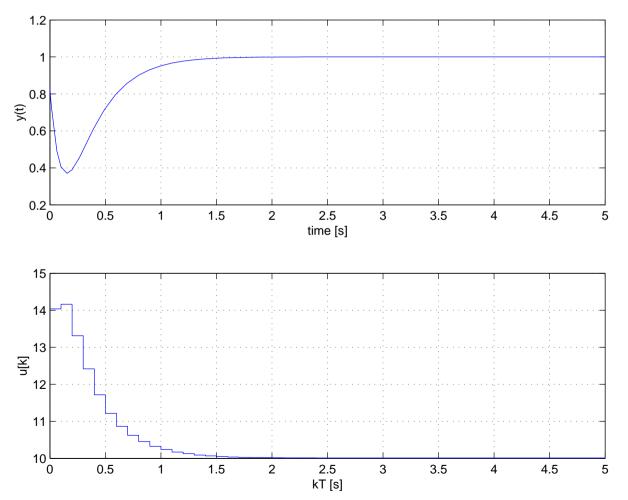
following SIMULINK diagram to run the simulations.



Notice the **sampled** signals coloured in red (check the option "Sample Time Colors" in the *Format* menu). All blocks with discrete-time signals include a sampler at their input.

Discrete-Time Control Design: Example

Example (Discrete design: simulations). The following plots show the response of the closed-loop sampled-data controlled system: continuous-time output y(t) and discrete-time control signal u[k].





Discrete-Time Design: Peculiarities

Two **special differences** in the discrete-time design procedure:

The gain N for steady state error compensation is, as in continuous-time, the inverse of the steady-state DC gain of the closed-loop transfer function. Notice though, that in discrete-time

$$N = \frac{1}{C(zI - A_d + B_d K_d)^{-1} B_d} = \frac{1}{C(I - A_d + B_d K_d)^{-1} B_d}$$

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If we implement robust tracking by adding Integral Action into the state feedback design, notice that the plant augmentation is different in discrete-time,

$$\mathbf{A}_{\alpha} = \begin{bmatrix} \mathbf{A}_{\mathbf{d}} & \mathbf{0} \\ -\mathbf{C} & \mathbf{1} \end{bmatrix}$$

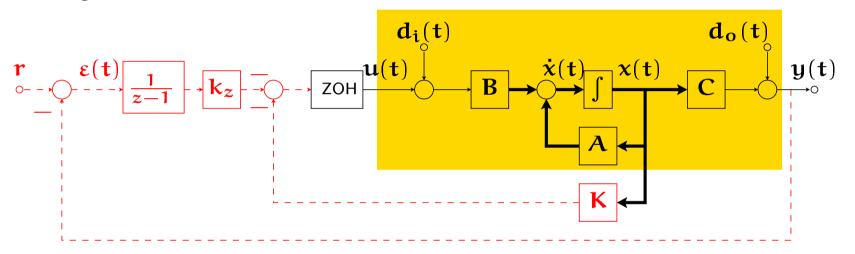
since the **discrete-integration** of the tracking error has to be implemented as

$$z[k+1] = z[k] + r - Cx[k]$$



Discrete-Time Design: Peculiarities

The implementation of the **discrete-time integral action** in the diagram should be consistent, i.e. **discrete-integration** of the tracking error.



Apart from these two differences, and the locations for the eigenvalues/poles, the discrete-time design is obtained by performing the same computations for continuous-time state feedback and observer design.

Outline

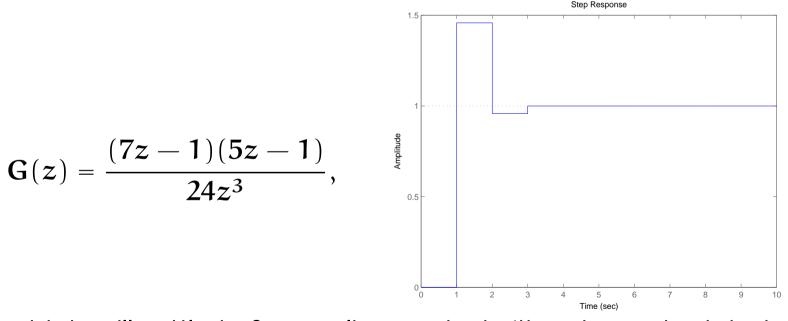
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Discrete-Time Design: Dead-Beat Control

A special discrete-time design that **has no correlate** in continuous-time is **dead-beat control**.

A **dead-beat** response is a response that settles in its final value at a **finite time**. This feature arises in a discrete-time system which has all its poles at z = 0, e.g.,



which will settle in 3 sampling periods (the dynamics is just a **delay** of 3 sample periods).



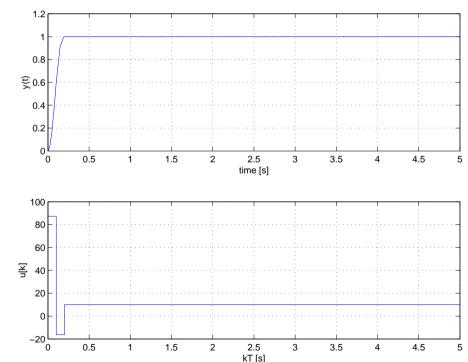
Discrete-Time Design: Dead-Beat Control

To design a **dead-beat** controller, we just have to find K_d to place all the closed-loop poles at z = 0. The discrete-time observer can also be designed **dead-beat**, with L_d to place all the observer poles at z = 0.

The plot shows the response of the DC motor of the example controlled to have **dead-beat** response (state feedback — no observer).

There is not much flexibility in dead-beat control, the only parameter to change the response is the sampling period.

Dead-beat usually requires **large control action**, which may saturate actuators.



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 - the choice of desired eigenvalues (which have to be in the unit circle for stability)
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