



Intro. Computer Control Systems: F9

State-feedback control and observers

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F8: Quiz!



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- 1) For an **observable** system
- a the effect of all $x(t)$ can be observed in $y(t)$ \uparrow
 - b we have $\det \mathcal{O} = 0$ \uparrow
 - c we have stability \downarrow



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- 2) If a state-space form of $G(s)$ is a **minimal realization**,
 - a A 's eigenvalues $<$ $G(s)$'s poles \uparrow
 - b A 's eigenvalues $=$ $G(s)$'s poles \uparrow
 - c there exist more compact state-space forms \downarrow



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- 3) For a **controllable** system with state-feedback control
 - a no information about the system is required \uparrow
 - b the poles of the closed-loop system can be designed arbitrarily \uparrow
 - c the closed-loop system is stable \downarrow

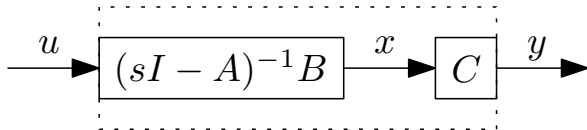


State-feedback control

State-feedback control

State-space form of linear time-invariant system

$$\begin{aligned} \dot{x} &= Ax + Bu \\ y &= Cx \end{aligned} \quad \Rightarrow \quad G(s) = C(sI - A)^{-1}B$$



State-feedback control

Controller using state feedback

$$u = -Lx + \ell_0 r$$

gives closed-loop system

$$\dot{x} = (A - BL)x + B\ell_0 r$$

$$y = Cx$$

where r is the reference signal.

State-feedback control

Controller using state feedback

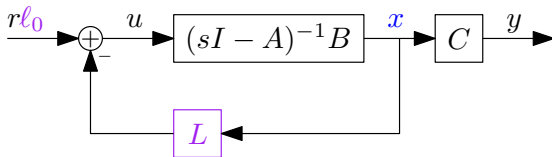
$$u = -Lx + l_0 r$$

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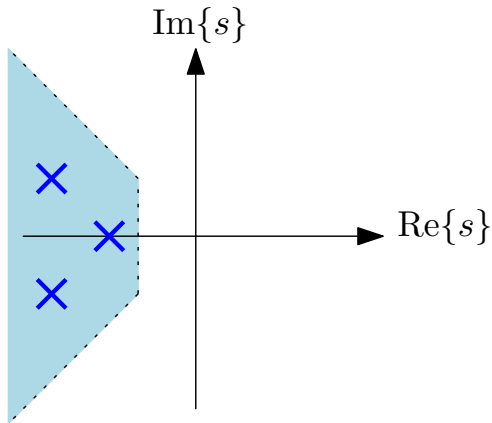


$$G_c(s) = C(sI - A + BL)^{-1} B l_0$$

Pole placement

Rules of thumb for designing L

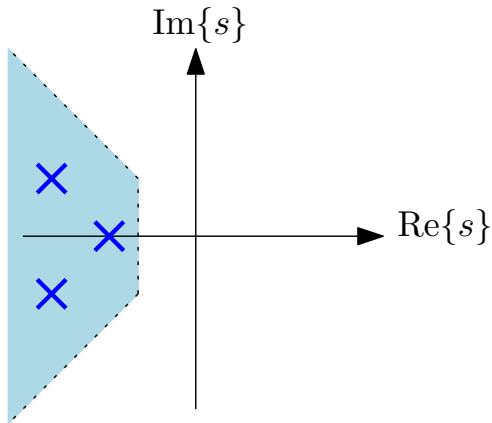
Eigenvalues/**poles** given by $\det(sI - A + BL) = 0$, which we can *design*



Pole placement

Rules of thumb for designing L

Eigenvalues/**poles** given by $\det(sI - A + BL) = 0$, which we can *design*



Distance to the origin: Quick system but also sensitive to disturbances



Estimating the states via simulation



Estimating the states

Via simulation

- ▶ Controller

$$u = -Lx + \ell_0 r$$

requires states x which are often **unknown**.

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- ▶ In practice, feedback using

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where \hat{x} is an **estimate** of x .

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where \hat{x} is an **estimate** of x .

- ▶ *Naive idea*: **Estimate** x by *simulating* the states

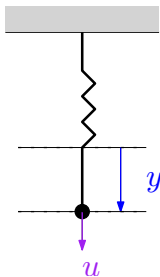
$$\dot{\hat{x}} = A\hat{x} + Bu, \quad \hat{x}(0) = \hat{x}_0$$

where \hat{x}_0 is an **initial guess**.

Build intuition from simple systems

State estimation via simulation

Ex.: Damper



State-space form:

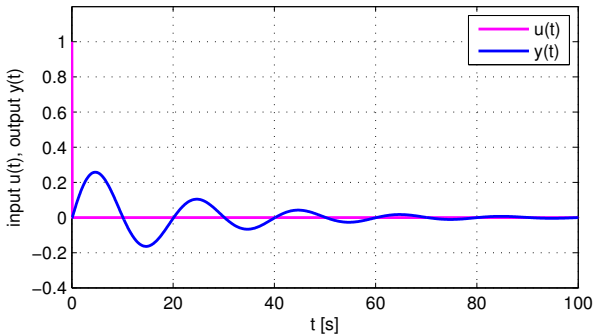
$$\dot{x}(t) = \begin{bmatrix} 0 & 1 \\ -k/m & 0 \end{bmatrix} x(t) + \begin{bmatrix} 0 \\ 1/m \end{bmatrix} u(t), \quad x(0) = x_0$$

$$y(t) = \begin{bmatrix} 1 & 0 \end{bmatrix} x(t)$$

Build intuition from simple systems

State estimation via simulation

Example using impulse u



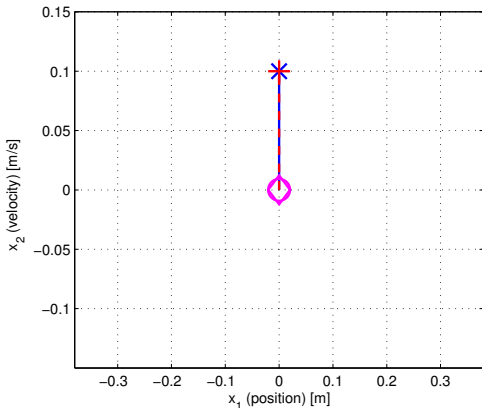
System with unknown **initial** state x_0

Build intuition from simple systems

State estimation via simulation

Naive estimate using *perfect* initial guess:

$$\dot{\hat{x}} = A\hat{x} + Bu, \quad \boxed{\hat{x}_0 = x_0}$$



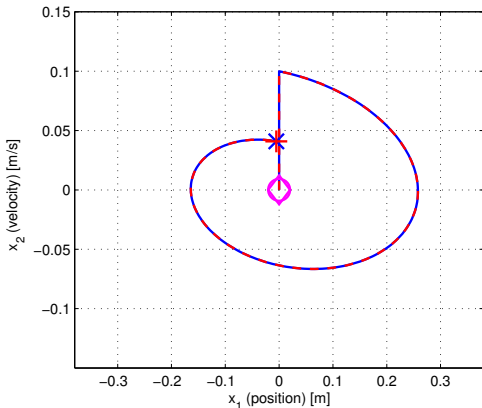
x versus \hat{x} at $t = 0^+$

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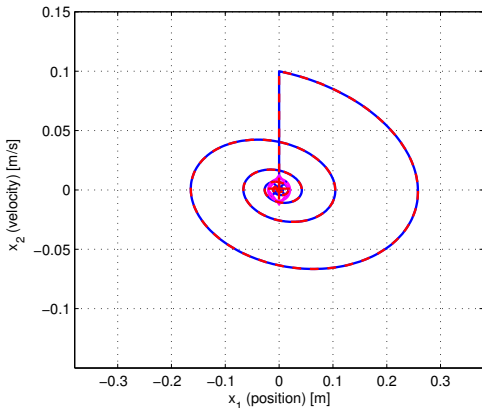
x versus \hat{x} at $t = 20$

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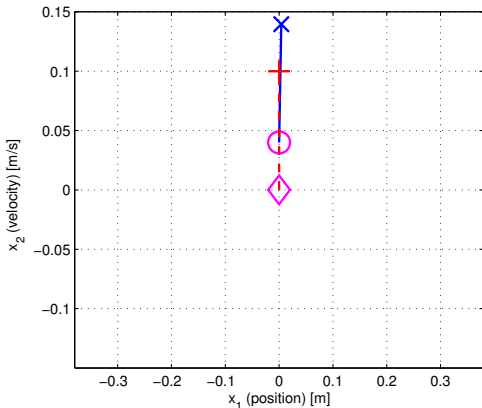
x versus \hat{x} at $t = 100$

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State estimation via simulation

Naive estimate using *wrong* initial guess:

$$\dot{\hat{x}} = A\hat{x} + Bu, \quad \hat{x}_0 \neq x_0$$



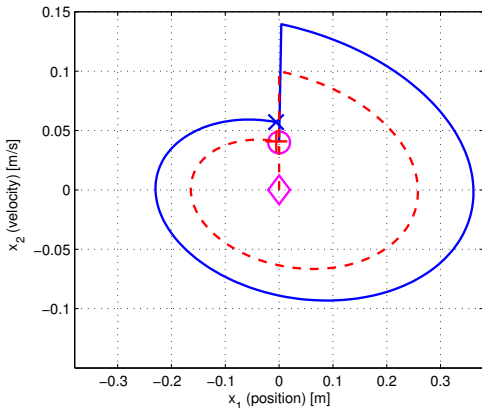
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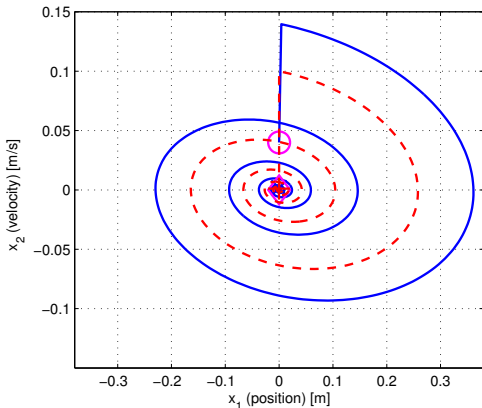
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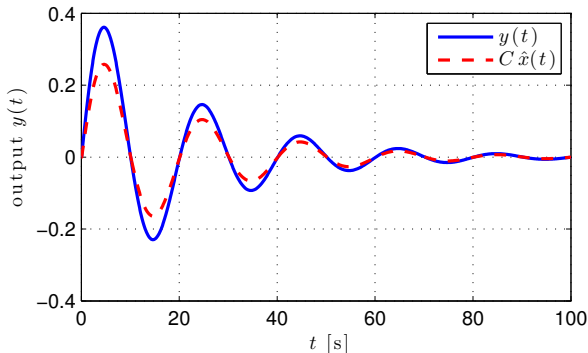
x versus \hat{x} at $t = 100$

Build intuition from simple systems

State estimation via simulation

x och \hat{x} correspond to different outputs:

$$y = Cx \quad \text{versus} \quad \hat{y} = C\hat{x}$$





Estimating the states via observer

Estimating the states

Correcting the state estimates

- ▶ *Idea*: Feedback the prediction error $y - C\hat{x}$ to correct \hat{x}
- ▶ **Observer**: an estimator with a *correction term*

$$\dot{\hat{x}} = A\hat{x} + Bu + \underbrace{K(y - C\hat{x})}_{\text{correction}}, \quad \hat{x}(0) = \hat{x}_0$$

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- ▶ Using matrix

$$K = \begin{bmatrix} k_1 \\ k_2 \\ \vdots \\ k_n \end{bmatrix}$$

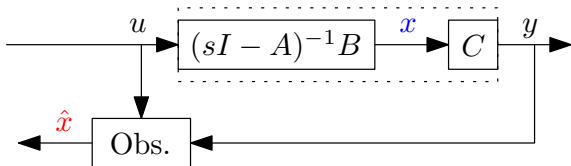
we can *design* the estimator.

Estimating the states

Correcting the state estimates

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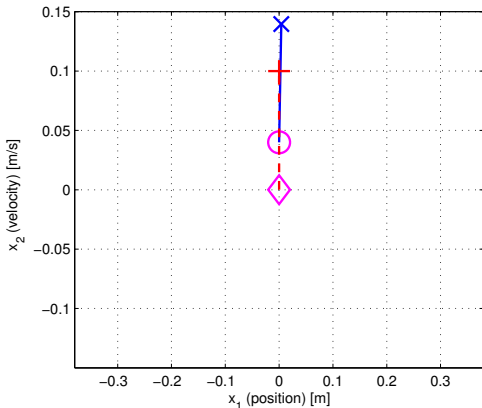


Build intuition using simple systems

State estimation using observer

Estimation using observer:

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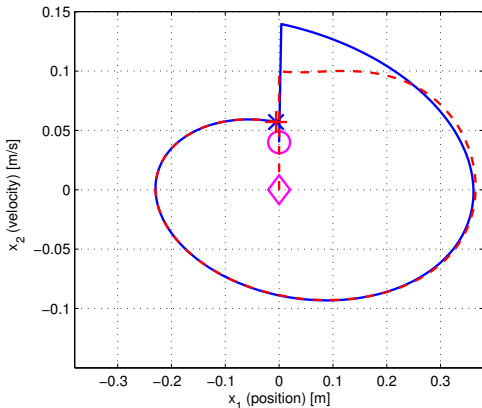
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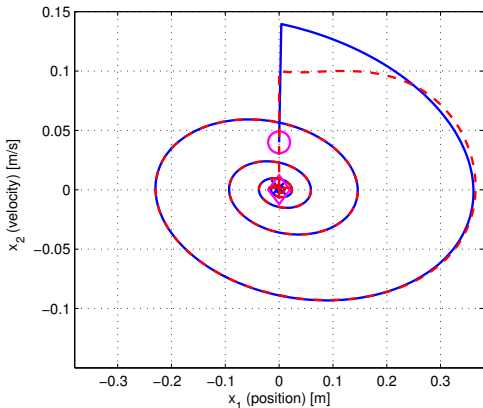
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x versus \hat{x} at $t = 100$



State estimation

Estimation error and observability

Estimation error:

$$\tilde{x} \triangleq x - \hat{x}$$

[Board: derive evolution of estimation errors]

State estimation

Estimation error and observability

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[Board: derive evolution of estimation errors]

Result

Errors of observer described as system

$$\tilde{x}(t) = e^{(A-KC)t} \tilde{x}(0)$$

and therefore $\|\tilde{x}(t)\|$ decays at a rate given by maximum

$$\operatorname{Re}\{\tilde{s}_i\}$$

where \tilde{s}_i are observer poles/eigenvalues of $(A - KC)$.



State estimation

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Result 9.2

State-space form is **observable** (cf. $\det \mathcal{O} \neq 0$) \Rightarrow matrix K can be chosen such that \tilde{x} vanish arbitrarily quick

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- ▶ K is solved by polynomial $\det(sI - A + KC) = 0$ with **desired roots** in left halfplane

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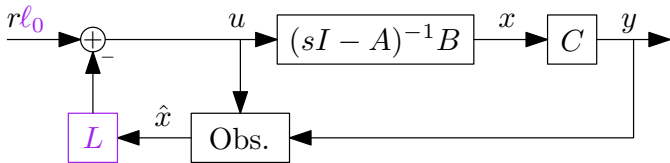
- ▶ K is solved by polynomial $\det(sI - A + KC) = 0$ with **desired roots** in left halfplane
- ▶ Quick observer \hat{x} is however **sensitive** to measurement noise!



Feedback with estimated states

Feedback using estimated states

Controller the Laplace domain

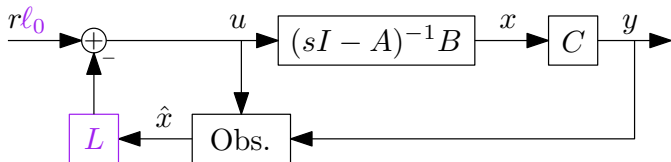


System and controller with observer:

$$\begin{cases} \dot{x} &= Ax + Bu \\ y &= Cx \end{cases} \quad \text{and} \quad \begin{cases} u &= -L\hat{x} + \ell_0 r \\ \dot{\hat{x}} &= A\hat{x} + Bu + K(y - C\hat{x}) \end{cases}$$

Feedback using estimated states

Controller the Laplace domain



Controller with observer:

$$\mathcal{L} : \begin{cases} U(s) &= -L\hat{X}(s) + l_0 R(s) \\ s\hat{X}(s) &= A\hat{X}(s) + BU(s) + K(Y(s) - C\hat{X}(s)) \end{cases}$$

[Board: solve for controller]

Feedback using estimated states

General linear feedback control

General linear feedback form, ch.9.5 G&L

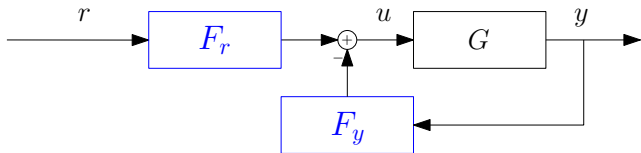
Controller with observer can be written as

$$U(s) = F_r(s)R(s) - F_y(s)Y(s),$$

where

$$F_r(s) = (1 - L(sI - A + KC + BL)^{-1}B)\ell_0$$

$$F_y(s) = L(sI - A + KC + BL)^{-1}K$$

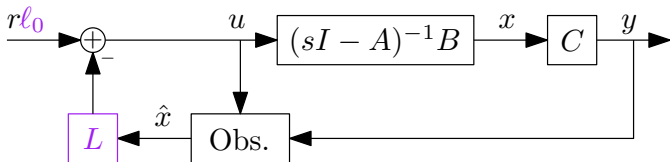




Resulting closed-loop system

Feedback using estimated states

Effect of estimation error



Study **system** and **controller with observer**:

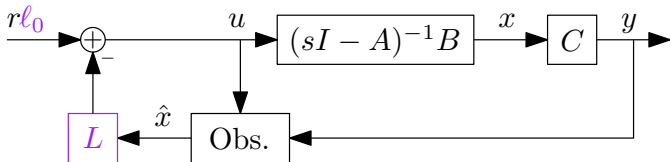
$$\begin{cases} \dot{x} &= Ax + Bu \\ y &= Cx \end{cases} \quad \text{and} \quad u = -L\hat{x} + \ell_0 r$$

by substituting $\hat{x} = x - \tilde{x}$

[Board: derive the closed-loop system with estimation error \tilde{x}]

Feedback using estimated states

Effect of estimation error



Yields closed-loop system:

$$\dot{x} = (A - BL)x + \overbrace{BL\tilde{x}}^{\text{effect of estimation error}} + Bl_0r$$

$$y = Cx$$

with additional error states

$$\dot{\tilde{x}} = (A - KC)\tilde{x} \rightarrow 0$$

[Board: write the closed-loop system in state-space form]

Feedback using estimated states

The closed-loop system with observer

The closed-loop system with estimation error can be written as

$$\begin{bmatrix} \dot{x} \\ \dot{\tilde{x}} \end{bmatrix} = \underbrace{\begin{bmatrix} A - BL & BL \\ 0 & A - KC \end{bmatrix}}_{\tilde{A}} \begin{bmatrix} x \\ \tilde{x} \end{bmatrix} + \underbrace{\begin{bmatrix} B \\ 0 \end{bmatrix}}_{\tilde{B}} \ell_0 r$$

$$y = \underbrace{\begin{bmatrix} C & 0 \end{bmatrix}}_{\tilde{C}} \begin{bmatrix} x \\ \tilde{x} \end{bmatrix}$$

with extended state vector.

This yields transfer function from r to y :

$$\Rightarrow G_c(s) = \tilde{C}(sI - \tilde{A})^{-1}\tilde{B}$$

Feedback using estimated states

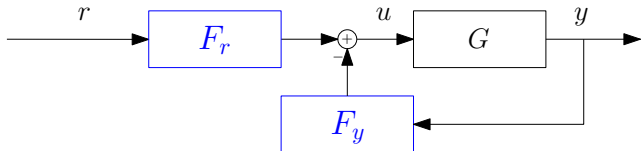
The closed-loop system with observer

Closed-loop system transfer function, ch.9.5 G&L

Insert matrices \tilde{A} , \tilde{B} and \tilde{C} yields

$$\begin{aligned}
 G_c(s) &= \tilde{C}(sI - \tilde{A})^{-1}\tilde{B} \\
 &= C(sI - A + BL)^{-1}B\ell_0
 \end{aligned}$$

with **same poles** as if states were **known** and K is gone!





Summary and recap

- ▶ Rules of thumb for pole placement
- ▶ Estimation using observer
- ▶ Feedback using estimated states
- ▶ Closed-loop system with observer