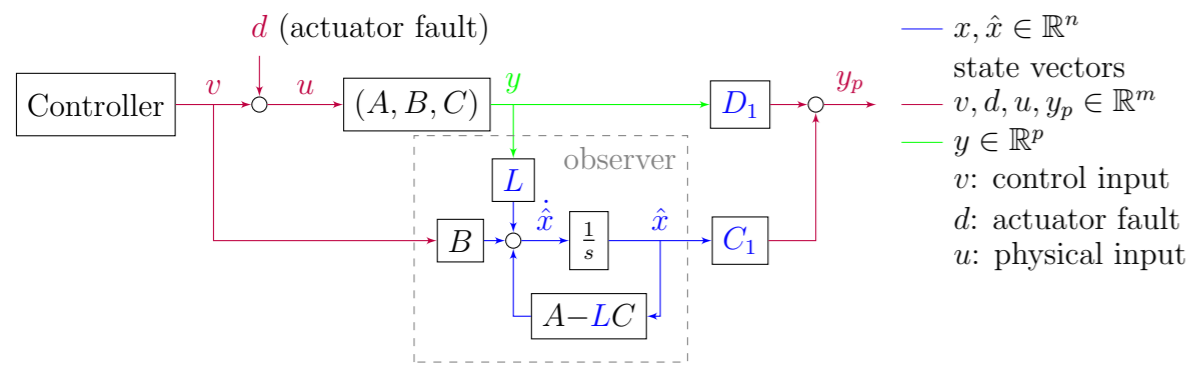
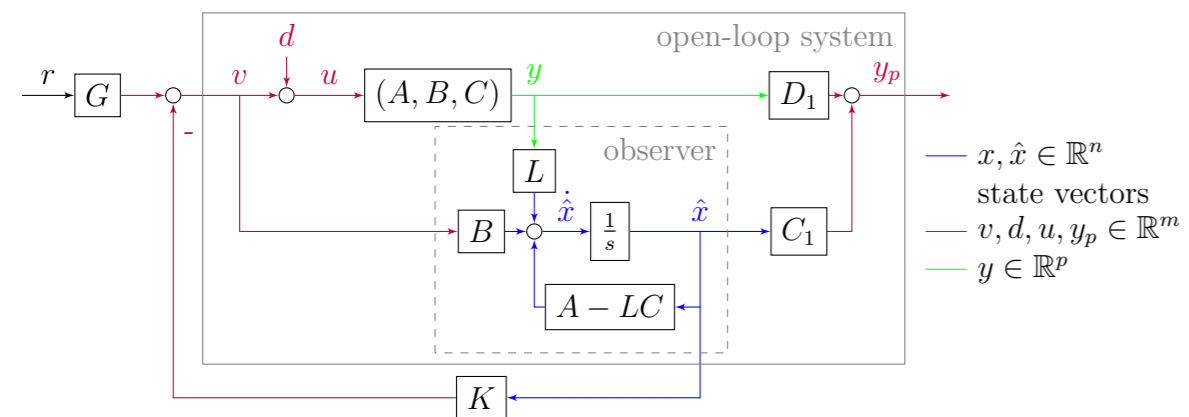


1 MIMO, $u \rightarrow y_p$ passivization, feedback equivalence with a passive system

Having an LTI MIMO system, which is somehow fed back through a controller (either tuned by output vector or by the full state vector). The actuator is assumed to be faulty. We intend to detect its fault using system inversion.]. However, the system is not invertible, since its zeros are unstable and/or its vector relative degree (v.r.d.) is more than 1, therefore, we augment the system with an additional (linear) dynamics, which is tuned by the system's output y and the designed control input u , hence its resemblance to an observer.



The goal is to choose matrices L, C_1, D_1 such that the system $v \rightarrow y_p$ be feedback equivalent to a passive system and the v.r.d. of $v \rightarrow y_p$ be 1.



1.1 Numerical examples – MIMO

Unstable MIMO system:

$$H_1(s) = \begin{pmatrix} \frac{s-1}{(s+1)(s-2)} & \frac{1}{s-3} \\ \frac{(s+2)(s-7)}{(s-5)(s-1)(s-2)} & \frac{s-6}{(s-2)(s-3)} \end{pmatrix} \quad (1)$$

$\text{tzero}(H_1) = (7.2854 \quad 0.0214 \quad 3 \quad 2 \quad 1.8361)$

Stable MIMO system:

$$H_2(s) = \begin{pmatrix} \frac{s-1}{(s+1)(s+2)} & \frac{1}{s+3} \\ \frac{s-7}{s^2+6s+5} & \frac{s-6}{s^2+5s+6} \end{pmatrix} \quad (2)$$

$\text{tzero}(H_2) = (2.5 - 7.1937i \quad 2.5 + 7.1937i)$

For both models $CB = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ is rank deficient ($r \neq \{1, 1\}$).

2 Separation principle, $X = \begin{pmatrix} x \\ \hat{x} \end{pmatrix}$ $K, L = \text{place}, G = I_m, C_1, D_1 = \text{sdpvar}$

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ \dot{\hat{x}} = LCx + (A - LC)\hat{x} + Bu \\ y_p = D_1Cx + C_1\hat{x} \end{cases} \Rightarrow \begin{cases} \hat{A}_o = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix}, \hat{B}_o = \begin{pmatrix} B \\ B \end{pmatrix} \\ \hat{C}_o = (D_1C \quad C_1) \end{cases} \quad (3)$$

$$\text{CLS: } u = -K\hat{x} + Gu \Rightarrow \begin{cases} \hat{A} = \begin{pmatrix} A & -BK \\ LC & A - LC - BK \end{pmatrix}, \hat{B} = \hat{B}_oG \\ \hat{C} = (D_1C \quad C_1) \end{cases} \quad (4)$$

$$\dot{V}(X) \leq r^T y_p + y_p^T r - y_p^T W y_p, \quad \text{where } W \succ 0, V(X) = X^T P X, P = P^T \succ 0 \quad (5)$$

$$\begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} \hat{A}^T P + P \hat{A} & P \hat{B} \\ \hat{B}^T P & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \leq \begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} -\hat{C}^T W \hat{C} & \hat{C}^T \\ \hat{C} & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \quad (6)$$

Equivalently: (it could be **only negative semi-definite**)

$$M_0 = \begin{pmatrix} \hat{A}^T P + P \hat{A} + \hat{C}^T W \hat{C} & P \hat{B} - \hat{C}^T \\ \hat{B}^T P - \hat{C} & 0 \end{pmatrix} \preceq 0, \quad \text{but } \not\prec 0 \quad (7)$$

Using Schur's complement lemma: (it could be **only negative semi-definite**)

$$M_1 = \begin{pmatrix} \hat{A}^T P + P \hat{A} & P \hat{B} - \hat{C}^T & \hat{C}^T \\ \hat{B}^T P - \hat{C} & 0 & 0 \\ \hat{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \not\prec 0 \quad (\text{LMI}) \quad (8)$$

Known variables: A, B, C, D (model matrices), furthermore, G, K, L are preliminarily designed.

Free variables: C_1, D_1 (output matrices), $P = \begin{pmatrix} Q \\ S \end{pmatrix}$

2.1 Numerical results (in case of $H_1(s)$ – unstable)

Before optimization:

$$p = (-2 \quad -1.8 \quad -1.6 \quad -1.4 \quad -1.2 \quad -1)$$

$$L^T = \text{place}(A^T, B^T, p)^T = \begin{pmatrix} -9.45 & -82.2 & 206 & 57.3 & -193 & -199 \\ -85.7 & 58.7 & -50.2 & -0.158 & 16 & 16 \end{pmatrix} \quad (9)$$

$$K = \text{place}(A, B, p) = \begin{pmatrix} 5.34 & -0.398 & 0.323 & -0.888 & 0.443 & -0.445 \\ -5.05 & 3.6 & -3.13 & 1.61 & 3.27 & -0.0894 \end{pmatrix}$$

$$G = I_2$$

Ezáltal kaptunk egy stabil observert és az observer által megfigyelt állapotot visszacsatolva egy stabil zárt hurkot.

Ezek után válasszuk meg C_1 és D_1 -et úgy, hogy a zárt rendszer passzív is legyen, tehát az open-loop rendszer feedback ekvivalens egy passzív rendszerrel és az $r = \{1, \dots, 1\}$ by construction teljesül, ezért az open-loop rendszer minimum fázisú.

Az optimalizációs eljárás után kapjuk, hogy:

$$C_1 = \begin{pmatrix} 2.81 & -0.758 & -7.06 & 2.13 & 5.98 & -12.3 \\ -1.39 & -2.21 & -0.601 & -1.98 & 6.4 & -6.2 \end{pmatrix}, \quad D_1 = \begin{pmatrix} -7.18 & -6.53 \\ -10.3 & -0.0594 \end{pmatrix} \quad (10)$$

Poles and zeros of the open-loop system:

$$\text{POLES} = (5 \quad -1 \quad 2 \quad 1 \quad 3 \quad 2), \quad \text{ZEROS} = (-0.31 + 1.14i \quad -0.31 - 1.14i \quad -0.35 \quad -0.22) \quad (11)$$

3 Separation principle (advanced, linearised), $X = \begin{pmatrix} x \\ e \end{pmatrix}$ $K = \text{place}, G = I_m, C_1, D_1 = \text{sdpvar}, P := \begin{pmatrix} Q & 0 \\ 0 & S \end{pmatrix}, N := SL$

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ \dot{\hat{x}} = LCx + (A - LC)\hat{x} + Bu \\ \dot{e} = \dot{x} - \hat{x} = (A - LC)e \\ y_p = D_1Cx + C_1\hat{x} \\ = (D_1C + C_1)x - C_1x + C_1\hat{x} \\ = (D_1C + C_1)x - C_1e \end{cases} \Rightarrow \begin{cases} \tilde{A}_o = \begin{pmatrix} A & 0 \\ 0 & A - LC \end{pmatrix}, \tilde{B}_o = \begin{pmatrix} B \\ 0 \end{pmatrix} \\ \tilde{C}_o = (D_1C + C_1 \quad -C_1) \end{cases} \quad (12)$$

$$\text{CLS: } u = -K\hat{x} + Gu \Rightarrow \begin{cases} \tilde{A} = \begin{pmatrix} A - BK & BK \\ 0 & A - LC \end{pmatrix}, \tilde{B} = \tilde{B}_oG \\ \tilde{C} = (D_1C + C_1 \quad -C_1) \end{cases} \quad (13)$$

Using Schur's complement lemma: (it could be **only negative semi-definite**)

$$M_1 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} - \tilde{C}^T & \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 & 0 \\ \tilde{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \not\prec 0 \quad (\text{will be an LMI}) \quad (14)$$

If we constrain $P = \begin{pmatrix} Q \\ S \end{pmatrix}$ to be a *block-diagonal* symmetric positive definite matrix, then

$$\tilde{A}^T P + P \tilde{A} = \begin{pmatrix} Q(A - BK) + (A - BK)^T Q & QBK \\ K^T B^T Q & SA + A^T S - SLC - C^T L^T S \end{pmatrix} \quad (15)$$

We can introduce $N := SL$ (not overdetermined), obtaining a linear matrix variable:

$$\tilde{A}^T P + P \tilde{A} = \begin{pmatrix} Q(A - BK) + (A - BK)^T Q & QBK \\ K^T B^T Q & SA + A^T S - NC - C^T N^T \end{pmatrix} \quad (16)$$

Known variables: A, B, C, D (model matrices), furthermore, K, G are preliminarily designed.

Free variables: C_1, D_1, L (model matrices of the augmented model), $P = \begin{pmatrix} Q \\ S \end{pmatrix}$

3.1 Numerical results (in case of $H_2(s)$ – stable)

We design a stabilizing feedback gain K for the original system ($u = -Kx$):

$$p = (-2 \quad -1.8 \quad -1.6 \quad -1.4 \quad -1.2 \quad -1)$$

$$K = \text{place}(A, B, p) = \begin{pmatrix} -1.5456 & -0.3636 & -0.1654 & 0.0854 & 0.0847 \\ 0.0668 & 0.0156 & -0.6457 & -0.677 & -0.6302 \end{pmatrix} \quad (17)$$

$$G = I_2$$

If we have a stable observer $\dot{\hat{x}} = CLx + (A - LC)\hat{x} + Bu$, applying $u = -K\hat{x}$ will stabilize the system (separation principle). Assuming that $G = I_2$, we chose L, C_1, D_1 such that the observer be stable, and the closed loop system with $y_p = D_1Cx + C_1\hat{x}$ performance input be strict output passive (from r to y_p). We obtain:

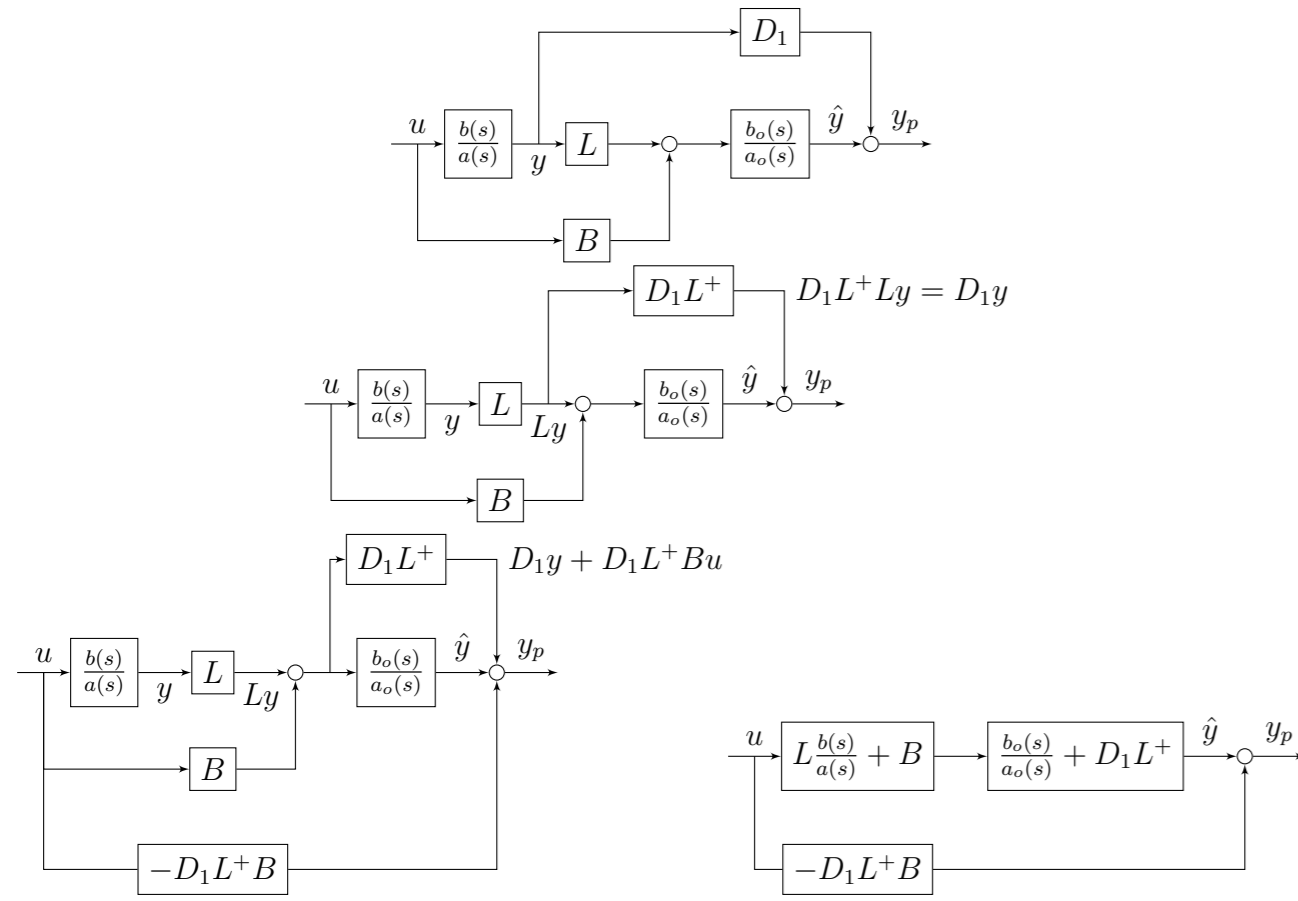
$$C_1 = \begin{pmatrix} 0.0048 & 0.0022 & 0.0015 & -0.001 & -0.0009 \\ -0.0005 & 0.0018 & 0 & 0.0044 & 0.0006 \end{pmatrix}, \quad D_1 = \begin{pmatrix} 0.0031 & 0.0001 \\ 0.0083 & -0.0005 \end{pmatrix} \quad (18)$$

$$L^T = \begin{pmatrix} -11.8203 & 11.0372 & 2.8894 & -13.6842 & 7.1631 \\ -3.5266 & -2.2015 & 7.9634 & -0.3426 & -0.4627 \end{pmatrix}$$

Poles and zeros of the open-loop system:

$$\text{POLES} = (-5 \quad -2 \quad -1 \quad -3 \quad -2), \quad \text{ZEROS} = (-3.3395 \quad -0.3282 \quad -1.2477) \quad (19)$$

4 Megvalósíthatósági tanulmány – operátortartománybeli analízis



Ahol

$$\frac{b_o(s)}{a_o(s)} = C_1(sI - A + LC)^{-1} \quad (20)$$

Szerintem ez MIMO esetben is jó ($a(s)$ és $a_o(s)$ skalár, minden egyéb mátrix):

$$\begin{aligned} y_p &= D_1 y + \hat{y} = \frac{1}{a(s)} D_1 b(s) u + \frac{1}{a_o(s)} b_o(s) \left(\frac{1}{a(s)} L b(s) + B \right) u \\ G_e(s) &= \frac{1}{a(s)} D_1 b(s) + \frac{1}{a_o(s)} b_o(s) \left(\frac{1}{a(s)} L b(s) + B \right) \\ G_e(s) &= \frac{1}{a(s) a_o(s)} D_1 b(s) a_o(s) + \frac{1}{a(s) a_o(s)} b_o(s) (L b(s) + B a(s)) \\ G_e(s) &= \frac{1}{a(s) a_o(s)} \left(D_1 b(s) a_o(s) + b_o(s) L b(s) + b_o(s) B a(s) \right) \end{aligned} \quad (21)$$

Tehát az új számláló:

$$b_e(s) = D_1 b(s) a_o(s) + b_o(s) L b(s) + b_o(s) B a(s) \quad (22)$$

A piros tagok lehetnek instabilak. Kék: megválasztható szabad constans. Mivel $a_o(s)$ skalár:

$$b_e(s) = \left(a_o(s) D_1 + b_o(s) L \right) b(s) + b_o(s) B a(s) \leftarrow \text{legyen stabil} \quad (23)$$

Numerikus példával illusztrálom, hogy ez valóban lehetséges:

$$G(s) = \frac{b(s)}{a(s)} = \frac{s-5}{s^2-3s+2}, \quad A = \begin{pmatrix} 3 & -2 \\ 1 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} 2 \\ 0 \end{pmatrix}, \quad C = (0.5 \quad -2.5)$$

$$\text{observer: } L^{\text{place}} := \begin{pmatrix} -13 \\ -5 \end{pmatrix}, \quad C_1 := (1 \quad -1) \Rightarrow G_o(s) = \frac{b_o(s)}{a_o(s)} = \frac{1}{s^2+3s+2} \begin{pmatrix} s+9 & -s-25 \end{pmatrix}$$

$$\begin{aligned} b_e(s) &= (D_1 + 2)s^3 + (4 - 2D_1)s^2 + (-13D_1 - 2)s - (10D_1 + 4) \Big|_{D_1=-1} \\ &= s^3 + 6s^2 + 11s + 6 = (s+3)(s+2)(s+1) \end{aligned}$$

$$\text{eredő átviteli függvény: } G_e(s) = \frac{s+3}{s^2-3s+2} \quad (24)$$

Másik példa (ami a relatív degreet is megjavítja):

$$G(s) = \frac{s-5}{(s-3)(s^2-3s+2)}, \quad L^T \text{place} := (-193 \quad -212 \quad -52), \quad C_1 := (1 \quad 1 \quad 1), \quad D_1 := 0.1$$

$$G_e(s) = \frac{s^2 + 4.1s + 3.5}{s^3 - 6s^2 + 11s - 6} \quad (25)$$

Tudom, hogy ennek nincs sok értelme ==>
de gondoltam megíscsak kiprobalom...

