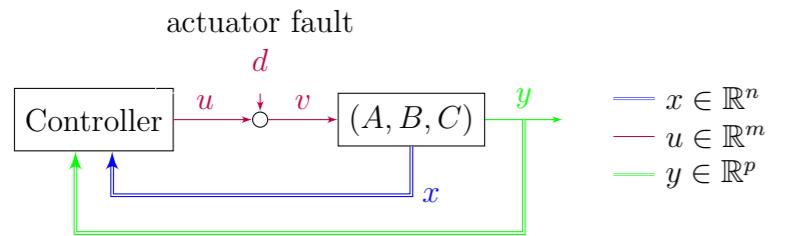
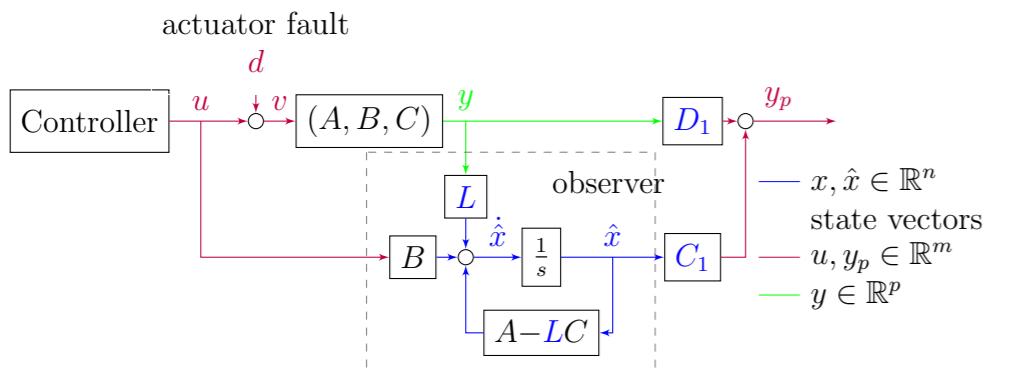


1 MIMO, $u \rightarrow y_p$ passivization, feedback equivalence with a passive system

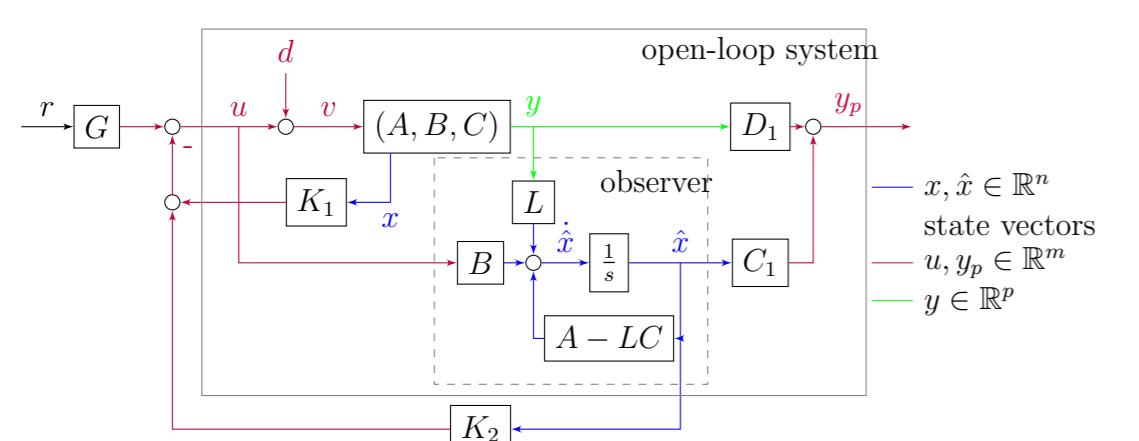


Having an LTI MIMO system, which is somehow fed back through a controller (either tuned by output vector or by the full state vector). The actuator is assumed to be faulty. We intend to detect its fault using system inversion. However, the system is not invertible, since its zeros are unstable and/or 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.



Proposition 1. (Vector relative degree) LTI MIMO
If matrix B is full column rank, $\exists C_1$ such that the system $u \rightarrow y_p$ will have r.d. $\{1, \dots, 1\}$.

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



1.1 Numerical example – MIMO

Having an LTI system with matrices:

$$A = \begin{pmatrix} 6.1 & 1.4 & -0.33 & -1.8 & 0.88 & -0.88 \\ -3.2 & 0.5 & 0.27 & 0.14 & -0.068 & 0.068 \\ 1.4 & -1.5 & 0.58 & 0.14 & -0.07 & 0.07 \\ -0.76 & -1.2 & -0.13 & 0.85 & 1.1 & -1.1 \\ 0.38 & 0.61 & 0.067 & 2.7 & 3.6 & -1.6 \\ -0.38 & -0.61 & -0.067 & 0.59 & 1.7 & 0.3 \end{pmatrix}, \quad B = \begin{pmatrix} 2 & 0 \\ 0 & 0 \\ 0 & 0 \\ -1.6 & 1 \\ 0.82 & 2 \\ -0.82 & 0 \end{pmatrix} \quad (1)$$

$$C = \begin{pmatrix} -0.15 & -0.24 & -0.026 & -0.2 & 0.6 & -0.6 \\ 0.39 & 0.17 & -1 & 0.58 & 0.21 & -1.2 \end{pmatrix}$$

Its transfer function and its zeros are:

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

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

This system does not have a relative degree $\{1, 1\}$, since

$$\dot{y} = CAx + CBu, \quad CB = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \quad (3)$$

2 $X = \begin{pmatrix} x \\ \hat{x} \end{pmatrix}$, $L = \text{place}$, partially linearised by $N := Q\tilde{C}^T$

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ \dot{\hat{x}} = LCx + (A - LC)\hat{x} + Bu \\ y_p = D_1Cx + C_1\hat{x} \end{cases} \Rightarrow \begin{cases} \tilde{A}_o = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix}, \quad \tilde{B}_o = \begin{pmatrix} B \\ B \end{pmatrix} \\ \tilde{C}_o = (D_1C \ C_1) \end{cases} \quad (4)$$

$$\text{CLS: } u = -KX + Gu \Rightarrow \begin{cases} \tilde{A} = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix} - BK \\ \tilde{B} = \tilde{B}_o G \\ \tilde{C} = (D_1C \ C_1) \end{cases} \quad (5)$$

$$\dot{V}(X) \leq r^T y_p + y_p^T r - y_p^T W y_p, \quad \text{where } W \succ 0, \quad V(X) = X^T P X, \quad P = P^T \succ 0 \quad (6)$$

$$\begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} \\ \tilde{B}^T P & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \leq \begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} -\tilde{C}^T W \tilde{C} & \tilde{C}^T \\ \tilde{C} & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \quad (7)$$

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

$$M_0 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} + \tilde{C}^T W \tilde{C} & P \tilde{B} - \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (8)$$

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

$$M_1 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} - \tilde{C}^T & \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 & 0 \\ \tilde{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (\text{BMI}) \quad (9)$$

This condition will remain bilinear since we co-design L, K but can efficiently be solved though.

$$\widehat{M}_1 := \begin{pmatrix} Q & & \\ & I & \\ & & I \end{pmatrix} M_1 \begin{pmatrix} Q & & \\ & I & \\ & & I \end{pmatrix} = \begin{pmatrix} Q\tilde{A}^T + \tilde{A}Q & \tilde{B} - Q\tilde{C}^T & Q\tilde{C}^T \\ \tilde{B}^T - \tilde{C}Q & 0 & 0 \\ \tilde{C}Q & 0 & -W^{-1} \end{pmatrix} \preceq 0 \quad (10)$$

The free decision variables of the bilinear problem are: $Q, K, G, N = Q\tilde{C}^T$. The bilinear problem is:

$$\widehat{M}_2 = \begin{pmatrix} Q\tilde{A}^T + \tilde{A}Q & \tilde{B} - N & N \\ \tilde{B}^T - N^T & 0 & 0 \\ N^T & 0 & -W^{-1} \end{pmatrix} \preceq 0 \quad (\text{BMI}) \quad (11)$$

$$Q \succeq 0, \quad G \neq 0 \quad (12)$$

After this:

$$\tilde{C} = (Q^{-1}N)^T = (\tilde{C}_1 \ \tilde{C}_2) \stackrel{!}{=} (D_1C \ C_1)$$

$$\text{thus: } \begin{cases} \tilde{C}_1 = D_1C \text{ (overdetermined)} \Rightarrow D_1 = \tilde{C}_1 C^T (CC^T)^{-1} \text{ if } \text{Im}(C^T) \stackrel{!!}{\subseteq} \text{Im}(\tilde{C}_1^T) \\ \tilde{C}_2 = C_1 \end{cases} \quad (13)$$

2.1 Numerical results

$$L^T = \begin{pmatrix} -9.45 & -82.2 & 206 & 57.3 & -193 & -199 \\ -85.7 & 58.7 & -50.2 & -0.158 & 16 & 16 \end{pmatrix}$$

$$C_1 = \begin{pmatrix} 0.0792 & -0.0938 & 0.143 & 0.46 & 0.108 & 0.751 \\ 1.42 & 1.34 & -0.379 & 2.09 & -1.09 & 3.34 \end{pmatrix}, D_1 = \begin{pmatrix} 0.702 & -0.293 \\ 5.25 & -2.07 \end{pmatrix}$$

$$K_1 = \begin{pmatrix} 205 & -4.62 & 55.1 & -96.9 & 14.5 & -89.5 \\ 52.8 & -4.71 & -32.7 & 9.28 & 151 & 173 \end{pmatrix}$$

$$K_2 = \begin{pmatrix} -38.2 & -97.5 & 25.2 & 70.4 & -7.57 & 65.7 \\ 7.32 & -1.27 & 38 & 8.68 & 81.9 & 89.1 \end{pmatrix}, G = \begin{pmatrix} 13.2 & 1.48 \\ 1.48 & 16 \end{pmatrix}$$

Problem. The closed loop system is indeed minimum phase, however, the open-loop system is NOT minimum phase, since $D_1 C = \tilde{C}_1$ (overdetermined).

If we can change C (i.e. we know the hole state vector), it may work.

$$3 \quad X = \begin{pmatrix} x \\ \hat{x} \end{pmatrix}, L = \text{sdpvar}, \text{ highly bilinear}$$

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ \dot{\hat{x}} = LCx + (A - LC)\hat{x} + Bu \\ y_p = D_1Cx + C_1\hat{x} \end{cases} \Rightarrow \begin{cases} \tilde{A}_o = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix}, \tilde{B}_o = \begin{pmatrix} B \\ B \end{pmatrix} \\ \tilde{C}_o = (D_1C \ C_1) \end{cases} \quad (14)$$

$$\text{CLS: } u = -KX + Gu \Rightarrow \begin{cases} \tilde{A} = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix} - BK, \tilde{B} = \tilde{B}_oG \\ \tilde{C} = (D_1C \ C_1) \end{cases} \quad (15)$$

$$\dot{V}(X) \leq r^T y_p + y_p^T r - y_p^T W y_p, \quad \text{where } W \succ 0, V(X) = X^T P X, P = P^T \succ 0 \quad (16)$$

$$\begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} \\ \tilde{B}^T P & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \leq \begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} -\tilde{C}^T W \tilde{C} & \tilde{C}^T \\ \tilde{C} & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \quad (17)$$

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

$$M_0 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} + \tilde{C}^T W \tilde{C} & P \tilde{B} - \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (18)$$

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

$$M_1 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} - \tilde{C}^T & \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 & 0 \\ \tilde{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (\text{BMI}) \quad (19)$$

$$Q \succeq 0, \quad G \neq 0 \quad (20)$$

The free decision variables of the bilinear problem are: P, C_1, D_1, L, K, G .

3.1 Numerical results

$$L^T = \begin{pmatrix} -9.45 & -82.2 & 206 & 57.3 & -193 & -199 \\ -85.7 & 58.7 & -50.2 & -0.158 & 16 & 16 \end{pmatrix}$$

$$C_1 = \begin{pmatrix} 0.0792 & -0.0938 & 0.143 & 0.46 & 0.108 & 0.751 \\ 1.42 & 1.34 & -0.379 & 2.09 & -1.09 & 3.34 \end{pmatrix}, D_1 = \begin{pmatrix} 0.702 & -0.293 \\ 5.25 & -2.07 \end{pmatrix}$$

$$K_1 = \begin{pmatrix} 205 & -4.62 & 55.1 & -96.9 & 14.5 & -89.5 \\ 52.8 & -4.71 & -32.7 & 9.28 & 151 & 173 \end{pmatrix}$$

$$K_2 = \begin{pmatrix} -38.2 & -97.5 & 25.2 & 70.4 & -7.57 & 65.7 \\ 7.32 & -1.27 & 38 & 8.68 & 81.9 & 89.1 \end{pmatrix}, G = \begin{pmatrix} 13.2 & 1.48 \\ 1.48 & 16 \end{pmatrix}$$

Problem. Solution not found.

$$4 \quad X = \begin{pmatrix} x \\ \hat{x} \end{pmatrix}, L = \text{place/sdpvar}, G = I_m, D_1, C_1 = \text{sdpvar}$$

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ \dot{\hat{x}} = LCx + (A - LC)\hat{x} + Bu \\ y_p = D_1Cx + C_1\hat{x} \end{cases} \Rightarrow \begin{cases} \tilde{A}_o = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix}, \tilde{B}_o = \begin{pmatrix} B \\ B \end{pmatrix} \\ \tilde{C}_o = (D_1C \ C_1) \end{cases} \quad (21)$$

$$\text{CLS: } u = -KX + Gu \Rightarrow \begin{cases} \tilde{A} = \begin{pmatrix} A & 0 \\ LC & A - LC \end{pmatrix} - BK, \tilde{B} = \tilde{B}_oG \\ \tilde{C} = (D_1C \ C_1) \end{cases} \quad (22)$$

$$\dot{V}(X) \leq r^T y_p + y_p^T r - y_p^T W y_p, \quad \text{where } W \succ 0, V(X) = X^T P X, P = P^T \succ 0 \quad (23)$$

$$\begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} \\ \tilde{B}^T P & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \leq \begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} -\tilde{C}^T W \tilde{C} & \tilde{C}^T \\ \tilde{C} & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \quad (24)$$

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

$$M_0 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} + \tilde{C}^T W \tilde{C} & P \tilde{B} - \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (25)$$

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

$$M_1 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} - \tilde{C}^T & \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 & 0 \\ \tilde{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (\text{BMI}) \quad (26)$$

$$P \succeq 0 \quad (27)$$

The free decision variables of the bilinear problem are: P, C_1, D_1, L, K ,

Problem. Solution not found.

5 Separability principle

D_1 should be full rank!

$$\text{OLS: } \begin{cases} \dot{x} = Ax + Bu \\ y_p = (D_1C + C_1)x \end{cases} \quad \text{CLS: } \begin{cases} \tilde{A} = A - BK, \tilde{B} = BG \\ \tilde{C} = D_1C + C_1 \end{cases} \quad (28)$$

$$\dot{V}(x) \leq r^T y_p + y_p^T r - y_p^T W y_p, \quad \text{where } W \succ 0, V(x) = x^T P x, P = P^T \succ 0 \quad (29)$$

$$\begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} \\ \tilde{B}^T P & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \leq \begin{pmatrix} X \\ r \end{pmatrix}^T \begin{pmatrix} -\tilde{C}^T W \tilde{C} & \tilde{C}^T \\ \tilde{C} & 0 \end{pmatrix} \begin{pmatrix} X \\ r \end{pmatrix} \quad (30)$$

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

$$M_0 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} + \tilde{C}^T W \tilde{C} & P \tilde{B} - \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (31)$$

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

$$M_1 = \begin{pmatrix} \tilde{A}^T P + P \tilde{A} & P \tilde{B} - \tilde{C}^T & \tilde{C}^T \\ \tilde{B}^T P - \tilde{C} & 0 & 0 \\ \tilde{C} & 0 & -W^{-1} \end{pmatrix} \preceq 0, \quad \text{but } \cancel{0} \quad (\text{BMI}) \quad (32)$$

This condition will remain bilinear since we co-design L, K but can efficiently be solved though.

$$\widehat{M}_1 := \begin{pmatrix} Q & I \\ I & I \end{pmatrix} M_1 \begin{pmatrix} Q & I \\ I & I \end{pmatrix} = \begin{pmatrix} Q \tilde{A}^T + \tilde{A}Q & \tilde{B} - Q \tilde{C}^T & Q \tilde{C}^T \\ \tilde{B}^T - \tilde{C}Q & 0 & 0 \\ \tilde{C}Q & 0 & -W^{-1} \end{pmatrix} \preceq 0 \quad (33)$$

The free decision variables of the bilinear problem are: $Q, K, G, N = Q \tilde{C}^T$. The bilinear problem is:

$$\widehat{M}_2 = \begin{pmatrix} Q \tilde{A}^T + \tilde{A}Q & \tilde{B} - N & N \\ \tilde{B}^T - N^T & 0 & 0 \\ N^T & 0 & -W^{-1} \end{pmatrix} \preceq 0 \quad (\text{BMI}) \quad (34)$$

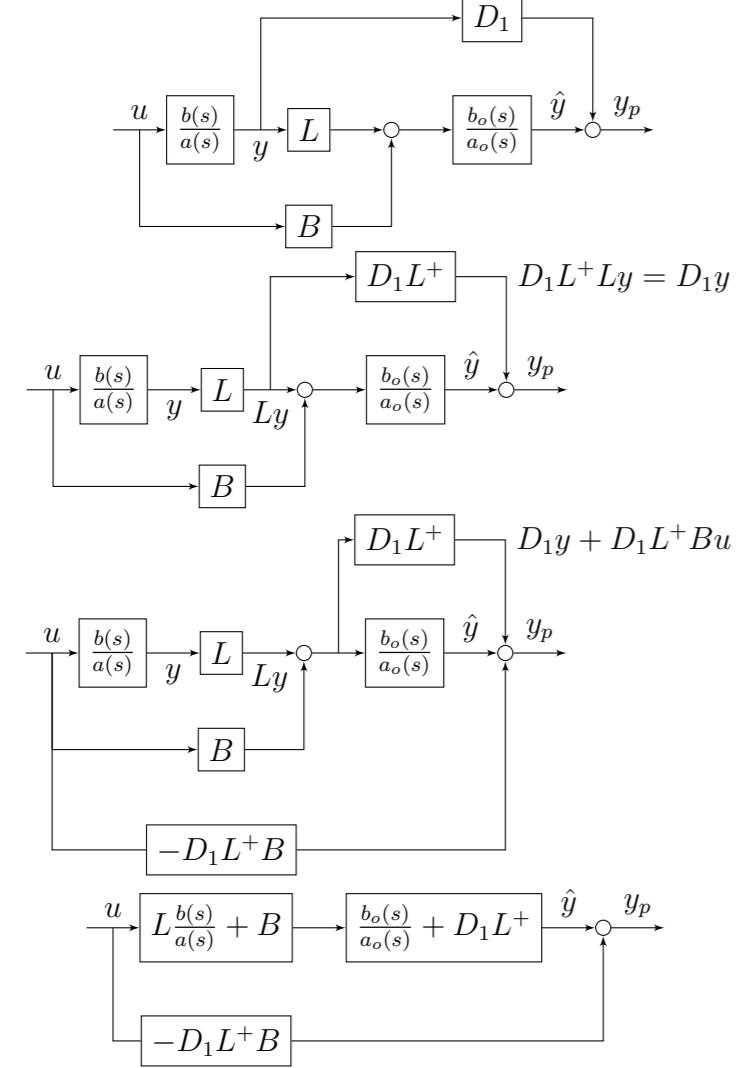
$$Q \succeq 0, \quad G \neq 0 \quad (35)$$

After this:

$$\tilde{C} = (Q^{-1}N)^T = (\tilde{C}_1 \ \tilde{C}_2) \stackrel{!}{=} (D_1C \ C_1)$$

$$\text{thus: } \begin{cases} \tilde{C}_1 = D_1C \text{ (overdetermined)} \Rightarrow D_1 = \tilde{C}_1 C^T (CC^T)^{-1} \text{ if } \text{Im}(C^T) \subseteq \text{Im}(\tilde{C}_1^T) \\ \tilde{C}_2 = C_1 \end{cases} \quad (36)$$

6 Megvalósíthatósági tanulmány – operátortartománybeli analízis



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

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

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

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

(41)

Másik példa (ami a relatív degeet is megjavítja):

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

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

(42)

Ahol

$$\frac{b_o(s)}{a_o(s)} = C_1(sI - A + LC)^{-1} \quad (37)$$

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

$$y_p = D_1y + \hat{y} = \frac{1}{a(s)}D_1b(s)u + \frac{1}{a_o(s)}b_o(s)\left(\frac{1}{a(s)}Lb(s) + B\right)u$$

$$G_e(s) = \frac{1}{a(s)}D_1b(s) + \frac{1}{a_o(s)}b_o(s)\left(\frac{1}{a(s)}Lb(s) + B\right)$$

$$G_e(s) = \frac{1}{a(s)a_o(s)}D_1b(s)a_o(s) + \frac{1}{a(s)a_o(s)}b_o(s)(Lb(s) + Ba(s))$$

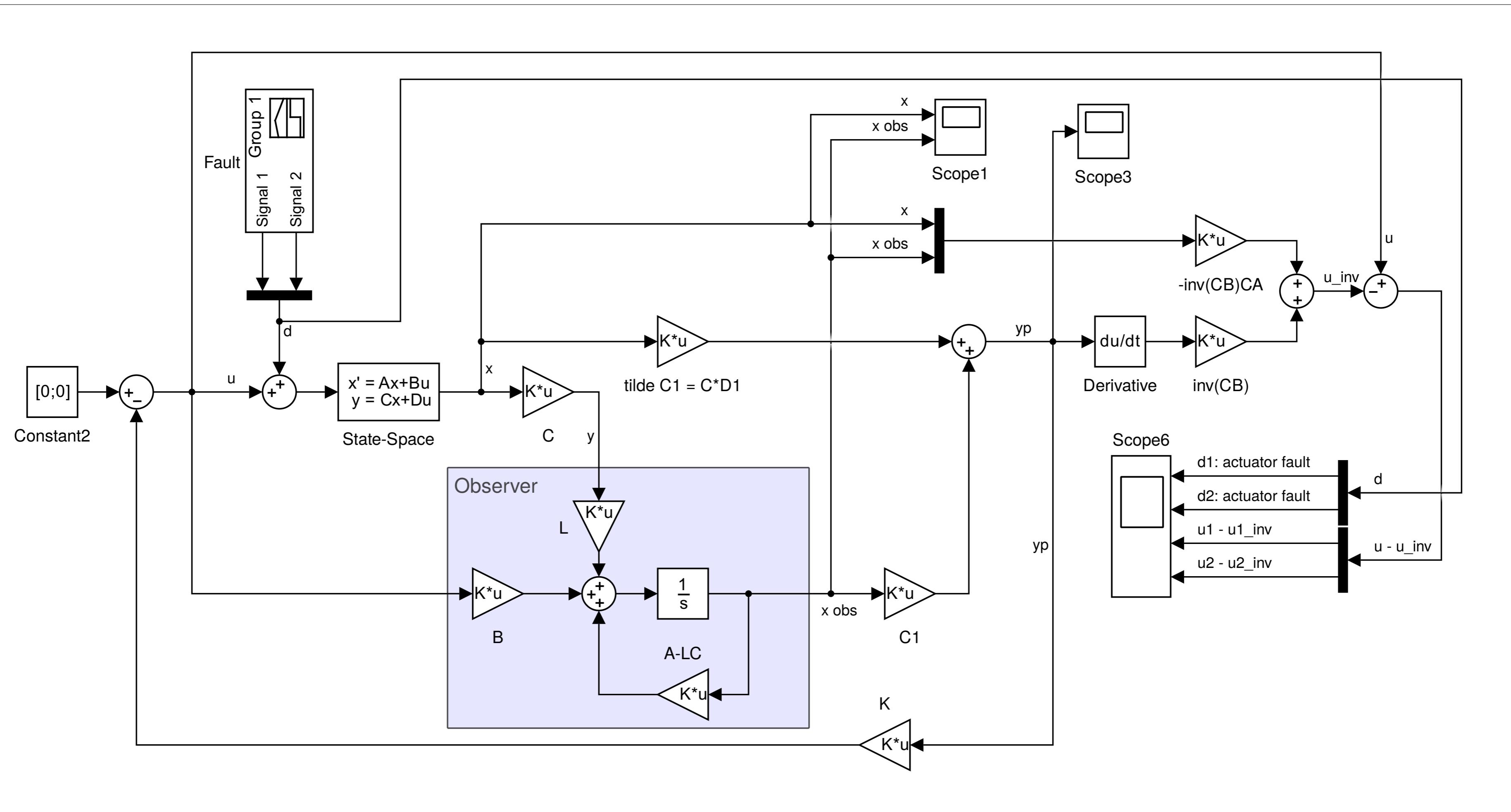
$$G_e(s) = \frac{1}{a(s)a_o(s)}\left(D_1b(s)a_o(s) + b_o(s)Lb(s) + b_o(s)Ba(s)\right)$$

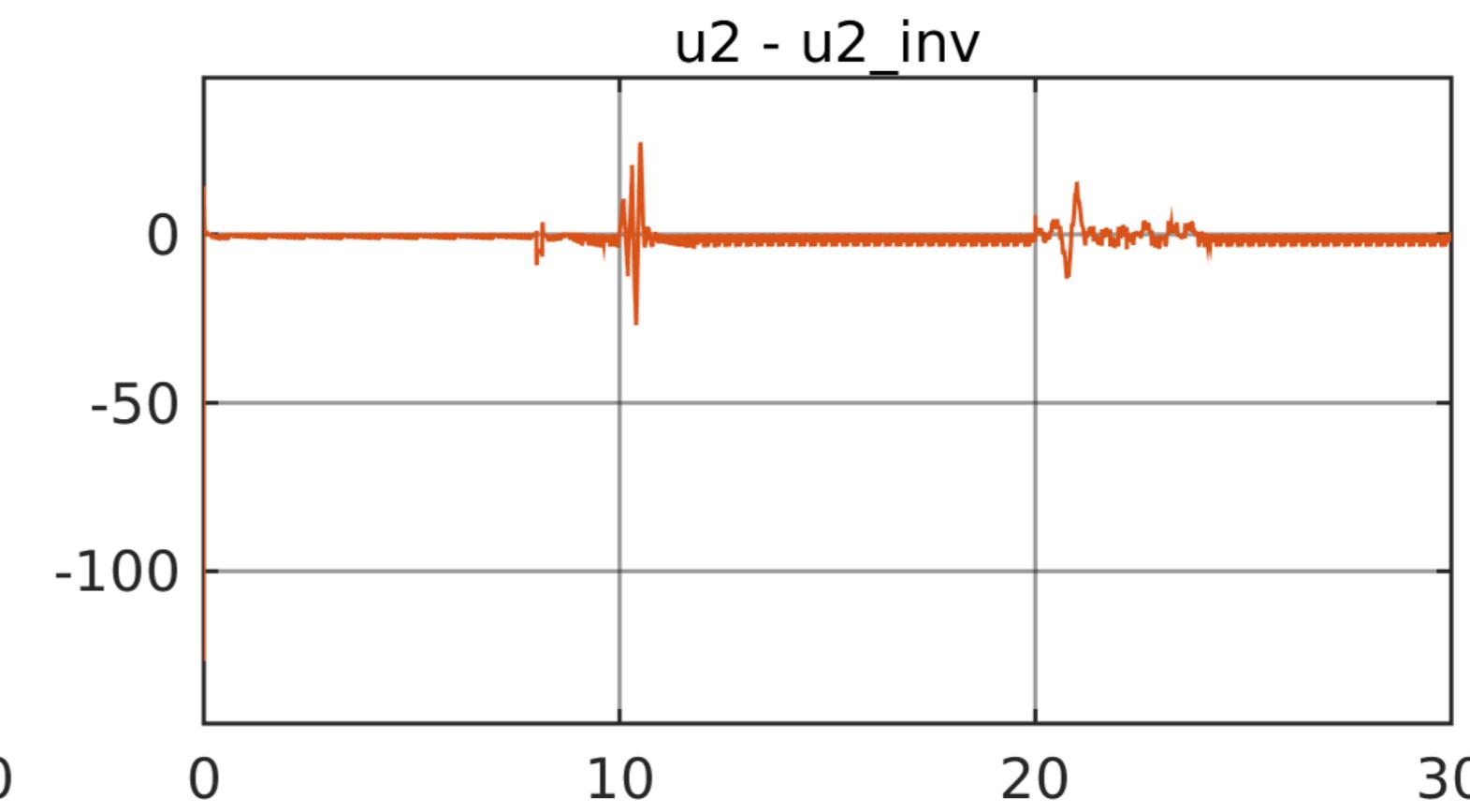
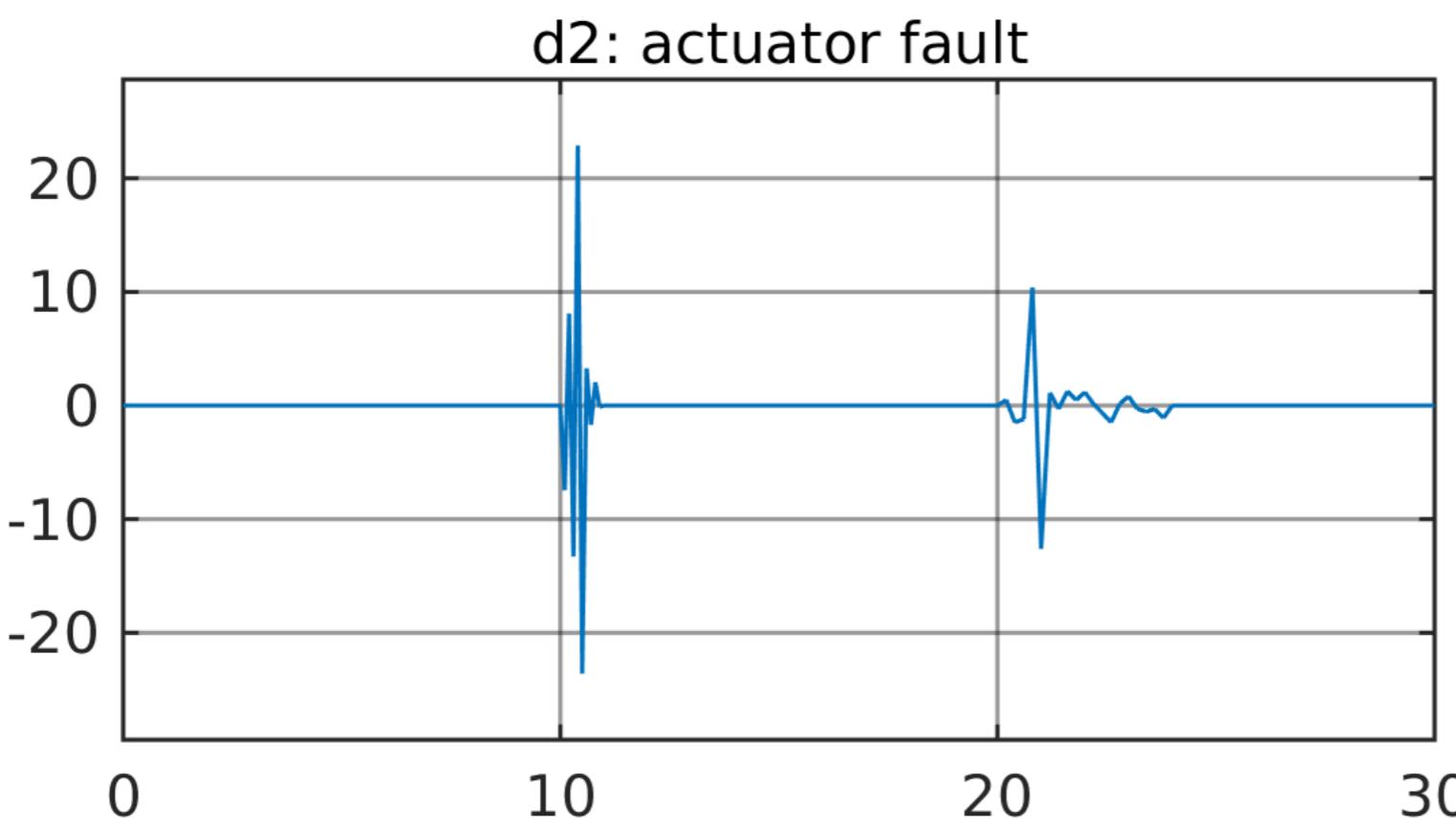
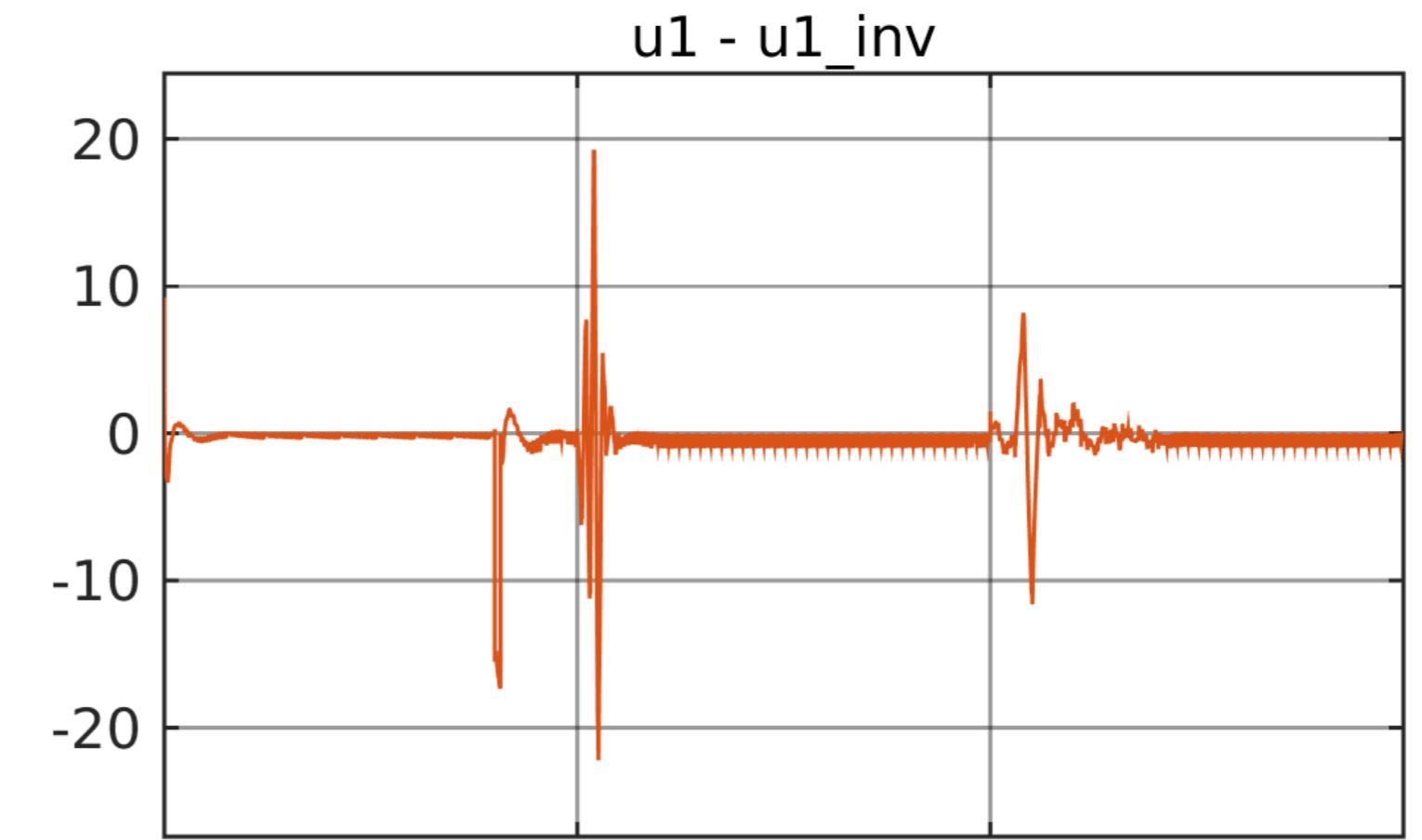
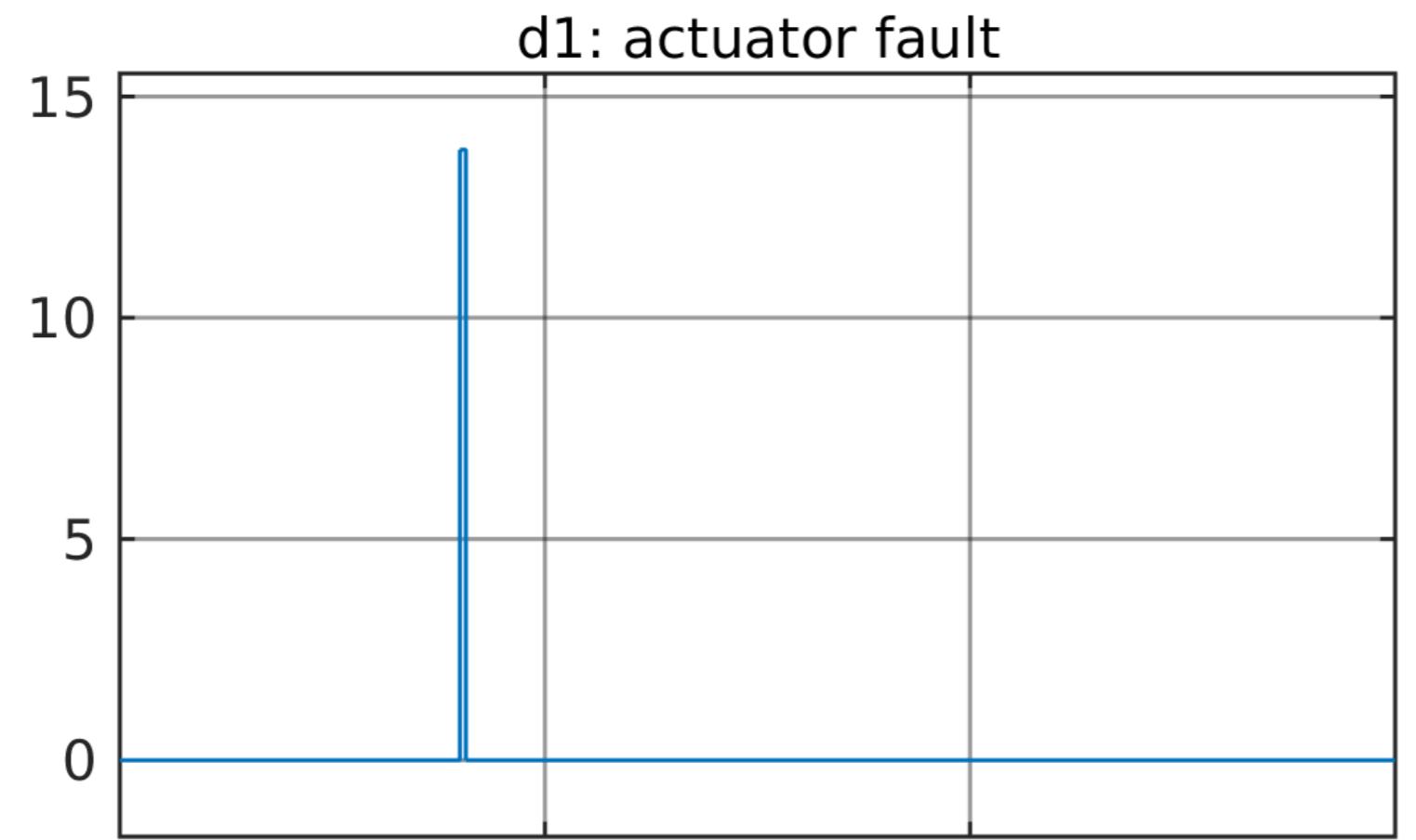
Tehát az új számláló:

$$b_e(s) = D_1b(s)a_o(s) + b_o(s)Lb(s) + b_o(s)Ba(s) \quad (39)$$

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

$$b_e(s) = \left(a_o(s)D_1 + b_o(s)L\right)b(s) + b_o(s)Ba(s) \leftarrow \text{legyen stabil} \quad (40)$$





Offset=0