# Computer controlled systems

#### Lecture 4

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# 1 State space transformation

As we shall already know, the state space model is not unique. For the given example, define a new SSM using a state space transformation.

$$A = \begin{pmatrix} -6 & -4 \\ 2 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 4 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 0 & 1 \end{pmatrix}$$

Let the linear transformation of the state vector be the following:

$$\bar{x}_1 = x_1 + x_2 \\ \bar{x}_2 = 3x_1 - 2x_2$$

In matrix form:

$$\begin{pmatrix} \bar{x}_1 \\ \bar{x}_2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 3 & -2 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

 $\bar{x} = Tx, \ x = T^{-1}\bar{x} \rightarrow \text{state state space equation can be written for the new state vector } \bar{x} \text{ as well}$ 

$$\dot{x} = Ax + Bu \quad \to \quad T^{-1}\dot{\bar{x}} = AT^{-1}\bar{x} + Bu$$

$$\dot{\bar{x}} = TAT^{-1}\bar{x} + TBu \quad \to \quad \bar{A} = TAT^{-1} \quad \bar{B} = TB$$

$$u = Cx = CT^{-1}\bar{x} \quad \to \quad \bar{C} = CT^{-1}$$

Returning to the example:

$$T = \begin{pmatrix} 1 & 1 \\ 3 & -2 \end{pmatrix} \quad T^{-1} = -\frac{1}{5} \begin{pmatrix} -2 & -1 \\ -3 & 1 \end{pmatrix}$$
 
$$\bar{A} = TAT^{-1} = \begin{pmatrix} -4 & 0 \\ -16 & -2 \end{pmatrix} \quad \bar{B} = TB = \begin{pmatrix} 4 \\ 12 \end{pmatrix} \quad \bar{C} = CT^{-1} = \begin{pmatrix} \frac{3}{5} & -\frac{1}{5} \end{pmatrix}$$

If the original and the transformed SSM are (A, B, C) and  $(\bar{A}, \bar{B}, \bar{C})$ , respectively, determine the transformation matrix T, which connects them.

$$A = \begin{pmatrix} 3 & 2 \\ -4 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0 \end{pmatrix} \tag{1}$$

$$\bar{A} = \begin{pmatrix} 1.8 & 1.6 \\ -4.4 & 2.2 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} 3 \\ 1 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} 0.4 & -0.2 \end{pmatrix}$$
 (2)

Solution.  $\bar{B} = TB$ ,  $\bar{A}\bar{B} = TAB \rightarrow T \cdot [B|AB] = [\bar{B}|\bar{A}\bar{B}] \rightarrow T \cdot [B|AB] = [\bar{B}|\bar{A}\bar{B}]$  where  $C_2 = [B|AB]$  and  $\bar{C}_2 = [\bar{B}|\bar{A}\bar{B}]$  are the controllability matrices of (1) and (2), respectively. Remark. B and AB are  $(2 \times 1)$  matrices.

$$C_2 = \begin{pmatrix} 1 & 5 \\ 1 & -3 \end{pmatrix}, \quad C_2^{-1} = \frac{1}{-8} \begin{pmatrix} -3 & -5 \\ -1 & 1 \end{pmatrix}, \quad \bar{C}_2 = \begin{pmatrix} 3 & 7 \\ 1 & -11 \end{pmatrix}$$

$$T = \frac{1}{-8} \cdot \begin{pmatrix} 3 & 7 \\ 1 & -11 \end{pmatrix} \cdot \begin{pmatrix} -3 & -5 \\ -1 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ -1 & 2 \end{pmatrix} \qquad T^{-1} = \frac{-1}{5} \begin{pmatrix} 2 & -1 \\ 1 & 2 \end{pmatrix}$$

Just as in the previous example, determine the transformation matrix T.

$$A = \begin{pmatrix} 2 & 0 \\ -1 & 2 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 0 \end{pmatrix} \tag{3}$$

$$\bar{A} = \begin{pmatrix} 3 & -1 \\ 1 & 1 \end{pmatrix} \quad \bar{B} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \quad \bar{C} = \begin{pmatrix} \frac{-1}{2} & \frac{1}{2} \end{pmatrix}$$
 (4)

Solution. 
$$T = \begin{pmatrix} -1 & 2 \\ 1 & 2 \end{pmatrix}$$
  $T^{-1} = \frac{-1}{4} \cdot \begin{pmatrix} 2 & -2 \\ -1 & -1 \end{pmatrix}$ 

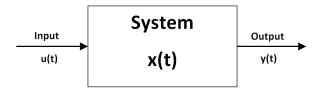
Remark. In case of SISO model, this method can be applied for an even higher dimensional state-space model, but then the controllability matrix will involve further rows. If the state vector is n-dimensional  $(A \in \mathbb{R}^{n \times n})$ , than  $\mathcal{C}_n = [B|AB|A^2B|\dots|A^{n-1}B]$ . To conclude, if the SSM is controllable:

$$T = \bar{\mathcal{C}}_n \cdot \mathcal{C}_n^{-1} \tag{5}$$

 $Megjegyz\acute{e}s$ : SISO modell esetén a fenti módszer több állapotváltozó esetén is alkalmazható, de ekkor több oszlopra van szükség. Ha  $A \in \mathbb{R}^{n \times n}$ , akkor a  $[B|AB|A^2B|\dots|A^{n-1}B]$  alakú mátrixokkal lehet számolni.

# 2 Controllability, observability

In general Given the following CT-LTI system: The question arouse: In the full knowledge of y(t) and



u(t) can we say something about the unknown state vector x(t)? In the other words is x(t) observable? The second question would be the following: is there an input function u(t), with which we can lead the system from the initial state  $x_0$  to state  $x_1$  in a finite time. If we can do so (for every possible initial and final states), we say that the system is **controllable**.

### 2.1 Observability

#### Theorem 1.

Sufficient and necessary condition for observability

A state space model described by matrices (A, B, C) is observable if and only if (iff) its observability matrix  $\mathcal{O}_n$  is full-rank:

$$\mathcal{O}_n = \begin{pmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{pmatrix}, \quad \operatorname{rank}(\mathcal{O}_n) = n$$

Remark. In SISO case  $\mathcal{O}_n$  is a square matrix, which is full-rank iff its determinant is nonzero.

**Example 1.** Is the system (A, B, C) observable?

$$A = \begin{pmatrix} 2 & 3 \\ 2 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad C = \begin{pmatrix} 0 & 1 \end{pmatrix}$$

The observability matrix is the following

$$CA = \begin{pmatrix} 2 & 1 \end{pmatrix} \rightarrow \mathcal{O}_2 = \begin{pmatrix} C \\ CA \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 2 & 1 \end{pmatrix}$$
,  $\det(\mathcal{O}_2) = -2 \neq 0 \Rightarrow \mathcal{O}_2$  is full-rank

Hence, x(t) is observable, namely, using y(t) and its time derivetive  $\dot{y}(t)$ , we can compute the actual value of x(t)

$$\begin{cases} y(t) = Cx(t) \\ \dot{y}(t) = CAx(t) + CBu(t) \end{cases} \Rightarrow x(t) = \mathcal{O}_2^{-1} \begin{pmatrix} y(t) \\ \dot{y}(t) - CBu(t) \end{pmatrix}$$
 (6)

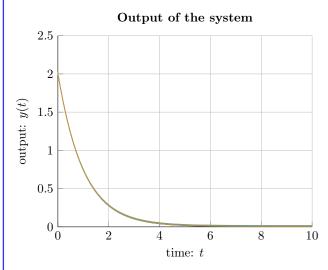
#### **Example 2.** Unobservable subspace (mathematical background presented in B.1)

Given the state space model:

$$A = \begin{pmatrix} 1 & 2 \\ -2 & -3 \end{pmatrix}, \quad B : \text{arbitrary}, \quad C = \begin{pmatrix} 1 & 1 \\ 1 & \end{pmatrix}, \quad \mathcal{O}_n = \begin{pmatrix} 1 & 1 \\ -1 & -1 \end{pmatrix}$$
 (7)

A basis for the kernel of  $\mathcal{O}_n$  is  $v_1 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$ . This means that

- $\rightarrow$  if there is a zero input and  $x(0) = \lambda v_1 \in \mathcal{O}_2$ , than  $x(t) \in \text{Ker}(\mathcal{O}_2)$  (Proposition 9) and y(t) = 0 for every t > 0.
- $\rightarrow$  for a given input u(t) and with an initial condition  $x(0) = x_0 + \lambda v_1 \in x_0 + \text{Ker}(\mathcal{O}_2)$  (where  $\lambda \in \mathbb{R}$  is arbitrary) the system will produce the same out y(t).



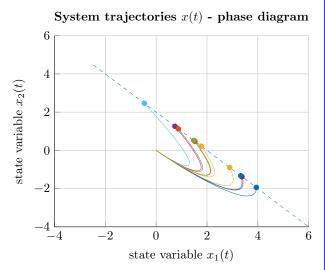


Figure 1. Simulation of system (7) from different initial conditions  $x(0) \in x_0 + \text{Ker}(\mathcal{O}_2)$  (denoted by dots) with zero input. As one can observe, the state trajectories are different, however this difference does not appear in the output of the system. In this example  $u \equiv 0$  and  $x_0 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . The blue dashed line in the right figure illustrates the actual unobservability subspace of the system corresponding to  $x_0$ .

# 2.2 Controllability

Given a strictly proper state space model (A, B, C) with  $x(t_0)$  initial and  $x(t_1) \neq x(t_0)$  final condition. The question arises, is there any input function u(t), which leads the system from  $x(t_0)$  to  $x(t_1)$  in a finite time.

Theorem 2. Controllability

A state space model described by matrices (A, B, C) is controllable iff its controllability matrix  $C_n$  is full-rank:

$$C_n = (B \quad AB \quad \cdots \quad A^{n-1}B) \quad , \quad \operatorname{rank}(C_n) = n$$

Remark. In SISO case  $C_n$  is a square matrix, which is full-rank iff its determinant is nonzero.

#### Example 3.

$$A = \begin{pmatrix} 2 & 3 \\ 2 & 1 \end{pmatrix}, \quad B = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad C = \begin{pmatrix} 0 & 1 \end{pmatrix}, \qquad \mathcal{C}_2 = \begin{pmatrix} B & AB \end{pmatrix} = \begin{pmatrix} 1 & 5 \\ 1 & 3 \end{pmatrix}$$

This system is controllable, since the determinant of  $C_2$  is nonzero. In this case the controllability subspace is the whole  $\mathbb{R}^2$  itself. If we start the system from zero initial condition, we can lead the system (with an appropriate input) to any other states of the controllability subspace in a finite time.

### Example 4. Controllable subspace (mathematical background presented in B.2)

Given the following state space system and its rank-deficient controllability matrix:

$$A = \begin{pmatrix} -\frac{1}{2} & \frac{2}{-6} & \frac{20}{3} \\ -\frac{1}{2} & -1 & -1 \end{pmatrix}, \quad B = \begin{pmatrix} 0 \\ 8 \\ 0 \end{pmatrix}, \quad \text{eigenvalues of } A \colon \begin{pmatrix} -2 \\ -2 \\ -4 \end{pmatrix}, \quad \mathcal{C}_3 = \begin{pmatrix} 0 & 16 & -96 \\ 8 & -48 & 224 \\ 0 & -8 & 48 \end{pmatrix}$$
 (8)

The basis vectors of  $\operatorname{Im}(\mathcal{C}_3)$  are:  $v_1 = \begin{pmatrix} 0.3832 \\ -0.9036 \\ -0.1916 \end{pmatrix}$ ,  $v_2 = \begin{pmatrix} 0.8082 \\ 0.4285 \\ -0.4041 \end{pmatrix}$ . They span a 2-dimensional subspace in  $\mathbb{R}^3$ , illustrated by the green plane in the Figure 2. If we start the system from an initial condition which is an element of this subspace  $x(0) \in \operatorname{Im}(\mathcal{C}_3)$ , the system trajectory will never leave this subspace. If the initial condition is outside of  $\operatorname{Im}(\mathcal{C}_3)$  and A is stable, the system trajectory will tend to this subspace.

#### System trajectories x(t) - phase diagram

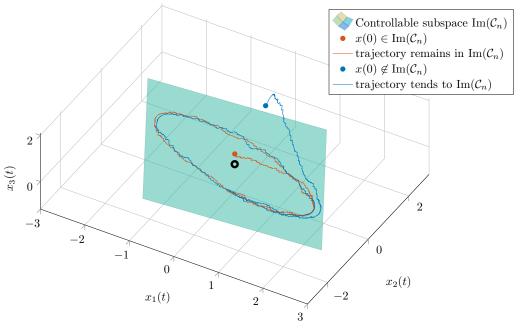


Figure 2. Simulation of system (7) from different initial conditions

#### Example 5.

Compute the controllable subspace of  $\dot{x} = Ax + Bu$ , where

$$A = \begin{pmatrix} 1 & 2 & -2 \\ -0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}, B = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}.$$
 (9)

To check your solutions, we give:

$$A^{2} = \begin{pmatrix} -1 & 4 & -4 \\ 0 & 1 & 0 \\ 2 & 2 & -1 \end{pmatrix}, \ \mathcal{O}_{3} = \begin{pmatrix} -1 & -1 & 1 \\ 0 & 0 & 0 \\ 0 & -1 & -2 \end{pmatrix}.$$
 (10)

# 2.3 Controllability and observability in case of a diagonal SSM

$$A = \begin{pmatrix} a_1 & 0 \\ 0 & a_2 \end{pmatrix} \quad B = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \quad AB = \begin{pmatrix} a_1b_1 \\ a_2b_2 \end{pmatrix} \quad C = \begin{pmatrix} c_1 & c_2 \end{pmatrix} \quad CA = \begin{pmatrix} c_1a_1 & c_2a_2 \end{pmatrix}$$
$$C_2 = \begin{pmatrix} b_1 & a_1b_1 \\ b_2 & a_2b_2 \end{pmatrix} \qquad C_2 = \begin{pmatrix} c_1 & c_2 \\ c_1a_1 & c_2a_2 \end{pmatrix}$$

SISO rendszer diagonális A mátrix esetén

irányítható  $\iff$  a főátlóbeli elemek páronként különbözőek, és  $\forall i \ b_i \neq 0$  megfigyelhető  $\iff$  a főátlóbeli elemek páronként különbözőek, és  $\forall j \ c_j \neq 0$ 

**Theorem 3.** The rank of  $\mathcal{O}_n$  and  $\mathcal{C}_n$  is invariant to the state space transformations.

Proof.

$$\bar{A} = TAT^{-1} \quad \bar{B} = TB \quad \bar{C} = CT^{-1}$$

$$\bar{C}_n = \begin{pmatrix} TB & TAT^{-1}TB \end{pmatrix} = T\begin{pmatrix} B & AB \end{pmatrix} = TC_n$$

$$\bar{O}_n = \begin{pmatrix} CT^{-1} \\ CT^{-1}TAT^{-1} \end{pmatrix} = \begin{pmatrix} C \\ CA \end{pmatrix} T^{-1} = \mathcal{O}_n T^{-1}$$

#### 2.4 Markov parameters

$$CA^iB$$

Markov parameters are invariant to the state space transformations.

$$\bar{CB} = CT^{-1}TB = CB$$
 
$$\bar{C}\bar{A}\bar{B} = CT^{-1}TAT^{-1}TB = CAB$$

# 3 Joint controllability and observability

- Egy  $H(s) = \frac{b(s)}{a(s)}$  (SISO) átviteli függvény n-edrendű realizációjának nevezzük az (A, B, C, D) állapottér-modellt, ha  $H(s) = C(sI A)^{-1}B + D$ , ahol  $A \in \mathbb{R}^{n \times n}$ ,  $B \in \mathbb{R}^{n \times 1}$ ,  $C \in \mathbb{R}^{1 \times n}$ ,  $D \in \mathbb{R}$  (nem egyértelmű!)
- $\bullet$  Egy H(s) átviteli függvény n-edrendű realizációját minimálisnak nevezzük, ha nem létezik nála kisebb rendű realizáció.

- Egy n-dimenziós (A, B, C, D) állapottér-modellt együttesen irányíthatónak és megfigyelhetnek nevezünk, ha teljesülnek rá az irányíthatóság és a megfigyelhetőség feltételei (azaz  $\mathcal{O}_n$  és  $\mathcal{C}_n$  teljes rangú).
- Egy ÁTM minimális  $\iff$  egyszerre irányítható és megfigyelhető.

**Example 6.** Is the state space representation minimal?

$$A = \begin{pmatrix} 2 & 3 \\ 2 & 1 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad C = \begin{pmatrix} 0 & 1 \end{pmatrix}$$

Transfer function:  $H(s) = \frac{s}{s^2 - 3s - 4}$ . This SSM is minimal, since H(s) is irreducible and the degree of the denominator is equal to the order of the state space realization (n = 2).

**Example 7.** Is the state space representation minimal?

$$A = \begin{pmatrix} -2 & -1 \\ -1 & -2 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 1 \end{pmatrix}$$

$$H(s) = C(sI - A)^{-1}B = \frac{s+1}{s^2 + 4s + 3} = \frac{s+1}{(s+1)(s+3)}$$

This SSM is not minimal, meaning the one of two properties is broken: the SSM is controllable but its is no observable.

Example 8. Is the state space representation minimal?

$$A = \begin{pmatrix} -6 & -4 \\ 2 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 4 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 0 & 1 \end{pmatrix}$$

Controllability matrix:

$$C_2 = \begin{pmatrix} B & AB \end{pmatrix} = \begin{pmatrix} 4 & -24 \\ 0 & 8 \end{pmatrix}$$

The determinant of matrix  $C_2$  is nonzero, therefore, it is controllable.

Observability matrix:

$$\mathcal{O}_2 = \begin{pmatrix} C \\ CA \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 2 & 0 \end{pmatrix}$$

The determinant of matrix  $\mathcal{O}_2$  is nonzero, therefore, it is observable. Consequently, the SSM is minimal.

**Example 9.** (MIMO case) Is the state space representation minimal?

$$A = \begin{pmatrix} 1 & 4 \\ -2 & 2 \end{pmatrix} \quad B = \begin{pmatrix} 1 & 4 & 1 \\ 2 & 3 & 0 \end{pmatrix} \quad C = \begin{pmatrix} 1 & 2 \\ 0 & 7 \end{pmatrix}$$

$$AB = \begin{pmatrix} 9 & 16 & 1 \\ 2 & -2 & -2 \end{pmatrix} \quad CA = \begin{pmatrix} -3 & 8 \\ -14 & 14 \end{pmatrix}$$

$$\mathcal{O}_2 = \begin{pmatrix} C \\ CA \end{pmatrix} = \begin{pmatrix} 1 & 2 \\ 0 & 7 \\ -3 & 8 \\ -14 & 14 \end{pmatrix} \quad \mathcal{C}_2 = \begin{pmatrix} B & AB \end{pmatrix} = \begin{pmatrix} 1 & 4 & 1 & 9 & 16 & 1 \\ 2 & 3 & 0 & 2 & -2 & -2 \end{pmatrix}$$

Matrix  $\mathcal{O}_2$  is full-column-rank, and  $\mathcal{C}_2$  is full row rank, meaning that the system is jointly controllable and observable and (A, B, C) is minimal.

**Example 10.** Is the SSM minimal? If not give a minimal representation.

$$A = \begin{pmatrix} -3 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 6 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 2 \\ 6 \end{pmatrix} \quad C = \begin{pmatrix} 3 & 0 & 4 \end{pmatrix}$$

REFERENCES REFERENCES

$$H(s) = \sum_{i=1}^{n} \frac{c_i b_i}{s - \lambda_i} = \frac{3 \cdot 1}{s + 3} + \frac{0 \cdot 2}{s - 4} + \frac{6 \cdot 4}{s - 6} = \frac{3(s - 6) + 24(s + 3)}{(s + 3)(s - 6)}$$

$$H(s) = \frac{27s + 54}{s^2 - 3s - 18}$$

The SSM is not minimal, because the transfer function can be reduced.

$$A^{2} = \begin{pmatrix} 9 & 0 & 0 \\ 0 & 16 & 0 \\ 0 & 0 & 36 \end{pmatrix} \qquad \mathcal{O}_{n} = \begin{pmatrix} 3 & 0 & 4 \\ -9 & 0 & 24 \\ 27 & 0 & 144 \end{pmatrix} \qquad \mathcal{C}_{n} = \begin{pmatrix} 1 & -3 & 9 \\ 2 & 8 & 32 \\ 6 & 36 & 216 \end{pmatrix}$$

A minimal SSM can be given by skipping the single degenerated state variable:

$$A = \begin{pmatrix} -3 & 0 \\ 0 & 6 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 6 \end{pmatrix} \quad C = \begin{pmatrix} 3 & 4 \end{pmatrix}$$

A minimal realization can also be given using the controller form:

$$A = \begin{pmatrix} 3 & 18 \\ 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 27 & 54 \end{pmatrix}$$

**Example 11.** It is given a SSM in the controller form. Is the SSM jointly controllable and observable?

$$A = \begin{pmatrix} 0 & 7 & -6 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \quad B = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad C = \begin{pmatrix} 0 & 3 & 9 \end{pmatrix}$$

Transfer function:

$$H(s) = \frac{3s+9}{s^3 - 7s + 6}$$

The realization is most be controllable, since it is given in controller form:

$$A^{2} = \begin{pmatrix} 7 & -6 & 0 \\ 0 & 7 & -6 \\ 1 & 0 & 0 \end{pmatrix} \qquad \mathcal{O}_{n} = \begin{pmatrix} 0 & 3 & 9 \\ 3 & 9 & 0 \\ 9 & 21 & -18 \end{pmatrix} \qquad \mathcal{C}_{n} = \begin{pmatrix} 1 & 0 & 7 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \qquad \underset{\text{rank}(\mathcal{O}_{n}) = 2}{\operatorname{rank}(\mathcal{O}_{n}) = 2}$$

However the SSM is not observable, because it is not minimal: H(s) is reducible by s+3. Using the controller form (on the irreducible form of H(s)), we can obtain a jointly controllable and observable realization Tehát nem együttesen megfigyelhető és irányítható a rendszer. The a unobservable subspace

$$\operatorname{Ker}(\mathcal{O}_n) = \left\{ \alpha \begin{pmatrix} 9 \\ -3 \end{pmatrix} \mid \alpha \in \mathbb{R} \right\}$$

Felhasználás: Állapotmegfigyelők tervezése

Bizonyos mennyiségeket (pl. szögsebesség) nem tudunk mérni, csak becsülni. Ld.: 3. ábra

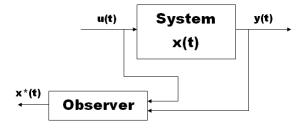


Figure 3. State observer design

### References

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[2] Lantos Béla. Irányítási rendszerek elmélete és tervezése I. Akadémiai Kiadó Budapest, 2001.

[3] A. D. Lewis. A Mathematical Approach to Classical Control. 2003.

# A Supplementary material in linear algebra (not needed for the exam)

#### Theorem 4.

The fundamental theorem of linear algebra

Let  $\mathcal{A}: \mathbb{R}^n \to \mathbb{R}^m$ ,  $\mathcal{A}(x) = Ax$ , where  $A \in \mathbb{R}^{m \times n}$ . Then the followings are true

$$\operatorname{Im}(A) = \operatorname{Ker}(A^{\top})^{\perp} \subset \mathbb{R}^{m}$$
(11a)

$$\operatorname{Im}(A^{\top}) = \operatorname{Ker}(A)^{\perp} \subset \mathbb{R}^{n}$$
(11b)

Furthermore

$$\operatorname{Im}(A) \otimes \operatorname{Ker}(A^{\top}) = \mathbb{R}^m \tag{12a}$$

$$\operatorname{Im}(A^{\top}) \otimes \operatorname{Ker}(A) = \mathbb{R}^{n} \tag{12b}$$

Remark. If r = rank(A), than

$$\dim \operatorname{Im}(A) = r, \qquad \dim \operatorname{Ker}(A^{\top}) = m - r \tag{13a}$$

$$\dim \operatorname{Im}(A^{\top}) = r, \qquad \dim \operatorname{Ker}(A) = n - r \tag{13b}$$

Proof. Proof of (11a) as presented in [1]. Let

$$A = \begin{pmatrix} \boldsymbol{a}_1 & \boldsymbol{a}_2 & \dots & \boldsymbol{a}_n \end{pmatrix} \Rightarrow A^{\top} = \begin{pmatrix} \boldsymbol{a}_1^{\top} \\ \boldsymbol{a}_2^{\top} \\ \dots \\ \boldsymbol{a}_n^{\top} \end{pmatrix}$$
 (14a)

$$\boldsymbol{x} \in \operatorname{Ker}\left(A^{\top}\right) \Rightarrow A^{\top}\boldsymbol{x} = \begin{pmatrix} \boldsymbol{a}_{1}^{\top}\boldsymbol{x} \\ \boldsymbol{a}_{2}^{\top}\boldsymbol{x} \\ \boldsymbol{a}_{n}^{\top}\boldsymbol{x} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ \dots \\ 0 \end{pmatrix}$$
 (14b)

$$\mathbf{y} \in \operatorname{Im}(A) \Rightarrow \exists \alpha_i \in \mathbb{R} \text{ such that } \mathbf{y} = \sum_{i=1}^n \alpha_i \mathbf{a}_i$$
 (14c)

Note that  $\boldsymbol{x}$  and  $\boldsymbol{y}$  are arbitrary vector elements of  $\operatorname{Ker}(A^{\top})$  and  $\operatorname{Im}(A)$ , respectively. Then we compute the dot product of  $\boldsymbol{x}$  and  $\boldsymbol{y}$ :

$$\langle \boldsymbol{x}, \boldsymbol{y} \rangle = \boldsymbol{y}^{\top} \boldsymbol{x} = \sum_{i=1}^{n} \alpha_{i} \boldsymbol{a}_{i}^{\top} \boldsymbol{x} = 0,$$
 (15)

since  $\mathbf{a}_i^{\top} \mathbf{x} = 0$ ,  $\forall i = \overline{1, n}$ . Consequently,  $\mathbf{x} \perp \mathbf{y}$  for all possible  $x \in \text{Ker}(A^{\top})$  and  $y \in \text{Im}(A)$ , which means that the two subspaces are the **orthogonal complement** for each other:

$$\operatorname{Im}(A) = \operatorname{Ker}(A^{\top})^{\perp}$$

$$\operatorname{Im}(A) \cap \operatorname{Ker}(A^{\top}) = \{0\}$$
(16)

Following this idea, we can conclude that the subspaces are linearly independent, therefore,

$$\dim\left(\operatorname{Im}(A)\otimes\operatorname{Ker}\left(A^{\top}\right)\right)=r+(m-r)=m. \tag{17}$$

This can only happend if *direct product* of the two spaces is  $\mathbb{R}^m$ , which completes the proof for (12a).

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**Proposition 5.** (Self-adjoint operator) Let  $\mathcal{A}: \mathbb{R}^n \to \mathbb{R}^n$ ,  $\mathcal{A}(x) = Ax$ , where  $A \in \mathbb{R}^{n \times n}$  is a symmetric matrix:  $A = A^{\top}$ . Than, as a consequence of Theorem 4, we have that

$$\operatorname{Im}(A) = \operatorname{Ker}(A)^{\perp} \text{ and } \operatorname{Im}(A) \otimes \operatorname{Ker}(A) = \mathbb{R}^{n}.$$

For more, see [2, Eq. (10.3)].

#### Proposition 6.

Singular value decomposition (SVD)

If we make the SVD for matrix  $A \in \mathbb{R}^{m \times n}$ 

$$A = U\Sigma V^{\top},\tag{18}$$

where

$$U \in \mathbb{R}^{m \times m}$$
 is unitary:  $U^*U = I_m$  (19a)

$$V \in \mathbb{R}^{n \times n}$$
 is unitary:  $V^*V = I_n$  (19b)

$$\Sigma \in \mathbb{R}^{m \times n}$$
 eigenvalues in the diagonal. (19c)

After this decomposition, the basis of the four subspaces (12) can be obtained as presented below.

Im(A): the first r colums of U

 $\operatorname{Ker}(A^{\top})$ : the last m-r columns of U

 $\operatorname{Im}(A^{\top})$ : the first r columns of V

Ker(A): the last n-r columns of V

In short

$${}_{"}A = \left[\operatorname{Im}(A) \operatorname{Ker}(A^{\top})\right] \Sigma \left[\operatorname{Im}(A^{\top}) \operatorname{Ker}(A)\right]^{\top}$$
(20)

# B Subspaces of the state space

Having a strictly proper (D = 0) MIMO LTI system:

$$\dot{x} = Ax + By 
y = Cx$$
(21)

The state space could be partitioned as follows:

$$X = X_{co} \otimes X_{c\bar{o}} \otimes X_{\bar{c}o} \otimes X_{\bar{c}\bar{o}} \tag{22}$$

where X.. are pairwise orthogonal subspaces of the state space, in other words:

$$X_{co} \perp X_{c\bar{o}}, \quad X_{co} \perp X_{\bar{c}o}, \quad X_{co} \perp X_{\bar{c}\bar{o}}, X_{c\bar{o}} \perp X_{\bar{c}o}, \quad X_{c\bar{o}} \perp X_{\bar{c}\bar{o}}, \quad X_{\bar{c}o} \perp X_{\bar{c}\bar{o}}.$$

$$(23)$$

# B.1 Unobservable subspace $X_{\overline{o}} = \operatorname{Ker}(\mathcal{O}_n)$ . Observable subspace $X_o = X_{\overline{o}}^{\perp} = \operatorname{Im}(\mathcal{O}_n^{\top})$ .

#### Lemma 7.

Linear independence of the first k rows of  $\mathcal{O}_n$ 

If rank  $(\mathcal{O}_n) = k \leq n$ , then the first k rows of  $\mathcal{O}_n$  are linearly independent, and any further rows of it can be expressed as the linear combination of the first k rows.

Formally:  $\forall i \in \mathbb{N} \ \exists \alpha \in \mathbb{R}^k$ , that  $CA^{k+i} = \alpha^{\top} \mathcal{O}_k$ , where  $\mathcal{O}_k \in \mathbb{R}^{k \times n}$  is defined as  $\mathcal{O}_k = \begin{pmatrix} C \\ CA \\ CA^{k-1} \end{pmatrix}$ . Remark.  $\mathbb{N} = \{0, 1, 2, ...\}, \ \mathbb{N}_+ := \mathbb{N} \setminus \{0\}$ .

*Proof.* The proof is given in the following three steps:

- (i) If k = n, the set of row vectors (also called as "covariant vectors")  $C, CA, ..., CA^{n-1}$  constitutes a linearly independent (covariant) basis for vector space  $\mathbb{R}^n$ , which means that any other row vectors in  $\mathbb{R}^n$  can be expressed by their linear combinations, the same as  $CA^{n+i}, \forall i \in \mathbb{N}$  can be.
- (ii) Let k be the first natural number, for which there exists  $\alpha \in \mathbb{R}^k$  such that  $CA^k = \alpha^\top \mathcal{O}_k$ . Then

 $CA^{k+1}$  can also be expressed by the covariant vectors of  $\mathcal{O}_k$ :

$$CA^{k+1} = \left(CA^{k}\right)A = \left(\sum_{j=1}^{k} \alpha_{j}CA^{j-1}\right)A = \sum_{j=1}^{k-1} \alpha_{j}CA^{j} + \alpha_{k}\sum_{j=1}^{k} \alpha_{j}CA^{j-1}$$
(24)

By induction, we have that for every every  $i \in \mathbb{N}$  there exists  $\alpha \in \mathbb{R}^k : CA^{k+i} = \alpha \mathcal{O}_k$ .

(iii) As a consequence of (ii), we can state that if  $\operatorname{rank}(\mathcal{O}_n) = k < n$ , that the first k rows of  $\mathcal{O}_n$  are linearly independent (i.e.  $\operatorname{rank}(\mathcal{O}_k) = k$ ).

**Lemma 8.** For every  $v \in \text{Im}(\mathcal{O}_n^{\top})$ , we have that  $A^{\top}v \in \text{Im}(\mathcal{O}_n^{\top})$ . In this sense, the observable subspace  $X_o = \text{Im}(\mathcal{O}_n^{\top}) = \text{Ker}(\mathcal{O}_n)^{\perp} \subseteq \mathbb{R}^n$  of the state space  $X = \mathbb{R}^n$  is invariant with respect to the linear transformation  $\mathcal{A}'(v) = A^{\top}v$ , i.e.  $\mathcal{A}'(X_o) = X_o$ .

*Proof.* Let  $a(s) = \det(sI - A) = a_0 + a_1s + ... + a_ns^n$ . Due to Cayley-Hamilton theorem, we have that

$$a(A) = 0 \implies A^n = \frac{1}{a_n} (a_0 I + a_1 A + \dots + a_{n-1} A^{n-1})$$
 (25)

### Proposition 9.

 $x(0) \in \text{Ker}(\mathcal{O}_n) \text{ and } u(t) = 0 \implies y(t) = 0$ 

Let rank  $(\mathcal{O}_n) = k < n$ . If  $x_0 \in \text{Ker}(\mathcal{O}_n)$  and  $u \equiv 0$ , than y(t) = 0 for every t > 0, i.e

$$x(t) = e^{At} x_0 \in \text{Ker}(\mathcal{O}_n)$$

In other words, if there is no input signal (u(t) = 0) and the initial condition  $x_0$  belongs to the unobservable subspace  $Ker(\mathcal{O}_n)$ , than the state response of the system  $x(t) = e^{At}x_0$  will remain in this subspace.

*Proof.* As a consequence of Proposition 7, we have that if  $CA^kx_0 = 0$  for  $k = \overline{0, n-1}$ , than  $CA^kx_0 = 0$  holds for every  $k \in \mathbb{N}$ . If we consider the Taylor expansion of matrix exponent  $e^{At}$ , we have:

$$CA^{k}e^{At}x_{0} = \sum_{j=0}^{\infty} \frac{t^{k}}{k!} \cdot \underbrace{CA^{k+j}x_{0}}_{0} = 0 \quad \forall k = \overline{0, n-1} \quad \Rightarrow \quad \mathcal{O}_{n}e^{At}x_{0} = 0 \Leftrightarrow e^{At}x_{0} \in \operatorname{Ker}(\mathcal{O}_{n})$$
 (26)

Consequently, for a given unobservable state space model (A,B,C,D) if we start the system from the unobservable subspace  $x(0) \in \text{Ker}(\mathcal{O}_n)$  and having a zero input  $(u \equiv 0)$  the output will be zero y(t) = 0, for every t > 0.

## Proposition 10.

Same output for all initial state of an unobservable class

Let us denote  $v_1, ..., v_{n-k} \in \mathbb{R}^n$ , k < n the basis vectors of the null space of  $\mathcal{O}_n$ :

$$\operatorname{Ker}(\mathcal{O}_n) = \left\{ \alpha_1 v_1 + \ldots + \alpha_{n-k} v_{n-k} = \alpha^\top N \mid \alpha \in \mathbb{R}^{n-k} \right\}, \text{ where } N := \begin{pmatrix} v_1 & \ldots & v_{n-k} \end{pmatrix} \in \mathbb{R}^{n \times (n-k)}$$

Matrix N is called an annihilator of  $\mathcal{O}_n$ , since  $\mathcal{O}_n N = 0_{n \times (n-k)}$ . Now we introduce the following notations:

$$x_0 + \operatorname{Ker}(\mathcal{O}_n) := \left\{ x_0 + \alpha^\top N \mid \alpha \in \mathbb{R}^{n-k} \right\}$$
 (27)

From any initial condition  $x(0) \in x_0 + \operatorname{Ker}(\mathcal{O}_n)$  and for a given input u(t), the system will produce the same output y(t).

*Proof*. The explicit solution of the state space model is

$$y(t) = Ce^{At}x(0) + C\int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$$
(28)

Considering an initial condition  $x(0) = x_0 + \alpha^\top N \in x_0 + \operatorname{Ker}(\mathcal{O}_n)$  with an arbitrary  $\alpha \in \mathbb{R}^{n-k}$ , and keeping in mind, that  $\alpha^\top N \in \operatorname{Ker}(\mathcal{O}_n)$  (i.e.  $CA^i\alpha^\top N = 0$  for all  $i \in \mathbb{N}$ ) we obtain:

$$y(t) = Ce^{At}\left(x_0 + \alpha^{\top}N\right) + C\int_0^t e^{A(t-\tau)}Bu(\tau)d\tau = Ce^{At}x_0 + C\int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$$
(29)

Finally, we can observe that the expression for y(t) does not depend on  $\alpha$ . It depends only on the input u(t) and on  $x_0$ , furthermore, for each  $x_0$  we obtain different outputs,  $x_0$  defines the unobservability class, that the system is actually in. If we can find a particular solution x(t) for the (under-determined) linear equation system

$$\mathcal{Y}(t) = \mathcal{O}_n x(t) + \mathcal{T}\mathcal{U}(t) \qquad [\text{lec\_03.pdf}, \text{pg. } 10/31]$$
(30)

we can determine the actual unobservability class of the system, but we have no further informations about the state vector itself.

Remark. Set  $x_0 + \text{Ker}(\mathcal{O}_n)$  is not a subspace of  $\mathbb{R}^n$ , since many properties of the vector space broke (eg. does not have a unity element), however, it is a k dimensional manifold (sokaság) in vector space  $\mathbb{R}^n$ .

# **B.2** Controllable subspace $X_c = \operatorname{Im}(\mathcal{C}_n)$ . Uncontrollable subspace $X_{\overline{c}} = X_c^{\perp} = \operatorname{Ker}(\mathcal{C}_n^{\top})$ .

**Lemma 11.** If (A, B, C) is not controllable rank $(C_n) = k < n$ , the first k columns of  $C_n$  are linearly independent.

**Lemma 12.** For every  $v \in \text{Im}(\mathcal{C}_n)$ , vector  $Av \in \text{Im}(\mathcal{C}_n)$ . In this sense, the controllable subspace  $X_c = \text{Im}(\mathcal{C}_n) \subseteq \mathbb{R}^n$  of the state space  $X = \mathbb{R}^n$  is invariant with respect to the linear transformation  $\mathcal{A}(v) = Av$ , i.e.  $\mathcal{A}(X_c) = X_c$ .

*Proof*. Let  $v \in X_c = \text{span} \langle B, AB, ..., A^{n-1}B \rangle$ , therefore, there exist real values  $\alpha_1, ..., \alpha_n \in \mathbb{R}$ , such that

$$v = \sum_{i=1}^{n} \alpha_i A^{i-1} B \Rightarrow Av = \sum_{i=1}^{n} \alpha_i A^i B. \tag{31}$$

It is obvious that  $A^iB \in X_c$  for all  $i = \overline{1, n-1}$ , furthermore, due to Lemma 11,  $A^nB$  can be expressed as the linear combination of vectors  $A^{i-1}$ , B,  $i = \overline{1, n}$ . Finally, we have that  $Av \in X_c$ .

# Proposition 13.

$$x_0 \in \operatorname{Im}(\mathcal{C}_n) \Rightarrow x(t) \in \operatorname{Im}(\mathcal{C}_n)$$

If the initial condition  $x(0) = x_0$  belongs to the controllable subspace of the state space, than the solution x(t) will also belong to it. Formally:

$$x_0 \in \operatorname{Im}(\mathcal{C}_n) \Rightarrow x(t) \in \operatorname{Im}(\mathcal{C}_n) \, \forall t \ge 0.$$
 (32)

If the initial condition is not an element of  $\text{Im}(\mathcal{C}_n)$ , but the system is stable, than the trajectory will tend exponentially to the controllable subspace of the state space, i.e.

$$A \prec 0 \Rightarrow x(t) \to \operatorname{Im}(\mathcal{C}_n)$$
 (33)

*Proof*. If  $x_0 \in \text{Im}(\mathcal{C}_n) = X_c$ , than

$$x(t) = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau = \sum_{k=0}^\infty \frac{t^k}{k!} \underbrace{A^k x_0}_{\in X} + \int_0^t \sum_{k=0}^\infty \frac{(t-\tau)^k}{k!} \underbrace{A^k B}_{\in X_c} u(\tau)d\tau \in X_c.$$
(34)

If  $x_0 \notin X_c$  but  $A \prec 0$  (is negative definite), than

$$x(t) = \underbrace{e^{At}x_0}_{\to 0} + \underbrace{\int_0^t e^{A(t-\tau)}Bu(\tau)d\tau}_{\in X_c} \to X_c.$$
(35)

So, the solution tends to the controllable subspace.

**Theorem 14.** (Control the system to a given state) If the system is controllable, there exists an input

$$u(t) = -B^{\top} e^{A^{\top}(t_1 - t)} P^{-1}(t_1) \left( e^{At_1} x_0 - x_1 \right), \text{ where } P(t) = \int_0^t e^{A\tau} B B^{\top} e^{A^{\top} \tau} d\tau, \quad t \in [0, t_1],$$
 (36)

which leads the system from x(0) to  $x(t_1) = x_1$  in a finite time  $t_1 < \infty$ .

*Proof*. A proof for it can be found in [3, Theorem 2.21].

# B.3 Controllability staircase form

#### Proposition 15.

Controllability staircase form

We construct the following transformation matrix  $T^{-1} = S = (v_1, ..., v_k, w_{k+1}, ..., w_n)$ , where  $[v] = [v_1, ..., v_k]$  is the orthonormal (ON) basis of  $X_c = \operatorname{Im}(\mathcal{C}_n)$  and  $[w] = [w_{k+1}, ..., w_n]$  is the ON basis of  $X_{\overline{c}} = \operatorname{Im}(\mathcal{C}_n)^{\perp} = \operatorname{Ker}(\mathcal{C}_n^{\top})$ . Then the transformed matrices will have the form:

$$\bar{A} = TAT^{-1} = \begin{pmatrix} \bar{A}_{11} & \bar{A}_{12} \\ 0_{(n-k)\times k} & \bar{A}_{22} \end{pmatrix}$$
 (37a)  $\bar{B} = TB = \begin{pmatrix} \bar{B}_1 \\ 0_{(n-k)\times 1} \end{pmatrix}$ 

Using SVD: 
$$C_n = U_c \Sigma_c V_c^{\top}, S := U_c$$

*Proof.* (For simplicity, only for SISO) Since  $X_c$  and  $X_{\overline{c}}$  are orthogonal complement of each other (i.e.  $X_c \otimes X_{\overline{c}} = \mathbb{R}^n$ ), [v, w] is an ON basis of  $\mathbb{R}^n$ . In other words: S is an orthogonal matrix with the well-known properties:

$$S^{\top}S = I_n \quad \Rightarrow \quad S^{-1} = S^{\top} = \begin{pmatrix} V^{\top} \\ W^{\top} \end{pmatrix}, \quad \text{where } V = \begin{pmatrix} v_1, ..., v_k \end{pmatrix} \text{ and } W = \begin{pmatrix} w_{k+1}, ..., w_n \end{pmatrix}$$
 (38)

Furthermore,  $V^{\top}W = 0_{k \times (n-k)}$  and  $W^{\top}V = 0_{(n-k) \times k}$  (39). Then the transformed matrix  $\bar{A}$  will be:

$$\bar{A} = TAT^{-1} = S^{\top}AS = \begin{pmatrix} V^{\top} \\ W^{\top} \end{pmatrix} A \begin{pmatrix} V & W \end{pmatrix} = \begin{pmatrix} V^{\top}AV & V^{\top}AW \\ W^{\top}AV & W^{\top}AW \end{pmatrix}. \tag{40}$$

The columns of V are elements of  $X_c$ , therefore, the columns of AV are also elements of  $X_c$ . The columns of W are the basis vectors of  $X_{\overline{c}} = X_c^{\perp}$ , therefore,  $W^{\top}AV = 0_{(n-k)\times k}$ . The transformed matrix  $\overline{B}$  will be:

$$\bar{B} = TB = S^{\top}B = \begin{pmatrix} V^{\top} \\ W^{\top} \end{pmatrix} B = \begin{pmatrix} V^{\top}B \\ W^{\top}B \end{pmatrix}. \tag{41}$$

Since  $B \in X_c$ ,  $w_j \in X_c^{\perp}$ ,  $W^{\top}B = 0_{(n-k)\times 1}$ ,  $j = \overline{k+1, n}$ .

## B.4 Observability staircase form

# Proposition 16.

Observability staircase form

We construct the following transformation matrix  $T^{-1} = S = (v_1, ..., v_k, w_{k+1}, ..., w_n)$ , where  $[v] = [v_1, ..., v_k]$  is the orthonormal (ON) basis of  $X_o = \text{Ker}(\mathcal{O}_n)^{\perp} = \text{Im}(\mathcal{O}_n^{\top})$  and  $[w] = [w_{k+1}, ..., w_n]$  is the ON basis of  $X_{\overline{o}} = \text{Ker}(\mathcal{O}_n)$ . Then the transformed matrices will have the form:

$$\bar{A} = TAT^{-1} = \begin{pmatrix} \bar{A}_{11} & 0_{k \times (n-k)} \\ \bar{A}_{21} & \bar{A}_{22} \end{pmatrix}$$
(42a) 
$$\bar{C} = CT^{-1} = (\bar{C}_1 \quad 0_{1 \times (n-k)})$$
(42b)

Using SVD: 
$$\mathcal{O}_n = U_o \Sigma_o V_o^{\top}, S := V_o$$

*Proof.* (For simplicity, only for SISO) Since  $X_o$  and  $X_{\overline{o}}$  are orthogonal complement of each other (i.e.  $X_o \otimes X_{\overline{o}} = \mathbb{R}^n$ ), [v, w] is an ON basis of  $\mathbb{R}^n$ . In other words: S is an orthogonal matrix with the well-known properties:

$$S^{\top}S = I_n \quad \Rightarrow \quad S^{-1} = S^{\top} = \begin{pmatrix} V^{\top} \\ W^{\top} \end{pmatrix}, \quad \text{where } V = \begin{pmatrix} v_1, ..., v_k \end{pmatrix} \text{ and } W = \begin{pmatrix} w_{k+1}, ..., w_n \end{pmatrix}$$
 (43)

Furthermore,  $V^{\top}W = 0_{k \times (n-k)}$  and  $W^{\top}V = 0_{(n-k) \times k}$  (44). The transformed matrix  $\bar{A}$  will be:

$$\bar{A} = TAT^{-1} = S^{\top}AS = \begin{pmatrix} V^{\top} \\ W^{\top} \end{pmatrix} A \begin{pmatrix} V & W \end{pmatrix} = \begin{pmatrix} V^{\top}AV & V^{\top}AW \\ W^{\top}AV & W^{\top}AW \end{pmatrix}. \tag{45}$$

The columns of V are elements of  $X_o$ , therefore, the columns of  $A^{\top}V$  are also elements of  $X_o$ . The columns of W are the basis vectors of  $X_{\overline{c}} = X_c^{\perp}$ , therefore,  $(A^{\top}V)^{\top}W = V^{\top}AW = 0_{k\times(n-k)}$ . The transformed matrix  $\overline{C}$  will be:

$$\bar{C} = CT^{-1} = CS = C \begin{pmatrix} V & W \end{pmatrix} = \begin{pmatrix} CV & CW \end{pmatrix}. \tag{46}$$

Since 
$$C^{\top} \in X_o$$
,  $w_j \in X_o^{\perp}$ ,  $CW^{\top} = 0_{1 \times (n-k)}$ ,  $j = \overline{k+1, n}$ .

**Proposition 17.** If (A, C) has unobservable mode (i.e. is unobservable), there exists  $x \in \mathbb{R}^n$ , such that  $Ax = \lambda x$  and Cx = 0. Consequently,  $\lambda$  is a "decoupling zero" of (A, B, C, D), since

$$M = \begin{pmatrix} A - \lambda I & B \\ C & 0 \end{pmatrix} \text{ is singular,} \tag{47}$$

namely there exists  $\xi = \begin{pmatrix} x \\ 0 \end{pmatrix} \neq 0$  such that  $M\xi = 0$ . Or in other words, the kernel space of M is not empty, meaning that M is singular.

Proposition 18. The input decoupling zeros are equal to the eigenvalues of the uncontrollable subsystem.

*Proof*. We assume that (A, B) is uncontrollable:

$$C_n = \begin{pmatrix} B & AB & \dots & A^{n-1}B \end{pmatrix} \in \mathbb{R}^{n \times mn}$$
(48)

is rank deficient, that implies a nonempty kernel space  $\operatorname{Ker}(\mathcal{C}_n^\top) \subset \mathbb{R}^n$ , namely, there exists  $x \in \mathbb{R}^n$  such that  $\mathcal{C}_n^\top x = 0$ . Alternatively, we have that

$$\begin{cases}
B^{\top} x = 0 \\
B^{\top} A^{\top} x = 0 \\
\dots \\
B^{\top} \left(A^{\top}\right)^{n-1} x = 0
\end{cases}$$
(49)

# B.5 Kalman decomposition

We produce a controllability staircase form decomposition on the system, than on both subsystems (controllable and uncontrollable) we produce an observability staircase form decomposition.