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Lecture Course: Advanced Systems Theory

Chapter 4-Lecture5: Disturbance decoupling problem

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Recapitulation-4.1 controlled invariant subspaces

- › **A-invariant** subspace and **invariant** subspace of $\dot{x} = Ax$.
- › Example: a subspace $\mathcal{V} \subseteq \text{im } A$ is **not necessarily** A -invariant.
- › Definition 4.1: **controlled invariant subspace** ((A, B) -invariant subspace)

Theorem (4.2)

The following is equivalent:

(i) \mathcal{V} is controlled invariant; (ii) $A\mathcal{V} \subseteq \mathcal{V} + \text{im } B$; (iii) $\exists F : (A + BF)\mathcal{V} \subseteq \mathcal{V}$.

- › Calculate **friend feedback** F , i.e., F such that $(A + BF)\mathcal{V} \subseteq \mathcal{V}$.
- › $\mathcal{V}^*(\mathcal{K})$ is the **largest controlled invariant subspace contained in \mathcal{K}** .

Recapitulation-4.2 Disturbance decoupling

› Consider a control system

$$\Sigma_{d,z} : \begin{cases} \dot{x} = Ax + Ed \\ z = Hx, \end{cases}$$

› The output $z(t) = He^{At}x_0 + \int_0^t T(t-\tau)d(\tau)d\tau$, where $T(t-\tau) = He^{A(t-\tau)}E$.

Definition

$\Sigma_{d,z}$ is called **disturbance decoupled** if $T(t) = He^{At}E = 0, \forall t \geq 0$, or if $G(s) = H(sI - A)^{-1}E = 0$.

Theorem (4.6)

$\Sigma_{d,z}$ is disturbance decoupled iff \exists an ***A*-invariant subspace** \mathcal{V} s.t.

$$\text{im } E \subseteq \mathcal{V} \subseteq \ker H.$$

Recapitulation-Questions

Question 1

Let $A \in \mathbb{R}^{2 \times 2}$, $B \in \mathbb{R}^2$ such that (A, B) is controllable, which one of the following subspaces is not (A, B) -invariant?

- (i) $\text{im } B$
- (ii) $\text{im } AB$
- (iii) $\text{im } A$
- (iv) $\text{im}(sI - A)$, $s \in \sigma(A)$.

Question 2

Let $A = \begin{bmatrix} 1 & -1 & 1 \\ 0 & 0 & 2 \\ 1 & 1 & 2 \end{bmatrix}$, $B = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ and $\mathcal{V} = \text{im} \begin{bmatrix} 0 & 1 \\ 0 & 1 \\ 1 & 0 \end{bmatrix}$, which F is a friend feedback of (A, B, \mathcal{V}) ?

- (i) $F = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix}$
- (ii) $F = \begin{bmatrix} 1 & -1 & 1 \end{bmatrix}$.
- (iii) $F = \begin{bmatrix} 1 & 0 & -1 \end{bmatrix}$.

Question 3

Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, $E = \begin{bmatrix} e \\ f \end{bmatrix}$, $H = \begin{bmatrix} g & h \end{bmatrix}$, $\mathcal{V} = \text{im} \begin{bmatrix} 0 \\ 1 \end{bmatrix}$, which of the following satisfies that $\text{im } E \subseteq \mathcal{V} \subseteq \ker H$ and \mathcal{V} is A -inv.?

- (i) $c = 0$, $f = 0$, $g = 0$.
- (ii) $b = 0$, $e = 0$, $h = 0$.
- (iii) $c = 0$, $e = 0$, $h = 0$.
- (iv) $b = 0$, $f = 0$, $g = 0$.

Question 4

Except for item (ii), which system in Question 3 is also disturbance decoupled?

- (i). (iii). (iv).

Disturbance decoupling problem (DDP)

Disturbance decoupling problem (DDP)

Given a system

$$\Sigma_{u,d,z} : \begin{cases} \dot{x} = Ax + Bu + Ed \\ z = Hx, \end{cases}$$

find $F : \mathcal{X} \rightarrow \mathcal{U}$ s.t. the state feedback $u = Fx$ renders the closed loop systems:

$$\Sigma_{d,z} : \begin{cases} \dot{x} = (A + BF)x + Ed \\ z = Hx, \end{cases}$$

is **disturbance decoupled**.

Remark: Recall from Thm 4.6 that the above system $\Sigma_{d,z}$ is disturbance decoupled iff \exists an $(A + BF)$ -invariant subspace \mathcal{V}_F s.t. $\text{im } E \subseteq \mathcal{V}_F \subseteq \ker H$.

Disturbance decoupling problem (DDP)

Disturbance decoupling problem (DDP)

Given a system $\Sigma_{u,d,z} = (A, B, E, H)$ find $F : \mathcal{X} \rightarrow \mathcal{U}$ s.t. the state feedback $u = Fx$ renders the closed loop systems $\Sigma_{d,z} = (A + BF, E, H)$ disturbance decoupled.

Q: What does the system block diagram for DDP look like ?

Theorem (4.8)

Given a system $\Sigma_{u,d,z} = (A, B, E, H)$, The DDP is solvable for $\Sigma_{u,d,z}$ iff there exists an (A, B) -invariant subspace with $\text{im } E \subseteq \mathcal{V} \subseteq \ker H$.

Disturbance decoupling problem (DDP)

Theorem (4.8)

The DDP is solvable iff there exists an (A, B) -invariant subspace with $\text{im } E \subseteq \mathcal{V} \subseteq \ker H$.

Proof.

Only if: DDP is solvable for $\Sigma_{u,d,z}$.

$\Rightarrow \exists F : \exists$ an $(A + BF)$ -invariant subspace \mathcal{V}_F s.t. $\text{im } E \subseteq \mathcal{V}_F \subseteq \ker H$.

Thm.4.2 $\Rightarrow \exists$ an (A, B) -invariant subspace \mathcal{V} with $\text{im } E \subseteq \mathcal{V} \subseteq \ker H$.

If: Suppose that \exists an (A, B) -invariant subspace \mathcal{V} with $\text{im } E \subseteq \mathcal{V} \subseteq \ker H$ and denote $\dim \mathcal{V} = k$.

For $\Sigma_{u,d,z} = (A, B, E, H)$, choose new coordinates

$$\tilde{x} = \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} = \begin{bmatrix} P_1 \\ P_2 \end{bmatrix} x = Px,$$

where $P_2 : \mathcal{X} \rightarrow \mathbb{R}^{n-k}$ s.t. $\text{rk } P_2 = n - k$ and $\text{im } P_2^T = \mathcal{V}^\perp$, and where $P_1 : \mathcal{X} \rightarrow \mathbb{R}^k$ is any s.t. P is invertible. □

Disturbance decoupling problem (DDP)

Proof of Thm 4.8 continue.

Then

$$\Sigma_{u,d,z} : \begin{cases} \dot{x} = Ax + Bu + Ed \\ z = Hx, \end{cases} \quad \tilde{x} \stackrel{u=Px}{\Rightarrow} \tilde{\Sigma}_{u,d,z} : \begin{cases} \dot{\tilde{x}} = PAP^{-1}\tilde{x} + PBu + PEd \\ z = HP^{-1}\tilde{x}. \end{cases}$$

\mathcal{V} is an (A, B) -invariant subspace $\Rightarrow \exists F : (A + BF)\mathcal{V} \subseteq \mathcal{V} \Rightarrow$
 $\exists F : P(A + BF)P^{-1}P\mathcal{V} \subseteq P\mathcal{V} \Rightarrow \exists \tilde{F} = FP^{-1} : (\tilde{A} + \tilde{B}\tilde{F})\tilde{\mathcal{V}} \subseteq \tilde{\mathcal{V}}.$

Thus the closed loop system of $\tilde{\Sigma}_{u,d,z}$ takes the form

$$\begin{cases} \dot{\tilde{x}} = PAP^{-1}\tilde{x} + PBu + PEd \\ z = HP^{-1}\tilde{x}. \end{cases} \quad u \stackrel{u=\tilde{F}x}{\Rightarrow} \begin{cases} \begin{bmatrix} \dot{\tilde{x}}_1 \\ \dot{\tilde{x}}_2 \end{bmatrix} = \begin{bmatrix} \tilde{A}_1 & \tilde{A}_2 \\ 0 & \tilde{A}_4 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} + \begin{bmatrix} E_1 \\ 0 \end{bmatrix} d \\ z = \begin{bmatrix} 0 & \tilde{H}_2 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \end{bmatrix} \end{cases}$$

Now it is clear that z is independent of d , i.e., $G_{dz}(s) = \tilde{H}(sI - \tilde{A})^{-1}\tilde{E} = 0.$ □

Recall that $\mathcal{V}^*(\ker H)$ is **the largest controlled invariant subspace contained in $\ker H$** , i.e., (i) \mathcal{V}^* is controlled invariant; (ii) $\mathcal{V}^* \subseteq \ker H$; (iii) any other $\mathcal{V} : A\mathcal{V} \subseteq \mathcal{V} + \text{im } B \Rightarrow \mathcal{V} \subseteq \mathcal{V}^*$.

Corollary 4.9

The DDP is solvable iff $\text{im } E \subseteq \mathcal{V}^*(\ker H)$.

Proof.

Only if. DDP is solvable $\stackrel{\text{Thm 4.8}}{\Rightarrow} \exists (A, B)\text{-inv. } \mathcal{V} : \text{im } E \subseteq \mathcal{V} \subseteq \ker H$ and $\mathcal{V} \subseteq \mathcal{V}^*(\ker H) \subseteq \ker H$ (by Thm 4.5)
 $\Rightarrow \text{im } E \subseteq \mathcal{V}^*(\ker H)$.

If. Take $\mathcal{V} = \mathcal{V}^*(\ker H)$, we have $\text{im } E \subseteq \mathcal{V} \subseteq \ker H \stackrel{\text{Thm 4.8}}{\Rightarrow}$ DDP is solvable. □

4.3 The invariant subspace algorithm (ISA)

- › Consider $\Sigma : \dot{x} = Ax + Bu$, given a subspace $\mathcal{K} \subseteq \mathcal{X}$, find $\mathcal{V}^*(\mathcal{K})$.
- › Recall definition of **inverse image/preimage**: $A^{-1}\mathcal{V} := \{x \mid Ax \in \mathcal{V}\}$.

Question 5a

Which one of the following is **correct**?

- (i) $AA^{-1}\mathcal{V} = \mathcal{V}$. (ii) $AA^{-1}\mathcal{V} \subseteq \mathcal{V}$. (iii) $AA^{-1}\mathcal{V} \supseteq \mathcal{V}$.

Question 5b

Which one of the following is **correct**?

- (i) $A^{-1}A\mathcal{V} = \mathcal{V}$. (ii) $A^{-1}A\mathcal{V} \subseteq \mathcal{V}$. (iii) $A^{-1}A\mathcal{V} \supseteq \mathcal{V}$.

Motivation of constructing $\mathcal{V}^*(\mathcal{K})$

$$\begin{aligned} \mathcal{V}^*(\mathcal{K}) &= \{x_0 \mid \exists u \in \mathcal{U} : x_u(x_0, t) \in \mathcal{K}, \forall t \geq 0\} \Rightarrow x(t) \in \mathcal{K} = \mathcal{V}_0 \\ &\Rightarrow \dot{x}(t) \in \mathcal{K} = \mathcal{V}_0 \Rightarrow Ax(t) \in \mathcal{V}_0 + \text{im } B \Rightarrow x(t) \in A^{-1}(\mathcal{V}_0 + \text{im } B) \\ &\stackrel{x(t) \in \mathcal{V}_0}{\Rightarrow} x(t) \in \mathcal{V}_1 = A^{-1}(\mathcal{V}_0 + \text{im } B) \Rightarrow \dot{x}(t) \in \mathcal{V}_1 \Rightarrow x(t) \in \mathcal{V}_2 \Rightarrow \dots \end{aligned}$$

4.3 The invariant subspace algorithm (ISA)

Algorithm

Given $A \in \mathbb{R}^{n \times n}$, $B \in \mathbb{R}^{n \times m}$ and $\mathcal{K} \subseteq \mathbb{R}^n$, define

$$\begin{cases} \mathcal{V}_0 := \mathcal{K}, \\ \mathcal{V}_{k+1} := \mathcal{K} \cap A^{-1}(\mathcal{V}_k + \text{im } B), \quad k = 1, 2, \dots \end{cases}$$

Theorem (4.10)

Let $\mathcal{K} \subseteq X$ and $\mathcal{V}_0, \mathcal{V}_1, \mathcal{V}_2, \dots$ as defined in the above algorithm. Then

- (i) $\mathcal{V}_0 \supseteq \mathcal{V}_1 \supseteq \mathcal{V}_2 \supseteq \dots$, (*non-increasing*)
- (ii) $\exists k \leq \dim \mathcal{K} : \mathcal{V}_k = \mathcal{V}_{k+1}$, (*stable index*)
- (iii) $\mathcal{V}_k = \mathcal{V}_{k+1} \Rightarrow \mathcal{V}_k = \mathcal{V}_l, \forall l \geq k$, (*stable*)
- (iv) $\mathcal{V}_k = \mathcal{V}_{k+1} \Rightarrow \mathcal{V}^*(\mathcal{K}) = \mathcal{V}_k$. (*limit*)

4.3 The invariant subspace algorithm (ISA)

Theorem (4.10)

Let $\mathcal{K} \subseteq X$ and $\mathcal{V}_0, \mathcal{V}_1, \mathcal{V}_2, \dots$ as defined in the above algorithm. Then

- (i) $\mathcal{V}_0 \supseteq \mathcal{V}_1 \supseteq \mathcal{V}_2 \supseteq \dots$, (ii) $\exists k \leq \dim \mathcal{K} : \mathcal{V}_k = \mathcal{V}_{k+1}$,
 (iii) $\mathcal{V}_k = \mathcal{V}_{k+1} \Rightarrow \mathcal{V}_k = \mathcal{V}_l, \forall l \geq k$, (iv) $\mathcal{V}_k = \mathcal{V}_{k+1} \Rightarrow \mathcal{V}^*(\mathcal{K}) = \mathcal{V}_k$.

Proof.

(i): $\mathcal{V}_0 \supseteq \mathcal{V}_1$ is clear by definition. Induction: suppose that $\mathcal{V}_k \supseteq \mathcal{V}_{k+1}$, then

$$\mathcal{V}_{k+2} = \mathcal{K} \cap A^{-1}(\mathcal{V}_{k+1} + \text{im } B) \stackrel{\mathcal{V}_{k+1} \subseteq \mathcal{V}_k}{\subseteq} \mathcal{K} \cap A^{-1}(\mathcal{V}_k + \text{im } B) = \mathcal{V}_{k+1}.$$

(iii): $\mathcal{V}_{k+1} = \mathcal{V}_k \Rightarrow \mathcal{V}_{k+2} = \mathcal{K} \cap A^{-1}(\mathcal{V}_{k+1} + \text{im } B) = \mathcal{K} \cap A^{-1}(\mathcal{V}_k + \text{im } B) = \mathcal{V}_{k+1}$.

(ii): (i)+(iii) $\Rightarrow \mathcal{V}_0 \supset \mathcal{V}_1 \supset \mathcal{V}_2 \supset \dots \supset \mathcal{V}_k = \mathcal{V}_{k+1} = \mathcal{V}_{k+2} = \dots$. Thus

$$0 \leq \dim \mathcal{V}_k \leq \dim \mathcal{V}_{k-1} - 1 \leq \dim \mathcal{V}_{k-1} - 1 \leq \dim \mathcal{V}_{k-2} - 2 \leq \dots \leq \dim \mathcal{V}_0 - k = \dim \mathcal{K} - k.$$

4.3 The invariant subspace algorithm (ISA)

Theorem (4.10)

Let $\mathcal{K} \subseteq X$ and $\mathcal{V}_0, \mathcal{V}_1, \mathcal{V}_2, \dots$ as defined in the above algorithm. Then (i) ..., (ii) ..., (iii) ..., (iv) $\mathcal{V}_k = \mathcal{V}_{k+1} \Rightarrow \mathcal{V}^*(\mathcal{K}) = \mathcal{V}_k$.

Proof.

(iv) $\mathcal{V}_k \subseteq \mathcal{V}_0 = \mathcal{K}$ and $A\mathcal{V}_k = A\mathcal{V}_{k+1} \subseteq \mathcal{V}_k + \text{im } B \Rightarrow \mathcal{V}_k$ is (A, B) -invariant subspace and $\mathcal{V}_k \subseteq \mathcal{K} \Rightarrow \mathcal{V}_k \subseteq \mathcal{V}^*(\mathcal{K})$.

We **claim** that $\mathcal{V}^*(\mathcal{K}) \subseteq \mathcal{V}_l, \forall l \geq 0$. Clearly, $\mathcal{V}^*(\mathcal{K}) \subseteq \mathcal{V}_0 = \mathcal{K}$. By induction,

$$\begin{aligned} A\mathcal{V}^*(\mathcal{K}) \subseteq \mathcal{V}^*(\mathcal{K}) + \text{im } B &\subseteq \mathcal{V}_{l-1} + \text{im } B \Rightarrow \mathcal{V}^*(\mathcal{K}) \subseteq A^{-1}(\mathcal{V}_{l-1} + \text{im } B) \\ &\Rightarrow \mathcal{K} \cap \mathcal{V}^*(\mathcal{K}) \subseteq \mathcal{K} \cap A^{-1}(\mathcal{V}_{l-1} + \text{im } B) = \mathcal{V}_l. \end{aligned}$$

Thus by **claim**, we have $\mathcal{V}^*(\mathcal{K}) \subseteq \mathcal{V}_l \subseteq \mathcal{V}_k$. □

4.3 The invariant subspace algorithm (ISA)

Example

Let

$$A = \begin{bmatrix} 3 & 2 & 0 \\ -3 & -1 & 1 \\ 4 & 2 & -1 \end{bmatrix}, \quad B = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}, \quad C = [1 \quad 1 \quad 0]$$

Calculate $\mathcal{V}^*(\ker C)$, i.e., the largest (A, B) -invariant subspace contained in $\ker C$.

Solution

$$\mathcal{V}_0 = \ker C = \text{im} \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathcal{V}_1 = \ker C \cap A^{-1}(\mathcal{V}_0 + \text{im } B) = \text{im} \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \cap \left(\begin{bmatrix} 3 & 2 & 0 \\ -3 & -1 & 1 \\ 4 & 2 & -1 \end{bmatrix}^{-1} \text{im} \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \right) =$$

$$\text{im} \begin{bmatrix} 1 & 0 \\ -1 & 0 \\ 0 & 1 \end{bmatrix} \cap \text{im} \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 0 & -1 \end{bmatrix} = \text{im} \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

$$\mathcal{V}_2 = \mathcal{V}_1 \Rightarrow \mathcal{V}^*(\ker C) = \mathcal{V}_2 = \mathcal{V}_1.$$