

# Linear autonomous systems partially ordered by the sharp partial order

Alicia Herrero   Néstor Thome

Instituto Universitario de Matemática Multidisciplinar,  
Universitat Politècnica de València

Ministerio de Economía, Industria y Competitividad of Spain  
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# Outline of the talk

- 1 Introduction
- 2 Sharp partial ordered autonomous systems
- 3 Conclusions and Future work

# Linear autonomous systems

## Linear autonomous system

A linear autonomous system can be given by

$$\begin{cases} x(k+1) = A x(k) \\ x(0) = x_0 \end{cases}$$

for  $A \in \mathbb{R}^{n \times n}$  and  $k = 0, 1, \dots$

## Solution

The solution of a linear autonomous system is

$$x(k) = A^k x_0$$

# Sharp partial ordering

## Group inverse

Given a matrix  $A \in \mathbb{R}^{n \times n}$ , with index less than or equal to 1. The **group inverse of  $A$**  is the matrix  $A^\#$  satisfying

$$AA^\#A = A, \quad A^\#AA^\# = A^\# \quad \text{and} \quad AA^\# = A^\#A$$

## Sharp partial ordering

Given  $A, B \in \mathbb{R}^{n \times n}$ , with  $\text{ind}(A) \leq 1$ ,  $\text{ind}(B) \leq 1$ ,

$$A \stackrel{\#}{\leq} B \iff \begin{cases} AA^\# = BA^\# \\ A^\#A = A^\#B \end{cases}$$

where  $A^\#$  is the group inverse of  $A$ .

# Properties

## Invariance under similarities

Given  $A, B \in \mathbb{R}^{n \times n}$ , with  $\text{ind}(A) \leq 1$  and  $\text{ind}(B) \leq 1$ , and  $P \in \mathbb{R}^{n \times n}$  nonsingular,

$$A \stackrel{\#}{\leq} B \implies PAP^{-1} \stackrel{\#}{\leq} PBP^{-1}$$

## Spectrum

Given  $A, B \in \mathbb{R}^{n \times n}$  such that  $A \stackrel{\#}{\leq} B$ , then

$$\text{Spectrum}(A) \subseteq \text{Spectrum}(B) \cup \{0\}.$$

# Characterizations of the sharp partial ordering

## First characterization

Given  $A, B \in \mathbb{R}^{n \times n}$ , with  $\text{ind}(A) \leq 1$  and  $\text{ind}(B) \leq 1$ , the following statements are equivalent:

- $A \overset{\#}{\leq} B$
- $A^2 = BA = AB$
- There exist nonsingular matrices  $P$ ,  $C_1$  and  $C_2$  such that

$$A = P \begin{bmatrix} C_1 & & \\ & O & \\ & & O \end{bmatrix} P^{-1} \quad \text{and} \quad B = P \begin{bmatrix} C_1 & & \\ & C_2 & \\ & & O \end{bmatrix} P^{-1}$$

Core-nilpotent decomposition

# Characterizations of the sharp partial ordering

## Second characterization

Given  $A, B \in \mathbb{R}^{n \times n}$ , with  $\text{ind}(A) \leq 1$  and  $\text{ind}(B) \leq 1$ , the following statements are equivalent:

- $A \overset{\#}{\leq} B$
- There exist an **idempotent** matrix  $Q$  such that

$$A = QB = BQ$$

## Characterizations of the sharp partial ordering

### Third characterization

Given  $A, B \in \mathbb{R}^{n \times n}$ , with  $\text{ind}(A) \leq 1$  and  $\text{ind}(B) \leq 1$ , the following statements are equivalent:

- $A \overset{\#}{\leq} B$
- There exist an **idempotent** matrix  $T$  such that

$$B = U \begin{bmatrix} \Sigma K & \Sigma L \\ 0 & 0 \end{bmatrix} U^T \quad \text{and} \quad A = U \begin{bmatrix} T \Sigma K & T \Sigma L \\ 0 & 0 \end{bmatrix} U^T$$

where

- $U$  is an **orthogonal** matrix,
  - $\Sigma$  is a **diagonal definite positive** matrix,
  - $K$  and  $L$  matrices such that  $KK^T + LL^T = I$ ,
  - $T \Sigma K = \Sigma K T$ .
- } Hartwig  
Spindelböck  
decomposition



# Sharp partial ordered autonomous systems

## Definition

Consider the autonomous systems

$$(1) \begin{cases} x(k+1) = A x(k) \\ x(0) = x_0 \end{cases} \quad \text{and} \quad (2) \begin{cases} \bar{x}(k+1) = B \bar{x}(k) \\ \bar{x}(0) = \bar{x}_0 \end{cases}$$

for  $k = 0, 1, \dots$  and  $A, B \in \mathbb{R}^{n \times n}$  having index 1.

The systems are **ordered under the sharp partial order** if  $A \stackrel{\#}{\leq} B$ .

- The system (1) is a **predecessor** of system (2) under the sharp partial order
- The system (2) is a **successor** of system (1) under the sharp partial order

## Solution of two ordered autonomous systems

First characterization:  $A \stackrel{\#}{\leq} B$

$$A = P \begin{bmatrix} C_1 & & \\ & O & \\ & & O \end{bmatrix} P^{-1} \quad \text{and} \quad B = A + P \underbrace{\begin{bmatrix} O & & \\ & C_2 & \\ & & O \end{bmatrix}}_{\Gamma} P^{-1}$$

### Theorem

Let  $A, B \in \mathbb{R}^{n \times n}$  be the state matrices of two sharp partial ordered autonomous systems.

The solutions of both systems are related by

$$\bar{x}(k) = x(k) + \Gamma^k x_0$$

provided that  $\bar{x}(0) = x(0) = x_0$ .

# Solution of two ordered autonomous systems

## Difference between both solutions

Let  $A, B \in \mathbb{R}^{n \times n}$  be the state matrices of two sharp partial ordered autonomous systems. Then,

$$\|\bar{x}(k) - x(k)\| \leq \|C_2\|_F^k \|x_0\|$$

where  $\|\cdot\|$  is compatible with the  $P$ -matrix norm.

## Remarks

- System (2) can be seen as a perturbation of System (1):  $B = A + \Gamma$ .
- If the perturbation  $C_2$  is “small” then  $\bar{x}(k)$  is close to  $x(k)$ .
- Since  $\sigma(B) = \sigma(A) \cup \sigma(C_2)$ , the stability of system (2) depends on stability of system (1) and on the perturbation.

## Solution of two ordered autonomous systems

### Algorithm

**Inputs:** The matrix  $A$  of index at most 1, the initial condition  $x_0$ , and the nonzero perturbation numbers  $\varepsilon_1, \dots, \varepsilon_\ell$ .

**Outputs:** The matrix  $B$  and the solutions  $x(k)$  and  $\bar{x}(k)$ .

- 1 Compute the core-nilpotent decomposition of  $A$ :

$$A = P \begin{bmatrix} C_1 & & \\ & O & \\ & & O \end{bmatrix} P^{-1}.$$

- 2 Select  $C_2 = \text{diag}(\varepsilon_1, \dots, \varepsilon_\ell)$ .

- 3 Construct  $\Gamma = P \begin{bmatrix} O & & \\ & C_2 & \\ & & O \end{bmatrix} P^{-1}$  and  $B = A + \Gamma$ .

- 4 The solutions are:  $x(k) = A^k x_0$  and  $\bar{x}(k) = x(k) + \Gamma^k x_0$ .

# Solution of two ordered autonomous systems

Example:  $x(k+1) = A x(k)$  and  $\bar{x}(k+1) = B \bar{x}(k)$

Let

$$A = \begin{bmatrix} 0.1739 & -0.0607 & -0.0662 & 0.0763 & 0.0596 & -0.0400 \\ 0.4338 & -0.0725 & -0.1948 & 0.2847 & 0.1164 & -0.0720 \\ -0.8839 & 0.3844 & 0.3080 & -0.2966 & -0.3343 & 0.2302 \\ -0.0239 & 0.1583 & -0.0471 & 0.1691 & -0.0697 & 0.0584 \\ 0.1986 & -0.1166 & -0.0579 & 0.0304 & 0.0875 & -0.0624 \\ -0.5959 & 0.2936 & 0.1947 & -0.1587 & -0.2395 & 0.1674 \end{bmatrix} =$$

$$= P \left[ \begin{array}{cc|cc} 0.5000 & -0.3333 & & \\ & 0 & 0.3333 & \\ \hline & & & 0 \\ & & & 0 \end{array} \right] P^{-1}$$

and

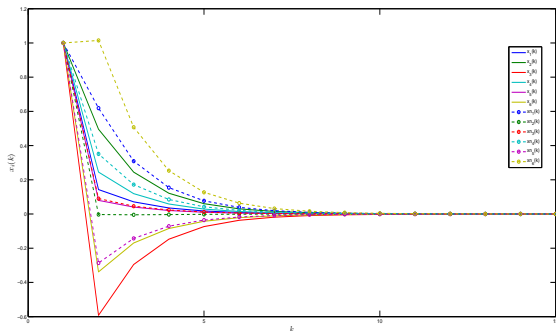
$$B = P \left[ \begin{array}{cc|cc|c} 0.5000 & -0.3333 & & & \\ & 0 & 0.3333 & & \\ \hline & & & 0.5000 & 0 \\ & & & 0 & 0.5000 \\ \hline & & & & 0 \end{array} \right] P^{-1}$$

## Solution of two ordered autonomous systems

Then

$$\|\bar{x}(k) - x(k)\| \leq \|C_2\|_F^k \|x_0\| = 0.7071 \|x_0\|.$$

Figure: Evolution of the 15th first iterations of  $x_i(k)$  and  $\bar{x}_i(k)$



## Solution of two ordered autonomous systems

Second characterization:  $A \stackrel{\#}{\leq} B$

$$A = QB$$

where  $Q$  is an idempotent matrix.

### Theorem

Let  $A, B \in \mathbb{R}^{n \times n}$  be the state matrices of two sharp partial ordered autonomous systems.

The solutions of both systems are related by

$$x(k) = Q\bar{x}(k)$$

provided that  $\bar{x}(0) = x(0) = x_0$ .

## Solution of two ordered autonomous systems

### Remarks

- Solution of System (1) is a **projection** of the solution of System (2):  
 $x(k) = Q\bar{x}(k)$ .
- All the idempotent matrices  $Q_Z$  satisfying  $A = Q_Z B = B Q_Z$  are

$$Q_Z = P \begin{bmatrix} I & & \\ & O & \\ & & Z \end{bmatrix} P^{-1}$$

In particular  $Z = O \Rightarrow Q_Z = AA^\# = A^\#A$ .

- Both solutions are as close as the magnitude of the matrix  $Q - I$ :

$$\|x(k) - \bar{x}(k)\| \leq \|Q - I\|_F \|x_0\|$$

where  $\|\cdot\|_F$  is the Frobenius norm.



## Solution of two ordered autonomous systems

Third characterization:  $A \stackrel{\#}{\leq} B$

$$B = U \begin{bmatrix} \Sigma K & \Sigma L \\ 0 & 0 \end{bmatrix} U^T \quad \text{and} \quad A = \underbrace{U \begin{bmatrix} T & 0 \\ 0 & I \end{bmatrix} U^T}_{\tilde{\Gamma}} B$$

### Theorem

Let  $A, B \in \mathbb{R}^{n \times n}$  be the state matrices of two **sharp partial ordered autonomous systems**.

The **solutions** of both systems are related by

$$x(k) = \tilde{\Gamma} \bar{x}(k)$$

provided that  $\bar{x}(0) = x(0) = x_0$ .

## Solution of two ordered autonomous systems

### Difference between both solutions

Let  $A, B \in \mathbb{R}^{n \times n}$  be the state matrices of two sharp partial ordered autonomous systems. Then,

$$\|x(k) - \bar{x}(k)\| \leq \|T - I\|_F \|\Sigma\|_F^k \|x_0\|$$

### Remarks

- Solution of System (1) is a **projection** of the solution of System (2):  
 $x(k) = \tilde{\Gamma} \bar{x}(k)$ .
- Both solutions are as close as the magnitude of the **singular value matrix**  $\Sigma$ .
- Since  $\sigma(T\Sigma K) \subseteq \sigma(\Sigma K)$ , the stability of System (2) implies the stability of System (1).

## Solution of two ordered autonomous systems

### Algorithm

*Inputs:* The matrix  $B$  of index at most 1 and the initial condition  $x_0$ .

*Outputs:* The matrix  $A$  and the solutions  $\bar{x}(k)$  and  $x(k)$ .

- 1 Compute the SVD of  $B$ :  $B = USV^T$  and  $r = \text{rank}(B)$ .
- 2 Assign to  $\Sigma$  the first  $r$  rows and the first  $r$  columns of  $S$ .
- 3 Compute  $M = SV^T U$ .
- 4 Assign to  $\tilde{M}$  the first  $r$  rows and the first  $r$  columns of  $M$ .
- 5 Compute  $R = \Sigma^{-1} \tilde{M}$ .
- 6 Assign to  $K$  the first  $r$  rows and the first  $r$  columns of  $R$ .
- 7 Assign to  $L$  the first  $r$  rows and the last  $n - r$  columns of  $R$ .

# Solution of two ordered autonomous systems

## Algorithm

*Until here we have constructed the Hartwig-Spindelböck decomposition of  $B$ :*

$$B = U \begin{bmatrix} \Sigma K & \Sigma L \\ O & O \end{bmatrix} U^T$$

- 8 Find a matrix  $T$  such that  $\Sigma K T = T \Sigma K$  and  $T^2 = T$ .
- 9 Construct

$$A = U \begin{bmatrix} T \Sigma K & T \Sigma L \\ O & O \end{bmatrix} U^T \quad \text{and} \quad \tilde{\Gamma} = U \begin{bmatrix} T & O \\ O & I \end{bmatrix} U^T.$$

- 10 The solutions are:  $\bar{x}(k) = B^k x_0$  and  $x(k) = \tilde{\Gamma} \bar{x}(k)$ .

## Solution of two ordered autonomous systems

Example:  $\bar{x}(k+1) = B \bar{x}(k)$  and  $x(k+1) = A x(k)$

Let

$$B = \begin{bmatrix} 0.3748 & -0.0548 & -0.1024 & 0.0447 & 0.2241 & 0.1326 \\ -0.2717 & 0.4732 & -0.2945 & -0.1694 & 0.0093 & 0.2499 \\ -0.3656 & 0.1359 & 0.3203 & -0.1150 & -0.1290 & 0.2435 \\ 0.2810 & -0.1305 & 0.0172 & 0.4182 & -0.0674 & -0.1675 \\ -0.2640 & 0.2224 & -0.1157 & -0.2485 & 0.0016 & 0.1178 \\ 0.3813 & -0.1425 & 0.2049 & 0.1514 & 0.1744 & 0.2454 \end{bmatrix}$$

Then

$$\Sigma = \begin{bmatrix} 1.0819 & 0 & 0 & 0 \\ 0 & 0.5784 & 0 & 0 \\ 0 & 0 & 0.5256 & 0 \\ 0 & 0 & 0 & 0.3085 \end{bmatrix}.$$

Choose

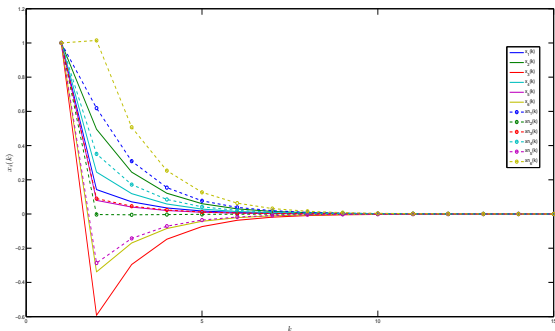
$$T = \begin{bmatrix} -0.1860 & -0.1428 & 0.0821 & 0.2657 \\ 0.6177 & 0.5894 & -0.3608 & -0.0162 \\ -1.0127 & -0.9133 & 0.5509 & 0.4261 \\ -0.1855 & -0.0384 & 0.0021 & 1.0455 \end{bmatrix}.$$

## Solution of two ordered autonomous systems

Then

$$\|x(k) - \bar{x}(k)\| \leq \|T - I\|_F^k \|\Sigma\|_F^k \|x_0\|.$$

Figure: Evolution of the 15th first iterations of  $x_i(k)$  and  $\bar{x}_i(k)$



# Conclusions

- We have introduced the concept of **sharp partial ordered autonomous systems**.
- The **successor system** can be seen as a **perturbation** of its predecessor. The difference between their solutions is given by the magnitude of the **perturbation matrix  $C_2$** .
- The **solution of the predecessor system** can be obtained as a **projection** of the solution of its successor system and, in this case, the difference between the solutions is given by the magnitude of the **singular value matrix  $\Sigma$** .

## Future work

- Define **ordered autonomous systems** for other matrix partial orders like minus, cn, star ...
- Extend the concept of **sharp partial ordered autonomous systems** to linear control systems.
- Study linear control systems for different orders.



THANK YOU VERY MUCH FOR  
YOUR ATTENTION

## Solution of $T\Sigma K = \Sigma K T$ and $T^2 = T$

### Proposition

Let  $\Sigma$  and  $K$  be the matrices of the Hartwig-Spindelböck decomposition of  $B$ .

There exists a **nontrivial idempotent matrix**  $T$  such that  $\Sigma K T = T \Sigma K$



There exists a **nonsingular matrix**  $S$  such that  $\Sigma K = S \begin{bmatrix} S_1 & O \\ O & S_2 \end{bmatrix} S^{-1}$

### Remarks

- The matrix  $T$  can always be constructed since the block partition of  $\Sigma K$  can always be done using its Jordan canonical form.
- If  $\Sigma K$  is diagonalizable, we can construct several matrices  $T$  with different rank by choosing adequate blocks in  $\Sigma K$ .
- Alternatively we can select  $T$  using the Schur decomposition of  $\Sigma K$ .