

# Stability of difference equations and applications to transport and wave propagation on networks

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joint work with Yacine Chitour and Mario Sigalotti

Stability of non-conservative systems

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# Outline

- 1 Introduction
- 2 Stability analysis
- 3 Application to a transport system
- 4 Relative controllability

# Introduction

## Linear difference equations

$$\Sigma(\Lambda, A) : \quad x(t) = \sum_{j=1}^N A_j(t)x(t - \Lambda_j), \quad t \geq 0.$$

- $\Lambda_1, \dots, \Lambda_N$ : positive delays.
- $A_1(t), \dots, A_N(t)$ : time-dependent  $d \times d$  matrices.
- $x(t) \in \mathbb{C}^d$ .
- Notation:  $\Lambda_{\min} = \min_i \Lambda_i$ ,  $\Lambda_{\max} = \max_i \Lambda_i$ .

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Motivation:

- Applications to some hyperbolic PDEs.
- Generalization of previous results:  $N = 1$ , autonomous.

# Introduction

Motivation: transport systems

Hyperbolic PDEs  $\rightarrow$  difference equations: [Cooke, Krumme; 1968], [Slemrod; 1971], [Greenberg, Li; 1984], [Coron, Bastin, d'Andréa Novel; 2008], [Fridman, Mondié, Saldivar; 2010], [Gugat, Sigalotti; 2010]...

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$$\left\{ \begin{array}{l} \partial_t u_i(t, \xi) + \partial_\xi u_i(t, \xi) + \alpha_i(t, \xi) u_i(t, \xi) = 0, \\ \qquad \qquad \qquad t \in \mathbb{R}_+, \xi \in [0, \Lambda_i], i \in \llbracket 1, N \rrbracket, \\ u_i(t, 0) = \sum_{j=1}^N m_{ij}(t) u_j(t, \Lambda_j), \quad t \in \mathbb{R}_+, i \in \llbracket 1, N \rrbracket. \end{array} \right.$$

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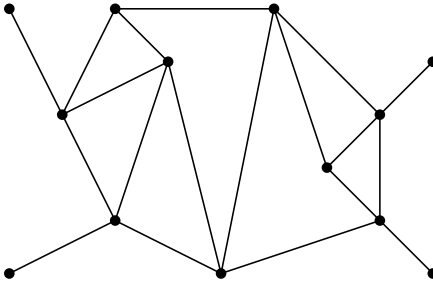
Method of characteristics: for  $t \geq \Lambda_{\max}$ ,

$$u_i(t, 0) = \sum_{j=1}^N m_{ij}(t) u_j(t, \Lambda_j) = \sum_{j=1}^N m_{ij}(t) e^{-\int_0^{\Lambda_j} \alpha_j(t-s, \Lambda_j-s) ds} u_j(t - \Lambda_j, 0).$$

Set  $x(t) = (u_i(t, 0))_{i \in \llbracket 1, N \rrbracket}$ . Then  $x$  satisfies a difference equation.

# Introduction

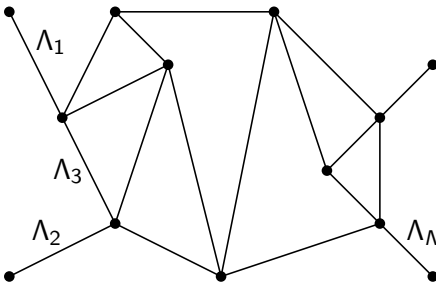
Motivation: wave propagation on networks





# Introduction

Motivation: wave propagation on networks



Edges:  $\mathcal{E}$

Vertices:  $\mathcal{V}$

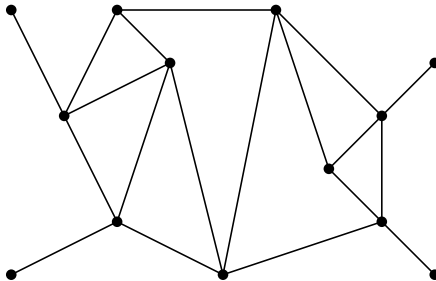
$$\partial_{tt}^2 u_i(t, \xi) = \partial_{\xi\xi}^2 u_i(t, \xi)$$

$$u_i(t, q) = u_j(t, q), \quad \forall q \in \mathcal{V}, \forall i, j \in \mathcal{E}_q$$

+ conditions on vertices.

# Introduction

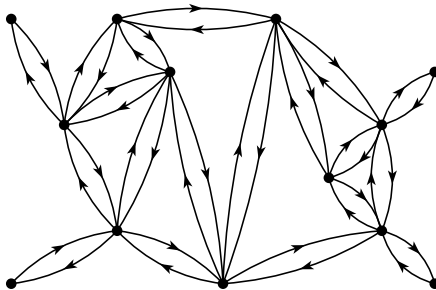
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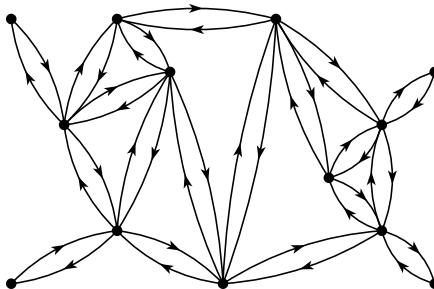
D'Alembert decomposition on travelling waves:



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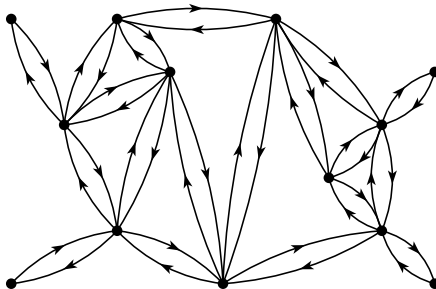


System of  $2N$  transport equations.

# Introduction

Motivation: wave propagation on networks

D'Alembert decomposition on travelling waves:



System of  $2N$  transport equations.

Can be reduced to a system of difference equations.

# Introduction

Motivation: case  $N = 1$

- When  $N = 1$ :  $x(t) = A(t)x(t - \Lambda)$ .
- Can be reduced to  $x_n = A_n x_{n-1}$ .

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## Autonomous system

$$x_n = Ax_{n-1}$$
$$A \in \mathcal{M}_d(\mathbb{C})$$

Exponential stability

$$\iff \rho(A) < 1$$

Finite-time stability

$$\iff \rho(A) = 0$$

$$\rho(A) = \lim_{n \rightarrow +\infty} |A^n|^{\frac{1}{n}}$$
$$= \max_{\lambda \in \sigma(A)} |\lambda|$$

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**Autonomous system**

$$x_n = Ax_{n-1}$$

$$A \in \mathcal{M}_d(\mathbb{C})$$

**Arbitrary switching**

$$x_n = A_n x_{n-1}$$

$$A_n \in \mathfrak{B} \subset \mathcal{M}_d(\mathbb{C})$$

Exponential stability

$$\iff \rho(A) < 1$$

Finite-time stability

$$\iff \rho(A) = 0$$

Uniform exponential stability

$$\iff \rho_J(\mathfrak{B}) < 1$$

Finite-time stability

$$\iff \rho_J(\mathfrak{B}) = 0$$

$$\rho(A) = \lim_{n \rightarrow +\infty} |A^n|^{\frac{1}{n}}$$

$$= \max_{\lambda \in \sigma(A)} |\lambda|$$

$$\rho_J(\mathfrak{B}) = \lim_{n \rightarrow +\infty} \sup_{A_1, \dots, A_n \in \mathfrak{B}} |A_1 A_2 \cdots A_n|^{\frac{1}{n}}$$

(cf. [Jungers; 2009])



# Introduction

Motivation: autonomous case

$$\Sigma^{\text{aut}}(\Lambda, A) : \quad x(t) = \sum_{j=1}^N A_j x(t - \Lambda_j), \quad t \geq 0$$

- [Cruz, Hale; 1970], [Henry; 1974], [Michiels et al.; 2009]...
- Studied through spectral methods.

- Stability: real parts of the roots of  $\det \left( \text{Id} - \sum_{j=1}^N A_j e^{-s\Lambda_j} \right) = 0$   
(*exponential polynomial*, see [Avellar, Hale; 1980]).

# Introduction

Motivation: autonomous case

$$\text{Let } \rho_{\text{HS}}(A) = \max_{(\theta_1, \dots, \theta_N) \in [0, 2\pi]^N} \rho \left( \sum_{j=1}^N A_j e^{i\theta_j} \right).$$

Theorem ([Hale; 1975], [Silkowski; 1976])

*The following are equivalent:*

- $\rho_{\text{HS}}(A) < 1$ ;
- $\Sigma^{\text{aut}}(\Lambda, A)$  is exponentially stable for some  $\Lambda \in (0, +\infty)^N$  with rationally independent components;
- $\Sigma^{\text{aut}}(\Lambda, A)$  is exponentially stable for every  $\Lambda \in (0, +\infty)^N$ .

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- 
- Still true if we replace  $\rho_{\text{HS}}(A) < 1$  by  $\rho_{\text{HS}}(A) = 0$  and exponential by finite-time stability.
  - For rationally dependent delays: [Michiels et al.; 2009].
  - Can this be generalized to the non-autonomous case?

# Introduction

## Main problem

Main problem: **exponential stability** of the non-autonomous system  $\Sigma(\Lambda, A)$  **uniformly** with respect to a given class  $\mathcal{A}$  of functions  $A : \mathbb{R} \rightarrow \mathcal{M}_d(\mathbb{C})^N$ .

- The techniques from the autonomous case cannot be applied.
- Our approach: **explicit formula** for solutions of  $\Sigma(\Lambda, A)$ .
- When  $\mathcal{A} = L^\infty(\mathbb{R}, \mathfrak{B})$ , we obtain a generalization of Hale–Silkowsky's Theorem.

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Exponential stability criteria:

	Autonomous	Arbitrary switching
$N = 1$	$\rho(A) < 1$	$\rho_J(\mathfrak{B}) < 1$
any $N$	$\rho_{\text{HS}}(A) < 1$	

# Stability analysis

## Explicit solution (I)

$$\Sigma(\Lambda, A) : \quad x(t) = \sum_{j=1}^N A_j(t)x(t - \Lambda_j), \quad t \geq 0$$

**Solution** with initial condition  $x_0 : [-\Lambda_{\max}, 0) \rightarrow \mathbb{C}^d$ :  $x$  satisfying  $\Sigma(\Lambda, A)$  for  $t \geq 0$  and  $x(t) = x_0(t)$  for  $-\Lambda_{\max} \leq t < 0$ .

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### Lemma

The solution  $x : [-\Lambda_{\max}, +\infty) \rightarrow \mathbb{C}^d$  of  $\Sigma(\Lambda, A)$  with initial condition  $x_0 : [-\Lambda_{\max}, 0) \rightarrow \mathbb{C}^d$  is, for  $t \geq 0$ ,

$$x(t) = \sum_{\substack{\mathbf{n} \in \mathbb{N}^N \\ t < \Lambda \cdot \mathbf{n} \leq t + \Lambda_{\max}}} \sum_{\substack{j \in \llbracket 1, N \rrbracket \\ \Lambda \cdot \mathbf{n} - \Lambda_j \leq t}} \Xi_{\mathbf{n} - e_j, t}^{\Lambda, A} A_j(t - \Lambda \cdot \mathbf{n} + \Lambda_j) x_0(t - \Lambda \cdot \mathbf{n}),$$

where the matrices  $\Xi_{\mathbf{n}, t}^{\Lambda, A}$  are defined recursively by

$$\Xi_{\mathbf{n}, t}^{\Lambda, A} = \sum_{\substack{k=1 \\ n_k \geq 1}}^N A_k(t) \Xi_{\mathbf{n} - e_k, t - \Lambda_k}^{\Lambda, A}, \quad \Xi_{0, t}^{\Lambda, A} = \text{Id}_d.$$

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$$x(t) = \sum_{\substack{n \in \mathbb{N}^N \\ t < \Lambda \cdot n \leq t + \Lambda_{\max}}} \Theta_{n,t}^{\Lambda, A} x_0(t - \Lambda \cdot n),$$

where the matrices  $\Xi_{n,t}^{\Lambda, A}$  are defined recursively by

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## Stability analysis (I)

$$\Sigma(\Lambda, A) : \quad x(t) = \sum_{j=1}^N A_j(t)x(t - \Lambda_j), \quad t \geq 0$$

- $X_p = L^p([-\Lambda_{\max}, 0], \mathbb{C}^d)$ ,  $p \in [1, +\infty]$
- $\mathcal{A}$ : set of **uniformly locally bounded** functions taking values in  $N$ -tuples of matrices
- $\Sigma(\Lambda, \mathcal{A})$ : family of systems  $\Sigma(\Lambda, A)$  for  $A \in \mathcal{A}$ .
- For  $x$  solution of  $\Sigma(\Lambda, A)$ ,  $x_t = x(t + \cdot)|_{[-\Lambda_{\max}, 0]} \in X_p$ .

# Stability analysis

## Stability analysis (I)

### Definition

$\Sigma(\Lambda, \mathcal{A})$  is of:

- **exponential type  $\gamma$**  in  $X_p$  if  $\forall \varepsilon > 0 \exists K > 0$  s.t.  $\forall A \in \mathcal{A}$ ,  $\forall x_0 \in X_p$ , the solution  $x$  satisfies  $\|x_t\|_{X_p} \leq Ke^{(\gamma+\varepsilon)t} \|x_0\|_{X_p}$ ;
- **$\Theta$ -exponential type  $\gamma$**  if  $\forall \varepsilon > 0 \exists K > 0$  s.t.  $\forall A \in \mathcal{A}$ ,  $\forall \mathbf{n} \in \mathbb{N}^N$ , a.e.  $t \in (\Lambda \cdot \mathbf{n} - \Lambda_{\max}, \Lambda \cdot \mathbf{n})$ , one has  $|\Theta_{\mathbf{n},t}^{\Lambda,A}| \leq Ke^{(\gamma+\varepsilon)t}$ ;
- **$\Xi$ -exponential type  $\gamma$**  if  $\forall \varepsilon > 0 \exists K > 0$  s.t.  $\forall A \in \mathcal{A}$ ,  $\forall \mathbf{n} \in \mathbb{N}^N$ , a.e.  $t \in \mathbb{R}$ , one has  $|\Xi_{\mathbf{n},t}^{\Lambda,A}| \leq Ke^{(\gamma+\varepsilon)\Lambda \cdot \mathbf{n}}$ .

**Exponential stability:** exponential type  $\gamma < 0$ .

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### Theorem (Chitour, M., Sigalotti; 2015)

Let  $\Lambda \in (0, +\infty)^N$  and  $\mathcal{A}$  be uniformly locally bounded.

- If  $\Sigma(\Lambda, \mathcal{A})$  is of  $\Theta$ -exponential type  $\gamma$  then  $\forall p \in [1, +\infty]$  it is of exponential type  $\gamma$  in  $X_p$ .

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- Suppose that  $\Lambda_1, \dots, \Lambda_N$  are **rationally independent**. If  $\exists p \in [1, +\infty]$  such that  $\Sigma(\Lambda, \mathcal{A})$  is of exponential type  $\gamma$  in  $X_p$ , then it is of  $\Theta$ -exponential type  $\gamma$ .

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- Suppose that  $\Lambda_1, \dots, \Lambda_N$  are **rationally independent**. If  $\exists p \in [1, +\infty]$  such that  $\Sigma(\Lambda, \mathcal{A})$  is of exponential type  $\gamma$  in  $X_p$ , then it is of  $\Theta$ -exponential type  $\gamma$ .
- Suppose that  $\mathcal{A}$  is **shift-invariant**. Then  $\Theta$ - and  $\Xi$ -exponential types  $\gamma$  are equivalent.



# Stability analysis

## Rational dependence of the delays

Let  $\Lambda = (\Lambda_1, \dots, \Lambda_N) \in (0, +\infty)^N$ . We define

$$Z(\Lambda) = \{\mathbf{n} \in \mathbb{Z}^N \mid \Lambda \cdot \mathbf{n} = 0\},$$

$$V(\Lambda) = \{L \in \mathbb{R}^N \mid Z(\Lambda) \subset Z(L)\}, \quad (\text{more rationally dependent})$$

$$W(\Lambda) = \{L \in \mathbb{R}^N \mid Z(\Lambda) = Z(L)\}, \quad (\text{as rationally dependent})$$

$$V_+(\Lambda) = V(\Lambda) \cap (0, +\infty)^N, \quad W_+(\Lambda) = W(\Lambda) \cap (0, +\infty)^N.$$

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Example:  $\Lambda = (1, \sqrt{2}, 1 + \sqrt{2})$ .

- $Z(\Lambda) = \{(n, m, -n - m) \mid n, m \in \mathbb{Z}\}$ ;
- $V(\Lambda) = \{(a, b, a + b) \mid a, b \in \mathbb{R}\}$ ;
- $W(\Lambda) = \{(a, b, a + b) \mid a, b \in \mathbb{R} \text{ rationally independent}\}$ .

# Stability analysis

## Rational dependence of the delays

For  $\Lambda \in (0, +\infty)^N$ , define the following equivalence relations on  $\llbracket 1, N \rrbracket$  and  $\mathbb{Z}^N$ ,

$$i \sim j \text{ iff } \Lambda_i = \Lambda_j, \quad \mathbf{n} \approx \mathbf{n}' \text{ iff } \Lambda \cdot \mathbf{n} = \Lambda \cdot \mathbf{n}',$$
$$\mathcal{J} = \llbracket 1, N \rrbracket / \sim, \quad \mathcal{Z} = \mathbb{Z}^N / \approx.$$

# Stability analysis

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For  $A : \mathbb{R} \rightarrow \mathcal{M}_d(\mathbb{C})^N$ ,  $L \in V_+(\Lambda)$ ,  $[\mathbf{n}] \in \mathcal{Z}$ ,  $[i] \in \mathcal{J}$ , and  $t \in \mathbb{R}$ ,

$$\hat{\Xi}_{[\mathbf{n}], t}^{L, \Lambda, A} = \sum_{\mathbf{n}' \in [\mathbf{n}]} \Xi_{\mathbf{n}', t}^{L, A}, \quad \hat{A}_{[i]}^\Lambda(t) = \sum_{j \in [i]} A_j(t),$$

$$\hat{\Theta}_{[\mathbf{n}], t}^{L, \Lambda, A} = \sum_{\substack{[j] \in \mathcal{J} \\ L \cdot \mathbf{n} - L_j \leq t}} \hat{\Xi}_{[\mathbf{n} - \mathbf{e}_j], t}^{L, \Lambda, A} \hat{A}_{[j]}^\Lambda(t - L \cdot \mathbf{n} + L_j).$$

# Stability analysis

## Explicit solution (II)

### Lemma (Chitour, M., Sigalotti; 2015)

Let  $\Lambda \in (0, +\infty)^N$ ,  $L \in V_+(\Lambda)$ ,  $A : \mathbb{R} \rightarrow \mathcal{M}_d(\mathbb{C})^N$ , and  $x_0 : [-L_{\max}, 0) \rightarrow \mathbb{C}^d$ . The corresponding solution  $x : [-L_{\max}, +\infty) \rightarrow \mathbb{C}^d$  of  $\Sigma(L, A)$  is, for  $t \geq 0$ ,

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# Stability analysis

## Stability analysis (II)

We can define exponential types for  $\hat{\Theta}$  and  $\hat{\Xi}$  similarly. Since they depend on  $\Lambda$ , we write  $(\hat{\Theta}, \Lambda)$ - and  $(\hat{\Xi}, \Lambda)$ -*exponential types*.

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Let  $\Lambda \in (0, +\infty)^N$  and  $\mathcal{A}$  be uniformly locally bounded.

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- Let  $L \in W_+(\Lambda)$ . If  $\exists p \in [1, +\infty]$  such that  $\Sigma(L, \mathcal{A})$  is of exponential type  $\gamma$  in  $X_p$ , then it is of  $(\hat{\Theta}, \Lambda)$ -exponential type  $\gamma$ .

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- Suppose that  $\mathcal{A}$  is *shift-invariant*. Then  $(\hat{\Theta}, \Lambda)$ - and  $(\hat{\Xi}, \Lambda)$ -exponential types  $\gamma$  are equivalent.

# Stability analysis

## Maximal Lyapunov exponent

### Definition

The *maximal Lyapunov exponent of  $\Sigma(L, \mathcal{A})$*  in  $X_p$  is

$$\lambda_p(L, \mathcal{A}) = \limsup_{t \rightarrow +\infty} \sup_{A \in \mathcal{A}} \sup_{\substack{x_0 \in X_p \\ \|x_0\|_{X_p} = 1}} \frac{\log \|x_t\|_{X_p}}{t}.$$

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### Proposition

$\lambda_p(L, \mathcal{A}) = \inf\{\gamma \in \mathbb{R} \mid \Sigma(L, \mathcal{A}) \text{ is of exponential type } \gamma \text{ in } X_p\}$ .

In particular,

$$\Sigma(L, \mathcal{A}) \text{ exponentially stable} \iff \lambda_p(L, \mathcal{A}) < 0.$$

By the previous results,  $\lambda_p(L, \mathcal{A})$  does not depend on  $p$ .

# Stability analysis

## Maximal Lyapunov exponent

### Theorem (Chitour, M., Sigalotti; 2015)

Let  $\Lambda \in (0, +\infty)^N$  and suppose that  $\mathcal{A}$  is shift-invariant. For every  $L \in W_+(\Lambda)$  and  $p \in [1, +\infty]$ ,

$$\lambda_p(L, \mathcal{A}) = \limsup_{|\mathbf{n}|_1 \rightarrow +\infty} \sup_{A \in \mathcal{A}} \operatorname{ess\,sup}_{t \in \mathbb{R}} \frac{\log \left| \hat{\Xi}_{\mathbf{n}, t}^{L, \Lambda, \mathcal{A}} \right|}{L \cdot \mathbf{n}}.$$

# Stability analysis

## Arbitrary switching

$$\Sigma(L, A) : \quad x(t) = \sum_{j=1}^N A_j(t)x(t - L_j), \quad t \geq 0.$$

- $\mathfrak{B} \subset \mathcal{M}_d(\mathbb{C})^N$ : bounded set of  $N$ -tuples of matrices.
- $\mathcal{A} = L^\infty(\mathbb{R}, \mathfrak{B})$ .
- $(A_1(t), \dots, A_N(t))$  is any measurable function taking values on  $\mathfrak{B}$ : **switched system with arbitrary switching signal**.
- In this case, one can obtain more precise results.



# Stability analysis

## Arbitrary switching

Using the recurrence relation for  $\Xi_{\mathbf{n},t}^{L,A}$ , we obtain:

$$\Xi_{[\mathbf{n}],t}^{L,\Lambda,A} = \sum_{\mathbf{n}' \in [\mathbf{n}] \cap \mathbb{N}^N} \sum_{v \in V_{\mathbf{n}'}} \prod_{k=1}^{|\mathbf{n}'|_1} A_{v_k} \left( t - \sum_{r=1}^{k-1} L_{v_r} \right).$$

$V_{\mathbf{n}}$ : set of all permutations of  $(\underbrace{1, \dots, 1}_{n_1 \text{ times}}, \underbrace{2, \dots, 2}_{n_2 \text{ times}}, \dots, \underbrace{N, \dots, N}_{n_N \text{ times}})$ .

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### Definition

$$\mu(\Lambda, \mathfrak{B}) = \limsup_{\xi \rightarrow +\infty} \sup_{\substack{B^r \in \mathfrak{B} \\ \xi \in \mathcal{L}(\Lambda) \text{ for } r \in \mathcal{L}_\xi(\Lambda)}} \left| \sum_{\substack{\mathbf{n} \in \mathbb{N}^N \\ \Lambda \cdot \mathbf{n} = \xi}} \sum_{v \in V_{\mathbf{n}}} \prod_{k=1}^{|\mathbf{n}|_1} B_{v_k}^{\Lambda_{v_1} + \dots + \Lambda_{v_{k-1}}} \right|^{\frac{1}{\xi}},$$

where  $\mathcal{L}(\Lambda) = \{\Lambda \cdot \mathbf{n} \mid \mathbf{n} \in \mathbb{N}^N\}$  and  $\mathcal{L}_\xi(\Lambda) = \mathcal{L}(\Lambda) \cap [0, \xi)$ .

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- for every  $L \in W_+(\Lambda)$ ,  $m_2 \lambda_p(\Lambda, \mathcal{A}) \leq \lambda_p(L, \mathcal{A}) \leq m_1 \lambda_p(\Lambda, \mathcal{A})$ .

Here,  $\{m_1, m_2\} = \left\{ \min_{j \in \llbracket 1, N \rrbracket} \frac{\Lambda_j}{L_j}, \max_{j \in \llbracket 1, N \rrbracket} \frac{\Lambda_j}{L_j} \right\}$ .

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### Corollary

The following statements are equivalent:

- $\mu(\Lambda, \mathfrak{B}) < 1$ ;
- $\Sigma(\Lambda, \mathcal{A})$  is exponentially stable in  $X_p$  for some  $p \in [1, +\infty]$ ;
- $\Sigma(L, \mathcal{A})$  is exponentially stable in  $X_p$  for every  $p \in [1, +\infty]$  and  $L \in V_+(\Lambda)$ .

# Stability analysis

## Conclusion

Exponential stability criteria:

	Autonomous	Arbitrary switching
$N = 1$	$\rho(A) < 1$	$\rho_{\mathfrak{B}} < 1$
any $N$	$\rho_{\text{HS}}(A) < 1$	$\mu(\Lambda, \mathfrak{B}) < 1$

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Exponential stability criteria:

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$N = 1$	$\rho(A) < 1$	$\rho_J(\mathfrak{B}) < 1$
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Interesting questions:

- Both  $\rho(A)$  and  $\rho_J(\mathfrak{B})$  are limits and  $\lim_{n \rightarrow +\infty}$  can be replaced by  $\inf_{n \in \mathbb{N}^*}$ . Is the same true for  $\mu(\Lambda, \mathfrak{B})$ ?

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- $\rho(A) = 0$ ,  $\rho_J(\mathfrak{B}) = 0$ , and  $\rho_{\text{HS}}(A) = 0$  are equivalent to convergence in finite time. Is this also true for  $\mu(\Lambda, \mathfrak{B})$ ?



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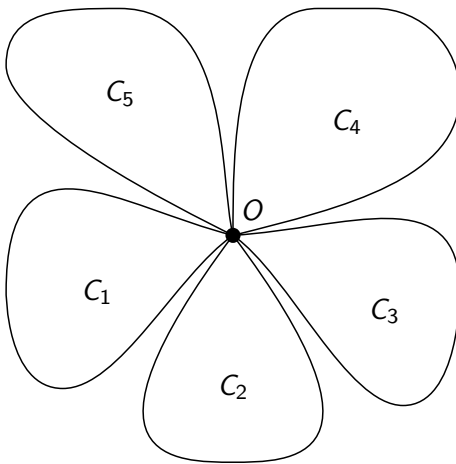
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- Can we numerically compute or approximate  $\mu$ ? (For  $\rho_J$ , this problem is NP-hard, Turing-undecidable, and non-algebraic, but several useful bounds and approximations exist, see [Jungers; 2009]).
- What can we say if  $\Lambda_1, \dots, \Lambda_N$  are time-dependent?

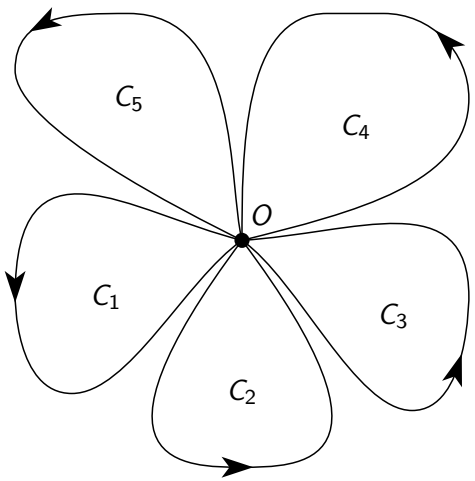
# Application to a transport system

## Transport system



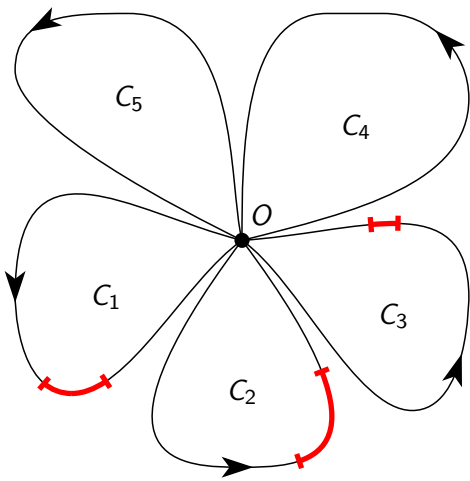
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$$\left\{ \begin{array}{l} \partial_t u_i(t, \xi) + \partial_\xi u_i(t, \xi) + \alpha_i(t) \chi_i(\xi) u_i(t, \xi) = 0, \\ \qquad \qquad \qquad t \in \mathbb{R}_+, \xi \in [0, L_i], i \in \llbracket 1, N_d \rrbracket, \\ \partial_t u_i(t, \xi) + \partial_\xi u_i(t, \xi) = 0, \quad t \in \mathbb{R}_+, \xi \in [0, L_i], i \in \llbracket N_d + 1, N \rrbracket, \\ u_i(t, 0) = \sum_{j=1}^N m_{ij} u_j(t, L_j), \quad t \in \mathbb{R}_+, i \in \llbracket 1, N \rrbracket. \end{array} \right.$$

- $\chi_i$ : characteristic function of an interval  $[a_i, b_i] \subset [0, L_i]$ .
- $M = (m_{ij})_{1 \leq i, j \leq N}$ : **transmission matrix**.
- $\alpha_i$  is **persistently exciting** for  $i \in \llbracket 1, N_d \rrbracket$ .

# Application to a transport system

## Persistence of excitation

- **Persistently exciting (PE) signals:** for  $T \geq \mu > 0$ , we say that  $\alpha \in \mathcal{G}(T, \mu)$  if  $\alpha \in L^\infty(\mathbb{R}; [0, 1])$  and

$$\forall t \in \mathbb{R}, \quad \int_t^{t+T} \alpha(s) ds \geq \mu.$$

- $\mathcal{G}(T, \mu)$  is shift-invariant.

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- $\mathcal{G}(T, \mu)$  is shift-invariant.
- Introduced in the context of identification and adaptive control [Anderson; 1977].
- Much studied in finite-dimensional control systems [Chitour, Sigalotti; 2010], [Chitour, M., Sigalotti; 2013].



# Application to a transport system

## Main result

Hypotheses:

- There exist  $i, j \in \llbracket 1, N \rrbracket$  such that  $\frac{L_i}{L_j} \notin \mathbb{Q}$ .
- $|M|_1 \leq 1$  and  $m_{ij} \neq 0$  for every  $i, j \in \llbracket 1, N \rrbracket$ .

### Theorem

$\forall T \geq \mu > 0, \exists C, \gamma > 0$  s.t.,  $\forall p \in [1, +\infty], \forall u_{i,0} \in L^p(0, L_i), i \in \llbracket 1, N \rrbracket$ , and  $\forall \alpha_k \in \mathcal{G}(T, \mu), k \in \llbracket 1, N_d \rrbracket$ , the corresponding solution satisfies

$$\sum_{i=1}^N \|u_i(t)\|_{L^p(0, L_i)} \leq C e^{-\gamma t} \sum_{i=1}^N \|u_{i,0}\|_{L^p(0, L_i)}, \quad \forall t \geq 0.$$

# Application to a transport system

## Technique of the proof

- For  $t \geq L_{\max}$ :

$$u_i(t, 0) = \sum_{j=1}^N m_{ij} u_j(t, L_j) = \sum_{j=1}^N m_{ij} e^{-\int_{t-L_j+a_j}^{t-L_j+b_j} \alpha_j(s) ds} u_j(t-L_j, 0)$$

- Set  $x(t) = (u_i(t, 0))_{i \in \llbracket 1, N \rrbracket}$ . Then  $x$  satisfies the difference equation

$$x(t) = \sum_{k=1}^N A_k(t) x(t - L_k)$$

with

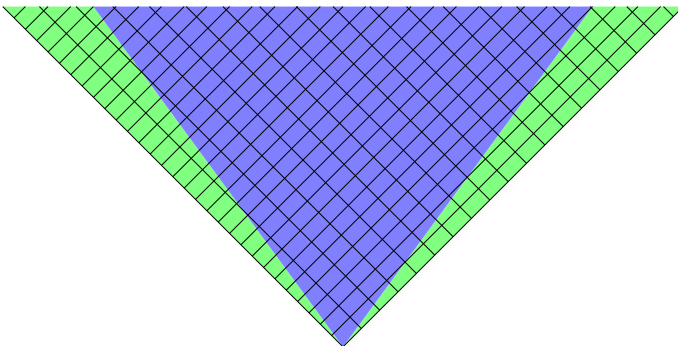
$$A_k(t) = \left( \delta_{jk} m_{ij} e^{-\int_{t-L_j+a_j}^{t-L_j+b_j} \alpha_j(s) ds} \right)_{i,j \in \llbracket 1, N \rrbracket}$$

- It suffices to show that such difference equation is  $(\hat{\Xi}, L)$ -exponentially stable. We study the behavior of the coefficients  $\Xi_{\mathbf{n}, t}^{L, A}$  as  $|\mathbf{n}|_1 \rightarrow +\infty$ .

# Application to a transport system

## Technique of the proof

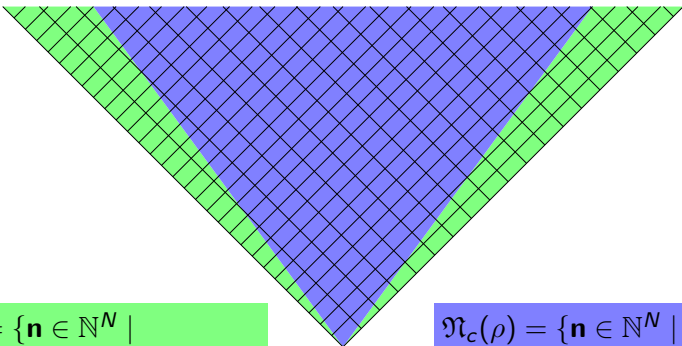
Decomposition of the set  $\mathbb{N}^N$ .



# Application to a transport system

## Technique of the proof

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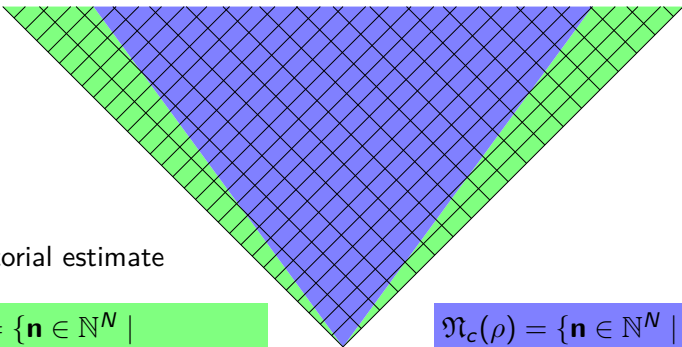
$$\mathfrak{N}_b(\rho) = \{ \mathbf{n} \in \mathbb{N}^N \mid \exists k \in \llbracket 1, N \rrbracket \text{ s.t. } n_k \leq \rho |\mathbf{n}|_1 \}$$

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combinatorial estimate

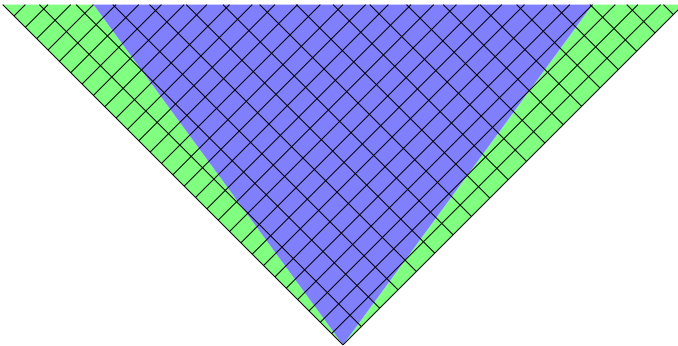
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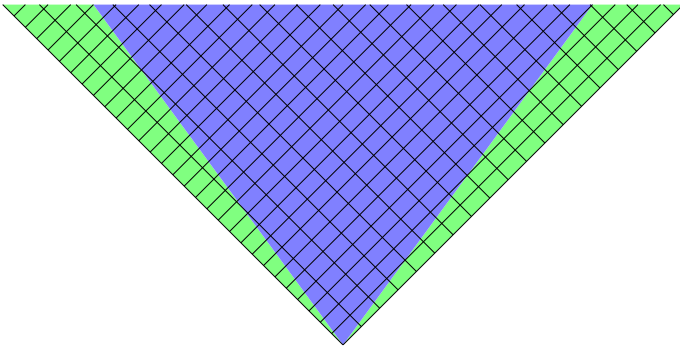
In  $\mathfrak{N}_c(\rho)$ : “box lemma”



# Application to a transport system

## Technique of the proof

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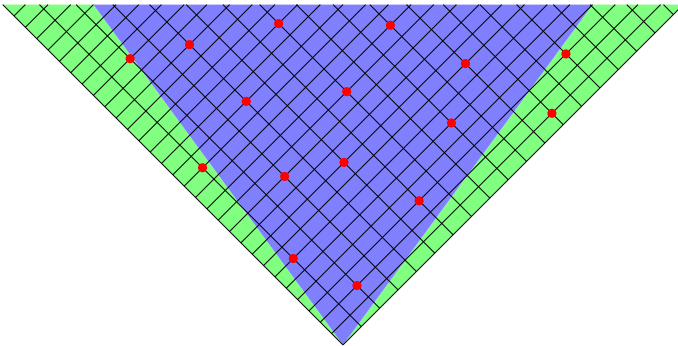


$$e^{-\int_{t-L\cdot n+a_k}^{t-L\cdot n+b_k} \alpha_k(s) ds}$$

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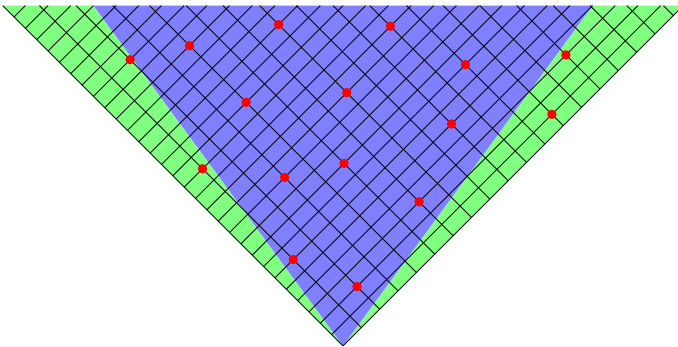
$$e^{-\int_{t-L\cdot n+a_k}^{t-L\cdot n+b_k} \alpha_k(s) ds} \leq \eta$$



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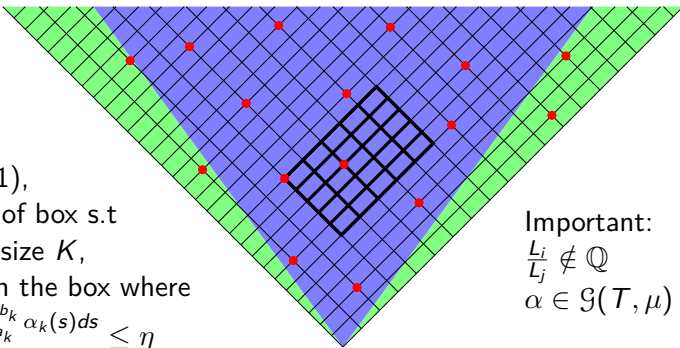


Find  $\eta \in (0, 1)$  such that  $e^{-\int_{t-L \cdot n + a_k}^{t-L \cdot n + b_k} \alpha_k(s) ds} \leq \eta$  “often enough”

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$\forall$  box of size  $K$ ,

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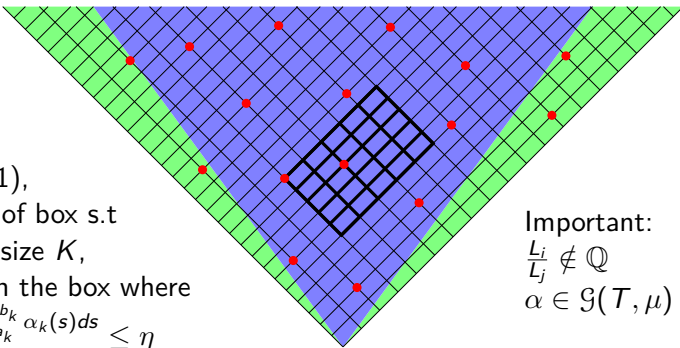
$$\frac{L_i}{L_j} \notin \mathbb{Q}$$

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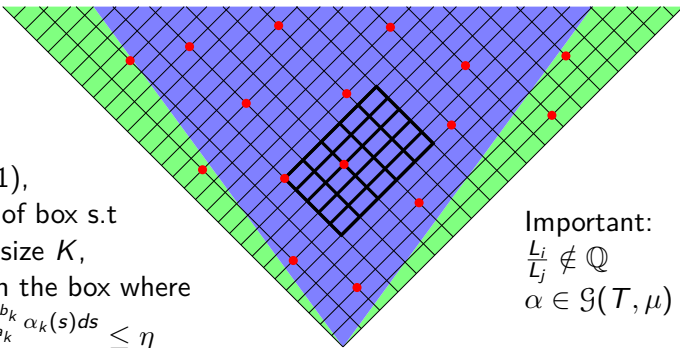
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$\implies$  the solutions converge exponentially ■

# Relative controllability

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We say that  $\Sigma_{\text{contr}}$  is **relatively controllable** in time  $T > 0$  if, for every  $x_0 : [-\Lambda_{\max}, 0) \rightarrow \mathbb{C}^d$  and  $x_1 \in \mathbb{C}^d$ , there exists  $u : [0, T] \rightarrow \mathbb{C}^m$  such that the unique solution  $x$  of  $\Sigma_{\text{contr}}$  with initial condition  $x_0$  and control  $u$  satisfies  $x(T) = x_1$ .

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## Explicit formula

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### Lemma (Explicit solution)

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$$x(t) = \sum_{\substack{[\mathbf{n}] \in \mathbb{Z} \\ \Lambda \cdot \mathbf{n} \leq t}} \hat{\Xi}_{\mathbf{n}, t}^{L, \Lambda, A} B u(t - \Lambda \cdot \mathbf{n}).$$

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By linearity, solution with initial condition  $x_0$  and control  $u$  is the sum of this formula with the previous one.

# Relative controllability

## Relative controllability criterion

### Theorem (M.; 2016)

*The following statements are equivalent:*

- $\Sigma_{\text{contr}}$  is relatively controllable in time  $T$ ;
- $\text{Span} \left\{ \hat{\Xi}_{[\mathbf{n}]}^{\Lambda, A} Bw \mid \mathbf{n} \in \mathbb{N}^N, \Lambda \cdot \mathbf{n} \leq T, w \in \mathbb{C}^m \right\} = \mathbb{C}^d$ ;

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- If  $\Lambda_1, \dots, \Lambda_N$  are rationally independent, then  $\Sigma_{\text{contr}}$  is relatively controllable in some time  $T > 0$  if and only if

$$\text{Span} \left\{ \Xi_{\mathbf{n}}^A B e_j \mid \mathbf{n} \in \mathbb{N}^N, |\mathbf{n}|_1 \leq d-1, j \in \llbracket 1, m \rrbracket \right\} = \mathbb{C}^d.$$

Introduction  
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Stability analysis  
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Application to a transport system  
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Relative controllability  
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