

# Scheduling using max-algebra

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# WHAT IS MAX-ALGEBRA

Definitions and basic properties

$$\begin{aligned}a \oplus b &= \max(a, b) \\ a \otimes b &= a + b \\ a, b &\in \overline{\mathbb{R}} := \mathbb{R} \cup \{-\infty\}\end{aligned}$$

Basic properties ( $\varepsilon = -\infty$ ,  $a^{-1} = -a$ ):

$$\begin{aligned}a \oplus b &= b \oplus a \\ (a \oplus b) \oplus c &= a \oplus (b \oplus c) \\ a \oplus \varepsilon &= a = \varepsilon \oplus a\end{aligned}$$

$$\begin{aligned}a \otimes b &= b \otimes a \\ (a \otimes b) \otimes c &= a \otimes (b \otimes c) \\ a \otimes \varepsilon &= \varepsilon = \varepsilon \otimes a \\ a \otimes 0 &= a = 0 \otimes a \\ a \otimes a^{-1} &= 0 = a^{-1} \otimes a\end{aligned}$$

$$\begin{aligned}(a \oplus b) \otimes c &= a \otimes c \oplus b \otimes c \\ a \oplus b &= a \text{ or } b\end{aligned}$$

# WHAT IS MAX-ALGEBRA

Definitions and basic properties

Extension to matrices and vectors:

$$\begin{aligned}A \oplus B &= (a_{ij} \oplus b_{ij}) \\A \otimes B &= \left( \sum_k^{\oplus} a_{ik} \otimes b_{kj} \right) \\ \alpha \otimes A &= (\alpha \otimes a_{ij})\end{aligned}$$

# WHAT IS MAX-ALGEBRA

## Definitions and basic properties

$$A \oplus B = B \oplus A$$

$$(A \oplus B) \oplus C = A \oplus (B \oplus C)$$

$$A \otimes \varepsilon = \varepsilon = \varepsilon \otimes A$$

$$[\text{not } A \otimes B = B \otimes A]$$

$$(A \otimes B) \otimes C = A \otimes (B \otimes C)$$

$$A \otimes I = A = I \otimes A$$

$$(A \oplus B) \otimes C = A \otimes C \oplus B \otimes C$$

$$A \otimes (B \oplus C) = A \otimes B \oplus A \otimes C$$

$A^{-1}$  exists  $\iff$   $A$  is a permutation matrix (= permuted diag)

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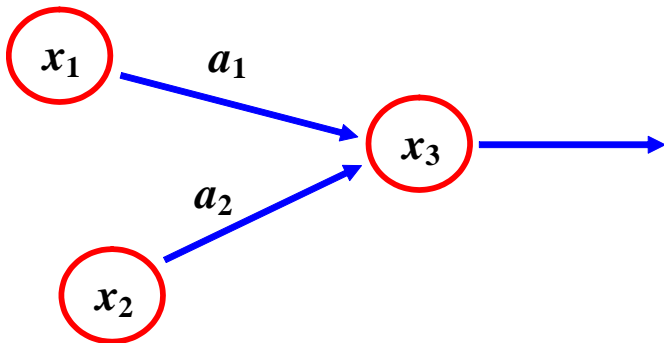
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- $a \oplus a = a$
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- $(A \oplus B)^k \neq A^k \oplus B^k$
- $(I \oplus A)^k = I \oplus A \oplus A^2 \oplus \dots \oplus A^k$

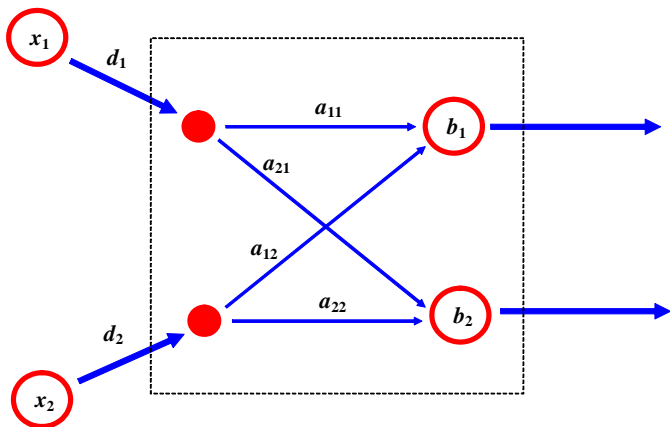


# SCHEDULING

## Example 1



$$\begin{aligned}x_3 &= \max(x_1 + a_1, x_2 + a_2) \\ &= a_1 \otimes x_1 \oplus a_2 \otimes x_2 \\ &= (a_1, a_2) \otimes \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = a^T \otimes x\end{aligned}$$



$$b_1 = \max(x_1 + d_1 + a_{11}, x_2 + d_2 + a_{12})$$

$$b_2 = \max(x_1 + d_1 + a_{21}, x_2 + d_2 + a_{22})$$

$$b = A \otimes x$$

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- If no solution then  $A \otimes x \leq b$  but as tight as possible

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- $\bar{x} = A^* \otimes' b$ , where  $A^* = -A^T$  and  $\otimes'$  is in min-algebra
- Then  $A \otimes \bar{x} \leq b$  and  $A \otimes \bar{x}$  is the best Chebyshev approximation of  $b$

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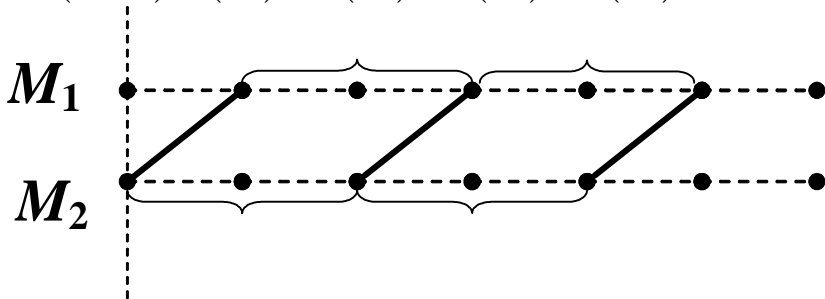


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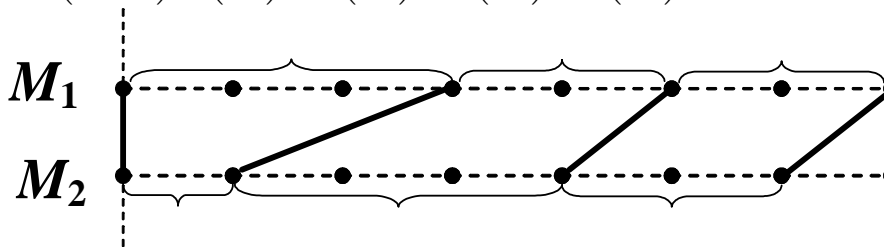
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$$A = \begin{pmatrix} 2 & 3 \\ 1 & 1 \end{pmatrix} : \begin{pmatrix} 1 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 3 \\ 2 \end{pmatrix} \rightarrow \begin{pmatrix} 5 \\ 4 \end{pmatrix} \rightarrow \begin{pmatrix} 7 \\ 6 \end{pmatrix} \rightarrow \dots$$



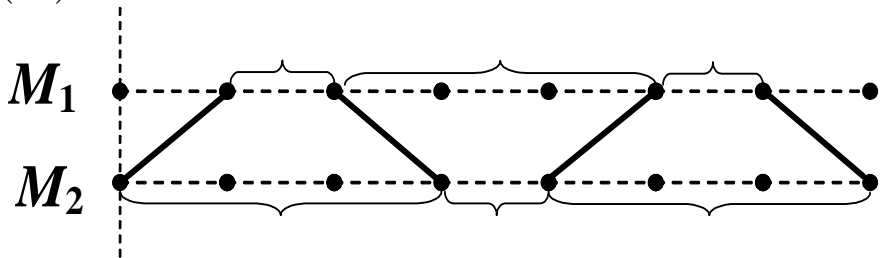
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- Since

$$x(r) = A \otimes x(r-1) = A^{(2)} \otimes x(r-2) = \dots = A^{(r)} \otimes x(0),$$

a steady regime is reached if and only if  $A^{(r)} \otimes x(0)$  “hits” an eigenvector of  $A$  for some  $r$ .



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## Problem formulation

Given:  $A = (a_{ij}) \in \overline{\mathbb{R}}^{n \times n}$

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- **Problem 3:** Is  $A$  *robust*? (That is for **every**  $x \in \overline{\mathbb{R}}^n$ ,  $x \neq \varepsilon$ , there is a  $k$  such that  $A^{(k)} \otimes x$  is an eigenvector of  $A$ )

# KEY PLAYERS

The maximum cycle mean and metric matrix

- $A = (a_{ij}) \in \overline{\mathbb{R}}^{n \times n} \longrightarrow D_A = (N, \{(i, j); a_{ij} > \varepsilon\}, (a_{ij})) \dots$   
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$$\lambda(A) = \max \left\{ \frac{a_{i_1 i_2} + a_{i_2 i_3} + \dots + a_{i_k i_1}}{k}; (i_1, \dots, i_k, i_1) \text{ cycle in } D_A \right\}$$

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- If

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- $\Gamma(A) = A \oplus A^2 \oplus \dots \oplus A^n$  (*metric matrix*)



# MMIPP: THE STEADY REGIME

## Classical results

### Theorem (R.A.Cuninghame-Green)

*If  $A \in \overline{\mathbb{R}}^{n \times n}$  is irreducible then  $\lambda(A)$  is the unique eigenvalue of  $A$  and the columns of  $\Gamma \left( (\lambda(A))^{-1} \otimes A \right)$  with zero diagonal entries are a complete set of generators of the eigenspace of  $A$ .*

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Theorem (G.Cohen, D.Dubois, J.-P.Quadrat, M.Viot)

*(Cyclicity Theorem) If  $A \in \overline{\mathbb{R}}^{n \times n}$  is irreducible, then  $A$  is ultimately periodic, that is there exist positive integers  $p$  and  $k_0$  such that:*

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*Let  $A \in \overline{\mathbb{R}}^{n \times n}$ ,  $A \neq \varepsilon$  be irreducible. Then the following are equivalent:*

- *$A$  is robust*

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# MMIPP: THE STEADY REGIME

## Classical results

Theorem (G.Cohen, D.Dubois, J.-P.Quadrat, M.Viot)

*(Cyclicity Theorem) If  $A \in \overline{\mathbb{R}}^{n \times n}$  is irreducible, then  $A$  is ultimately periodic, that is there exist positive integers  $p$  and  $k_0$  such that:*

$$A^{(k+p)} = (\lambda(A))^{(p)} \otimes A^{(k)} \text{ for all } k \geq k_0.$$

$p$  smallest ...  $p$  is the *period* of  $A$ ,  $p = p(A)$

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- $A \otimes x = b$  (*feasibility*) ...  $O(n)$  [CG]

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- $A \otimes x = b$  (*feasibility*) ...  $O(n)$  [CG]
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- *Optimizing starting times* ... polynomial/NP-complete [PB+KPT]



# WHAT IS MAX-ALGEBRA

Key players: The principal solution

$$\bullet \begin{pmatrix} -2 & 2 & 2 \\ -5 & -3 & -2 \\ \varepsilon & \varepsilon & 3 \\ -3 & -3 & 2 \\ 1 & 4 & \varepsilon \end{pmatrix} \otimes \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ -2 \\ 1 \\ 0 \\ 5 \end{pmatrix}$$

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$$\bullet M_2 \cup M_3 = M \text{ hence}$$

$$S(A, b) = \left\{ (x_1, x_2, x_3)^T \in \overline{\mathbb{R}}^3; x_1 \leq 3, x_2 = 1, x_3 = -2 \right\}.$$

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Theorem (S.Gaubert, 1992)

(Spectral Theorem) If  $A \in \overline{\mathbb{R}}^{n \times n}$  then

$$\Lambda(A) = \{\lambda(A_{ii}); \lambda(A_{ii}) \geq \lambda(A_{jj}) \text{ if } N_j \rightarrow N_i\}$$

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- $\lambda(A) = 5$
- $N_1, N_4$  are spectral

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$N_i$  is called *trivial* if  $A_{ii}$  is  $1 \times 1$  matrix ( $\varepsilon$ )

### Theorem

(Robustness criterion) If  $A \in \overline{\mathbb{R}}^{n \times n}$ ,  $A \neq \varepsilon$  then  $A$  is robust if and only if

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- $\lambda(A_{ii}) = \lambda(A_{jj})$  if  $N_i, N_j$  are non-trivial,  $N_i \rightarrow N_j$  and  $N_j \rightarrow N_i$  ( $i, j \in R$ )

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- The reason:  $N_1 \nrightarrow N_2$  and  $N_2 \nrightarrow N_1$  but  $\lambda(N_1) \neq \lambda(N_2)$ .

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# References

## Historical remarks

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