

COMPUTATION OF THE SECOND MAXIMUM PATH WEIGHT IN A MAX-PLUS MATRIX

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ABSTRACT

The problem of efficient computation of all second maximum path weights for a given square matrix A of dimension n in max-plus algebra is solved by computing so-called double power sequence of a specific extension \tilde{A} of A . An algorithm using this method is presented, by which the second maximum cycle mean value $\lambda_2(A)$ can be computed in $O(n^4)$ time.

Keywords: orbit period, max-plus algebra

1. INTRODUCTION

Matrix computations using binary operations of maximum and plus, in analogy to operation plus and times known in classical linear algebra, are often involved in optimization problems, see [2–5, 11, 17]. Max-plus matrices are important in the study of discrete events systems (DES), see [3, 4, 11]. The steady states of DES correspond to eigenvectors of max-plus matrices, and the periodical behavior of DES is reflected by so-called linear periodicity of the power sequence of the transient max-plus matrix, see [6, 7, 14, 15]. The eigenvalues of max-plus matrices are closely connected with the maximum (or dually: minimum) cycle mean, see [10, 12]. It has been shown in [13, 16] that the linear pre-period of a max-plus matrix can be arbitrary long, regardless of the matrix dimension. For the estimation of the linear pre-period length of a given matrix, the values of the maximum cycle mean and the second maximum cycle mean in the associated digraph are important, see [8, 9].

Closely related question of efficient computation of all second maximum path weights for a given square matrix A of dimension n in max-plus algebra is considered in this paper. The main tool for solving the problem are so-called double matrices and their multiplication. An algorithm for computing the second maximum cycle mean value $\lambda_2(A)$ in $O(n^4)$ time is described in the last section. Similar problems using different approach were considered in [1, 5].

2. NOTIONS AND NOTATION

By max-plus algebra we understand the set \mathbb{R}^* of all real numbers with added element $\varepsilon = -\infty$, together with binary operation of maximum denoted as \oplus , and the operation of addition denoted as \otimes (clearly, the element $-\infty$ is absorbing in respect to the operation \otimes). We remark that the results in this paper remain valid in more general situation, when \mathbb{R}^* is substituted by arbitrary linearly ordered commutative and divisible group G^* in additive notation, with the least element absorbing in respect to the group operation. Then \oplus and \otimes are the operation of maximum and the group operation on G^* , respectively.

For a fixed natural number n we denote $N = \{1, 2, \dots, n\}$. The set of all $n \times n$ matrices (n -dimensional

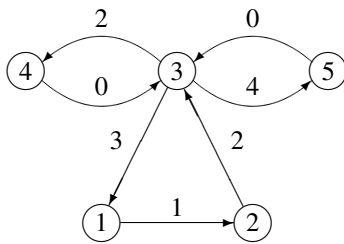
vectors) over \mathbb{R}^* is denoted by $\mathbb{R}^*(n, n)$ ($\mathbb{R}^*(n)$). Matrix operations over \mathbb{R}^* are performed with respect to operations \oplus, \otimes formally in the same way as matrix operations over any field. Thus for $A, B \in \mathbb{R}^*(m, n)$ we define the sum $A \oplus B = C \in \mathbb{R}^*(m, n)$, with $c_{ij} = a_{ij} \oplus b_{ij}$, for all $i \in M$ and $j \in N$. Further for $A \in \mathbb{R}^*(m, r)$ and $B \in \mathbb{R}^*(r, n)$ we define the product $A \otimes B = C \in \mathbb{R}^*(m, n)$ with $c_{ij} = \sum_k^{\oplus} a_{ik} \otimes b_{kj}$, for all $i \in M$ and $j \in N$. A digraph $\mathcal{G} = \mathcal{G}(A)$ is associated to every matrix $A \in \mathbb{R}^*(n, n)$, with the vertex set $V(\mathcal{G}) = N$ and with the edge set $E(\mathcal{G}) = \{(i, j); a_{ij} \neq \varepsilon\}$. The edges in $\mathcal{G}(A)$ are evaluated by the weight function w defined by the formula $w(i, j) = a_{ij}$, and the weight $w(p)$ of a path $p = (p_0, p_1, \dots, p_r)$ of length $|p| = r$ in $\mathcal{G}(A)$ is defined as the sum of all weights $w(p_{i-1}, p_i)$ for $i = 1, 2, \dots, r$. The weight mean $\bar{w}(p)$ is computed by the formula $\bar{w}(p) = w(p)/r$, for paths of positive length.

It has been proved in [4] that the maximum weight of all paths connecting node i with node j of a given length r for a matrix A in max-plus algebra can be found by computing the r^{th} formal power $A^{(r)}$ of the matrix A as illustrated in the following example.

Example 2.1 Input matrix $A \in \mathbb{R}^*(n, n)$, with $n = 5$,

$$A = \begin{bmatrix} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \end{bmatrix}$$

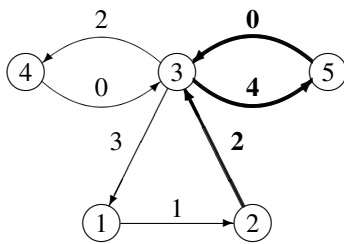
and the corresponding digraph $\mathcal{G}(A)$



For instance, if we are interested in finding the maximum weight of a path of length $r = 3$ from the node $i = 2$ to the node $j = 3$, then the maximum weight is represented by the element $a_{23}^{(3)} = 6$ in $A^{(3)}$,

$$A^{(3)} = \begin{bmatrix} 6 & \varepsilon & \varepsilon & 5 & 7 \\ \varepsilon & 6 & 6 & \varepsilon & \varepsilon \\ 7 & \varepsilon & 6 & 6 & 8 \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon \end{bmatrix}.$$

It can be easily seen in this simple example that the corresponding path with the maximum weight is the path $(2, 3, 5, 3)$ (marked in digraph $\mathcal{G}(A)$ below by printing the arcs of the path and their weights in bold).



Similarly, we can easily see that the path of length $r = 3$ from the node $i = 2$ to the node $j = 3$ with the second maximum weight is the path $(2, 3, 4, 3)$. Now a natural question arises: how can be the second maximum weight of all paths of a given length connecting a given node with another given node computed in general situation? The second maximum cycle mean (as well as the maximum cycle mean) is, namely, one of the tools to estimate the length of the non-periodic parth (the so-called pre-period) of a linear periodic matrix in max-plus algebra.

3. DOUBLE MATRIX

To answer the question put at the end of the previous paragraph, we start with extending the operation of finding the maximum element of a subset of \mathbb{R}^* to finding the two largest values.

Definition 3.1 Let $S \subseteq \mathbb{R}^*$ be a finite set. We define $\max^2 S = (s_1, s_2)$ by setting $s_1 = \max S$ and $s_2 = \max(S \setminus \{s_1\})$.

Lemma 3.1 Let be $S \subseteq \mathbb{R}^*$, $\max^2 S = (s_1, s_2)$. Then the following statements hold true

- (i) if $S = \emptyset$, or if $S = \{\varepsilon\}$, then $s_1 = s_2 = \varepsilon$,

- (ii) if $S = \{a\}$, $a > \varepsilon$, or if $S = \{a, \varepsilon\}$, $a > \varepsilon$, then $s_1 = a$, $s_2 = \varepsilon$,

- (iii) if there are $a, b \in S$, $a > b > \varepsilon$, then $s_1 > s_2 > \varepsilon$.

Proof. All statements in the lemma follow directly from Definition 3.1 and from the fact that maximum of the empty subset in \mathbb{R}^* is ε (i.e. the least element in \mathbb{R}^*).

□

Main tool for computing the second maximum cycle mean value $\lambda_2(A)$ of a given square matrix over max-plus algebra is the notion of double matrix with a special operation of double matrix multiplication introduced in this section.

Definition 3.2 An ordered pair of matrices $\bar{A} = ({}^1A, {}^2A)$ is called a double matrix (of dimension n), if ${}^1A, {}^2A \in \mathbb{R}^*(n, n)$ and if ${}^1a_{ij} > {}^2a_{ij}$, or ${}^1a_{ij} = {}^2a_{ij} = \varepsilon$, holds true for any $i, j \in N$. 1A is the left-hand matrix of \bar{A} and 2A is the right-hand matrix of \bar{A} . The set of all double matrices of a fixed dimension n is denoted as $\mathcal{D}(n, n)$. The double max-plus product $\bar{A} \otimes \bar{B} = \bar{C} = ({}^1C, {}^2C)$ is defined for any two matrices $\bar{A} = ({}^1A, {}^2A)$, $\bar{B} = ({}^1B, {}^2B) \in \mathcal{D}(n, n)$ by putting

$$S_{ij}(\bar{A}, \bar{B}) = \{ {}^1a_{ik} + {}^1b_{kj}, {}^1a_{ik} + {}^2b_{kj}, {}^2a_{ik} + {}^1b_{kj}; k \in N \} \tag{3.1}$$

and

$$({}^1c_{ij}, {}^2c_{ij}) = \max^2 S_{ij}(\bar{A}, \bar{B}) \tag{3.2}$$

for any $i, j \in N$.

The double max-plus multiplication of two double matrices results in a double matrix, as illustrated in the next example and proved in the following lemma.

Example 3.1 Let $\bar{A}, \bar{B} \in \mathcal{D}(2, 2)$.

$$\bar{A} = \left[\begin{array}{cc|cc} 1 & 2 & 0 & \varepsilon \\ \varepsilon & 0 & \varepsilon & -2 \end{array} \right], \quad \bar{B} = \left[\begin{array}{cc|cc} 3 & 2 & 1 & -1 \\ 1 & \varepsilon & 0 & \varepsilon \end{array} \right].$$

Let us find the double product of the given matrices $\bar{A} \otimes \bar{B} = \bar{C}$. Using formulas (3.1) and (3.2) from Definition 3.2 the elements of the resulting matrix \bar{C} can be computed.

$$S_{11}(\bar{A}, \bar{B}) = \{ {}^1a_{1k} + {}^1b_{k1}, {}^1a_{1k} + {}^2b_{k1}, {}^2a_{1k} + {}^1b_{k1}; k = 1, 2 \} = \{ 4, 3, 2, \varepsilon \},$$

$$({}^1c_{11}, {}^2c_{11}) = \max^2 S_{11}(\bar{A}, \bar{B}) = (4, 3),$$

$$S_{12}(\bar{A}, \bar{B}) = \{ {}^1a_{1k} + {}^1b_{k2}, {}^1a_{1k} + {}^2b_{k2}, {}^2a_{1k} + {}^1b_{k2}; k = 1, 2 \} = \{ 3, 2, 0, \varepsilon \},$$

$$({}^1c_{12}, {}^2c_{12}) = \max^2 S_{12}(\bar{A}, \bar{B}) = (3, 2),$$

$$S_{21}(\bar{A}, \bar{B}) = \{ {}^1a_{2k} + {}^1b_{k1}, {}^1a_{2k} + {}^2b_{k1}, {}^2a_{2k} + {}^1b_{k1}; k = 1, 2 \} = \{ 1, 0, -1, \varepsilon \},$$

$$({}^1c_{21}, {}^2c_{21}) = \max^2 S_{21}(\bar{A}, \bar{B}) = (1, 0),$$

$$S_{22}(\bar{A}, \bar{B}) = \{ {}^1a_{2k} + {}^1b_{k2}, {}^1a_{2k} + {}^2b_{k2}, {}^2a_{2k} + {}^1b_{k2}; \\ k = 1, 2 \} = \{ \varepsilon \}, \\ ({}^1c_{22}, {}^2c_{22}) = \max^2 S_{22}(\bar{A}, \bar{B}) = (\varepsilon, \varepsilon).$$

Thus

$$\bar{C} = \left[\begin{array}{cc|cc} 4 & 3 & 3 & 2 \\ 1 & \varepsilon & 0 & \varepsilon \end{array} \right].$$

Lemma 3.2 The set $\mathcal{D}(n, n)$ is closed under the double max-plus multiplication \otimes .

Proof. Let $\bar{A}, \bar{B} \in \mathcal{D}(n, n)$. We shall show that the product $\bar{C} = \bar{A} \otimes \bar{B}$ belongs to $\mathcal{D}(n, n)$. It is necessary to show that for any $i, j \in N$, the condition ${}^1c_{ij} > {}^2c_{ij}$, or ${}^1c_{ij} = {}^2c_{ij} = \varepsilon$ holds true. The condition follows immediately from the statements of Lemma 3.1, if we put $S = S_{ij}(\bar{A}, \bar{B})$, $s_1 = {}^1c_{ij}$ and $s_2 = {}^2c_{ij}$.

□

Double powers of a double matrix $\bar{A} \in \mathcal{D}(n, n)$ are defined by recursion

$$\bar{A}^{(1)} = \bar{A}, \\ \bar{A}^{(r+1)} = \bar{A}^{(r)} \otimes \bar{A}, \quad \text{for } r = 1, 2, \dots$$

The equations (3.1) and (3.2) from Definition 3.2 turn for any $i, j \in N$ to

$$S_{ij}(\bar{A}^{(r)}, \bar{A}) = \{ {}^1a_{ik}^{(r)} + {}^1a_{kj}, {}^1a_{ik}^{(r)} + {}^2a_{kj}, {}^2a_{ik}^{(r)} + {}^1a_{kj}; k \in N \} \quad (3.3)$$

and

$$({}^1a_{ij}^{(r+1)}, {}^2a_{ij}^{(r+1)}) = \max^2 S_{ij}(\bar{A}^{(r)}, \bar{A}), \quad (3.4)$$

for $r = 1, 2, \dots$

4. MATRIX EXTENSION

The matrix extension of a matrix A to a double matrix \bar{A} is introduced in this section. An interpretation of the sequence of double powers of the matrix extension \bar{A} of a given matrix A is described.

Any matrix $A \in \mathbb{R}^*(n, n)$ can be formally extended to a double matrix $\bar{A} = ({}^1A, {}^2A) \in \mathcal{D}(n, n)$ by setting ${}^1A = A$ and ${}^2A = E$ with all input values in E equal to ε . Clearly, the condition ${}^1a_{ij} > {}^2a_{ij}$, or ${}^1a_{ij} = {}^2a_{ij} = \varepsilon$, is fulfilled for any $i, j \in N$.

In the computation of the sequence of double powers of a matrix extension we can use the fact that the right-hand matrix of the matrix extension \bar{A} consists of the values ${}^2a_{kj} = \varepsilon$. Hence, for matrix extensions the computation can be simplified, in comparison with the definition above, as follows:

$$S_{ij}(\bar{A}^{(r)}, \bar{A}) = \{ {}^1a_{ik}^{(r)} + {}^1a_{kj}, {}^2a_{ik}^{(r)} + {}^1a_{kj}; k \in N \} \quad (4.1)$$

and

$$({}^1a_{ij}^{(r+1)}, {}^2a_{ij}^{(r+1)}) = \max^2 S_{ij}(\bar{A}^{(r)}, \bar{A}), \quad (4.2)$$

for $r = 1, 2, \dots$

Example 4.1 Input matrix $A \in \mathbb{R}^*(n, n)$, with $n = 5$,

$$A = \left[\begin{array}{ccccc} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \end{array} \right].$$

Let us find the power sequence of length 5 of the extension \bar{A} of the given matrix

$$\bar{A} = \left[\begin{array}{ccccc|ccccc} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{array} \right].$$

The successive powers of \bar{A} will be computed according to (4.1) and (4.2).

For $r = 1$

$$\bar{A}^{(2)} = \bar{A}^{(1)} \otimes \bar{A} = \bar{A} \otimes \bar{A},$$

$$S_{ij}(\bar{A}, \bar{A}) = \{ {}^1a_{ik} + {}^1a_{kj}, {}^2a_{ik} + {}^1a_{kj}; k = 1, 2, \dots, 5 \}, \\ ({}^1a_{ij}^{(2)}, {}^2a_{ij}^{(2)}) = \max^2 S_{ij}(\bar{A}, \bar{A}),$$

$$\bar{A}^{(2)} = \left[\begin{array}{ccccc|ccccc} \varepsilon & \varepsilon & 3 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ 5 & \varepsilon & \varepsilon & 4 & 6 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \end{array} \right].$$

For $r = 2$

$$\bar{A}^{(3)} = \bar{A}^{(2)} \otimes \bar{A},$$

$$S_{ij}(\bar{A}^{(2)}, \bar{A}) = \{ {}^1a_{ik}^{(2)} + {}^1a_{kj}, {}^2a_{ik}^{(2)} + {}^1a_{kj}; k = 1, 2, \dots, 5 \}, \\ ({}^1a_{ij}^{(3)}, {}^2a_{ij}^{(3)}) = \max^2 S_{ij}(\bar{A}^{(2)}, \bar{A}),$$

$$\bar{A}^{(3)} = \left[\begin{array}{ccccc|ccccc} 6 & \varepsilon & \varepsilon & 5 & 7 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & 6 & 6 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 4 & \varepsilon & \varepsilon \\ 7 & \varepsilon & 6 & 6 & 8 & 5 & \varepsilon & \varepsilon & 4 & 6 \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \end{array} \right].$$

For $r = 3$

$$\bar{A}^{(4)} = \bar{A}^{(3)} \otimes \bar{A},$$

$$S_{ij}(\bar{A}^{(3)}, \bar{A}) = \{ {}^1a_{ik}^{(3)} + {}^1a_{kj}, {}^2a_{ik}^{(3)} + {}^1a_{kj}; k = 1, 2, \dots, 5 \}, \\ ({}^1a_{ij}^{(4)}, {}^2a_{ij}^{(4)}) = \max^2 S_{ij}(\bar{A}^{(3)}, \bar{A}),$$

$$\bar{A}^{(4)} = \left[\begin{array}{ccccc|ccccc} \varepsilon & 7 & 7 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 5 & \varepsilon & \varepsilon \\ 9 & \varepsilon & 8 & 8 & 10 & 7 & \varepsilon & \varepsilon & 6 & 8 \\ 9 & 8 & 8 & 8 & 10 & \varepsilon & 6 & 6 & \varepsilon & \varepsilon \\ 7 & \varepsilon & 6 & 6 & 8 & 5 & \varepsilon & \varepsilon & 4 & 6 \\ 7 & \varepsilon & 6 & 6 & 8 & 5 & \varepsilon & \varepsilon & 4 & 6 \end{array} \right].$$

For $r = 4$

$$\bar{A}^{(5)} = \bar{A}^{(4)} \otimes \bar{A},$$

$$S_{ij}(\bar{A}^{(4)}, \bar{A}) = \{ {}^1a_{ik}^{(4)} + {}^1a_{kj}, {}^2a_{ik}^{(4)} + {}^1a_{kj}; k = 1, 2, \dots, 5 \}, \\ ({}^1a_{ij}^{(5)}, {}^2a_{ij}^{(5)}) = \max^2 S_{ij}(\bar{A}^{(4)}, \bar{A}),$$

$$\bar{A}^{(5)} = \left[\begin{array}{ccccc|ccccc} 10 & \varepsilon & 9 & 9 & 11 & 8 & \varepsilon & \varepsilon & 7 & 9 \\ 11 & 10 & 10 & 10 & 12 & \varepsilon & 8 & 8 & \varepsilon & \varepsilon \\ 11 & 10 & 10 & 10 & 12 & 9 & \varepsilon & 8 & 8 & 10 \\ 9 & 8 & 8 & 8 & 10 & \varepsilon & 6 & 6 & \varepsilon & \varepsilon \\ 9 & 8 & 8 & 8 & 10 & \varepsilon & 6 & 6 & \varepsilon & \varepsilon \end{array} \right].$$

Next theorem describes the properties of the sequence of double powers of the extension \bar{A} of a matrix A .

Theorem 4.1 *Let double matrix $\bar{A} \in \mathcal{D}(n, n)$ be the extension of $A \in \mathbb{R}^*(n, n)$. Then the sequence of double powers $\bar{A}^{(r)} = ({}^1A^{(r)}, {}^2A^{(r)})$, $r = 1, 2, \dots$ satisfies the following statements for any $i, j \in N$*

- (i) ${}^1a_{ij}^{(r)} = a_{ij}^{(r)}$ is the maximum weight of all paths in $\mathcal{G}(A)$ from i to j , of length r ,
- (ii) if ${}^1a_{ij}^{(r)} > \varepsilon$ and ${}^2a_{ij}^{(r)} > \varepsilon$, then ${}^2a_{ij}^{(r)}$ is the second maximum weight of all paths in $\mathcal{G}(A)$ from i to j , of length r ,
- (iii) if ${}^1a_{ij}^{(r)} > \varepsilon$ and ${}^2a_{ij}^{(r)} = \varepsilon$, then there is no path in $\mathcal{G}(A)$ from i to j , with weight less than ${}^1a_{ij}^{(r)} = a_{ij}^{(r)}$, of length r ,
- (iv) if ${}^1a_{ij}^{(r)} = \varepsilon$, then ${}^2a_{ij}^{(r)} = \varepsilon$ and there is no path in $\mathcal{G}(A)$ from i to j , of length r .

Proof. The proof proceeds on recursion with respect to r . For $r = 1$ and for any $i, j \in N$ there is exactly one path of length 1 from i to j in digraph $\mathcal{G}(A)$, if $a_{ij} > \varepsilon$, and there is no such path, if $a_{ij} = \varepsilon$. In the first case, the unique path consists of the single arc (i, j) and its weight is a_{ij} , which is the maximum weight, while the second maximum weight is ε (maximum of the empty set). In the second case, when there is no arc from i to j in $\mathcal{G}(A)$, both the maximum and the second maximum weight are equal to ε . Notice that ${}^2a_{ij} = \varepsilon$ by the definition of formal extension \bar{A} , for all i, j . Thus the assumption ${}^2a_{ij}^{(r)} > \varepsilon$ in statement (ii) is not fulfilled, for any i, j if $r = 1$.

Let us assume by recursion, that the statements (i), (ii), (iii) and (iv) are satisfied for some fixed value r . We shall prove that the statements are fulfilled also when r is replaced by $r + 1$.

(i) By the recursion assumption and by the definition of formal extension \bar{A} , the equalities ${}^1a_{ik}^{(r)} = a_{ik}^{(r)}$, ${}^1a_{kj} = a_{kj}$ hold true for every $k \in N$. Therefore,

$$\begin{aligned} {}^1a_{ij}^{(r+1)} &= \max \left\{ {}^1a_{ik}^{(r)} + {}^1a_{kj}, {}^2a_{ik}^{(r)} + {}^1a_{kj}; k \in N \right\} = \\ &= \max \left\{ {}^1a_{ik}^{(r)} + {}^1a_{kj}; k \in N \right\} = \\ &= \max \left\{ a_{ik}^{(r)} + a_{kj}; k \in N \right\} = a_{ij}^{(r+1)}. \end{aligned}$$

By the well-known properties of matrix powers in max-plus algebra, $a_{ij}^{(r+1)}$ is the maximum weight of a path in $\mathcal{G}(A)$ from i to j , of length $r + 1$.

(ii) Let us suppose that ${}^1a_{ij}^{(r+1)} > \varepsilon$. In view of the computation in the previous part of the proof, there is a non-empty set $K \subseteq N$ such that ${}^1a_{ij}^{(r+1)} = {}^1a_{ik}^{(r)} + {}^1a_{kj}$ for every

$k \in K$ and ${}^1a_{ij}^{(r+1)} > {}^1a_{ik}^{(r)} + {}^1a_{kj}$ for all $k \in N \setminus K$. Therefore,

$$\begin{aligned} {}^2a_{ij}^{(r+1)} &= \max \left(\left\{ {}^2a_{ik}^{(r)} + {}^1a_{kj}; k \in K \right\} \cup \right. \\ &\quad \left. \cup \left\{ {}^1a_{ik}^{(r)} + {}^1a_{kj}, {}^2a_{ik}^{(r)} + {}^1a_{kj}; k \in N \setminus K \right\} \right). \end{aligned}$$

Using the definition of \bar{A} , we have ${}^1a_{kj} = a_{kj}$ for all $k \in N$, hence we get

$$\begin{aligned} {}^2a_{ij}^{(r+1)} &= \max \left(\left\{ {}^2a_{ik}^{(r)} + a_{kj}; k \in K \right\} \cup \right. \\ &\quad \left. \cup \left\{ {}^1a_{ik}^{(r)} + a_{kj}, {}^2a_{ik}^{(r)} + a_{kj}; k \in N \setminus K \right\} \right). \end{aligned}$$

By recursion assumption (ii), we have ${}^1a_{ik}^{(r)} \geq {}^2a_{ik}^{(r)}$, which gives

$$\begin{aligned} {}^2a_{ij}^{(r+1)} &= \max \left(\left\{ {}^2a_{ik}^{(r)} + a_{kj}; k \in K \right\} \cup \right. \\ &\quad \left. \cup \left\{ {}^1a_{ik}^{(r)} + a_{kj}; k \in N \setminus K \right\} \right). \end{aligned} \quad (4.3)$$

In view of the statement (i) proved above, the assumption ${}^1a_{ij}^{(r+1)} > \varepsilon$ implies $a_{ij}^{(r+1)} > \varepsilon$, i.e. in $\mathcal{G}(A)$ there exists a path from i to j , of length $r + 1$, with the maximum weight $a_{ij}^{(r+1)}$. Let us consider a path p with the second maximum weight from i to j , of length $r + 1$ (if such a path exists in $\mathcal{G}(A)$). The path p consists of two shorter paths: a sub-path p' from node i to some node $k \in N$, of length r , and an arc (k, j) . Then $w(p) = w(p') + a_{kj}$. Clearly, ${}^1a_{ik}^{(r)} \geq w(p') \geq {}^2a_{ik}^{(r)}$. We shall consider two cases.

(a) If $k \in K$, then ${}^1a_{ik}^{(r)} > w(p') = {}^2a_{ik}^{(r)}$ and $w(p) = w(p') + a_{kj} = {}^2a_{ik}^{(r)} + a_{kj}$.

(b) If $k \in N \setminus K$, then ${}^1a_{ik}^{(r)} = w(p')$ and $w(p) = w(p') + a_{kj} = {}^1a_{ik}^{(r)} + a_{kj}$.

According to equation (4.3), we get ${}^2a_{ij}^{(r+1)} \geq w(p)$ in both cases.

By assumption ${}^2a_{ij}^{(r+1)} > \varepsilon$, there is $k \in K$ such that ${}^2a_{ij}^{(r+1)} = {}^2a_{ik}^{(r)} + a_{kj} > \varepsilon$, or there is $k \in N \setminus K$ such that ${}^2a_{ij}^{(r+1)} = {}^1a_{ik}^{(r)} + a_{kj} > \varepsilon$. This implies, in view of the recursion assumption, that there exists a path p from i to j , of length $r + 1$ in $\mathcal{G}(A)$, with ${}^1a_{ij}^{(r+1)} > w(p) = {}^2a_{ij}^{(r+1)}$. Hence p has the second maximum weight and the statement (ii) holds true for $r + 1$.

(iii) The proof goes similarly as above, only by assumption ${}^2a_{ij}^{(r+1)} = \varepsilon$ we get that there is no path p from i to j , of length $r + 1$ in $\mathcal{G}(A)$, with ${}^1a_{ij}^{(r+1)} > w(p)$.

(iv) The statement follows directly from Lemma 3.1.

□

Theorem 4.1 answers the question posed at the end of Section 2: how to compute the second maximum weight of all paths of a given length connecting a given node with another given node in the digraph corresponding to a given matrix? In the next example we will find the second maximum weight of all paths of length $r = 3$ connecting node 2 with node 3 in the digraph $\mathcal{G}(A)$ of the matrix A from Example 2.1.

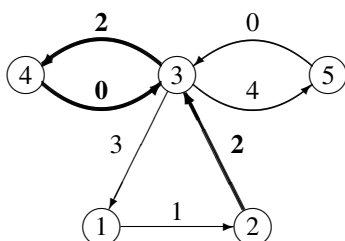
Example 4.2 Input matrix $A \in \mathbb{R}^*(n, n)$, with $n = 5$,

$$A = \begin{bmatrix} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \end{bmatrix}.$$

According to Theorem 4.1 we shall use the third power of the matrix extension of the given matrix. Since the considered matrix has been extended to a double matrix and the mentioned power has already been computed in Example 4.1, we can immediately find the answer. The second maximum weight of all paths of length $r = 3$ connecting node 2 with node 3 is represented by the element ${}^2a_{23}^{(3)} = 4$ in the right-hand matrix of $\bar{A}^{(3)}$.

$$\bar{A}^{(3)} = \left[\begin{array}{ccccc|ccccc} 6 & \varepsilon & \varepsilon & 5 & 7 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & 6 & 6 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & \mathbf{4} & \varepsilon & \varepsilon \\ 7 & \varepsilon & 6 & 6 & 8 & 5 & \varepsilon & \varepsilon & 4 & 6 \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ \varepsilon & 4 & 4 & \varepsilon & \varepsilon & \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \end{array} \right].$$

The corresponding path with the second maximum weight is the path (2,3,4,3) (marked in digraph $\mathcal{G}(A)$ below by printing the arcs of the path and their weights in bold).



5. THE SECOND MAXIMUM CYCLE MEAN

The previous results will be used in this section for computing the second maximum cycle mean value. In accordance with the notation introduced in Section 2, the weight $w(c)$ of a cycle c in digraph $\mathcal{G}(A)$ is defined as the sum of the weights of all arcs in c . If the length of c is positive, then the ratio $\bar{w}(c) := w(c)/|c|$ is called the cycle mean. The *maximum cycle mean* in $\mathcal{G}(A)$ is denoted by

$$\lambda(A) := \max \{ \bar{w}(c); c \text{ cycle in } \mathcal{G}(A) \}.$$

Since the cycle mean of a cycle can not exceed the cycle mean of its elementary subcycles, and since the number of elementary cycles is finite, $\lambda(A)$ always exists. If there is no cycle of positive length, then $\lambda(A) = \varepsilon$.

When considering cycles of arbitrary length, we can create combinations of cycles with the maximum cycle mean $\lambda(A)$ with cycles of lower cycle mean, and the cycle mean value of such combinations can be made arbitrarily close to $\lambda(A)$. Hence, strictly speaking, the second maximum cycle mean value does not exist for cycles of unrestricted length. For that reason we restrict our investigation

to cycles in $\mathcal{G}(A)$ with lengths not exceeding the dimension of the given square matrix A (which includes all elementary cycles).

The *second maximum cycle mean* in $\mathcal{G}(A)$ is denoted by

$$\lambda_2(A) := \max \{ \bar{w}(c); c \text{ cycle in } \mathcal{G}(A), |c| \leq n, \bar{w}(c) < \lambda(A) \}.$$

If there is no cycle of positive length with $\bar{w}(c) < \lambda(A)$, then $\lambda_2(A) = \varepsilon$.

Theorem 5.1 There is an algorithm which for every $A \in \mathbb{R}^*(n, n)$ computes the second maximum cycle mean value $\lambda_2(A)$ in $O(n^4)$ time.

Proof. The algorithm consists of the following steps:

Step 1. For a given matrix $A \in \mathbb{R}^*(n, n)$ compute the double matrix $\bar{A} = ({}^1A, {}^2A) \in \mathcal{D}(n, n)$ by setting ${}^1A = A$ and ${}^2A = E$ (all inputs in E are ε).

Step 2. Compute the double powers $\bar{A}^{(1)} = \bar{A}$, $\bar{A}^{(r)} = ({}^1A^{(r)}, {}^2A^{(r)})$, for $r = 2, 3, \dots, n$.

Step 3. For every $r = 1, 2, \dots, n$ compute the maximum and the second maximum mean value for cycles of length r in $\mathcal{G}(A)$ by formulas

$$(\lambda_1^{(r)}(A), \lambda_2^{(r)}(A)) = \max^2 \{ {}^1a_{ii}^{(r)}/r, {}^2a_{ii}^{(r)}/r; i \in N \}.$$

Step 4. Compute the maximum mean value for cycles of any length and the second maximum mean value for cycles of length at most n in $\mathcal{G}(A)$

$$(\lambda(A), \lambda_2(A)) = \max^2 \{ \lambda_1^{(r)}(A), \lambda_2^{(r)}(A); r = 1, 2, \dots, n \}.$$

It is easy to verify that the computational complexity for Step 1 is $O(n^2)$, for Step 2 the complexity is $O(n^4)$ and for Steps 3, 4 again $O(n^2)$. Hence, the whole computation is performed in $O(n^4)$ time. The correctness of the algorithm follows from Lemma 3.1 and Theorem 4.1.

□

Below, the algorithm is demonstrated on the input matrix A used in the previous examples.

Example 5.1 Input matrix $A \in \mathbb{R}^*(n, n)$, with $n = 5$,

$$A = \begin{bmatrix} \varepsilon & 1 & \varepsilon & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 2 & \varepsilon & \varepsilon \\ 3 & \varepsilon & \varepsilon & 2 & 4 \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \\ \varepsilon & \varepsilon & 0 & \varepsilon & \varepsilon \end{bmatrix}.$$

Let us compute the second maximum cycle mean value of all cycles of length at most 5. In the first and the second step of the algorithm the matrix extension \bar{A} and the first five consecutive powers of \bar{A} should be computed. The results of these computations can be found in Example 4.1, hence we can pass directly to the third step of the algorithm and compute the maximum and second maximum cycle mean value for cycles of length $r = 1, 2, \dots, 5$ using the formula

$$(\lambda_1^{(r)}(A), \lambda_2^{(r)}(A)) = \max^2 \{ {}^1a_{ii}^{(r)}/r, {}^2a_{ii}^{(r)}/r; i \in N \}.$$

Thus

$$\begin{aligned}(\lambda_1^{(1)}(A), \lambda_2^{(1)}(A)) &= \max^2\{\varepsilon\} = (\varepsilon, \varepsilon), \\(\lambda_1^{(2)}(A), \lambda_2^{(2)}(A)) &= \max^2\{2, 1\} = (2, 1), \\(\lambda_1^{(3)}(A), \lambda_2^{(3)}(A)) &= \max^2\{2\} = (2, \varepsilon), \\(\lambda_1^{(4)}(A), \lambda_2^{(4)}(A)) &= \max^2\{2, \frac{3}{2}\} = (2, \frac{3}{2}), \\(\lambda_1^{(5)}(A), \lambda_2^{(5)}(A)) &= \max^2\{2, \frac{8}{5}\} = (2, \frac{8}{5}).\end{aligned}$$

Finally, we compute the maximum cycle mean value of cycles of any length and the second maximum cycle mean value of cycles of length at most 5 in $\mathcal{G}(A)$ by formula

$$(\lambda(A), \lambda_2(A)) = \max^2\left\{\lambda_1^{(r)}(A), \lambda_2^{(r)}(A); r = 1, 2, \dots, 5\right\}.$$

Thus

$$(\lambda(A), \lambda_2(A)) = \max^2\left\{\varepsilon, 2, 1, \frac{3}{2}, \frac{8}{5}\right\} = \left(2, \frac{8}{5}\right).$$

References

- [1] BOULMAKOUL, A.: *Generalized path-finding algorithms on semirings and the fuzzy shortest path problem*, J. of Computational and Applied Math., 162 (2004), pp. 263–272.
- [2] BRUALDI, R. A. – RYSER, H. J.: *Combinatorial Matrix Theory*, in: Encyclopedia of Mathematics and its Applications, Vol. 39, Cambridge Univ. Press, Cambridge, 1991.
- [3] COHEN, G. – DUBOIS, D. – QUADRAT, J. P. – VIOT, M.: *A linear-system-theoretic view of discrete event processes and its use for performance evaluation in manufacturing*, IEE Transactions on Automatic Control, AC-30 (1985), pp. 210–220.
- [4] CUNINGHAME-GREEN, R. A.: *Minimax algebra*, Lecture Notes in Econom. and Math. Systems 166, Springer-Verlag, Berlin, 1979.
- [5] GONDRAN, M. – MINOUX, M.: *Graphes et algorithmes, 3ième edition*, Eyrolles, Paris, 1995.
- [6] GAVALEC, M.: *Linear matrix period in max-plus algebra*, Linear Algebra and its App. 307 (2000), pp. 167–182.
- [7] GAVALEC, M.: *Polynomial algorithm for linear matrix period in max-plus algebra*, Central Europ. J. Oper. Res. 8 (2000), pp. 247–258.
- [8] GAVALEC, M. – MOLNÁROVÁ, M.: *Computing linear orbit period in max-plus algebra*, Abstracts of the 22nd Europ. Conf. on Oper. Res. EURO 2007, Prague (2007), p. 35/MA-05-2.
- [9] GAVALEC, M. – MOLNÁROVÁ, M.: *Results on orbit periods in max-plus algebra*, 16th Inter. Sci. Conf. Math. Methods in Economy and Industry 2009, České Budějovice (2009).
- [10] GAVALEC, M. – PLAVKA, J.: *An $O(n^2)$ algorithm for maximum cycle mean of Monge matrices in max-algebra*, Discrete Appl. Math. 127 (2003), pp. 651–656.
- [11] HEIDERGOTT, B. – OLSDER, G. J. – van der WOUDE, J.: *Max Plus at Work*, Princeton University Press, Princeton, New Jersey, 2006.
- [12] KARP, R. M.: *A characterization of the minimum cycle mean in a digraph*, Discrete Math. 23 (1978), pp. 309–311.
- [13] MOLNÁROVÁ, M.: *Computational complexity of Nachtigall's representation*, Optimization 52 (2003), pp. 93–104.
- [14] MOLNÁROVÁ, M.: *Generalized matrix period in max-plus algebra*, Linear Algebra and its App. 404 (2005), pp. 345–366.
- [15] MOLNÁROVÁ, M. – PRIBIŠ, J.: *Matrix period in max-algebra*, Discrete Appl. Math. 103 (2000), pp. 167–175.
- [16] NACHTIGALL, K.: *Powers of Matrices over an Extremal Algebra with Applications to Periodic Graphs*, Math. Methods of Oper. Research 46 (1997), pp. 87–102.
- [17] ZIMMERMANN, U.: *Linear and Combinatorial Optimization in Ordered Algebraic Structures*, North Holland, Amsterdam, 1981.

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