

A Semi-Tensor Product of Tensors and Applications

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Abstract. A semi-tensor product of matrices is proposed as a generalization of usual matrix product in the case where the dimensions of two factor matrices do not match. The properties of the semi-tensor product of tensors and swap tensors based on the Einstein product are studied. Applications of this new tensor product in image restoration and in finite dimensional algebras are discussed.

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1. Introduction

In recent years various concepts of matrix theory including eigenvalues, multi-linear systems, tensor decompositions, data mining, degree theory have been extended to problems involving tensors [1–3, 8, 9, 11, 16, 17, 23]. The semi-tensor product of matrices proposed by Cheng [5] found applications in the control design of dynamic systems, finite automata, graph theory, differential geometry, algebra, and data science [5–7, 15].

Definition 1.1 (cf. Cheng [5], Cheng & Zhang [7]). Let $\mathbf{x} = [x_1, \dots, x_s]^T \in \mathbb{R}^s$, $\mathbf{y} = [y_1, \dots, y_t]^T \in \mathbb{R}^t$.

- (1) If $s = t \cdot n$, $n \in \mathbb{Z}_+$, we split \mathbf{x}^T into t equal blocks, $\mathbf{x}_1^T, \dots, \mathbf{x}_t^T$. Each block is an n -dimensional row vector. The (left) semi-tensor product of \mathbf{x}^T and \mathbf{y} is the n -dimensional row vector defined by

$$\mathbf{x}^T \ltimes \mathbf{y} := \sum_{k=1}^t y_k \mathbf{x}_k^T \in \mathbb{R}^{1 \times n}.$$

- (2) If $t = s \cdot n$, $n \in \mathbb{Z}_+$, we split \mathbf{y} into s equal blocks $\mathbf{y}_1, \dots, \mathbf{y}_s$. Each block is an n -dimensional column vector. The (left) semi-tensor product of \mathbf{x}^T and \mathbf{y} is the n -dimensional column vector defined by

$$\mathbf{x}^T \ltimes \mathbf{y} := \sum_{k=1}^s x_k \mathbf{y}_k \in \mathbb{R}^n.$$

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Definition 1.2 (cf. Cheng [5], Cheng & Zhang [7]). Let $M \in \mathbb{R}^{m \times n}$ and $N \in \mathbb{R}^{p \times q}$. If n is a divisor of p or p is a divisor of n , then the (left) semi-tensor product $C = [C^{ij}]$ of M and N , denoted by $C = M \ltimes N$, is a matrix that consists of $m \times q$ blocks, where each block is defined by

$$C^{ij} = M(i, :) \ltimes N(:, j), \quad i = 1, \dots, m, \quad j = 1, \dots, q.$$

For example, if $A \in \mathbb{R}^{m \times n}$, $\mathbf{x} \in \mathbb{R}^p = \mathbb{R}^{p \times 1}$, and n is a divisor of p , say $p = t \cdot n$, then $A \ltimes \mathbf{x} \in \mathbb{R}^{tm}$ is a column vector. If p is a divisor of n , say $n = s \cdot p$, then $A \ltimes \mathbf{x} \in \mathbb{R}^{m \times s}$ is an $m \times s$ matrix.

We also note the following properties of the semi-tensor products.

- (a) Setting $t = s$ in Definition 1.1, we obtain the usual Euclidean inner product, while setting $n = p$ in Definition 1.2, we obtain the usual matrix product.
- (b) The right semi-tensor product of matrices was introduced in [7]. However, it is analogous to Definition 1.2 and is not considered here.

Let I_n denote the $n \times n$ identity matrix and \otimes_K the Kronecker product.

Lemma 1.1 (Cheng *et al.* [5–7]). Let A, B, C be matrices such that the corresponding semi-tensor products are well defined. Then we have

- (i) If $A \in \mathbb{R}^{m \times np}$, $B \in \mathbb{R}^{p \times q}$, then $A \ltimes B = A(B \otimes_K I_n) \in \mathbb{R}^{m \times nq}$.
- (ii) If $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{np \times q}$, then $A \ltimes B = (A \otimes_K I_p)B \in \mathbb{R}^{mp \times q}$.
- (iii) $(A \ltimes B) \ltimes C = A \ltimes (B \ltimes C)$.
- (iv) $A \ltimes (\alpha B + \beta C) = \alpha A \ltimes B + \beta A \ltimes C$, where α and β are constants.
- (v) $(\alpha B + \beta C) \ltimes A = \alpha B \ltimes A + \beta C \ltimes A$, where α and β are constants.
- (vi) $(A \ltimes B)^T = B^T \ltimes A^T$.
- (vii) $(A \ltimes B)^{-1} = B^{-1} \ltimes A^{-1}$, where A and B are invertible.
- (viii) $\text{tr}(A \ltimes B) = \text{tr}(B \ltimes A)$, where $\text{tr}(M)$ denotes the trace of a square matrix M .

Now we consider an $mn \times mn$ matrix $W_{[m,n]}$, which plays an important role in semi-tensor products — cf. [5]. It is a permutation matrix, called the swap matrix, and constructed in the following way. Let

$$(11, 12, \dots, 1n, \dots, m1, m2, \dots, mn)$$

denote the columns of $W_{[m,n]}$ and

$$(11, 21, \dots, m1, \dots, 1n, 2n, \dots, mn)$$