PAPER DETAILS

TITLE: Pseudo matrix multiplication

AUTHORS: Osman KEÇILIOGLU, Halit GÜNDOGAN

PAGES: 37-43

ORIGINAL PDF URL: https://dergipark.org.tr/tr/download/article-file/774343

Article Electronically published on January 12, 2017

Commun.Fac.Sci.Univ.Ank.Series A1 Volume 66, Number 2, Pages 37-43 (2017) DOI: 10.1501/Commun1_0000000798 ISSN 1303-5991 http://communications.science.ankara.edu.tr/index.php?series=A1



PSEUDO MATRIX MULTIPLICATION

OSMAN KEÇİLİOĞLU AND HALİT GÜNDOĞAN

ABSTRACT. In this paper, a new matrix multiplication is defined in $\mathbb{R}^{m,n} \times \mathbb{R}^{n,p}$ by using scalar product in \mathbb{R}^n , where $\mathbb{R}^{m,n}$ is set of matrices of *m* rows and *n* columns. With this multiplication it has been shown that $\mathbb{R}^{n,n}$ is an algebra with unit. By considering this new multiplication we define eigenvalues and eigenvectors of square $n \times n$ matrix *A* and also present some applications.

1. INTRODUCTION

In [1], Lorentzian matrix multiplication was introduced. For some applications related to this multiplication, we refer the papers [2-5].

In the present paper we aim to define a new matrix multiplication using scalar product on \mathbb{R}^n of which index is ν . We generalize some properties given for ordinary matrix multiplication. As one of the most important properties of this new multiplication we don't need to use sign matrix to obtain orthogonal, symmetric matrix etc. In the third section, we examine the concepts of eigenvalues and its eigenvectors of square $n \times n$ matrix A. Finally, we study on diagonalizable matrix. We start with some basis expected potential.

We start with some basic concepts and notations.

Let $\mathbb{R}^{m,n}$ be the set of all $m \times n$ matrices. $\mathbb{R}^{m,n}$ with the matrix addition and the scalar-matrix multiplication is a real vector space. More properties of the ordinary matrix multiplication can be found in [7].

Let \mathbb{R}^n_{ν} be pseudo-Euclidean space over the real field \mathbb{R} equipped with a scalar product $\langle x, y \rangle_{\nu}$ which is symmetric, non degenerate bilinear form;

$$\langle x, y \rangle_{\nu} = -\sum_{i=1}^{\nu} x_i y_i + \sum_{i=\nu+1}^{n} x_i y_i$$

where $x, y \in \mathbb{R}^n$ and ν is an integer with $0 \le \nu \le n$ [8].

Communications de la Faculté des Sciences de l'Université d'Ankara. Séries A1. Mathematics and Statistics.

Received by the editors: July 28, 2016; Accepted: November 04, 2016.

²⁰¹⁰ Mathematics Subject Classification. 15A18.

Key words and phrases. p-matrix multiplication, eigenvalue, eigenvector.

^{©2017} Ankara University

³⁷

2. Pseudo Matrix Multiplication and Properties

Let A_1, \ldots, A_m denote the row vectors of $A = [a_{ij}] \in \mathbb{R}^{m,n}$ and B_1, \ldots, B_p denote the column vectors of $B = [b_{jk}] \in \mathbb{R}^{n,p}$. Then we define a new matrix multiplication denoted by " \bullet_{ν} ", as

$$A \bullet_{\nu} B = \begin{bmatrix} \langle A_1, B_1 \rangle_{\nu} & \langle A_1, B_2 \rangle_{\nu} & \cdots & \langle A_1, B_p \rangle_{\nu} \\ \langle A_2, B_1 \rangle_{\nu} & \langle A_2, B_2 \rangle_{\nu} & \cdots & \langle A_2, B_p \rangle_{\nu} \\ \vdots & \vdots & & \vdots \\ \langle A_m, B_1 \rangle_{\nu} & \langle A_m, B_2 \rangle_{\nu} & \cdots & \langle A_m, B_p \rangle_{\nu} \end{bmatrix}$$
$$= \begin{bmatrix} -\sum_{j=1}^{\nu} a_{ij} b_{jk} + \sum_{j=\nu+1}^{n} a_{ij} b_{jk} \end{bmatrix}.$$

We call this multiplication as pseudo matrix multiplication and if we let A_i to be *i*th row of A and B^j to be *j*th column of B then (i, j) entry of $A \bullet_{\nu} B$ is $\langle A_i, B^j \rangle_{\mu}$. Note that $A \bullet_{\nu} B$ is an $m \times p$ matrix. We will denote $\mathbb{R}^{m,n}$ with pseudo matrix multiplication by $\mathbb{R}^{m,n}_{\nu}$. In the special case of ν we get followings:

- (1) For $\nu = 0$, $A \bullet_{\nu} B$ coincide with usual matrix multiplication.
- (2) For $\nu = 1, A \bullet_{\nu} B$ coincide with Lorentzian matrix multiplication defined in [1].

Also, in the results and definitions given in throughout the paper one can easily obtain the classical ones when $\nu = 0$.

In the sequel we present some properties of new type matrix multiplication and give the analogous of definitions, in the classical matrix multiplication, by \bullet_{ν} .

Theorem 1. The following statements are satisfied.

- i) For every $A \in \mathbb{R}^{m,n}_{\nu}, B \in \mathbb{R}^{n,p}_{\nu}, C \in \mathbb{R}^{p,r}_{\nu}, A \bullet_{\nu} (B \bullet_{\nu} C) = (A \bullet_{\nu} B) \cdot_{L} C$
- *ii)* For every $A \in \mathbb{R}^{m,n}_{\nu}$, $B, C \in \mathbb{R}^{n,p}_{\nu}$, $A \bullet_{\nu} (B+C) = A \bullet_{\nu} B + A \bullet_{\nu} C$ *iii)* For every $A, B \in \mathbb{R}^{m,n}_{\nu}$, $C \in \mathbb{R}^{n,p}_{\nu}$, $(A+B) \bullet_{\nu} C = A \bullet_{\nu} C + B \bullet_{\nu} C$
- iv) For every $k \in \mathbb{R}$, $A \in \mathbb{R}^{m,n}_{\nu}$, $B \in \mathbb{R}^{n,p}_{\nu}$, $k(A \bullet_{\nu} B) = (kA) \bullet_{\nu} B = A \bullet_{\nu} (kB)$

Definition 1. $n \times n$ identity matrix according to pseudo matrix multiplication, denoted by $I_n = [i_{ij}]$, is defined by

$$i_{ij} = \begin{cases} -1 &, i = j \text{ and } 1 \le i, j \le \nu \\ 1 &, i = j \text{ and } \nu + 1 \le i, j \le n \\ 0 &, i \ne j \end{cases},$$

that is

$$I_n = \begin{bmatrix} -1 & \cdots & 0 & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & -1 & 0 & \cdots & 0 \\ 0 & \cdots & 0 & 1 & \cdots & 0 \\ \vdots & \ddots & \vdots & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & 0 & \cdots & 1 \end{bmatrix}$$

Note that for every $A \in \mathbb{R}^{m,n}_{\nu}$, $I_m \bullet_{\nu} A = A \bullet_{\nu} I_n = A$.

Corollary 1. $\mathbb{R}^{n,n}_{\nu}$ with pseudo matrix multiplication is an algebra with unit.

Definition 2. An $n \times n$ matrix A is called p-invertible, if there exists an $n \times n$ matrix B such that $A \bullet_{\nu} B = B \bullet_{\nu} A = I_n$. Then B is called p-inverse of A and is shown by A^{-1} .

Definition 3. Let $A = [a_{ij}]$ be an $m \times n$ matrix. Then the transpose of A is the $n \times m$ matrix A^T obtained by interchanging the rows and columns of A, so that the (i, j) th entry of A^T is a_{ji} .

Theorem 2. Let A and B be matrices of the appropriate sizes so that the following operations make sense, and c be a scalar. Then

- (1) $(A + B)^{T} = A^{T} + B^{T}$ (2) $(A \bullet_{\nu} B^{T}) = B^{T} \bullet_{\nu} A^{T}$ (3) $(cA)^{T} = cA^{T}$ (4) $(A^{T})^{T} = A.$

Definition 4. Let $A \in \mathbb{R}^{n,n}_{\nu}$. If $A^T = A$, $A^T = -A$ and $A^{-1} = A^T$ then A is said to be symmetric, skew-symmetric and p-orthogonal matrix, respectively.

Based on this definition, we obtain the following result.

Theorem 3. Let $A \in \mathbb{R}^{n,n}_{\nu}$. Then

- (1) A is p-orthogonal if and only if the row vectors of A form an orthonormal basis of \mathbb{R}^n_{ν} under the scalar product; and
- (2) A is p-orthogonal if and only if the column vectors of A form an orthonormal basis of \mathbb{R}^n_{ν} under the scalar product.

Proof. We shall only prove (1), since the proof of (2) is almost identical. Let A_1, \ldots, A_n denote the row vectors of A. Then

$$A \bullet_{\nu} A^{T} = \begin{bmatrix} \langle A_{1}, A_{1} \rangle_{\nu} & \cdots & \langle A_{1}, A_{n} \rangle_{\nu} \\ \vdots & & \vdots \\ \langle A_{n}, A_{1} \rangle_{\nu} & \cdots & \langle A_{n}, A_{n} \rangle_{\nu} \end{bmatrix}$$

It follows that $A \bullet_{\nu} A^T = I_n$ if and only if for every $i, j = 1, \ldots, n$

$$\langle A_i, A_j \rangle = \begin{cases} -1 &, i = j \text{ and } 1 \le i, j \le \nu \\ 1 &, i = j \text{ and } \nu + 1 \le i, j \le n \\ 0 &, i \ne j \end{cases}.$$

Then $\{A_1, \ldots, A_n\}$ is an orthonormal basis of \mathbb{R}^n_{ν} .

Definition 5. The determinant of a matrix $A = [a_{ij}] \in \mathbb{R}^{n,n}_{\nu}$ is denoted by det A and defined as

$$\det A = \sum_{\sigma \in S_n} s(\sigma) a_{\sigma(1)1} a_{\sigma(2)2} \cdots a_{\sigma(n)n},$$

where S_n is set of all permutations of the set $\{1, 2, \dots, n\}$ and $s(\sigma)$ is sign of the permutation σ .

Theorem 4. For every $A, B \in \mathbb{R}^{n,n}_{\nu}$, $\det(A \bullet_{\nu} B) = (-1)^{\nu} \det A \cdot \det B$.

Proof. Let $A = [a_{ij}]$, $B = [b_{jk}]$ and $A \bullet_{\nu} B = C$. Let us denote the *i*th column of matrix A by A^i and the kth column of matrix C by C^k

$$C^{k} = -b_{1k}A^{1} - b_{2k}A^{2} - \dots - b_{\nu k}A^{\nu} + b_{(\nu+1)k}A^{\nu+1} + \dots + b_{nk}A^{n}, \quad 1 \le k \le n.$$

Then

$$\det(A \bullet_{\nu} B) = \det[-b_{11}A^{1} - b_{21}A^{2} - \dots - b_{\nu 1}A^{\nu} + b_{(\nu+1)1}A^{\nu+1} + \dots + b_{n1}A^{n}, \\ \dots, -b_{1n}A^{1} - b_{2n}A^{2} - \dots - b_{\nu n}A^{\nu} + b_{(\nu+1)n}A^{\nu+1} + \dots + b_{nn}A^{n}] \\ = \sum_{\sigma \in S_{n}} (-1)^{\nu} b_{\sigma(1)1}b_{\sigma(2)2} \cdots b_{\sigma(n)n} \det[A^{\sigma(1)}, A^{\sigma(2)}, \dots, A^{\sigma(n)}] \\ = (-1)^{\nu} \sum_{\sigma \in S_{n}} s(\sigma) b_{\sigma(1)1}b_{\sigma(2)2} \cdots b_{\sigma(n)n} \det[A^{1}, A^{2}, \dots, A^{n}] \\ = (-1)^{\nu} \det A \sum_{\sigma \in S_{n}} s(\sigma) b_{\sigma(1)1}b_{\sigma(2)2} \cdots b_{\sigma(n)n} \\ = (-1)^{\nu} \det A \det B.$$

3. Some Applications OF Pseudo Matrix Multiplication

Eigenvalues and eigenvectors play an important role in matrix theory because of its application in the areas of mathematics, physics and engineering. By this aim, we define the eigenvalues and eigenvectors of square $n \times n$ matrix A by pseudo matrix multiplication.

Definition 6. Let $A \in \mathbb{R}^{n,n}_{\nu}$. An eigenvector of A is a nonzero vector x in \mathbb{R}^{n}_{ν} such that

$$A \bullet_{\nu} x = \lambda x$$

for some scalar λ . The scalar λ is called an eigenvalue of the matrix A, and we say that the vector x is an eigenvector belonging to the eigenvalue λ .

Theorem 5. The eigenvectors of a symmetric matrix $A \in \mathbb{R}^{n,n}_{\nu}$ corresponding to different eigenvalues are orthogonal to each other.

Proof. For the eigenvectors x, y corresponding to two different eigenvalues λ, μ of the matrix A, we can say that $A \bullet_{\nu} x = \lambda x$ and $A \bullet_{\nu} y = \mu y$, so

$$y^{T} \bullet_{\nu} A \bullet_{\nu} x = \lambda y^{T} \bullet_{\nu} x = \lambda \langle x, y \rangle_{\nu}.$$
(3.1)

But numbers are always their own transpose, so

$$y^{T} \bullet_{\nu} A \bullet_{\nu} x = x^{T} \bullet_{\nu} A \bullet_{\nu} y$$
$$= x^{T} \bullet_{\nu} \mu y$$
$$= \mu \left(x^{T} \bullet_{\nu} y \right)$$
$$y^{T} \bullet_{\nu} A \bullet_{\nu} x = \mu \left\langle x, y \right\rangle_{\nu}.$$
(3.2)

From (3.1) and (3.2), we get

$$(\lambda - \mu) \langle x, y \rangle_{\nu} = 0.$$

So $\lambda = \mu$ or $\langle x, y \rangle_{\nu} = 0$, and it isn't the former, so x and y are orthogonal.

Example 1. Let

$$A = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 2 & -1 \\ 0 & -1 & 1 \end{bmatrix} \in \mathbb{R}_2^{3,3}.$$

A is a symmetric matrix. Then the eigenvalues of A are $\lambda_1 = 0, \lambda_2 = \sqrt{2} - 1$ and $\lambda_3 = -\sqrt{2} - 1$. Some eigenvectors of A corresponding to λ_1, λ_2 and λ_3 are

$$u_{1} = \begin{bmatrix} 1\\ -1\\ 1 \end{bmatrix} , \quad u_{2} = \begin{bmatrix} -1 + \sqrt{2}\\ -2 + \sqrt{2}\\ 1 \end{bmatrix} , \quad u_{3} = \begin{bmatrix} -1 - \sqrt{2}\\ -2 - \sqrt{2}\\ 1 \end{bmatrix}$$

respectively. For $i \neq j$, we get

$$\langle u_i, u_j \rangle_2 = u_i^T \bullet_2 u_j = 0.$$

Then $\{u_1, u_2, u_3\}$ form an orthogonal basis of \mathbb{R}^3_2 .

Definition 7. A matrix A is diagonalizable if there exists a nonsingular matrix P and a diagonal matrix D such that

$$D = P^{-1} \bullet_{\nu} A \bullet_{\nu} P.$$

Theorem 6. Let all the eigenvalues of $A \in \mathbb{R}^{n,n}_{\nu}$ are real. Then A diagonalizable if and only if it has n linearly independent eigenvectors.

Proof. Let x_1, \ldots, x_n be *n* linearly independent eigenvectors of *A* associated with the eigenvalues $\lambda_1, \ldots, \lambda_n$. That is,

$$A \bullet_{\nu} x_i = \lambda_i x_i \quad , \quad i = 1, \dots, n \; .$$

Now, we denote $P = \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}$. Since the columns of P are linearly independent, P is invertible. Let D be $diag[-\lambda_1, \ldots, -\lambda_{\nu}, \lambda_{\nu+1}, \ldots, \lambda_n]$. Then

$$A \bullet_{\nu} P = A \bullet_{\nu} \begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix}$$

= $\begin{bmatrix} \lambda_1 x_1 & \cdots & \lambda_n x_n \end{bmatrix}$
= $\begin{bmatrix} x_1 & \cdots & x_n \end{bmatrix} \bullet_{\nu} diag [-\lambda_1, \dots, -\lambda_{\nu}, \lambda_{\nu+1}, \dots, \lambda_n]$
= $P \bullet_{\nu} D.$

Since $A \bullet_{\nu} P = P \bullet_{\nu} D$, it follows that $D = P^{-1} \bullet_{\nu} A \bullet_{\nu} P$ which shows that A is diagonalizable.

To prove the other direction we assume that A is diagonalizable. Then there exists a nonsingular matrix P and a diagonal matrix $D = diag [-\lambda_1, \ldots, -\lambda_{\nu}, \lambda_{\nu+1}, \ldots, \lambda_n]$ such that

$$D = P^{-1} \bullet_{\nu} A \bullet_{\nu} P$$

If we multiply above equation with P from the left, we get

$$A \bullet_{\nu} P = P \bullet_{\nu} D \tag{3.3}$$

which implies

$$A \bullet_{\nu} v_i = \lambda_i v_i \quad , \quad i = 1, \dots, n \tag{3.4}$$

where v_i are columns of P. The equations (3.4) show that v_1, \ldots, v_n are eigenvectors of A corresponding to eigenvalues $\lambda_1, \ldots, \lambda_n$. Furthermore, since P is invertible, $\{v_1, \ldots, v_n\}$ are linearly independent.

Example 2. Consider the matrix

$$A = \begin{bmatrix} 1 & 2 & 0 \\ 1 & 0 & -1 \\ 1 & -2 & -2 \end{bmatrix} \in \mathbb{R}_2^{3.3}.$$

The eigenvalues of A are $\lambda_1 = -2, \lambda_2 = 0, \lambda_3 = -1$ and eigenvectors corresponding these eigenvalues are

$$u_1 = \begin{bmatrix} 2\\1\\0 \end{bmatrix} \quad , \quad u_2 = \begin{bmatrix} 2\\-1\\-2 \end{bmatrix} \quad , \quad u_3 = \begin{bmatrix} 1\\0\\-1 \end{bmatrix}$$

respectively. Therefore

$$P = \begin{bmatrix} 2 & 2 & 1 \\ 1 & -1 & 0 \\ 0 & -2 & -1 \end{bmatrix} \quad and \quad P^{-1} = \begin{bmatrix} \frac{1}{2} & 0 & -\frac{1}{2} \\ \frac{1}{2} & -1 & -\frac{1}{2} \\ 1 & -2 & -2 \end{bmatrix} .$$

42

Finally

$$P^{-1} \bullet_{\nu} A \bullet_{\nu} P = \begin{bmatrix} 2 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{bmatrix}.$$

References

- Gündoğan, H., Keçilioğlu, O., "Lorentzian matrix multiplication and the motions on Lorentzian plane", Glas. Mat. Ser. III 41(61) (2006), no. 2, 329–334
- [2] Keçilioğlu, O., Özkaldı, S., Gündoğan, H., "Rotations and Screw Motion with Timelike Vector in 3-Dimensional Lorentzian Space", Adv. Appl. Cliff ord Algebras, 22 (2012), 1081–1091.
- [3] Özkaldı, S., Gündoğan, H., "Dual split quaternions and screw motion in 3-dimensional Lorentzian space", Adv. Appl. Clifford Algebr. 21 (2011), no. 1, 193–202.
- [4] Gündoğan, H. Özkaldı, S., "Clifford product and Lorentzian plane displacements in 3dimensional Lorentzian space", Adv. Appl. Clifford Algebr. 19 (2009), no. 1, 43–50.
- [5] Yıldırım, H., Yüce, S., Kuruoğlu, N., "Holditch theorem for the closed space curves in Lorentzian 3-space", Acta Mathematica Scientia 2011,31B(1):172–180.
- [6] Hegedüs, G., Moore, B., "The Minkowski Planar 4R Mechanism", Int. Electron. J. Geom. 5 (2012), no. 1, 1–35.
- [7] Lang, S., "Linear Algebra", Addison-Wesley Publishing Co., Reading, Mass.-London-Don Mills, Ont. 1971.
- [8] O'Neill, B., "Semi-Riemannian Geometry With Applications to Relativity", Academic Press, Inc., New York, 1983.

Current address: Department of Statistics, Kırıkkale University, Kırıkkale, TURKEY E-mail address, Osman Keçilioğlu: okecilioglu@yahoo.com E-mail address, Halit Gündoğan: hagundogan@hotmail.com