UNIV. BEOGRAD. PUBL. ELEKTROTEHN. FAK. Ser. Mat. 15 (2004), 13–25. Available electronically at http://matematika.etf.bg.ac.yu

# APPLICATIONS OF THE HYPER-POWER METHOD FOR COMPUTING MATRIX PRODUCTS

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We introduce representations for  $\{1, 2, 3\}$ ,  $\{1, 2, 4\}$ -inverses in terms of matrix products involving the MOORE-PENROSE inverse. We also use representations of  $\{2, 3\}$  and  $\{2, 4\}$ -inverses of a prescribed rank, introduced in [6] and [9]. These representations can be computed by means of the modification of the hyper-power iterative process which is used in computing matrix products involving the MOORE-PENROSE inverse, introduced in [8]. Introduced methods have arbitrary high orders  $q \geq 2$ . A few comparisons with the known modification of the hyper-power method from [17] are presented.

### 1. INTRODUCTION

Let  $\mathbb{C}^n$  be the *n*-dimensional complex vector space,  $\mathbb{C}^{m \times n}$  the set of  $m \times n$ complex matrices, and  $\mathbb{C}_r^{m \times n} = \{X \in \mathbb{C}^{m \times n} : \operatorname{rank}(X) = r\}$ . We use  $\mathcal{N}(A)$ to denote the kernel and  $\mathcal{R}(A)$  to denote the range of A, and  $\rho(A)$  to denote the spectral radius of A. If  $A \in \mathbb{C}^{n \times n}$  and L, M are complementary subspaces of  $\mathbb{C}^n$ , then  $P_{L,M}$  denotes the projector on L along M.

For any  $A \in \mathbb{C}^{m \times n}$  PENROSE defined the following equations in X:

(1) 
$$AXA = A$$
, (2)  $XAX = X$ , (3)  $(AX)^* = AX$ , (4)  $(XA)^* = XA$ .

For a subset S of the set  $\{1, 2, 3, 4\}$  the set of matrices obeying the conditions represented in S will be denoted by  $A\{S\}$ . A matrix G in  $A\{S\}$  is called an Sinverse of A and denoted by  $A^{(S)}$ . In particular, the set  $A\{1, 2, 3, 4\}$  consists of a single element, the MOORE-PENROSE inverse of A, denoted by  $A^{\dagger}$ . The set of  $\{2,3\}$  and  $\{2,4\}$ -inverses of a given rank 0 < s < r is denoted by  $A\{2,3\}_s$  and  $A\{2,4\}_s$ , as in [5], [6] and [9].

<sup>2000</sup> Mathematics Subject Classification: 15A09

Keywords and Phrases: Hyper-power method, generalized inverses, full-rank factorization.

An application of the following hyper-power method of the order 2

$$X_{k+1} = X_k(2I - AX_k) = (2I - X_k A)X_k$$

in usual matrix inversion dates back to the well-known paper of SCHULZ [15]. BEN-ISRAEL and COHEN shown that this iterative process converges to  $A^{\dagger}$  provided that  $X_0 = \alpha A^*$ , where  $\alpha$  is a positive and sufficiently small real number [2], [3], [4]. The hyper-power iterative method of an arbitrary order  $q \ge 2$  was originally devised by ALTMAN [1] for inverting a nonsingular bounded operator in a BANACH space. In [11] the convergence of the same method is proved under the condition which is weaker than the one assumed in [1], and better error estimates are derived.

ZLOBEC in [21] defined two hyper-power iterative methods of an arbitrary high order  $q \ge 2$ :

(1.1)  

$$Y_{0} = \alpha A^{*},$$

$$T_{k} = I - Y_{k}A,$$

$$M_{k} = I + T_{k} + \dots + T_{k}^{q-1}$$

$$Y_{k+1} = M_{k}Y_{k}, \quad k = 0, 1, \dots$$

(1.1')  

$$Y'_{0} = \alpha A^{*},$$

$$T_{k}' = I - AY_{k}',$$

$$M'_{k} = I + T'_{k} + \dots + T'_{k}^{q-1}$$

$$Y_{k+1}' = Y_{k}'M'_{k}, \quad k = 0, 1, \dots$$

It is well known that if we take

$$0 < \alpha \le \frac{2}{\operatorname{Tr} A^* A},$$

then  $Y_k \to A^{\dagger}$  and  $Y'_k \to A^{\dagger}$  [21].

If A is  $m \times N$  complex matrix, then the process (1.1) is superior with respect to (1.1') when m > N [8].

The hyper-power iterative method of the order 2 is investigated in [16] in view of the singular value decomposition of A. Recently, this method is investigated in [11] and [13]. In [13] several error estimates of the method are investigated. In [11] the hyper-power method of the order 2 is implemented by means of parallel computing, and several acceleration procedures are introduced.

In [20] there are given necessary and sufficient conditions for the starting approximation of the hyper-power iterative method, ensuring the convergence of these methods to an arbitrary  $\{1, 2\}$ -inverse. Modifications of the hyper-power method for computing various subclasses of  $\{1, 2\}$ -inverses are introduced in [17].

In [8] are introduced two methods for computing the matrix products  $A^{\dagger}B$ and  $BA^{\dagger}$ , involving the MOORE-PENROSE inverse, where  $A \in \mathbb{C}^{m \times N}$  and  $B \in$   $\mathbb{C}^{m \times n}$  are arbitrary complex matrices with equal number of rows. The starting matrix  $Y_0$  is chosen such that

(1.2) 
$$Y_0 = A^* W A^*, \text{ for some } W \in \mathbb{C}^{m \times N} \text{ provided that} \\ \rho \left( P_{\mathcal{R}(A)} - A Y_0 \right) < 1,$$

where  $P_{\mathcal{R}(A)}$  is the orthogonal projection on the range of A. The sequence  $\{X_k\}$ , defined by the following modification of the hyper-power method:

(1.3)  

$$Y_{0} \text{ is given by (1.2),} \\
X_{0} = Y_{0}B, \\
T_{0} = I - Y_{0}A, \\
M_{k} = I + T_{k} + T_{k}^{2} + \ldots + T_{k}^{q-1}, \\
X_{k+1} = M_{k}X_{k}, \\
T_{k+1} = T_{k}^{q} = I + M_{k}[T_{k} - I].$$

converges to  $A^{\dagger}B$  [8].

In [8] it is shown that (1.3) is an improvement (over using (1.1) to find  $A^{\dagger}$  and then forming  $A^{\dagger}B$ ) only when N > n.

In [19] we develop an iterative method for computing the *best approximate* solution and the *basic solution* of a given system of linear equations. This method is an adaptation of the modified hyper-power method (1.3). In this paper we introduce several modifications of the iterative process (1.3), applicable in computing  $\{1, 2, 3\}$ ,  $\{1, 2, 4\}$  and  $\{2, 3\}$ ,  $\{2, 4\}$  generalized inverses of a given rank.

In the second section we introduce representations for  $\{1, 2, 3\}$  and  $\{1, 2, 4\}$  inverses of a given complex matrix, in terms of matrix products involving the MOORE-PENROSE inverse. We also restate usual representations for  $\{2, 3\}$  and  $\{2, 4\}$ -inverses from [5], [6] and [9].

In view of these representations, we propose several modifications of the hyper-power method (1.3), which can be used in computation of  $\{2,3\}$ ,  $\{1,2,3\}$  and  $\{2,4\}$ ,  $\{1,2,4\}$ -inverses. Methods have arbitrary high order  $q \ge 2$ . Representations for  $\{i, j, k\}$  inverses of a matrix of rank 1 are also investigated. Introduced methods can be considered as a continuation of the papers [8] and [19].

In the third section we describe main implementation details in the package MATHEMATICA and present an illustrative example.

## 2. ITERATIVE METHODS FOR COMPUTING $\{i, j, k\}$ INVERSES

The following representations for  $\{2,3\}$   $\{2,4\}$ -inverses are restated from [5, p. 56–58], [6, p. 47–48] and [9].

**Proposition 2.1.** Let  $A \in \mathbb{C}_r^{m \times N}$  and 0 < s < r be a chosen integer. Then the following is valid:

- (a)  $A\{2,4\}_s = \{(W_2A)^{\dagger}W_2 : W_2 \in \mathbb{C}^{s \times m}, W_2A \in \mathbb{C}^{s \times N}_s\}.$
- (b)  $A\{2,3\}_s = \{W_1(AW_1)^{\dagger}: W_1 \in \mathbb{C}^{N \times s}, AW_1 \in \mathbb{C}_s^{m \times s}\}.$

In the following theorem we investigate similar representations of  $\{1, 2, 3\}$ and  $\{1, 2, 4\}$ -inverses, in terms of matrix products involving the MOORE-PENROSE inverse.

**Theorem 2.1.** Let  $A \in \mathbb{C}_r^{m \times N}$  and A = PQ be a full-rank factorization of A. Then the following statements about the sets  $A\{1,2,3\}$  and  $A\{1,2,3\}$  are valid:

- (a) The set of  $\{1, 2, 4\}$ -inverses of A can be represented as follows:
  - $A\{1,2,4\} = \left\{ (W_2 A)^{\dagger} W_2 : \quad W_2 \in \mathbb{C}^{r \times m}, \quad W_2 P \text{ is invertible} \right\}.$
- (b) The set of  $\{1, 2, 3\}$ -inverses of A can be represented as follows:
  - $A\{1,2,3\} = \{W_1(AW_1)^{\dagger}: W_1 \in \mathbb{C}^{N \times r}, QW_1 \text{ is invertible}\}.$
- (c) Particularly,

$$A^{\dagger} = (P^*A)^{\dagger}P^* = Q^*(AQ^*)^{\dagger}.$$

**Proof.** (a) Consider an arbitrary matrix  $W_2 \in \mathbb{C}^{r \times m}$ , such that  $W_2 P$  is invertible. Since the matrix  $X = (W_2 A)^{\dagger} W_2$  is  $\{2, 4\}$  inverse of A, we must to verify the equation AXA = A. We use the following important property of the MOORE– PENROSE inverse [7]:  $(UV)^{\dagger} = V^{\dagger}U^{\dagger}$  if and only if both of the following two conditions are satisfied

(2.1) 
$$U^{\dagger}UVV^{*}U^{*} = VV^{*}U^{*}, \quad VV^{\dagger}U^{*}UV = U^{*}UV.$$

The matrix  $U = W_2 P$  is invertible and V = Q is the right invertible. So the conditions (2.1) are satisfied in this case, and we get

$$AXA = PQ(W_2PQ)^{\dagger}W_2PQ = PQQ^{\dagger}(W_2P)^{\dagger}W_2PQ = PQ = A$$

In this way,  $X \in A\{1, 2, 4\}$ .

On the other hand, consider an arbitrary matrix  $X \in A\{1, 2, 4\}$ . Using the general representation of  $\{1, 2, 4\}$ -inverses from [14], and [18], we conclude that X can be represented in the form

$$X = Q^* (QQ^*)^{-1} (YP)^{-1} Y = Q^{\dagger} (YP)^{-1} Y, \quad Y \in \mathbb{C}_r^{r \times m},$$

where A = PQ is a full-rank factorization of A. Since the conditions (2.1) are satisfied for U = YP, V = Q, we get

$$X = (YA)^{\dagger}Y \in \left\{ (W_2A)^{\dagger}W_2 : W_2 \in \mathbb{C}^{r \times m}, W_2P \text{ is invertible} \right\}.$$

(b) It is sufficient to apply the property (2.1) with U = P and V = QY,  $Y \in \mathbb{C}^{N \times r}$ , and the full-rank representation of  $\{1, 2, 3\}$ -inverses from [14] and [18].

(c) This part of the proof follows from the following transformations:

$$\begin{split} (P^*A)^{\dagger}P^* &= Q^{\dagger}(P^*P)^{\dagger}P^*Q^*(QQ^*)^{-1}(P^*P)^{-1}P^* = A^{\dagger}, \\ Q^*(AQ^*)^{\dagger} &= Q^*(QQ^*)^{\dagger}P^{\dagger}Q^*(QQ^*)^{-1}(P^*P)^{-1}P^* = A^{\dagger}. \end{split}$$

In order to compare general representations of  $\{1,2,3\}$  and  $\{1,2,4\}$  inverses with known results [5, p. 58] and [6, p. 48] we state the following corollary.

Corollary 2.1. Under the assumptions of Theorem 2.2 we have

$$\begin{split} A\{1,2,4\} &= \left\{ (W_2 A)^{\dagger} W_2 : \quad W_2 \in \mathbb{C}^{r \times m}, \quad W_2 P \text{ is invertible} \right\} \\ &= \left\{ (W_2 A)^{\dagger} W_2 : \quad W_2 A \in \mathbb{C}_r^{r \times N} \right\} \\ A\{1,2,3\} &= \left\{ W_1 (AW_1)^{\dagger} : \quad W_1 \in \mathbb{C}^{N \times r}, \quad QW_1 \text{ is invertible} \right\} \\ &= \left\{ W_1 (AW_1)^{\dagger} : \quad AW_1 \in \mathbb{C}_r^{m \times r} \right\}. \end{split}$$

REMARK 2.1. Sharper versions of Proposition 2.1 and Theorem 2.1 are proved in [10].

Let  $A \in \mathbb{C}_r^{m \times N}$  and 0 < s < r. Then

$$\begin{split} A\{2,4\}_s &= \left\{ (W_2 A)^{(1,4)} W_2 : \quad W_2 A \in \mathbb{C}_s^{s \times N} \right\} \\ A\{2,3\}_s &= \left\{ W_1 (A W_1)^{(1,3)} : \quad A W_1 \in \mathbb{C}_s^{m \times s} \right\} \\ A\{1,2,4\}_s &= \left\{ (W_2 A)^{(1,4)} W_2 : \quad W_2 A \in \mathbb{C}_r^{r \times N} \right\} \\ A\{1,2,3\}_s &= \left\{ W_1 (A W_1)^{(1,3)} : \quad A W_1 \in \mathbb{C}_r^{m \times r} \right\} \end{split}$$

However, in this paper we use representations from Proposition 2.1 and Theorem 2.1, because the hyper-power method computes the MOORE-PENROSE inverse.

Introduced representations of generalized inverses are convenient for the application of the modified hyper-power iterative method (1.3). Using this idea, we introduce two modifications of the hyper-power method for construction of  $\{1, 2, 3\}$  and  $\{1, 2, 4\}$  generalized inverses, and two modifications of the hyper-power method for computing subsets of  $\{2, 3\}$  and  $\{2, 4\}$ -inverses. In these algorithms we consider an arbitrary matrix A of the order  $m \times N$ . Also, it is assumed that rank  $A = r \ge 2$  and  $q \ge 2$  is any integer. Algorithm A24 can be used in construction of  $\{2, 4\}$ -inverses, and Algorithm A23 can be used in construction of  $\{2, 3\}$ -inverses.

# Algorithm A24.

(2.1)  

$$Y_{0} = (W_{2}A)^{*}W(W_{2}A)^{*}, \text{ for some } W_{2} \in \mathbb{C}^{s \times N} \text{ such that}$$

$$\rho \left(P_{\mathcal{R}(W_{2}A)} - W_{2}AY_{0}\right) < 1, \quad 1 < s \leq \operatorname{rank}A, \quad W_{2}A \in \mathbb{C}_{s}^{s \times N}$$

$$X_{0} = Y_{0}W_{2},$$

$$T_{0} = I - Y_{0}W_{2}A,$$

$$M_{k} = I + T_{k} + T_{k}^{2} + \dots + T_{k}^{q-1},$$

$$X_{k+1} = M_{k}X_{k} = (I + T_{k} + T_{k}^{2} + \dots + T_{k}^{q-1})X_{k},$$

$$T_{k+1} = T_{k}^{q} = I + M_{k}[T_{k} - I],$$

$$k = 0, 1, \dots$$

This algorithm is an improvement (over using a modification of (1.1) to find  $(W_2A)^{\dagger}$ and then forming  $(W_2A)^{\dagger}W_2$ ) only in the case N > m.

# Algorithm A23.

(2.3)  

$$Y_{0} = (AW_{1})^{*}W(AW_{1})^{*}, \text{ for some } W_{1} \in \mathbb{C}^{m \times s} \text{ such that}$$

$$\rho \left(P_{\mathcal{R}(AW_{1})} - AW_{1}Y_{0}\right) < 1, \quad 1 < s \leq \operatorname{rank} A, \quad AW_{1} \in \mathbb{C}^{m \times s}_{s}$$

$$X_{0} = W_{1}Y_{0},$$

$$T_{0} = I - AW_{1}Y_{0},$$

$$M_{k} = I + T_{k} + T_{k}^{2} + \ldots + T_{k}^{q-1},$$

$$X_{k+1} = X_{k}M_{k} = X_{k}(I + T_{k} + T_{k}^{2} + \cdots + T_{k}^{q-1}),$$

$$T_{k+1} = T_{k}^{q} = I + M_{k}[T_{k} - I],$$

$$k = 0, 1, \ldots$$

This algorithm is an improvement (over using a modification of (1.1) to find  $(AW_1)^{\dagger}$ and then forming  $W_1(AW_1)^{\dagger}$ ) in the case m > N.

REMARK 2.2. Instead of the initial approximations  $Y_0$ , used in (2.2) and (2.3), we can use the following approximations:

(2.2') 
$$Y_0 = \alpha (W_2 A)^*, \qquad 0 < \alpha \le \frac{2}{\operatorname{Tr} ((W_2 A)^* W_2 A)},$$

(2.3') 
$$Y_0 = \alpha (AW_1)^*, \quad 0 < \alpha \le \frac{2}{\operatorname{Tr} ((AW_1)^* AW_1)}.$$

Initial approximations (2.2) and (2.3) are more general, but (2.2') and (2.3') are simpler for computation.

**Theorem 2.2.** For an arbitrary matrix  $A \in \mathbb{C}_r^{m \times N}$ , any integer  $1 < s \leq r$  and arbitrary matrices  $W_2 \in \mathbb{C}^{s \times m}$ ,  $W_1 \in \mathbb{C}^{N \times s}$  the following statements are valid:

(a) In general, the sequence  $X_k$ , k = 0, 1, ..., defined in Algorithm A24 converges to

(2.4) 
$$X_k \to (W_2 A)^{\dagger} W_2 \in A\{2,4\}_s \text{ if and only if } W_2 A \in \mathbb{C}_s^{s \times N}$$

(b) In the case s = r the sequence  $X_k$ , k = 0, 1, ..., defined in Algorithm A24 satisfies

(2.4') 
$$X_k \to (W_2 A)^{\dagger} W_2 \in A\{1, 2, 4\}$$
 if and only if  $W_2 P$  is invertible

(c) The optimal order q of methods (a) and (b) minimizes the function

$$f(q) = (m/N + q - 1)/\ln q.$$

(d) In general, the sequence  $X_k$ , k = 0, 1, ..., defined in Algorithm A23 satisfies

(2.5) 
$$X_k \to W_1(AW_1)^{\dagger} \in A\{2,3\}_s \text{ if and only if } AW_1 \in \mathbb{C}_s^{m \times s}.$$

(e) In the case s = r the sequence  $X_k$ , k = 0, 1, ..., defined in Algorithm A23 satisfies

(2.5') 
$$X_k \to W_1(AW_1)^{\dagger} \in A\{1,2,3\}$$
 if and only if  $QW_1$  is invertible.

(f) The optimal order q of methods in (d) and (e) minimizes the function

$$f(q) = (N/m + q - 1)/\ln q.$$

**Proof.** (a), (b) It is not difficult to verify

$$X_k = Y_k W_2, \quad k = 0, 1, \dots$$

where the sequence  $\{Y_k\}$  is defined as in the following:

$$Y_{0} = \alpha(W_{2}A)^{*}, \quad 0 < \alpha \leq \frac{2}{\operatorname{Tr}\left((W_{2}A)^{*}W_{2}A\right)}, \quad W_{2}A \in \mathbb{C}_{s}^{s \times N}$$

$$T_{0} = I - Y_{0}W_{2}A,$$

$$M_{k} = I + T_{k} + T_{k}^{2} + \dots + T_{k}^{q-1},$$

$$Y_{k+1} = M_{k}Y_{k} = (I + T_{k} + T_{k}^{2} + \dots + T_{k}^{q-1})Y_{k},$$

$$T_{k+1} = T_{k}^{q} = I + M_{k}[T_{k} - I],$$

$$k = 0, 1, \dots$$

Since the sequence  $\{Y_k\}$  is defined by applying the usual hyper-power method (1.1) on the matrix  $W_2A$ , we conclude  $Y_k \to (W_2A)^{\dagger}$ . Hence, we get  $X_k \to (W_2A)^{\dagger}W_2$ . Then statements (2.4) and (2.4') follows from Proposition 2.1 and Theorem 2.1, respectively.

(c) The optimal order q can be determined using the known results from [8]. The parts (d), (e) and (f) can be proved in a similar way.

In the case rank A = 1 the set of  $\{1, 2, 3\}$  and  $\{1, 2, 4\}$ -inverses can be generated using the next known proposition from [17]:

$$A^{\dagger} = \frac{1}{\operatorname{Tr}\left(A^*A\right)} A^*.$$

**Corollary 2.2.** If A is  $m \times N$  matrix satisfying rank A = 1, then the following statements are valid:

(a)  $X = W_1(AW_1)^{\dagger} \in A\{1, 2, 3\}$  is given by

$$X = \frac{1}{Tr((AW_1)^*AW_1)} W_1(AW_1)^*.$$

(b) 
$$Y = (W_2 A)^{\dagger} W_2 \in A\{1, 2, 4\}$$
 is given by  

$$Y = \frac{1}{Tr((W_2 A)^* W_2 A)} (W_2 A)^* W_2.$$

## **3. IMPLEMENTATION METHOD**

In this section we describe implementation details of the introduced algorithms, in the package MATHEMATICA. In the following function norm [a] we  $\int m$ m

compute the norm 
$$\sqrt{\sum_{i=1}^{m} \sum_{j=1}^{n} a_{ij}^2}$$
 of the matrix  $a$ .

```
norm[a_]:
          Block[{m,n,i,j,u},
      {m,n\}=Dimensions[a];
     u=Sum[a[[i,j]]^2,{i,m},{j,n}];
     Return[N[Sqrt[u],20]]
]
```

In the function HyperPower 124 is implemented Algorithm A24.

```
(* A^(2,4)=(W2A)^+ W2 *)
HyperPower124[a_,w2_,q_,eps_]:
         Block[{tk,tk1,c=w2.a,e,x0,x1,y,alpha,s,k=1,m,n,nor=1},
         alpha=2/trace[Transpose[Conjugate[c]].c];
         y=alpha Conjugate[Transpose[c]];
         x0=y.w2;
         e=IdentityMatrix[n];
         tk=s=e-y.c;
         While[nor>=eps,
               s=e; tk1=tk; Do[s+=tk1; tk1=tk1.tk,{i,q-1}];
               x1=s.x0;
```

```
tk=tk1;
nor=norm[x1-x0];
x0=x1; k=k+1
];
N[x1]
```

]

In the function HyperPower 123 is implemented Algorithm A23.

```
(* A^ {}(2,3)=W1(AW1)^ {}+ *)
HyperPower123[a_,w1_,q_,eps_]:
         Block[{tk,tk1,c=a.w1,e,x0,x1,y,alpha,s,k=1,m,n,nor=1},
         alpha=2/trace[Transpose[Conjugate[c]].c];
         y=alpha Conjugate[Transpose[c]];
         x0=w1.y;
         e=IdentityMatrix[n];
         tk=s=e-c.y;
         While[nor>=eps,
               s=e; tk1=tk; Do[s+=tk1; tk1=tk1.tk,\{i,q-1\}];
               x1=x0.s;
               tk=tk1;
               nor=norm[x1-x0];
               x0=x1; k=k+1
              ];
         N[x1]
   ]
```

EXAMPLE 3.1. In this example we construct  $\{1, 2, 4\}$  and  $\{1, 2, 3\}$ -inverse of the matrix

$$A = \begin{vmatrix} -1 & 0 & 1 & 2 \\ -1 & 1 & 0 & -1 \\ 0 & -1 & 1 & 3 \\ 0 & 1 & -1 & -3 \\ 1 & -1 & 0 & 1 \\ 1 & 0 & -1 & -2 \end{vmatrix}$$

which are generated, respectively, by the matrices

$$(3.1) W_2 = \begin{bmatrix} 3 & 1 & 3 & 1 & 2 & -1 \\ 0 & -1 & 0 & 0 & -2 & 1 \end{bmatrix}, W_1 = \begin{bmatrix} 2 & 0 \\ 0 & 1 \\ 1 & 0 \\ 4 & 2 \end{bmatrix}.$$

By means of the expression HyperPower124[ $a, w2, 3, 10^{(-11)}$ ] we apply the Algorithm A24 for the matrices  $A, W_2$ , using the order q = 3 and the precision  $10^{-11}$ . In this case is  $s = \operatorname{rank} A = 2$ , and the resulting matrix

Γ-	-0.647059	0.843137	-0.647059	-0.215686	1.68627	-0.843137
	0.411765	-0.627451	0.411765	0.137255	-1.2549	0.627451
	0.235294	-0.215686	0.235294	0.0784314	-0.431373	0.215686
L	0.0588235	0.196078	0.0588235	0.0196078	0.392157	-0.196078

is an  $\{1, 2, 4\}$ -inverse of A.

Similarly, by means of the expression  $HyperPower123[a, w1, 3, 10^{(-11)}]$  we apply the *Algorithm* A23 for the matrices  $A, W_1$ , using the modified Hyper-power method of the order 3, with the precision  $10^{-11}$ . The resulting  $\{1, 2, 3\}$ -inverse of A is equal to

-0.117647	-0.176471	0.0588235	-0.0588235	0.176471	0.117647	
0.186275	0.196078	-0.00980392	0.00980392	-0.196078	-0.186275	
-0.0588235	-0.08882353	0.0294118	-0.0294118	0.0882353	0.0588235	1
0.137255	0.0392157	0.0980392	-0.0980392	-0.0392157	-0.137255	

Consider now the following matrix of rank 3

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

and the matrices  $W_1$ ,  $W_2$  as in (3.1). In this case is  $s = 2 < \operatorname{rank} A = 3$ , and the expression  $HyperPower124[a, w2, 4, 10^{(-12)}]$  produces the following  $\{2, 4\}$ -inverse of A:

_					_	
-0.205821	0.243243	-0.205821	-0.0686071	0.486486	-0.243243	
-0.380457,	0.540541	-0.380457	-0.126819	1.08108	-0.540541	
0.0997921	-0.027027	0.0997921	0.033264	-0.0540541	0.027027	
0.0935551	0.162162	0.0935551	0.031185	0.324324	-0.162162	

Also, the result of the expression  $HyperPower123[a, w1, 4, 10^{(-12)}]$  is the following  $\{2, 3\}$ -inverse of A:

-0.038633	-0.170877	0.0401189,	-0.0401189	0.0787519	0.038633	
0.0903913	0.20109	0.0151065	-0.0151065	-0.0752848	-0.0903913	
-0.0193165	-0.0854383	0.0200594	-0.0200594	0.0393759	0.0193165	·
0.103517	0.060426	0.110451	-0.110451	0.00693413	-0.103517	

## 4. CONCLUSION

In this section we present a few concluding remarks and comparisons of the introduced method with the modification of the hyper-power method introduced in [17].

REMARK 4.1. We point out the following advantages of defined algorithms with respect to modifications of the hyper-power method which are introduced in [17]:

- (a) In the iterations defined in this paper it is not necessary to multiply the matrix  $M_k X_k$  (or the matrix  $X_k M_k$ ) by the matrices  $W_1$  and  $W_2$  from the left and right, respectively.
- (b) Iterations defined in [17] require a full-rank factorization of A in computation of  $\{1, 2, 3\}$  and  $\{1, 2, 4\}$ -inverses. On the other hand, the iterations (2.2) and (2.3) do not require the full-rank factorization.

In order to demonstrate these advantages we consider, in parallel, iterations introduced in this paper and in [17].

By means of Algorithm A24, the sets  $A\{2,4\}_s$  and  $A\{1,2,4\}$  can be generated as follows:

$$Y_0 = \alpha (W_2 A)^*, \quad 0 < \alpha \le \frac{2}{\text{Tr} ((W_2 A)^* W_2 A)}, \quad W_2 P \text{ is invertible},$$
  

$$X_0 = Y_0 W_2, \quad T_0 = I - Y_0 W_2 A,$$
  

$$X_{k+1} = (I + T_k + T_k^2 + \dots + T_k^{q-1}) X_k,$$
  

$$T_{k+1} = T_k^q, \quad k = 1, \dots.$$

Applying iterations from [17], Lemma 2.1], we must compute a full-rank factorization A = PQ and then generate the following iterations:

$$Y_0 = \alpha (W_2 A Q^*)^*, \qquad 0 < \alpha \le \frac{2}{\operatorname{Tr} \left( (W_2 A Q^*)^* W_2 A Q^* \right)}$$
$$T_k = I - Y_k W_2 A Q^* = T_{k-1}^q$$
$$Y_{k+1} = (I + T_k + T_k^2 + \dots + T_k^{q-1}) Y_k,$$
$$X_{k+1} = W_1 Y_{k+1} W_2, \quad k = 0, 1, \dots$$

Similarly, by means of Algorithm A23, the set  $A\{1, 2, 3\}$  can be generated as follows:

$$Y_{0} = \alpha (AW_{1})^{*}, \quad 0 < \alpha \leq \frac{2}{\operatorname{Tr} ((AW_{1})^{*}AW_{1})}, \quad QW_{1} \text{ is invertible}, X_{0} = W_{1}Y_{0}, \quad T_{0} = I - AW_{1}Y_{0}, X_{k+1} = X_{k}(I + T_{k} + T_{k}^{2} + \dots + T_{k}^{q-1}), T_{k+1} = T_{k}^{q}, \quad k = 0, 1, \dots$$

Also, using the method defined in [17, Lemma 2.1], we must compute a full-rank factorization A = PQ and generate the following iterations:

$$Y_0 = \alpha (P^* A W_1)^*, \qquad 0 < \alpha \le \frac{2}{\text{Tr} \left( (P^* A W_1)^* P^* A W_1 \right)}$$
$$T_k = I - P^* A W_1 Y_k = T_{k-1}^q$$
$$Y_{k+1} = Y_k (I + T_k + T_k^2 + \dots + T_k^{q-1}),$$
$$X_{k+1} = W_1 Y_{k+1} W_2, \quad k = 0, 1, \dots.$$

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(Received January 27, 2000)