# FULL RANK DECOMPOSITION OF MATRICES AND ITS APPLICATION OVER SKEW FIELDS 

Junqing Wang*, Xin Fan and Zhang Jingling<br>Department of Mathematics, Dezhou University, Shandong 253055, PR China.

(Received on: 05-12-13; Revised \& Accepted on: 18-12-13)


#### Abstract

Reference full rank factorization theory over fields, this paper will extend the theory of full rank decomposition of matrices over fields to skew fields. This paper also gives some applications, such as the proof of classical inequality and the compute of the Drazin inverse and $M-P$ inverse.


Keywords: Full rank decomposition; Inequality; Drazin inverse; $M-P$ inverse.

## 1. INTRODUCTION

A matrix is expressed by the multiplication of a number of simple structural matrices, which is called the decomposition of matrices. It is important for us to reveal the full rank decomposition and using this theory to solve some problem of matrices. This paper expands the theory of full rank decomposition of matrices. Using this theory to certify Forbeniua inequality and Sylvester inequality and calculate the Drazin inverse and $M-P$ inverse.

In this paper, $K$ denotes a skew field, $K^{m \times n}$ represents the set of $m \times n$ order matrix, $M_{n}(K)$ represents the set of $n \times n$ order matrix over $K, G L_{n}(K)$ means the all of the inverse matrix of $n$ order , $A^{*}$ means the conjugate transpose matrix of $A$.

## 2. PREPARATION KNOWLEDGE

Definition: $1^{[1]}$ Let $R$ is a non-commutative principal ideal domian with involutorial anti-automorphism $\sigma, A \in R^{m \times n}$, if there exits $X \in R^{n \times m}$ such that:

$$
A X A=A, X A X=X,(A X)^{*}=A X,(X A)^{*}=X A
$$

Then, we say $X$ is a Moore - Penrose inverse of $A$, denoted $A^{+}$.
Definition: $2^{[1]}$ Let $A \in M_{n}(K)$, if there is a $X \in M_{n}(K)$ such that:

$$
A^{k}=A^{k+1} X, X=X^{2} A, A X=X A
$$

Then, we say $X$ is a $\operatorname{Drazin}$ inverse of $A$, denoted $A^{D}$.
Lemma: $1^{[2]}$ Let $D$ is a non-singular matrix, $\left[\begin{array}{ll}A & B \\ C & D\end{array}\right] \in M_{n}(K)$, then

$$
r\left[\begin{array}{ll}
A & B \\
C & D
\end{array}\right]=r(D)+r\left(A-B D^{-1} C\right)
$$

*Corresponding author: Junqing Wang*
Department of Mathematics, Dezhou University, Shandong 253055, PR China.

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Lemma: $2^{[1]}$ Let $A \in K^{m \times n}, B \in K^{t \times m}, C \in K^{s \times n}$ and $\operatorname{rank} A=n, \operatorname{rank} B=m, \operatorname{rank} C=s$, then,

$$
\operatorname{rank} A=\operatorname{rank} B A=\operatorname{rank}^{2} C^{T}=\operatorname{rank}^{2} A C^{T}
$$

Lemma: $3^{[1]}$ Let $A \in K^{m \times n}$, if there is $T \in M_{n}(K), S \in M_{n}(K)$, then

$$
\operatorname{rank} A=\operatorname{rank}(T A)=\operatorname{rank}(A S)=\operatorname{rank}(T A S)
$$

## 3. MAIN CONCLUSION

Theorem: 1 If $A \in K^{m \times n}, \quad \operatorname{rank} A=r(r>0)$, then $A=C D$, where $C \in K^{m \times r}$ is a row full rank matrix, $D \in K^{r \times n}$ is a column full rank matrix.

Proof: Because rankA $r$, by the series of elementary row transformations, we can transfer $A$ into a row stepformed matrix $B$, where $B=\binom{D}{0}, D \in K^{r \times n}$, and rank $A=r$.. Then exists a product of many elementary matrix, such that $P A=B$, namely $A=P^{-1} B$, we write $P^{-1}=\left(\begin{array}{ll}C & M\end{array}\right), C \in K^{m \times r}, M \in K^{m \times(m-r)}$, then $\mathrm{A}=\left(\begin{array}{ll}C & M\end{array}\right)\binom{D}{0}=C D, C$ is a $m \times r$ column full rank matrix, $D$ is a $r \times n$ row full rank matrix. So $A=C D$ is a column full rank decomposition of $A$.

Theorem: 2 Let $A \in K^{m \times n}$, rank $A=r(r>0)$, if $A$ is a line-simplify matrix, the column $i_{1}, i_{2} \cdots i_{r}$ of matrix $D$ in $B=\binom{D}{0}$ is the first $r$ columns of the identify matrix $I_{r}$ in the proper order. We set $C=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)$, then $A=C D$ is the full rank decomposition of $A$.

Proof: Actually, finding the full rank decomposition of $A$ is to find the maximal linearly independent group of column vector group $\alpha_{1}, \alpha_{2} \cdots \alpha_{n}$. Let $C=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)$, then, it is a $m \times r$ column full rank matrix. As to any column vector $\alpha_{i j}$, we have $\alpha_{i j}=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\left(\begin{array}{c}k_{1 j} \\ k_{2 j} \\ \vdots \\ k_{r j}\end{array}\right)\right.$,
Note $K=\left[k_{i j}\right]_{r \times(n-r)}$, and take $n$ order permutation matrix $\left(\varepsilon_{i_{1}}, \varepsilon_{i_{2}} \cdots \varepsilon_{i_{n}}\right)$, then $A=\left(\alpha_{1}, \alpha_{2} \cdots \alpha_{n}\right)=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)\left(I_{k}, K\right)=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)\left[\left(I_{k}, K\right) P_{1}\right]$. We let permutation matrix $P_{1}=\left(\varepsilon_{i_{1}}, \varepsilon_{i_{2}} \cdots \varepsilon_{i_{n}}\right)^{-1}=\left(\varepsilon_{i_{1}}, \varepsilon_{i_{2}} \cdots \varepsilon_{i_{n}}\right)^{H} . D=\left(I_{r}, K\right) P_{1}$ is a $r \times n$ row full rank matrix.

For $D=\left(I_{r}, K\right) P_{1}$ we consider it as follows: according to a series of element row operation, that is there is a universe matrix $P$, we can transform $A$ into $P A=B$, where $B=\left(\beta_{1}, \beta_{2} \cdots \beta_{n}\right)$,that is $\alpha_{i}=P^{-1} \beta_{i}$. We know $\left(\alpha_{1}, \alpha_{2} \cdots \alpha_{n}\right)$ and $\left(\beta_{1}, \beta_{2} \cdots \beta_{n}\right)$ have the same pertinence relativity pertinence. So for $A=P^{-1} B$, we have $B=\left(\beta_{1}, \beta_{2}, \cdots \beta_{n}\right)=\left(\beta_{i_{1}}, \beta_{i_{2}} \cdots \beta_{i_{r}}\right)\left(I_{r}, K\right) P$.

So according to a series of element row operations, we can obtain a matrix $B$ as referred in the theory and $B=\binom{I_{r}}{0}\left(I_{r}, 0\right) P_{1}=\binom{\left(I_{r}, K\right) P_{1}}{0}=\binom{D}{0} . D=\left(I_{r}, K\right) P_{1}$ is a $r \times n$ row full rank matrix. We take the maximal linearly independent column group $\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)$ of $A$ to form a $m \times r$ column full rank matrix $C=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)$. Namely $A=\left(\alpha_{i_{1}}, \alpha_{i_{2}} \cdots \alpha_{i_{r}}\right)\left[\left(I_{k}, K\right) P_{1}\right]=C D$.

Example: 1 Let $A=\left[\begin{array}{llll}1 & i & j & k \\ 0 & i & 0 & k \\ 1 & 0 & j & 0\end{array}\right]$, compute the column rank decomposition of it. By the series of elementary row transformations, we can transfer $A$ into a row step-formed matrix $B$. we records the product of elementary matrix.

$$
\begin{aligned}
& {\left[A, I_{3}\right] } \rightarrow\left[\begin{array}{lllllll}
1 & i & j & k & 1 & 0 & 0 \\
0 & i & 0 & k & 0 & 1 & 0 \\
1 & 0 & j & k & 0 & 0 & 1
\end{array}\right] \rightarrow\left[\begin{array}{ccccccc}
1 & i & j & k & 1 & 0 & 0 \\
0 & i & 0 & k & 0 & 1 & 0 \\
0 & -i & 0 & -k & -1 & 0 & 1
\end{array}\right] \\
& \rightarrow\left[\begin{array}{lllllll}
1 & i & j & k & 1 & 0 & 0 \\
0 & i & 0 & k & 0 & 1 & 0 \\
0 & 0 & 0 & 0 & -1 & 1 & 1
\end{array}\right]=[B, P] \\
& B=\left[\begin{array}{llll}
1 & i & j & k \\
0 & i & 0 & k \\
0 & 0 & 0 & 0
\end{array}\right], P=\left[\begin{array}{ccc}
1 & 0 & 0 \\
0 & 1 & 0 \\
-1 & 1 & 1
\end{array}\right], P^{-1}=\left[\begin{array}{ccc}
1 & 0 & 0 \\
0 & 1 & 0 \\
1 & -1 & 1
\end{array}\right]
\end{aligned}
$$

According to the theorem 1, we can obtain
$A=C D=\left[\begin{array}{cc}1 & 0 \\ 0 & 1 \\ 1 & -1\end{array}\right]\left[\begin{array}{llll}1 & i & j & k \\ 0 & i & 0 & k\end{array}\right]$.
Example: 2 Let $A=\left[\begin{array}{cccc}i & j & k & 0 \\ 0 & j & 0 & -i \\ i & 0 & k & j\end{array}\right]$, compute the column rank decomposition of it. According to the method of theorem 2,
$A=\left[\begin{array}{cccc}i & j & k & 0 \\ 0 & j & 0 & -i \\ i & 0 & k & j\end{array}\right] \rightarrow\left[\begin{array}{cccc}i & j & k & 0 \\ 0 & j & 0 & -i \\ 0 & 0 & 0 & 0\end{array}\right] \rightarrow\left[\begin{array}{cccc}1 & -k & j & 0 \\ 0 & 1 & 0 & k \\ 0 & 0 & 0 & 0\end{array}\right] \rightarrow\left[\begin{array}{cccc}1 & 0 & j & -1 \\ 0 & 1 & 0 & k \\ 0 & 0 & 0 & 0\end{array}\right]$
Then, $A=\left[\begin{array}{ll}i & j \\ 0 & j \\ i & 0\end{array}\right]\left[\begin{array}{cccc}1 & 0 & j & -1 \\ 0 & 1 & 0 & k\end{array}\right]=C D$.
Theorem: $\mathbf{3}$ Let $A \in M_{n}(K)$, and $\operatorname{rank} A=\operatorname{rank}\left(I_{n}-A\right)=n$,then $A$ is a idempotent matrix.
Proof: In order to certify $A^{2}=A$, only $r\left(A^{2}-A\right)=0$. Let $r(A)=r$, then $r-n+r\left(I_{n}-A\right)=0$. The full rank decomposition of $A=H L$, where $H \in K^{m \times r}$ is a row full rank matrix, $L \in K^{r \times n}$ is a column full rank matrix.

Then according to lemma 1 :

$$
\begin{aligned}
r\left(A^{2}-A\right) & =r(H L H L-H L)=r\left[H\left(I_{r}-L I_{n} H\right) L\right]=r\left(I_{r}-L I_{n} H\right) \\
& =\left[\begin{array}{cc}
I_{n} & H \\
L & I_{r}
\end{array}\right]-r\left(I_{n}\right)=\left[\begin{array}{cc}
I_{n}-H L & H \\
0 & I_{r}
\end{array}\right]-n \\
& =r+r\left[I_{n}-H L\right]-n=r-n+r\left[I_{n}-A\right]=0
\end{aligned}
$$

So, $A$ is a idempotent matrix.

Theorem: 4 (Sylvester inequality) Let $A \in K^{m \times n}, B \in K^{n \times t}$, then,

$$
\operatorname{rank}(A B) \geq \operatorname{rank} A+\operatorname{rank} B-n
$$

Proof: Let rank $A=r(r>0)$, let $A=H L$, where $H \in K^{m \times r}$ is a row full rank matrix, $L \in K^{r \times n}$ is a column full rank matrix. Then there is $L \in K^{(n-r) \times n}$ such as $\binom{L}{L_{1}} \in M_{n}(K)$, where $r a n k L_{1}=n-r$.
According to lemma 2: $\operatorname{rank}(A B)=\operatorname{rank}(H L B)=\operatorname{rank}(L B)=\operatorname{rank}\left[\binom{L B}{L_{1} B}-\binom{0}{L_{1} B}\right]$. According to lemma 3, we know:

$$
\begin{aligned}
\operatorname{rank}(A B) & \geq \operatorname{rank}\left[\binom{L}{L_{1}} B\right]-\operatorname{rank}\left(L_{1} B\right)=\operatorname{rank} B-\operatorname{rank}\left(L_{1} B\right) \\
& \geq \operatorname{rank} B-\operatorname{rank} L_{1}=\operatorname{rank} B-(n-r)=\operatorname{rank} A+\operatorname{rank} B-n .
\end{aligned}
$$

Theorem: 5 (Forbeniua inequality) Let $A \in K^{m \times n}, B \in K^{n \times t}, C \in K^{t \times s}$ then,

$$
\operatorname{rank}(A B C) \geq \operatorname{rank}(A B)+\operatorname{rank}(B C)-B
$$

Proof: Let rank $B=r(r>0)$, let $B=H L$, where $H \in K^{m \times r}$ is a row full rank matrix, $L \in K^{r \times n}$ is a column full rank matrix. According to the theorem 4,

$$
\operatorname{rank}(A B C)=\operatorname{rank}(A H)(L C) \geq \operatorname{rank}(A H)+\operatorname{rank}(L C)-r
$$

According to lemma 2, we can know,

$$
\begin{aligned}
& \operatorname{rank}(A H)=\operatorname{rank}(A H L)=\operatorname{rank}(A B) \\
& \operatorname{rank}(L C)=\operatorname{rank}(H L C)=\operatorname{rank}(B C)
\end{aligned}
$$

So, $\operatorname{rank}(A B C) \geq \operatorname{rank}(A B)+\operatorname{rank}(B C)-B$.
Theorem:6 Let $A \in M_{n}(K)$, let $A=H L$, where $H \in K^{m \times r}$ is a row full rank matrix, $L \in K^{r \times n}$ is a column full rank matrix. Then the existences of $A^{\#}$ if and only if $L H$ is a non-singular matrix, and $A^{\#}=H(L H)^{-2} L$.

Proof: If $A^{\#}$ exists, according to $A=A^{2} X$, we can know rank $A \leq \operatorname{rank} A^{2}$, because $r a n k A^{2} \leq r a n k A$,
So

$$
\operatorname{rank} A=\operatorname{rank} A^{2}=r .
$$

$$
r=\operatorname{rank}(H L)=\operatorname{rank}(H L H L)=\operatorname{rank} H(L H) L \leq \operatorname{rank}(H L)
$$

$\operatorname{rank}(H L) \leq \operatorname{rankL} \leq r$, so $\operatorname{rank}(L H)=r$. We can know $L H$ is a non-singular matrix.
In addition, because $L H$ is a non-singular matrix, it is meaningful to $X=H(L H)^{-2} L$. It is easy for us to certify $X$ satisfies four conditions of Drazin inverse.

The theorem have certified.
Theorem: 7 Let $A \in K^{m \times n}$, let $A=H L$, where $H \in K^{m \times r}$ is a row full rank matrix, $L \in K^{r \times n}$ is a column full rank matrix, then $A^{+}=L^{*}\left(H^{*} A L^{*}\right)^{-1} H^{*}$.

Proof: First $H^{*} A L^{*}=H^{*} H L L^{*}=\left(H^{*} H\right)\left(L L^{*}\right)$, and $H^{*} H$ and $L L^{*}$ are square matrices. So $H^{*} A L^{*}$ is a nonsingular matrix.

$$
A^{+}=L^{*}\left(H^{*} A L^{*}\right)^{-1} H^{*}=L^{*}\left(L L^{*}\right)^{-1}\left(H^{*} H\right)^{-1} H^{*}
$$

Let $X=L^{*}\left(H^{*} A L^{*}\right)^{-1} H^{*}, A X=H\left(H^{*} H\right)^{-1} H^{*}, X A=L^{*}\left(L L^{*}\right)^{-1} L$. Then,

$$
A X A=A, X A X=X,(A X)^{*}=A X,(X A)^{*}=X A
$$

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## Source of Support: Nil, Conflict of interest: None Declared

