# ON THE SOLUTIONS OF THE MATRIX EQUATION <br> $f\left(X, X^{*}\right)=g\left(X, X^{*}\right)$ <br> BY <br> P. BASAVAPPA 

It is well known that the matrix identities $X X^{*}=I, X=X^{*}$ and $X X^{*}=X^{*} X$, where $X$ is a square matrix with complex elements, $X^{*}$ is the conjugate transpose of $X$ and $I$ is the identity matrix, characterize unitary, hermitian and normal matrices respectively. These identities are special cases of more general equations of the form (a) $f\left(X, X^{*}\right)=A$ and (b) $f\left(X, X^{*}\right)=g\left(X, X^{*}\right)$, where $f(x, y)$ and $g(x, y)$ are monomials of one of the following four forms: xyxy...xyxy, xyxy...xyx, $y x y x \ldots y x y x$, and $y x y x \ldots y x y$. In this paper all equations of the form (a) and (b) are solved completely. It may be noted a particular case of $f\left(X, X^{*}\right)=A$, viz. $X X^{\prime}=A$, where $X$ is a real square matrix and $X^{\prime}$ is the transpose of $X$ was solved by Weitzenböck [3]. The distinct equations given by (a) and (b) are enumerated and solved.

Most of the terminology is standard. All the matrices are matrices of complex numbers. By a projection is meant a matrix $E$ such that $E=E^{*}=E^{2}$.

The main tools used in the solutions of the equations are: (1) the principal axis theorem for a nonhermitian matrix [1] and (2) the polar decomposition of a matrix [2]. These are stated as lemmas for later use.

Lemma 1. Let $X$ be any rectangular matrix. Then there exist unitary matrices $U$ and $V$ such that

$$
U X V=\operatorname{diag}\left(x_{1}, \ldots, x_{r}, 0, \ldots, 0\right)
$$

where $x_{1}, \ldots, x_{r}, 0, \ldots, 0$ are singular values of $X$.
Lemma 2. Let $X$ be any square matrix. Then $X$ can be written as

$$
X=H U(V K)
$$

where $H(K)$ is p.s.d. and is unique and $U(V)$ is a unitary matrix. Moreover $H(K)$ and $U(V)$ commute if and only if $X$ is normal.

Theorem 1. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p}=A, \quad p \geq 1
$$

iff

$$
X=U^{*} \operatorname{diag}\left(\alpha_{1}^{1 / 2 p}, \ldots, \alpha_{r}^{1 / 2 p}, 0, \ldots, 0\right) V^{*}
$$

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where $U$ is any unitary matrix such that

$$
U A U^{*}=\operatorname{diag}\left(\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0\right)
$$

and $V$ is any unitary matrix.
Proof. Assume $X$ is a solution of the equation. By Lemma 1 we have

$$
\begin{equation*}
U X V=\operatorname{diag}\left(x_{1}, \ldots, x_{r}, 0, \ldots, 0\right) \tag{1}
\end{equation*}
$$

Since $X$ is a solution of the equation, we get

$$
\begin{equation*}
\left((U X V)(U X V)^{*}\right)^{p}=U A U^{*} \tag{2}
\end{equation*}
$$

Therefore if $\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0$ are the characteristic roots of $A$, by making use of (1) in (2) we get

$$
\operatorname{diag}\left(x_{1}^{2 p}, \ldots, x_{r}^{2 p}, 0, \ldots, 0\right)=\operatorname{diag}\left(\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0\right)
$$

which implies that

$$
U X V=\operatorname{diag}\left(\alpha_{1}^{1 / 2 p}, \ldots, \alpha_{r}^{1 / 2 p}, 0, \ldots, 0\right)
$$

or

$$
X=U^{*} \operatorname{diag}\left(\alpha_{1}^{1 / 2 p}, \ldots, \alpha_{r}^{1 / 2 p}, 0, \ldots, 0\right) V^{*}
$$

It is easily verified $X$ in the above form satisfies the equation. Note that if $p=1$ and $X$ is a real square matrix, we get Weitzenböck's result.

Theorem 2. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p} X=A
$$

iff

$$
X=U^{*} \operatorname{diag}\left(\alpha_{1}^{1 /(2 p+1)}, \ldots, \alpha_{r}^{1 /(2 p+1)}, 0, \ldots, 0\right) V^{*}
$$

where $\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0$ are the singular values of $A$ and $U$ and $V$ are the unitary matrices such that

$$
U A V=\operatorname{diag}\left(\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0\right)
$$

Proof. Let $X$ be a solution of the equation. Let $U$ and $V$ be matrices as in Theorem 1. From the given equation, we get

$$
\left((U X V)(U X V)^{*}\right)^{p} U X V=U A V .
$$

or

$$
\operatorname{diag}\left(x_{1}^{2 p+1}, \ldots, x_{r}^{2 p+1}, 0, \ldots, 0\right)=U A V
$$

Thus if the singular values of $A$ are $\alpha_{1}, \ldots, \alpha_{r}, 0, \ldots, 0$ we get

$$
\operatorname{diag}\left(x_{1}, \ldots, x_{r}, 0, \ldots, 0\right)=\operatorname{diag}\left(\alpha_{1}^{1 /(2 p+1)}, \ldots, \alpha_{r}^{1 /(2 p+1}, 0, \ldots, 0\right)
$$

Thus by (1) we get

$$
X=U^{*} \operatorname{diag}\left(\alpha_{1}^{1 /(2 p+1)}, \ldots, \alpha_{r}^{1 /(2 p+1)}, 0, \ldots, 0\right) V^{*}
$$

If $X$ is in the above form, it follows easily that $X$ satisfies the equation.
Theorem 3. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p}=\left(X X^{*}\right)^{q}, \quad p>q \geq 1
$$

iff each nonzero singular value of $X$ is 1 .
Proof. Assume $X$ is a solution of the equation.
By making use of Lemma 1 and using the same method as in the proof of Theorem 1, we get

$$
\operatorname{diag}\left(x_{1}^{2 p}, \ldots, x_{r}^{2 p}, 0, \ldots, 0\right)=\operatorname{diag}\left(x_{1}^{2 q}, \ldots, x_{r}^{2 q}, 0, \ldots, 0\right)
$$

which implies $x_{1}=\cdots=x_{r}=1$.
Now suppose $X$ is a matrix with each of its nonzero singular value 1 .
Then by Lemma 1, we have

$$
U X V=\operatorname{diag}(1, \ldots, 1,0, \ldots, 0)
$$

or

$$
X=U^{*} \operatorname{diag}(1, \ldots, 1,0, \ldots, 0) V^{*}
$$

It is easily checked $X$ in the above form satisfies the equation.
Theorem 4. A matrix $X$ is a solution of the equation

$$
\left(X X^{*}\right)^{p} X=\left(X X^{*}\right)^{q} X, \quad p>q \geq 0
$$

iff each nonzero singular value of $X$ is 1 .
Proof. The proof as in Theorem 3 works.
Theorem 5. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p}=\left(X^{*} X\right)^{q}, \quad p, q \geq 1, \quad p \neq q
$$

iff $X$ is normal with each of its nonzero singular value 1.
Proof. Note here $X$ must be square. Assume $X$ is a solution of the equation. By Lemma 2, we can write

$$
X=U H
$$

Thus from the given equation, we get

$$
\begin{equation*}
U H^{2 p} U^{*}=H^{2 q} \tag{3}
\end{equation*}
$$

It follows that $H^{2 p}$ and $H^{2 q}$ have the same eigenvalues. If the eigenvalues of $H$ are $\lambda_{1} \geq \cdots \geq \lambda_{r}>\lambda_{r+1}=\cdots=\lambda_{n}=0$, we get

$$
\lambda_{i}^{2 p}=\lambda_{i}^{2 q}, \quad i=1, \ldots, r
$$

Thus $\lambda_{1}=\cdots=\lambda_{r}=1$. Hence $H$ is a projection and from (3) we see that $H$ and $U$ commute. Therefore by Lemma 2, $H$ is normal. If $X$ is normal with each of its nonzero singular value one, as in the proof of Theorem 3, it follows that $X$ satisfies the equation.

Theorem 6. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p} X=\left(X X^{*}\right)^{q}
$$

iff $X$ is a projection.
Proof. Assume $X$ is a solution of the equation. Then by making use of Lemma 2, from the given equation, we get

$$
\begin{equation*}
H^{2 p+1} U=H^{2 q}=H^{2 q} I \tag{4}
\end{equation*}
$$

By the uniqueness of $H$, we get

$$
H^{2 p+1}=H^{2 q}
$$

Thus $H$ is a projection. From (4), we have

$$
\begin{gathered}
H U=H \\
X=H U=H
\end{gathered}
$$

is a projection.
The converse is obvious.
Theorem 7. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p} X=\left(X^{*} X\right)^{p} X^{*}(p \geq 0)
$$

iff $X$ is hermitian.
Proof. Assume $X$ is a solution. Then

$$
\begin{aligned}
\left(X X^{*}\right)^{2 p+1} & =\left(\left(X X^{*}\right)^{p} X\right)\left(X^{*}\left(X X^{*}\right)^{p}\right) \\
& =\left(\left(X^{*} X\right)^{p} X^{*}\right)\left(X\left(X^{*} X\right)^{p}\right) \\
& =\left(X^{*} X\right)^{2 p+1}
\end{aligned}
$$

Therefore by the uniqueness of root extraction we get

$$
X X^{*}=X^{*} X
$$

Hence $X$ is normal. Therefore each eigenvalue of $X$ satisfies

$$
|\lambda|^{p} \lambda=|\lambda|^{p} \bar{\lambda} .
$$

It follows that nonzero eigenvalues of $X$ are real. Hence $X$ is hermitian. The converse is obvious.

Theorem 8. A matrix $X$ is a solution of the matrix equation

$$
\left(X X^{*}\right)^{p} X=\left(X^{*} X\right)^{q} X^{*}, \quad p \neq q
$$

iff $X$ is hermitian with $0,1,-1$ as the only eigenvalues.
Proof. Assume $X$ is a solution of the equation. As in the proof of Theorem 7, we get

$$
\begin{equation*}
\left(X X^{*}\right)^{2 p+1}=\left(X^{*} X\right)^{2 q+1} \tag{5}
\end{equation*}
$$

Since the eigenvalues of $X X^{*}$ and $X^{*} X$ coincide, it follows that any eigenvalue of $X X^{*}$ is 1 or 0 .

Therefore $X X^{*}$ and $X^{*} X$ are projections. From (5) we see that

$$
\begin{equation*}
X X^{*}=X^{*} X \tag{6}
\end{equation*}
$$

From (6) $X$ is normal, therefore the eigenvalues $\lambda$ of $X$ satisfy the same equation as $X$ :

$$
|\lambda|^{p} \lambda=|\lambda|^{q} \bar{\lambda} .
$$

Since $X X^{*}$ is a projection, $|\lambda|=1$ or 0 . These facts immediately combine to give $\lambda=0,1$, or -1 . The converse is easily checked.

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## References

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