ON THE SOLUTIONS OF THE MATRIX EQUATION $f(X, X^*) = g(X, X^*)$

by P. BASAVAPPA

It is well known that the matrix identities $XX^* = I$, $X = X^*$ and $XX^* = X^*X$, 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) $f(X, X^*) = A$ and (b) $f(X, X^*) = g(X, X^*)$, where f(x, y) and g(x, y)are monomials of one of the following four forms: xyxy...xyx, xyxy...xyx, yxyx...yxy, and yxyx...yxy. In this paper all equations of the form (a) and (b) are solved completely. It may be noted a particular case of $f(X, X^*) = A$, viz. XX' = A, where X is a real square matrix and X' 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

$$UXV = \text{diag}(x_1, \ldots, x_r, 0, \ldots, 0),$$

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

$$X = HU(VK)$$

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

$$(XX^*)^p = A, \quad p \ge 1,$$

iff

$$X = U^* \operatorname{diag} (\alpha_1^{1/2p}, \ldots, \alpha_r^{1/2p}, 0, \ldots, 0) V^*$$

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where U is any unitary matrix such that

$$UAU^* = \operatorname{diag}\left(\alpha_1, \ldots, \alpha_r, 0, \ldots, 0\right)$$

and V is any unitary matrix.

Proof. Assume X is a solution of the equation. By Lemma 1 we have

(1)
$$UXV = \text{diag}(x_1, \ldots, x_r, 0, \ldots, 0).$$

Since X is a solution of the equation, we get

(2)
$$((UXV)(UXV)^*)^p = UAU^*.$$

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

diag
$$(x_1^{2p}, \ldots, x_r^{2p}, 0, \ldots, 0) = \text{diag}(\alpha_1, \ldots, \alpha_r, 0, \ldots, 0)$$

which implies that

$$UXV = \text{diag}(\alpha_1^{1/2p}, \ldots, \alpha_r^{1/2p}, 0, \ldots, 0)$$

or

$$X = U^* \operatorname{diag} (\alpha_1^{1/2p}, \ldots, \alpha_r^{1/2p}, 0, \ldots, 0) V^*.$$

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

 $(XX^*)^p X = A$

iff

$$X = U^* \operatorname{diag} (\alpha_1^{1/(2p+1)}, \ldots, \alpha_r^{1/(2p+1)}, 0, \ldots, 0) V^*$$

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

$$UAV = \text{diag}(\alpha_1, \ldots, \alpha_r, 0, \ldots, 0).$$

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

$$((UXV)(UXV)^*)^p UXV = UAV.$$

or

diag
$$(x_1^{2p+1}, \ldots, x_r^{2p+1}, 0, \ldots, 0) = UAV.$$

Thus if the singular values of A are $\alpha_1, \ldots, \alpha_r, 0, \ldots, 0$ we get

diag
$$(x_1, \ldots, x_r, 0, \ldots, 0)$$
 = diag $(\alpha_1^{1/(2p+1)}, \ldots, \alpha_r^{1/(2p+1)}, 0, \ldots, 0)$.

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Thus by (1) we get

$$X = U^* \operatorname{diag} (\alpha_1^{1/(2p+1)}, \ldots, \alpha_r^{1/(2p+1)}, 0, \ldots, 0) V^*.$$

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

$$(XX^*)^p = (XX^*)^q, \quad p > q \ge 1$$

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

diag
$$(x_1^{2p}, \ldots, x_r^{2p}, 0, \ldots, 0)$$
 = diag $(x_1^{2q}, \ldots, x_r^{2q}, 0, \ldots, 0)$

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

$$UXV = \text{diag}(1, ..., 1, 0, ..., 0).$$

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

$$(XX^*)^p X = (XX^*)^q X, \quad p > q \ge 0$$

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

$$(XX^*)^p = (X^*X)^q, \ p,q \ge 1, \ p \ne q$$

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

$$X = UH.$$

Thus from the given equation, we get

$$UH^{2p}U^* = H^{2q}.$$

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It follows that H^{2p} and H^{2q} have the same eigenvalues. If the eigenvalues of H are $\lambda_1 \ge \cdots \ge \lambda_r > \lambda_{r+1} = \cdots = \lambda_n = 0$, we get

$$\lambda_i^{2p} = \lambda_i^{2q}, \quad i = 1, \dots, r.$$

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

$$(XX^*)^p X = (XX^*)^q$$

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

(4) $H^{2p+1}U = H^{2q} = H^{2q}I$

By the uniqueness of H, we get

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

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

HU = HX = HU = H

is a projection.

The converse is obvious.

THEOREM 7. A matrix X is a solution of the matrix equation

$$(XX^*)^p X = (X^*X)^p X^* (p \ge 0)$$

iff X is hermitian.

Proof. Assume X is a solution. Then

$$(XX^*)^{2p+1} = ((XX^*)^p X)(X^*(XX^*)^p) = ((X^*X)^p X^*)(X(X^*X)^p) = (X^*X)^{2p+1}.$$

Therefore by the uniqueness of root extraction we get

$$XX^* = X^*X.$$

Hence X is normal. Therefore each eigenvalue of X satisfies

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

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

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THEOREM 8. A matrix X is a solution of the matrix equation

$$(XX^*)^p X = (X^*X)^q X^*, \quad p \neq q,$$

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

(5)
$$(XX^*)^{2p+1} = (X^*X)^{2q+1}.$$

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

Therefore XX^* and X^*X are projections. From (5) we see that

$$XX^* = X^*X.$$

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

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

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

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