

The eigenvalue problem of a specially updated matrix

Jiu Ding ^{a,*}, Guangming Yao ^{b,1}

^a Department of Mathematics, The University of Southern Mississippi, Hattiesburg, MS 39406-5045, USA

^b School of Mathematics and Computation, Harbin University, Harbin, China

Abstract

We study the eigenvalue problem for a specially structured rank- k updated matrix, based on the Sherman–Morrison–Woodbury formula.

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1. Introduction

In a recent paper [2], the eigenvalue problem of a special rank-one updated matrix was studied and the result therein provided an alternative proof of the eigenvalue theorem [3–6] for the Google matrix whose eigenvector associated with eigenvalue 1 is the so-called PageRank for the Google web search engine [1,7,8]. The main result of [2] is the following theorem.

Theorem 1.1. *Let A be an $n \times n$ real matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ counting algebraic multiplicities, and let u and v be two n -dimensional real column vectors such that v is a left eigenvector of A associated with eigenvalue λ_1 . Then, the eigenvalues of the matrix*

$$B = A + uv^T$$

are

$$\{\lambda_1 + u^T v, \lambda_2, \lambda_3, \dots, \lambda_n\}.$$

In this paper, we generalize Theorem 1.1 by considering the eigenvalue problem of the matrix

$$B = A + u_1 v_1^T + u_2 v_2^T + \dots + u_k v_k^T$$

with $2 \leq k \leq n$, where u_1, \dots, u_k and v_1, \dots, v_k are real column vectors such that v_1, \dots, v_k are left eigenvectors of A .

* Corresponding author.

E-mail address: Jiu.Ding@usm.edu (J. Ding).

¹ Present address: Department of Mathematics and Computation, Harbin University, Harbin, China.

Instead of the Sherman–Morrison formula used in [2] for the inverse of a rank-one updated matrix, we need the following Sherman–Morrison–Woodbury formula for the inverse of a rank- k updated matrix for our purpose.

Lemma 1.1. *If A is an invertible $n \times n$ real matrix, and U, V are two $n \times k$ real matrices, then the $n \times n$ matrix $A + UV^T$ is invertible if and only if the $k \times k$ matrix $I + V^T A^{-1} U$ is invertible, and then*

$$(A + UV^T)^{-1} = A^{-1} - A^{-1}U(I + V^T A^{-1}U)^{-1}V^T A^{-1}. \tag{1}$$

Although the proof of our main result **Theorem 3.1** works for any $k \leq n$, in the next section we prove the special case $k = 2$ first to illustrate the basic idea of our approach. Then we give the general result in Section 3.

2. Eigenvalues of rank-2 updated matrices

We first consider the special case $k = 2$ in this section. Let A be an $n \times n$ real matrix and let u_1, u_2, v_1, v_2 be n -dimensional real column vectors. We consider the eigenvalue problem for the matrix

$$B = A + u_1 v_1^T + u_2 v_2^T.$$

In the proof of our theorems below, we use the standard notation in matrix theory. For example, $N(A)$ denotes the null space of A and M^\perp is the orthogonal complement of a subset M in R^n .

Theorem 2.1. *Let A be an $n \times n$ real matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ counting algebraic multiplicities, and let u_1, u_2, v_1, v_2 be real column vectors such that v_1 and v_2 are linearly independent left eigenvectors of A corresponding to eigenvalues λ_1 and λ_2 , respectively. Then the eigenvalues of the matrix $B = A + u_1 v_1^T + u_2 v_2^T$ are*

$$\{\mu, \nu, \lambda_3, \dots, \lambda_n\},$$

where μ and ν are the eigenvalues of the 2×2 matrix

$$W = \begin{bmatrix} \lambda_1 + u_1^T v_1 & u_1^T v_2 \\ u_2^T v_1 & \lambda_2 + u_2^T v_2 \end{bmatrix} = \text{diag}(\lambda_1, \lambda_2) + U^T V \tag{2}$$

with U and V denoting the $n \times 2$ matrices

$$U = [u_1, u_2], \quad V = [v_1, v_2].$$

Proof. For any complex number λ ,

$$B - \lambda I = A - \lambda I + UV^T. \tag{3}$$

We first show that μ and ν are eigenvalues of B . Since $v_1^T A = \lambda_1 v_1^T$ and $v_2^T A = \lambda_2 v_2^T$, we have

$$v_1^T B = v_1^T A + v_1^T u_1 v_1^T + v_1^T u_2 v_2^T = (\lambda_1 + u_1^T v_1) v_1^T + (u_2^T v_1) v_2^T$$

and

$$v_2^T B = v_2^T A + v_2^T u_1 v_1^T + v_2^T u_2 v_2^T = (u_1^T v_2) v_1^T + (\lambda_2 + u_2^T v_2) v_2^T.$$

That is,

$$V^T B = W^T V^T, \tag{4}$$

where W is the 2×2 matrix as defined by (2). Suppose λ is an eigenvalue of W . Then there is a nonzero 2-dimensional column vector η such that $\eta^T(W^T - \lambda I) = 0$, which and (4) imply that

$$\eta^T V^T (B - \lambda I) = \eta^T (W^T - \lambda I) V^T = 0.$$

Since v_1, v_2 are linearly independent, $V\eta$ is a left eigenvector of B associated with eigenvalue λ . Therefore, both μ and ν are eigenvalues of B .

Next we show that a complex number λ is an eigenvalue of B if and only if $\lambda \in \{\mu, v, \lambda_3, \dots, \lambda_n\}$. We consider four cases separately.

- (i) Assume $\lambda \neq \lambda_i$ for all $i = 1, \dots, n$. Then the inverse matrix $(A - \lambda I)^{-1}$ exists. If λ is an eigenvalue of B , then

$$\det[I + V^T(A - \lambda I)^{-1}U] = 0 \tag{5}$$

by the Sherman–Morrison–Woodbury formula applied to (3). Since $V^T A = \text{diag}(\lambda_1, \lambda_2)V^T$, $V^T(A - \lambda I) = \text{diag}(\lambda_1 - \lambda, \lambda_2 - \lambda)V^T$,

which implies that

$$V^T(A - \lambda I)^{-1} = \text{diag}\left(\frac{1}{\lambda_1 - \lambda}, \frac{1}{\lambda_2 - \lambda}\right)V^T.$$

It follows that

$$\begin{aligned} \det[I + V^T(A - \lambda I)^{-1}U] &= \det\left[I + \text{diag}\left(\frac{1}{\lambda_1 - \lambda}, \frac{1}{\lambda_2 - \lambda}\right)V^T U\right] \\ &= \det\left[\text{diag}\left(\frac{1}{\lambda_1 - \lambda}, \frac{1}{\lambda_2 - \lambda}\right)\right] \cdot \det[\text{diag}(\lambda_1, \lambda_2) + U^T V - \lambda I] \\ &= \frac{1}{(\lambda_1 - \lambda)(\lambda_2 - \lambda)} \cdot \det(W - \lambda I). \end{aligned}$$

So, equality (5) implies that $\det(W - \lambda I) = 0$. That is, λ is an eigenvalue of the 2×2 matrix W . Therefore $\lambda = \mu$ or v . This proves that all the other eigenvalues of B are inside the eigenvalue set $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ of A .

- (ii) Suppose $\lambda = \lambda_i$ for some $i = 3, \dots, n$. Then the matrix $A - \lambda I$ is singular. Since

$$A - \lambda I = B - \lambda I - UV^T, \tag{6}$$

if λ_i is not an eigenvalue of B , then the Sherman–Morrison–Woodbury formula applied to (6) implies that

$$\det[I - V^T(B - \lambda I)^{-1}U] = 0. \tag{7}$$

Since $V^T(B - \lambda I) = (W^T - \lambda I)V^T$ from (4),

$$V^T(B - \lambda I)^{-1} = (W^T - \lambda I)^{-1}V^T.$$

Thus,

$$\begin{aligned} \det[I - V^T(B - \lambda I)^{-1}U] &= \det[I - (W^T - \lambda I)^{-1}V^T U] = \det[(W^T - \lambda I)^{-1}] \cdot \det[\text{diag}(\lambda_1 - \lambda, \lambda_2 - \lambda)] \\ &= \frac{(\lambda_1 - \lambda)(\lambda_2 - \lambda)}{\det(W - \lambda I)}. \end{aligned}$$

It follows from (7) that $\lambda = \lambda_1$ or λ_2 , which is a contradiction if $\lambda_i \neq \lambda_1, \lambda_2$. This says that λ_i is an eigenvalue of B under the additional assumption that $\lambda_i \neq \lambda_1, \lambda_2$.

- (iii) Now suppose $\lambda = \lambda_i = \lambda_1$ for some $i = 3, \dots, n$. Then the algebraic multiplicity of the eigenvalue λ for A is at least two. First assume that $\dim N(A - \lambda I) < \dim N[(A - \lambda I)^2]$. Then there is a nonzero vector $u \in N[(A - \lambda I)^2]$ such that $v \equiv (A - \lambda I)u \neq 0$. Since $(\lambda_2 - \lambda_1)^2 v_2^T u = v_2^T (A - \lambda I)^2 u = 0$, there holds $(\lambda_2 - \lambda_1)v_2^T u = 0$. Therefore, by (3),

$$\begin{aligned} (B - \lambda I)v &= (B - \lambda I)(A - \lambda I)u = [(A - \lambda I)^2 + UV^T(A - \lambda I)]u = (A - \lambda I)^2 u + (\lambda_2 - \lambda_1)u_2 v_2^T u \\ &= (\lambda_2 - \lambda_1)v_2^T u \cdot u_2 = 0. \end{aligned}$$

That is, λ is an eigenvalue of B with eigenvector v . Next assume that $\dim N(A - \lambda I) = \dim N[(A - \lambda I)^2] \geq 2$. If $\lambda_1 = \lambda_2$, then $\dim N(A - \lambda I) \geq 3$. Since $\dim\{v_1, v_2\}^\perp = n - 2$, $\dim N(B - \lambda I) \geq 1$ by (3). In other words, λ is an eigenvalue of B . If $\lambda_1 \neq \lambda_2$, then there is $u \neq 0$ such that $(A - \lambda I)u = 0$ and $v_1^T u = 0$ since $\dim\{v_1\}^\perp = n - 1$. Since $(\lambda_2 - \lambda_1)v_2^T u = v_2^T(A - \lambda I)u = 0$, we also have $v_2^T u = 0$. Hence,

$$(B - \lambda I)u = (A - \lambda I)u + UV^T u = 0.$$

That is, λ is an eigenvalue of B with eigenvector u . By the same token, if $\lambda_i = \lambda_2$ for some $i = 3, \dots, n$, then λ_i is an eigenvalue of B .

- (iv) Finally, we show that for $\lambda = \lambda_1$ or λ_2 , if $\lambda \neq \lambda_i$ for all $i = 3, \dots, n$, and if λ is not an eigenvalue of W , then λ is not an eigenvalue of B . We prove the claim for $\lambda = \lambda_1$ only since the proof for the other case is exactly the same. What we need to show is that the matrix $B - \lambda I$ is nonsingular. Let $w \in R^n$ be a non-zero vector. First assume that $w = V\eta$ for some nonzero vector $\eta \in R^2$. Since $\lambda \neq \mu, \nu$, the 2×2 matrix $W - \lambda I$ is nonsingular, so $\eta^T(W^T - \lambda I) \neq 0$. Since the rank of V is 2, equality (4) gives

$$w^T(B - \lambda I) = \eta^T V^T(B - \lambda I) = \eta^T(W^T - \lambda I)V^T \neq 0.$$

Now assume that $w \notin \text{span}\{v_1, v_2\}$. Suppose $w^T(B - \lambda I) = w^T(A - \lambda I) + w^T UV^T = 0$. Then $w^T(A - \lambda_1 I) = -w^T UV^T$, so

$$w^T(A - \lambda_2 I)(A - \lambda_1 I)^2 = -w^T UV^T(A - \lambda_1 I)(A - \lambda_2 I) = 0 \tag{8}$$

since $V^T(A - \lambda_1 I)(A - \lambda_2 I) = 0$. Eq. (8) and the fact that the algebraic multiplicity of the eigenvalue λ_1 of A is at most 2 imply that $w^T(A - \lambda_2 I) = \eta^T V^T$ for some $\eta \in R^2$. Therefore,

$$\eta^T V^T = w^T(A - \lambda_2 I) = w^T(A - \lambda_1 I) + (\lambda_1 - \lambda_2)w^T = -w^T UV^T + (\lambda_1 - \lambda_2)w^T. \tag{9}$$

If $\lambda_1 = \lambda_2$, then $w^T(A - \lambda_1 I)^2 = 0$ by (9), which contradicts the fact that $\dim N[(A - \lambda_1 I)^2] = 2$. If $\lambda_1 \neq \lambda_2$, then (9) gives

$$w = \frac{V(\eta + U^T w)}{\lambda_1 - \lambda_2},$$

which contradicts the assumption that $w \notin \text{span}\{v_1, v_2\}$. This concludes the proof of the theorem. \square

Remark 2.1. Actually, in case (iv) of the above proof, since the algebraic multiplicity of λ_1 is at most 2, by the theory of Jordan forms for matrices, if $w \notin \text{span}\{v_1, v_2\}$, then (8) implies that $w^T(A - \lambda_2 I)(A - \lambda_1 I)^2 \neq 0$. This observation will be used later to shorten the proof of Theorem 3.1.

An immediate consequence of Theorem 2.1 is

Corollary 2.1. *Suppose in addition that $u_1^T v_2 = 0$ and $u_2^T v_1 = 0$. Then the eigenvalues of B are*

$$\{\lambda_1 + u_1^T v_1, \lambda_2 + u_2^T v_2, \lambda_3, \dots, \lambda_n\}.$$

3. Eigenvalues of rank- k updated matrices

The same idea as used in the proof of Theorem 2.1 can be applied to establishing the following general result.

Theorem 3.1. *Let A be an $n \times n$ real matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ counting algebraic multiplicities, and for $1 \leq k \leq n$ let u_1, \dots, u_k and v_1, \dots, v_k be real column vectors such that v_1, \dots, v_k are linearly independent left eigenvectors of A corresponding to the eigenvalues $\lambda_1, \dots, \lambda_k$, respectively. Then the eigenvalues of the matrix $B = A + \sum_{i=1}^k u_i v_i^T$ are*

$$\{\mu_1, \mu_2, \dots, \mu_k, \lambda_{k+1}, \dots, \lambda_n\},$$

where μ_1, \dots, μ_k are the eigenvalues of the $k \times k$ matrix

$$W = \text{diag}(\lambda_1, \dots, \lambda_k) + U^T V \tag{10}$$

and

$$U = [u_1, \dots, u_k], \quad V = [v_1, \dots, v_k].$$

Proof. The special cases $k = 1$ and $k = 2$ have been covered by Theorems 1.1 and 2.1, respectively, so we assume $k \geq 2$. Exactly the same process in the proof of Theorem 2.1 can be used again to show that

- (i) μ_1, \dots, μ_k are eigenvalues of B ,
- (ii) if $\lambda \notin \{\lambda_1, \lambda_2, \dots, \lambda_n\}$, then λ is not an eigenvalue of B except for $\lambda = \mu_j$ for some $j = 1, \dots, k$, and
- (iii) λ_i is an eigenvalue of B for $i = k + 1, \dots, n$ if $\lambda_i \neq \lambda_j$ for all $j = 1, \dots, k$.

Now suppose $\lambda_i = \lambda_j$ for some $k + 1 \leq i \leq n$ and $1 \leq j \leq k$. We show that λ_j is an eigenvalue of B . We just prove the case $j = 1$ since the proof for $j = 2, \dots, k$ is exactly the same. Without loss of generality, we may assume that $\lambda_1 = \lambda_2 = \dots = \lambda_l$ and $\lambda_1 \neq \lambda_j$ for all $j = l + 1, \dots, k$. Then, since $\lambda_1 = \lambda_j$, the algebraic multiplicity of the eigenvalue λ_1 for A is at least $l + 1$. Let $u \in N[(A - \lambda_1 I)^2]$. Since $v_j^T (A - \lambda_1 I)^2 = (\lambda_j - \lambda_1)^2 v_j^T$,

$$v_j^T u = \frac{1}{(\lambda_j - \lambda_1)^2} v_j^T (A - \lambda_1 I)^2 u = 0, \quad j = l + 1, \dots, k. \tag{11}$$

Suppose first that $\dim N(A - \lambda_1 I) < \dim N[(A - \lambda_1 I)^2]$. Then there is $u \in N[(A - \lambda_1 I)^2]$ such that $v \equiv (A - \lambda_1 I)u \neq 0$. Since $v_j^T (A - \lambda_1 I) = 0$ for $j = 1, \dots, l$ and $v_j^T (A - \lambda_1 I) = (\lambda_j - \lambda_1)v_j^T$ for $j = l + 1, \dots, k$, we have

$$(B - \lambda_1 I)(A - \lambda_1 I) = (A - \lambda_1 I)^2 + UV^T(A - \lambda_1 I) = (A - \lambda_1 I)^2 + \sum_{j=l+1}^k (\lambda_j - \lambda_1)u_j v_j^T.$$

The above equality and (11) imply that $(B - \lambda_1 I)v = 0$. That is, λ_1 is an eigenvalue of B with eigenvector v .

Now suppose $\dim N(A - \lambda_1 I) = \dim N[(A - \lambda_1 I)^2]$. Then $N(A - \lambda_1 I) = N[(A - \lambda_1 I)^2]$ with dimension at least $l + 1$. Since $\dim\{v_1, \dots, v_l\}^\perp = n - l$, there is a nonzero vector $u \in N(A - \lambda_1 I)$ such that $u \in \{v_1, \dots, v_l\}^\perp$. Moreover, by (11), $u \in \{v_{l+1}, \dots, v_k\}^\perp$. Therefore,

$$(B - \lambda_1 I)u = (A - \lambda_1 I)u + UV^T u = 0.$$

So λ_1 is an eigenvalue of B with eigenvector u .

Finally we show that if for some $j = 1, \dots, k$, λ_j is not an eigenvalue of W defined by (10) and $\lambda_j \neq \lambda_i$ for all $i = k + 1, \dots, n$, then λ_j is not an eigenvalue of B . We assume $j = 1$ for the sake of simplicity of notation.

First let $w = V\eta$ for some nonzero vector $\eta \in R^k$. Since $W - \lambda_1 I$ is nonsingular, $\eta^T (W^T - \lambda_1 I) \neq 0$. The assumption that the rank of V is k and equality (10) give

$$w^T (B - \lambda_1 I) = \eta^T V^T (B - \lambda_1 I) = \eta^T (W^T - \lambda_1 I) V^T \neq 0.$$

Next assume $w \notin \text{span}\{v_1, \dots, v_k\}$. Suppose $w^T (B - \lambda_1 I) = w^T (A - \lambda_1 I) + w^T UV^T = 0$. Then $w^T (A - \lambda_1 I) = -w^T UV^T$, so

$$w^T \prod_{j=2}^k (A - \lambda_j I) \cdot (A - \lambda_1 I)^2 = -w^T UV^T \prod_{j=1}^k (A - \lambda_j I) = 0 \tag{12}$$

since $V^T \prod_{j=1}^k (A - \lambda_j I) = 0$. On the other hand, since $\lambda_1 \neq \lambda_i$ for all $i = k + 1, \dots, n$, the condition $w \notin \text{span}\{v_1, \dots, v_k\}$ implies that $w^T \prod_{j=2}^k (A - \lambda_j I) \cdot (A - \lambda_1 I)^2 \neq 0$. This gives a contradiction to (12). \square

In particular, we have

Corollary 3.1. Let A be an $n \times n$ real matrix with eigenvalues $\lambda_1, \lambda_2, \dots, \lambda_n$ counting algebraic multiplicities. For $i = 1, 2, \dots, k$ let u_i and v_i be n -dimensional real column vectors such that v_i 's are linearly independent left eigenvectors of A associated with eigenvalues λ_i respectively. If $u_i^T v_j = 0$ for all $i \neq j$, then the eigenvalues of the matrix

$$A + \sum_{i=1}^k u_i v_i^T$$

are

$$\{\lambda_1 + u_1^T v_1, \dots, \lambda_k + u_k^T v_k, \lambda_{k+1}, \dots, \lambda_n\}.$$

Remark 3.1. We point out that the assumption that v_1, \dots, v_k are linearly independent is not necessary and can be removed from the fact that eigenvalues of a matrix are continuous functions of its entries and any square matrix is a limit of a sequence of matrices with all distinct eigenvalues so that their corresponding eigenvectors are linearly independent.

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