

Computing for Eigenpairs on Globally Convergent Iterative Method for Hermitian Matrices

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Abstract. Let $A = A^* \in M_n$ and $\mathcal{L} = \{(U_k, \lambda_k) \mid U_k \in \mathbb{C}^n, \|U_k\| = 1 \text{ and } \lambda_k \in \mathbb{R}\}$ for $k = 1, \dots, n$ be the set of eigenpairs of A . In this paper we develop a modified Newton method that converges to a point in \mathcal{L} starting from any point in a compact subset $\mathcal{D} \subseteq \mathbb{C}^{n+1}$, $\mathcal{L} \subseteq \mathcal{D}$.

1 Introduction

We denote by M_n the space of n -by- n complex matrices. We denote by $\sigma(A)$ the set of eigenvalues of $A \in M_n$. Let $A \in M_n$ be Hermitian. Then there is a unitary

$$U = [U_1 \cdots U_n] \in M_n \text{ such that } A = U \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} U^*, \quad \lambda_k \in \mathbb{R}.$$

We assume that the eigenvalues λ_k of A are arranged in decreasing order, i.e., $\lambda_1 \geq \dots \geq \lambda_n$ [3, chapter 4].

Let $\mathcal{L} = \left\{ \begin{bmatrix} U_k \\ \lambda_k \end{bmatrix} \mid U_k \in \mathbb{C}^n, \|U_k\| = 1, \text{ and } \lambda_k \in \mathbb{R} \right\}_{k=1, \dots, n}$ be the set of eigen-

pairs of A , and suppose $\mathcal{D} = \left\{ \begin{bmatrix} X \\ \alpha \end{bmatrix} \mid X \in \mathbb{C}^n, \|X\|_2 = 1, \text{ and } \alpha \in [a, b] \right\}$ be

a compact subset of $\mathbb{C}^n \times \mathbb{R}$ such that $\mathcal{L} \subseteq \mathcal{D}$. The purpose of this paper is to compute on globally convergent iteration method which converges to an eigen-

pair of A , i.e., an element of \mathcal{L} , starting from any arbitrary point $\begin{bmatrix} X \\ \alpha \end{bmatrix} \in \mathcal{D}$. The

following is the usual Newton method for obtaining an eigenpair of a hermitian matrix $A \in M_n$ [4]: Consider $G: \mathbb{C}^n \times \mathbb{R} \rightarrow \mathbb{C}^n \times \mathbb{R}$ such that

$$G \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) = \begin{bmatrix} (\alpha I - A)X \\ X^*X - 1 \end{bmatrix}. \quad (1)$$

Then \mathcal{L} is the set of solutions for $G \begin{pmatrix} X \\ \alpha \end{pmatrix} = 0$. Assuming the matrix $\begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix} \in M_{n+1}$ is invertible, then the usual newton iteration is

$$\begin{bmatrix} X' \\ \alpha' \end{bmatrix} = \begin{bmatrix} X \\ \alpha \end{bmatrix} - \begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix}^{-1} \begin{bmatrix} (\alpha I - A)X \\ X^*X - 1 \end{bmatrix} \quad (2)$$

It is well known that the newton's method has a local quadratic convergence rate [2], that is there is a small neighborhood N_{ϵ_k} for each eigenpair $\begin{bmatrix} U_k \\ \lambda_k \end{bmatrix}$ such that if $\begin{bmatrix} X \\ \alpha \end{bmatrix} \equiv \begin{bmatrix} X^{(0)} \\ \alpha^{(0)} \end{bmatrix} \in N_{\epsilon_k}$ then $\| \begin{bmatrix} X^{(i+1)} \\ \alpha^{(i+1)} \end{bmatrix} - \begin{bmatrix} U_k \\ \lambda_k \end{bmatrix} \|_2 \leq C \| \begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} - \begin{bmatrix} U_k \\ \lambda_k \end{bmatrix} \|_2^2$ for all $i = 0, 1, \dots$, where $C < \infty$ is a positive constant.

We call N_{ϵ_k} the quadratic convergence neighborhood of the eigenpair $\begin{bmatrix} U_k \\ \lambda_k \end{bmatrix}$. Although the specific determination of each N_{ϵ_k} is an extremely difficult task, if the method converges to a point in \mathcal{L} then we know the rate of convergence will eventually be quadratic. It can be shown easily that the newton's method is not global. We provide an example:

Example 1. Let $A = \begin{bmatrix} 1.1 + \epsilon & 0 \\ 0 & 0.9 \end{bmatrix}$, $\epsilon > 0$ be the objective matrix with eigenpairs $\begin{bmatrix} U_1 \\ \lambda_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 1.1 + \epsilon \end{bmatrix}$ and $\begin{bmatrix} U_2 \\ \lambda_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0.9 \end{bmatrix}$. Suppose the initial points are $\begin{bmatrix} X_1 \\ \alpha_1 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix}$ and $\begin{bmatrix} X_2 \\ \alpha_2 \end{bmatrix} = \begin{bmatrix} \frac{-1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix}$.

Then the newton iteration (2) becomes

$$\begin{aligned} \begin{bmatrix} X'_1 \\ \alpha'_1 \end{bmatrix} &= \begin{bmatrix} X_1 \\ \alpha_1 \end{bmatrix} - \begin{bmatrix} \alpha_1 I - A & X_1 \\ 2X_1^* & 0 \end{bmatrix}^{-1} \begin{bmatrix} (\alpha_1 I - A)X_1 \\ X_1^*X_1 - 1 \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix} - \begin{bmatrix} -0.1 - \epsilon & 0 & \frac{1}{\sqrt{2}} \\ 0 & 0.1 & \frac{1}{\sqrt{2}} \\ \frac{2}{\sqrt{2}} & \frac{2}{\sqrt{2}} & 0 \end{bmatrix}^{-1} \begin{bmatrix} \frac{-(0.1+\epsilon)}{\sqrt{2}} \\ \frac{\sqrt{2}}{0.1} \\ \frac{\sqrt{2}}{0} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix} - 1/\epsilon \begin{bmatrix} -1 & 1 & \frac{-0.1}{\sqrt{2}} \\ 1 & -1 & \frac{0.1+\epsilon}{\sqrt{2}} \\ \frac{-0.2}{\sqrt{2}} & \frac{2(0.1+\epsilon)}{\sqrt{2}} & -0.1(0.1+\epsilon) \end{bmatrix} \begin{bmatrix} \frac{-(0.1+\epsilon)}{\sqrt{2}} \\ \frac{\sqrt{2}}{0.1} \\ 0 \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix} - 1/\epsilon \begin{bmatrix} \frac{(0.2+\epsilon)}{\sqrt{2}} \\ \frac{-(0.2+\epsilon)}{\sqrt{2}} \\ 0.02 + 0.2\epsilon \end{bmatrix} \end{aligned}$$

Thus if ϵ goes to zero, the iteration diverges. Similarly, for the initial eigenpair $\begin{bmatrix} X_2 \\ \alpha_2 \end{bmatrix}$. We modify the newton method in order to have a global convergence. There are several considerations to give for the modification. First, under the modification we desire the pair $\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix}$ gets closer to an eigenpair at each step of the iteration, i.e., $d_{\mathcal{L}}\left(\begin{bmatrix} X^{(i+1)} \\ \alpha^{(i+1)} \end{bmatrix}\right) \leq d_{\mathcal{L}}\left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix}\right)$ where $d_{\mathcal{L}}$ is a suitable distance measure from a point to \mathcal{L} . It will ensure the points under the iteration remain in \mathcal{D} . Second, we want to modify the method the least amount as possible in order to preserve the original properties of the newton's Method, for example, local quadratic convergence. Third, the modified method should be simple to implement, requires almost the same procedures as the original newton iteration.

2 Modification for Global Newton Iteration

Consider the newton iteration (2): $\begin{bmatrix} X' \\ \alpha' \end{bmatrix} = \begin{bmatrix} X \\ \alpha \end{bmatrix} - \begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix}^{-1} \begin{bmatrix} (\alpha I - A)X \\ X^*X - 1 \end{bmatrix}$
 Then, assuming $\alpha \neq 0$ and $\alpha \notin \sigma(A)$

$$\begin{aligned} \begin{bmatrix} X' \\ \alpha' \end{bmatrix} &= \left(I - \begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix}^{-1} \begin{bmatrix} \alpha I - A & 0 \\ X^* & -1/\alpha \end{bmatrix} \right) \begin{bmatrix} X \\ \alpha \end{bmatrix} \\ &= \begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix}^{-1} \begin{bmatrix} 0 & X \\ X^* & 1/\alpha \end{bmatrix} \begin{bmatrix} X \\ \alpha \end{bmatrix}, \text{ or } \begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix} \begin{bmatrix} X' \\ \alpha' \end{bmatrix} = \begin{bmatrix} 0 & X \\ X^* & 1/\alpha \end{bmatrix} \begin{bmatrix} X \\ \alpha \end{bmatrix}. \end{aligned}$$

Choose a parameter $t > 0$ so that the method takes the form [1]

$$\begin{bmatrix} \alpha I - A & X \\ 2X^* & 0 \end{bmatrix} \begin{bmatrix} X' \\ \alpha' \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & t \end{bmatrix} \begin{bmatrix} 0 & X \\ X^* & 1/\alpha \end{bmatrix} \begin{bmatrix} X \\ \alpha \end{bmatrix}. \quad (3)$$

Then

$$(\alpha I - A)X' + \alpha'X = \alpha X \quad (4)$$

and

$$2X^*X' = t(X^*X + 1) \quad (5)$$

From (4), we have $X' = (\alpha - \alpha')(\alpha I - A)^{-1}X$, and hence from (5) $X^*X' = \frac{1}{2}t(X^*X + 1) = (\alpha - \alpha')X^*(\alpha I - A)^{-1}X$. Set $\beta \equiv X^*(\alpha I - A)^{-1}X$. Then $\frac{t}{2}(X^*X + 1) = \beta(\alpha - \alpha')$, or $(\alpha - \alpha') = \frac{t}{2}(X^*X + 1)\frac{1}{\beta}$. Thus, $X' = \frac{t}{2}(X^*X + 1)\frac{1}{\beta}(\alpha I - A)^{-1}X$, and $\alpha' = \alpha - \frac{t}{2}(X^*X + 1)\frac{1}{\beta}$. If we normalize the vector $X \in \mathbb{C}^n$ in each step of the iteration, $\frac{1}{2}(X^*X + 1) = 1$.

Thus we have $X' = \frac{1}{\|\frac{t}{\beta}(\alpha I - A)^{-1}X\|_2} \frac{t}{\beta}(\alpha I - A)^{-1}X$, and $\alpha' = \alpha - \frac{t}{\beta}$. Set

$$\hat{\alpha} \equiv \|(\alpha I - A)^{-1}X\|_2 = (X^*(\alpha I - A)^{-2}X)^{\frac{1}{2}}.$$

Then we have $X' = \frac{1}{\hat{\lambda}} \frac{|\beta|}{\beta} (\alpha I - A)^{-1} X$, and $\alpha' = \alpha - \frac{t}{\beta}$. Since $\frac{|\beta|}{\beta} = \pm 1$, we ignore the sign. Then the parameterized newton method takes a form:

$$X' = \frac{1}{\hat{\lambda}} (\alpha I - A)^{-1} X, \text{ and} \quad (6)$$

$$\alpha' = \alpha - \frac{t}{\beta}. \quad (7)$$

Now, suppose $\begin{bmatrix} X \\ \alpha \end{bmatrix} \in \mathcal{D}$ and Let $\mathcal{L} = \left\{ \begin{bmatrix} U_k \\ \lambda_k \end{bmatrix} \mid U_k \in \mathbb{C}^n, \|U_k\|_2 = 1, \text{ and } \lambda_k \in \mathbb{R} \right\}$ be the set of all eigenpairs of A . Define a distance measure from a point $\begin{bmatrix} X \\ \alpha \end{bmatrix}$ to \mathcal{L} by

$$d_{\mathcal{L}} \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) \equiv \|(\alpha I - A)X\|_2. \quad (8)$$

Clearly, $d_{\mathcal{L}} \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) \geq 0$, $d_{\mathcal{L}} \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) = 0$ implies $\begin{bmatrix} X \\ \alpha \end{bmatrix} \in \mathcal{L}$, and $d_{\mathcal{L}} : \mathcal{D} \rightarrow \mathbb{R}^+$ is continuous (since \mathcal{D} is compact, $d_{\mathcal{L}}$ is actually uniformly continuous)[5].

We have the following.

Lemma 1. *Let $A \in M_n$ be Hermitian. Consider the parameterized newton's method $X' = \frac{1}{\hat{\lambda}} (\alpha I - A)^{-1} X$ and $\alpha' = \alpha - \frac{t}{\beta}$, where $\beta = X^* (\alpha I - A)^{-1} X$ and $\hat{\beta} = \|(\alpha I - A)^{-1} X\|_2 = (X^* (\alpha I - A)^{-2} X)^{1/2}$. Then $d_{\mathcal{L}} \left(\begin{bmatrix} X' \\ \alpha' \end{bmatrix} \right)$ is minimized at $t = \left(\frac{\beta}{\hat{\lambda}} \right)^2$ with $\min_t d_{\mathcal{L}} \left(\begin{bmatrix} X' \\ \alpha' \end{bmatrix} \right) = \frac{1}{\hat{\lambda}} \left(1 - \left(\frac{\beta}{\hat{\lambda}} \right)^2 \right)^{\frac{1}{2}}$.*

Proof: Suppose $X' = \frac{1}{\hat{\lambda}} (\alpha I - A)^{-1} X$ and $\alpha' = \alpha - \frac{t}{\beta}$. Then

$$\begin{aligned} d_{\mathcal{L}}^2 \left(\begin{bmatrix} X' \\ \alpha' \end{bmatrix} \right) &= \|((\alpha - \frac{t}{\beta})I - A) \frac{1}{\hat{\lambda}} (\alpha I - A)^{-1} X\|_2^2 \\ &= \frac{1}{\hat{\lambda}^2} X^* (I - 2\frac{t}{\beta} (\alpha I - A)^{-1} + (\frac{t}{\beta})^2 (\alpha I - A)^{-2}) X \\ &= \frac{1}{\hat{\lambda}^2} (1 - 2\frac{t}{\beta} \cdot \beta + t^2 \left(\frac{\hat{\beta}}{\beta} \right)^2). \end{aligned}$$

Thus, $d_{\mathcal{L}}^2 \left(\begin{bmatrix} X' \\ \alpha' \end{bmatrix} \right)$ is minimized at $t = \left(\frac{\beta}{\hat{\lambda}} \right)^2$ with $\min_t d_{\mathcal{L}} \left(\begin{bmatrix} X' \\ \alpha' \end{bmatrix} \right) = \frac{1}{\hat{\lambda}} \left(1 - \left(\frac{\beta}{\hat{\lambda}} \right)^2 \right)^{\frac{1}{2}}$. □

Therefore, we have the following modification of the newton's method:

$$X' = \frac{1}{\hat{\lambda}}(\alpha I - A)^{-1}X. \quad (9)$$

$$\alpha' = \alpha - \frac{\beta}{\hat{\lambda}^2}. \quad (10)$$

The following result shows that the modified iteration (9), (10) is bounded.

Lemma 2. *Let $A \in M_n$ be a Hermitian such that $\sigma(A) = \{\lambda_1 \geq \dots \geq \lambda_n, \lambda_k \in \mathbb{R}\}$. Let $\mathcal{D} = \left\{ \begin{bmatrix} X \\ \alpha \end{bmatrix} \mid X \in \mathbb{C}^n, \|X\|_2 = 1, \text{ and } \alpha \in [a, b] \right\}$ be such that $a \leq \lambda_n$*

and $b \geq \lambda_1$. Suppose $\mathcal{D}' = \left\{ \begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right\}_{i=t, \infty, \dots}$ is the sequence of iterates of (9)

and (10), i.e., $X^{(i+1)} = \frac{1}{\hat{\lambda}^{(i)}}(\alpha^{(i)}I - A)^{-1}X^{(i)}$, and $\alpha^{(i+1)} = \alpha^{(i)} - \frac{\beta^{(i)}}{\left(\hat{\lambda}^{(i)}\right)^2}$.

Then $\mathcal{D}' \subset \mathcal{D}$ whenever $\begin{bmatrix} X^{(0)} \\ \alpha^{(0)} \end{bmatrix} \in \mathcal{D}$.

Theorem 3. *Suppose $\mathcal{D}' = \left\{ \begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right\}_{i=t, \infty, \dots}$ is the sequence of iterates of (9)*

and (10). Then the sequence $\left\{ d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) \right\}_{i=0, 1, \dots}$ is convergent.

Note from Theorem 3 that since $\left\{ d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) \right\}$ is a monotone decreasing sequence that is bounded below by zero, The sequence $\left\{ d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) \right\}$ converges to either (i) zero, or (ii) a positive constant L .

Suppose $\lim_{i \rightarrow \infty} d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) = 0$. Then clearly, $\left\{ d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) \right\}$ converges to an

eigenpair of A . In the following section we discuss the case $\lim_{i \rightarrow \infty} d_{\mathcal{L}} \left(\begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right) =$

$L > 0$ which requires some detailed analysis. We summarize the results. Under the modified newton iteration, the sequence $\{d_{\mathcal{L}}\}$ converges to either zero or

$L > 0$. If $\{d_{\mathcal{L}}\}$ converges to $L > 0$, then iterates $\left\{ \begin{bmatrix} X^{(i)} \\ \alpha^{(i)} \end{bmatrix} \right\}$ has an accumulation

point $\begin{bmatrix} X \\ \alpha \end{bmatrix}$ where $d_{\mathcal{L}} \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) = L > 0$ such that the point $\alpha \in \mathbb{R}$ lies exactly at

the midpoint of two distinct eigenvalues (each eigenvalue may have the algebraic multiplicity more than one) such that corresponding components of the vector

U^*X have equal weights that are $\frac{1}{2}$ each, see Figure 1.

$$\begin{array}{ccc} \lambda_t & \alpha & \lambda_s \\ (U^*X)_t^2 = \frac{1}{2} & & (U^*X)_s^2 = \frac{1}{2} \end{array}$$

Fig. 1.

Therefore, $d_{\mathcal{L}}^2 = \frac{1}{\hat{\beta}} = \sum_{k=1}^n (\alpha - \lambda_k)^2 |y_k|^2 = (\alpha - \lambda_k)^2 = \frac{(\lambda_s - \lambda_t)^2}{2}$, $\lambda_s > \lambda_t$.

We conclude this section with the following consequence of above results and an example.

Theorem 4. Suppose $d_{\mathcal{L}} \left(\begin{bmatrix} X \\ \alpha \end{bmatrix} \right) = L$. Then both $\alpha + \frac{1}{\hat{\beta}}$ and $\alpha - \frac{1}{\hat{\beta}}$ are the eigenvalues of A and $\frac{1}{\hat{\beta}} [(\alpha + \frac{1}{\hat{\beta}})I - A]^{-1}X$ and $\frac{1}{\hat{\beta}''} [(\alpha - \frac{1}{\hat{\beta}})I - A]^{-1}X$ are

the corresponding eigenvector of A , where $\hat{\beta}' = \left\| \left[\begin{array}{c} (\alpha + \frac{1}{\hat{\beta}})I - A \\ X \end{array} \right]^{-1} \right\|_2$, and

$$\hat{\beta}'' = \left\| \left[\begin{array}{c} (\alpha - \frac{1}{\hat{\beta}})I - A \\ -1X \end{array} \right] \right\|_2.$$

Example 2. Consider the Example 1. Let $A = \begin{bmatrix} 1.1 + \epsilon & 0 \\ 0 & 0.9 \end{bmatrix}$, $\epsilon > 0$. Suppose

we start with initial eigenpair $\begin{bmatrix} X_1^{(0)} \\ \alpha_1^{(0)} \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \\ 1 \end{bmatrix}$. Then

$$\begin{aligned} \beta &= X^*(\alpha I - A)^{-1}X = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} -(0.1 - \epsilon) & 0 \\ 0 & 0.1 \end{bmatrix}^{-1} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} -1/(0.1 + \epsilon) & 0 \\ 0 & 10 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \\ &= \frac{-1}{2(0.1 + \epsilon)} + 5 = 5 - \frac{1}{0.2 + 2\epsilon}, \text{ and} \end{aligned}$$

$$\begin{aligned} \hat{\beta} &= \|(\alpha I - A)^{-1}X\|_2 = \left\| \begin{bmatrix} -1/(0.1 + \epsilon) & 0 \\ 0 & 10 \end{bmatrix} \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \right\|_2 \\ &= \left(\frac{1}{2} \left(\frac{-1}{0.1 + \epsilon} \right)^2 + 50 \right)^{1/2} = \left(\frac{1}{0.02 + 0.4\epsilon + \epsilon^2} + 50 \right)^{1/2}. \end{aligned}$$

If ϵ goes to zero, then $\beta \rightarrow 0$ and $\hat{\beta} \rightarrow 10$. Notice that for $\epsilon \cong 0$

$$d_{\mathcal{L}} \left(\begin{bmatrix} X^{(0)} \\ \alpha^{(0)} \end{bmatrix} \right) = d_{\mathcal{L}} \left(\begin{bmatrix} X^{(m)} \\ \alpha^{(m)} \end{bmatrix} \right) = \dots = 0.1.$$

Therefore by Theorem 4, we have $d_{\mathcal{L}} \left(\begin{bmatrix} X^{(n)} \\ \alpha^{(n)} \end{bmatrix} \right) = \frac{1}{\hat{\beta}}$.

Hence $\lambda_1(A) = 1 + 1/\hat{\beta} = 1 + .1 = 1.1$ and $\lambda_2(A) = 1 - 1/\hat{\beta} = 1 - .1 = .9$.

We obtain $X_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ and $X_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ by solving $(A - \lambda_1 I)X_1 = 0$ and $(A - \lambda_2 I)X_2 = 0$.

3 Examples with Modified Newton’s Iterations

Example 3. Let $H = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{3} & \dots & \frac{1}{n} \\ \frac{1}{2} & \frac{1}{3} & & & \vdots \\ \vdots & & \ddots & & \\ \vdots & & & \ddots & \\ \frac{1}{n} & \frac{1}{n+1} & \dots & \dots & \frac{1}{2n-1} \end{bmatrix}$ be a n by n Hilbert matrix H .

The Hilbert matrix is a well-known example of an ill-conditioned positive definite matrix. Because the smallest eigenvalue λ_{12} of $H \in M_{12}$ is so near zero, many conventional algorithms produce $\lambda_{12} = 0$. Our method gives the following experimental results: Set the convergence criterion, $\epsilon = 2 \times 10^{-16}$, i.e., $\|(H - \alpha_k^{(i)} I)X_k^{(i)}\|_2 < \epsilon$.

Suppose $\mathcal{D} = \left\{ \begin{bmatrix} X \\ \alpha \end{bmatrix} \mid X \in \{e_1, \dots, e_{12}\}, \text{ and } \alpha \in \{h_{1,1}, \dots, h_{12,12}\} \right\}$ is the initial set of points where e_i is the i th column of the identity matrix and $h_{i,i}$ is the i th diagonal entry of H .

Table 1. Eigenvalues of H_{12} by Modified Newton Iteration

Eigenvalues	Eigenvalues of H	$\ (H - \alpha_k^{(i)} I)X_k^{(i)}\ _2$
1st	1.7953720595620	4.5163365159057D-17
2nd	0.38027524595504	9.5107769421299D-17
3rd	4.4738548752181D-02	9.3288777118150D-17
4th	3.7223122378912D-03	9.5107769421299D-17
5th	2.3308908902177D-04	6.5092594645258D-17
6th	1.1163357483237D-05	6.6374428417771D-17
7th	4.0823761104312D-07	1.9236667674542D-16
8th	1.1228610666749D-08	4.9553614188006D-17
9th	2.2519644461451D-10	6.0015952254039D-17
10th	3.1113405079204D-12	6.5125904614112D-17
11th	2.6487505785549D-14	1.0932505712948D-16
12th	1.1161909467844D-16	6.0015952254039D-17

Example 4. Let $H_3 = \begin{bmatrix} 10^{40} & 10^{19} & 10^{19} \\ 10^{19} & 10^{20} & 10^9 \\ 10^{19} & 10^9 & 1 \end{bmatrix}$ be a 3 by 3 graded matrix. Suppose $\mathcal{D} = \left\{ \begin{bmatrix} X \\ \alpha \end{bmatrix} \mid X \in \{e_1, e_2, e_3\}, \text{ and } \alpha \in \{h_{1,1}, h_{2,2}, h_{3,3}\} \right\}$ is the initial set of points where e_i is the i th column of the identity matrix and $h_{i,i}$ is the i th diagonal entry of H . The graded matrix is a well-known example of an ill-conditioned symmetric matrix. The following is the result obtained by MATLAB and our method.

Table 2. Eigenvalues of H_3 by Modified Newton Iteration

Eigenvalues	Modified Newton Method		MATLAB
	Eigenvalues of H	Iteration	Eigenvalues of H
λ_1	1.00000000000000D+40	2	9.99999999999999e+39
λ_2	1.00000000000000D+20	1	0
λ_3	0.98000000000000D+20	2	-1.0000000936789517+20

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