

## SOME POSTULATES ON SPECTRUM OF EIGEN VALUES REGARDING MODIFIED POWER METHOD

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### ABSTRACT

*Eigen value problems arise naturally from a wide variety of scientific and engineering applications. There exist some smart methods in literature where only the Eigen values were found but not the corresponding Eigen vectors. Power method is very simple but a powerful tool for finding largest Eigen value and corresponding Eigen vector (Eigen-pair). On the other hand Inverse Power method is applied to find out smallest Eigen-pair and/or desire Eigen-pairs. But it is known that inverse power method is computationally very costly. Again by using shifting property in Power method, we can find further Eigen-pairs. Exploiting the shifting property in Power Method, we have proposed four postulates concerning Eigen spectrum. Moreover incorporating these postulates upon Power method, we have proposed a modified algorithm of power method. Typical examples are presented to demonstrate the validity of the postulates as well as the proposed algorithm. By exploiting the postulates, the modified algorithm of power method is able to find out both (largest and smallest) Eigen-pairs successfully in some relevant cases.*

**KeyWords:** *Eigen value, Eigen vector, Power method, Inverse Power method, Modified Algorithm of Power Method.*

### 1. INTRODUCTION

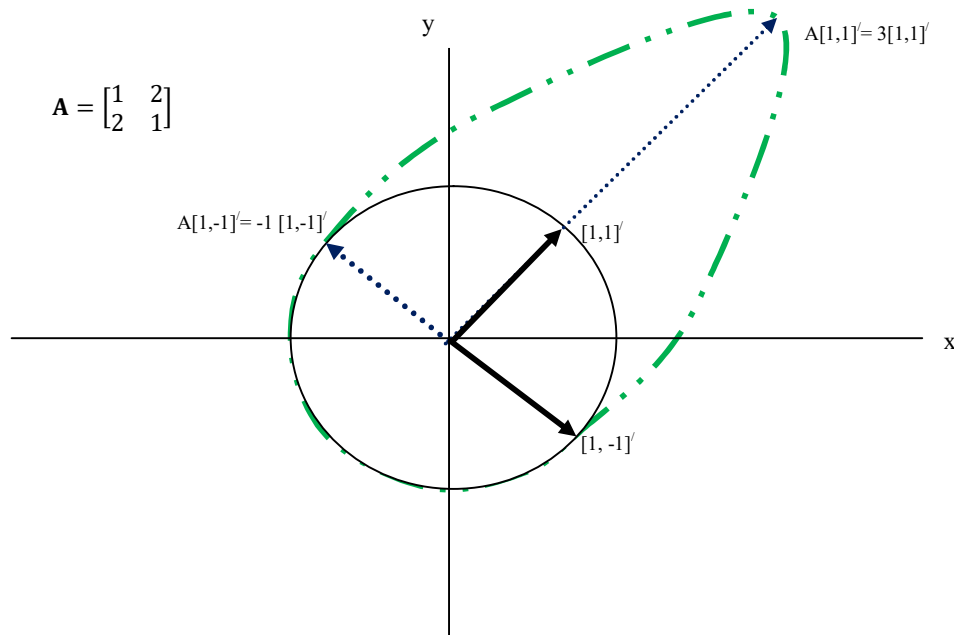
The theory and computation of Eigenvalue and Eigen vectors are the most successful and widely used tools of applied mathematics and scientific computing. Matrix Eigen value problems arise naturally from a wide variety of scientific and engineering applications such common applications including structural dynamics, quantum chemistry, quantum mechanics, electrical networks, control theory and design, material science, the vibrations of membranes, in the separation of variables for the problems of heat conduction or acoustics or in the hydrodynamic stability analysis (Chu and Golub 2002), (Mehrmann and Schroder, 2011). Acoustics, earthquake engineering, Markov chains, pattern recognition, graph theory, stability analysis, the dynamics of elastic bodies, the physics of rotating bodies, small oscillations of vibrating systems, system identification, seismic tomography, principal component analysis, exploration and remote sensing, antenna array processing, geophysics, molecular spectroscopy, particle physics, structure analysis, circuit theory, Hopfield neural networks, mechanical system simulation and many other areas (Michiels and Niculescu, 2008; Sachdev, 1999; Singh and Ram, 2002 and Unger. G, 2013).

Eigen values are a special set of scalars associated with a linear system of equations (i.e., a matrix equation) that are sometimes also known as characteristic roots, characteristic values, proper values, or latent roots (Hoffman and Kunze, 1971). Each Eigen value is paired with a corresponding Eigenvector. Mathematically, the Eigen value problem for square matrices  $\mathbf{A}$  is defined as follows

$$\mathbf{A}\mathbf{x} = \lambda\mathbf{x} \quad (1)$$

where  $\lambda$  is called Eigen value and  $\mathbf{x} (\neq 0)$  is called corresponding Eigen vector. The schematic view of

Eigen value problem is illustrated in Fig.1; where  $\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ . Then we have  $(\lambda, \mathbf{x}) = \{(3, [1, 1]^T), (-1, [1, -1]^T)\}$ .



**Figure 1:** Geometry of Eigen value Problem

Some efficient methods are available in the literature but they are able to find out only Eigen values. For finding corresponding Eigen vectors, much more efforts are necessary. The best known direct method, which is able to find all Eigen values, is the QR-Algorithm, based on the QR decomposition of a matrix which is also implemented in the Matlab function `eig()`. The QR Iteration algorithm for computing the Eigen values of a general matrix came from an elegantly simple idea that was proposed by Heinz Rutishauser in 1958 and refined by Francis in 1961-1962 (Jian, 2004 and Liao *et. al.* 2010). But this method is failed to find out corresponding Eigen vectors.

Power method is applicable in various field of application where only largest Eigen-pairs (Eigen value and corresponding Eigen vector) are important. On the other hand Inverse Power method is used to find out smallest Eigen-pair. Moreover using shifting property, Inverse Power method is also applicable to find out desire Eigen-pair. It is important to note that various areas of science and engineering seek multiple Eigen-pairs (largest as well as smallest Eigen-pairs) for the reasons other than algorithmic gains.

Besides Power method and Inverse Power method, there are several methods available in the literature for finding both Eigen values and Eigen vectors. Chang and Wong (2008) proposed several refinements of the Power method that enable the computation of multiple extremely Eigen-pairs of very large matrices by using Monte Carlo simulation method. Panju, 2011 examined some numerical iterative methods for computing the Eigen values and Eigen vectors of real matrices. He examined five methods – from the simple Power iteration method to the more complicated QR iteration method. The derivations, procedure, and advantages of each method are briefly discussed in that paper. Jamali and Sah Alam, 2015 proposed modified iterative method based on Power method and Inverse Power method for finding largest and smallest Eigen pairs.

## 2. PROPOSED POSTULATES

Before state the postulates, it is worthwhile to introduce some properties of Power method which are related with the proposed postulates.

**Property (a) (Shifting Property):** if  $(\lambda, \mathbf{x})$  be any Eigen-pair of equation (1) and  $\alpha$  be any scalar quantity, then  $(\lambda - \alpha)$  be an Eigen value of  $(\mathbf{A} - \alpha \mathbf{I})$ . So we have  $(\mathbf{A} - \alpha \mathbf{I})\mathbf{x} = (\lambda - \alpha)\mathbf{x}$  (2)

Here  $(\lambda - \alpha, \mathbf{x})$  be the Eigen-pair of the shifting Eigen problem (2).

**Property (b):** The Eigen value obtained by Power method is largest in magnitude. This Eigen value is denoted here as *firstEigen* value.

**Property (c):** Using shifting property, shifting with first(largest) Eigen value, the Power method is able to find out another Eigenvalue and corresponding Eigen vector. This Eigen value is termed as *secondEigen* value.

Now the postulates, which are associated with the Power method as well as modified (using shifting property) Power method, have been given below.

**Postulate 1.** If *first* (largest) Eigen value is positive and *secondEigen* value is also positive (produced by shifting largest one) then *firstEigen* value be the largest both in magnitude as well as in actual value. On the other hand the *secondEigen* value is the smallest both in magnitude as well as in actual value. In consequence, all Eigen values are positive in sign.

**Postulate 2.** If *first* (largest) Eigen value is positive and *secondEigen* value is negative (produced by shifting largest one) then the *firstEigen* value be the largest both in magnitude as well as in actual value. On the other hand the *secondEigen* value (obtained by the algorithm) is the smallest in actual value but it is the largest in magnitude among all negative Eigen values (if any). In consequence, some Eigen values along with largest Eigen value are positive and some Eigen values are negative in sign (if exist).

**Postulate 3.** If *first* (largest) Eigen value is negative and *secondEigen* value is also negative (produced by shifting largest one) then *firstEigen* value be the largest in magnitude but smallest in actual value. On the other hand the *secondEigen* value is the smallest in magnitude but largest in actual value. In consequence, all Eigen values are negative in sign.

**Postulate 4.** If *first* (largest in magnitude) Eigen value is negative and *secondEigen* value is positive (produced by shifting largest one) then *firstEigen* value be the largest in magnitude but smallest in actual value. On the other hand the *secondEigen* value is the largest in actual value and it is also largest among all positive Eigen values (if any). In consequence, some Eigen values along with largest Eigen values are negative in sign and some Eigen values are positive in sign (if exist).

### 3. POWER METHOD AND MODIFIED POWER METHOD

By exploiting the postulates upon the Power method, we will develop an algorithm to find out Eigen-pairs. As the proposed algorithm is developed on the basis of Power method, so it is worthwhile to present the algorithm of Power method. Table 1 represents the algorithm of Power method.

|                                      |                                                               |
|--------------------------------------|---------------------------------------------------------------|
| <b>Power Method ( ) {</b>            |                                                               |
| Step (1): read A                     |                                                               |
| set                                  | $\mathbf{y} = \mathbf{x}_0$                                   |
| set                                  | $\xi = \xi_0$                                                 |
| set                                  | $I_{\max}$                                                    |
| for $k = 1, 2, \dots, I_{\max}$ do { |                                                               |
| Step (2):                            | $\mathbf{v} = \mathbf{y} / \ \mathbf{y}\ _2$                  |
| Step (3):                            | $\mathbf{y} = \mathbf{A}\mathbf{v}$                           |
| Step (4):                            | $\theta = \mathbf{v}^* \mathbf{y}$                            |
| Step (5):                            | if $\ \mathbf{y} - \theta \mathbf{v}\ _2 \leq \xi  \theta $ , |
|                                      | set $(\lambda, \mathbf{x}) = (\theta, \mathbf{v})$            |
| else continue                        |                                                               |
| }end for                             |                                                               |
| Step (6):                            | accept $(\lambda, \mathbf{x}) = (\theta, \mathbf{v})$         |
| }                                    |                                                               |

**Table 1:** Algorithm of Power method

Now we will develop the modified power method by incorporating the shifting property. Actually in first iteration, the proposed algorithm runs the power method to find out the *firstlargest* Eigen value and corresponding Eigen vector. In second iteration, the algorithm uses shifting property where shifting parameter is the first Eigen value. Then, the algorithm again implement Power method on shifted Matrix and able to find out the second Eigen value which is off course the largest Eigen value of the shifted matrix and corresponding Eigen vector. Now transfer the second Eigen value with shifting parameter and this is being another (second) Eigen value of the original matrix but Eigen value of vector remains unchanged. Now algorithm compares the signs of the two Eigen values and then comment about the two Eigen values

as well as the sign of the spectrum of Eigen values according to the postulates. A pseudo code of the proposed modified algorithm of Power method is given below:

**Modified Algorithm of Power Method ( )**

```
{
  Step (1): read A
  Set B=A and  $\{\lambda, \mathbf{x}\} = \{\lambda_0, \mathbf{x}_0\}$ 
  for  $r=1, 2$  do {
    if  $r=1$  {
      Step (2): apply Power method ( )
                output  $\{\lambda_1, \mathbf{x}_1\}$ 
      Step (3) : find  $s_1$ , such that  $\lambda_1 = s_1 |\lambda_1|$ 
      Step (4): output  $\{\lambda_1, \mathbf{x}_1, s_1\}$ 
       $r = r+1$ 
    }
    else if  $r=2$  {
      Step (5): set  $\mathbf{B} = \mathbf{A} - \lambda_1 \mathbf{I}$ 
      Step (6): apply Power method ( )
                output  $\{\sigma_2, \mathbf{y}_2\}$ 
      Step (7) :  $\lambda_2 = \sigma_2 + \lambda_1$ 
                find  $s_2$ , such that  $\lambda_2 = s_2 |\lambda_2|$ 
      Step (8): output  $\{\lambda_2, \mathbf{x}_2, s_2\}$ 
    }
  }
  end for
  Step (9) : if ( $s_1 = s_2$  and  $> 0$ )
    {
      output :  $\{(\lambda_1, \mathbf{x}_1),$  // Largest (in magnitude) Eigen pair
               $(\lambda_2, \mathbf{x}_2),$  // Smallest (in magnitude) Eigen pair
              (all  $\lambda_i \geq 0\}$  // All Eigen values are positive
      Step (10) : Stop
    }
  else if ( $s_1 = s_2$  and  $< 0$ )
    {
      output :  $\{(\lambda_1, \mathbf{x}_1),$  // Largest (in magnitude) Eigen pair
               $(\lambda_2, \mathbf{x}_2),$  // Smallest (in magnitude) Eigen pair
              (all  $\lambda_i \leq 0\}$  // All Eigen values are negative
      Step (11) : Stop
    }
  else if ( $s_1 \neq s_2$  and  $s_1 > 0$ )
    {
      output :  $\{(\lambda_1, \mathbf{x}_1),$  // Largest (in magnitude) Eigen pair
               $(\lambda_2, \mathbf{x}_2),$  // Not smallest (in magnitude) Eigen pair
              (sign of all  $\lambda_i\}$  // Eigen values are both positive and negative
      Step (12) : Stop
    }
  else if ( $s_1 \neq s_2$  and  $s_1 < 0$ )
    {
      output :  $\{(\lambda_1, \mathbf{x}_1),$  // Largest (in magnitude) Eigen pair
               $(\lambda_2, \mathbf{x}_2),$  // Not smallest (in magnitude) Eigen pair
              (sign of all  $\lambda_i\}$  // Eigen values are both positive and negative
      Step (13) : Stop
    }
}
```

#### 4. NUMERICAL EXPERIMENTS

To examine the validity of the first postulate, we have considered the following typical example 1.

**Example 1:** 
$$\mathbf{A} = \begin{bmatrix} 4 & 3 & 1 & 0 \\ 3 & 4 & 2 & 1 \\ 1 & 2 & 7 & 1 \\ 1 & 1 & 1 & 8 \end{bmatrix}$$

At first, we have carried out an experiment on the proposed algorithm. After performing the experiment, the proposed algorithm has obtained the first Eigen-pair =  $(\mathbf{10.1375}, [0.498278, 0.709471, 0.929782, 1])'$  and the second Eigen-pair:  $(\mathbf{0.9125}, [-0.913116, 1, -0.180717, 0.0132391])'$ . Moreover, as the first Eigen value is positive and second Eigen value is also positive so according to the postulate 1 the second Eigen value should be the smallest Eigen value. Moreover as both first and second Eigen values are positive so all real Eigen values of the problem should be positive.

Now to justify the above argument of the postulate as well as to verify the proposed algorithm, we have solved the problem 1 by using MatLab solver to find out the spectrum of Eigen values. From MatLab, we have the following Eigenvalues: **10.1375**, 6.9741, 4.9759 and **0.9125**. It is observed that the experimental results agree with the argument of postulate 1. Moreover proposed algorithm is able to find out the desired results. That is the proposed algorithm is successfully able to find out both largest and smallest (in magnitude) Eigen values and corresponding Eigen vectors.

To investigate the validity of the second postulate as well as proposed algorithm, we have considered the following typical example 2.

**Example 2:** 
$$\mathbf{B} = \begin{bmatrix} 2 & 2 & 1 & 2 \\ 8 & 2 & 1 & 4 \\ 9 & 2 & 2 & 5 \\ 1 & 2 & 5 & 6 \end{bmatrix}$$

Again we have carried out numerical experiment on the proposed algorithm. After performing the experiment, the proposed algorithm has obtained the first Eigen-pair =  $(\mathbf{12.7568}, [0.418653, 0.77234, 0.958699, 1])'$  and second Eigen-pair =  $(-\mathbf{2.6332}, [-0.303505, 0.972362, 1, -0.769269])'$ . It is observed that the sign of first Eigen value is positive and the sign of the second Eigen value is negative. So first Eigen value should be positive largest Eigen value and according to the postulate 2, the second one should be the smallest Eigen value according to the sign but it should not be necessarily smallest in magnitude. Moreover as first one is positive and second one is negative so some Eigen values are positive and some Eigen values (if more) should be negative in sign.

Now to justify the above argument, we have solved the problem 2 by using MatLab solver to find out the spectrum of Eigen values. From MatLab, we have the following Eigen values: **12.7568**, 1.4718, 0.4045,  $-\mathbf{2.6332}$ . It is observed that the largest Eigen value is 12.7568 which is positive and smallest one is  $-2.6332$  which is not smallest in magnitude. Therefore the experimental results agree with the argument of postulate 2. It is noticed that the smallest (in magnitude) Eigen value is **0.4045**.

Now to justify the postulate 3 and to test the validation of the proposed algorithm we have considered the typical example 3.

**Example 3:** 
$$\mathbf{C} = \begin{bmatrix} -26 & -26 & -17 & -4 \\ -27 & -30 & -26 & -14 \\ -18 & -26 & -55 & -17 \\ -16 & -17 & -18 & -66 \end{bmatrix}$$

Again after carried out an experiment on the proposed algorithm, we have first Eigen-pair =  $(-\mathbf{102.7695}, [0.498278, 0.709471, 0.929782, 1])'$  and second Eigen-pair =  $(-\mathbf{0.8326}, [-0.913117, 1, -0.180716, 0.0132394])'$ . It is observed that the sign of first the Eigen value is negative and the sign of the second Eigen value is also negative. So first Eigen value should be negative but largest Eigen value in magnitude and according to the postulate 3, the second one should be the smallest Eigen value according to both in sign and in magnitude. Moreover as first one is negative and second one is also negative so according to the postulate 3, all real Eigen values should be negative in sign.

Now to test the validity of the postulate as well as algorithm, we have solved the problem 3 by using MatLab solver to find out the spectrum of Eigen values. From MatLab, we have the following Eigenvalues:  $-\mathbf{102.7695}$ ,  $-48.6385$ ,  $-24.7594$ ,  $-\mathbf{0.8326}$ . It is observed that the largest (in magnitude) Eigen value is  $-102.7695$  which is negative and the smallest one is  $-0.8326$  which is also negative and is the smallest in magnitude too. Therefore the experimental results agree with the argument of postulate 3. In the same time the proposed algorithm is able to find both the largest as well as the smallest (in magnitude) Eigen values and corresponding Eigen vectors successfully.

Now for the justification of the postulate 4 and to test the validity of the algorithm, we have considered the following typical example 4.

**Example 4:** 
$$\mathbf{D} = \begin{bmatrix} -6 & 2 & 1 & -2 \\ 8 & 2 & 1 & 4 \\ 9 & 2 & 2 & 5 \\ 1 & 7 & 5 & -6 \end{bmatrix}$$

Again we have carried out experiment on the proposed algorithm. After performing experiment, we have first Eigen-pair =  $(-12.4503, [0.57957, -0.554055, -0.6303, 1]^T)$  and second Eigen-pair =  $(7.9628, [0.0746226, 0.770871, 1, 0.749899]^T)$ . It is observed that the sign of first Eigen value is negative and the sign of the second Eigen value is positive. So first Eigen value should be largest (in magnitude) Eigen value and according to the postulate 4, the second one should be the largest Eigen value according to the sign but it should not be necessarily smallest and of course not be the largest in magnitude. Moreover as first one is negative and second one is positive so some Eigen value are positive and some Eigen values (if more) should be negative in sign.

Now to justify the above argument, we have solved the problem 4 by using MatLab solver to find out the spectrum of Eigen values. From MatLab, we have the following Eigenvalues:  $-12.4503, -4.1452, 0.6327, 7.9628$ . It is observed that the largest (in magnitude) Eigen value is  $-12.4503$  which is negative and the largest positive Eigen value one is  $7.9628$  which is not smallest in magnitude. Therefore the experimental results agree with the argument of postulate 4. It is noticed that the smallest (in magnitude) Eigen value is  $0.6327$ .

## 5. CONCLUSION

We have proposed four postulates as well as a modified algorithm based on Power iterative method and shifting property. All four postulates are verified by the corresponding typical instances. To investigate the effectiveness of the proposed algorithm, numerical experiments have been carried out. It is observed that proposed algorithm is able to find out the largest Eigen-pair and also smallest Eigen-pairs efficiently according to the relevant postulates 1 and 3. It is observed that in the case of all positive Eigen values as well as in the case of all negative Eigen values the modified Power method successfully is able to find out both largest (in magnitude) as well as smallest (in magnitude) Eigen-pairs. Moreover after finding second Eigen value by the modified Power method, it is easy to classify the nature (regarding sign) of the Eigen values. But it worthwhile to mention here that as the proposed algorithm is developed based on Power method so inherently it preserves some shortcoming of Power methods such as in the case of complex Eigen values.

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