

Power Method for Dominant Eigenvalue

Let $\lambda_1, \lambda_2, \lambda_3$, and λ_n be the eigenvalues of an $n \times n$ matrix \mathbf{A} . λ_1 is called the dominant eigenvalue of \mathbf{A} if

$$|\lambda_1| > |\lambda_i|, \quad i = 2, 3, \dots, n$$

The eigenvectors corresponding to λ_1 are called dominant eigenvectors of \mathbf{A} .

Procedure

1. Choose an $n \times n$ matrix

Otherwise-Pop up error – the number of rows and columns should be the same (or matrix dimension mismatched)

2. Like the Jacobi and Gauss-Seidel methods, the power method for approximating eigenvalues is iterative. First, we assume that matrix \mathbf{A} has a dominant eigenvalue with corresponding dominant eigenvectors. Then we choose an initial approximation \mathbf{x}_0 of one of the dominant eigenvectors of \mathbf{A} . This initial approximation must be a nonzero vector in \mathbb{R}^n

Finally, we form the sequence given by

$$\mathbf{x}_1 = \mathbf{A}\mathbf{x}_0$$

$$\mathbf{x}_2 = \mathbf{A}\mathbf{x}_1 = \mathbf{A}(\mathbf{A}\mathbf{x}_0) = \mathbf{A}^2\mathbf{x}_0$$

$$\mathbf{x}_3 = \mathbf{A}\mathbf{x}_2 = \mathbf{A}(\mathbf{A}^2\mathbf{x}_0) = \mathbf{A}^3\mathbf{x}_0$$

...

$$\mathbf{x}_n = \mathbf{A}\mathbf{x}_{n-1} = \mathbf{A}(\mathbf{A}^{n-1}\mathbf{x}_0) = \mathbf{A}^n\mathbf{x}_0$$

(In the above, \mathbf{x}_1 denotes the value of vector \mathbf{x} at the first iteration and so on)

Compare the updated value of \mathbf{x} with its previous value (obtained from the previous iteration)

For large powers of k , and by properly scaling this sequence, we will see that we obtain a good approximation of the dominant eigenvector of \mathbf{A} .

3. Repeat the iteration process until convergence

4. The formula for finding the corresponding eigenvalue from eigenvector \mathbf{x} .

If \mathbf{x} is an eigenvector of \mathbf{A} , then its corresponding eigenvalue is given by

$$\lambda = (\mathbf{Ax} \cdot \mathbf{x} / \mathbf{x} \cdot \mathbf{x})$$

5. If they do not converge even after many iterations (maybe after 1000 iterations), then show pop-up message

Otherwise-Pop up error – Entered matrix has no dominant eigenvalue

Example

$$\mathbf{A} = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix}$$

We begin with an initial nonzero approximation of

$$\mathbf{x}_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

We then obtain the following approximations

$$\mathbf{x}_1 = \mathbf{Ax}_0 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -10 \\ -4 \end{bmatrix} = -4 \begin{bmatrix} 2.50 \\ 1.00 \end{bmatrix}$$

$$\mathbf{x}_2 = \mathbf{Ax}_1 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} -10 \\ -4 \end{bmatrix} = \begin{bmatrix} 28 \\ 10 \end{bmatrix} = 10 \begin{bmatrix} 2.80 \\ 1.00 \end{bmatrix}$$

$$\mathbf{x}_3 = \mathbf{Ax}_2 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 28 \\ 10 \end{bmatrix} = \begin{bmatrix} -64 \\ -22 \end{bmatrix} = -22 \begin{bmatrix} 2.91 \\ 1.00 \end{bmatrix}$$

$$\mathbf{x}_4 = \mathbf{Ax}_3 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} -64 \\ -22 \end{bmatrix} = \begin{bmatrix} 136 \\ 46 \end{bmatrix} = 46 \begin{bmatrix} 2.96 \\ 1.00 \end{bmatrix}$$

$$\mathbf{x}_5 = \mathbf{Ax}_4 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 136 \\ 46 \end{bmatrix} = \begin{bmatrix} -280 \\ -94 \end{bmatrix} = -94 \begin{bmatrix} 2.98 \\ 1.00 \end{bmatrix}$$

$$\mathbf{x}_6 = \mathbf{Ax}_5 = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} -280 \\ -94 \end{bmatrix} = \begin{bmatrix} 568 \\ 190 \end{bmatrix} = 190 \begin{bmatrix} 2.99 \\ 1.00 \end{bmatrix}$$

Note that the approximations in Example appear to be approaching scalar multiples of $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$

So, the obtained dominant eigenvector from the above iterations is

$$\mathbf{x} = \begin{bmatrix} 3 \\ 1 \end{bmatrix}$$

Now, we'll find the corresponding eigenvalue from the obtained eigenvector

Formula

If \mathbf{x} is an eigenvector of \mathbf{A} , then its corresponding eigenvalue is given by

$$\lambda = (\mathbf{Ax} \cdot \mathbf{x} / \mathbf{x} \cdot \mathbf{x})$$

$$\mathbf{Ax} = \begin{bmatrix} 2 & -12 \\ 1 & -5 \end{bmatrix} \begin{bmatrix} 2.99 \\ 1.00 \end{bmatrix} = \begin{bmatrix} -6.02 \\ -2.01 \end{bmatrix}$$

$$\text{Then, } \mathbf{Ax} \cdot \mathbf{x} = \begin{bmatrix} -6.02 \\ -2.01 \end{bmatrix} \cdot \begin{bmatrix} 2.99 \\ 1.00 \end{bmatrix} = -20.0 \text{ (approx.)}$$

$$\text{And } \mathbf{x} \cdot \mathbf{x} = \begin{bmatrix} 2.99 \\ 1.00 \end{bmatrix} \cdot \begin{bmatrix} 2.99 \\ 1.00 \end{bmatrix} = 9.94 \text{ (approx.)}$$

So, the corresponding eigenvalue, $\lambda = (-20.0 / 9.94) = -2$ (approx.)

Some more discussion about eigenvalue & eigenvector

Let's assume a square matrix \mathbf{A}

The characteristic equation,

$$|\mathbf{A} - \lambda \mathbf{I}| = 0$$

(where \mathbf{I} is an identity matrix)

After calculating the values of λ s we attempt to find eigenvectors for corresponding eigenvalues like this

For eigenvalue, $\lambda = \lambda_1$

$$\mathbf{A} \cdot \mathbf{x} = \lambda_1 \cdot \mathbf{I} \cdot \mathbf{x} \quad (\text{where, } \mathbf{x} \text{ is an unknown vector})$$

$$\text{Or, } (\mathbf{A} - \lambda_1 \cdot \mathbf{I}) \cdot \mathbf{x} = 0$$

The value of \mathbf{x} is the corresponding eigenvector of λ_1

We've already discussed the topic of 'eigenvalue and eigenvector' with the proper example in 'Jordan Decomposition'. Please go thru it if you want.