

Eigenvalue power method → used to find largest eigenvalue and its eigenvector

Suppose A is an $n \times n$ matrix with n real distinct eigenvalues.

now, order eigenvalues by magnitude (with their eigenvector)

$$|\lambda_1| > |\lambda_2| > |\lambda_3| > \dots > |\lambda_n| \rightarrow \lambda_1 \text{ is largest (dominant)}$$

$$e_1 \quad e_2 \quad e_3 \quad \dots \quad e_n \text{ — eigenvector (linearly independent)}$$

Start with any initial vector x_0 (can be any).

any vector can be written as a combination of the eigenvectors: $x_0 = c_1 e_1 + c_2 e_2 + \dots + c_n e_n$

$$\therefore x_0 = \sum_{i=1}^n c_i e_i = c_1 e_1 + c_2 e_2 + c_3 e_3 + \dots + c_n e_n$$

$$\text{Now multiply by } A: x_1 = A x_0 = A \sum_{i=1}^n c_i e_i = \sum_{i=1}^n (c_i A e_i)$$

recall $Ax = \lambda x$ where x is an eigenvector

$$\therefore \sum_{i=1}^n c_i A e_i = \sum_{i=1}^n c_i \lambda_i e_i$$

now, if we multiply x_0 by A^p we have:

$$x_p = A^p x_0 = \sum_{i=1}^n c_i A^p e_i = \sum_{i=1}^n c_i \lambda_i^p e_i$$

\therefore since λ_1 is the largest eigenvalue, so λ_1^p is the largest number.

Factoring out λ_1 term:

$$x_p = c_1 \lambda_1^p e_1 + \sum_{i=2}^n c_i \lambda_i^p e_i$$

$$\text{or } x_p = \lambda_1^p \left[c_1 e_1 + \sum_{i=2}^n c_i \left(\frac{\lambda_i}{\lambda_1} \right)^p e_i \right]$$

These terms tend to 0 as λ_i in denominator is largest eigenvalue

$$\therefore x_p \approx \lambda_1^p c_1 e_1$$

multiply this by A again to get:

$$x_{p+1} = \lambda_1^{p+1} c_1 e_1$$

$$\text{or } \lambda_1 = \frac{x_p^T x_{p+1}}{x_p^T x_p} \quad \left. \vphantom{\frac{x_p^T x_{p+1}}{x_p^T x_p}} \right\} \text{ This is to get rid of } c_1 \text{ and } e_1 \text{ and get back the dominant } \lambda_1.$$

$e_1 = x_{p+1} \rightarrow$ because they are the same basic vectors (this x_p is normalized)

Note - overflow can happen $\lambda_1 > 1$ and is multiplied too many times by itself and becomes larger than the largest machine number.

OR there can be underflow when $\lambda_1 < 0$ and too many multiplications by itself can cause underflow (number is lower than real min).

We can avoid this by normalizing x_p :

$$x_p \text{ is saved with each iteration as } x_p = \frac{x_p}{(x_p^T x_p)^{1/2}}$$

Power method example

$$A = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \longrightarrow \lambda_1 = 10, v_1 = \begin{bmatrix} 5/4 \\ 1 \end{bmatrix} \oplus \lambda_2 = 1, v_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

First we need an initial vector to multiply by A

$$x_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\therefore x_1 = Ax_0$$

$$\text{or, } x_1 = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$$

now continue to multiply the resultant vectors by A :

$$x_2 = Ax_1$$

$$\text{or, } x_2 = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 6 \\ 4 \end{bmatrix} = \begin{bmatrix} 56 \\ 44 \end{bmatrix}$$

$$x_3 = Ax_2$$

$$x_3 = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 56 \\ 44 \end{bmatrix} = \begin{bmatrix} 556 \\ 444 \end{bmatrix}$$

$$x_4 = Ax_3 = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 556 \\ 444 \end{bmatrix} = \begin{bmatrix} 5556 \\ 4444 \end{bmatrix}$$

$$x_5 = Ax_4 = \begin{bmatrix} 6 & 5 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} 5556 \\ 4444 \end{bmatrix} = \begin{bmatrix} 55556 \\ 44444 \end{bmatrix}$$

$$\therefore \lambda_1 = \frac{x_4^T x_5}{x_4^T x_4} = \frac{506,178,271}{50,618,272} \approx 10$$

$$\therefore \text{Eigenvector } e_1 = x_p$$

$$\text{or, } e_1 = x_5 = \begin{bmatrix} 55556/44444 \\ 1 \end{bmatrix} \approx \begin{bmatrix} 5/4 \\ 1 \end{bmatrix}$$

Another example : Use the power method to find dominant eigenvalue of : $A = \begin{bmatrix} -5 & 6 \\ 5 & -4 \end{bmatrix}$

Solution: start with an initial vector $x_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$\therefore x_1 = Ax_0 = \begin{bmatrix} -5 & 6 \\ 5 & -4 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} -5 \\ 5 \end{bmatrix}$$

$$x_2 = Ax_1 = \begin{bmatrix} -5 & 6 \\ 5 & -4 \end{bmatrix} \begin{bmatrix} -5 \\ 5 \end{bmatrix} = \begin{bmatrix} 35 \\ -45 \end{bmatrix} \quad \therefore \lambda_1 = \frac{x_4^T x_5}{x_4^T x_4}$$

$$x_3 = Ax_2 = \begin{bmatrix} -545 \\ 455 \end{bmatrix}$$

$$x_4 = Ax_3 = \begin{bmatrix} 5455 \\ -4545 \end{bmatrix}$$

$$x_5 = \begin{bmatrix} -54545 \\ 45455 \end{bmatrix}$$

$$\text{or, } \lambda_1 = \frac{\begin{bmatrix} 54545 & 45455 \end{bmatrix} \begin{bmatrix} -54545 \\ 45455 \end{bmatrix}}{\begin{bmatrix} 54545 & 45455 \end{bmatrix} \begin{bmatrix} 5455 \\ -4545 \end{bmatrix}}$$

$$\text{or, } \lambda_1 = \frac{5041314050}{-504135950} \approx -10$$

$$\therefore e_1 = x_5 (\text{normalized})$$

$$\therefore e_1 = \begin{bmatrix} -1.2 \\ 1 \end{bmatrix}$$