

$$Ax = \lambda x \rightarrow \text{eigenvectors.}$$

\downarrow
 eigenvalues eigenpairs.

\Rightarrow normalized eigenvectors.

$$\|x\|_2 = 1, \quad (\text{numpy.linalg.eig}$$

using $p=2$ norm).

$$\underline{\det(A - \lambda I) = 0.} \quad / \quad A - \lambda I \text{ is singular.}$$

we want to approximate the eigenvalues numerically.

a $n \times n$ matrix A with n linearly independent eigenvectors U is said to be diagonalizable.

$$\begin{cases} Au_1 = \lambda_1 u_1 \\ Au_2 = \lambda_2 u_2 \\ \dots \\ Au_n = \lambda_n u_n \end{cases}
 \quad
 \begin{bmatrix} A \end{bmatrix}
 \underbrace{\begin{bmatrix} | & | & \dots & | \\ u_1 & u_2 & \dots & u_n \\ | & | & \dots & | \end{bmatrix}}_U
 =
 \begin{bmatrix} | & | & \dots & | \\ \lambda_1 u_1 & \lambda_2 u_2 & \dots & \lambda_n u_n \\ | & | & \dots & | \end{bmatrix}$$

$$=
 \underbrace{\begin{bmatrix} | & | & \dots & | \\ u_1 & u_2 & \dots & u_n \\ | & | & \dots & | \end{bmatrix}}_U
 \underbrace{\begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}}_{ID.}$$

$$AU = U ID. \Rightarrow \underline{A = U ID U^{-1}} \quad \text{i.e. } A \text{ is similar to } ID$$

defective (not diagonalizable).

if $n \times n$ symmetric matrix A has n distinct eigenvalues then A is diagonalizable.

Propose a vector x is lin comb of eigenvectors.

$$x = \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n.$$

$$Ax = A\alpha_1 u_1 + A\alpha_2 u_2 + \dots + A\alpha_n u_n$$

$$= \lambda_1 \alpha_1 u_1 + \lambda_2 \alpha_2 u_2 + \dots + \lambda_n \alpha_n u_n.$$

(Assume $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n|$).

Goal is to find an eigenvector u_i of A

$$x_0 = \alpha_1 u_1 + \alpha_2 u_2 + \dots + \alpha_n u_n.$$

$$x_1 = Ax_0 = \lambda_1 \alpha_1 u_1 + \lambda_2 \alpha_2 u_2 + \dots + \lambda_n \alpha_n u_n.$$

$$x_2 = Ax_1 = \lambda_1^2 \alpha_1 u_1 + \lambda_2^2 \alpha_2 u_2 + \dots + \lambda_n^2 \alpha_n u_n.$$

$$x_k = Ax_{k-1} = \alpha_1 \lambda_1^k u_1 + \alpha_2 \lambda_2^k u_2 + \dots + \alpha_n \lambda_n^k u_n.$$

Power Iteration: \Downarrow

$$x_k = (\lambda_1)^k \left[\alpha_1 u_1 + \alpha_2 \left(\frac{\lambda_2}{\lambda_1} \right)^k u_2 + \dots + \alpha_n \left(\frac{\lambda_n}{\lambda_1} \right)^k u_n \right]$$

Since $|\lambda_1| > |\lambda_2|$, we have $\left(\frac{\lambda_2}{\lambda_1} \right)^k \ll 1$ when k is large.

\Rightarrow Hence, as k increases, x_k converges to a multiple of the first eigenvector u_1 , i.e.

$$k \rightarrow +\infty \Rightarrow x_k \approx \lambda_1^k \alpha_1 u_1$$