

Eigenvectors and Eigenvalues

(21)

- if A is $n \times n$, a nonzero $\vec{x}$ s.t. for some scalar λ (possibly complex — even if A real) we have $A\vec{x} = \lambda\vec{x}$ is called an eigenvector of A ; λ is associated eigenvalue.

- e.g. $\begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 3 \end{bmatrix} = 3 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$

so $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is eigenvector associated to $\lambda = 3$

$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ -i \end{bmatrix} = \begin{bmatrix} i \\ 1 \end{bmatrix} = i \begin{bmatrix} 1 \\ -i \end{bmatrix}$$

so $\begin{bmatrix} 1 \\ -i \end{bmatrix}$ " " $\lambda = i$

How many eigenthings?

- A always has n eigenvalues if you count multiplicity (possible repeated eigenvalues in char. polynomial)

- usually A has n distinct eigenvalues

- eigenvectors assoc. to distinct eigenvalues always independent (can prove by induction)

- a repeated eigenvalue $\lambda_1 = \lambda_2$ might or might not have 2 independent eigenvectors.

- eigenvectors/values help us understand dynamics of A :

- usually n indep. eigenvectors $\vec{x}_1, \dots, \vec{x}_n$ assoc. to $\lambda_1, \dots, \lambda_n$

- then any $\vec{v} \in \mathbb{R}^n$ is linear combo:

$$\vec{v} = c_1 \vec{x}_1 + \dots + c_n \vec{x}_n$$

- so: $A\vec{v} = c_1 \lambda_1 \vec{x}_1 + \dots + c_n \lambda_n \vec{x}_n$

$$A^k \vec{v} = c_1 \lambda_1^k \vec{x}_1 + \dots + c_n \lambda_n^k \vec{x}_n$$

- so: if we knew n (indep.) eigenvectors for A and assoc. eigenvalues, knew all there is to know about A (and powers of A !)

- even more, in this situation can diagonalize A . What's this mean?

- put indep. eigenvectors as columns of matrix $X = [\vec{x}_1 | \dots | \vec{x}_n]$

- then $AX = [A\vec{x}_1 | \dots | A\vec{x}_n] = [\lambda_1 \vec{x}_1 | \dots | \lambda_n \vec{x}_n]$

$$= [\vec{x}_1 | \dots | \vec{x}_n] \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

(zeros)

$= XD$ ^{diagonal} with eigenvalues on main diagonal.

- X is invertible (since $\vec{x}_1, \dots, \vec{x}_n$ indep. by hypothesis)

- So: $AX = XD \Rightarrow A = XDX^{-1}$

↳ Now: easy to compute powers of A

$$A^2 = XDX^{-1}XDX^{-1} = XD^2X^{-1}$$

$$A^k = XD^kX^{-1}$$

← squares of eigenvalues on diag.

- Conversely: If $A = XDX^{-1}$ for some diagonal D , then reversing above we see: entries on diag of D are eigenvalues and columns of X are assoc. (indep.) eigenvectors.

We've proved:

Theorem An $n \times n$ matrix A is diagonalizable,

i.e. $A = XDX^{-1}$ for some diagonal D ,

(iff A has n independent eigenvectors)

In this case, eigenvalues of A appear on diagonal of D .