

3.3 Diagonalization and Eigenvalues

- Eigenvalues and Eigenvectors

Def. An $n \times n$ matrix D is called a *diagonal matrix* if D has the form

$$D = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \cdots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} = \text{diag}(\lambda_1, \lambda_2, \cdots, \lambda_n).$$

Observations:

1. If $A = PBP^{-1}$ then $A^k = PB^kP^{-1}$ for $k = 1, 2, \dots$.
2. If $D = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$ is a diagonal matrix, then

$$D^k = \text{diag}(\lambda_1^k, \lambda_2^k, \dots, \lambda_n^k).$$

These observations suggest an efficient way to calculate A^k . More precisely, if we can find a **diagonal** matrix D s.t. $A = PDP^{-1}$, then $A^k = PD^kP^{-1}$.

Def. $A : n \times n$ matrix

A number λ is called an **eigenvalue** of A if \exists a nonzero $X \in \mathbf{F}^n$ s.t.

$$AX = \lambda X,$$

and such an X is called an **eigenvector** of A corresponding to the eigenvalue λ , or **λ -eigenvector** for short.

Eg. If $A = \begin{bmatrix} 3 & 5 \\ 1 & -1 \end{bmatrix}$ and $X = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$, then $AX = 4X$. Thus

$\lambda = 4$ is an eigenvalue of A and $X = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$ is an eigenvector of A corresponding to $\lambda = 4$.

Question : How can we get all the eigenvalues and eigenvectors of a matrix?

Def. $A : n \times n$ matrix.

The *characteristic polynomial* $c_A(x)$ of A is defined by

$$c_A(x) = \det(xI - A).$$

A characteristic polynomial is indeed a polynomial of degree n .

Thm. $A : n \times n$ matrix.

1. The *eigenvalues* λ of A are the *roots* of $c_A(x)$.
2. The λ -*eigenvectors* X are the nonzero *solutions* of

$$(\lambda I - A)X = O.$$

Proof. Assume that λ is an eigenvalue of A with an eigenvector X . Then $AX = \lambda X = \lambda IX$ and $(\lambda I - A)X = O$. Hence, X is a nontrivial solution of $(\lambda I - A)X = O$. Since the equation has a nontrivial solution, we have

$$\det(\lambda I - A) = c_A(\lambda) = 0$$

and so λ is a root of the characteristic polynomial.

Conversely, a similar argument shows that a root λ of $c_A(x)$ is indeed an eigenvalue and that a solution of $(\lambda I - A)X = O$ is an eigenvector. \square

Eg. Consider $\begin{bmatrix} 3 & 5 \\ 1 & -1 \end{bmatrix}$. Then

$$xI - A = \begin{bmatrix} x & 0 \\ 0 & x \end{bmatrix} - \begin{bmatrix} 3 & 5 \\ 1 & -1 \end{bmatrix} = \begin{bmatrix} x - 3 & -5 \\ -1 & x + 1 \end{bmatrix},$$

$c_A(x) = \det(xI - A) = x^2 - 2x - 8 = (x - 4)(x + 2)$,
and so $\lambda_1 = 4$ and $\lambda_2 = -2$ are eigenvalues.

i) $\lambda_1 = 4$

$$(\lambda_1 I - A)X = \begin{bmatrix} 1 & -5 \\ -1 & 5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

$$\begin{bmatrix} 1 & -5 \\ -1 & 5 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & -5 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = t \begin{bmatrix} 5 \\ 1 \end{bmatrix} \quad (t \neq 0).$$

$$ii) \lambda_2 = -2$$

$$\lambda_2 I - A = \begin{bmatrix} -5 & -5 \\ -1 & -1 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = t \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad (t \neq 0).$$

Eg.

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 1 & 0 & 1 \\ 4 & -4 & 5 \end{bmatrix}, \quad xI - A = \begin{bmatrix} x-1 & -2 & 1 \\ -1 & x & -1 \\ -4 & 4 & x-5 \end{bmatrix},$$

$c_A(x) = \det(xI - A) = x^3 - 6x^2 + 11x - 6 = (x-1)(x-2)(x-3)$,
and so $\lambda_1 = 1$, $\lambda_2 = 2$, $\lambda_3 = 3$ are eigenvalues.

$$i) \lambda_1 = 1$$

$$\lambda_1 I - A = \begin{bmatrix} 0 & -2 & 1 \\ -1 & 1 & -1 \\ -4 & 4 & -4 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & -\frac{1}{2} \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \\ 1 \end{bmatrix}.$$

$$ii) \lambda_2 = 2$$

$$\lambda_2 I - A = \begin{bmatrix} 1 & -2 & 1 \\ -1 & 2 & -1 \\ -4 & 4 & -3 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{2} \\ 0 & 1 & -\frac{1}{4} \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{4} \\ 1 \end{bmatrix}.$$

iii) $\lambda_3 = 3$

$$\lambda_3 I - A = \begin{bmatrix} 2 & -2 & 1 \\ -1 & 3 & -1 \\ -4 & 4 & -2 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{4} \\ 0 & 1 & -\frac{1}{4} \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} -\frac{1}{4} \\ \frac{1}{4} \\ 1 \end{bmatrix}.$$

Eg.

$$A = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \quad c_A(x) = \det(xI - A) = \begin{vmatrix} x & -1 \\ 1 & x \end{vmatrix} = x^2 + 1.$$

If $\mathbf{F} = \mathbf{R}$, A has no eigenvalue.

If $\mathbf{F} = \mathbf{C}$, A has eigenvalues $\pm i$.

Eg. Consider an upper triangular matrix

$$A = \begin{bmatrix} \lambda_1 & * & \cdots & * \\ 0 & \lambda_2 & \cdots & * \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}.$$

What are the eigenvalues of A ?

Eg. A and A^T have the same eigenvalues. Indeed,

$$\begin{aligned} c_{A^T}(x) &= \det(xI - A^T) = \det(xI^T - A^T) \\ &= \det[(xI - A)^T] = \det(xI - A) = c_A(x). \end{aligned}$$

- Diagonalization

Def. An $n \times n$ matrix A is called *diagonalizable* if \exists an invertible P s.t. $P^{-1}AP = D$ is diagonal.

Assume that $P^{-1}AP = D$. Write $P = [X_1 \ X_2 \ \cdots \ X_n]$ and $AP = PD$.

$$AP = A [X_1 \ X_2 \ \cdots \ X_n] = [AX_1 \ AX_2 \ \cdots \ AX_n].$$

$$\begin{aligned} PD &= [X_1 \ X_2 \ \cdots \ X_n] \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix} \\ &= [\lambda_1 X_1 \ \lambda_2 X_2 \ \cdots \ \lambda_n X_n]. \end{aligned}$$

Comparing columns, we have

$$AX_i = \lambda_i X_i \text{ for each } i.$$

Thus the diagonal entries of D are eigenvalues of A , and the columns of P are corresponding eigenvectors.

Conversely, assume that we have $AX_i = \lambda_i X_i$ for each i with $P = [X_1 \ X_2 \ \cdots \ X_n]$ invertible. Write $D = \text{diag}(\lambda_1, \lambda_2, \cdots, \lambda_n)$. It's easy to see that $AP = PD$ and $P^{-1}AP = D$.

Thm. $A : n \times n$ matrix.

A is *diagonalizable* if and only if it has *eigenvectors* X_1, X_2, \cdots, X_n s.t. $P = [X_1 \ X_2 \ \cdots \ X_n]$ is *invertible*. In this case, $P^{-1}AP = D = \text{diag}(\lambda_1, \lambda_2, \cdots, \lambda_n)$, where λ_i is the *eigenvalue* of A corresponding to X_i .

Eg.

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 1 & 0 & 1 \\ 4 & -4 & 5 \end{bmatrix}, \quad \lambda_1 = 1, \quad X_1 = \begin{bmatrix} -1 \\ 1 \\ 2 \end{bmatrix};$$

$$\lambda_2 = 2, \quad X_2 = \begin{bmatrix} -2 \\ 1 \\ 4 \end{bmatrix}; \quad \lambda_3 = 3, \quad X_3 = \begin{bmatrix} -1 \\ 1 \\ 4 \end{bmatrix}.$$

$$P = \begin{bmatrix} -1 & -2 & -1 \\ 1 & 1 & 1 \\ 2 & 4 & 4 \end{bmatrix}, \quad P^{-1}AP = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}.$$

Eg.

$$A = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}, \quad c_A(x) = \begin{vmatrix} x & -1 & -1 \\ -1 & x & -1 \\ -1 & -1 & x \end{vmatrix} = (x-2)(x+1)^2.$$

The eigenvalues are $\lambda_1 = 2$ and $\lambda_2 = -1$.

i) $\lambda_1 = 2$

$$\lambda_1 I - A = \begin{bmatrix} 2 & -1 & -1 \\ -1 & 2 & -1 \\ -1 & -1 & 2 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = t \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}.$$

$$ii) \lambda_2 = -1$$

$$\lambda_2 I - A = \begin{bmatrix} -1 & -1 & -1 \\ -1 & -1 & -1 \\ -1 & -1 & -1 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix},$$

$$\begin{cases} x_1 + x_2 + x_3 = 0 \\ x_2 = s \\ x_3 = t \end{cases}, \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = s \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}.$$

$$P = \begin{bmatrix} 1 & -1 & -1 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}, \quad P^{-1}AP = \begin{bmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -1 \end{bmatrix}.$$

Eg. $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$, $c_A(x) = (x - 1)^2$, and so we have one eigenvalue $\lambda = 1$.

$$\lambda I - A = \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = t \begin{bmatrix} 1 \\ 0 \end{bmatrix}.$$

A is *not* diagonalizable.

Eg. Let $A = \begin{bmatrix} 1 & 2 \\ 3 & 2 \end{bmatrix}$. Compute A^{100} .

$$c_A(x) = \begin{bmatrix} x - 1 & -2 \\ -3 & x - 2 \end{bmatrix} = (x - 4)(x + 1).$$

The eigenvalues are $\lambda_1 = 4$ and $\lambda_2 = -1$.

$$i) \lambda_1 = 4$$

$$\lambda_1 I - A = \begin{bmatrix} 3 & -2 \\ -3 & 2 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & -\frac{2}{3} \\ 0 & 0 \end{bmatrix},$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = t \begin{bmatrix} \frac{2}{3} \\ 1 \end{bmatrix}, \quad X_1 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}.$$

$$ii) \lambda_2 = -1$$

$$\lambda_2 I - A = \begin{bmatrix} -2 & -2 \\ -3 & -3 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix},$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = t \begin{bmatrix} -1 \\ 1 \end{bmatrix}, \quad X_2 = \begin{bmatrix} -1 \\ 1 \end{bmatrix}.$$

$$P = \begin{bmatrix} 2 & -1 \\ 3 & 1 \end{bmatrix}, \quad P^{-1} = \frac{1}{5} \begin{bmatrix} 1 & 1 \\ -3 & 2 \end{bmatrix}, \quad P^{-1}AP = D = \begin{bmatrix} 4 & 0 \\ 0 & -1 \end{bmatrix}.$$

$$\begin{aligned} A^{100} &= PD^{100}P^{-1} = \begin{bmatrix} 2 & -1 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} 4^{100} & 0 \\ 0 & 1^{100} \end{bmatrix} \frac{1}{5} \begin{bmatrix} 1 & 1 \\ -3 & 2 \end{bmatrix} \\ &= \frac{1}{5} \begin{bmatrix} 2 \cdot 4^{100} & -1 \\ 3 \cdot 4^{100} & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ -3 & 2 \end{bmatrix} \\ &= \frac{1}{5} \begin{bmatrix} 2 \cdot 4^{100} + 3 & 2 \cdot 4^{100} - 2 \\ 3 \cdot 4^{100} - 3 & 3 \cdot 4^{100} + 2 \end{bmatrix}. \end{aligned}$$