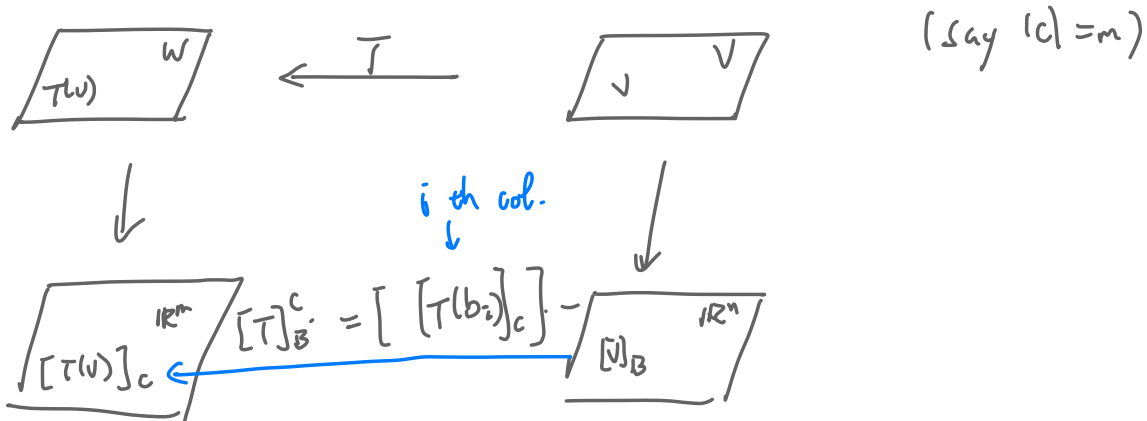


Last time: · mentioned the matrix $[T]_B^C$ of a linear map $T: V \rightarrow W$
w.r.t. a basis $B = \{b_1, b_2, \dots, b_n\}$ of V and a basis C of W .



Today : · Start Ch5. Diagonalization : motivation and strategy.

1. Two motivating examples

$$\begin{bmatrix} -1 & 3 \\ 3 & 1 \end{bmatrix} \quad \begin{bmatrix} (-1/10 \cdot (-7))^k & 3/10 \cdot (-7)^k \\ 3/10 \cdot 3^k & 1/10 \cdot 3^k \end{bmatrix}$$

a) Computing matrix powers

Eg. Find a closed formula for $\begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}^k$ \rightarrow hard ...

Useful Fact: Find a closed formula for $\begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}^k$ where k is an arbitrary pos. int.

$$\text{Useful Fact: } \underbrace{\begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}}_A = \underbrace{\begin{bmatrix} -1 & 3 \\ 3 & 1 \end{bmatrix}}_P \underbrace{\begin{bmatrix} -7 & 0 \\ 0 & 3 \end{bmatrix}}_{D, \text{ diagonal}} \underbrace{\begin{bmatrix} -1/10 & 3/10 \\ 3/10 & 1/10 \end{bmatrix}}_{P^{-1}}$$

Thus, $A^k = (PDP^{-1})^k = \underbrace{PDP^{-1}}_{\text{easy}} \underbrace{PDP^{-1}}_{\text{easy}} \dots \underbrace{PDP^{-1}}_{\text{easy}} \text{ (k copies)}$

$$= P \underbrace{D^k}_{\text{easy}} P^{-1} = \begin{bmatrix} -1 & 3 \\ 3 & 1 \end{bmatrix} \cdot \begin{bmatrix} (-7)^k & 0 \\ 0 & 3^k \end{bmatrix} \cdot \begin{bmatrix} -1/10 & 3/10 \\ 3/10 & 1/10 \end{bmatrix} = \frac{1}{10} \begin{bmatrix} (-7)^k + 3^{k+2} & \dots \\ -3 \cdot (-7)^k + 3^{k+1} & \dots \end{bmatrix}$$

Point: Having A in the form $A = PDP^{-1}$ for D diagonal and P inv. is helpful for computing A^k .

b) "Nice" matrix representations of linear maps

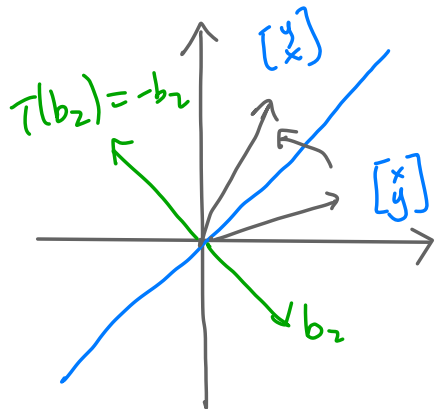
E.g. Consider the map $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ given by refl. w.r.t the line $y=x$.

Consider the two bases $B_1 = \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$ $B_2 = \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\}$ of \mathbb{R}^2 .

Then T is linear, and

$$[T]_{B_1}^{B_1} = \left[[T(e_1)]_{B_1} \mid [T(e_2)]_{B_1} \right] = \left[\begin{bmatrix} -1 \\ 0 \end{bmatrix} \mid \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right]_{B_1}$$

Fact: if B is the standard basis, then $[T]_B^B$ is the usual standard = $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ ← standard matrix



$$[T]_{B_2}^{B_2} = \left[[T(b_1)]_{B_2} \mid [T(b_2)]_{B_2} \right] = \left[\begin{bmatrix} 1 \\ 1 \end{bmatrix} \mid \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right]_{B_2} = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \leftarrow \text{considered "nicer" since it's diagonal}$$

Both the above examples are related to the problem of diagonalizing square matrices (We'll explain the relationship in more details later).

Def: The diagonalization of a square matrix A refers to the process of trying to find a diagonal matrix D and an invertible matrix P s.t. $A = P D P^{-1}$.

Fact: Such a pair (D, P) does not always exist. If it does, we also refer to the decomp. $A \stackrel{*}{=} P D P^{-1}$ as a diagonalization of A .

$$\text{e.g. } \begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix} = \begin{bmatrix} -1 & 3 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} -7 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} -1/10 & 3/10 \\ 3/10 & 1/10 \end{bmatrix}$$

↓

To get $*$, we'll try to get the cols of P & the diag entries of D .

- We'll get the diagonal entries of D as "eigenvalues of A ".
- We'll get the cols of P as "eigenvectors of A ".

2. Def of eigenvalues and eigenvectors

Def. Let A be an $n \times n$ matrix. A scalar $\lambda \in \mathbb{R}$ is called an eigenvalue of A if there is a nonzero vector $v \in \mathbb{R}^n$ s.t. $Av = \lambda v$. In this case, i.e., if λ is an eigenvalue of A , then any vector w s.t. $Aw = \lambda w$ is called an eigenvector of A w/ eigenvalue λ .

Eg. Let $A = \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}$, $v = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, $w = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$, $u = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$

$$Av = \begin{bmatrix} 3 \\ 6 \end{bmatrix} = 3 \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 3 \cdot v, \quad Aw = \begin{bmatrix} 0 \\ 0 \end{bmatrix} = 0 \cdot w, \quad Au = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \neq \lambda \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

So v is an e-vec of A w/ e-value 3, w is an e-vec of A w/ e-value 0, u is not an e-vec of A . In particular, 3 and 0 are e-values of A .

3. Finding eigenvalues via characteristic polynomials

Def. (characteristic polynomial) For every $n \times n$ matrix A , the characteristic polynomial of A is the polynomial (in x)

$$\chi(A) = \det(A - xI_n).$$

E.g. $\chi\left(\begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix}\right) = \det\left(\begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} - \begin{bmatrix} x & 0 \\ 0 & x \end{bmatrix}\right) = \det\left(\begin{bmatrix} 1-x & 1 \\ 2 & 2-x \end{bmatrix}\right)$

$$= (1-x)(2-x) - 1 \cdot 2 = x^2 - 3x + 2 - 2 = x^2 - 3x$$

Note, $(x^2 - 3x) = x(x-3)$, so the roots of the char. poly. (i.e., the solns of $\chi(A) = 0$) are 0 and 3.

Why is the char. poly. important ?

Thm 1. The eigenvalues of A are precisely the roots of the characteristic poly for any square matrix A !

Ex. 1. Find all e-values of $A = \begin{bmatrix} 2 & 3 \\ 3 & -6 \end{bmatrix}$.

Ex. 2. Find all e-values of $B = \begin{bmatrix} 1 & 3 & 3 \\ -3 & -5 & -3 \\ 3 & 3 & 1 \end{bmatrix}$, assuming 1 is an e-value.

Next time:

- Ex. 2.
- proof of Thm 1.
- finding eigenvectors.