

Applications of $A = SDS^{-1}$

Linear $f: V \rightarrow V$, $A_f =$ any matrix representation of f (for a basis $\{e_i\}$)

If eigenvalues of f are different, $\exists D$ diagonal, $\exists S$ invertible \iff eigenvalue $\{e_i\}$

(same as those of A_f)

$$S = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

s.t. $A = SDS^{-1}$

How can we use $A = SDS^{-1}$

① We want $\det(A)$: we know $\det(XY) = \det(X)\det(Y)$ \leftarrow X, Y $n \times n$ matrices

If $A = SDS^{-1}$, then $\det(A) = \det(S) \det(D) \det(S^{-1})$
 $= \det(S) \det(D) \frac{1}{\det(S)}$

$\det(A) = \det(D) = \lambda_1 \dots \lambda_n$ \leftarrow product of all eigenvalues

② Taking powers of A : compute $A, A^2, A^3, \dots, A^{2006}, \dots, A^{10^{100}}$

Ex. $\begin{pmatrix} 2 & 2 \\ 5 & -1 \end{pmatrix} = A$, what is $A^3 = A \cdot A \cdot A$

$$\begin{aligned} A \cdot A &= \begin{pmatrix} 2 & 2 \\ 5 & -1 \end{pmatrix} \begin{pmatrix} 2 & 2 \\ 5 & -1 \end{pmatrix} \\ &= \begin{pmatrix} 14 & 2 \\ 5 & 11 \end{pmatrix} \begin{pmatrix} 2 & 2 \\ 5 & -1 \end{pmatrix} \\ &= \begin{pmatrix} 32 & 26 \\ 65 & -1 \end{pmatrix} \rightarrow \text{complicated} \end{aligned}$$

If we use $A = SDS^{-1}$, what is A^2 ?

It is $A^2 = A \cdot A = (SDS^{-1})(SDS^{-1}) = SD^2S^{-1}$

Lemma 1: $A^n = SD^nS^{-1}$, note that $D^n = \begin{pmatrix} \lambda_1^n & & \\ & \ddots & \\ & & \lambda_n^n \end{pmatrix}$

Ex. $\begin{pmatrix} 2 & 2 \\ 5 & -1 \end{pmatrix}^{10} = \begin{pmatrix} 2 & 2 \\ -5 & 1 \end{pmatrix} \begin{pmatrix} -3 & 0 \\ 0 & 4 \end{pmatrix}^{10} \cdot \frac{1}{7} \begin{pmatrix} 1 & -1 \\ 5 & 2 \end{pmatrix}$
 $= \begin{pmatrix} 2 & 2 \\ -5 & 1 \end{pmatrix} \begin{pmatrix} (-3)^{10} & 0 \\ 0 & (4)^{10} \end{pmatrix} \begin{pmatrix} 1/7 & -1/7 \\ 5/7 & 2/7 \end{pmatrix}$
 $= \begin{pmatrix} 2(-3)^{10} & (4)^{10} \\ -5(-3)^{10} & (4)^{10} \end{pmatrix} \begin{pmatrix} 1/7 & -1/7 \\ 5/7 & 2/7 \end{pmatrix}$
 $= \begin{pmatrix} \frac{2}{7}(-3)^{10} + \frac{5}{7}(4)^{10} & -\frac{2}{7}(-3)^{10} + \frac{2}{7}(4)^{10} \\ -\frac{5}{7}(-3)^{10} + \frac{5}{7}(4)^{10} & \frac{5}{7}(-3)^{10} + \frac{2}{7}(4)^{10} \end{pmatrix} \rightarrow \text{easy to compute}$

③ Computing $\exp(A) = e^A$: recall, if $x \in \mathbb{R}$, then $e^x = \sum_{n=0}^{\infty} \frac{1}{n!} x^n = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \frac{x^4}{4!} + \dots$

For a matrix A , define $e^A := \text{Id}_{n \times n} + A + \frac{A^2}{2!} + \frac{A^3}{3!} + \frac{A^4}{4!} + \dots +$
matrix x

Lemma 2:

If $A = SDS^{-1}$, then $e^A = S \cdot e^D \cdot S^{-1}$

Proof:

$$\begin{aligned} e^A &= e^{SDS^{-1}} \stackrel{\text{lemma 1}}{=} \text{Id} + SDS^{-1} + \frac{(SDS^{-1})^2}{2!} + \frac{(SDS^{-1})^3}{3!} + \dots \\ &= S(\text{Id} + D + \frac{D^2}{2!} + \frac{D^3}{3!} + \dots)S^{-1} \\ &= S e^D S^{-1} \end{aligned}$$

Application of computing exp of a matrix,

Ex. (22B week 1) $x(t), x: \mathbb{R} \rightarrow \mathbb{R}$, we do not know $x(t)$, but we know something about $x'(t)$.

Eg. $x'(t) = 3x$, then $3t + c = x(t)$

$$x'(t) = 3x(t) \leadsto x(t) = e^{3t} \cdot c$$

In several,

$$\text{variables: } A \cdot \begin{pmatrix} x_1(t) \\ \vdots \\ x_n(t) \end{pmatrix} = \begin{pmatrix} x_1'(t) \\ \vdots \\ x_n'(t) \end{pmatrix} \Leftrightarrow \begin{pmatrix} a_{11}x_1(t) + \dots + a_{1n}x_n(t) = x_1'(t) \\ \vdots \\ a_{n1}x_1(t) + \dots + a_{nn}x_n(t) = x_n'(t) \end{pmatrix}$$

matrix

$$A \cdot x(t) = \vec{x}'(t) \quad \therefore \text{Sol: } \vec{x} = e^{At}$$

④ Physical meaning of λ : in quantum mechanics \rightarrow vector spaces are ∞ -dim

In the basic setting, you have Schrodinger's eqⁿ (analogy of $F=ma$)

$$\partial_t \Psi(x,t) = \hat{H} \Psi(x,t), \text{ where } H = KE + PE \text{ and } \hat{H}(x,t) = \partial x^2 + V(x) \leftarrow \text{linear maps}$$

unknown $C^\infty(\mathbb{R})$

$\frac{1}{2}(x)''$

$x' = p$

eigenvalues

\rightarrow probability for a certain eigenstate