

EIGENVALUES, EIGENVECTORS, DIAGONALIZATION

if A is diagonalizable ...

$$A = P D P^{-1}$$

A is related to D by
a similarity transform.

$$= \begin{bmatrix} V_1 & \cdots & V_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix}$$

right eigenvectors diagonal eigenvalues left eigenvectors

$w_i^T V_j = 1$

$P^T P = I$

$$\rho(A): \text{spectrum of } A \quad \rho(A) := \{\lambda_1, \dots, \lambda_n\}$$

Right eigenvector: $A V_i = \lambda_i V_i$

Left eigenvector: $w_i^T A = w_i^T \lambda_i$

select $w_i \in V_i$ so
 $w_i^T V_i = 1$

Geometrically how do V_i 's & w_i 's relate

$$P^{-1} P = \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix} \begin{bmatrix} V_1 & \cdots & V_n \end{bmatrix} = \begin{bmatrix} w_1^T V_1 & \cdots & w_1^T V_n \\ w_n^T V_1 & \cdots & w_n^T V_n \end{bmatrix}$$

$$w_i^T V_j = 0 \quad j \neq i$$

off diagonal terms are 0

$$= \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & 0 \\ \vdots & & \ddots & 0 \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

a right eigenvector v_i is orthogonal to the subspace spanned by all other left eigenvectors w_j $j \neq i$

a left eigenvector w_i is orthogonal to the subspace spanned by all other right eigenvectors v_j $j \neq i$

$$Av_i = [v_1 \dots v_n] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix} v_i \quad \leftarrow$$

$$\underline{[v_1 \dots v_n] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} 0 \\ w_1^T v_i \\ \vdots \\ w_n^T v_i \end{bmatrix}} = \begin{bmatrix} w_1^T v_i \\ \vdots \\ w_n^T v_i \end{bmatrix}$$

$$\underline{\underline{[v_1 \dots v_n] \begin{bmatrix} 0 \\ \lambda_i \\ 0 \end{bmatrix}}}$$

$$Av_i = \lambda_i v_i = v_1 0 + v_i \lambda_i + v_n 0$$

$$\begin{aligned} w_i^T A &= w_i^T [v_1 \dots v_n] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix} \\ &= \underline{\underline{[w_i^T v_1 \dots w_i^T v_n] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix}}} \\ &\quad \underline{\underline{[\underbrace{0}_{1} w_i^T v_i 0] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix}}} \\ &= \underline{\underline{[0 \ \lambda_i \ 0] \begin{bmatrix} -w_1^T \\ \vdots \\ -w_n^T \end{bmatrix}}} = \lambda_i w_i^T \end{aligned}$$

IN GENERAL: For eigenvector pairs
 not true that $w_i^T v_i \neq 1$
 eigenvectors really should be called
 eigensubspaces \rightarrow we don't know $|w_i|$ or $|v_i|$
 can select $w_i \in v_i$ so that $w_i^T v_i = 1$

$$A = [v_1 \dots v_n] \begin{bmatrix} \lambda_1 & 0 \\ 0 & \ddots & \lambda_n \end{bmatrix} \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix}$$

$$\underline{A} = \sum_i v_i w_i^T \lambda_i = \sum_i \underbrace{\begin{bmatrix} v_i \end{bmatrix}}_{\text{dyadic expansion}} \underline{\lambda_i} \underline{[-w_i^T]}$$

$$Av_i = \sum_j \underbrace{\begin{bmatrix} v_j \end{bmatrix}}_{0 \text{ if } j \neq i} \underline{\lambda_j} \underline{[-w_j^T]} v_i$$

coords
w.r.t the
eigenvector
basis

EIGENVECTOR COORDINATES:

$$y = \underline{Ax}$$

$$x = \underline{Px'} = [v_1 \dots v_n] \begin{bmatrix} x'_1 \\ \vdots \\ x'_n \end{bmatrix}$$

$$y = \underline{Py'}$$

basis of eigen vectors writing x as a linear comb of right eigenvectors

$$Py' = APx'$$

|

$$\textcircled{P} \mathbf{y}' = P D P^{-1} \textcircled{P} \mathbf{x}' = \textcircled{P} D \mathbf{x}' \Rightarrow \mathbf{y}' = D \mathbf{x}'$$

Powers of Matrices:

$$\begin{bmatrix} y_1' \\ y_n' \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_n \end{bmatrix} \begin{bmatrix} x_1' \\ x_n' \end{bmatrix}$$

$$A = P D P^{-1}$$

$$A^k = \overbrace{A \times \dots \times A}^k$$

$$\rightarrow = P D \underbrace{P^{-1} \times P D \underbrace{P^{-1} \times \dots \times P D}_{k-1} P^{-1}}$$

$$\rightarrow \underline{\underline{A^k}} = \underline{\underline{P D^k P^{-1}}} = P \underbrace{\begin{bmatrix} \lambda_1^k & 0 \\ 0 & \lambda_n^k \end{bmatrix}}_{D^k} P^{-1}$$

later

Polynomials of A:

$$f(X) = X^m + \alpha_{m-1} X^{m-1} + \dots + \alpha_0 I$$

$$f(A) = P D^m P^{-1} + \alpha_{m-1} P D^{m-1} P^{-1} + \dots + \alpha_0 P P^{-1}$$

$$\rightarrow = P \underbrace{D^m + \alpha_{m-1} D^{m-1} + \dots + \alpha_0 I}_{P^{-1}}$$

$$= P \begin{bmatrix} \lambda_1^m + \alpha_{m-1} \lambda_1^{m-1} + \dots + \alpha_0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

PREVIEW

Jordan
Form

$$A = P J P^{-1}$$

MOST
GENERAL
VERSION
OF
DIAGONALIZING

$$J^k = \begin{bmatrix} \lambda_1^k & & \\ & \ddots & \\ & & \lambda_n^k \end{bmatrix}$$

$$= \begin{bmatrix} 0 & & \\ & \ddots & \\ & & \lambda_n^m + \alpha_{m-1} \lambda_n^{m-1} + \dots + \alpha_0 \end{bmatrix} P^{-1}$$

Remember: $\chi_A(s) = \det(sI - A) = s^n + \underbrace{\alpha_{n-1}s^{n-1} + \dots + \alpha_0}_{\lambda_i \text{ were roots}}$

$$\underline{\chi_A(\lambda_i) = 0}$$

λ_i 's were roots
of $\chi_A(s)$

$$\begin{aligned}\underline{\chi_A(A)} &= P \underline{\chi_A(D)} P^{-1} = P \begin{bmatrix} \underline{\chi_A(\lambda_1)} & & 0 \\ 0 & \ddots & \\ & & \underline{\chi_A(\lambda_n)} \end{bmatrix} P^{-1} \\ &= P \begin{bmatrix} 0 \end{bmatrix} P^{-1} = 0\end{aligned}$$

$\chi_A(s)$
char
poly:

$$\underline{\chi_A(A) = 0} \quad \text{Cayley-Hamilton Theorem}$$

(for general matrices true but
slightly different proof)

A result of Cayley Hamilton $A \in \mathbb{R}^{n \times n}$
is that any polynomial function of any degree
of a matrix can be written
as a polynomial of degree $\underline{n-1}$.

$$\underline{\chi_A(s) = \det(sI - A) = s^n + \underbrace{\alpha_{n-1}s^{n-1} + \dots + \alpha_0}_{-}}$$

$$\underline{|A|^n} = - \underbrace{\alpha_{n-1}A^{n-1}}_{-} - \dots - \underbrace{\alpha_0 I}_{-}$$

$$\begin{aligned}
 A^{n+2} &= A^n A^2 = \left(-\underbrace{\alpha_{n-1} A^{n-1} - \dots - \alpha_0 I}_{= -\alpha_{n-1} A^{n+1} + \dots} \right) A^2 \\
 &= -\alpha_{n-1} A^{n+1} + \dots \\
 &= -\alpha_{n-1} A^n A + \dots \\
 &= -\alpha_{n-1} \left(-\underbrace{\alpha_{n-1} A^{n-1} - \dots - \alpha_0 I}_{= -\alpha_{n-1} (-\alpha_{n-1} A^{n-1} - \dots - \alpha_0 I)} \right) A \\
 &= -\alpha_{n-1} \left(-\alpha_{n-1} A^n - \dots \right) \\
 &= -\alpha_{n-1} \left(-\alpha_{n-1} \left(-\underbrace{\alpha_{n-1} A^{n-1} - \dots - \alpha_0 I}_{= -\alpha_{n-1} (-\alpha_{n-1} A^{n-1} - \dots - \alpha_0 I)} \right) \right)
 \end{aligned}$$

$$\begin{aligned}
 \chi_A(s) &= s^n + \alpha_{n-1} s^{n-1} + \dots + \alpha_1 s + \alpha_0 \\
 s^n + \alpha_{n-1} s^{n-1} + \dots + \alpha_1 s &= 0 \\
 (s^{n-1} + \alpha_{n-1} s^{n-2} + \dots + \alpha_1) s &= 0 \\
 s &= 0
 \end{aligned}$$

Clever way to write A^{-1} as a polynomial?
using $\chi_A(A) = 0$

$$\begin{aligned}
 \chi_A(A) &= A^n + \alpha_{n-1} A^{n-1} + \dots + \alpha_1 A + \alpha_0 I = 0 \quad A^{-1} \\
 A^{n-1} + \alpha_{n-1} A^{n-2} + \dots + \alpha_1 I + \alpha_0 A^{-1} &= 0
 \end{aligned}$$

$$A^{-1} = \frac{1}{\alpha_0} \left[-A^{n-1} - \alpha_{n-1} A^{n-2} - \dots - \alpha_1 I \right]$$

$$\underline{A^{-1}} : A = P D P^{-1} \Rightarrow$$

$$\bar{A}^{-1} = (\bar{P}^{-1})^{-1} \bar{D}^{-1} \bar{P}^{-1} = \bar{P} \bar{D}^{-1} \bar{P}^{-1} = \bar{P} \begin{bmatrix} \frac{1}{\lambda_1} & & \\ & \ddots & \\ & & \frac{1}{\lambda_n} \end{bmatrix} \bar{P}^{-1}$$

inverting a matrix is the same
as inverting the eigenvalues

square A is invertible iff $\lambda_i \neq 0 \forall i$

Determinant & Trace — Eigenvalues $\rightarrow \det(A) \neq 0$

$$\det(A) = \det(PDP^{-1}) = \det(P)\det(D)\det(P^{-1})$$

$$= \cancel{\det(P)} \det(D) \frac{1}{\cancel{\det(P)}} = \underline{\det(D)}$$

$$\boxed{\det(A) = \prod_{i=1}^n \lambda_i}$$

for diagonal matrix

$\det(D) = \text{product of diagonal}$

$$\text{tr}(A) = \text{tr}(PDP^{-1}) = \text{tr}(P^{-1}PD) = \text{tr}(D)$$

$$\boxed{\text{tr}(A) = \sum_{i=1}^n A_{ii} = \sum_{i=1}^n \lambda_i} \leftarrow$$

Taylor Expansions of Matrices...

Matrix Exponential:

$$\begin{aligned} e^{At} &:= I + At + \frac{1}{2}(At)^2 + \frac{1}{3!}(At)^3 + \dots \\ &= \sum_{k=0}^{\infty} \frac{1}{k!}(At)^k = \sum_{k=0}^{\infty} \frac{1}{k!} A^k t^k \end{aligned}$$

$$\underline{e^{At}} = \underline{Pe^{Dt}P^{-1}} = P \begin{bmatrix} e^{\lambda_1 t} & 0 \\ 0 & e^{\lambda_n t} \end{bmatrix} P^{-1}$$

ORDINARY DIFFERENTIAL EQUATIONS (ODEs)

scalar: $\dot{x}(t) = \frac{dx}{dt} = \underline{\lambda x(t)} \quad x(0) = x_0$

$x(t) \in \mathbb{R}$

$$x(t) = \underline{\text{const}} \times e^{\lambda t} \quad \frac{d}{dt} e^{\lambda t} = e^{\lambda t}$$

$$\dot{x} = \text{const} \frac{d}{dt} e^{\lambda t} = \lambda \text{const} \frac{e^{\lambda t}}{x(t)}$$

$$\text{const} = x_0$$

Solution:

$$x(t) = e^{\lambda t} x_0$$

vector case

$x(t) \in \mathbb{R}^n$

$$\dot{x}(t) = \frac{dx}{dt} = \underline{Ax(t)} \quad x(0) = \underline{x_0}$$

$$x(t) = e^{\lambda t} x_0$$

$$x(t) = Pe^{Dt}P^{-1}x_0$$

Change
coords:

$$x(t) = Pz(t)$$

$$z(t) = e^{Dt} z(0)$$

writing
 x in the eigenvector
coordinates

$$z_i(t) = e^{\lambda_i t} z_{i,0}$$

$$z_n(t) = e^{\lambda_n t} z_{n,0}$$

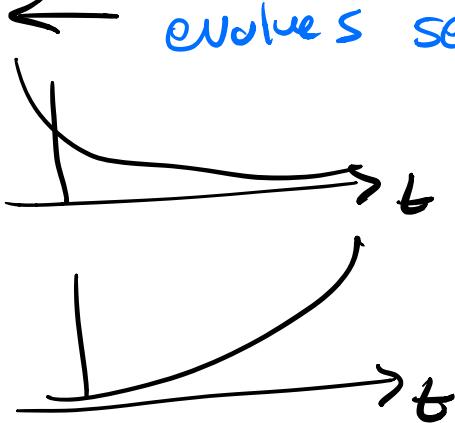
if $\operatorname{Re}(\lambda_i) < 0$

$$e^{\lambda_i t} \rightarrow 0$$

if $\operatorname{Re}(\lambda_i) > 0$

$$e^{\lambda_i t} \rightarrow \infty$$

in ODE's an eigenvector evolves separately



Stability
of linear
systems

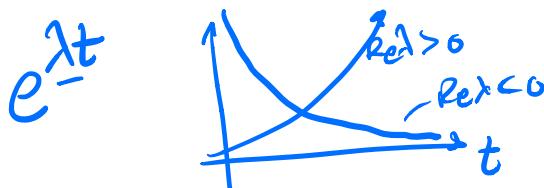
SUMMARY

SPECTRAL
MAPPING :
THM

$$\varphi(A) = \{\lambda_1, \dots, \lambda_n\}$$

$f(\cdot)$: analytic function
(polynomials, etc.)

$$\varphi(f(A)) = \{f(\lambda_1), \dots, f(\lambda_n)\}$$



SCRIBBLES

