

eg. Markov chain

$$p(t+1) = P \cdot p(t)$$

every column sum is 1 ($1^T P = 1^T$)

$$\begin{bmatrix} p_u \\ p_v \end{bmatrix} \quad \begin{bmatrix} p(1) & p(2) \\ p_u(1) & p_v(1) \end{bmatrix}$$

1 is a left eigenvector, with eigenvalue 1
 $\det(P - I) = 0$, $\therefore$ there's also a right eigenvector with eigenvalue 1

have $Pv = v$, v is the equilibrium distribution / steady state

Diagonalization

$A = T \Lambda T^{-1}$ cols of T is independent set of eigenvectors

$$\Lambda = T^{-1} A T$$

A is called defective if not diagonalizable

not all matrices are diagonalizable eg. $A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$

if A has n distinct eigenvalues, then diagonalizable

Diagonalization and left eigenvector

$$A = T \Lambda T^{-1} \quad T^{-1} A = \Lambda T^{-1} \quad \text{let } T^{-1} = \begin{bmatrix} w_1^T \\ \vdots \\ w_n^T \end{bmatrix} \quad T = \begin{bmatrix} | & & | \\ v_1 & \dots & v_n \\ | & & | \end{bmatrix}$$

$$\begin{bmatrix} -w_1^T \\ -w_2^T \\ \vdots \\ -w_n^T \end{bmatrix} A = \begin{bmatrix} -w_1^T A \\ \vdots \\ -w_n^T A \end{bmatrix} = \begin{bmatrix} -\lambda_1 w_1^T \\ \vdots \\ -\lambda_n w_n^T \end{bmatrix}$$

the row of T^{-1} are left eigenvectors of A , with eigenvalue $\lambda_1 \dots \lambda_n$

$$T^{-1} T = I \Rightarrow w_i^T v_j = 1 \{i=j\} \quad \text{left \& right eigenvectors are dual bases}$$

- Modal Form

$$A = T \Lambda T^{-1}$$

define new coordinate $\tilde{x} = T^{-1}x$

$$T^{-1} A T = \Lambda$$

$$T^{-1} A T \tilde{x} = \Lambda \tilde{x}$$

$$\widetilde{Ax} = \Lambda \tilde{x}$$

$$T^{-1}(Ax) = \Lambda \tilde{x}$$

the trajectories consist of n independent modes

$$\tilde{x}_i(t) = e^{\lambda_i t} \tilde{x}_i(0)$$

when eigenvalues are complex, $\lambda_i = \sigma_i + j\omega_i$, system can be put in real modal form

$$S^{-1} A S = \begin{bmatrix} \boxed{\lambda_1} & & \\ & \begin{bmatrix} \sigma_1 & \omega_1 \\ -\omega_1 & \sigma_1 \end{bmatrix} & \\ & & \ddots \\ & & & \begin{bmatrix} \sigma_n & \omega_n \\ -\omega_n & \sigma_n \end{bmatrix} \end{bmatrix}$$

diagonalization simplifies many expressions

$$\begin{aligned} (sI - A)^{-1} &= (sI T^{-1} - T \Lambda T^{-1})^{-1} \\ &= [T (sI - \Lambda) T^{-1}]^{-1} \\ &= T (sI - \Lambda)^{-1} T^{-1} \\ &= T \operatorname{diag}\left(\frac{1}{s - \lambda_i}\right) T^{-1} = \sum_i \frac{v_i w_i^T}{s - \lambda_i} \end{aligned}$$

$$\begin{aligned} e^{At} &= I + A + \frac{A^2}{2!} + \frac{A^3}{3!} \dots \\ &= I + T \Lambda T^{-1} + \frac{T \Lambda^2 T^{-1}}{2!} \dots \\ &= T e^{\Lambda t} T^{-1} \\ &= T \operatorname{diag}(e^{\lambda_i t}) T^{-1} \end{aligned}$$

$$f(A) = T \operatorname{diag}(f(\lambda_i)) T^{-1} \quad \text{when } f \text{ is a polynomial}$$

• Solution via Diagonalization

$$\dot{x} = Ax \quad A = T\Lambda T^{-1}$$

$$x(t) = e^{tA} x(0)$$

$$= T e^{t\Lambda} T^{-1} x(0)$$

$$= T \operatorname{diag}(e^{t\lambda_i}) T^{-1} x_0$$

$$= \sum_i e^{t\lambda_i} \cdot (v_i \cdot w_i^T) x_0$$

$$= \sum_i e^{t\lambda_i} \cdot (w_i^T x_0) v_i$$

$$x(0) = T T^{-1} x(0) = \sum_i v_i w_i^T x(0) = \sum_i (w_i^T x(0)) v_i$$

any trajectory can be expressed as linear combination of modes

decompose $x(0)$ into modes, propagate each mode, reassemble

eg. divide the eigenvalues into

$$R(\lambda_1) \dots R(\lambda_s) < 0 < R(\lambda_{s+1}) \dots R(\lambda_n)$$

$$x(t) = \sum_i e^{\lambda_i t} (w_i^T x(0)) v_i$$

$$\Rightarrow \sum_i e^{\sigma_i t} \cos(\omega_i t + \phi) (w_i^T x(0)) v_i$$

first s terms exponentially decay
last $n-s$ terms exponentially grow

condition for $x(t) \rightarrow 0$: $x(0) \in \operatorname{span}\{v_1, \dots, v_s\}$ stable eigenspace

Stability of Discrete-Time System

$$A = T \Lambda T^{-1} \quad x(t+1) = A \cdot x(t)$$

$$x(t) = A^t x(0) = \sum_{i=1}^n \lambda_i^t (W_i^T x(0)) V_i$$

V_i : part decays if $|\lambda_i| < 1$ (complex magnitude)
 (true even if A not diagonalizable)

$x^+ = A x$ is stable if all λ_i has magnitude ≤ 1

Jordan Canonical Form

"as close as a diagonalization when non diagonalizable"

mostly a conceptual tool, rarely used in practice

Any matrix $A \in \mathbb{R}^{n \times n}$ can be written in Jordan canonical form by a similarity transformation

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

Similarity transformation does not change the determinant $\det AB = \det BA$

$$J = \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_r \end{bmatrix}$$

$$\sum_{i=1}^r n_i = n$$

$$J_i = \begin{bmatrix} \lambda_i & 1 & & \\ & \lambda_i & 1 & \\ & & \ddots & \ddots \\ & & & \lambda_i \end{bmatrix} \in \mathbb{C}^{n_i \times n_i}$$

Jordan block with eigenvalue λ_i

J is upper-bidiagonal

J diagonal is the special case of n Jordan blocks of size $n_i = 1 \forall i$

Jordan form is unique (up to permutation of blocks)

can have multiple blocks with same eigenvalue

the characteristic polynomial $\chi(s) = \det(sI - A) = (s - \lambda_1)^{n_1} \dots (s - \lambda_r)^{n_r}$
 (distinct eigenvalue $\Rightarrow n_i = 1 \Rightarrow J$ is diagonal)

eg. Jordan blocks with same eigenvalue

$$J = \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix}$$

$$J = \begin{bmatrix} 1 & 1 & & \\ & 1 & 1 & \\ & & \ddots & \ddots \\ & & & 1 \end{bmatrix}$$

these matrix has eigenvalues $\{1, 1, 1\}$

$$J = \begin{bmatrix} \boxed{1} & & \\ & \boxed{1} & \\ & & \boxed{1} \end{bmatrix}$$

$\dim \text{Null}(\lambda I - A)$ is the number of Jordan blocks with eigenvalue λ

$$T^{-1}(\lambda I - A)T = \lambda I - J = \begin{bmatrix} \boxed{\lambda_i - 1} & & \\ & \boxed{\lambda_i - 1} & \\ & & \ddots \end{bmatrix} - \lambda I$$

gives 0 column

eg matrix with eigenvalue $\{1, 1, 1, 3, 5\}$ may have Jordan form

$$\begin{bmatrix} 1 & & & & \\ & 1 & & & \\ & & 1 & & \\ & & & 3 & \\ & & & & 5 \end{bmatrix} \quad \begin{bmatrix} \boxed{1} & & & & \\ & \boxed{1} & & & \\ & & \boxed{1} & & \\ & & & \boxed{3} & \\ & & & & \boxed{5} \end{bmatrix}$$

3 and 5 has Jordan Block 1x1
since they both have multiplicity 1

Generalized Eigenvectors

$$T^{-1}AT = J = \text{diag}(J_1, \dots, J_k)$$

$$T = \begin{bmatrix} || & || & \\ \hline T_1 & T_2 & \dots & T_k \\ || & || & \end{bmatrix} \quad T_i = \begin{bmatrix} | & | & | \\ \hline v_{i1} & v_{i2} & v_{in_i} \\ | & | & | \end{bmatrix}$$

$$Av_{i1} = \lambda_i v_{i1}$$

$$Av_{ij} = v_{i,j-1} + \lambda_i v_{ij}$$

$v_{i1} \dots v_{in_i}$ are generalized eigenvectors

Jordan Form LDS

$$\dot{x} = Ax \quad A = TJT^{-1}$$

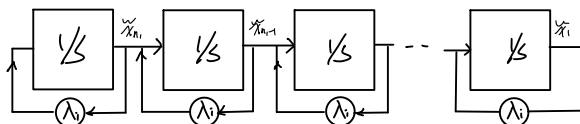
$$\text{let } \tilde{x} = T^{-1}x$$

$$\dot{\tilde{x}} = A\tilde{x} = TJ\tilde{x} \quad \dot{\tilde{x}} = J\tilde{x}$$

$$\begin{bmatrix} \boxed{\lambda_1 - 1} & & \\ & \boxed{\lambda_1 - 1} & \\ & & \ddots \end{bmatrix}$$

$$\tilde{x} = \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \\ \vdots \end{bmatrix}$$

System is decomposed into independent "Jordan Block System" $\dot{\tilde{x}}_i = J_i \tilde{x}_i$



• Resolvent, Exponential of Jordan Block

$$(sI - J_\lambda)^{-1} = \begin{bmatrix} s-\lambda & 1 & & \\ & s-\lambda & 1 & \\ & & \ddots & 1 \\ & & & s-\lambda \end{bmatrix}^{-1} = \begin{bmatrix} (s-\lambda)^{-1} & (s-\lambda)^{-2} & \dots & (s-\lambda)^{-k} \\ & (s-\lambda)^{-1} & \dots & (s-\lambda)^{-k+1} \\ & & \ddots & \\ & & & (s-\lambda)^{-1} \end{bmatrix}$$

$$= (s-\lambda)^{-1} I + (s-\lambda)^{-2} F_1 + \dots + (s-\lambda)^{-k} F_{k-1} \quad F_k = \begin{bmatrix} & 1 & & \\ & & \ddots & \\ & & & 1 \\ & & & & \ddots & \\ & & & & & 1 \end{bmatrix}$$

$$\begin{aligned} e^{tJ_\lambda} &= I + tJ_\lambda + \frac{(tJ_\lambda)^2}{2!} + \dots \\ &= \mathcal{L}^{-1} \left[\frac{I}{s} + \frac{tJ_\lambda}{s^2} + \dots + \frac{(tJ_\lambda)^{k-1}}{s^k} + \dots \right] \\ &= \mathcal{L}^{-1} \left[(sI - tJ_\lambda)^{-1} \right] \\ &= \mathcal{L}^{-1} \left[(s-\lambda)^{-1} \right] I + \mathcal{L}^{-1} \left[(s-\lambda)^{-2} \right] F_1 + \dots \\ &= e^{\lambda t} \left[I + \frac{t}{1!} + \dots + \frac{t^{k-1}}{(k-1)!} F_{k-1} \right] \\ &= e^{\lambda t} \begin{bmatrix} 1 & t & \dots & \frac{t^{k-1}}{(k-1)!} \\ & 1 & t & \dots & \frac{t^{k-2}}{(k-2)!} \\ & & \ddots & \ddots & \vdots \end{bmatrix} \end{aligned}$$

Jordan blocks correspond to $t^k e^{\lambda t}$ terms

eg

$$\dot{x} = Ax$$

$$A = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix} \quad x(t) = e^{At} \begin{bmatrix} 1 & t & \dots & \frac{t^{k-1}}{(k-1)!} \\ & 1 & t & \dots & \frac{t^{k-2}}{(k-2)!} \\ & & \ddots & \ddots & \vdots \end{bmatrix} x(0)$$

will see ($\lambda=0$)

$$\begin{aligned} t \cdot e^{\lambda t} &= t \\ t^2 e^{\lambda t} &= t^2 \\ t^3 e^{\lambda t} &= t^3 \end{aligned}$$

$$A = \begin{bmatrix} \boxed{0} & & \\ & \boxed{0} & \\ & & \boxed{0 \ 1 \\ 0 & 0} \end{bmatrix} \quad x(t) = \begin{bmatrix} 1 & & \\ & 1 & \\ & & \boxed{1 \ t \\ & 1} \end{bmatrix} \quad \text{will see } t \cdot e^{\lambda t}$$

