

$$x(t) = e^{At} x_0 \quad x(0) = x_0$$

$$e^{At} = \sum_{i=1}^n e^{\lambda_i t} x_i y_i^H$$

we assumed A has n L.I. eigenvectors x_i in order to get this.

$$x(t) = \sum_{i=1}^n e^{\lambda_i t} x_i (y_i^H x_0) \quad \text{scalar}$$

$$= e^{\lambda_1 t} x_1 y_1^H x_0 + \dots + e^{\lambda_k t} x_k y_k^H x_0 + \dots + e^{\lambda_n t} x_n y_n^H x_0$$

Claim:

$$\text{If } x_0 = x_k, \text{ then } x(t) = e^{\lambda_k t} x_k$$

"proof"

$$x(t) = e^{\lambda_1 t} x_1 y_1^H x_k + \dots + e^{\lambda_k t} x_k y_k^H x_k + \dots + e^{\lambda_n t} x_n y_n^H x_k$$

Recall:

$$X X^{-1} = I$$

$$\begin{bmatrix} \text{---} & y_1^H & \text{---} \\ \text{---} & y_k^H & \text{---} \\ \text{---} & y_n^H & \text{---} \end{bmatrix} \begin{bmatrix} | & & | \\ x_1 & \dots & x_n \\ | & & | \end{bmatrix} = \begin{bmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{bmatrix}$$

$$y_i^H x_j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$

$$y_1^H x_1 = 1$$

$$y_1^H x_2 = 0$$

$$x(t) = e^{\lambda_k t} x_k y_k^H x_k = e^{\lambda_k t} x_k$$

"end of Proof"

⇒ Example

$$\dot{x} = Ax \quad x(0) = x_0$$

$$A = \begin{bmatrix} 0 & 1 \\ -2 & -3 \end{bmatrix}$$

Let's study the natural modes of this system

a) What are the eigenvalues?

$$\begin{aligned} \det(\lambda I - A) &= 0 \\ &= \det \begin{bmatrix} \lambda & -1 \\ 2 & \lambda+3 \end{bmatrix} = \lambda(\lambda+3) - (-2) \\ &= \lambda^2 + 3\lambda + 2 \\ &= (\lambda+1)(\lambda+2) = 0 \end{aligned}$$

$$\boxed{\lambda_1 = -1}$$

$$\boxed{\lambda_2 = -2}$$

* negative eigenvalues (poles) correspond to decaying exponentials in the time domain.
 e^{-t} and e^{-2t}

These are stable modes

b) Compute Right Eigenvectors

→ will be linearly independent because two distinct eigenvalues.

$$(\lambda I - A)x = 0$$

$$\begin{bmatrix} \lambda_i & -1 \\ 2 & \lambda_i+3 \end{bmatrix} x = 0 \quad \rightsquigarrow \quad x = \begin{bmatrix} 1 \\ \lambda_i \end{bmatrix}$$

$$x_1 = \begin{bmatrix} 1 \\ \lambda_1 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$x_2 = \begin{bmatrix} 1 \\ \lambda_2 \end{bmatrix} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

Right Eigenvectors are NOT unique!

c) compute Left Eigenvectors

$$x^{-1} = \begin{bmatrix} 1 & 1 \\ x_1 & x_2 \\ 1 & 1 \end{bmatrix}^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix}^{-1}$$

$$x^{-1} = \begin{bmatrix} z & 1 \\ -1 & -1 \end{bmatrix} = \begin{bmatrix} -y_1^n & - \\ -y_2^n & - \end{bmatrix}$$

Recall

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix}^{-1} = \frac{1}{ad-bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

d) Solve for $x(t)$

$$x(t) = e^{At} x_0 = \sum_{i=1}^2 e^{\lambda_i t} x_i y_i^n x_0$$

$$= e^{-t} \begin{bmatrix} 1 \\ 1 \end{bmatrix} [z \ 1] x_0 + e^{-2t} \begin{bmatrix} 1 \\ -2 \end{bmatrix} [-1 \ -1] x_0$$

Note:

We could have found $x(t)$ by computing the inverse Laplace Transform of $(sI - A)^{-1} x_0 = X(s)$.

Analysis

If $x_0 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ then $x(t) = e^{-t} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

If $x_0 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$ then $x(t) = e^{-2t} \begin{bmatrix} 1 \\ -2 \end{bmatrix}$

Example