

#### **ECE 602: LUMPED LINEAR SYSTEMS**

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Solution of Discrete-Time LTI Systems

# **Discrete-Time Autonomous LTI Systems**

Discrete-time LTI system

$$x[k+1] = Ax[k], k = 0, 1, ...$$

with initial condition x[0] has the solution

$$x[k] = A^k x[0] := \Phi[k] x[0], \quad k = 0, 1, \dots$$

 $\Phi[k] = A^k$  is called the fundamental matrix of the DT LTI system

- For any  $k_0$ ,  $x[k_0 + k] = \Phi[k]x[k_0]$
- $\Phi[k]$  propagates the solution from any initial time to k steps later
- $\Phi[k+\ell] = \Phi[k] \cdot \Phi[\ell] = \Phi[\ell] \cdot \Phi[k], \quad k,\ell = 0,1,\ldots$
- $\Phi[k]$  may be singular (different from  $\Phi(t)$  for CT LTI systems)

# System Modes: Diagonalizable A Case

Suppose  $A \in \mathbb{R}^{n \times n}$  is diagonalizable:  $A = T \Lambda T^{-1}$ 

- Diagonal entries of  $\Lambda = \operatorname{diag}(\lambda_1, \dots, \lambda_n)$  are the eigenvalues of A
- Column of  $T = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}$  are (right) eigenvectors of A
- Rows of  $T^{-1} = \begin{bmatrix} w_1 & \cdots & w_n \end{bmatrix}^T$  are left eigenvectors of A

The solution to x[k+1] = Ax[k] with initial state x[0] is

$$x[k] = T\Lambda^k T^{-1}x[0] = \left(w_1^T x[0]\right) \lambda_1^k v_1 + \dots + \left(w_n^T x[0]\right) \lambda_n^k v_n$$

- $\lambda_1^k v_1, \dots, \lambda_n^k v_n$  are the modes of the system
- Any solution  $x[\cdot]$  is a linear combination of these n modes

# System Modes: General A Case

Using the JCF: 
$$A = TJT^{-1} = \begin{bmatrix} T_1 & \cdots & T_r \end{bmatrix} \begin{bmatrix} J_1 & & \\ & \ddots & \\ & & J_r \end{bmatrix} \begin{bmatrix} S_1^T \\ \vdots \\ S_r^T \end{bmatrix}$$

Solution to x[k+1] = Ax[k] with the initial state x[0]:

$$x(t) = TJ^{k}T^{-1}x[0] = \sum_{i=1}^{r} T_{i}J_{i}^{k} (S_{i}^{T}x[0])$$

• Columns of  $T_i J_i^k \in \mathbb{R}^{n \times n_i}$  are modes corresponding to eigenvalue  $\lambda_i$ , whose weights in x[k] are given by entries of vector  $S_i^T x[0] \in \mathbb{R}^{n_i}$ 

$$J_{i} = \begin{bmatrix} \lambda_{i} & 1 & & \\ & \ddots & \ddots & \\ & & \ddots & 1 \\ & & & \lambda_{i} \end{bmatrix} \Rightarrow J_{i}^{k} = \begin{bmatrix} \lambda_{i}^{k} & k\lambda_{i}^{k-1} & \cdots & \frac{k(k-1)\cdots(k-n_{i}+2)}{(n_{i}-1)!}\lambda_{i}^{k-n_{i}+1} \\ & \lambda_{i}^{k} & \ddots & \vdots \\ & & \ddots & k\lambda_{i}^{k-1} \\ & & & \lambda_{i}^{k} \end{bmatrix}$$

### **Example**

$$x[k+1] = Ax[k] \text{ with } A = \underbrace{\begin{bmatrix} -1 & 1 & 1 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{T} \underbrace{\begin{bmatrix} -2 & 1 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -1 \end{bmatrix}}_{J} \underbrace{\begin{bmatrix} 0 & -1 & 0 \\ 1 & -1 & -1 \\ 0 & 0 & 1 \end{bmatrix}}_{T^{-1}}$$