

# SC223 - Linear Algebra

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Lecture 33



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## Summary of Lecture 32

- To compute eigenvalues and eigenvectors of  $T$ , find roots of the characteristic polynomial:  $c(x) = \det([T]_{\beta}^{\beta} - xI_n)$ . For any root  $\lambda$ ,  $u \neq \theta$ ,  $u \in N(T - \lambda I)$  is an eigenvector.
- For  $T \in \mathcal{L}(U)$ ,  $\dim(U) = n$ , if a diagonal matrix representation is preferred, we need an *eigenbasis* of the linear operator:  $e = \{u_1, \dots, u_n\}$ .
- The process of similarity transformation on a matrix  $A$ , using eigenvectors as columns of a matrix, say  $E$ , to get a diagonal matrix  $\Lambda$ :  $\Lambda = E^{-1}AE$  is called **matrix diagonalization**.
- **Algebraic multiplicity of  $\lambda$** : Multiplicity of  $\lambda$  as a root of  $c(x)$ .  
Denoted as  $AM(\lambda)$
- **Geometric Multiplicity of  $\lambda$** :  $GM(\lambda) = \dim(N(T - \lambda I))$ .

## Example

- Let  $S_N = \{x : \mathbb{Z} \rightarrow \mathbb{C} \mid x[n+N] = x[n], \forall n \in \mathbb{Z}, x[n] \in \mathbb{C}, \forall n \in \mathbb{Z}\}$ .  
 $\dim(S_N) = N$ .
- Let  $\beta = \{\delta_0, \dots, \delta_{N-1}\}$  denote the basis of  $S^N$ , where  
 $\forall k = 0, \dots, N-1$ .
- Let  $x \in S_N, [x]_\beta \in \mathbb{C}^N, [x]_\beta = (x[0], \dots, x[N-1])$ .
- ~~Delay/Shift operator:  $D \in \mathcal{L}(S_N)$  be defined as~~  
 $D(x)[n] = x[n-1], \forall n \in \mathbb{Z}$ .
- Shift-invariant operator:  $T \in \mathcal{L}(S_N)$  be a linear operator such that  
 $T \cdot D = D \cdot T$ .
- 

$$[Dx]_\beta[n]$$

$$= x[n-1 \bmod N]$$

$$[T]_\beta^\beta = \begin{bmatrix} h[0] & h[N-1] & \dots & h[1] \\ h[1] & h[0] & \dots & h[2] \\ \vdots & \vdots & \ddots & \vdots \\ h[N-1] & h[N-2] & \dots & h[0] \end{bmatrix}$$

- Such a matrix is called a *Circulant matrix*.

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$$f_p[k] = f_p[k+N], \quad \forall k \in \mathbb{Z}.$$

$$f_p[k] = \sum_{n=0}^{N-1} w^{np} \delta_n[k]$$

$$f_p[k+N] = \sum_{n=0}^{N-1} w^{np} \delta_n[k+N] = \sum_{n=0}^{N-1} w^{np} \delta_n[k] = f_p.$$

$p=0, \dots, N-1.$

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$$y[n] = \sum_{k=0}^{N-1} f_p[(n-k) \bmod N] h[k]$$

$$y[n] = \sum_{k=0}^{N-1} x[(n-k) \bmod N] h[k]$$

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$\downarrow$   
 $f, p$  

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- Let  $H(p) := \left( \sum_{k=0}^{N-1} h[k] w^{-kp} \right)$ ,  $p = 0, \dots, N-1$ .
- We have shown that  $Tf_p = H(p)f_p$ ,  $p = 0, \dots, N-1$ , and so  $[T]_{\beta}^{\beta}[f_p]_{\beta} = H(p)[f_p]_{\beta}$ ,  $p = 0, \dots, N-1$ .

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$$[T]_{\beta}^{\beta} \begin{bmatrix} \begin{array}{c} | \\ [f_0]_{\beta} \\ | \end{array} & \begin{array}{c} | \\ [f_1]_{\beta} \\ | \end{array} & \dots & \begin{array}{c} | \\ [f_{N-1}]_{\beta} \\ | \end{array} \end{bmatrix}$$



- Assuming (for now)  $F = \{f_0, \dots, f_{N-1}\}$  are linearly independent,  $F$  forms an eigenbasis for any shift-invariant linear operator on  $S_N$ , and thus

free domain.

$$[T]_{\beta}^{\beta} \begin{bmatrix} | & | & \dots & | \\ [f_0]_{\beta} & [f_1]_{\beta} & \dots & [f_{N-1}]_{\beta} \\ | & | & \dots & | \end{bmatrix} = \begin{bmatrix} | & | & \dots & | \\ [f_0]_{\beta} & [f_1]_{\beta} & \dots & [f_{N-1}]_{\beta} \\ | & | & \dots & | \end{bmatrix} \begin{bmatrix} H(0) & 0 & \dots & 0 \\ 0 & H(1) & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & H(N-1) \end{bmatrix}$$

$$[T]_{\beta}^{\beta} [x]_{\beta} = [y]_{\beta}$$

$$\underbrace{[N]_{\beta}^{\beta} [T]_{\beta}^F [N]_{\beta}^F}_{[T]_{\beta}^{\beta}} [x]_{\beta} = [y]_{\beta}$$

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- $H(p) = \sum_{n=0}^{N-1} h[n]w^{-np}$ ,  $p = 0, \dots, N-1$  is called the *Discrete Fourier Transform (DFT)* of the sequence  $h$ .

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- Also, in the basis  $F$ ,

$$[T]_F^F = \begin{bmatrix} H(0) & 0 & \dots & 0 \\ 0 & H(1) & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & H(N-1) \end{bmatrix}$$

## Results on Eigenvectors and Eigenvalues

- **Proposition 20:** Let  $\lambda_1, \dots, \lambda_m$  be distinct eigenvalues of  $T \in \mathcal{L}(U)$ . Then the eigenvectors  $v_1, \dots, v_m$  associated with these eigenvalues are linearly independent.

Assume  $\{v_1, \dots, v_m\}$  are LI.

Smallest  $k \leq m$  such that  $\{v_1, \dots, v_{k-1}\}$  is LI  
but  $\{v_1, \dots, v_k\}$  is LD.

$$v_k = c_1 v_1 + c_2 v_2 + \dots + c_{k-1} v_{k-1} \rightarrow \textcircled{1}$$

$$\lambda_k v_k = c_1 \lambda_1 v_1 + c_2 \lambda_2 v_2 + \dots + c_{k-1} \lambda_{k-1} v_{k-1} \rightarrow \textcircled{2}$$

$$0 = \underline{c_1(\lambda_1 - \lambda_k)v_1 + c_2(\lambda_2 - \lambda_k)v_2 + \dots + c_{k-1}(\lambda_{k-1} - \lambda_k)v_{k-1}}$$

—END OF CLASS—

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- **Proposition 21:** For  $T \in \mathcal{L}(U)$ ,  $\dim(U) = n$ , if  $\sum_{i=1}^m AM(\lambda_i) = n$ , and  $GM(\lambda_i) = AM(\lambda_i)$ ,  $i = 1, \dots, m$ , then  $T$  is diagonalizable.

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- **Proposition 22:** For  $T \in \mathcal{L}(U)$ ,  $\dim(U) = n$ , for any eigenvalue  $\lambda$  of  $T$ ,  $GM(\lambda) \leq AM(\lambda)$

# Applications

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$$X(t) = \exp(At) X(0)$$

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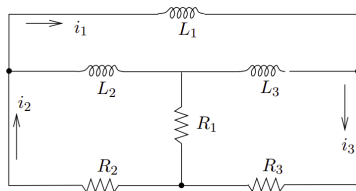


Figure: Source: Differential Equations and Linear Algebra, by GB Gustafson.

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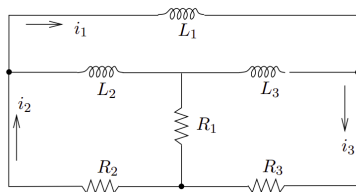


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$$\begin{aligned}i_1' &= - \left( \frac{R_2}{L_1} \right) i_2 - \left( \frac{R_3}{L_1} \right) i_3, \\i_2' &= - \left( \frac{R_2}{L_2} + \frac{R_2}{L_1} \right) i_2 + \left( \frac{R_1}{L_2} - \frac{R_3}{L_1} \right) i_3, \\i_3' &= \left( \frac{R_1}{L_3} - \frac{R_2}{L_1} \right) i_2 - \left( \frac{R_1}{L_3} + \frac{R_3}{L_1} + \frac{R_3}{L_3} \right) i_3\end{aligned}$$



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- Coupled Differential Equations:
- Epidemics Modeling:
  - ▶ Population can be divided into: Susceptible/Healthy (S), Infected (I), and Dead (D) classes.
  - ▶ Susceptible  $\xrightarrow{-a}$  Infected  $\xrightarrow{b}$  Susceptible:  $\frac{d}{dt}S = -aS(t) + rI(t)$
  - ▶ Infected  $\xrightarrow{-d}$  Dead:  $\frac{d}{dt}D = dI(t)$
  - ▶ Infected:  $\frac{d}{dt}I = aS(t) - dI(t) - rI(t)$

# Applications

- Page Rank/Importance:

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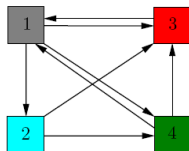
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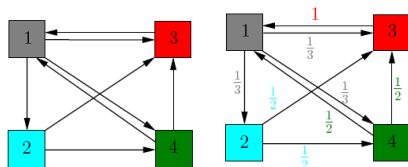
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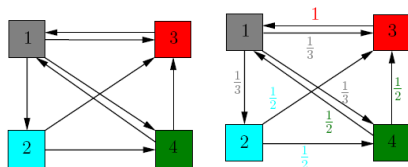
**Figure:** Model: If  $P_2$  has links to  $P_3$  and  $P_4$ , a surfer go to these pages with equal probability. Source: [pi.math.cornell.edu](http://pi.math.cornell.edu)

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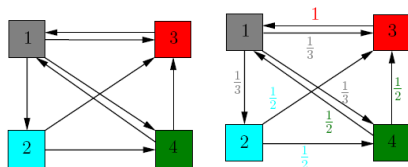
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- $\lim_{n \rightarrow \infty} A^n x_0 = y, Ay = y, y = [0.38, 0.12, 0.29, 0.19]^T$ .