

### Section 4.6: Bases and Dimensions

- Finding a basis for  $\text{Ker}(A)$  \*Note,  $\text{Ker}(A)$  is always a V.S.

- ① Take r.r.e.f and solve systems
- ② write all equations as combinations of free variables
- ③ These combos are the  $\vec{v}$  that span the basis

OR "eye the Kernel"

- Column Space ( $\text{ColSpace}(A)$ )

- a span of  $(v_1, \dots, v_k)$  where  $A = [v_1 \dots v_n]$

- ① Compute r.r.e.f. (need to find pivot columns)
- ② Go back to original matrix, columns corresponding w/ pivot columns span the  $\text{ColSpace}(A)$

\* Row Space

- conserved w/ row reduction
- take r.r.e.f.
- linearly independent rows span row-space  $(A)$

### Section 4.8: Rank Nullity Theorem

$$\dim(\text{ColSpace}(A)) + \dim(\text{Ker}(A)) = n \quad \leftarrow \text{total \# of columns}$$

$$\text{Rank} + \text{Nullity} = n$$

$$\text{Pivot Columns} + \text{Free Variables} = n$$

### Section 4.7: Change of Basis

$$\left. \begin{aligned} B_1 &= [v_1 \dots v_n] \\ B_2 &= [w_1 \dots w_n] \end{aligned} \right\} \text{Basis \& V.S. for } V$$

How can we relate  $[u]_{B_1}$  w/  $[u]_{B_2}$

$$[u]_{B_1} = P_{B_1 \leftarrow B_2} [u]_{B_2}$$

Change of basis matrix from  $B_2 \rightarrow B_1$

$$P_{B_1 \leftarrow B_2} = \begin{bmatrix} | & & | \\ [w_1]_{B_1} & \dots & [w_n]_{B_1} \\ | & & | \end{bmatrix}$$

Basis vectors of  $B_2$  (starting basis) in terms of  $B_1$  (ending basis)

Fact:  $P_{B_2 \leftarrow B_1} = \left[ P_{B_1 \leftarrow B_2} \right]^{-1}$

inverse to go in reverse order

### Section 4.9: Invertible Matrix Theorem

- Key idea: there are a bunch of ways to convey the same idea, that a matrix is invertible

- A has n pivots
- A has 0 free variables
- A has rank n
- $Ax=0$  has only the trivial solution
- dimension of the  $\text{Ker}(A)=0$
- $\det(A) \neq 0$
- $n \neq 1 \implies A$

∴ A is invertible

dimension of the ker(A) = 0  
 det(A) ≠ 0  
 0 ≠ λ of A  
 Range / Colspace spans ℝ<sup>n</sup>

•• T is invertible

## Section 6.1: Linear Transformations

Given  $V, W$  (Vector Spaces) a function

$T: V \rightarrow W$  is called a linear transformation if

$v_1, v_2 \in V$   $T(v_1 + v_2) = T(v_1) + T(v_2)$  (closed under addition)

$c \in \mathbb{R}$   
 $v_1 \in V$   $T(cv_1) = cT(v_1)$  (closed under scalar multiplication)

Linear transformations can be written as matrix multiplication

$T(v) = A \cdot v$  where  $A = \begin{bmatrix} | & | & & | \\ T(e_1) & T(e_2) & \dots & T(e_n) \\ | & | & & | \end{bmatrix}$  ← linearly transform all of the basis vectors

2D Rotation Matrix =  $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$

Section 6.5

## Section 7.1: Eigenvalue / Eigenvector

Problem: Given linear transformation  $T: V \rightarrow V$  what

is the best basis of  $V$  such that  $[T]_{\mathcal{B}}$

← what is easiest matrix to work with? diagonal matrix?

Eigen Values (λ) / Eigenvectors

Given  $T: V \rightarrow V$  a non zero vector  $\vec{v}$

is an eigen vector if  $T\vec{v} = \lambda\vec{v}$

↓  
where  $\lambda$  is the eigen value corresponding to  $\vec{v}$

If  $V$  has a basis of eigen vectors, it

is called an eigen basis

① How to find e-values

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

solve for all possible  $\lambda$ 's by solving the characteristic polynomial

② Find eigen vectors

solve  $\text{Ker}(A - \lambda I)$  for  $\lambda$ 's

the vectors in the ker are the eigen vectors

Section 7.2

\* Not all transformations have an eigen base

$[T]_{\mathcal{C}} = P_{\mathcal{C} \leftarrow \mathcal{B}} [T]_{\mathcal{B}} P_{\mathcal{B} \leftarrow \mathcal{C}}$

similar matrices  $\Rightarrow A$  is similar to  $\mathcal{D}$  if you can write  $B = S^{-1}AS$

## Section 7.3: Diagonalization

What basis should we choose for  $\mathbb{C}^n$  to be as easy to work with as possible?

$\mathbb{C}^n$  is diagonal iff  $A$  is a basis of eigen vectors

A linear map  $T: \mathbb{C}^n \rightarrow \mathbb{C}^n$  is diagonalizable if there is a basis for  $\mathbb{C}^n$  consisting of eigen vectors of  $T$

$$D = S^{-1} \cdot A \cdot S \quad \text{or} \quad A = S \cdot D \cdot S^{-1}$$

↑  
Diagonal matrix of eigen values  
 $\begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$

↑  
original Transformation Matrix

↑  
Change of basis from  $\mathbb{C}^n$  to eigen basis matrices (opposite orders)

↓  
if  $\mathcal{C}$  is standard basis

$$S = \begin{bmatrix} | & | & & | \\ v_1 & v_2 & \dots & v_n \\ | & | & & | \end{bmatrix} \quad \text{where } v_i \text{ are eigen vectors}$$

$\therefore D$  and  $A$  are similar

Facts:  $A$  is  $n \times n$  with e-values  $\lambda_1, \dots, \lambda_n$

$$\text{Tr}(A) = \sum_{i=1}^n \lambda_i$$

$$\det(A) = \prod_{i=1}^n \lambda_i = \lambda_1 \lambda_2 \dots \lambda_n$$

The algebraic multiplicity of an eigen value is always  $\geq$  the geometric multiplicity  
algebraic multiplicity is power of  $(\lambda - \lambda_i)^*$

geometric multiplicity is  $\dim(\ker(A - \lambda_i I))$

## Section 7.4: Matrix Exponentials

$$e^{At} = S \cdot e^{Dt} \cdot S^{-1} \quad \text{for } A = S \cdot D \cdot S^{-1}$$

$$e^{At} = S \cdot \begin{bmatrix} e^{\lambda_1 t} & & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} S^{-1}$$

For diagonalizable matrices

For JC form,  $e^{Jt} \rightarrow$  split into Jordan blocks

$$\text{Exponential Jordan Block size } k = \begin{bmatrix} e^{\lambda t} & t e^{\lambda t} & \frac{t^2}{2} e^{\lambda t} & \dots & \frac{t^{k-1}}{(k-1)!} e^{\lambda t} \\ & \ddots & \ddots & \ddots & \vdots \\ & & & & e^{\lambda t} \end{bmatrix}$$

## Section 7.6 Jordan Canonical Form

- what if matrix isn't diagonalizable? what's the next simplest form?

Jordan Block

for our class, always an eigen value

## Jordan Block

- Block of size  $K$  corresponding to a  $\lambda$  is the  $K \times K$  matrix

$$\begin{pmatrix} \lambda & & 0 \\ & \ddots & \\ 0 & & \lambda \end{pmatrix}$$

$\lambda$ 's on diagonal

1's on super diagonal (above diagonal)

0's everywhere else

\*  $1 \times 1$  Jordan blocks exist and look like  $[\lambda]$

for our class, always an eigen value

## Jordan Canonical Form

- a matrix is in Jordan canonical form if  $A$  is a block diagonal matrix consisting of only Jordan blocks.

For EVERY linear map  $T$ , there exists a basis

$B$  such that  $[T]_B$  is in Jordan Form

\* The Jordan form is unique up to permutations of the blocks

## Generalized eigen vectors

- non-zero vector  $\vec{v} \in \mathbb{C}^n$  such that  $(A - \lambda I)^K \vec{v} = 0$

- normal eigen vectors are generalized eigen vectors where  $K=1$

## Cycle length

- cycle length  $K$  for generalized e-vectors generated by  $\vec{v}$  is

$$\left\{ (A - \lambda I)^{K-1} \vec{v}, \dots, (A - \lambda I) \vec{v}, \vec{v} \right\}$$

$$\left. \begin{matrix} (A - \lambda I)^{K-1} \vec{v} \neq 0 \\ (A - \lambda I)^K \vec{v} = 0 \end{matrix} \right\} \text{Conditions for } \vec{v}$$

## Basis $B$

- The basis  $B$  is a union of cycles of generalized eigen vectors

- every cycle of length  $K$  corresponds to a Jordan block of size  $K$

# of Jordan blocks = # of linearly independent eigen vectors

## Section 9.1: Linear Differential Equations

(\*) Equation, (IC) set of initial conditions

$$(*) = y^{(n)}(t) + a_1(t)y^{(n-1)}(t) + a_2(t)y^{(n-2)}(t) + \dots + a_{n-1}(t)y'(t) + a_n(t)y(t) = F(t)$$

$$(IC) = \left\{ y^{(n-1)}(t_0) = c_1, y^{(n-2)}(t_0) = c_2, \dots, y(t_0) = c_n \right\}$$

$c_1, \dots, c_n \in \mathbb{R}^n$ ,  $a_1(t), \dots, a_n(t)$  and  $F(t)$  are all continuous

$n = \text{order of } (*)$ , if  $F=0$  ODE is "homogeneous" otherwise inhomogeneous

## Concept behind solving Linear Transformation

1. Rephrase  $(*)$  as  $Ly = F$
2. Solve  $Ly = 0$  (find  $\text{Ker}(L)$ )
  - find basis  $\{y_1, \dots, y_n\}$  for  $\text{Ker}(L)$  so all solutions of  $Ly = 0$  can be written in terms of the basis  $(c_1 y_1 + \dots + c_n y_n)$
3. Show general solution of  $Ly = F$  is  $c_1 y_1 + \dots + c_n y_n + y_p(x)$  any particular solution to  $Ly = F$
4. Match initial conditions by solving a system (solve for  $c_1, \dots, c_n$ )

$-0: C^1([a,b]) \rightarrow C^0([a,b])$

$\bullet \bullet Df = f'$

$-0^k: C^k([a,b]) \rightarrow C^0([a,b])$

Note:  $C^N([a,b])$  is V.S. of all functions  $f: [a,b] \rightarrow \mathbb{R}$  with  $n$  continuous derivatives

- Linear Differential Operator

- a linear map  $L: C^k([a,b]) \rightarrow C^0([a,b])$

$L = D^k + a_{k-1}(x)D^{k-1} + \dots + a_0$

- Linear differential equation is an equation that can be written in the form  $Ly = F$  for some differentiable operator,  $L$

## - Existence/Uniqueness

For an initial value problem, any linear ODE where the coefficients are continuous functions, there exists a unique solution satisfying initial conditions

## Section 8.2: Constant Coefficient, Homogeneous L. ODE's

- A linear ODE is a constant coefficient if all of its are constants instead of functions

- Fact, linear differential operators in general don't commute but polynomial differential operators do

$P(D)y = 0$  want to find basis of  $\text{Ker}(P(D))$

$(D-r)y = 0$

① basis =  $e^{rx}$  Gen solution =  $C e^{rx}$

$(D-r)^m y = 0$

②  $C_1 e^{rx} + C_2 x e^{rx} + \dots + C_m x^{m-1} e^{rx}$

at least one roots

③  $e^{ax} \cos bx$  and  $e^{ax} \sin bx$  are solutions

if ODE is order  $n$ , need  $n$  linearly independent vectors for the basis

For order two  $b^2 - 4ac$

- $\rightarrow > 0$
- $\rightarrow = 0$
- $\rightarrow < 0$

when you have a multiplicity of root