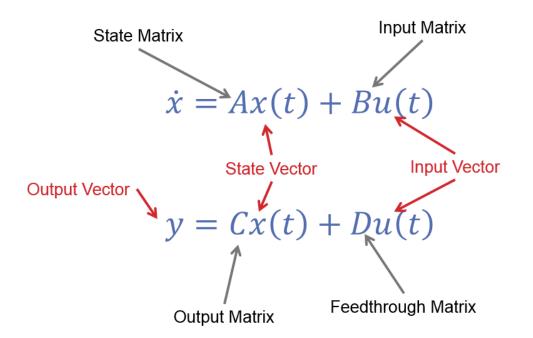
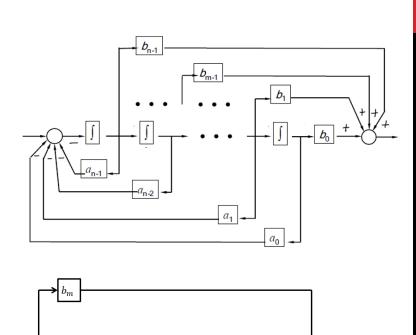
#### FROM LAST TIME...

#### **Intro to State-Space Models**

- Review of State Space
- Transfer Functions and State Space
- Canonical Forms





#### **LINEAR ALGEBRA PRIMER**

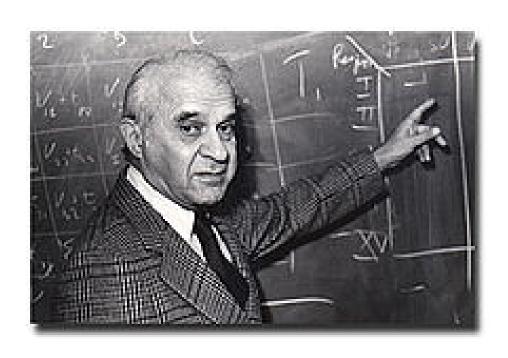
#### **Topics**

- Matrix Inverses
- Eigenvalues and Eigenvectors
- Jordan Canonical Form

#### At the end of this sections, students should be able to:

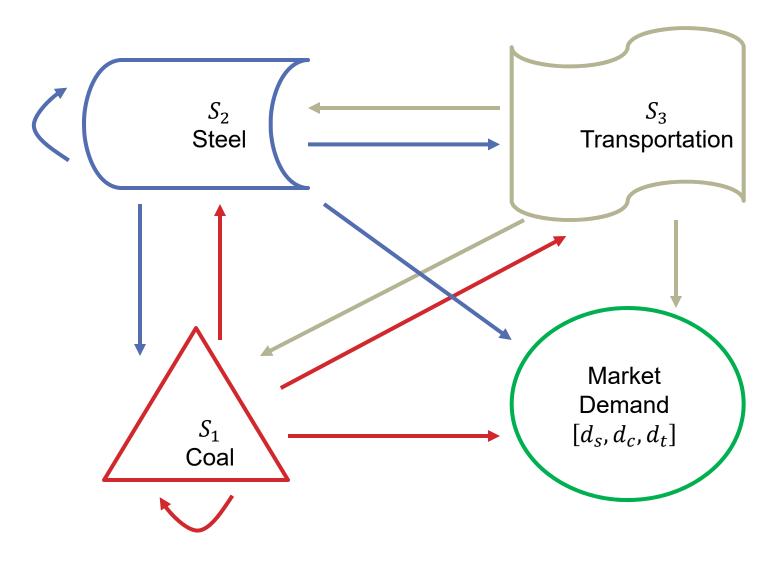
- Compute matrix inverses (up to 3x3).
- Describe eigenvalues and eigenvectors.
- Transform matrices to Jordan Canonical Form.

# WASSILY LEONTIEF WON THE NOBEL MEMORIAL PRIZE IN ECONOMIC SCIENCES IN 1973



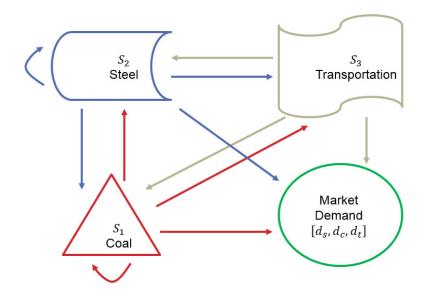
- Input-output tables analyze
   the process by which inputs
   from one industry produce
   outputs for consumption or
   for inputs for another
   industry
- With the input-output table, one can estimate the change in demand for inputs resulting from a change in production of the final good

# CONSIDER A SIMPLIFIED ECONOMY WITH FOUR SECTORS



# WE CAN WRITE THESE INPUT-OUTPUT RELATIONSHIPS AS LINEAR EQUATIONS...

$$c = 0.02c + 2s + 0.01t + D_c$$
$$s = 0.1c + 0.01s + 0.1t + D_s$$
$$t = 0.3c + 0.5s + 0t + D_t$$



- c is the amount of coal (in tons)
- s is the amount of steel (in tons)
- t is the amount of transportation

## ...WHICH CAN BE WRITTEN AS A MATRIX EQUATION!

$$\begin{bmatrix} c \\ s \\ t \end{bmatrix} = \begin{bmatrix} 0.02 & 2 & 0.01 \\ 0.1 & 0.01 & 0.1 \\ 0.3 & 0.5 & 0 \end{bmatrix} \begin{bmatrix} c \\ s \\ t \end{bmatrix} + \begin{bmatrix} D_c \\ D_s \\ D_t \end{bmatrix}$$

 We can solve for the required amount of each good using the matrix inverse

$$\begin{bmatrix} c \\ s \\ t \end{bmatrix} = \begin{pmatrix} I_3 - \begin{bmatrix} 0.02 & 2 & 0.01 \\ 0.1 & 0.01 & 0.1 \\ 0.3 & 0.5 & 0 \end{bmatrix} \end{pmatrix}^{-1} \begin{bmatrix} D_c \\ D_s \\ D_t \end{bmatrix}$$

 : Matrix inverses are worth at least US\$1.2 million (current Nobel Prize monetary value)

James A. Mynderse

### WHY DO WE CARE ABOUT LINEAR ALGEBRA FOR CONTROLLER DESIGN?

#### **System properties**

- Poles (eigenvalues)
- Controllability/Observability (rank)

#### **System response**

- SS to TF (matrix inverse)
- Free response (matrix exponential)

#### Controller design

- Observer/Controller design (eigenvalues, inverse, matrix multiplication)
- Optimal control (calculus of variations, transpose, inverse)

#### **LINEAR ALGEBRA PRIMER**

# GENERAL MATRIX STUFF

### THE DIMENSIONS OF A MATRIX ARE GIVEN IN ROWS X COLUMNS

$$A = \begin{bmatrix} 2 & 0 & -1 \\ -3 & 9 & 11 \end{bmatrix}$$

Dimension of A is 2x3

MATLAB: [r,c] = size(A)

# THE RANK OF A MATRIX IS THE NUMBER OF LINEARLY INDEPENDENT ROWS OR COLUMNS

$$A = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 2 & 1 \\ 2 & 7 & 8 \end{bmatrix} \quad \Rightarrow A_{ref} = \begin{bmatrix} 1 & 2 & 1 \\ 0 & 1 & 2 \\ 0 & 0 \end{bmatrix}$$

$$rank(A) = 2$$

- MATLAB: [k] = rank(A)
- If all rows/columns are linearly independent, the matrix is full rank

#### MATRIX MULTIPLICATION IS ASSOCIATIVE AND DISTRIBUTIVE BUT NOT COMMUTATIVE

$$A \cdot (B \cdot C) = (A \cdot B) \cdot C$$
$$A(B + C) = AB + AC$$
$$A \cdot B \neq B \cdot A$$

• MATLAB: D = A\*(B\*C)

# THE MATRIX TRANSPOSE IS FLIPPED OVER THE DIAGONAL

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix} \qquad A^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$$

$$(A^T)^T = A$$
$$(A + B)^T = A^T + B^T$$
$$(AB)^T = B^T A^T$$

- Symmetric:  $A = A^T$
- Skew-symmetric:  $A = -A^T$
- Orthogonal:  $A^{-1} = A^T$

MATCAB A'

#### **LINEAR ALGEBRA PRIMER**

### MATRIX INVERSE

### RECALL THE CALCULATION OF A MATRIX INVERSE

$$A^{-1} = \frac{\operatorname{adj}(A)}{|A|} = \frac{\left[c_{ij}\right]^T}{|A|}$$

$$c_{ij} = (-1)^{i+j} M_{ij}$$

Minor: Determinant of submatrix that is formed by deleting the  $i^{th}$  row and  $j^{th}$  column

Matrix is not invertible (singular) if the determinant is zero

#### **LET'S TRY A SIMPLE EXAMPLE**

$$A = \begin{bmatrix} 7 & 2 & 1 \\ 3 & -1 \\ -3 & 4 & -2 \end{bmatrix}$$

1. For the each term, find the minor

$$M_{11} = \begin{vmatrix} 3 & -1 \\ 4 & -2 \end{vmatrix} = -6 - (-4) = -2$$

2. Now complete the c matrix j  $j \rightarrow lolum$   $c_{11} = (-1)^{1+1}M_{11} = -2$ 

$$c = \begin{bmatrix} -2 & 3 & 9 \\ 8 & -11 & -34 \\ -5 & 9 & 21 \end{bmatrix}$$

$$\begin{pmatrix} -1 \end{pmatrix}^{1+2} M_{12} = -1 (-3) = 3$$

$$\begin{vmatrix} 0 & -1 \\ -3 & -2 \end{vmatrix} = 0 + 3 = 3$$

#### **CONTINUING THE EXAMPLE**

3. Completed *c* matrix

$$A = \begin{bmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{bmatrix} \qquad c = \begin{bmatrix} -2 & 3 & 9 \\ 8 & -11 & -34 \\ -5 & 7 & 21 \end{bmatrix}$$

Find determinant of A

$$|A| = \begin{vmatrix} 7 & 2 & 1 \\ 0 & 3 & -1 \\ -3 & 4 & -2 \end{vmatrix} = 1$$

Apply formula to determine inverse of A

$$A^{-1} = \frac{\begin{bmatrix} c_{ij} \end{bmatrix}^T}{|A|} = \frac{1}{1} \begin{bmatrix} -2 & 3 & 9 \\ 8 & -11 & -34 \\ -5 & 7 & 21 \end{bmatrix}^T = \begin{bmatrix} -2 & 8 & -5 \\ 3 & -11 & 7 \\ 9 & -34 & 21 \end{bmatrix}$$

### OUR INTEREST IS WITH A VERY SPECIAL MATRIX INVERSE

$$\frac{Y(s)}{U(s)} = C(sI - A)^{-1}B + D$$



$$\frac{Y(s)}{U(s)} = \frac{C \operatorname{adj}(sI - A) B}{|sI - A|} + D$$

- CHARECTERISTIC

Poles are determined by |sI - A| = 0 which is an eigenvalue problem!

# WE CAN APPLY MATRIX INVERSES TO CONVERTING STATE-SPACE MODELS INTO TRANSFER FUNCTIONS

$$\dot{X} = \dot{A}X + BU \longrightarrow \dot{D}DUC \qquad A = \begin{bmatrix} 0 & 1 \\ -\frac{k}{m} & -\frac{b}{m} \end{bmatrix}, \qquad B = \begin{bmatrix} 0 \\ \frac{1}{m} \end{bmatrix}$$

$$\dot{Y} = \dot{C}X + DU \longrightarrow \dot{D}UCDUC \qquad C = \begin{bmatrix} 1 & 0 \end{bmatrix}, \qquad D = 0$$

$$\dot{Y} = \dot{D} = 0$$

$$\dot{Y} = \dot{D} = 0$$

$$\dot{Y} = \dot{Z} = 0$$

$$\dot{Y} = \dot{Z} = 0$$

$$\dot{Y} = \dot{Z} = 0$$

 $\frac{Y(s)}{U(s)} = C(sI - A)^{-1}B + D$   $\lambda I - A = \begin{bmatrix} \lambda I - O & 1 & 1 \\ -k/m & \lambda I + b \\ m & m \end{bmatrix} \rightarrow \begin{bmatrix} \lambda & -1 & 1 \\ k/m & \lambda A & k \\ m & m \end{bmatrix}$   $= \begin{bmatrix} 1 & GT \begin{bmatrix} s & -1 & 1 \\ k/m & s & B/m \end{bmatrix} \begin{bmatrix} 0 & 1 \\ k/m & s & B/m \end{bmatrix}$ 

A LOVO PASS
HIGH VALUE

#### THE MATRIX INVERSION LEMMA IS OFTEN USEFUL WHEN PROVING STATE-SPACE FORMULATIONS

$$= \frac{1}{S^{2} + W_{m}S + k/m} \quad ms^{2} \rightarrow bs + k$$

$$(A + BC)^{-1} = A^{-1} - A^{-1}B(I + CA^{-1}B)^{-1}CA^{-1}$$
Not found D USE THIS

- Also called the Sherman–Morrison–Woodbury formula or Woodbury formula
- Used in Schur complement and Kalman filter

Ly Work to THIS Ly WILL TALK ABOUT

#### LINEAR ALGEBRA PRIMER

# EIGENVALUES AND EIGENVECTORS

### RECALL THE SOLUTION OF A FIRST ORDER ODE

$$\dot{x} = ax$$
  $x = ce^{at}$ 

#### Now, for a system of first order differential equations,

$$\dot{x}_1 = a_{11}x_1 + a_{12}x_2$$

$$\dot{x}_2 = a_{21}x_1 + a_{22}x_2$$

$$\lambda v_1 e^{\lambda t} = e^{\lambda t}(a_{11}v_1 + a_{12}v_2)$$

$$\lambda v_2 e^{\lambda t} = e^{\lambda t}(a_{21}v_1 + a_{22}v_2)$$

$$\lambda v_1 = a_{11}v_1 + a_{12}v_2$$

$$\lambda v_2 = a_{21}v_1 + a_{22}v_2$$

$$\lambda v_2 = a_{21}v_1 + a_{22}v_2$$

$$\lambda v = Av$$

$$A$$
EIGEN VECTORS & EIGENUATUES

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# THE EIGENVALUES OF STATE MATRIX A ARE THE ROOTS OF THE CHARACTERISTIC EQUATION

$$Ax = \lambda x$$

$$X + \lambda x + \delta$$

$$(\lambda x + \lambda x + \delta)$$

$$\lambda x + \lambda x + \delta$$

$$\lambda x +$$

$$\begin{cases} \gamma \in \mathcal{C} & \text{ones} & \text{B} = \mathbb{C} = \mathbb{D} = 0 \\ \gamma \in \mathcal{C} & \text{ones} & \text{B} = \mathbb{C} = \mathbb{D} = 0 \end{cases}$$

$$\begin{cases} \chi = \mathbb{C} & \text{ones} & \text{one$$

Z poles @ -1 3 -2, THE system is stable Ly 1st Order NO IMAG. poles

$$\lambda_{1} = -1$$

$$\begin{bmatrix} \lambda & -1 \\ 2 & \lambda + 3 \end{bmatrix}$$

$$\lambda_{1} = -1$$

$$\begin{bmatrix} -1 & -1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} \chi_{1} \\ \chi_{2} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \chi_{1} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_{1} = -2$$

$$\begin{bmatrix} -2 & -1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} \chi_{1}^{2} \\ \chi_{2}^{2} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \qquad \chi^{2} = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

#### **EIGENVECTORS**

 $bX_i = yX_i$ 

When initial conditions align with an eigenvector, the solution

When initial conditions align with an eigenvector, the solution stays on that eigenvector, i.e., if the system starts in a given mode, it follows that mode to equilibrium.

WE that mode to equilibrium.

Solution

Soluti

Eigenvectors diagonalize the A matrix.

L, FREQ

#### **EIGENVECTORS**

#### State vector written in terms of state variables:

$$x = x_1 \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix} + \dots + x_n \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

#### Represent state vector in terms of eigenvectors:

$$x = \alpha_1 x^1 + \alpha_2 x^2 + \dots + \alpha_n x^n$$
 Assume:  $\|x^i\| = 1$ : unit length WEIGTED Sum of the Vectors

#### **EIGENVECTORS**

$$\dot{X} = Ax$$

$$\frac{d}{dt}x = Ax = \alpha_1 Ax^1 + \alpha_2 Ax^2 + \dots + \alpha_n Ax^n$$

$$Ax^1 = \lambda i X^i$$

$$\dot{\mathbf{X}} = \lambda_1 \alpha_1 x^1 + \lambda_2 \alpha_2 x^2 + \dots + \lambda_n \alpha_n x^n$$

#### **LINEAR ALGEBRA PRIMER**

### MATRIX EXPONENTIAL

# THE MATRIX EXPONENTIAL WILL BE USED TO CALCULATE FORCED AND FREE RESPONSE

$$\dot{x} = Ax$$
  $x(t) = e^{At}x(0)$ 

More detail coming next section

# TO CALCULATE THE MATRIX EXPONENTIAL, ONE TRICK IS TO DIAGONALIZE THE MATRIX

$$A = T\Lambda T^{-1} \stackrel{\text{definition}}{=} T \qquad \qquad \ddots \qquad \qquad \lambda_n \end{bmatrix} T^{-1}$$

$$e^{At} = e^{T\Lambda T^{-1}t} = Te^{\Lambda t} T^{-1} = T \begin{bmatrix} e^{\lambda_1 t} & 0 \\ & \ddots & \\ 0 & & e^{\lambda_n t} \end{bmatrix} T^{-1}$$

So how do we diagonalize a matrix?

### WE NEED A STATE TRANSFORMATION TO DIAGONALIZE THE STATE MATRIX...

$$\dot{x} = Ax$$

$$T\dot{z} = ATz$$

$$\dot{z} = T^{-1}ATz$$

Choose T such that

ch that 
$$T^{-1}AT = \Lambda = \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}$$

 This is called a modal representation, poles appear on the diagonal and modes are decoupled

# BUT HOW DO WE CHOOSE THE TRANSFORMATION MATRIX TO MAKE THIS WORK?

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

$$A[v^1 \quad v^2 \quad \cdots \quad v^n] = \begin{bmatrix} v^1 & v^2 & \cdots & v^n \end{bmatrix} \begin{bmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{bmatrix}$$

$$T = \begin{bmatrix} v^1 & v^2 & \cdots & v^n \end{bmatrix}$$

 Assuming that the eigenvalues are real and distinct, choose T from the eigenvectors!

#### **RECALL THE PREVIOUS EXAMPLE**

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

$$\lambda_1 = -1, \qquad x^1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_1 = -2, \qquad x^2 = \begin{bmatrix} 1 \\ -2 \end{bmatrix}$$

$$T = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \qquad T^{-1} = \frac{1}{-1} \cdot \begin{bmatrix} -2 & -1 \\ 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

$$CHRNGE$$

$$CHRNGE$$

$$SIGN$$

$$T^{-1}AT = \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 \\ -1 & 2 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 0 & 0 \\ -1 & 2 & 0 & 0 & 0 \\ -1 & 2 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 0 & 0 \\ -1 & 2 & 0 & 0 & 0 \\ -1 & 2 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 2 & 1 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ -2$$

$$Z_{1}(t) = Z_{10}e^{-t}$$
 $Z_{2}(t) = Z_{20}e^{-2t}$ 
 $X_{1} = Z_{10}e^{-t}$ 
 $X_{1} = Z_{10}e^{-t}$ 
 $X_{1} = Z_{10}e^{-t}$ 
 $X_{1} = Z_{10}e^{-t}$ 
 $X_{1} = Z_{10}e^{-t}$ 

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X,H>=(2x=+ x20)e=+ (-X1=

#### **DOES THIS ALWAYS WORK?**

#### **Definitions**

- The multiplicity of an eigenvalue is called the algebraic multiplicity
- The number of linearly independent eigenvectors corresponding to a single eigenvalue is called the geometric multiplicity

### A matrix that has an eigenvalue whose geometric multiplicity is less than its algebraic multiplicity is called defective

- A defective matrix doesn't have enough eigenvectors
- A defective matrix is not completely diagonalizable
- But... a defective matrix can be almost diagonalized

### CONSIDER A GENERALIZED EIGENVECTOR

$$AT = T \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix}$$
$$T = \begin{bmatrix} v^1 & v^2 \end{bmatrix}$$

$$A[v^1 \quad v^2] = \begin{bmatrix} v^1 & v^2 \end{bmatrix} \begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix} = \begin{bmatrix} \lambda v^1 & v^1 + \lambda v^2 \end{bmatrix}$$

$$Av^{1} = \lambda v^{1}$$

$$Av^{2} = v^{1} + \lambda v^{2} \qquad \Rightarrow (A - \lambda I)v^{2} = v^{1}$$

#### THEN THE JORDAN CANONICAL FORM **ALLOWS ALMOST DIAGONALIZATION**

$$J_{i} = \begin{bmatrix} \lambda_{i} & 1 & 0 & \cdots & 0 \\ 0 & \lambda_{i} & 1 & \cdots & 0 \\ 0 & 0 & \lambda_{i} & \cdots & \vdots \\ \vdots & \vdots & \vdots & \ddots & 1 \\ 0 & 0 & 0 & \cdots & \lambda_{i} \end{bmatrix}$$

$$l_i = m_i - (n - r_i) + 1$$

$$J_{i} = \begin{bmatrix} \lambda_{i} & 1 & 0 & \cdots & 0 \\ 0 & \lambda_{i} & 1 & \cdots & 0 \\ 0 & 0 & \lambda_{i} & \cdots & \vdots \\ \vdots & \vdots & \vdots & \ddots & 1 \\ 0 & 0 & 0 & \cdots & \lambda_{i} \end{bmatrix} \qquad J = \begin{bmatrix} J_{1}(\lambda_{1}) & 0 & \cdots & 0 \\ 0 & J_{2}(\lambda_{2}) & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & J_{p}(\lambda_{p}) \end{bmatrix}$$

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

#### **EXAMPLE**

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

$$\lambda = -2, -2$$

$$\chi^{2}$$

### THIS RESULTS IN THE FOLLOWING TRANSFORMATION MATRIX

$$T = \begin{bmatrix} 1 & 1 & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$J = T^{-1}AT = \begin{bmatrix} 2 & 1 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & F2 \end{bmatrix}$$

$$Bcock$$

$$Bcock$$

$$Bcock$$

# A MATRIX IS ALMOST DIAGONALIZED, DOES THIS WORK FOR THE MATRIX EXPONENTIAL?

$$e^{At} = e^{T\Lambda T^{-1}t} = Te^{\Lambda t}T^{-1} = T\begin{bmatrix} e^{\lambda_1 t} & 0 \\ & \ddots & \\ 0 & e^{\lambda_n t} \end{bmatrix}T^{-1}$$

$$e^{At} = e^{TJT^{-1}t} = Te^{Jt}T^{-1} = T\begin{bmatrix} e^{J_1t} & 0 \\ & \ddots & \\ 0 & e^{J_nt} \end{bmatrix}T^{-1}$$

$$e^{J_{i}t} = \begin{bmatrix} e^{\lambda t} & te^{\lambda t} & \frac{t^{2}}{2}e^{\lambda t} & \cdots & \frac{t^{k-1}}{(k-1)!}e^{\lambda t} \\ 0 & e^{\lambda t} & te^{\lambda t} & \cdots & \vdots \\ 0 & 0 & e^{\lambda t} & \cdots & \frac{t^{2}}{2}e^{\lambda t} \\ \vdots & \vdots & \vdots & \ddots & te^{\lambda t} \\ 0 & 0 & 0 & \cdots & e^{\lambda t} \end{bmatrix}$$

#### RETURN TO THE PREVIOUS EXAMPLE

$$A = \begin{bmatrix} -2 & 3 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix} \qquad T = \begin{bmatrix} 1 & 1 & 0 \\ 0 & \frac{1}{3} & 0 \\ 0 & 0 & 1 \end{bmatrix} \qquad J = T^{-1}AT = \begin{bmatrix} -2 & 1 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & -2 \end{bmatrix}$$

$$e^{At} = Te^{Jt}T^{-1}$$

$$=T$$

$$T^{-1}$$

#### **COMING UP...**

#### **Solution of LTI State Equations**

- State Transition Matrix
- Free Response
- Forced Response

#### **Controllability**

- Definition of Controllability
- Controllable Canonical Form
- Controllable Decomposition
- Stabilizability