

16.31 Fall 2005  
Lecture Wed 28-Sep-05 ver 1.1

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September 28, 2005

**TODAY:**

Finish investigating the structure of  $\dot{x} = Ax$  given a transformation  $x = Mq$ .

**GOAL: Diagonalize the system matrix  $A$**

- Modal of form of

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

which is

$$\dot{q} = M^{-1}AMq + M^{-1}Bu \tag{1}$$

$$y = CMq + Du \tag{2}$$

is useful for analysis

- We want to find a state variable transformation  $x = Mq$  such that  $A$  is transformed into diagonal form

$$\Lambda = M^{-1}AM = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$$

- A transformation  $M$  that will do this is to let the columns of  $M$  be the set of eigenvectors  $x_i$  of the matrix  $A$ .

- We need to show that in the case that the system matrix  $A$  has  $n$  distinct eigenvalues  $\lambda_i$ , that  $A$  is diagonalized into  $\Lambda$ , but more importantly, we must show that the set of eigenvectors  $x_i$  are linearly independent, and thus form a basis for the state space.
- We will then show that in the case that  $A$  does not have  $n$  distinct eigenvalues, that it *may not* be possible to diagonalize  $A$  into  $\Lambda$ . In that case  $A$  can be placed into Jordan form  $J$ , and a set of  $n$  linearly independent generalized eigenvectors  $x_i$  can be found to form the columns of  $M$ .

## Review of Basis

- A basis of a vector space  $V$  of dimension  $n$  is a set of  $n$  linearly independent vectors. Any vector in the vector space is then a unique linear combination of the basis vectors. (DeRusso et al., p 75)
- Proof of uniqueness is by contradiction:  
Assume that the representation of a vector  $y \in V$  is not unique: That there are two sets of constants  $k_1, \dots, k_n$  and  $k'_1, \dots, k'_n$  not all equal to each other such that  $y$  is represented by both sets of constants.

$$y = k_1x_1 + \dots + k_nx_n \quad (3)$$

$$y = k'_1x_1 + \dots + k'_nx_n \quad (4)$$

$$\text{Subtract both representations from each other} \quad (5)$$

$$0 = y - y \quad (6)$$

$$= (k'_1 - k_1)x_1 + \dots + (k'_n - k_n)x_n \quad (7)$$

$$x_i \neq 0 \quad \text{implies} \quad (8)$$

$$k_i = k'_i \quad \text{which is a contradiction} \quad (9)$$

- Also note,  $x_i$ 's linearly independent implies the matrix

$$M = \begin{pmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{pmatrix}$$

has full rank  $r = n$  and nullity  $q = 0$  so that the only vector in its nullspace is 0.

- Alternative proof:

Let  $k = [k_1, \dots, k_n]^T$  and  $k' = [k'_1, \dots, k'_n]^T$ . Then

$$y = Mk \quad (10)$$

$$y = Mk' \quad (11)$$

$$y - y = M(k - k') \quad (12)$$

$$0 = M(k - k') \quad (13)$$

$$\text{but the only vector in } N(M) \text{ is } 0, \text{ so} \quad (14)$$

$$(k - k') = 0 \quad \text{which is a contradiction} \quad (15)$$

## Eigenvectors of $A$ form a basis for the State Space

- We need to show that the  $n$  distinct eigenvectors  $x_i$  each corresponding to the distinct eigenvalues  $\lambda_i$  form a basis, that is, they are linearly independent. Proof by contradiction: Assume that they are not linearly independent
- Proof by contradiction:

$$0 = k_1x_1 + \dots + k_nx_n \quad k_i \text{ not all zero}$$

Assume  $k_1 \neq 0$  and form the matrix

$$(A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)$$

. Premultiply both sides of the equation.

$$(A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)0 = (A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)(k_1x_1 + \dots + k_nx_n)$$

LHS is zero.

$$0 = (A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)(k_1x_1 + \dots + k_nx_n)$$

RHS

$$0 = k_1(A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)x_1 + \dots + \quad (16)$$

$$k_i(A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)x_i + \dots + \quad (17)$$

$$k_n(A - \lambda_2 I)(A - \lambda_3 I) \dots (A - \lambda_n I)x_n \quad (18)$$

Distributing  $x_i$

$$0 = k_1(\lambda_1 - \lambda_2)(\lambda_1 - \lambda_3) \cdots (\lambda_1 - \lambda_n)x_1 + \cdots + \quad (19)$$

$$k_i(\lambda_i - \lambda_2) \cdots (\lambda_i - \lambda_i) \cdots (\lambda_i - \lambda_n)x_i + \cdots + \quad (20)$$

$$k_n(\lambda_n - \lambda_2)(\lambda_n - \lambda_3) \cdots (\lambda_n - \lambda_n)x_n \quad (21)$$

Leaving

$$0 = k_1(\lambda_1 - \lambda_2)(\lambda_1 - \lambda_3) \cdots (\lambda_1 - \lambda_n)x_1$$

But  $\lambda_i \neq \lambda_j$  for  $i \neq j$  implies  $k_1 = 0$  which is a contradiction. Q.E.D.

- Same trick applies in the more general case of the Jordan form where the repeated eigenvalues *may* result in a degeneracy of eigenvectors. In that case we create a combination of  $n$  linearly independent eigenvalues and generalized eigenvectors, and the proof by contradiction proceeds in a similar manner!

## **A is Diagonalizable if it has $n$ distinct eigenvalues**

- Proof by construction:

Let  $M$  be matrix whose columns are the  $n$  linearly independent eigenvectors  $x_i$ .

We want so that

$$\Lambda = M^{-1}AM$$

It is easier to equivalently show that

$$M\Lambda = AM$$

$$\begin{pmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{pmatrix} \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix} = \begin{pmatrix} & & \\ A & & \\ & & \end{pmatrix} \begin{pmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{pmatrix}$$

$$\begin{pmatrix} | & & | \\ \lambda_1 x_1 & \cdots & \lambda_n x_n \\ | & & | \end{pmatrix} = \begin{pmatrix} | & & | \\ Ax_1 & \cdots & Ax_n \\ | & & | \end{pmatrix}$$

$$\begin{pmatrix} | & & | \\ \lambda_1 x_1 & \cdots & \lambda_n x_n \\ | & & | \end{pmatrix} = \begin{pmatrix} | & & | \\ \lambda_1 x_1 & \cdots & \lambda_n x_n \\ | & & | \end{pmatrix}$$

Q.E.D.

## Diagonalizability

- Diagonalizability is concerned with the eigenvectors
- (Invertibility is concerned with the eigenvalues!)
- The only connection between eigenvalues and diagonalizability is that diagonalization *can* fail if there are repeated eigenvalues.
- The test is to check, for an eigenvalue that is repeated  $q$  times, whether there are  $q$  independent eigenvectors.
- Examples

$$A = \begin{pmatrix} 3 & 1 \\ 0 & 3 \end{pmatrix} \quad \text{Repeated positive e-values. Not diagonalizable! (22)}$$

$$I = \begin{pmatrix} 1 & & \\ & 1 & \\ & & 1 \end{pmatrix} \quad \text{Repeated positive e-values. Diagonalizable! (23)}$$

$$0 = \begin{pmatrix} 0 & & \\ & 0 & \\ & & 0 \end{pmatrix} \quad \text{Repeated zero e-values. Diagonalizable! Not invertible! (24)}$$

## Jordan Form

- If a matrix  $A$  has  $s$  linearly independent eigenvectors, then it is similar to a matrix in Jordan form with  $s$  square blocks on the diagonal:

$$J = \begin{pmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{pmatrix}$$

- Each block  $J_i$  has one eigenvector, one eigenvalue, and 1's just above the diagonal.

$$J_i = \begin{pmatrix} \lambda_i & 1 & \\ & \ddots & 1 \\ & & \lambda_i \end{pmatrix}$$

- $J$  is equal to

$$J = M^{-1}AM$$

where the columns  $M$  contain the  $s$  linearly independent eigenvectors and  $n - s$  generalized eigenvectors.

- This implies

$$AM = MJ$$

or

$$\begin{pmatrix} & A \\ & \end{pmatrix} \begin{pmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{pmatrix} = \begin{pmatrix} | & & | \\ x_1 & \cdots & x_n \\ | & & | \end{pmatrix} \begin{pmatrix} J_1 & & \\ & \ddots & \\ & & J_s \end{pmatrix}$$

- Multiplying through we find that

$$Ax_i = \lambda_i x_i \quad \text{or} \quad Ax_i = \lambda_i x_i + x_{i-1}$$

- DeRusso et al., pp 140–143, gives an algorithm for calculating generalized eigenvectors.
- Each  $J_i$  can also be written as  $\Lambda_i + N_i$  where  $N_i$  is a matrix of zeros along the diagonal and 1's along the superdiagonal.

$$J_i = \Lambda_i + N_i$$

$$\begin{pmatrix} \lambda_i & 1 & \\ & \ddots & 1 \\ & & \lambda_i \end{pmatrix} = \begin{pmatrix} \lambda_i & & \\ & \ddots & \\ & & \lambda_i \end{pmatrix} + \begin{pmatrix} 0 & 1 & \\ & \ddots & 1 \\ & & 0 \end{pmatrix}$$

- For argument lets say that  $J_i$ ,  $\Lambda_i$ , and  $N_i$  are  $n \times n$  matrices and drop the  $i$  subscript for now. Then

$$N^n = 0$$

- We can now calculate  $e^{Jt}$

$$e^{Jt} = e^{(\Lambda+N)t} \tag{25}$$

$$= e^{\Lambda t} e^{tN} \tag{26}$$

$$= e^{\lambda t} \sum_{k=0}^{n-1} \frac{t^k N^k}{k!} \tag{27}$$

or in matrix form

$$e^{Jt} = e^{\lambda t} \begin{pmatrix} 1 & t & t^2/2! & \cdots & t^{n-1}/(n-1)! \\ & 1 & t & & t^{n-2}/(n-2)! \\ & & 1 & & \\ & & & \ddots & \\ & & & & t \\ & & & & 1 \end{pmatrix}$$

- We will rarely use Jordan form in class, but it is important to
  1. Know that it exists and that it is the most general form of diagonalization
  2. Know and understand the form of  $e^{Jt}$
- *However*, the Jordan form does lead us to an interesting discussion about system models and numerical computation with system models

## FRIDAY:

- Look at an interesting modeling/numerical issue that can arise and is related to our discussion on diagonalization and Jordan forms.
- Summarize transformations
- Find a better way to calculate  $e^{At}$  given that it shows up so much in the solutions to

$$\dot{x} = Ax + Bu \tag{28}$$

$$y = Cx + Du \tag{29}$$