

## LECTURE 33

Diagonal Form of a Matrix.

Hamilton's Theorem

The Internal Input-Output Processing  
Structure of a Matrix

Diagonalizable Matrices.

## I. DIAGONAL FORM OF A MATRIX

Read String 5.2

33.1

One has

$$A \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix} = \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix}$$

The essential properties of a  $n \times n$  matrix with a set of  $n$  linearly independent eigen vectors are expressed in terms of its eigenvalues and the transition matrix formed from these eigen vectors. This is expressed by the following

Proposition 33.1

Let  $A : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be an  $n \times n$  matrix.

Let  $\vec{A}$  have a set of  $n$  lin. indep. eigen vectors  $\{\vec{e}_i\}$

Let  $S$  be the transition matrix formed from these eigenvalues.

Then

$$S^{-1}AS = \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \equiv \Lambda$$

One says that  $A$  can be diagonalized.

proof: Consider the equations

$$A\vec{e}_i = \lambda_i \vec{e}_i \quad i=1, 2, \dots, n$$

satisfied by the given eigenvectors.

33.2

$$\begin{aligned} A \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix} &= \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix} \\ &= \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix} \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} \begin{bmatrix} \vec{e}_1 & \dots & \vec{e}_n \end{bmatrix}^{-1} \end{aligned}$$

$$A = S \Lambda S^{-1} \quad \text{where } S \text{ is invertible (why)}$$

$$\text{Thus } S^{-1}AS = \Lambda \quad \text{or} \quad A = S\Lambda S^{-1}$$

A is diagonalizable by means of a similarity transformation.

Comments:

1. It is easy to compute powers of a diagonalizable matrix

$$\begin{aligned} A^m &= (S\Lambda S^{-1})^m \quad (\Lambda \text{ is diagonal}) \\ &= S \Lambda^m S^{-1} \end{aligned}$$

2. One can compute matrix power series!
- Consider a function defined by a convergent power series

$$g(x) = a_0 + a_1 x + a_2 x^2 + \dots ; \text{ e.g. } e^x = 1 + x + \frac{x^2}{2} + \dots$$

3.3.3

Then  $g(A)$  is defined and convergent in an appropriate sense:

$$\begin{aligned} e^{At} &= I + At + A^2 \frac{t^2}{2} + \dots \\ &= ST^{-1} + SAtS^{-1} + SAt^2 \frac{S^{-1}}{2!} + \dots \\ &= S(I + At + \frac{At^2}{2!} + \dots)S^{-1} \\ &= S \begin{bmatrix} e^{At} & & & \\ & e^{At} & 0 & \\ & 0 & e^{At} & \\ & & & e^{At} \end{bmatrix} S^{-1} \end{aligned}$$

$$\text{Thus } g(A) = S^{-1} f(At) S$$

4. Question: True or False:  $\text{tr } AB = \text{tr } BA$ ?

Answer: True. Indeed one has

$$\text{tr } AB = \sum_{i,j} A_{ij}B_{ji} = \sum_{i,j} B_{ji}A_{ij} = \text{tr } BA$$

3. Consider the secular (a, b, a, characteristic) polynomial

$$\det |A - \lambda I| = a_0(-\lambda) + \dots + a_m(\lambda)^m$$

The secular polynomial in the powers of  $\lambda$  vanishes:

$$f(A) = S^{-1} f(\lambda) S =$$

$$= S^{-1} \begin{bmatrix} (\lambda_1 - \lambda)(\lambda_2 - \lambda) & \dots & (\lambda_1 - \lambda) \\ \vdots & \ddots & \vdots \\ 0 & \dots & (\lambda_m - \lambda) \end{bmatrix} S = S^{-1} \begin{bmatrix} 0 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 0 \end{bmatrix} = [0]_{m \times m}$$

This validates Eq. (#).

We have

$$\begin{aligned} f(\lambda) &= \lambda_1 \dots \lambda_n + (\lambda_1 \lambda_2 + \dots + \lambda_1 \lambda_n) \lambda_1^2 + \dots + (\lambda_1 + \lambda_2 + \dots + \lambda_n) \lambda_1^n + (-\lambda)^n \\ f(A) &= a_0 + a_1 f(A) + a_2 f(A)^2 + \dots + a_n f(A)^n + (-A)^n = [0]_{m \times m} \end{aligned}$$

This illustrates the Cayley-Hamilton theorem:

Theorem 33.1: A matrix satisfies its characteristic equation

Here the coefficients

$$a_0 = f(0) = \lambda_1 + \lambda_2 + \dots + \lambda_n = \text{det } A$$

$$a_{n-1} = \lambda_1 + \lambda_2 + \dots + \lambda_n = \sum_{i=1}^n \lambda_i = \text{trace } A \quad (\#)$$

depend only on the eigenvalues of  $A$ .

4. Question: True or False:  $\text{tr } AB = \text{tr } BA$ ?

Answer: True.

$$\text{tr } AB = \sum_{i,j} A_{ij}B_{ji} = \sum_{i,j} B_{ji}A_{ij} = \text{tr } BA$$

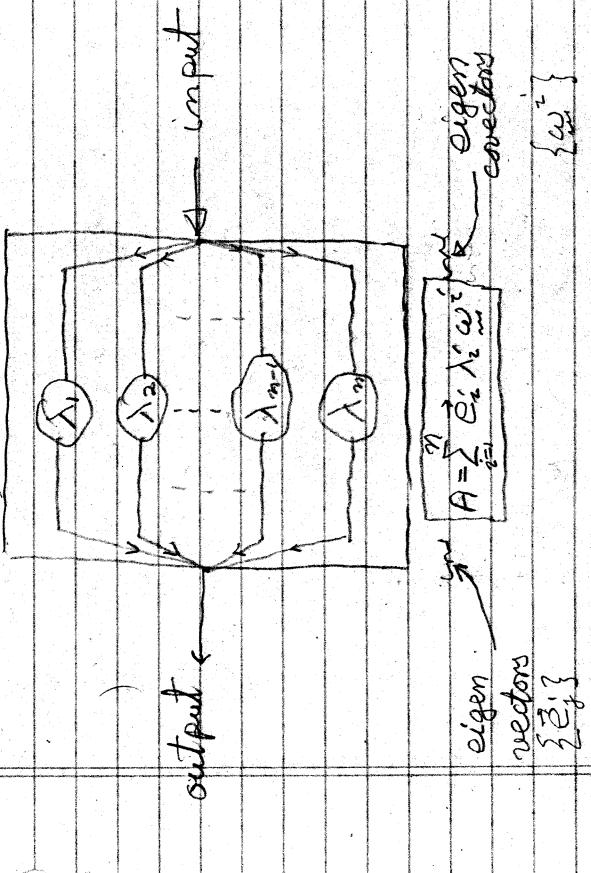
Application: For  $S^{-1}AS = \Lambda$  one has

$$\begin{aligned} \lambda_1 + \lambda_2 + \dots + \lambda_n &= \text{tr } \Lambda \\ &= \text{tr } S^{-1}AS \\ &= \text{tr } A SS^{-1} \\ &= \text{tr } A \\ &= \text{tr } A \quad [\text{tr } 1 = \text{tr } A] \end{aligned}$$

33.5

4. A diagonalizable matrix  $A$  is a linear input-output system whose internal structure is characterized by its eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_m$

Contact with its exterior is provided by its eigenvector basis and the corresponding dual basis. Diagrammatically one has



33.6

A matrix, say  $A$ , is a linear input-output device. Its output exhibits a causal relation ship to the input. As depicted in the diagram this relationship is the result of a three-stage process.

1. The input is decomposed by the eigenvector basis { $e_i$ } (see page 33.8) into

its eigenvector components,

2. Each eigenvector component gets multiplies ("amplified", "phase-shifted", "filtered") by its respective eigenvalue  $\lambda_i$ .

3. The results of this modification get recombined by the eigenvector basis { $e_i$ }. This recombination is the output from  $A$ .

(b) The mathematical (here algebraic) formulation of the diagram on page 33.6 is achieved by the following

(a) This diagram conceptualizes the following physical process:

4-step process:

33.7

Let  $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_m\}$  be the eigenvector basis corresponding to  $\{A, \lambda_1, \lambda_2, \dots, \lambda_m\}$

Let  $\{\vec{e}_1^*, \vec{e}_2^*, \dots, \vec{e}_m^*\}$  be the corresponding cobasis for the dual space  $V^*$  (See Lecture 8).

The two bases are related by the duality relation

$$\langle \vec{e}_i | \vec{e}_j^* \rangle = \delta_{ij}.$$

Explicitly one has

$$\vec{e}_j^* = \begin{bmatrix} e_1^* \\ e_2^* \\ \vdots \\ e_m^* \end{bmatrix}, \text{ the } j^{\text{th}} \text{ basis vector;}$$
$$A \vec{e}_j^* = \lambda_j \vec{e}_j^*.$$

STEP II Construct the cobasis elements

$$\vec{e}_i^* = [\omega_1^i, \omega_2^i, \dots, \omega_m^i]$$

from the duality relation

$$\delta_{ij}^* = \langle \vec{e}_i^* | \vec{e}_j^* \rangle \equiv [\omega_1^i, \dots, \omega_m^i] \begin{bmatrix} e_1^* \\ e_2^* \\ \vdots \\ e_m^* \end{bmatrix}$$

STEP III

Exhibit the matrix

$$A = \sum_{i=1}^m \vec{e}_i \cdot \lambda_i \vec{e}_i^*$$

STEP IV

Validate the correctness of this representation by applying  $A$  to the vector

$$\vec{x} = \sum_{j=1}^m \vec{e}_j < \omega_1^j | \vec{x}^* > = \sum_{j=1}^m \vec{e}_j \cdot x^j$$

we have

33.9

$$A(\vec{x}) = \sum_i \vec{e}_i \lambda_i \vec{e}_i^T (\vec{x})$$

$$\begin{aligned} &= \sum_i \vec{e}_i \lambda_i \langle \vec{e}_i^T | \vec{e}_j \vec{x} \rangle \\ &= \sum_i \vec{e}_i \lambda_i \langle \vec{e}_i | \vec{e}_j \rangle \vec{x}^j \end{aligned}$$

$$\stackrel{i}{\vec{x}}$$

$$= \sum_{i=1}^n \vec{e}_i \lambda_i \vec{x}^i$$

If  $\vec{x}$  is one of the eigenvectors, say

$$\vec{x} = \vec{e}_k$$

Then

$$A \vec{e}_k = \sum_i \vec{e}_i \lambda_i \langle \vec{e}_i^T | \vec{e}_k \rangle$$

$$\stackrel{i}{\vec{e}_k}$$

$$A \vec{e}_k = \vec{e}_k \lambda_k \quad (\text{not sum over } k)$$

$$\text{Thus } A = \sum_{i=1}^n \vec{e}_i \lambda_i \vec{e}_i^T$$

mathematically the diagram on page  
33.5 indeed

33.10

5. The eigenvectors  $\{\vec{e}_i\}$ ,

$$A \vec{e}_j = \lambda_j \vec{e}_j \quad j = 1, \dots, n$$

are right eigenvectors. By contrast

$\{\vec{e}_k\}$  are left eigenvectors. Indeed,

$$\vec{e}_k^T A = \langle \vec{e}_k^T | \sum_i \vec{e}_i \lambda_i \vec{e}_i^T | \vec{e}_m \rangle$$

$$= \sum_i \langle \vec{e}_k^T | \vec{e}_i \rangle \lambda_i \vec{e}_m^T$$

$$\boxed{\vec{e}_k^T A = \lambda_k \vec{e}_m^T}$$

### Comment

33.1/

Notice that the decomposition of  $A$ ,

$$A = \sum_{i=1}^n \lambda_i w_i w_i^\top$$

and hence the eigenvalue  $\lambda_i$  in

$$A\vec{x} = \lambda \vec{x}$$

did not in any way depend on the existence of an inner product.

The only necessary ingredient was the construction of the dual eigenbasis  $\{w_i^\top\}$  from the already known eigenbasis  $\{\vec{e}_i\}$  of  $A$ .

### II. DIAGONALIZABLE MATRICES

33.2

It is not obvious that all matrices are

diagonalizable. In fact it is not even true as exemplified by the ("defective")

matrix  $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ , whose characteristic

polynomial  $(-\lambda)^2 = 0$  has degenerate roots,

However if the roots are distinct, then

there exists a corresponding set of eigen-vectors. This fact is made precise by the following

Theorem 33.2

A matrix with  $n$  distinct eigenvalues has a set of  $n$  eigenvectors which is linearly independent.

**Nota Bene:** The existence of each such eigenvector is guaranteed by THEOREM 23.1 on page 23.4.

33.13  
Because of this, one has the following

### Corollary

A matrix with distinct eigenvalues can be diagonalized.

### Comment

1. The converse is not true. There are many many diagonalizable matrices with degenerate eigenvalues.

For example,  $\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

2. The theorem does not apply to "defective" matrices.

33.14  
Proof of the linear independence of the set of n eigenvectors having n distinct eigenvalues.

### Step I

Consider

$$c_1 \vec{e}_1 + c_2 \vec{e}_2 = \vec{0} \quad (1)$$

Apply A to both sides and obtain

$$c_1 \lambda_1 \vec{e}_1 + c_2 \lambda_2 \vec{e}_2 = \vec{0}$$

Multiply Eq.(1) by  $\lambda_2$  and subtract. One obtains

$$c_1 (\lambda_2 - \lambda_1) \vec{e}_1 = \vec{0}$$

By hypothesis  $\lambda_2 \neq \lambda_1$ . Thus  $c_1 = 0$ .

Introduce this result into Eq.(1), one finds  $c_2 \vec{e}_2 = \vec{0}$ .

Thus  $c_2 = 0$ . It follows that the set of eigenvectors in Eq.(1),

$\{\vec{e}_1, \vec{e}_2\}$  is a linearly independent set.]

### Step II

One now repeats this line of reasoning by considering the linear combination

$$c_1 \vec{e}_1 + c_2 \vec{e}_2 + c_3 \vec{e}_3 = \vec{0}$$

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to conclude that

$\{e_1, e_2, e_3\}$  is a l.i. set.

Step III. More generally one assumes that

$\{e_1, e_2, \dots, e_k\}$  is a lin. indep. set

and considers

$$c_1 \vec{e}_1 + \dots + c_k \vec{e}_k + c_{k+1} \vec{e}_{k+1} = \vec{0} \quad (3)$$

in order to show that  
 $\{e_1, e_2, \dots, e_k, e_{k+1}\}$  is a lin. indep. set.

This is certainly true for  $k=1$  and  $k=2$ .

Step IV.

Apply A to Eq.(3) and obtain

$$c_1 \lambda_1 \vec{e}_1 + \dots + c_k \lambda_k \vec{e}_k + c_{k+1} \lambda_{k+1} \vec{e}_{k+1} = \vec{0}. \quad (4)$$

Multiply Eq.(3) by  $\lambda_{k+1}$ , subtract the result

from Eq.(4) and obtain

$$c_1 (\lambda_1 - \lambda_{k+1}) \vec{e}_1 + \dots + c_k (\lambda_k - \lambda_{k+1}) \vec{e}_k = \vec{0}$$

By hypotheses  $\lambda_1 \neq \lambda_{k+1}, \dots, \lambda_k \neq \lambda_{k+1}$ . Thus

Eq.(2) implies

$$c_1 = c_2 = \dots = c_k = 0,$$

33.16

Introduce this result into Eq.(3); thus

$$c_{k+1} \vec{e}_{k+1} = \vec{0}$$

Consequently, all coefficients vanish:

$$c_1 = \dots = c_k = c_{k+1} = 0$$

This means that

$\{e_1, e_2, \dots, e_k, e_{k+1}\}$  is also a lin. indep. set.

Step V.

Repeat steps III and IV until one has exhausted all n eigenvalues.

The result is that

$$\{\vec{e}_1, \dots, \vec{e}_n\}$$
 is a lin. indep. set.