

## MTH 100 : Lecture 33

Theorem (Diagonalization Theorem):

- (a) An  $n \times n$  matrix  $A$  is diagonalizable if and only if  $A$  has  $n$  linearly independent eigenvectors.
- (b) In this case,  $A = PDP^{-1}$  where the columns of  $P$  are  $n$  linearly independent eigenvectors of  $A$ , and the diagonal entries of  $D$  are eigenvalues corresponding to these eigenvectors.

Another way to express the above Theorem

An  $n \times n$  matrix  $A$  is diagonalizable if and only if it has enough (linearly independent) eigenvectors to form a basis of  $\mathbb{R}^n$ .

Such a basis is called an eigenvector basis.

Note: Check if the matrix considered in the last lecture is diagonalizable.

## Proof of the Diagonalization Theorem:

(a)  $\Rightarrow$ : Suppose  $A$  is diagonalizable  
Want to show that  $A$  has  $n$  linearly independent eigen vectors.

Now  $A = PDP^{-1}$  for some diagonal matrix  $D$  and some invertible matrix  $P$ .

$$\text{Therefore } AP = PD \quad \dots \dots \dots \quad (1)$$

Let  $P = [v_1 \ v_2 \ \dots \ v_n]$  in column form

and Let  $D = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$  where the  $\lambda$ 's need not be distinct.

Then (1) becomes

$$A[v_1 \ v_2 \ \dots \ v_n] = [v_1 \ v_2 \ \dots \ v_n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$$\Rightarrow [Av_1 \ Av_2 \ \dots \ Av_n] = [\lambda_1 v_1 \ \lambda_2 v_2 \ \dots \ \lambda_n v_n]$$

$$\Rightarrow Av_i = \lambda_i v_i \text{ for } i=1, 2, \dots, n$$

Therefore  $v_1, v_2, \dots, v_n$  are eigen vectors

corresponding to eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$

Also the vectors  $v_1, v_2, \dots, v_n$  being columns of an invertible matrix  $P$  are linearly independent.

$\Leftarrow$ : Suppose  $A$  has  $n$  linearly independent eigen vectors.

Want to show that  $A$  is diagonalizable.

Let  $v_1, v_2, \dots, v_n$  be the  $n$  linearly independent eigen vectors of  $A$  corresponding to the eigen values  $\lambda_1, \lambda_2, \dots, \lambda_n$  (not necessarily distinct).

Then  $A v_i = \lambda_i v_i$  for  $i=1, 2, \dots, n$

We form the matrix  $P$  with the  $v_i$ 's as columns.

$$\text{i.e. } P = [v_1 \ v_2 \ \dots \ v_n]$$

$$\text{Then } AP = A[v_1 \ v_2 \ \dots \ v_n]$$

$$= [Av_1 \ Av_2 \ \dots \ Av_n]$$

$$= [\lambda_1 v_1 \ \lambda_2 v_2 \ \dots \ \lambda_n v_n]$$

$$= [v_1 \ v_2 \ \dots \ v_n] \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

$$= PD \text{ where } D = \begin{bmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{bmatrix}$$

Now since  $v_1, v_2, \dots, v_n$  are linearly independent, the matrix  $P = [v_1 \ v_2 \ \dots \ v_n]$  is invertible and  $A = P D P^{-1}$ . Hence  $A$  is diagonalizable.

(b) Part (b) of the theorem has been proved enroute to proving part (a) of the theorem. (QED)

In practice, we can distinguish three cases:

- Case 1: An  $n \times n$  matrix A has  $n$  distinct (real) eigen values.

Then we have the following result.

Proposition: An  $n \times n$  matrix A with  $n$  distinct eigen values is diagonalizable.

Proof: By an earlier proposition, eigen vectors corresponding to distinct eigenvalues are linearly independent. Therefore in this case A has  $n$  linearly independent eigen vectors. Hence by Diagonalization Theorem, A is diagonalizable.

Two Preliminary Definitions:

Given an eigenvalue  $\lambda_1$  for a matrix A

We define

- The algebraic multiplicity of  $\lambda_1$  is the

power of the factor  $(\lambda - \lambda_1)$  in the characteristic polynomial of  $A$ .

- The geometric multiplicity of  $\lambda_1$  is the dimension of the eigen space corresponding to  $\lambda_1$ .

Note: Algebraic multiplicity applies to polynomials in general (not only characteristic polynomial). The geometric multiplicity applies specifically to the characteristic polynomial (since its roots are eigenvalues which have corresponding eigenspaces).

Case 2: An  $n \times n$  matrix  $A$  has  $p < n$  distinct eigenvalues, but counting the (algebraic) multiplicities, there are  $n$  real eigenvalues (not distinct).

We then have a weaker result for this case.

Proposition: Let  $A$  be an  $n \times n$  matrix with  $n$  (real) eigenvalues (counting algebraic multiplicities) of which only  $\lambda_1, \lambda_2, \dots, \lambda_p$  are distinct ( $p < n$ ).

Then the following hold:

- For  $1 \leq k \leq p$ , the geometric multiplicity

of  $\lambda_k$  is less than or equal to the algebraic multiplicity of  $\lambda_k$ .

(b)  $A$  is diagonalizable if and only if the sum of the dimensions of the distinct eigenspaces is  $n$  and this happens if and only if the geometric multiplicity for each  $\lambda_k$  equals its algebraic multiplicity.

(c) If  $A$  is diagonalizable and  $B_k$  is a basis for the eigenspace corresponding to  $\lambda_k$  for each  $k$ , then the total collection of vectors in  $B_1, B_2, \dots, B_p$  forms an eigenvector basis for  $\mathbb{R}^n$ .

Note: This result is weaker because unlike case 1,  $A$  is not automatically diagonalizable.  $A$  has to satisfy the additional condition (b) and this may not happen for all matrices.

Ex ① Let  $A = \begin{bmatrix} 42 & -33 \\ 22 & -13 \end{bmatrix}$

Then  $\det(A - \lambda I) = \begin{vmatrix} 42-\lambda & -33 \\ 22 & -13-\lambda \end{vmatrix}$

$$= (42-\lambda)(-13-\lambda) + 22(33)$$

$$= -546 - 42\lambda + 13\lambda + \lambda^2 + 726$$

$$= 180 - 29\lambda + \lambda^2 = (\lambda - 20)(\lambda - 9)$$

Thus here there are 2 distinct eigenvalues

$$\lambda_1 = 20, \quad \lambda_2 = 9$$

For  $\lambda_1 = 20$   $A - \lambda_1 I = \begin{bmatrix} 42-20 & -33 \\ 22 & -13-20 \end{bmatrix}$

$$= \begin{bmatrix} 22 & -33 \\ 22 & -33 \end{bmatrix} \xrightarrow{R_2 \rightarrow R_2 - R_1} \begin{bmatrix} 22 & -33 \\ 0 & 0 \end{bmatrix} \xrightarrow{R_1 \rightarrow \frac{1}{22}R_1} \begin{bmatrix} 1 & -\frac{3}{2} \\ 0 & 0 \end{bmatrix} = \text{RREF matrix}$$

so,  $(A - \lambda_1 I)x = 0 \Rightarrow \begin{cases} x_1 = \frac{3}{2}x_2 \\ x_2 = x_2 \end{cases} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} \frac{3}{2} \\ 1 \end{bmatrix}$

Let us take  $v_1 = 2 \begin{bmatrix} \frac{3}{2} \\ 1 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$  as an eigen vector corresponding to  $\lambda_1 = 20$   
(by taking  $x_2 = 2$ )

Check  $A v_1 = \begin{bmatrix} 42 & -33 \\ 22 & -13 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 60 \\ 40 \end{bmatrix} = 20 \begin{bmatrix} 3 \\ 2 \end{bmatrix} = 20 v_1$

$$\text{For } \lambda_2 = 9 \quad A - \lambda_2 I = \begin{bmatrix} 42-9 & -33 \\ 22-9 & -13-9 \end{bmatrix} = \begin{bmatrix} 33 & -33 \\ 22 & -22 \end{bmatrix}$$

$$\text{RREF matrix} = \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix} \xleftarrow{\substack{R_2 \rightarrow R_2 - 22R_1}} \begin{bmatrix} 1 & -1 \\ 22 & -22 \end{bmatrix} \xleftarrow{R_1 \rightarrow \frac{1}{33}R_1} \begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}$$

$$\text{So, } (A - \lambda_2 I)x = 0 \Rightarrow \begin{cases} x_1 = x_2 \\ x_2 = x_2 \end{cases} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Let us take  $v_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  as an eigen-vector  
 (By taking  $x_2 = 1$ ) corresponding to eigenvalue  $\lambda_2 = 9$

Check  $A v_2 = \begin{bmatrix} 42 & -33 \\ 22 & -13 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 9 \\ 9 \end{bmatrix} = 9 v_2$

Note that we should get  $A = P D P^{-1}$   
 where  $D = \begin{bmatrix} 20 & 0 \\ 0 & 9 \end{bmatrix}$  and  $P = [v_1, v_2] = \begin{bmatrix} 3 & 1 \\ 2 & 1 \end{bmatrix}$

Easier to check  $AP = PD$

$$\text{Now } AP = \begin{bmatrix} 42 & -33 \\ 22 & -13 \end{bmatrix} \begin{bmatrix} 3 & 1 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} 60 & 9 \\ 40 & 9 \end{bmatrix}$$

$$\text{and } PD = \begin{bmatrix} 3 & 1 \\ 2 & 1 \end{bmatrix} \begin{bmatrix} 20 & 0 \\ 0 & 9 \end{bmatrix} = \begin{bmatrix} 60 & 9 \\ 40 & 9 \end{bmatrix}$$

$$\text{So, } AP = PD \quad \text{as desired.}$$