

**Lecture 19: 5.3 Diagonalization.**

Let us clarify one point. Suppose that  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear map with matrix  $A$ , and that  $\mathcal{B}$  is a new basis for  $\mathbb{R}^n$  and  $\mathcal{B}'$  is a new basis for  $\mathbb{R}^m$ .

$$\begin{array}{ccc} \mathbf{x} & \xrightarrow{A} & T(\mathbf{x}) \\ P_{\mathcal{B}} \uparrow & & \uparrow P_{\mathcal{B}'} \\ [\mathbf{x}]_{\mathcal{B}} & \xrightarrow{C} & [T(\mathbf{x})]_{\mathcal{B}'} \end{array}$$

If  $C$  is the matrix of  $T$  in the new bases  $\mathcal{B}$  and  $\mathcal{B}'$ , then

$$A = P_{\mathcal{B}'} C P_{\mathcal{B}}^{-1}.$$

**Special Cases: Row reduction:**  $A = EC$  where  $C$  is the reduced Echelon form and  $E$  is a product of elementary transformations. Then  $E$  represents the changing the basis for  $\mathbb{R}^m$ . The basis for  $\mathbb{R}^n$  remains the same.

**Diagonalization:** If  $A$  is an  $n \times n$  matrix, we say it can be **diagonalized** if we can write  $A = PDP^{-1}$  where  $D$  is a diagonal matrix (i.e. the entries off the main diagonal are all zeros). If we think of  $P$  as a change of basis matrix, then we are changing the basis of the map  $\mathbf{x} \rightarrow A\mathbf{x}$  from  $\mathbb{R}^n$  to  $\mathbb{R}^n$  in the same way on both the domain and the image. The new basis consists of eigenvectors of the original matrix  $A$ . Diagonalizing  $A$  can be used to compute  $A^k$ , for large  $k$ , which is useful in the applications.

(If multiplying by  $A$  represents the evolution of a system during one time unit then multiplying by  $A^k$  represents the evolution after  $k$  time units)

**Ex** Let  $D = \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix}$ . Compute  $D^2$ ,  $D^3$  and  $D^k$ .

**Sol**  $D^2 = \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix} = \begin{bmatrix} 5^2 & 0 \\ 0 & 4^2 \end{bmatrix}$ ,  $D^3 = DD^2 = \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} 5^2 & 0 \\ 0 & 4^2 \end{bmatrix} = \begin{bmatrix} 5^3 & 0 \\ 0 & 4^3 \end{bmatrix}$ ,  
 $D^k = \begin{bmatrix} 5^k & 0 \\ 0 & 4^k \end{bmatrix}$

**Ex** Let  $A = \begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix}$ . Compute  $A^k$ .

**Sol** The trick to computing  $A^k$  in general is to diagonalize the matrix  $A$  and to do this we compute the eigenvalues and eigenvectors. Indeed,

$$\begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad \begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 4 \begin{bmatrix} 1 \\ 2 \end{bmatrix}.$$

so the columns of  $P$  are made out of the eigenvectors of  $A$  and the diagonal entries of  $D$  are the eigenvalues of  $A$ . We can put this to equations together in one matrix equation:

$$\begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 5 & 4 \\ 5 & 8 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix},$$

i.e.

$$\begin{bmatrix} 6 & -1 \\ 2 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 4 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}^{-1},$$

We have  $A^2 = PDP^{-1}PDP^{-1} = PDIDP^{-1} = PD^2P^{-1}$ ,

$$A^k = PD^kP^{-1} = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} 5^k & 0 \\ 0 & 4^k \end{bmatrix} \begin{bmatrix} 2 & -1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 2 \cdot 5^k - 4^k & -5^k + 4^k \\ 2 \cdot 5^k - 2 \cdot 4^k & -5^k + 2 \cdot 4^k \end{bmatrix}$$

A square matrix  $A$  is called **diagonalizable** if it can be written  $A = PDP^{-1}$ , where  $D$  is diagonal and  $P$  is invertible.

When is  $A$  diagonalizable and if it is how do we find  $D$  and  $P$ ?

In general if  $A$  is an  $n \times n$  matrix with  $n$  linearly independent eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  and eigenvalues  $\lambda_1, \dots, \lambda_n$  then

$$A[\mathbf{v}_1 \ \cdots \ \mathbf{v}_n] = [A\mathbf{v}_1 \ \cdots \ A\mathbf{v}_n] = [\lambda_1\mathbf{v}_1 \ \cdots \ \lambda_n\mathbf{v}_n] = [\mathbf{v}_1 \ \cdots \ \mathbf{v}_n] \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix}$$

$$\Rightarrow A = [\mathbf{v}_1 \ \cdots \ \mathbf{v}_n] \begin{bmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{bmatrix} [\mathbf{v}_1 \ \cdots \ \mathbf{v}_n]^{-1}$$

**Diagonalization Theorem** An  $n \times n$  matrix is diagonalizable  $A$  if and only if it has  $n$  linearly independent eigenvectors.

**Ex** If possible, diagonalize  $A = \begin{bmatrix} 2 & 0 & 0 \\ 1 & 2 & 1 \\ -1 & 0 & 1 \end{bmatrix}$ .

**Sol** The eigenvalues  $\det(A - \lambda I) = \begin{vmatrix} 2 - \lambda & 0 & 0 \\ 1 & 2 - \lambda & 1 \\ -1 & 0 & 1 - \lambda \end{vmatrix} = (2 - \lambda)^2(1 - \lambda) = 0$ .

Basis for  $\lambda = 1$ :  $\mathbf{v}_1 = \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix}$ .

Basis for  $\lambda = 2$ :  $\mathbf{v}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$ ,  $\mathbf{v}_3 = \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ .

Construct  $P = [\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3] = \begin{bmatrix} 0 & 0 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$ ,  $D = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$ .  $A = PDP^{-1}$ .

**Ex** If possible, diagonalize  $A = \begin{bmatrix} 2 & 4 & 6 \\ 0 & 2 & 2 \\ 0 & 0 & 4 \end{bmatrix}$ .

**Sol** The eigenvalues  $\det(A - \lambda I) = (\lambda - 2)^2(\lambda - 4) = 0$ .

Basis for  $\lambda = 2$ :  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ .

Basis for  $\lambda = 4$ :  $\mathbf{v}_2 = \begin{bmatrix} 5 \\ 1 \\ 1 \end{bmatrix}$ .

There are not three linearly independent eigenvectors so  $A$  can not be diagonalized.

**Th** If  $\lambda_1, \dots, \lambda_n$  are distinct eigenvalues of an  $n \times n$  matrix  $A$  with corresponding eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$ , then  $\mathbf{v}_1, \dots, \mathbf{v}_n$  are linearly independent.

**Th** If  $A$  is symmetric matrix  $A^T = A$  then  $A$  has  $n$  linearly independent Eigenvectors.