Linear Algebra [1]

## 5.3 Similarity and Diagonalization

Diagonalization Revisited

Thm. [A]  $A: n \times n$  matrix.

A is diagonalizable if and only if it has eigenvectors  $X_1, X_2, \dots, X_n$  s.t.  $P = \begin{bmatrix} X_1 & X_2 & \cdots & X_n \end{bmatrix}$  is invertible. In this case,  $P^{-1}AP = D = \operatorname{diag}(\lambda_1, \lambda_2, \cdots, \lambda_n)$ , where  $\lambda_i$  is the eigenvalue of A corresponding to  $X_i$ .

Thm. [A']  $A: n \times n$  matrix.

A is diagonalizable if and only if  $\mathbf{F}^n$  has a basis  $\{X_1, X_2, \cdots, X_n\}$  of eigenvectors of A.

**Thm.** [B] Let  $X_1, X_2, \dots, X_k$  be eigenvectors corresponding to distinct eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_k$  of A. Then  $\{X_1, X_2, \dots, X_k\}$  is linearly independent.

**Proof.** Assume that  $\{X_1, X_2, \cdots, X_k\}$  is linearly dependent. We can find j s.t.  $\{X_1, X_2, \cdots, X_{j-1}\}$  is linearly independent, and  $\{X_1, X_2, \cdots, X_j\}$  is linearly dependent. Then we have

$$(*) a_1 X_1 + a_2 X_2 + \dots + a_j X_j = O,$$

where not all  $a_i$ 's are zero, and in particular  $a_j \neq 0$ . Multiplying (\*) by A from the left, we get

$$a_1\lambda_1 X_1 + a_2\lambda_2 X_2 + \dots + a_j\lambda_j X_j = O.$$

On the other hand, multiplying (\*) by  $\lambda_j$ , we obtain

$$a_1\lambda_j X_1 + a_2\lambda_j X_2 + \dots + a_j\lambda_j X_j = O.$$

Subtracting two equations, we have

$$a_1(\lambda_1 - \lambda_j)X_1 + a_2(\lambda_2 - \lambda_j)X_2 + \dots + a_{j-1}(\lambda_{j-1} - \lambda_j)X_{j-1} = O,$$

and  $a_1(\lambda_1 - \lambda_j) = a_2(\lambda_2 - \lambda_j) = \cdots = a_{j-1}(\lambda_{j-1} - \lambda_j) = 0$ . Since  $\lambda_i$ 's are distinct, we have

$$a_1 = a_2 = \dots = a_{j-1} = 0,$$

and  $a_j X_j = 0$ ,  $a_j = 0$ , a contradiction.

Therefore,  $\{X_1, X_2, \cdots, X_k\}$  is linearly independent.  $\square$ 

Linear Algebra [4]

**Cor.** [B'] If A is an  $n \times n$  matrix with n distinct eigenvalues, then A is diagonalizable.

**Fact.** If one chooses linearly independent sets of eigenvectors corresponding to distinct eigenvalues, and combines them into a single set, then that combined set will be linearly independent.

**Def.** An eigenvalue  $\lambda$  of A is said to have multiplicity m if it occurs m times as a root of  $c_A(x)$ .

**Def.** The set

$$E_{\lambda}(A) = \{X \in \mathbf{F}^n | AX = \lambda X\}$$

of  $\lambda$ -eigenvectors is a subspace of  $\mathbf{F}^n$  called the eigenspace of A corresponding to  $\lambda$ .

Note that an eigenspace  $E_{\lambda}(A)$  is merely the null space of  $\lambda I - A$ .

**Thm.** [C]  $A: n \times n$  matrix.

A is diagonalizable if and only if  $\dim E_{\lambda}(A)$  is equal to the multiplicity of  $\lambda$  for every eigenvalue  $\lambda$  of A.

**Proof.**  $(\Rightarrow)$  We omit it.

( $\Leftarrow$ ) Let  $\lambda_1, \lambda_2, \cdots, \lambda_k$  be distinct eigenvalues. Assume that  $\dim E_{\lambda_i}(A)$  is equal to the multiplicity of  $\lambda_i$  for each  $i=1,2,\cdots,k$ . Choose a basis  $B_i$  of  $E_{\lambda_i}(A)$  for each  $\lambda_i$ . Let  $B=B_1\cup B_2\cup\cdots\cup B_k$ . Then |B|=n and B is linearly independent from **Fact**. Thus B is a basis of  $\mathbf{F}^n$ , and A is diagonalizable by **Thm A'**.  $\square$ 

Thm. [C']  $A: n \times n$  matrix.

A is diagonalizable if and only if every eigenvalue  $\lambda$  of multiplicity m yields m basic solutions of the equation

$$(\lambda I - A)X = O.$$

**Fact.** Let  $\lambda$  be an eigenvalue of multiplicity of m of A. Then

$$\dim E_{\lambda}(A) \leq m$$
.

• Diagonalization Algorithm

Let A be an  $n \times n$  matrix.

- 1. Find all the eigenvalues  $\lambda$  of A.
- 2. For each  $\lambda$ , compute the basic solutions of  $(\lambda I A)X = O$ . If there are n basic solutions in total, A is diagonalizable.
- 3. Construct the matrix P whose columns are (scalar multiples of) basic solutions.
- 4.  $P^{-1}AP$  is diagonal. (P is invertible.)

**Eg.** 
$$A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}$$
,  $c_A(x) = x(x-1)^2$ . For  $\lambda = 1$ ,

$$\lambda I - A = \begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & -2 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}.$$

A is not diagonalizable.

$$B = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{bmatrix}$$
 is diagonalizable.

Linear Algebra [10]

## Similar Matrices

**Def.**  $A, B: n \times n$  matrices

We say that A and B are similar if  $B = P^{-1}AP$  for some invertible P. We will write  $A \sim B$  for similar matrices A and B.

**Eg.** 
$$\begin{bmatrix} 1 & -2 \\ -3 & 2 \end{bmatrix}$$
 and  $\begin{bmatrix} 4 & 0 \\ 0 & -1 \end{bmatrix}$  are similar.

Indeed, for  $P = \begin{bmatrix} \frac{2}{3} & 1 \\ -1 & 1 \end{bmatrix}$ , we have

$$P^{-1} \begin{bmatrix} 1 & -2 \\ -3 & 2 \end{bmatrix} P = \begin{bmatrix} 4 & 0 \\ 0 & -1 \end{bmatrix}.$$

## Observations:

1. A is diagonalizable if and only if A is similar to a diagonal matrix.

- 2. Assume that A and B are similar. Then  $A^{-1} \sim B^{-1}$ ,  $A^T \sim B^T$ ,  $A^k \sim B^k$ . If one of A and B is diagonalizable, then the other is also diagonalizable.
- 3. If A is diagonalizable, then  $A^{-1}$ ,  $A^T$  and  $A^k$  are also diagonalizable.

**Def.** Let  $A = [a_{ij}]$ . The trace of an  $n \times n$  matrix A is defined by

$$tr A = \sum_{i=1}^{n} a_{ii} = a_{11} + a_{22} + \dots + a_{nn}.$$

Linear Algebra [12]

## Prop.

1.  $\operatorname{tr}(A+B) = \operatorname{tr}A + \operatorname{tr}B$ ,

- 2.  $\operatorname{tr}(kA) = k \operatorname{tr} A$ ,
- 3.  $\operatorname{tr}(A^T) = \operatorname{tr} A$ ,
- 4.  $\operatorname{tr}(AB) = \operatorname{tr}(BA)$ .

**Proof.**  $A = [a_{ij}], B = [b_{ij}], AB = [c_{ij}], and BA = [d_{ij}].$ 

$$\operatorname{tr}(AB) = \sum_{i=1}^{n} c_{ii} = \sum_{i=1}^{n} \sum_{k=1}^{n} a_{ik} b_{ki}$$
$$= \sum_{k=1}^{n} \sum_{i=1}^{n} b_{ki} a_{ik} = \sum_{k=1}^{n} d_{kk} = \operatorname{tr}(BA).$$

Linear Algebra [13]

**Thm.** If  $A \sim B$ , then A and B have the same determinant, rank, trace, characteristic polynomial, and eigenvalues.

**Proof.** Let  $B = P^{-1}AP$  for some invertible P.

$$\det B = \det(P^{-1}AP) = \det P^{-1} \det A \det P = \det A.$$

$$trB = tr(P^{-1}AP) = tr((AP)P^{-1}) = trA.$$

$$c_B(x) = \det(xI - B) = \det(P^{-1}xIP - P^{-1}AP)$$
  
=  $\det[P^{-1}(xI - A)P] = \det(xI - A) = c_A(x).$ 

$$rankB = rank(P^{-1}AP) = rank(AP) = rankA.$$