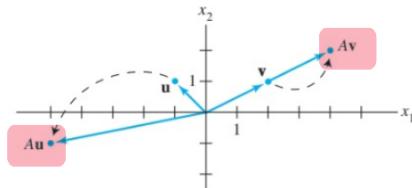


**EXAMPLE 1** Let  $A = \begin{bmatrix} 3 & -2 \\ 1 & 0 \end{bmatrix}$ ,  $\mathbf{u} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ . The images of  $\mathbf{u}$  and  $\mathbf{v}$  under multiplication by  $A$  are shown in Figure 1. In fact,  $A\mathbf{v}$  is just  $2\mathbf{v}$ . So  $A$  only "stretches," or dilates,  $\mathbf{v}$ . ■

FIGURE 1 Effects of multiplication by  $A$ .

→  $\mathbf{u}$ 와  $\mathbf{v}$ 의  $A$ 로 곱할 때  $A\mathbf{v}$ 는  $\mathbf{v}$ 의 동일한 길이의 2배.

### Eigenvalue, Eigenvector 정의.

#### DEFINITION

An **eigenvector** of an  $n \times n$  matrix  $A$  is a nonzero vector  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$  for some scalar  $\lambda$ . A scalar  $\lambda$  is called an **eigenvalue** of  $A$  if there is a nontrivial solution  $\mathbf{x}$  of  $A\mathbf{x} = \lambda\mathbf{x}$ ; such an  $\mathbf{x}$  is called an *eigenvector corresponding to  $\lambda$* !

$A\mathbf{x} = \lambda\mathbf{x}$  만족하는 nonzero vector  $\mathbf{x}$ 가 eigenvector,

$A\mathbf{x} = \lambda\mathbf{x}$ 인 nontrivial solution의 대응하는 scalar  $\lambda$ 가 eigenvalue,

이때  $\mathbf{x}$ 를  $\lambda$ 에 대응하는 eigenvector라고 함.

**EXAMPLE 2** Let  $A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix}$ ,  $\mathbf{u} = \begin{bmatrix} 6 \\ -5 \end{bmatrix}$ , and  $\mathbf{v} = \begin{bmatrix} 3 \\ -2 \end{bmatrix}$ . Are  $\mathbf{u}$  and  $\mathbf{v}$  eigenvectors of  $A$ ?

#### SOLUTION

$$A\mathbf{u} = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} 6 \\ -5 \end{bmatrix} = \begin{bmatrix} -24 \\ 20 \end{bmatrix} = -4 \begin{bmatrix} 6 \\ -5 \end{bmatrix} = -4\mathbf{u}$$

$$A\mathbf{v} = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} \begin{bmatrix} 3 \\ -2 \end{bmatrix} = \begin{bmatrix} -9 \\ 11 \end{bmatrix} \neq \lambda \begin{bmatrix} 3 \\ -2 \end{bmatrix}$$

$\mathbf{u}$ 는  $A$ 의 eigenvector.

$-4$ 는  $A$ 의 eigenvalue.

Thus  $\mathbf{u}$  is an eigenvector corresponding to an eigenvalue  $(-4)$ , but  $\mathbf{v}$  is not an eigenvector of  $A$ , because  $A\mathbf{v}$  is not a multiple of  $\mathbf{v}$ . ■

**EXAMPLE 3** Show that 7 is an eigenvalue of matrix  $A$  in Example 2, and find the corresponding eigenvectors.

$$A = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix},$$

→ 7은 Eigenvalue?

⇒  $Ax = 7x$  를 만족해야 한다.

$$Ax - 7x = 0,$$

$$(A - 7I)x = 0 \text{ 으로 정리한 다음에},$$

Eigenvalue 대응되는  $x$ 는 nonzero vector.

nontivial solution이 존재하지 않으면

(Free variable이 존재하거나)

$$A - 7I = \begin{bmatrix} 1 & 6 \\ 5 & 2 \end{bmatrix} - \begin{bmatrix} 7 & 0 \\ 0 & 7 \end{bmatrix} = \begin{bmatrix} -6 & 6 \\ 5 & -5 \end{bmatrix}$$

$$\begin{bmatrix} -6 & 6 & 0 \\ 5 & -5 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \Rightarrow x_1 = x_2.$$

$$\Rightarrow x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix} \text{ (general solution)}$$

∴ nontivial solution 존재 → 7은 A의 eigenvalue.

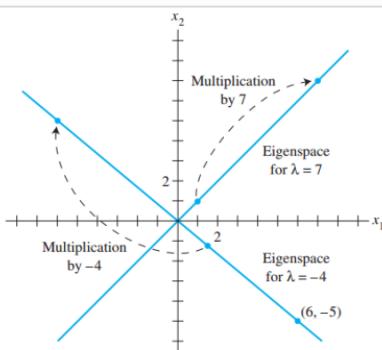


FIGURE 2 Eigenspaces for  $\lambda = -4$  and  $\lambda = 7$ .

λ가 A의 Eigenvalue라면  $(A - \lambda I)x = 0$ 의

nontivial solution을 찾는데,

$x$ 는 nonzero vector 이므로  $A - \lambda I = 0$ 의 solution을

찾으면되며,  $A - \lambda I$  matrix의 null space가

A의 eigenspace가 된다.

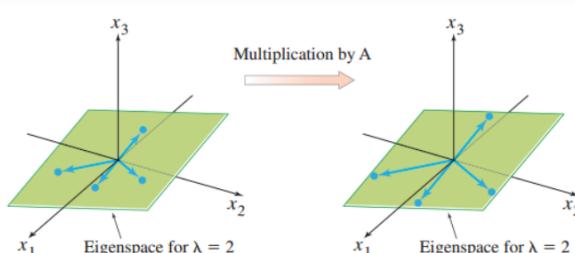


FIGURE 3  $A$  acts as a dilation on the eigenspace.

Eigen space 위에 벡터들을 A로

linear transform하는 경우를 생각해보자

양수와 음수로만 확장되므로 eigenspace는

유지됨.

**THEOREM 1**

The eigenvalues of a triangular matrix are the entries on its main diagonal.

triangular matrix의 eigenvalues는 대각선.

- eigenvectors는 nonzero vector이며 다음과 같다.

Eigenvalue는 0인지를 묻는다.

$$Ax = \lambda x = 0x = 0.$$

$\Rightarrow Ax = 0$  이고  $x$ 는 0이 아닌 nontrivial solution을 가진다.

eigenvalue가 0인가 가능. 'A is not invertible..'

**THEOREM 2**

If  $\mathbf{v}_1, \dots, \mathbf{v}_r$  are eigenvectors that correspond to distinct eigenvalues  $\lambda_1, \dots, \lambda_r$  of an  $n \times n$  matrix  $A$ , then the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly independent.

(기저로 (정렬되지 않는) Eigenvalue에 해당하는 eigenvector set은

linearly independent이다

**5.1.****EXERCISES.**

1. Is  $\lambda = 2$  an eigenvalue of  $\begin{bmatrix} 3 & 2 \\ 3 & 8 \end{bmatrix}$ ? Why or why not?

Sol).  $Ax = \lambda x = 2x$ .

$$Ax = 2x.$$

$$\underline{(A - 2I)x = 0}.$$

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

$$\Rightarrow \begin{bmatrix} 1 & 2 & 0 \\ 3 & 6 & 0 \end{bmatrix} \text{ row } 1 \times (-3) + \text{row } 2.$$

$$\begin{bmatrix} 1 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$x_2$  is free variable.

$\therefore \lambda = 2$  is eigenvalue.

## 5.2. THE CHARACTERISTIC EQUATION.

A scalar  $\lambda$  is an eigenvalue of an  $n \times n$  matrix  $A$  if and only if  $\lambda$  satisfies the characteristic equation

$$\det(A - \lambda I) = 0$$

가지 characteristic equation  $\det(A - \lambda I) = 0$  을 만족하면  
가능 행렬  $A$ 의 eigenvalue 이다.

- Similarity  $\Rightarrow$  어떤 invertible matrix P가 있을 때

$B = PAP^{-1}$  이 성립하면  $A$ 와  $B$ 는 similar하다는 것.

equivalently,  $A = PBP^{-1}$ .

A와 B가 similar 때,

$A$ 를  $PAP^{-1}$ 로 변환하는 것을 similarity transform이라 한다.

### THEOREM 4

If  $n \times n$  matrices  $A$  and  $B$  are similar, then they have the same characteristic polynomial and hence the same eigenvalues (with the same multiplicities).

$A$ 와  $B$ 가 similar 때, 같은 eigenvalue를 가진다.

#### WARNINGS:

1. The matrices

$$\begin{bmatrix} 2 & 1 \\ 0 & 2 \end{bmatrix} \text{ and } \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$$

are not similar even though they have the same eigenvalues.

2. Similarity is not the same as row equivalence. (If  $A$  is row equivalent to  $B$ , then  $B = EA$  for some invertible matrix  $E$ .) Row operations on a matrix usually change its eigenvalues.

A와 B가 similar 때, eigenvalue가 같지만,  
eigenvalue가 같다고 해서 similar한 것은 아니다.

## 5.2. EXERCISES

In Exercises 21 and 22,  $A$  and  $B$  are  $n \times n$  matrices. Mark each statement True or False. Justify each answer.

21. a. The determinant of  $A$  is the product of the diagonal entries in  $A$ .
- b. An elementary row operation on  $A$  does not change the determinant.
- c.  $(\det A)(\det B) = \det AB$
- d. If  $\lambda + 5$  is a factor of the characteristic polynomial of  $A$ , then 5 is an eigenvalue of  $A$ .

Sol).

a.  $\det A$  는  $A$ 의 대각원소들의 곱인가?

$\Rightarrow$  False.

triangular matrix의 경우에만 대각원소들의 곱으로  $\det$  구함.

b. row operation은 determinant 값을 변경시켜야 한다?

$\Rightarrow$  False.

row operation  
 [ replacement  $\rightarrow$  same det  
 [ Interchange  
 [ Scaling  $\rightarrow$  change det. ] ] ]

c.  $(\det A)(\det B) = \det AB$  ?

$\Rightarrow$  True.

(Theorem 3).

d.  $A$ 의 characteristic polynomial이  $(\lambda+5)^2$ 를 성립할 때,

$A$ 의 eigenvalue는  $5$ ?

$\Rightarrow$  False.

$(\lambda - (-5))$  칭하지 않음

eigenvalue =  $-5$ ,

**EXAMPLE 3** Find the characteristic equation of

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

**SOLUTION** Form  $A - \lambda I$ , and use Theorem 3(d):

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{bmatrix} 5 - \lambda & -2 & 6 & -1 \\ 0 & 3 - \lambda & -8 & 0 \\ 0 & 0 & 5 - \lambda & 4 \\ 0 & 0 & 0 & 1 - \lambda \end{bmatrix} \\ &= (5 - \lambda)(3 - \lambda)(5 - \lambda)(1 - \lambda) \end{aligned}$$

The characteristic equation is

$$(5 - \lambda)^2(3 - \lambda)(1 - \lambda) = 0$$

or

$$(\lambda - 5)^2(\lambda - 3)(\lambda - 1) = 0$$

Expanding the product, we can also write

$$\lambda^4 - 14\lambda^3 + 68\lambda^2 - 130\lambda + 75 = 0$$

In Examples 1 and 3,  $\det(A - \lambda I)$  is a polynomial in  $\lambda$ . It can be shown that if  $A$  is an  $n \times n$  matrix, then  $\det(A - \lambda I)$  is a polynomial of degree  $n$  called the **characteristic polynomial** of  $A$ .

## 5.3. DIAGONALIZATION.

[Ex.1].

$$D = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix}$$

diagonal matrix.

$$D^2 = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 5^2 & 0 \\ 0 & 3^2 \end{bmatrix}$$

$$D^3 = DD^2 = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} 5^2 & 0 \\ 0 & 3^2 \end{bmatrix} = \begin{bmatrix} 5^3 & 0 \\ 0 & 3^3 \end{bmatrix}$$

diagonal matrix의 제곱

$$D^k = \begin{bmatrix} 5^k & 0 \\ 0 & 3^k \end{bmatrix} \quad \text{for } k \geq 1$$

= 대각원소들의 제곱.

따라서,  $A = PDP^{-1}$  일 때 diagonal matrix라면

(A가 diagonal matrix와 similar하다면)  $A^k$ 도 쉽게 구할 수 있다.

[Ex.2]

$$A = \begin{bmatrix} 7 & 2 \\ -4 & 1 \end{bmatrix}$$

$$A = PDP^{-1}$$

$$P = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \quad \text{and} \quad D = \begin{bmatrix} 5 & 0 \\ 0 & 3 \end{bmatrix}$$

$$P^{-1} = \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

$$A^2 = (PDP^{-1})(PDP^{-1}) = PD(P^{-1}P)DP^{-1} = PDDP^{-1}$$

$$= PD^2P^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} 5^2 & 0 \\ 0 & 3^2 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix}$$

$$A^3 = (PDP^{-1})A^2 = (PDP^{-1})PD^2P^{-1} = PDD^2P^{-1} = PD^3P^{-1}$$

$$\begin{aligned} A^k &= PD^kP^{-1} = \begin{bmatrix} 1 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} 5^k & 0 \\ 0 & 3^k \end{bmatrix} \begin{bmatrix} 2 & 1 \\ -1 & -1 \end{bmatrix} \\ &= \begin{bmatrix} 2 \cdot 5^k - 3^k & 5^k - 3^k \\ 2 \cdot 3^k - 2 \cdot 5^k & 2 \cdot 5^k - 3^k \end{bmatrix} \end{aligned}$$

$\Rightarrow$  Square matrix A가 diagonal matrix와 similar하면  
A는 Diagonalizable이다.

(Eigenvalue decomposition이라고도 하겠다).

## THEOREM 5

### The Diagonalization Theorem

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

In fact,  $A = PDP^{-1}$ , with  $D$  a diagonal matrix, if and only if the columns of  $P$  are  $n$  linearly independent eigenvectors of  $A$ . In this case, the diagonal entries of  $D$  are eigenvalues of  $A$  that correspond, respectively, to the eigenvectors in  $P$ .

A는 대각화 가능한데

A는 n개의 linearly independent eigenvectors를 갖게 되어야 함.

D는 대각 행렬이고  $A = PDP^{-1}$  일 때, P의 Column은

A의 n개의 linearly independent eigenvector를 갖게 되어야 함.

이때 D의 대각 entries는 P의 Column의 eigenvectors인

행을, A의 eigen value.

•  $A = PDP^{-1}$ ,

D는 대각 행렬, 대각 entries = Eigenvalue of A 일 때

① A와 D는 유사  $\Rightarrow$  A의 eigenvalue = D의 eigenvalue.

A의 determinant = D의 determinant.

② D는 대각 행렬  $\Rightarrow$  대각 entries = Eigenvalue.

대각 entries  $\prod$  = determinant.

A의 eigenvalue = D의 eigenvalue = D의 대각 entries.

$$\Rightarrow \det(A) = \det(D) = D의 대각 entries의 곱.$$

$$= D의 eigenvalue들의 곱$$

$$= A의 eigenvalue들의 곱.$$

- Diagonalization 하는 방법.

- ① Eigenvalues를 찾는다.
- ② Eigenvectors를 찾는다.
- ③ Column으로 eigenvectors로 구성된 행렬  $P$ 를 만든다.
- ④ 그래서 구한 eigenvalues로 diagonal entries eigenvalues인 diagonal matrix  $D$ 를 만든다.

### THEOREM 6

An  $n \times n$  matrix with  $n$  distinct eigenvalues is diagonalizable.

=  $L \cdot I$  eigenvector

n개의 eigenvalue가 distinct 하면 diagonalizable 한다  
 diagonalizable 한 모드행렬이 n개의 다른 eigenvalues를  
 가지는 것은 아니다.

$p \leq n$

### THEOREM 7

Let  $A$  be an  $n \times n$  matrix whose distinct eigenvalues are  $\lambda_1, \dots, \lambda_p$ .

- For  $1 \leq k \leq p$ , the dimension of the eigenspace for  $\lambda_k$  is less than or equal to the multiplicity of the eigenvalue  $\lambda_k$ .
- The matrix  $A$  is diagonalizable if and only if the sum of the dimensions of the eigenspaces equals  $n$ , and this happens if and only if (i) the characteristic polynomial factors completely into linear factors and (ii) the dimension of the eigenspace for each  $\lambda_k$  equals the multiplicity of  $\lambda_k$ .
- 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 the sets  $B_1, \dots, B_p$  forms an eigenvector basis for  $\mathbb{R}^n$ .



$\dim \text{Nul}(A - \lambda_k I) \leq \lambda_k$ 의 증복회수.  
 = eigenspace for  $\lambda_k$ .

$A$  is diagonalizable

$$\equiv \sum_k \dim \text{Nul}(A - \lambda_k I) = n$$

$$\equiv \dim \text{Nul}(A - \lambda_k I) = \lambda_k$$
 증복회수

$\equiv$  Eigenvectors로  $\mathbb{R}^n$ 의 basis 형성.

## 5.3. EXERCISES.

In Exercises 1 and 2, let  $A = PDP^{-1}$  and compute  $A^4$ .

$$1. P = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix}, D = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\text{Sol). } A = PDP^{-1}$$

$$\Rightarrow P = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$P^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

$$\therefore P^{-1} = \frac{1}{(5 \times 3) - (11 \times 2)} \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix}$$

$$= \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix}$$

$$A^4 = P D^4 P^{-1} \text{ 을 활용함으로써}$$

$$A^4 = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 2^4 & 0 \\ 0 & 1^4 \end{bmatrix} \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix} = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 16 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 5 \times 16 + 7 \times 0 & 5 \times 0 + 7 \times 1 \\ 2 \times 16 + 3 \times 0 & 2 \times 0 + 3 \times 1 \end{bmatrix} \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix} = \begin{bmatrix} 80 & 7 \\ 32 & 3 \end{bmatrix} \begin{bmatrix} 3 & -11 \\ -2 & 5 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 80 \times 3 + 7 \times (-2) & 80 \times (-11) + 7 \times 5 \\ 32 \times 3 + 3 \times (-2) & 32 \times (-11) + 3 \times 5 \end{bmatrix} = \begin{bmatrix} 226 & -525 \\ 90 & -209 \end{bmatrix}$$