

Hence  $c_i \|\underline{v}_i\|^2 = 0 \Rightarrow c_i = 0$  because  $\|\underline{v}_i\|^2 > 0$  (set of vectors are non-zero,  $\Rightarrow \Leftarrow$ ). Thus, the assumption leads to a contradiction.  
 Therefore: there exists no set of non-zero orthogonal vectors which are linearly dependent. ■

## Orthogonal Complements

Def'n The vector  $\underline{u}$  is orthogonal to a subspace  $W$  of vector space  $V$ , provided that,  $\langle \underline{u}, \underline{w} \rangle = 0$  for  $\forall \underline{w} \in W$ .

Def'n The orthogonal complement  $W^\perp$  of  $W$  is the set of all vectors in  $V$  that are orthogonal to  $W$ .  
 (Note that here  $W$  does not have to be a subspace!)

ex/(i)  $W$  is a subspace of  $\mathbb{R}^3$ :  $W = \left\{ \underline{x} : \underline{x} = \alpha_1 \underline{u}_1 + \alpha_2 \underline{u}_2, \alpha_1 \in \mathbb{R}, \alpha_2 \in \mathbb{R} \right\}$   
 where  $\underline{u}_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ ,  $\underline{u}_2 = \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$ .

How can you find an orthogonal vector to  $W$ ?

Thm Let  $W$  be a subspace of  $\mathbb{R}^n$ .

- (i)  $W^\perp$  is also a subspace
- (ii)  $W \cap W^\perp = \{ \underline{0} \}$
- (iii)  $(W^\perp)^\perp = W$
- (iv)  $S$  is a spanning set for  $W$ , then:  $\underline{u} \in W^\perp \Leftrightarrow \underline{u}$  is orthogonal to  $S$ .

Thm  $A_{m \times n}$ :  $(\text{Row}(\underline{A}))^\perp = \text{Null}(\underline{A})$ .

Proof: If  $\underline{x} \in \text{Null}(\underline{A}) \Rightarrow \underline{A}\underline{x} = \underline{0} \Rightarrow \begin{bmatrix} \underline{r}_1^T \\ \vdots \\ \underline{r}_m^T \end{bmatrix} \underline{x} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$   
 $\Rightarrow \underline{x} \perp \text{span}\{\underline{r}_1, \dots, \underline{r}_m\} \Rightarrow \underline{x} \in (\text{Row}(\underline{A}))^\perp$   
 $\Rightarrow \text{Null}(\underline{A}) \subset (\text{Row}(\underline{A}))^\perp$

$$\begin{aligned}
 \text{If } \underline{x} \in (\text{Row}(\underline{A}))^\perp &\Rightarrow \underline{x} \perp \underline{r}_i, \quad 1 \leq i \leq m \\
 &\Rightarrow \underline{A}\underline{x} = \underline{0} \\
 &\Rightarrow \underline{x} \in \text{Null}(\underline{A}) \\
 &\Rightarrow (\text{Row}(\underline{A}))^\perp \subset \text{Null}(\underline{A})
 \end{aligned}$$

$$\Rightarrow (\text{Row}(\underline{A}))^\perp = \text{Null}(\underline{A})$$

## Chapter 6, Eigenvalues and Eigenvectors

Def'n  $\underline{A}\underline{v} = \lambda\underline{v}$  ← eigenvector,  $\underline{v} \neq \underline{0}$ .

↑  
square matrix      ↑  
eigenvalue

ex /  $\underline{A} = \begin{bmatrix} 5 & -6 \\ 2 & -2 \end{bmatrix}$ ,  $\underline{A} \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 4 \\ 2 \end{bmatrix} = 2 \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  ← eigenvalue.

Aslo:  $\underline{A} \cdot \begin{bmatrix} 3 \\ 2 \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix} = 1 \cdot \begin{bmatrix} 3 \\ 2 \end{bmatrix}$  ← eigenvalue.

## The characteristic Equation

$$\underline{A}\underline{v} = \lambda\underline{v} = \lambda\underline{I}\underline{v} \Rightarrow (\underline{A} - \lambda\underline{I})\underline{v} = \underline{0}$$

A non zero solution exists for  $(\underline{A} - \lambda\underline{I})\underline{v} = \underline{0} \Leftrightarrow \det(\underline{A} - \lambda\underline{I}) = 0$   
↳ Characteristic Equ.

ex /  $\underline{A} = \begin{bmatrix} 5 & 7 \\ -2 & -4 \end{bmatrix}$ ,  $\underline{A} - \lambda\underline{I} = \begin{bmatrix} 5-\lambda & 7 \\ -2 & -4-\lambda \end{bmatrix}$

$$|\underline{A} - \lambda\underline{I}| = (5-\lambda)(-4-\lambda) + 14 = \lambda^2 - \lambda - 6 = (\lambda-3)(\lambda+2)$$

$$\Rightarrow |\underline{A} - \lambda\underline{I}| = 0 \Rightarrow \lambda = 3 \text{ or } \lambda = -2.$$

Eigenvector associated with  $\lambda = -2$  :

$$\begin{bmatrix} 5+2 & 7 \\ -2 & -4+2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 7 & 7 \\ -2 & -2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = t \cdot \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad t \in \mathbb{R}. \text{ a non-zero eigenvector is } \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

eigenvector associated with  $\lambda = 3$  :

$$\begin{bmatrix} 5-3 & 7 \\ -2 & -4-3 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 2 & 7 \\ -2 & -7 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = t \cdot \begin{bmatrix} -7/2 \\ 1 \end{bmatrix}, \text{ can choose } \underline{v}_2 = \begin{bmatrix} -7 \\ 2 \end{bmatrix}.$$

Thm  $\lambda$  is an eigenvalue of  $\underline{A} \Leftrightarrow \det(\underline{A} - \lambda \underline{I}) = 0$ .

- Note that  $\det(\underline{A} - \lambda \underline{I})$  is a polynomial in  $\lambda$  of degree  $n$  for  $\underline{A}_{n \times n}$ .

Hence, we have  $n$  (possibly complex) roots for the characteristic eqn.

$$\text{ex/ } \underline{A} = \begin{bmatrix} 2 & 3 \\ 0 & 2 \end{bmatrix}, \det \begin{bmatrix} 2-\lambda & 3 \\ 0 & 2-\lambda \end{bmatrix} = (2-\lambda)^2 = 0 \Rightarrow \lambda = 2 \text{ double root.}$$

$$(\underline{A} - 2\underline{I}) \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 0 & 3 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} = t \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \text{ hence can choose only one eigenvector } \begin{bmatrix} 1 \\ 0 \end{bmatrix} \text{ which is linearly independent in this solution set.}$$

## Eigenspaces

Solution set of  $(\underline{A} - \lambda \underline{I}) \underline{v} = \underline{0}$  for  $\lambda$ , an eigenvalue, is called as the eigenspace of  $\underline{A}$  associated with eigenvalue  $\lambda$ .

## Diagonalization of Matrices

If  $\underline{A}_{n \times n}$  has  $n$  linearly independent eigenvectors, then:

$$\underline{A} \cdot \underbrace{\begin{bmatrix} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_n \end{bmatrix}}_{\underline{P}} = \underbrace{\begin{bmatrix} \underline{v}_1 & \underline{v}_2 & \dots & \underline{v}_n \end{bmatrix}}_{\underline{P}} \underbrace{\begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix}}_{\underline{D}}$$

$$\Rightarrow \underline{A} \underline{P} = \underline{P} \underline{D} \Rightarrow \underline{A} = \underline{P} \underline{D} \underline{P}^{-1} \text{ or } \underline{P}^{-1} \underline{A} \underline{P} = \underline{D}$$

( $\underline{P}^{-1}$  exists because its columns are linearly independent).

$$\text{ex/ } \underline{A} = \begin{bmatrix} 5 & -6 \\ 2 & -2 \end{bmatrix}, \lambda_1 = 2 \Rightarrow \underline{v}_1 = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\lambda_2 = 1 \Rightarrow \underline{v}_2 = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$

$\underline{v}_1$  &  $\underline{v}_2$  are linearly independent. Thus:

$$\underline{A} = \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix} \cdot \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 3 \\ 1 & 2 \end{bmatrix}^{-1}$$

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

Def'n  $\underline{A}_{n \times n}$  and  $\underline{B}_{n \times n}$  are called as similar if:

$$\underline{B} = \underline{P}^{-1} \underline{A} \underline{P},$$

## Criterion for diagonalizability:

$\underline{A}_{n \times n}$  is diagonalizable  $\Leftrightarrow$  it has  $n$  linearly independent eigenvectors.

### Theorem

Let  $\underline{v}_1, \dots, \underline{v}_k$  be eigenvectors of  $\underline{A}$  associated with distinct eigenvalues  $\lambda_1, \dots, \lambda_k$ . Then,  $\underline{v}_1, \dots, \underline{v}_k$  are linearly independent.

### proof by induction

$k=1$ :  $\underline{v}_1$  forms a linearly independent set because  $\underline{v}_1 \neq \underline{0}$ .

Assume that any set of  $(k-1)$  eigenvectors are linearly independent if they are associated with distinct eigenvalues. Now, consider following linear combination.

$$c_1 \underline{v}_1 + c_2 \underline{v}_2 + \dots + c_n \underline{v}_n = \underline{0} \quad (\star)$$

Need to show that  $c_1 = c_2 = \dots = c_n = 0$ .

Multiply both sides by  $(\underline{A} - \lambda_1 \underline{I})$  to get:

$$(\underline{A} - \lambda_1 \underline{I})(c_1 \underline{v}_1 + \dots + c_n \underline{v}_n) = \underline{0}$$

$$\text{Note that } (\underline{A} - \lambda_1 \underline{I}) \cdot \underline{v}_j = \begin{cases} \underline{0} & \text{if } j=1 \\ (\lambda_j - \lambda_1) \underline{v}_j & \text{if } j \neq 1. \end{cases}$$

Hence, we get:

$$c_2 (\lambda_2 - \lambda_1) \underline{v}_2 + \dots + c_n (\lambda_n - \lambda_1) \underline{v}_n = \underline{0}$$

Since these  $n$  vectors are linearly independent:

$$c_j (\lambda_j - \lambda_1) = 0 \quad \text{for } j > 1.$$

But  $\lambda_j \neq \lambda_1 \Rightarrow c_j = 0$  for  $j > 1$ .

Thus, in  $\star$ , the only non-zero coefficient can be  $c_1$ .

But this is a contradiction because:

$$c_1 \underline{v}_1 = \underline{0} \Rightarrow c_1 = 0.$$

Hence, the only combination in  $\star$  that results  $\underline{0}$  is  $c_1 = \dots = c_n = 0$ . Thus, the eigenvectors associated with distinct eigenvalues are lin. independent.

Thm  $\underline{A}_{n \times n}$  has distinct eigenvalues  $\Rightarrow \underline{A}$  is diagonalizable.

Have seen that if the eigenvectors of  $\underline{A}$  are linearly independent then  $\underline{A}$  is diagonalizable. If  $\underline{A}$  has distinct eigenvalues, then it has a linearly independent set of eigenvectors. Thus, it is diagonalizable.

Thm Let  $\lambda_1, \lambda_2, \dots, \lambda_k$  be the distinct eigenvalues of  $\underline{A}_{n \times n}$ . Let  $S_i$  be a basis for the eigenspace associated with  $\lambda_i$ . Then  $S = S_1 \cup S_2 \cup \dots \cup S_k$  is a linearly independent set of eigenvectors of  $\underline{A}$ .

Proof To keep the notation simple, assume  $k=3$ .

$$S_1 = \{ \underline{u}_1, \dots, \underline{u}_p \}$$

$$S_2 = \{ \underline{v}_1, \dots, \underline{v}_q \}$$

$$S_3 = \{ \underline{w}_1, \dots, \underline{w}_r \}$$

Assume that the union set  $S = S_1 \cup S_2 \cup S_3$  is dependent. Then, there exists nonzero coefficients such that:

$$\underbrace{a_1 \underline{u}_1 + a_2 \underline{u}_2 + \dots + a_p \underline{u}_p}_{\underline{u}} + \underbrace{b_1 \underline{v}_1 + b_2 \underline{v}_2 + \dots + b_q \underline{v}_q}_{\underline{v}} + \underbrace{c_1 \underline{w}_1 + c_2 \underline{w}_2 + \dots + c_r \underline{w}_r}_{\underline{w}} = \underline{0}$$

where  $\underline{u}$  is an eigenvector associated with  $\lambda_1$

"  $\underline{v}$  " " " "  $\lambda_2$

"  $\underline{w}$  " " " "  $\lambda_3$

Hence,  $\underline{u}$ ,  $\underline{v}$  and  $\underline{w}$  must be linearly independent if they are not all zeros. Thus:  $\underline{u} = \underline{0}$ ,  $\underline{v} = \underline{0}$  and  $\underline{w} = \underline{0}$ .

But  $\underline{u}$  can only be zero if  $a_1 = \dots = a_p = 0$  because  $\underline{u}_1, \dots, \underline{u}_p$  forms a basis (hence they are linearly independent). Likewise,  $b_1 = \dots = b_q = 0$  and  $c_1 = \dots = c_r = 0$ .

Thus, we end up with a contradiction. Hence, the theorem has the right statement.

## Matrix Powers

If  $\underline{A}_{n \times n}$  has linearly independent eigenvectors, then:

$$\underline{A} = \underline{P} \underline{D} \underline{P}^{-1}$$

$$\text{Thus: } \underline{A}^k = \underbrace{\underline{A} \dots \underline{A}}_{k \text{ times}} = (\underline{P} \underline{D} \underline{P}^{-1}) (\underline{P} \underline{D} \underline{P}^{-1}) \dots (\underline{P} \underline{D} \underline{P}^{-1}) \\ = \underline{P} \underline{D}^k \underline{P}^{-1}$$

$$\text{Since } \underline{D} = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix}$$

$$\underline{D}^k = \begin{bmatrix} \lambda_1^k & & \\ & \ddots & \\ & & \lambda_n^k \end{bmatrix}$$

## The Cayley-Hamilton Thm

If  $\underline{A}_{n \times n}$  has the characteristic polynomial:

$$p(\lambda) = (-1)^n \lambda^n + c_{n-1} \lambda^{n-1} + \dots + c_1 \lambda + c_0$$

$$\text{then: } p(\underline{A}) = (-1)^n \underline{A}^n + c_{n-1} \underline{A}^{n-1} + \dots + c_1 \underline{A} + c_0 \underline{I} = \underline{0}$$

proof The theorem is valid for any  $\underline{A}_{n \times n}$ . But we will provide here proof for  $\underline{A}$  which is diagonalizable:  $\underline{A} = \underline{P} \underline{D} \underline{P}^{-1}$ .

$$p(\underline{D}) = (-1)^n \underline{D}^n + c_{n-1} \underline{D}^{n-1} + \dots + c_1 \underline{D} + c_0 \underline{I}$$

$$= (-1)^n \begin{bmatrix} \lambda_1^n & & \\ & \ddots & \\ & & \lambda_n^n \end{bmatrix} + c_{n-1} \begin{bmatrix} \lambda_1^{n-1} & & \\ & \ddots & \\ & & \lambda_n^{n-1} \end{bmatrix} + \dots + c_1 \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} + \begin{bmatrix} c_0 & & \\ & \ddots & \\ & & c_0 \end{bmatrix}$$

$$= \begin{bmatrix} p(\lambda_1) & & \\ & p(\lambda_2) & \\ & & \ddots \\ & & & p(\lambda_n) \end{bmatrix} = \underline{0}$$

$$p(\underline{A}) = \underline{P} \cdot p(\underline{D}) \underline{P}^{-1} = \underline{P} \cdot \underline{0} \underline{P}^{-1} = \underline{0}. \quad \blacksquare$$

$$\text{ex/ } \underline{A} = \begin{bmatrix} 4 & -2 & 1 \\ 2 & 0 & 1 \\ 2 & -2 & 3 \end{bmatrix}$$

$$p(\lambda) = -\lambda^3 + 7\lambda^2 - 16\lambda + 12$$

$$\text{Thus: } -\underline{A}^3 + 7\underline{A}^2 - 16\underline{A} + 12\underline{I} = \underline{0}.$$

$$\Rightarrow \underline{A}^3 = 7\underline{A}^2 - 16\underline{A} + 12\underline{I}.$$

If we want to compute  $\underline{A}^4$ :

$$\begin{aligned} \underline{A}^4 &= \underline{A} \cdot \underline{A}^3 = 7\underline{A}^3 - 16\underline{A}^2 + 12\underline{A} \\ &= 7(7\underline{A}^2 - 16\underline{A} + 12\underline{I}) - 16\underline{A}^2 + 12\underline{A} \\ &= 33\underline{A}^2 - 100\underline{A} + 84\underline{I}. \end{aligned}$$

Or if we want to compute  $\underline{A}^{-1}$ :

$$\underline{A}^{-1}(-\underline{A}^3 + 7\underline{A}^2 - 16\underline{A} + 12\underline{I}) = \underline{0}$$

$$\Rightarrow -\underline{A}^2 + 7\underline{A} - 16\underline{I} + 12\underline{A}^{-1} = \underline{0}$$

$$\Rightarrow \underline{A}^{-1} = \frac{1}{12}(\underline{A}^2 - 7\underline{A} + 16\underline{I}) = \frac{1}{6} \begin{bmatrix} 1 & 2 & -1 \\ -2 & 5 & -1 \\ -2 & 2 & 2 \end{bmatrix}.$$