

# Session 8



6.4. Theorem.

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn} \end{bmatrix}$$

← triangle matrix

$$\lambda_1 = a_{11}, \lambda_2 = a_{22}, \dots, \lambda_n = a_{nn}$$

$$\left| [A - \lambda I | 0] \right|$$

$$\left| \begin{bmatrix} a_{11}-\lambda & a_{12} & \dots & 0 \\ 0 & a_{22}-\lambda & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_{nn}-\lambda \end{bmatrix} \right| \stackrel{\text{set}}{=} 0.$$

$$(a_{11}-\lambda)(a_{22}-\lambda) \cdots (a_{nn}-\lambda) \stackrel{\text{set}}{=} 0$$

ex. 6.5 Find the eigenvalues of

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

$$\lambda_1 = 2 \quad \lambda_2 = 1 \quad \lambda_3 = 3 \quad \lambda_4 = -2$$

6.6. Theorem  $\exists A_{n \times n}^{-1}$  iff.  $\lambda_i \neq 0$  ( $i \in \{1, \dots, n\}$ )

6.7. Theorem. Let  $A_{n \times n}$  have eigenpair  $(\lambda_i, \vec{v}_i)$   $i = 1, 2, \dots, n$

$\Rightarrow$  (a)  $\lambda^n$  is eigenvalue of  $A^n$  w/ the same  $\vec{x}$ .  $(\lambda^n, \vec{x})$  is a eigenpair of  $A^n$

(b)  $\frac{1}{\lambda}$  is eigenvalue of  $A^{-1}$  with the same  $\vec{x}$ .  $(\frac{1}{\lambda}, \vec{x})$  is a eigenpair of  $A^{-1}$

(c)  $\alpha \lambda$  is eigenvalues of  $\alpha A$

(d)  $g(\lambda)$  is eigenvalue of  $g(A)$ ,  $g$  is polynomial

Theorem 6.8. An  $n \times n$  has pairs  $(\lambda_i, \vec{v}_i)(\lambda_2, \vec{v}_2)$  ...  $(\lambda_n, \vec{v}_n)$ . If  $\vec{x} \in \mathbb{R}^n$ ,  $\exists$

$$\vec{x} = c_1 \vec{v}_1 + c_2 \vec{v}_2 + \cdots + c_n \vec{v}_n *$$

$$\Rightarrow \forall k \in \mathbb{N}$$

$$A^k \vec{x} = c_1 \lambda_1^k \vec{v}_1 + c_2 \lambda_2^k \vec{v}_2 + \cdots + c_n \lambda_n^k \vec{v}_n$$

6.7. ex Compute  $\begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}^{10} \begin{bmatrix} 5 \\ 1 \end{bmatrix}$

Step 1. Solve for  $\lambda$ 's.

$$|A - \lambda I| = 0$$

$$\begin{vmatrix} -\lambda & 1 \\ 2 & 1-\lambda \end{vmatrix} = 0$$

$$(-\lambda)(1-\lambda) - 2 = 0$$

$$\lambda^2 - \lambda - 2 = 0$$

$$(\lambda-2)(\lambda+1) = 0$$

$$\lambda_1 = -1, \lambda_2 = 2$$

Step 2. solve for eigenvectors w.r.t  $\lambda_1 = -1, \lambda_2 = 2$ .

(i) when  $\lambda_1 = -1$

solve for  $[A - \lambda I | 0]$

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

$$x_1 + x_2 = 0 \Rightarrow x_1 = -x_2 \quad \text{set } x_2 = t \in \mathbb{R}$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} t \\ -t \end{bmatrix} = t \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \vec{v}_1 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

Step 4  $\begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}^{10} \begin{bmatrix} 5 \\ 1 \end{bmatrix}$

$$= 3 \cdot (-1)^{10} \begin{bmatrix} 1 \\ -1 \end{bmatrix} + 2 \cdot 2^{10} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 3+2^{11} \\ -3+2^{12} \end{bmatrix}$$

Step 3. Solve for  $c_1, c_2$ .

$$c_1 \begin{bmatrix} 1 \\ -1 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 1 \end{bmatrix}$$

Solve for  $\begin{bmatrix} 1 & 1 & 5 \\ -1 & 2 & 1 \end{bmatrix}$

$$R_1 + R_2 \rightarrow \begin{bmatrix} 1 & 1 & 5 \\ 0 & 3 & 6 \end{bmatrix} \xrightarrow{R_2/3} \begin{bmatrix} 1 & 1 & 5 \\ 0 & 1 & 2 \end{bmatrix}$$

$$R_1 - R_2 \rightarrow \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & 2 \end{bmatrix} \quad c_1 \Rightarrow c_2 = 2$$

$$\vec{x} = 3 \vec{v}_1 + 2 \vec{v}_2$$

(ii)  $\lambda_2 = 2$

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

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

$$-2x_1 + x_2 = 0$$

$$x_1 = t \quad x_2 = 2x_1 = 2t$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} t \\ 2t \end{bmatrix} = t \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$\vec{v}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Algorithm for computing  $A^k \vec{x}$

Step 1. solve for  $\lambda$ 's.



$$\text{solve } |A - \lambda I| = 0$$

Step 2. solve for eigenvector  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$   
 $\downarrow \quad \downarrow \quad \downarrow$   
 $\lambda_1 \quad \lambda_2 \quad \lambda_n$

Step 3. solve for  $c_1, \dots, c_n$



$$\text{solve for } [\vec{v}_1 | \vec{v}_2 | \dots | \vec{v}_n | \vec{x}] \Leftrightarrow c_1 \vec{v}_1 + \dots + c_n \vec{v}_n = \vec{x}.$$

hint: RE form

Step 4. Compute  $A^k \vec{x} = c_1 \lambda_1^k \vec{v}_1 + c_2 \lambda_2^k \vec{v}_2 + \dots + c_n \lambda_n^k \vec{v}_n$

## 7). Determinants.

7.1. Def. For  $3 \times 3$   $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$

$$\det A = |A| = \boxed{a_{11}} a_{22} a_{23} - a_{12} a_{21} a_{33} + a_{13} a_{21} a_{32}$$

where  $\begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} a_{22} - a_{12} a_{21}$

$$|A| = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11} a_{22} - a_{12} a_{21}$$

7.2. ex.  $A = \begin{bmatrix} 5 & -3 & 2 \\ 1 & 0 & 2 \\ 2 & -1 & 3 \end{bmatrix}$

$$\det A = 5 \cdot \begin{vmatrix} 0 & 2 \\ -1 & 3 \end{vmatrix} - (-3) \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} + 2 \begin{vmatrix} 1 & 0 \\ 2 & -1 \end{vmatrix}$$

$$= 5 \cdot (0+2) + 3 \cdot (3-4) + 2 \cdot (-1)$$

$$= 10 - 3 - 2 = 5$$

7.3. Def.  $A_{ij} \stackrel{\text{def}}{=} \text{matrix } A \text{ w.o. } i\text{th row and } j\text{th column}$

$$\begin{aligned} |A| &\stackrel{\text{def}}{=} a_{11} |A_{11}| - a_{12} |A_{12}| + \cdots + (-1)^{1+n} a_{1n} |A_{1n}| \\ &= -a_{11} |A_{21}| + a_{22} |A_{22}| - \cdots - (-1)^{1+n} a_{2n} |A_{2n}| \\ &\quad + \quad - \quad + \cdots + \\ &\quad - \quad + \quad - \cdots - \\ &\left[ \begin{array}{cccc} + & - & + & - \cdots \\ - & + & - & + \cdots \\ + & - & + & - \cdots \\ \vdots & \vdots & \vdots & \vdots \end{array} \right] \end{aligned}$$

7.4. The Laplace Expansion Theorem.

$$\det A = a_{11} c_{11} + \cdots + a_{nn} c_{nn}$$

$$c_{ij} = (-1)^{i+j} |A_{ij}|$$

$$c_{21} = (-1)^{2+1} A_{21} = -A_{21}$$

7.5 ex. Compute  $|A|$ .  $A = \begin{bmatrix} 1 & -3 & 2 \\ 1 & 0 & 2 \\ 2 & -1 & 3 \end{bmatrix}$

$$\begin{bmatrix} + & (-) & + \\ - & (+) & - \\ + & (-) & + \end{bmatrix}$$

$$|A| = -(-2) \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} - (-1) \begin{vmatrix} 5 & 2 \\ 1 & 2 \end{vmatrix}$$

$$= 3(3-4) + (10-2) = -3 + 8 = 5.$$

7.6. Thm.  $A = \begin{bmatrix} a_{11} & & 0 \\ * & * & \\ * & * & * \\ \vdots & \vdots & a_{nn} \end{bmatrix} \quad A^* = \begin{bmatrix} a_{11} & & 1 \\ 0 & \ddots & \\ & * & a_{nn} \end{bmatrix}$

$$|A| = \prod_{i=1}^n a_{ii} = \prod_{i=1}^n \lambda_i$$

7.7. Theorem.  $A_{n \times n}$

a. If  $A$  has a zero <sup>column</sup> row  $\Rightarrow |A|=0$

b.  $A = \begin{bmatrix} \vec{a}_1^T \\ \vec{a}_2^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix} \quad A^* = \begin{bmatrix} \vec{a}_1^T \\ (\vec{a}_n^T) \\ \vdots \\ \vec{a}_1^T \end{bmatrix}$

$$|A| = -|A^*|$$

c. If  $A$  has two identical <sup>column</sup> rows  $\Rightarrow |A|=0$

d.  $B = \begin{bmatrix} \vec{b}_1^T \\ \vec{b}_2^T \\ \vdots \\ \vec{b}_n^T \end{bmatrix} \quad |B| = k|A| \quad (\forall k)$

$$A = \begin{bmatrix} \vec{a}_1^T \\ \vdots \\ \vec{a}_k^T \\ \vec{a}_{k+1}^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix}$$

e.  $A = \begin{bmatrix} \vec{a}_1^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix} \quad B = \begin{bmatrix} \vec{b}_1^T \\ \vdots \\ \vec{b}_n^T \end{bmatrix} \quad C = \begin{bmatrix} \vec{c}_1^T \\ \vdots \\ \vec{c}_n^T \\ \vec{c}_k^T + \vec{b}_k^T \end{bmatrix}$

$$\vec{a}_i^T = \vec{b}_i^T = \vec{c}_i^T \text{ except } \vec{c}_k^T = \vec{a}_k^T + \vec{b}_k^T$$

$$|C| = |A| + |B|$$

f.  $B = \begin{bmatrix} \vec{a}_1^T \\ \vdots \\ \vec{a}_k^T + k\vec{a}_j^T \\ \vdots \\ \vec{a}_n^T \end{bmatrix} \quad |A| = |B|$

Ex1  $\begin{vmatrix} 1 & 2 & 3 \\ 0 & 1 & 5 \\ 1 & 5 & 1 \end{vmatrix} + \begin{vmatrix} 1 & 2 & 3 \\ 1 & 0 & 1 \\ 1 & 5 & 1 \end{vmatrix} = \begin{vmatrix} 1 & 2 & 3 \\ 1 & 1 & 6 \\ 1 & 5 & 1 \end{vmatrix}$

Ex2  $\begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} + \begin{vmatrix} 1 & 0 \\ 1 & 0 \end{vmatrix} = \begin{vmatrix} 1 & 0 \\ 1 & 1 \end{vmatrix}$

7.8. Ex.

$$(i) A = \begin{bmatrix} 2 & 3 & -1 \\ 0 & 5 & 3 \\ -4 & -6 & 2 \end{bmatrix} \xrightarrow{R_3 + 2R_1} \begin{bmatrix} 2 & 3 & -1 \\ 0 & 5 & 3 \\ 0 & 0 & 0 \end{bmatrix}$$

$$|A| = 0$$

$$(ii) B = \begin{bmatrix} 0 & 2 & -4 & 5 \\ 3 & 0 & -3 & 6 \\ 2 & 4 & 5 & 7 \\ 5 & -1 & -3 & 1 \end{bmatrix}$$

$$\begin{array}{l} R_1 \leftrightarrow R_2 \\ B \rightarrow \begin{bmatrix} 3 & 0 & -3 & 6 \\ 0 & 2 & -4 & 5 \\ 2 & 4 & 5 & 7 \\ 5 & -1 & -3 & 1 \end{bmatrix} \xrightarrow{R_1/3} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & 2 & -4 & 5 \\ 2 & 4 & 5 & 7 \\ 5 & -1 & -3 & 1 \end{bmatrix} \end{array}$$

Compute determinant  
of  $A_{n \times n}$  ( $n \geq 4$ )

Use the following operations  
as done w.l. R.E. form

- (i) Interchange  $R_i$  w/  $R_j$ :
- (ii) Add  $kR_i$  to  $R_j$
- (iii) multiply C on  $R_i$ :  divide  
C on outside

$A \rightarrow L$  or  $U$ .

Using  $|L| = L_{11} \cdot L_{22} \cdots L_{nn}$

$$\begin{array}{l} R_3 \rightarrow R_1 \\ R_4 \rightarrow R_1 \\ g. \quad \xrightarrow{-3} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & 2 & -4 & 5 \\ 0 & 4 & 7 & 3 \\ 0 & -1 & 2 & -9 \end{bmatrix} \xrightarrow{R_2 \leftrightarrow R_4} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & -1 & 2 & -9 \\ 0 & 4 & 7 & 3 \\ 0 & 2 & -4 & 5 \end{bmatrix} \end{array}$$

$$\begin{array}{l} R_1 + 4R_2 \\ R_2 + 2R_3 \\ g. \quad \xrightarrow{3} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & -1 & 2 & -9 \\ 0 & 0 & 15 & -33 \\ 0 & 0 & 0 & -13 \end{bmatrix} \xrightarrow{3 \cdot 15} \begin{bmatrix} 1 & 0 & -1 & 2 \\ 0 & -1 & 2 & -9 \\ 0 & 0 & 1 & -33 \\ 0 & 0 & 0 & -13 \end{bmatrix} \end{array}$$

$$|B| = 3 \cdot 1 \cdot (-1) \cdot 15 \cdot (-13) = 3 \cdot 15 \cdot 13$$

$$\underline{3 \cdot 15 \cdot (-1) \cdot 1 \cdot (-13)}$$

### 7.9. Theorem

- a.  $A_{n \times n}$  is invertible iff.  $|A| \neq 0$ .
- b.  $|kA_{n \times n}| = k^n |A|$

c.  $|A_{n \times n} B_{n \times n}| = |A_{n \times n}| |B_{n \times n}|$

d.  $A$  is invertible  $\Rightarrow |A^{-1}| = \frac{1}{|A|}$

e.  $|A| = |A^T|$

### 7.10. Cramer's Rule

$A_{n \times n}, \vec{b} \in \mathbb{R}^n$ . Then the unique  $\stackrel{+R_1+R_2+R_3+R_4}{\text{solution of}} A\vec{x} = \vec{b}$  is given as

$$\vec{x} = [x_1 \dots x_n]^T$$

$$x_i = \frac{\det(A_i(\vec{b}))}{\det A}$$

$$A_i(\vec{b}) = \begin{bmatrix} a_1 & \dots & \underset{i}{a_i} & \dots & a_n \end{bmatrix}_{n \times n}$$

$$i=1, \dots, n.$$

$$[a_1 | \dots | a_{i-1} | \vec{b} | a_{i+1} \dots a_n]$$

### 7.11 ex Use Cramer's Rule to solve

$$\begin{aligned} x_1 + 2x_2 &= 2 \\ -x_1 + 4x_2 &= 1 \end{aligned} \quad \vec{b} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\text{Find } |A| \quad A = \begin{vmatrix} 1 & 2 \\ -1 & 4 \end{vmatrix} = 4 - (-2) = 6$$

$$|A_i(\vec{b})| = \begin{vmatrix} 2 & 2 \\ 1 & 4 \end{vmatrix} = 8 - 2 = 6$$

$$|A_{12}(\vec{b})| = \begin{vmatrix} 1 & 2 \\ -1 & 2 \end{vmatrix} = 1 + 2 = 3$$

$$x_1 = \frac{|A_i(\vec{b})|}{|A|} = \frac{6}{6} = 1 \quad x_2 = \frac{|A_{12}(\vec{b})|}{|A|} = \frac{3}{6} = \frac{1}{2} \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ \frac{1}{2} \end{bmatrix}$$

Compute determinant of

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

$$\xrightarrow{R_1+R_2} \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 3 & 7 \\ 0 & 1 & 0 & 2 \\ 0 & -1 & 1 & 0 \end{bmatrix}$$

$$\xrightarrow{R_2+R_3} \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 3 & 7 \\ 0 & 0 & -3 & -5 \\ 0 & 0 & 1 & 7 \end{bmatrix}$$

$$\xrightarrow{R_3-R_2} \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 3 & 7 \\ 0 & 0 & -6 & -12 \\ 0 & 0 & 1 & 7 \end{bmatrix}$$

$$1 \cdot 1 \cdot (-3) \cdot \frac{1}{6} =$$

7.11. Adjoint

$$C_{ji} \equiv (-1)^{j+i} |A_{ji}| = \begin{vmatrix} a_{11} & a_{12} & \cdots & 0 & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & 0 & \cdots & a_{2n} \\ a_{j1} & a_{j2} & \cdots & 1 & \cdots & a_{jn} \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & 0 & \cdots & a_{nn} \end{vmatrix}$$

$\rightarrow j^{\text{th}} \text{ row}$

is called the  $(j, i)$ th cofactor of  $A$

$$[C_{ij}] = [C_{ij}]^T \begin{bmatrix} c_{11} & \cdots & c_{1n} \\ \vdots & \ddots & \vdots \\ c_{n1} & \cdots & c_{nn} \end{bmatrix}$$

is called adjoint (adjugate),  $C = \text{adj } A$

7.12. Theorem  $A^{-1} = \frac{1}{|A|} \cdot \text{adj } A$

7.13. ex.  $A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 2 & 4 \\ 1 & 3 & -3 \end{bmatrix} \quad \begin{bmatrix} + & - & + \end{bmatrix}$

$$\begin{aligned} |A| &= 1 \begin{vmatrix} 2 & 4 \\ 3 & -3 \end{vmatrix} - 2 \begin{vmatrix} 2 & 4 \\ 1 & -3 \end{vmatrix} + (-1) \begin{vmatrix} 2 & 2 \\ 1 & 3 \end{vmatrix} \\ &= -6 - 12 + 2((6) + 4) + (-1)(6 - 2) \end{aligned}$$

$$= -18 + 20 - 4 = -2$$

$$C_{11} = + \begin{vmatrix} 2 & 4 \\ 3 & -3 \end{vmatrix} = -18 \quad C_{12} = - \begin{vmatrix} 2 & 4 \\ 1 & -3 \end{vmatrix} = 10 \quad C_{13} = + \begin{vmatrix} 2 & 2 \\ 1 & 3 \end{vmatrix} = 4$$

$$C_{21} = - \begin{vmatrix} 2 & -1 \\ 3 & -3 \end{vmatrix} = 3 \quad C_{22} = + \begin{vmatrix} 1 & -1 \\ 1 & -3 \end{vmatrix} = -2 \quad C_{23} = - \begin{vmatrix} 1 & 2 \\ 1 & 3 \end{vmatrix} = -1$$

$$C_{31} = + \begin{vmatrix} 2 & -1 \\ 2 & 4 \end{vmatrix} = 10 \quad C_{32} = - \begin{vmatrix} 1 & -1 \\ 2 & 4 \end{vmatrix} = -6 \quad C_{33} = + \begin{vmatrix} 1 & 2 \\ 2 & 2 \end{vmatrix} = 2$$

$$\text{adj } A = \begin{bmatrix} -18 & 10 & 4 \\ 3 & -2 & -1 \\ 10 & -6 & 2 \end{bmatrix}^T \Rightarrow A^{-1} = \frac{1}{-2} \begin{bmatrix} -18 & 3 & 10 \\ 10 & -2 & -6 \\ 4 & -1 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} 9 & -3/2 & -5 \\ -5 & 1/2 & 3 \\ -2 & 1/2 & 1 \end{bmatrix}$$

## Summary

1.  $2 \times 2$ .  $A^{-1} = \frac{1}{ad-bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$

2.  $[A | I] \xrightarrow{\text{R.E.}} [I | A^{-1}]$

3.  $A^{-1} = \frac{1}{|A|} (\text{adj } A)$

## 8). Diagonalization

8.1. Def.  $A_{n \times n}, B_{n \times n}$   $A$  is similar to  $B$

if  $\exists P_{n \times n} \Rightarrow P^{-1}AP = B$

denoted,  $A \sim B$

$\Rightarrow A \xrightarrow{\text{R.E.}} B$ .

8.2. Theorem. a.  $A \sim A$

b.  $A \sim B, B \sim C \Rightarrow A \sim C$

c.  $A \sim B \Rightarrow B \sim A$

8.3. Properties  $A_{n \times n} \sim B_{n \times n} \Rightarrow$

- a.  $|A| = |B|$
- b.  $\exists A^{-1}$  iff.  $\exists B^{-1}$
- c.  $\text{rank}(A) = \text{rank}(B)$
- d.  $A, B$  have the same eigenvalues
- e.  $A^m \sim B^m \quad m \in \mathbb{N}$
- f.  $\exists A^{-1} \quad A^m \sim B^m$  for any  $m \in \mathbb{Z}$

$$P^T A P = B$$

$$P P^T A P P^{-1} = P B P^{-1}$$

$$(P P^{-1}) A (P^{-1}) = P B P^{-1}$$

I

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

8.4. Def.  $A_{n \times n}$  is diagonalizable if

$$\boxed{\exists D \ni A \sim D} \quad D \text{ is diagonal matrix}$$

↓

$$\boxed{\text{i.e. } \exists P_{n \times n} \ni P^T A P = D}$$

8.5. Theorem.  $A$  is diagonalizable iff.  $A$  has  $n$  IND eigenvectors

Moreover  $A$  is diagonalizable  $\Rightarrow \exists P \ni P^{-1} A P = D$

where  $P = [\vec{v}_1 \dots \vec{v}_n]$ ,  $\vec{v}_1, \dots, \vec{v}_n$  are IND.

$$D = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \quad \downarrow \quad \text{eigenvectors of } A$$

$\lambda_1, \dots, \lambda_n$  are eigenvalues

8.6. ex.  $A = \begin{bmatrix} -1 & 0 & 1 \\ 3 & 0 & -3 \\ 1 & 0 & -1 \end{bmatrix}$

Step 1. Solve for eigenvalues.

Solve  $|A - \lambda I| = 0$

$$\begin{vmatrix} -1-\lambda & 0 & 1 \\ 3 & -\lambda & -3 \\ 1 & 0 & -1-\lambda \end{vmatrix} = 0$$

$$\begin{bmatrix} + & - & + \\ - & + & - \\ + & - & + \end{bmatrix}$$

$$+(-\lambda) \begin{vmatrix} -1-\lambda & 1 \\ 1 & -1-\lambda \end{vmatrix} = 0$$

$$\lambda [(\lambda + 2)^2 - 1] = 0$$

$$\lambda [\lambda^2 + 2\lambda] = 0$$

$$\lambda^2(\lambda + 2) = 0$$

$$\lambda_1 = \lambda_2 = 0 \quad \lambda_3 = -2$$

Step 2. Solve for eigenvectors.

Solve  $[A - \lambda I | 0]$

$$\lambda_1 = 0 \quad \begin{bmatrix} -1 & 0 & 1 & | & 0 \\ 3 & 0 & -3 & | & 0 \\ 1 & 0 & -1 & | & 0 \end{bmatrix} \xrightarrow{\text{R}_2 + 3\text{R}_1, \text{R}_3 + \text{R}_1} \begin{bmatrix} 1 & 0 & 0 & | & 0 \\ 0 & 0 & 0 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix} \Leftrightarrow -x_1 + x_3 = 0 \quad x_1 = x_3.$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \stackrel{x_1=t}{=} \begin{bmatrix} t \\ s \\ t+s \end{bmatrix} = \begin{bmatrix} t \\ 0 \\ t \end{bmatrix} + \begin{bmatrix} 0 \\ s \\ 0 \end{bmatrix} \quad x_3 \text{ can be any real number}$$

set  $x_3 = t$ .

$x_2$  can be any real number

set  $x_2 = s$ .

$$\lambda_3 = -2. \quad \begin{bmatrix} -1 & 0 & 1 & | & 0 \\ 3 & 2 & -3 & | & 0 \\ 1 & 0 & 1 & | & 0 \end{bmatrix} \xrightarrow{\text{R}_2 - 3\text{R}_1, \text{R}_3 - \text{R}_1} \begin{bmatrix} 1 & 0 & 1 & | & 0 \\ 0 & 2 & -6 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix} \xrightarrow{\text{R}_2 / 2} \begin{bmatrix} 1 & 0 & 1 & | & 0 \\ 0 & 1 & -3 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$$

$$x_1 + x_3 = 0 \quad x_1 = -x_3 \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \stackrel{x_1=-t}{=} \begin{bmatrix} -1 \\ 3 \\ 1 \end{bmatrix} + t \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} \quad v_3$$

$$x_2 - 3x_3 = 0 \quad x_2 = 3x_3$$

$$\text{Step 3. } P \equiv [\vec{v}_1 \vec{v}_2 \vec{v}_3]$$

$$P \equiv \begin{bmatrix} \lambda_1 & & 0 \\ 0 & \lambda_2 & \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

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

$$\boxed{\begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix} = P^{-1} A \begin{bmatrix} \vec{v}_1 & \vec{v}_2 & \vec{v}_3 \end{bmatrix}}$$

→ Algorithm  
to diagonalize  
matrix A

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

Remark: to find eigen vectors, which  
equivalent to  $[A - \lambda I | 0]$ .

**Case 1:** If  $\lambda_1, \lambda_2, \dots, \lambda_n$  are distinct  
plug in  $\lambda_i$ :  $[A - \lambda_i I | 0]$   
and solve  $\vec{v}_i$  one-by-one

**Case 2:** If  $\lambda_i = \lambda_j \exists i, j \in \{1, 2, \dots, n\}$

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = t \vec{v}_i + s \vec{v}_j$$

We extract  $\vec{v}_i$  as eigenvector for  $\lambda_i$   
 $\vec{v}_j \dots$  for  $\lambda_j$

**Case 3:**  $\lambda_i = \lambda_j = \lambda_k$

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = + \vec{v}_i + s \vec{v}_j + h \vec{v}_k$$

8.7 Theorem. Anxn w/ n distinct eigenvalues

~~(1)~~  $\Rightarrow$  (1) Anxn is diagonalizable

(2)  $V_1, \dots, V_k$  are IND.

8.8. Algorithm to compute  $A^k$

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

$$\Rightarrow A^k = P D^k P^{-1}$$

8.9 ex.  $A = \begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}$  find  $A^{10}$ .

Step 1. Solve for  $\lambda_1, \lambda_2$

$$|A - \lambda I| = 0$$

$$\begin{vmatrix} -\lambda & 1 \\ 2 & 1-\lambda \end{vmatrix} = 0$$

$$(-\lambda)(1-\lambda) - 2 = 0$$

$$\lambda^2 - \lambda - 2 = 0$$

$$(\lambda-2)(\lambda+1) = 0$$

$$\lambda_1 = -1 \quad \lambda_2 = 2$$

Step 2. solve  $\vec{v}_1, \vec{v}_2$

$$\lambda_1 = -1 \quad |A - \lambda I| \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

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

$$x_1 + x_2 = 0 \quad x_1 = -x_2 \stackrel{\text{set}}{=} t$$

$$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} t \\ -t \end{bmatrix} = t \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$\lambda_2 = 2 \quad |A - \lambda I| \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} -2 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow -2x_1 + x_2 = 0$$

$$2x_1 = x_2 \stackrel{\text{set}}{=} 2t \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} + t \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Step 3.  $P = [\vec{v}_1 \vec{v}_2] = \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix}$

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

Step 4.  $\begin{bmatrix} 0 & 1 \\ 2 & 1 \end{bmatrix}^{10} = \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} -1 & 0 \\ 0 & 2 \end{bmatrix}^{10} \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix}^{-1}$

$$= \begin{bmatrix} t & 1 \\ -t & 2 \end{bmatrix} \begin{bmatrix} (-1)^{10} & 0 \\ 0 & 2^{10} \end{bmatrix} \frac{1}{2+t} \begin{bmatrix} 2 & -1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} \frac{2}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

## Additional Theorems

1. Let  $A_{m \times n}$ ,  $B_{n \times m}$ . Then  $AB$  is  $m \times m$   $BA$  is  $n \times n$

Suppose  $m \leq n$

$\Rightarrow$  eigenvalues of  $BA$  are the  $m$  eigenvalues of  $AB$   
w.l. extra eigenvalues being 0.

2. Def. A symmetric matrix  $A$  is  
idempotent if  $A = AA$

Thm. If  $A$  is idempotent iff.  
all its eigenvalues are either 0 or 1.

The # of eigenvalues equal to 1 = Trace(A)

3. Thm. All eigenvalues of a symmetric  
matrix are real

4. Thm. Any symmetric matrix is diag'ble

5. Def Matrix Exponential

$$e^A \equiv \sum_{k=0}^{\infty} \frac{A^k}{k!}$$

properties (i)  $e^0 = 1$

$$(ii) e^{-A} = (e^A)^{-1}$$

$$(iii) e^{(a+b)A} = e^{aA} e^{bA}$$

$$(iv) \text{ if } AB = BA, e^{A+B} = e^A e^B$$

$$(v) \text{ if } \exists A^{-1}, e^{A^{-1}BA} = A^{-1} e^B A$$

$$(vi) |e^A| = e^{\text{tr}(A)}$$

$$\text{tr}(A) = a_{11} + a_{22} + \dots + a_{nn}$$

## 6. Condition number VIF

$\lambda_1, \dots, \lambda_n$  are eigenvalues of

$$\mathbf{X}^T \mathbf{X}$$

$$k \equiv \frac{\lambda_{\max}}{\lambda_{\min}} \quad \begin{cases} (0, 100] & \text{Good} \\ (100, 1000] & \text{moderate multicollinearity} \\ (1000, \infty) & \text{serious mult.} \end{cases}$$

$$\begin{bmatrix} 1 & L_{M_1} & L_{T_1} & K_1 \\ 1 & L_{M_2} & L_{T_2} & K_2 \\ \vdots & \vdots & \vdots & \vdots \\ 1 & L_{M_n} & L_{T_n} & K_n \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \\ \vdots \\ \beta_p \end{bmatrix} = \begin{bmatrix} Y \\ \vdots \\ Y \end{bmatrix} \quad n \times p$$

## Multicollinearity problems

- (1) large variance and covariances
- (2) explosive and sensitive  $\beta$ 's

## Treatment of multicollinearity

- (1) More data
- (2) Model Respecification
- (3) Ridge Regression
- (4) PCA
- (5)  $X_i / X_j$  | Delete  $X$