## Lecture 11 (sections 1.6,1.7)

## · · · Continue with

$$A \cdot \overline{X} = \overline{b}$$

- $\implies$  what if either:
- A is not a square matrix?
- A is a square matrix but is not invertible?
- $\implies$  then linear system is consistent only for some choice of  $\overline{b}$ .

## Example:

$$\begin{cases} x_1 + x_2 + 2x_3 = b_1 \\ x_1 + x_3 = b_2 \\ 2x_1 + x_2 + 3x_3 = b_3 \end{cases}$$

 $\implies$  we determine consistency conditions on  $\bar{b}$  using augmented matrix approach:

$$\begin{bmatrix} 1 & 1 & 2 & b_1 \\ 1 & 0 & 1 & b_2 \\ 2 & 1 & 3 & b_3 \end{bmatrix} \rightarrow \qquad \begin{matrix} r_2 \to r_2 - r_1 \\ r_3 \to r_3 - 2r_1 \end{matrix} \longrightarrow \begin{bmatrix} \textcircled{1} & 1 & 2 & b_1 \\ 0 & -1 & -1 & b_2 - b_1 \\ 0 & -1 & -1 & b_3 - 2b_1 \end{bmatrix}$$

"REF-like" because the last row  $(r_3)$  is not entirely zero in general.

 $\implies$  focus on  $r_3$ : system is consistent only if

$$|b_3 - b_1 - b_2| = 0$$

Assuming

$$|b_3 - b_1 - b_2| = 0$$

 $\Longrightarrow$ 

$$\begin{bmatrix}
1 & 1 & 2 & b_1 \\
0 & 1 & 1 & b_1 - b_2 \\
0 & 0 & 0 & 0
\end{bmatrix} 
\rightarrow 
r_1 \rightarrow r_1 - r_2 
\rightarrow
\begin{bmatrix}
1 & 0 & 1 & b_2 \\
0 & 1 & 1 & b_1 - b_2 \\
0 & 0 & 0 & 0
\end{bmatrix} 
\Leftarrow= RREF$$

- leading variables:  $x_1, x_2$
- free variable:  $x_3$
- $\implies$  There are infinitely many solutions for any

$$\overline{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 = b_1 + b_2 \end{bmatrix}$$

parameterized by  $x_3 = s$ :

$$\overline{X} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_2 - s \\ b_1 - b_2 - s \\ s \end{bmatrix}$$

However, if

$$b_3 - b_1 - b_2 \neq 0$$

 $\implies$  the system of linear equations

$$\begin{cases} x_1 + x_2 + 2x_3 = b_1 \\ x_1 + x_3 = b_2 \\ 2x_1 + x_2 + 3x_3 = b_3 \end{cases}$$

is inconsistent (there is no solution)

## Section 1.7 — diagonal, triangular and symmetric matrices

A square matrix A is diagonal if the only non-zero elements are on the main diagonal:

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}, \qquad B = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$
diagonal matricing

 $\implies$  Note:

$$A^{2} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 9 \end{bmatrix} , \qquad B^{2} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

In general, if

$$A = \underbrace{\begin{bmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ & & & a_n \end{bmatrix}}_{n \times n \text{ diagonal matrix}} \overset{def}{\equiv} \operatorname{diag}(a_1, \cdots a_n)$$

then

$$A^k = \underbrace{A \cdots A}_{k-\text{times}} = \underbrace{\begin{bmatrix} a_1^k & & \\ & a_2^k & & \\ & & \ddots & \\ & & & a_n^k \end{bmatrix}}_{n \times n \text{ diagonal matrix}}$$

Lemma:

Diagonal matrix is invertible if all diagonal matrix elements are non-zero:

$$A = \begin{bmatrix} a_1 & & & \\ & a_2 & & \\ & & \ddots & \\ & & & a_n \end{bmatrix} \equiv \operatorname{diag}(a_1, \dots a_n)$$

If

$$a_1 \cdot a_2 \cdots a_n \neq 0$$

then

$$A^{-1} = \begin{bmatrix} \frac{1}{a_1} & & & \\ & \frac{1}{a_2} & & \\ & & \ddots & \\ & & & \frac{1}{a_n} \end{bmatrix}$$

A square matrix is upper triangular (UTM) if all elements below the main diagonal are zero

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

A square matrix is lower triangular (LTM) if all elements above the main diagonal are zero

$$A^{LTM} = \begin{bmatrix} \mathbf{1} & 0 & 0 & 0 \\ 3 & \mathbf{2} & 0 & 0 \\ 4 & 1 & \mathbf{0} & 0 \\ 7 & -3 & 1 & \mathbf{3} \end{bmatrix}$$

 $\Longrightarrow$  Properties:

• Transpose of a LTM is an UTM:

$$\left(A^{LTM}\right)^T = B^{UTM}$$

• Transpose of UTM is LTM:

$$\left(A^{UTM}\right)^T = B^{LTM}$$

• A triangular matrix (LTM or UTM)

$$A = \{a_{ij}\},$$
  $i = 1 \cdots n$   
 $j = 1 \cdots n$ 

is invertible if all of its diagonal elements are non-zero:

$$a_{11} \cdot a_{22} \cdots a_{nn} \neq 0$$

- Products of LTMs is an LTM; products of UTMs is an UTM
- Inverses of LTMs or UTMs (if exist) are triangular matrices of the same type

A is a symmetric matrix if

$$A^T = A$$

⇒ Properties:

• if A and B are symmetric  $n \times n$  matrices and  $k \in \mathbb{R}$ , then the following matrices:

$$A+B$$
,  $A-B$ ,  $k\cdot A$ 

are symmetric.

• Note: if A and B are symmetric,  $A \cdot B$  is not necessarily symmetric. Indeed:

$$(A \cdot B)^T = B^T \cdot A^T = B \cdot A$$
 enly if the matrices commute

Example:

$$\begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} \cdot \begin{bmatrix} -4 & 1 \\ 1 & 0 \end{bmatrix} = \underbrace{\begin{bmatrix} -2 & 1 \\ -5 & 2 \end{bmatrix}}_{A \cdot B} \longleftarrow (A \cdot B)^T \neq A \cdot B$$

$$\begin{bmatrix} -4 & 1 \\ 1 & 0 \end{bmatrix} \cdot \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix} = \underbrace{\begin{bmatrix} -2 & -5 \\ 1 & 2 \end{bmatrix}}_{B \cdot A} \longleftarrow B \cdot A \neq A \cdot B$$

- If A is a symmetric invertible matrix,  $A^{-1}$  is symmetric
- Note that for any  $n \times n$  matrix A, matrices

$$A \cdot A^T$$
, and  $A^T \cdot A$ 

are symmetric.

Indeed:

$$(A \cdot A^T)^T = (A^T)^T \cdot A^T = A \cdot A^T$$

$$(A^T \cdot A)^T = A^T \cdot (A^T)^T = A^T \cdot A$$