

Diagonal, Triangular and Symmetric §1.7

Definition:

A diagonal matrix is a square matrix D such that $D_{ij} = 0$ if $i \neq j$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

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$\text{Diag}(1, 2, 3)$

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

Diagonal

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

↑
not
Diagonal

• $\text{Diag}(a_1, \dots, a_n)$ is the diagonal matrix

$$\begin{pmatrix} a_1 & & 0 \\ & a_2 & \\ 0 & & a_n \end{pmatrix}$$

i.e. $(\text{Diag}(a_1, \dots, a_n))_{ii} = a_i$

• An upper triangular matrix is a square matrix T such that $T_{ij} = 0$ if $i > j$

e.g.

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix}$$

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

• A lower triangular matrix is a square matrix T such that $T_{ij} = 0$ if $i < j$

Remarks:

• $\text{Diag}(a_1, \dots, a_n) \text{Diag}(b_1, \dots, b_n) = \text{Diag}(a_1 b_1, \dots, a_n b_n)$

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} \begin{pmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 3 & 0 & 0 \\ 0 & 4 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

• Stage I of Gauss-Jordan elimination gives an upper triangular matrix

• The product of two upper triangular matrices is upper triangular

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 4 & 5 \\ 0 & 0 & 6 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 2 \end{pmatrix} = \begin{pmatrix} \bullet \\ \bullet \\ \bullet \end{pmatrix}$$

... lower ...
... lower ...

Definition: A symmetric matrix is a square matrix A such that $A^T = A$

Recall: $(A^T)_{ij} = A_{ji}$

so symmetric $\Leftrightarrow A_{ij} = A_{ji}$

$$\begin{pmatrix} 1 & 4 & 5 \\ 4 & 2 & 8 \\ 5 & 8 & 3 \end{pmatrix}$$

Lemma: (i) For any matrix A , $(A^T)^T = A$

(ii) For any matrices A, B such that AB is defined

$$(AB)^T = B^T A^T$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} 2 & 8 & 1 \\ 7 & 6 & 0 \\ 2 & 3 & 4 \end{pmatrix} = \begin{pmatrix} \bullet \\ \bullet \\ \bullet \end{pmatrix}$$

$$\begin{pmatrix} 2 & 7 & 2 \\ 8 & 6 & 3 \\ 1 & 0 & 4 \end{pmatrix} \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix} = \begin{pmatrix} \cdot & & \end{pmatrix}$$

$B^T \quad A^T$

(iii) If A is invertible

then so is A^T

and $(A^T)^{-1} = (A^{-1})^T$

(indeed: $(A^{-1})^T A^T = (A A^{-1})^T = I^T = I$

since $A^T (A^{-1})^T = I$)

$$\begin{pmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{pmatrix} \begin{pmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{pmatrix}$$

Remarks: • If A and B are symmetric
then so is $A+B$

• If $AB=BA$ and A, B are symmetric
then AB is symmetric:

$$(AB)^T = (BA)^T = A^T B^T = AB$$

• If AB is symmetric and A and B are symmetric
then $AB=BA$:

$$BA = B^T A^T = (AB)^T = AB$$

• If A is symmetric and invertible

then A^{-1} is symmetric

since $(A^{-1})^T = (A^T)^{-1} = A^{-1}$

Resumé,

Given a linear system $A\underline{x} = \underline{b}$:

Gauss-Jordan yields a sequence of elementary row operations reducing A to an rref matrix R .

Applying these operations to the identity matrix, we obtain an invertible E (product of the corresponding elementary matrices) such that $EA = R$

Now since E is invertible,

$$A\underline{x} = \underline{b} \Leftrightarrow EA\underline{x} = E\underline{b} \Leftrightarrow R\underline{x} = E\underline{b}$$

(using augmented matrices simultaneously computes R and $E\underline{b}$)

$$\left(\begin{array}{c|c} A & \underline{b} \\ \hline \end{array} \right) \longrightarrow \left(\begin{array}{c|c} R & E\underline{b} \\ \hline \end{array} \right)$$

~~III~~ RAI

A is invertible $\Leftrightarrow R = I$

so $R\underline{x} = E\underline{b}$

reads $\underline{x} = E\underline{b}$

and $EA = I$ so $E = A^{-1}$

so it just says $\underline{x} = A^{-1}\underline{b}$ (unique solution for any $\underline{b}$)

• If R has a row of zeroes

then $R\underline{x} = E\underline{b}$ ~~has~~ has no solution for some $\underline{b}$

hence $A\underline{x} = \underline{b}$

✓

• If R has a column not containing a leading 1
then (by setting ~~that~~ the corresponding variable to
an arbitrary value)
 $R \underline{x} = \underline{0}$ has multiple (infinitely many) solutions
so $Ax = E\underline{0} = \underline{0}$ does too.

If A is square and not invertible,
then R has a row of zeroes and a column without
a leading 1.