

## Lecture #8

- Last time → Addition and multiplication by scalars for matrices
  - ↳ Multiplication of matrices
  - ↳ Transpose of a matrix (reverts rows ↔ columns)

Q1: Is there a nonzero column vector  $\vec{x} \in \mathbb{R}^n$  such that  $\vec{x}^T \cdot \vec{x} = 0$ ? (A: No)

Q2: If  $A$  is a  $k \times m$  matrix, and  $B$  is an  $m \times 3$  matrix whose last column is the sum of the first two. What can be said about  $A \cdot B$ ?

(A:  $A \cdot B$  is a  $k \times 3$  matrix whose last column is the sum of the first two)

### § 2.2 The inverse of a matrix

Def: An  $n \times n$  matrix  $A$  is called invertible if there is another  $n \times n$  matrix  $C$  such that:

$$\boxed{C \cdot A = I_n = A \cdot C}$$

the  $n \times n$  identity matrix

Q3: Can there be several matrices  $C$  satisfying these properties?

(A: NO. If  $C_1 \cdot A = I_n = A \cdot C_2$ , then:

$$C_1 = C_1 \cdot I_n = C_1 \cdot (A \cdot C_2) = (C_1 \cdot A) \cdot C_2 = I_n \cdot C_2 = C_2 \Rightarrow C_1 = C_2)$$

! Thus, if  $C$  exists, then it is unique and is denoted by  $A^{-1}$   
the inverse of  $A$

Ex1: Does the matrix  $\begin{pmatrix} 1 & 3 \\ 0 & 0 \end{pmatrix}$  admit an inverse?

(A: NO. If  $C = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  was an inverse, we would have e.g.  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 3 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix}$   
But  $\begin{pmatrix} 1 & 3 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a+3c & b+3d \\ 0 & 0 \end{pmatrix}$  has the (2,2)-entry equal to 0 NOT 1!

Def: An  $n \times n$  matrix  $A$  is called singular if it is not invertible.

For  $2 \times 2$  matrices, this becomes very explicit, due to the following:

CLAIM: 1) The matrix  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  is invertible iff  $ad - bc \neq 0$

2) If  $ad - bc \neq 0$ , then  $A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$

Def: The above quantity  $\det(A) := ad - bc$  is called the determinant of  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$

Q4: Compute the inverse of  $A = \begin{pmatrix} 1 & 2 \\ 3 & 7 \end{pmatrix}$

(A:  $A^{-1} = \begin{pmatrix} 7 & -2 \\ -3 & 1 \end{pmatrix}$ )

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### • Properties of inverse

1)  $A$ -invertible  $\Rightarrow A^{-1}$ -invertible :  $(A^{-1})^{-1} = A$

2)  $A, B$ -invertible  $n \times n$  matrices  $\Rightarrow AB$ -invertible :  $(AB)^{-1} = B^{-1} \cdot A^{-1}$

3)  $A$ -invertible  $\Rightarrow A^T$ -invertible :  $(A^T)^{-1} = (A^{-1})^T$

CLAIM : If  $A$  is an invertible  $n \times n$  matrix, then for any  $\vec{b} \in \mathbb{R}^n$  there is a unique solution of  $A\vec{x} = \vec{b}$ , given explicitly by  $\vec{x} = A^{-1} \cdot \vec{b}$

Question: How to compute  $A^{-1}$  (know the  $2 \times 2$  case so far)?

CLAIM : An  $n \times n$  matrix  $A$  is invertible iff  $A$  is row equivalent to  $I_n$ .

In this case, any sequence of elementary row operations that reduces  $A$  to  $I_n$ , also transforms  $I_n$  to  $A^{-1}$ .

In order not to do the same work twice, the above claim can be interpreted as the following algorithm to find inverse  $A^{-1}$ :

Algorithm : Row reduce the  $n \times n$  matrix  $(A; I_n)$ .

If  $A$  is row equivalent to  $I_n$ , then  $(A; I_n)$  is row equivalent to  $(I_n; A^{-1})$ . Otherwise,  $A$  is singular

### Another perspective to this algorithm

The row reduction of  $(A; I_n)$  to  $(I_n; A^{-1})$  can be viewed as a simultaneous solution of  $n$  vector equations

$$A\vec{x} = \vec{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad A\vec{x} = \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \quad \dots \quad A\vec{x} = \vec{e}_n = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$$

But by definition of the inverse and product, we exactly know that the  $i^{\text{th}}$  column of  $B$  is a solution of the  $i^{\text{th}}$  equation above.



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### §2.3 Characterizations of invertible matrices

The notion of "invertible matrix" is closely related with previously discussed notions & questions due to the following important result:

CLAIM: Let  $A$  be an  $n \times n$  matrix. The following are equivalent:

- 1)  $A$ -invertible
- 2)  $A$ -row equivalent to  $I_n$
- 3)  $A$  has  $n$  pivot positions
- 4)  $A\vec{x} = \vec{0}$  has only the trivial solution
- 5) The columns of  $A$  form a linearly independent set
- 6) The linear transformation  $\vec{x} \mapsto A\vec{x}$  is one-to-one
- 7) The linear transformation  $\vec{x} \mapsto A\vec{x}$  maps  $\mathbb{R}^n$  onto  $\mathbb{R}^n$ .
- 8) The columns of  $A$  span  $\mathbb{R}^n$ .
- 9) The equation  $A\vec{x} = \vec{b}$  has at least (and actually exactly) one solution for any  $\vec{b} \in \mathbb{R}^n$ .
- 10) There is an  $n \times n$  matrix  $C$  such that  $CA = I_n$
- 11) There is an  $n \times n$  matrix  $D$  such that  $AD = I_n$
- 12)  $A^T$  is invertible

Comment on 10) & 11):  $\left. \begin{array}{l} \text{If } CA = I_n \xrightarrow{\text{CLAIM}} A\text{-invertible} \Rightarrow C = A^{-1} \\ AD = I_n \xrightarrow{\text{CLAIM}} D = A^{-1} \end{array} \right\} \Rightarrow$

CLAIM: If  $A, B$  are  $n \times n$  matrices satisfying  $AB = I_n$ , then  $A, B$  - invertible with  $A^{-1} = B$  and  $B^{-1} = A$ .

! Most efficient way (so far) is to apply 2) or 3) to decide if  $A$  is invertible or not.

Ex 2: Is the matrix  $A = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{pmatrix}$  invertible?

$A \sim \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & -4 & -8 & -12 \\ 0 & -8 & -16 & -24 \\ 0 & -12 & -24 & -35 \end{pmatrix} \sim \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \Rightarrow$  not invertible as we got zero row!

! The above Claim refers only to square matrices.

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Q: What is the meaning of this notion "invertible" in terms of the corresponding linear transformations  $\mathbb{R}^n \rightarrow \mathbb{R}^n$ ?

First, let me give a definition:

Def: A linear transformation  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  is said to be invertible if there exists a transformation  $S: \mathbb{R}^n \rightarrow \mathbb{R}^n$  such that

$$\boxed{T(S(\vec{x})) = \vec{x} = S(T(\vec{x})) \text{ for any } \vec{x} \in \mathbb{R}^n.} \quad (*)$$

CLAIM: Let  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a linear transformation,  $A$  - its standard matrix. Then,  $T$ -invertible iff  $A$ -invertible matrix. In that case, the linear transformation  $S: \vec{x} \mapsto A^{-1}\vec{x}$  is the unique solution  $S$  satisfying  $(*)$

So:  $S$  is unique & linear (if it exists).

Ex 3: a) Is it true that a one-to-one linear transformation  $\mathbb{R}^1 \rightarrow \mathbb{R}^3$  is onto?

b) Is it true that a one-to-one linear transformation  $\mathbb{R}^3 \rightarrow \mathbb{R}^3$  is onto?

▶ a) No, e.g.  $T: x \mapsto \begin{pmatrix} x \\ x \\ x \end{pmatrix}$  is one-to-one, but  $\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \in \mathbb{R}^3$  is not in the range.

b) Yes: by Claim on p. 3: the standard matrix is invertible  $\Rightarrow T$  is onto

Ex 4: Assume  $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$  is a linear transformation such that

$$T(\vec{e}_1) = T(3\vec{e}_2). \text{ Can } T \text{ be onto?}$$

▶

$T(\vec{e}_1) = T(3\vec{e}_2) \Rightarrow T$  is not one-to-one  $\Rightarrow$  its standard matrix  $A$  is not invertible  
 $\Downarrow$   
 $T$  is not onto

Let us now move to a new concept which is closely related with matrices and matrix equations.

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## § 2.8 Subspaces of $\mathbb{R}^n$

Def: A set  $H \subseteq \mathbb{R}^n$  is called a subspace of  $\mathbb{R}^n$  if the following holds:

- 1)  $\vec{0} \in H$
- 2)  $\forall \vec{u}, \vec{v} \in H \Rightarrow \vec{u} + \vec{v} \in H$
- 3)  $\forall \vec{u} \in H \Rightarrow c \cdot \vec{u} \in H$  for any  $c \in \mathbb{R}$

Examples:  $\{\vec{0}\}$ , line  $\{\mathbb{R} \cdot \vec{v}\}$ , coordinate plane  $\mathbb{R}^2_{x_1, x_2} \subseteq \mathbb{R}^3_{x_1, x_2, x_3}$



Generalizing the above examples, we see that

Given any collection  $\{\vec{v}_1, \dots, \vec{v}_k\}$  of vectors in  $\mathbb{R}^n$ , their span  $\text{Span}\{\vec{v}_1, \dots, \vec{v}_k\}$  is a subspace of  $\mathbb{R}^n$

! The converse is also true - shall be discussed below.

Let's now discuss two constructions producing subspaces given a matrix

Def: The column space of a matrix  $A$ , denoted Col  $A$ , is the set of all linear combinations of the columns of  $A$  (i.e. the span of columns)

For an  $m \times n$  matrix  $A \rightsquigarrow \text{Col } A$  is thus a subspace of  $\mathbb{R}^m$   
(! Col  $A$  coincides with the range of the corresponding linear transformation  $\mathbb{R}^n \rightarrow \mathbb{R}^m$ )

Def: The null space of a matrix  $A$ , denoted Nul  $A$ , is the set of all solutions of the homog. equation  $A\vec{x} = \vec{0}$

Clearly, given an  $m \times n$  matrix  $A$ , Nul  $A$  is a subspace of  $\mathbb{R}^n$ .

Note: While Col  $A$  is defined explicitly, Nul  $A$  is defined above quite implicitly. To get an explicit description of Nul  $A$ , one needs to write sol-ns of  $A\vec{x} = \vec{0}$  in parametric form.



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Def: A basis of a subspace  $H$  of  $\mathbb{R}^n$  is a linearly independent set in  $H$  which spans  $H$ .

Regarding the above two constructions, we have:

CLAIM 1: Solving homogeneous matrix equation  $A\vec{x} = \vec{0}$  in the parametric vector form automatically yields a basis in  $\text{Nul } A$ .

CLAIM 2: The pivot columns of a matrix  $A$  form a basis of  $\text{Col } A$ .

! Note that we say "a basis" as there are too many ways to pick one.

Ex 5: For  $A = \begin{pmatrix} 1 & 2 & 7 & 10 & 12 \\ 2 & 5 & 8 & 11 & 6 \\ 4 & 6 & 9 & 5 & 0 \end{pmatrix}$  find a basis of  $\text{Col } A$  and  $\text{Nul } A$ .