

## Lecture 9: §1.7 Linear Independence & Nonsingular Matrices

Last Time: • Uniqueness of inverses of  $n \times n$  matrices (whenever they exist)

- Algorithm To find inverse for  $A$  of size  $n \times n$

$$\text{Compute: } [A | I_n] \sim [A' | B'] \quad \underline{A' \text{ REF}}$$

- ① If  $A' = I_n$ , then  $A$  is invertible &  $B' = A^{-1}$
- ② If  $A' \neq I_n$ , then  $A$  is not invertible.

- Prop: If  $A \cdot \underline{x} = \underline{b}$   $A$  of size  $n \times n$  is invertible, then the system has a unique solution  $\underline{x} = A^{-1} \underline{b}$  for any  $\underline{b}$  in  $\mathbb{R}^n$ .

TODAY: • Understand better how the Algorithm (\*) works (via nonsingular matrices)

- Define & study linear independence of vectors in  $\mathbb{R}^n$
- Find other tests for checking when a matrix is invertible.

## Nonsingular Matrices

Def: An  $n \times n$  matrix  $A$  is nonsingular whenever  $A\underline{x} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$  has a unique solution (the zero solution). Otherwise, we say  $A$  is singular.

Prop:  $A$  is nonsingular if, and only if, all  $A\underline{x} = \underline{b}$  are always consistent, with unique solutions (for all choices of constant vectors  $\underline{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ ).

- This is consistent with our algorithm: pick  $n$  values  $\underline{b} = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$
- If  $A$  is nonsingular, then  $[A | I_n] \sim [I_n | \overset{A^{-1}}{B'}]$ , & so  $A$  is invertible
- Since  $A\underline{x} = \underline{b}$  for  $A$  invertible has unique soln  $\underline{x} = A^{-1}\underline{b}$ , we get:

Theorem 1:  $A$  of size  $n \times n$  is nonsingular if, and only if,  $A$  is invertible

Useful Lemma: Fix  $P, Q$  of size  $n \times n$ , & write  $R = PQ$ .

[ If  $P$  or  $Q$  are singular, then so is  $R$ .

Consequence: If  $AB = I_n$  with  $A, B$  of size  $n \times n$ , then  $BA = I_n$ .

Why?

**Unit Vectors**

In  $\mathbb{R}^n$  we have  $n$  special vectors of norm 1.

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad e_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \dots, \quad e_n = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{bmatrix} \quad (\text{Standard Unit Vectors})$$

## Linear Independence in $\mathbb{R}^n$

Def. • A set of vectors  $\{\vec{v}_1, \dots, \vec{v}_p\}$  in  $\mathbb{R}^n$  is linearly independent (li) if the only solution  $(a_1, \dots, a_p)$  to the vector equation:

$$(*) \quad a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_p \vec{v}_p = \vec{0} \quad \text{in } \mathbb{R}^n$$

is the trivial one, that is  $a_1 = a_2 = \dots = a_p = 0$ .

• If a non-trivial solution to  $(*)$  exists, we say the set  $\{\vec{v}_1, \dots, \vec{v}_p\}$

is linearly dependent (ld)

Example: ①  $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$  is

②  $\left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$  is

③  $\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -1 \end{bmatrix} \right\}$  is

Example:  $\vec{v}_1 = \begin{bmatrix} 2 \\ 3 \\ 1 \end{bmatrix}$ ,  $\vec{v}_2 = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}$ ,  $\vec{v}_3 = \begin{bmatrix} 8 \\ 11 \\ 8 \end{bmatrix}$ ,  $\vec{v}_4 = \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}$

①  $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$  is l.d.

②  $\{\vec{v}_1, \vec{v}_2, \vec{v}_4\}$  is l.i.

Q Why study independence?

Proposition 1: The system  $A \cdot \underline{x} = \underline{0}$  has a unique solution if and only if the  $p$  columns of  $A$  are li. vectors in  $\mathbb{R}^n$ .

Proposition 2:  $A$  of size  $n \times p$ ,  $\underline{b} \in \mathbb{R}^n$ . Then, the system  $A \underline{x} = \underline{b}$  has a solution if and only if  $\underline{b}$  is a linear combination of the  $p$  columns of  $A$  ( $\underline{x}_1 \text{col}_1(A) + \underline{x}_2 \text{col}_2(A) + \dots + \underline{x}_p \text{col}_p(A) = \underline{b}$ )

The solution is unique only when the columns of  $A$  are l.i.

Theorem: If  $\{\vec{v}_1, \dots, \vec{v}_p\}$  are l.d with  $a_1 \vec{v}_1 + \dots + a_p \vec{v}_p = \underline{0}$  &  $a_i \neq 0$ , then  $\vec{v}_i$  is a linear comb. of the remaining vectors.

## Useful Properties

Prop 0: Standard unit vectors are li

Prop 1: Reordering vectors preserves linear independency / li.

Prop 2: A subset of a li set is li

Theorem 3: If  $\{\vec{v}_1, \dots, \vec{v}_p\}$  in  $\mathbb{R}^n$  &  $p > n$ , then  $\{\vec{v}_1, \dots, \vec{v}_p\}$  is li.

Theorem 4: A matrix is non-singular if and only if cols are li

## Summary of Results

Fix  $A$  of size  $n \times n$ . The following statements are equivalent.

- ①  $A$  is nonsingular ( $A \underline{x} = \underline{0}$  has a unique solution)
- ② The  $n$  columns of  $A$  are li
- ③  $A \underline{x} = \underline{b}$  has a unique solution for all  $\underline{b}$  in  $\mathbb{R}^n$ .
- ④  $A$  is invertible
- ⑤  $A$  is row equivalent to  $I_n$