

Lecture #2

08/27/2020

- Last time: 1) Linear eq-n, linear system, solution of a linear system
2) Elimination of variables to solve a linear system via elementary row operations, applied to the augmented matrix.
↓
(Recall them!)
3) Notions of echelon AND reduced echelon forms

• Warming up

Ex 1: For which values of numbers c, d the linear system

$$(*) \begin{cases} x_1 + cx_2 = 3 \\ 2x_1 - 4x_2 = d \end{cases}$$

- 1) has exactly one solution
- 2) has infinitely many solutions
- 3) is inconsistent (i.e. has no solutions)

$$\begin{pmatrix} 1 & c & | & 3 \\ 2 & -4 & | & d \end{pmatrix} \xrightarrow{R_2 \leftarrow R_2 - 2R_1} \begin{pmatrix} 1 & c & | & 3 \\ 0 & -4-2c & | & d-6 \end{pmatrix} \rightsquigarrow \text{corresponds to system } \begin{cases} x_1 + cx_2 = 3 \\ (-4-2c)x_2 = d-6 \end{cases}$$

Case 1: $-4-2c \neq 0$ (equivalently, $c \neq -2$)

Then, x_2 is uniquely recovered: $x_2 = \frac{d-6}{-4-2c}$. Plugging this value of x_2 into the 1st equation uniquely recovers x_1 . Thus: there is exactly 1 solution.

Case 2: $-4-2c = 0$ (i.e. $c = -2$)

I) if $d-6 \neq 0$ (i.e. $d \neq 6$), then there is no x_2 such that $0 \cdot x_2 = d-6$
 \Rightarrow no solutions

II) if $d-6 = 0$ (i.e. $d = 6$), then 2nd equation reads $0 \cdot x_2 = 0$ and any value of x_2 works. For any x_2 , the 1st equation recovers x_1 uniquely. So: there are infinitely many solutions.

Conclusion:
1) (*) has exactly one solution iff $c \neq -2$
2) (*) has infinitely many solutions iff $c = -2$ & $d = 6$
3) (*) has no solutions iff $c = -2$ AND $d \neq 6$

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Before we proceed to the next example, let's recap 2 definitions:

// Def: A pivot position in a matrix A is a location that corresponds to a leading entry 1 in the reduced echelon form.

// Def: A pivot column is a column of A that contains a pivot position.

! Note: Pivot position and pivot column are already seen from an echelon form

e.g. in $\begin{pmatrix} \boxed{2} & 3 & -5 & 7 & 10 \\ 0 & \boxed{5} & 0 & 6 & 12 \\ 0 & 0 & 0 & \boxed{-1} & 3 \end{pmatrix}$ the boxed positions are pivot positions

↑
pivot columns

Ex 2: Apply elementary row operations to transform

$$\begin{pmatrix} 0 & 1 & 2 & -3 & 1 & 5 \\ 2 & 1 & 8 & 5 & 0 & 16 \\ 1 & 2 & 7 & -2 & 0 & 17 \end{pmatrix}$$

first into echelon form and then into the reduced echelon form.

$$\begin{pmatrix} 0 & 1 & 2 & -3 & 1 & 5 \\ 2 & 1 & 8 & 5 & 0 & 16 \\ 1 & 2 & 7 & -2 & 0 & 17 \end{pmatrix} \xrightarrow{R_1 \leftrightarrow R_3} \begin{pmatrix} 1 & 2 & 7 & -2 & 0 & 17 \\ 2 & 1 & 8 & 5 & 0 & 16 \\ 0 & 1 & 2 & -3 & 1 & 5 \end{pmatrix} \xrightarrow{R_2 \mapsto R_2 - 2R_1}$$

$$\xrightarrow{\begin{pmatrix} 1 & 2 & 7 & -2 & 0 & 17 \\ 0 & -3 & -6 & 9 & 0 & -18 \\ 0 & 1 & 2 & -3 & 1 & 5 \end{pmatrix}} \xrightarrow{R_2 \leftrightarrow R_3} \begin{pmatrix} 1 & 2 & 7 & -2 & 0 & 17 \\ 0 & 1 & 2 & -3 & 1 & 5 \\ 0 & -3 & -6 & 9 & 0 & -18 \end{pmatrix} \xrightarrow{R_3 \mapsto R_3 + 3R_2}$$

$$\xrightarrow{\begin{pmatrix} \boxed{1} & 2 & 7 & -2 & 0 & 17 \\ 0 & \boxed{1} & 2 & -3 & 1 & 5 \\ 0 & 0 & 0 & 0 & \boxed{3} & -3 \end{pmatrix}} \xrightarrow{R_3 \mapsto \frac{1}{3}R_3} \begin{pmatrix} 1 & 2 & 7 & -2 & 0 & 17 \\ 0 & 1 & 2 & -3 & 1 & 5 \\ 0 & 0 & 0 & 0 & 1 & -1 \end{pmatrix} \xrightarrow{R_2 \mapsto R_2 - R_3}$$

already in echelon form
BUT not in reduced echelon

$$\xrightarrow{\begin{pmatrix} 1 & 2 & 7 & -2 & 0 & 17 \\ 0 & 1 & 2 & -3 & 0 & 6 \\ 0 & 0 & 0 & 0 & 1 & -1 \end{pmatrix}} \xrightarrow{R_1 \mapsto R_1 - 2R_2} \begin{pmatrix} \boxed{1} & 0 & 3 & 4 & 0 & 5 \\ 0 & \boxed{1} & 2 & -3 & 0 & 6 \\ 0 & 0 & 0 & 0 & \boxed{1} & -1 \end{pmatrix}$$

reduced echelon form!

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Let us now relate these row operations back to solving linear systems.

Ex 3: Solve the following linear system

$$\begin{cases} x_2 + 2x_3 - 3x_4 + x_5 = 5 \\ 2x_1 + x_2 + 8x_3 + 5x_4 = 16 \\ x_1 + 2x_2 + 7x_3 - 2x_4 = 17 \end{cases}$$

▶ We start by writing down the corresponding augmented matrix:

$$\left(\begin{array}{ccccc|c} 0 & 1 & 2 & -3 & 1 & 5 \\ 2 & 1 & 8 & 5 & 0 & 16 \\ 1 & 2 & 7 & -2 & 0 & 17 \end{array} \right)$$

As this is exactly the matrix from Ex 2, we know by now that it is row equivalent to the following reduced echelon matrix:

$$\left(\begin{array}{ccccc|c} 1 & 0 & 3 & 4 & 0 & 5 \\ 0 & 1 & 2 & -3 & 0 & 6 \\ 0 & 0 & 0 & 0 & 1 & -1 \end{array} \right)$$

Let us now write down the corresponding linear system:

$$\begin{cases} x_1 + 3x_3 + 4x_4 = 5 \\ x_2 + 2x_3 - 3x_4 = 6 \\ x_5 = -1 \end{cases}$$

The 3rd equation determines x_5 uniquely: $x_5 = -1$

But for any values of x_3 and x_4 , the values of x_2 and x_1 are uniquely determined by 2nd & 1st equations:

$$x_2 = 6 - 2x_3 + 3x_4 \quad \text{AND} \quad x_1 = 5 - 3x_3 - 4x_4.$$

NOTE: x_3, x_4 - free, i.e. can take any values!

Answer:

$$\begin{cases} x_1 = 5 - 3x_3 - 4x_4 \\ x_2 = 6 - 2x_3 + 3x_4 \\ x_3 \text{ is free} \\ x_4 \text{ is free} \\ x_5 = -1 \end{cases}$$

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In the above examples, the variables x_1, x_2, x_5 , corresponding to pivot columns are called "basic variables", while the other variables x_3, x_4 are called "free variables".

Rmk: a) The "free variables" can take any values, while "basic variables" will be uniquely determined by them.

b) Returning to the system

$$\begin{cases} x_1 + 3x_3 + 4x_4 = 5 \\ x_2 + 2x_3 - 3x_4 = 6 \\ x_5 = -1 \end{cases}$$

we could also let x_2, x_3 to take any values, then the 2nd equation would uniquely determine x_4 , and then the 1st equation would uniquely determine x_1 .

! However, it is customary to make "free variables" to take any values.

Rmk: While reducing a matrix to an echelon form we move from top rows to bottom rows, when transforming echelon to a reduced echelon form, we move backwards: from bottom rows to top rows (note: this is equivalent to substitution algorithm)

Q: What type of solutions does the system of Ex3 have?
(Answer: infinitely many)

Using the above argument, one arrives at the following result:

CLAIM: a) A linear system is consistent iff the rightmost column of the augmented matrix is NOT a pivot column, i.e. there is no row $(0 \dots 0 \mid b \neq 0)$ in its echelon form

b) If a linear system is consistent, then:

- I) it has a unique solution iff there are no free variables
- II) it has infinitely many solutions when there is at least one free variable.

↖ Present / Discuss why this is so...

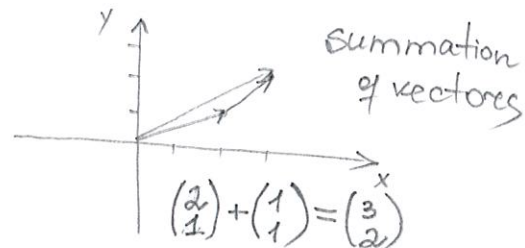
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Let us now move to

§ 1.3 Vector equations

Def: A matrix with only one column is called a column vector (or just a vector)

Examples: $\begin{pmatrix} 1 \\ 3 \end{pmatrix} \in \mathbb{R}^2$
↑ 2-dimensional real plane
symbol "∈" means "belongs to".



Def: For any positive integer n , \mathbb{R}^n denotes the collection of all $n \times 1$ column vectors.

There are two basic operations one can do with vectors:

1) Addition, e.g. $\begin{pmatrix} a_1 \\ b_1 \\ c_1 \end{pmatrix} + \begin{pmatrix} a_2 \\ b_2 \\ c_2 \end{pmatrix} = \begin{pmatrix} a_1 + a_2 \\ b_1 + b_2 \\ c_1 + c_2 \end{pmatrix}$

Both are defined coordinate-wise.

2) Scalar Multiplication, e.g. $d \cdot \begin{pmatrix} a \\ b \\ c \end{pmatrix} = \begin{pmatrix} d \cdot a \\ d \cdot b \\ d \cdot c \end{pmatrix}$

Natural Properties (assume $\vec{u}, \vec{v}, \vec{w} \in \mathbb{R}^n$ while $a, b \in \mathbb{R}$):

$$\begin{aligned} \vec{u} + \vec{v} &= \vec{v} + \vec{u} \\ (\vec{u} + \vec{v}) + \vec{w} &= \vec{u} + (\vec{v} + \vec{w}) \\ \vec{u} + \vec{0} &= \vec{u} = \vec{0} + \vec{u} \\ a(\vec{u} + \vec{v}) &= a\vec{u} + a\vec{v} \\ (a+b)\vec{u} &= a\vec{u} + b\vec{u} \\ a(b\vec{u}) &= (ab)\vec{u} \\ 1 \cdot \vec{u} &= \vec{u} \end{aligned}$$

we use arrows above to indicate that those are vectors

← here: $\vec{0} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$ - so-called zero vector

Def: Given vectors $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \in \mathbb{R}^n$ and scalars $c_1, \dots, c_k \in \mathbb{R}$, we call $\vec{w} = c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k$ a linear combination of $\vec{v}_1, \dots, \vec{v}_k$ (with weights c_1, \dots, c_k).

Def: Given $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k \in \mathbb{R}^n$, the span of $\vec{v}_1, \dots, \vec{v}_k$, denoted $\text{Span}\{\vec{v}_1, \dots, \vec{v}_k\}$ is a subset of \mathbb{R}^n consisting of all linear combinations of $\vec{v}_1, \dots, \vec{v}_k$ (i.e. a subset of all $\vec{w} \in \mathbb{R}^n$ which can be written as $\vec{w} = c_1\vec{v}_1 + \dots + c_k\vec{v}_k$)

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Q: Describe the span of

a) $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \in \mathbb{R}^3$

b) $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \in \mathbb{R}^3$

c) $\begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \in \mathbb{R}^3$

Let us conclude by relating vector equation

$$x_1 \cdot \vec{v}_1 + x_2 \cdot \vec{v}_2 + \dots + x_k \cdot \vec{v}_k = \vec{w}$$

to linear systems - whose augmented matrix has columns $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k, \vec{w}$.

Ex 4: Let $\vec{v}_1 = \begin{pmatrix} 1 \\ 3 \\ -2 \end{pmatrix}, \vec{v}_2 = \begin{pmatrix} -2 \\ 1 \\ 4 \end{pmatrix}, \vec{w} = \begin{pmatrix} -4 \\ 9 \\ 8 \end{pmatrix}$.

Determine if \vec{w} belongs to $\text{Span}\{\vec{v}_1, \vec{v}_2\}$ or not.

► This boils down to the question if there are scalars x_1, x_2 such that

$$x_1 \begin{pmatrix} 1 \\ 3 \\ -2 \end{pmatrix} + x_2 \begin{pmatrix} -2 \\ 1 \\ 4 \end{pmatrix} = \begin{pmatrix} -4 \\ 9 \\ 8 \end{pmatrix}, \text{ equiv. solving } \begin{cases} x_1 - 2x_2 = -4 \\ 3x_1 + x_2 = 9 \\ -2x_1 + 4x_2 = 8 \end{cases}$$

Augmented matrix: $\left(\begin{array}{cc|c} 1 & -2 & -4 \\ 3 & 1 & 9 \\ -2 & 4 & 8 \end{array} \right) \xrightarrow{\substack{R_2 \rightarrow R_2 - 3R_1 \\ R_3 \rightarrow R_3 + 2R_1}} \left(\begin{array}{cc|c} 1 & -2 & -4 \\ 0 & 7 & 21 \\ 0 & 0 & 0 \end{array} \right)$

The latter encodes the linear system $\begin{cases} x_1 - 2x_2 = -4 \\ 7x_2 = 21 \\ 0 = 0 \end{cases} \Rightarrow \begin{matrix} x_2 = 3 \\ \Downarrow \\ x_1 = 2 \end{matrix}$

Answer: Yes ($\vec{w} = 2\vec{v}_1 + 3\vec{v}_2$)