

## Lecture 2: Vector equations, Matrix equations.

(book: 1.3, 1.4)

Previous lecture: geometric / row point of view to an SLE  
+ Gaussian elimination.

Today: column point of view to an SLE.

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$$\begin{aligned}x + y &= 30 \\ 2x + 4y &= 74\end{aligned}$$

$$A = \begin{bmatrix} 1 & 1 \\ 2 & 4 \end{bmatrix} \quad \underline{b} = \begin{bmatrix} 30 \\ 74 \end{bmatrix}$$

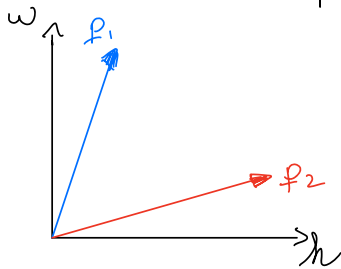
From a machine learning perspective: columns of  $A$  are feature vectors (vectors that collect features that we measured for each animal).

For instance, height and weight  $\underline{p} = \begin{bmatrix} h \\ w \end{bmatrix}$

Collect data from  $N$  animals  $\rightarrow N$  feature vectors:

$$\underline{p}_1 = \begin{bmatrix} h_1 \\ w_1 \end{bmatrix}, \underline{p}_2 = \begin{bmatrix} h_2 \\ w_2 \end{bmatrix}, \dots, \underline{p}_N = \begin{bmatrix} h_N \\ w_N \end{bmatrix}.$$

These vectors are points in the feature space ( $\mathbb{R}^2$ ).



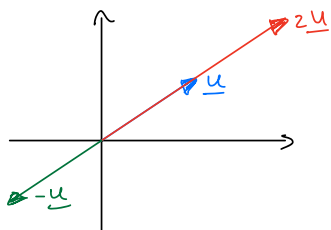
In general:  $n$  features  $\rightarrow (n \times 1)$  vector  $\underline{v} = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \in \mathbb{R}^n$

$$\begin{bmatrix} 2 \\ 3 \end{bmatrix} \neq \begin{bmatrix} 3 \\ 2 \end{bmatrix}.$$

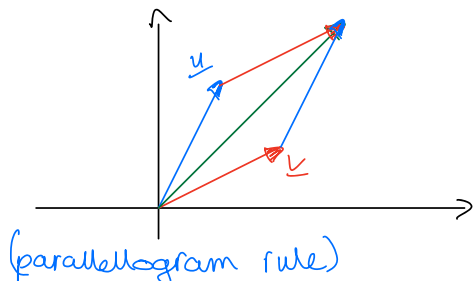
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We can perform operations with vectors:

\* scaling  $\underline{u} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$   $2\underline{u} = \begin{bmatrix} 6 \\ 4 \end{bmatrix}$   $-\underline{u} = (-1) \cdot \underline{u} = \begin{bmatrix} -3 \\ -2 \end{bmatrix}$ .



\* addition  $\underline{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$   $\underline{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$   $\underline{u} + \underline{v} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$



Algebraic properties of  $\mathbb{R}^n$ : book p.3.

Combining these two operations:

Given vectors  $\{\underline{v}_1, \underline{v}_2, \dots, \underline{v}_p\}$  in  $\mathbb{R}^n$  and  $c_1, \dots, c_p$  the vector

$$\underline{y} = c_1 \cdot \underline{v}_1 + c_2 \cdot \underline{v}_2 + \dots + c_p \cdot \underline{v}_p$$

is called a **linear combination** of  $\underline{v}_1, \dots, \underline{v}_p$  with weights  $c_1, \dots, c_p$ .

Example:  $\underline{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$   $\underline{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$   $\underline{b} = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$

Q: can  $\underline{b}$  be written as a linear combination of  $\underline{u}$  and  $\underline{v}$ ?  
i.e., can we find  $c_1$  and  $c_2$  such that  $c_1 \cdot \underline{u} + c_2 \cdot \underline{v} = \underline{b}$ ?

$$c_1 \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix} \quad \leftarrow \text{vector equation}$$

$$\begin{bmatrix} c_1 \\ 2c_1 \end{bmatrix} + \begin{bmatrix} 2c_2 \\ c_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$$

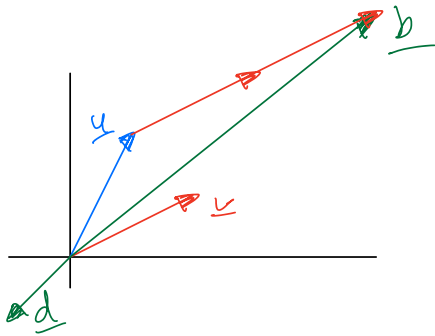
$$\begin{bmatrix} c_1 + 2c_2 \\ 2c_1 + c_2 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$$

$$\begin{cases} c_1 + 2c_2 = 5 \\ 2c_1 + c_2 = 4 \end{cases}$$

$\leftarrow$  hey, that is an **SLE!**

\* Every SLE can be written as a vector equation, and the other way around.

\* Solving an SLE means investigating whether  $\underline{b}$  can be written as a linear combination of the columns of  $A$ .  
(column point of view).



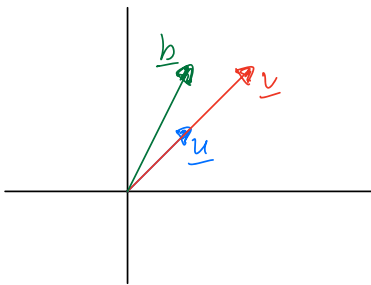
$$\underline{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad \underline{v} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$\underline{b} = \underline{u} + 2 \cdot \underline{v}$$

$$\underline{d} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

$$-\frac{1}{3} \underline{u} + (-\frac{1}{3}) \underline{v}$$

Example: (no sol.)  $\underline{u} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \underline{v} = \begin{bmatrix} 2 \\ 2 \end{bmatrix} \quad \underline{b} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$



Hence, there is no way to obtain  $\underline{b}$  by taking a linear combination of  $\underline{u}$  and  $\underline{v}$ .

Example ( $\infty$  many solutions in  $\mathbb{R}^2$ ):

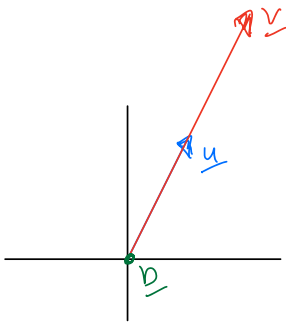
$$\underline{u} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad \underline{v} = \begin{bmatrix} 2 \\ 4 \end{bmatrix} \quad \underline{b} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$2 \cdot \underline{u} + (-1) \cdot \underline{v} = \underline{b}$$

$$0 \cdot \underline{u} + 0 \cdot \underline{v} = \underline{b}$$

$$\underline{u} + (-\frac{1}{2}) \cdot \underline{v} = \underline{b}$$

There are  $\infty$  many ways to linearly combine  $\underline{u}$  and  $\underline{v}$ .



$$\begin{cases} x_1 + 2x_2 = 0 \\ 2x_1 + 4x_2 = 0 \end{cases}$$

$$\left[ \begin{array}{cc|c} 1 & 2 & 0 \\ 2 & 4 & 0 \end{array} \right] \xrightarrow{R_2: R_2 - 2 \cdot R_1} \left[ \begin{array}{cc|c} 1 & 2 & 0 \\ 0 & 0 & 0 \end{array} \right]$$

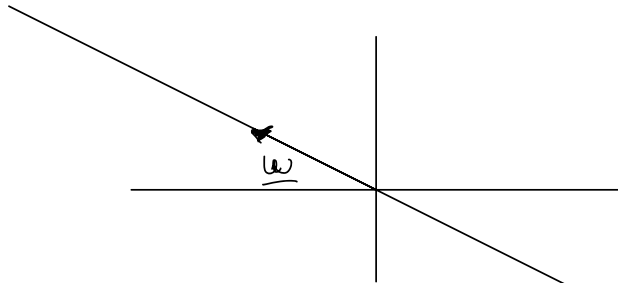
$$\begin{cases} x_1 = -2x_2 \\ x_2 \text{ is free} \end{cases} \quad (\text{parametric form})$$

$$\underline{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -2x_2 \\ x_2 \end{bmatrix} = x_2 \cdot \begin{bmatrix} -2 \\ 1 \end{bmatrix} \quad (\text{parametric vector form})$$

Hence, the solution set is  $x_2 \cdot \underline{w}$ , where  $\underline{w} = \begin{bmatrix} -2 \\ 1 \end{bmatrix}$ .

i.e.,  $\text{Span}\{\underline{w}\}$

↳ any scalar multiple of  $\underline{w}$ .



any solution on this line is a solution to the SUE.

Example (∞ many solutions in  $\mathbb{R}^3$ )

$$\begin{aligned} x_1 - 3x_2 + 2x_3 &= 0 \\ 2x_1 - 6x_2 + 4x_3 &= 0 \end{aligned}$$

$$\begin{bmatrix} 1 & -3 & 2 & | & 0 \\ 2 & -6 & 4 & | & 0 \end{bmatrix} \xrightarrow{R_2: R_2 - 2 \cdot R_1} \begin{bmatrix} 1 & -3 & 2 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$$

$$\begin{cases} x_1 = 3x_2 - 2x_3 \\ x_2 \text{ is free} \\ x_3 \text{ is free} \end{cases} \quad \text{parametric form}$$

$$\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 3x_2 - 2x_3 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \cdot \begin{bmatrix} 3 \\ 1 \\ 0 \end{bmatrix} + x_3 \cdot \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}$$

$\underline{w}_1$                        $\underline{w}_2$

Hence, a solution is any linear combination of  $\underline{w}_1$  and  $\underline{w}_2$ .

$\text{Span}\{\underline{w}_1, \underline{w}_2\}$

So, the solution set is a plane in  $\mathbb{R}^3$ .

Given a set of  $p$  vectors  $\{\underline{v}_1, \dots, \underline{v}_p\}$  in  $\mathbb{R}^n$ , the span of this set of vectors is the set of all possible linear combinations of the vectors in this set.

→  $\text{Span}\{\underline{v}_1, \dots, \underline{v}_p\}$  contains any vector  $\underline{y}$  that can be written as

$$\underline{y} = c_1 \cdot \underline{v}_1 + c_2 \cdot \underline{v}_2 + \dots + c_p \cdot \underline{v}_p.$$

Hence, solving an SUE boils down to investigating whether  $\underline{b}$  belongs to the Span of the columns of  $A$ .

We can see a linear combination of vectors as the product of a matrix ( $A$ ) and a vector ( $\underline{x}$ ).

Definition of  $A\underline{x}$ : linear combination of the columns of  $A$  with the entries of  $\underline{x}$  being the weights.

$$A\underline{x} = \begin{bmatrix} 1 & 2 & -1 \\ 0 & -5 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix} = 4 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 3 \cdot \begin{bmatrix} 2 \\ -5 \end{bmatrix} + 7 \cdot \begin{bmatrix} -1 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix}.$$

More efficient  $A\underline{x} = \begin{bmatrix} 1 & 2 & -1 \\ 0 & -5 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix} = \begin{bmatrix} 1 \cdot 4 + 2 \cdot 3 + (-1) \cdot 7 \\ 0 \cdot 4 + (-5) \cdot 3 + 3 \cdot 7 \end{bmatrix} = \begin{bmatrix} 3 \\ 6 \end{bmatrix}$

We need: # columns of  $A$  = # row/entries of  $\underline{x}$ .

Properties of the matrix-vector product: p. 65

Three things with the same solution set:

\* the SUE with augmented matrix  $[\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_n \ | \ \underline{b}]$

\* the vector equation  $x_1 \cdot \underline{a}_1 + x_2 \cdot \underline{a}_2 + \dots + x_n \cdot \underline{a}_n = \underline{b}$ .

\* the matrix equation  $[\underline{a}_1 \ \underline{a}_2 \ \dots \ \underline{a}_n] \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \underline{b}$ .  $A\underline{x} = \underline{b}$

Example:  $A = \begin{bmatrix} 1 & 5 & -3 \\ 0 & 1 & -2 \\ -2 & -10 & 6 \end{bmatrix}$

Q: Is  $A\underline{x} = \underline{b}$  consistent for every  $\underline{b} \in \mathbb{R}^3$ ? No!

$$\begin{bmatrix} 1 & 5 & -3 & | & b_1 \\ 0 & 1 & -2 & | & b_2 \\ -2 & -10 & 6 & | & b_3 \end{bmatrix} \sim \begin{bmatrix} 1 & 5 & -3 & | & b_1 \\ 0 & 1 & -2 & | & b_2 \\ 0 & 0 & 0 & | & b_3 + 2b_1 \end{bmatrix} \quad R_3: R_3 + 2 \cdot R_1$$

The SUE is consistent iff  $b_3 + 2b_1 = 0$ .

For example, the SUE is inconsistent if  $\underline{b} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$

consistent if  $\underline{b} = \begin{bmatrix} 1 \\ 2 \\ -2 \end{bmatrix}$

4 equivalent statements:

- \* The columns of  $A$  span  $\mathbb{R}^m$ .
- \* Each  $\underline{b} \in \mathbb{R}^m$  is a linear combination of the columns of  $A$ .
- \* For each  $\underline{b} \in \mathbb{R}^m$ ,  $A\underline{x} = \underline{b}$  has a solution.
- \*  $A$  has a pivot position in every row.  $\hookrightarrow$  (coefficient matrix!)