

EECS 16A Lecture 5.

Today:

- Proofs continued
- Matrices as linear transformations

- Tech Survey: Please fill out
- Study groups survey.
- Watch Fa2019 lectures
- Wednesday HW Party.

Thm: If the columns of A are linearly dependent, then $A\vec{x} = \vec{b}$ does not have a unique solution.

Last time: 3×3 case.

$n \times n$ case is identical.

Known:

$$A = \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \\ 1 & 1 & & 1 \end{bmatrix}$$

n columns. Rows don't matter

Columns are linearly dependent.

So some column \vec{a}_k

$$\vec{a}_k = c_1 \vec{a}_1 + c_2 \vec{a}_2 + \dots + c_{k-1} \vec{a}_{k-1} + c_{k+1} \vec{a}_{k+1} + \dots + c_n \vec{a}_n \quad (*)$$

\rightarrow Defn of linear dependence.

\rightarrow We want to connect this to the matrix A , and solutions to $A\vec{x} = \vec{b}$.

Want:

$A\vec{x} = \vec{b}$ does not have a unique solution.

i.e. either there is no solution or there are ≥ 2 solutions.

If possible let \vec{x}_*

be a unique solution.

We will show this is not possible.

How can we go from information about

individual columns to information about A ?

Can we rewrite the equation (*) in terms of A ?

To write a column \vec{a}_k in terms of A ,

we can say

$$\underbrace{\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \end{bmatrix}}_A \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{bmatrix} \leftarrow \text{kth position} = \vec{a}_k$$

$$\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = \vec{a}_1$$

$$\begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_n \end{bmatrix} \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = \vec{a}_1 c_1 + \dots + \vec{a}_n c_n$$

$$\begin{bmatrix} A \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_{k-1} \\ 0 \\ c_{k+1} \\ \vdots \\ c_n \end{bmatrix} = c_1 \vec{a}_1 + c_2 \vec{a}_2 + \dots + c_{k-1} \vec{a}_{k-1} + \underbrace{c_{k+1} \vec{a}_{k+1}}_{\text{no } \vec{a}_k} + \dots + c_n \vec{a}_n$$

kth position

So we can rewrite (*) as

$$A \cdot \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = A \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ 0 \\ c_n \end{bmatrix} \rightarrow A \left(\begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{bmatrix} - \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{bmatrix} \right) = 0$$

or equivalently

$$A \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_{k-1} \\ -1 \\ c_{k+1} \\ \vdots \\ c_n \end{bmatrix} = 0$$

kth position

call this vector \vec{w} .

At least one entry of \vec{w} is non-zero. So $\vec{w} \neq \vec{0}$.

Is $\vec{w} = \vec{0}$? \rightarrow NO.

$$\vec{0} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Now: $A \cdot \vec{w} = \vec{0}$. (**)

But \vec{w} is not the zero vector.

\vec{w} has a non zero entry at the kth position.

A is not a zero matrix.

Interesting....

How can this help us? We want to show that if \vec{x}_* is a solution, there must be another solution as well.

We know: $A\vec{x}_* = \vec{b}$.

$$\Rightarrow A\vec{x}_* + \vec{0} = \vec{b} \quad \text{Eq. (*)}$$

$$\Rightarrow A\vec{x}_* + A\vec{w} = \vec{b}$$

$$\Rightarrow A(\vec{x}_* + \vec{w}) = \vec{b} \quad \vec{x}_* + \vec{w} + \vec{x}_*$$

$$\Rightarrow \vec{x}_* + \vec{w} \text{ is also a solution!}$$

Now \vec{x}_* is not the only solution!

We can say:

$\vec{x}_* + \vec{w} + \vec{w} + \vec{w} + \vec{w} + \dots$ is a solⁿ

□ QED

Quod et demonstrandum.

So if the columns of A corresponding to your tomography machine are linearly dependent, then you cannot recover an accurate image.

Thm: If $A\vec{x} = \vec{b}$ has two or more solutions, then the columns of A are linearly dependant.

Note: This is related to the previous theorem, but it is very different!

$$A = \begin{bmatrix} 1 & 2 \\ 1 & 2 \end{bmatrix}$$

$$\vec{x}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\vec{x}_2 = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$$

Proof:

Known:

$A\vec{x} = \vec{b}$ has two distinct solutions, say \vec{x}_1 and \vec{x}_2 .
 $\vec{x}_1 \neq \vec{x}_2$

$$\Rightarrow \begin{cases} A\vec{x}_1 = \vec{b} \\ A\vec{x}_2 = \vec{b} \end{cases} \quad \begin{aligned} A\vec{x}_1 - A\vec{x}_2 &= \vec{b} - \vec{b} = \vec{0} \end{aligned}$$

$$A\vec{x}_1 - A\vec{x}_2 = \vec{0}$$

$$A(\vec{x}_1 - \vec{x}_2) = \vec{0}$$

Let \vec{y} have entries

$$\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} \quad \boxed{\vec{y} \neq \vec{0}}$$

To show:

Columns of A are linearly dependant.

Mathematically, some \vec{a}_k

$$\vec{a}_k = c_1 \vec{a}_1 + c_2 \vec{a}_2 + \dots + c_n \vec{a}_n$$

↑
no \vec{a}_k

$$\boxed{A\vec{y} = \vec{0}}$$

Can this give me information about the columns of A ?

$$\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \dots & \vec{a}_n \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \vec{0}$$

$y_1, y_2, \dots, y_n \in \mathbb{R}$
scalars.

$$\Rightarrow y_1 \vec{a}_1 + y_2 \vec{a}_2 + \dots + y_n \vec{a}_n = \vec{0}$$

Because $\vec{y} \neq \vec{0}$, at least one of the entries is not equal to zero. Let us call this y_l .

$$\Rightarrow y_l \vec{a}_l = -y_1 \vec{a}_1 - \dots - y_{l-1} \vec{a}_{l-1} - y_{l+1} \vec{a}_{l+1} - \dots - y_n \vec{a}_n$$

Because $y_l \neq 0$

$$\Rightarrow \vec{a}_l = \frac{-y_1 \vec{a}_1}{y_l} + \dots + \frac{-y_n \vec{a}_n}{y_l}$$

\vec{a}_l is a lin comb of other columns. \square QED.

Linear dependence Alternate definition.

$\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$ are linearly dependent if there exist constants c_1, c_2, \dots, c_n not all zero such that $c_1\vec{a}_1 + c_2\vec{a}_2 + \dots + c_n\vec{a}_n = \vec{0}$.

↳ Equivalent to the other definition.

If $c_i \neq 0$, then:

$$c_1\vec{a}_1 + c_2\vec{a}_2 + \dots + c_{i-1}\vec{a}_{i-1} + c_{i+1}\vec{a}_{i+1} + \dots + c_n\vec{a}_n = -c_i\vec{a}_i$$

$$\Rightarrow \frac{-c_1}{c_i}\vec{a}_1 - \frac{c_2}{c_i}\vec{a}_2 - \dots - \frac{c_n}{c_i}\vec{a}_n = \vec{a}_i$$

i.e. \vec{a}_i is represented as a linear combination of other columns.

$\{\vec{a}_1, \vec{a}_2, \vec{0}\}$

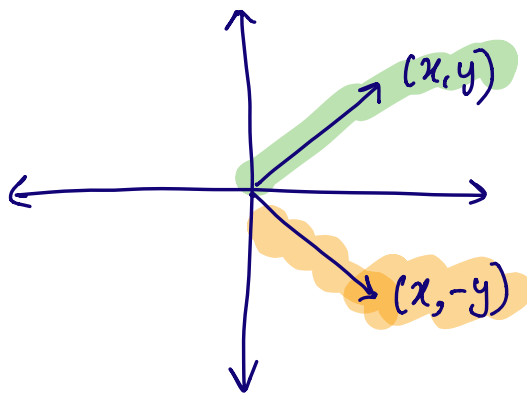
① Def ①: Write any column as a linear combination of other columns.

$$\vec{0} = 0 \cdot \vec{a}_1 + 0 \cdot \vec{a}_2$$

Matrix - vector multiplication is useful for representing systems of equations. But more broadly, matrices are "operators" that transform a vector into another vector.

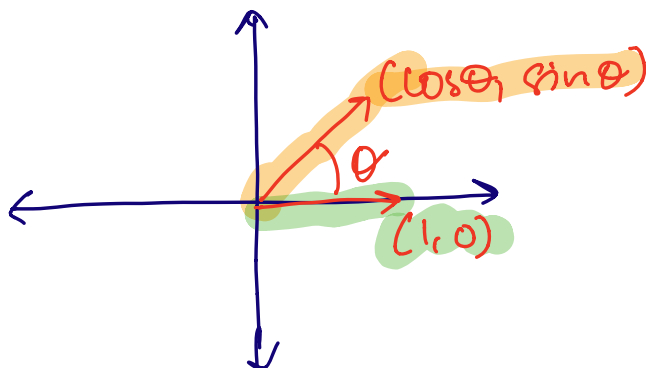
→ Discussion section (more examples)

eg. $\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} x \\ -y \end{bmatrix}$ Reflection matrix.



In particular, matrices are linear transformations.

eg. $\begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}$



Linear transformation

f : is a linear transformation if:

$$f(\vec{x} + \vec{y}) = f(\vec{x}) + f(\vec{y})$$

$$f(\alpha \cdot \vec{x}) = \alpha \cdot f(\vec{x}) \quad \alpha \in \mathbb{R} \text{ (scalar)}$$

e.g. $f(\vec{x}) = 2\vec{x}$ is linear.

$f(x) = x^2$ is not linear.

$$A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$$

$$A(\alpha \cdot \vec{x}) = \alpha \cdot (A\vec{x})$$

Check: Do matrix-vector mult. satisfy?

Vectors are used to represent "state" of a system

e.g. the "state" of a car $\vec{s} = \begin{bmatrix} x \\ y \\ v \end{bmatrix}$ $\begin{matrix} \rightarrow x \text{ position} \\ \rightarrow y \text{ position} \\ \rightarrow \text{velocity} \end{matrix}$

If this is changing with time

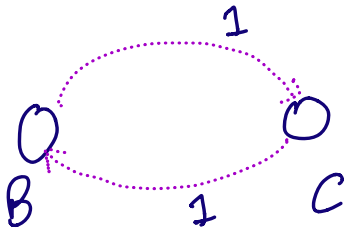
$$\vec{s}(t) = \begin{bmatrix} x(t) \\ y(t) \\ v(t) \end{bmatrix}$$

A system of reservoirs and pumps.



A, B, C are tanks of water

What is the "state" of such a system?




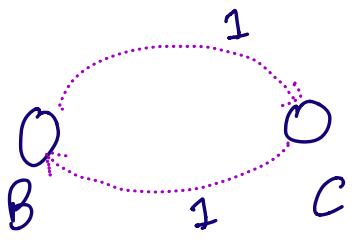
$\vec{x}(t) = \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix}$ → Water in A at time t.

$\vec{x}(1) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$

Say we interconnect these using some pumps.

Pumps run every time the clock ticks (e.g. every second)

 : Every time the pump runs all water from A moves back into A-

 : Everytime the pump runs all water_{fraction(1)} from B goes into C and all water (fraction 1) from C goes into B.

How can I represent this mathematically?

$$x_A(t+1) = x_A(t).$$

$$x_B(t+1) = x_C(t)$$

$$x_C(t+1) = x_B(t).$$

System of equations that describes evolution of the state.

Write in matrix form:

$$\underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}}_Q \begin{bmatrix} x_A(t) \\ x_B(t) \\ x_C(t) \end{bmatrix} = \begin{bmatrix} x_A(t+1) \\ x_B(t+1) \\ x_C(t+1) \end{bmatrix}$$

$\vec{x}(t)$ $\vec{x}(t+1)$

$$Q \cdot \vec{x}(t) = \vec{x}(t+1)$$

What happens when we run the pumps twice? **Guess?**

$$\boxed{Q} \cdot \vec{x}(t+1) = \underline{\vec{x}(t+2)}$$

$$\boxed{Q} \cdot (\boxed{Q} \cdot \vec{x}(t)) = \boxed{\vec{x}(t+2)}$$

What is $\boxed{Q \cdot Q}$?

Matrix - Matrix multiplication

2x2

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

A \vec{b}_1 B \vec{b}_2

Two vectors stacked.

$$= \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} \vec{b}_1 & \vec{b}_2 \end{bmatrix}$$

A

$A \cdot \vec{b}_1 = \text{vector}$
 $A \vec{b}_2 = \text{vector}$

$$= \begin{bmatrix} A \vec{b}_1 & A \vec{b}_2 \end{bmatrix} = \begin{bmatrix} a_{11} b_{11} + a_{12} b_{21} & a_{11} b_{12} + a_{12} b_{22} \\ a_{21} b_{11} + a_{22} b_{21} & a_{21} b_{12} + a_{22} b_{22} \end{bmatrix}$$

2x2 matrix.
2 stacked vectors.

In general: $A B$

$$= A \cdot \begin{bmatrix} \vec{b}_1 & \vec{b}_2 & \dots & \vec{b}_n \end{bmatrix}$$

$$= \begin{bmatrix} A \vec{b}_1 & A \vec{b}_2 & \dots & A \vec{b}_n \\ | & | & & | \end{bmatrix}$$

↪ Go back to
example.