

Matrices as linear transformations

$T: \mathbb{R}^m \rightarrow \mathbb{R}^m$ is a linear transformation

$$T(\alpha_1 \vec{v}_1 + \alpha_2 \vec{v}_2) = \alpha_1 T(\vec{v}_1) + \alpha_2 T(\vec{v}_2)$$

Simplest (only) example: M is $m \times m$ matrix

$$T(\vec{x}) = M\vec{x}$$

Matrix multiplication =
composition of linear transformations

$$\mathbb{R}^n \xrightarrow{T} \mathbb{R}^m \xrightarrow{S} \mathbb{R}^p$$

$\underbrace{\hspace{10em}}_{S \circ T}$

T is represented by M

S is represented by N

$$\boxed{NM}$$

matrix mult.

$\Rightarrow S \circ T$ is represented by

\hookrightarrow Theorem.

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Fact: Every choice of basis yields a different matrix representing the linear transformation.

Matrix Inverses: M is $(n \times n)$ square

and there another N ($n \times n$) with

$$MN = \begin{bmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{bmatrix} = I \Rightarrow N \text{ is}$$

called the inverse of M and written $N = M^{-1}$

$$\Rightarrow M^{-1}M = M M^{-1} = I$$

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So me matrices have inverses

$$\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

so $\begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1}$

$$\begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}^{-1} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \quad \begin{pmatrix} 2 & 4 \\ 1 & 2 \end{pmatrix}$$

Some matrices don't eg

SAY $\begin{pmatrix} 2 & 4 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

$$\begin{cases} 2a + 4c = 1 \\ a + 2c = 0 \end{cases} \rightarrow \begin{cases} a + 2c = 1/2 \\ a + 2c = 0 \end{cases} \rightarrow \text{contradiction}$$

If M has an inverse it is called

invertible or nonsingular

it is called

If M doesn't have an inverse

non-invertible or singular.



Theorem: M is $n \times n$, The following are equivalent

- (1) M is invertible
- (2) $\text{rank}(M) = n$
- (3) $\text{null}(M) = \{0\}$
- (4) $\det(M) \neq 0$
- (5) columns of M form a basis for \mathbb{R}^n
- (6) rows of M form a basis for \mathbb{R}^n
- (7) ... etc.

Solving linear equations

$$A_{11}x_1 + \dots + A_{1n}x_n = b_1$$

$$A_{21}x_1 + \dots + A_{2n}x_n = b_2$$

⋮

$$A_{m1}x_1 + \dots + A_{mn}x_n = b_m$$

n unknowns

m eq.

More Succinctly

$$A \vec{x} = \vec{b}$$

A is $m \times n$

\vec{x} is $n \times 1$

\vec{b} is $m \times 1$

1st question! Is there any soln?

a unique soln? many soln?

Simplest case A is $n \times n$ and invertible

$$A\vec{x} = \vec{b}$$

$$A^{-1}A\vec{x} = A^{-1}\vec{b}$$

$$I\vec{x} = A^{-1}\vec{b}$$

$$\vec{x} = A^{-1}\vec{b}$$

unique soln

General case - geometrically.

$$\vec{b} = A\vec{x} = \begin{bmatrix} \vec{c}_1 & \vec{c}_2 & \dots & \vec{c}_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$= x_1 \vec{c}_1 + x_2 \vec{c}_2 + \dots + x_n \vec{c}_n$$

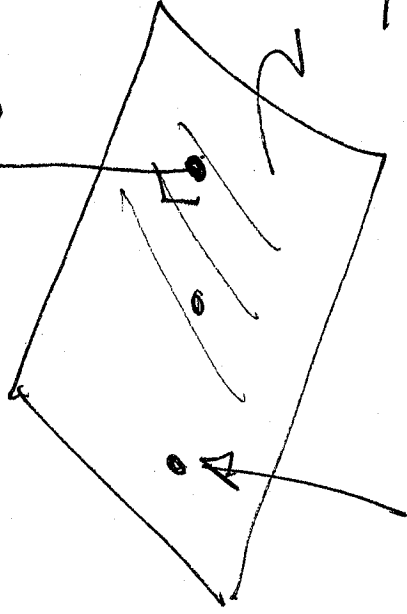
So existence soln is same as whether \vec{b} is a linear comb. of the col of A .

OR whether \vec{b} is in the col space of A

OR \vec{b} is in the range of the linear

transformation $T\vec{x} = A\vec{x}$

$$T = Ax^2 \rightarrow$$



range T
 $= \text{col}(A)$

$\vec{b} \leftarrow$ a soln
(many)

$$\text{rank}(A) = 2$$

How do you solve linear equations? — Row reduction
= Gaussian elimination = LU decomposition

Moves

in row reduction

put in for

→ (1) $C \cdot \text{Row}_i + \text{Row}_j \rightarrow \text{Row}_j$

(2) swap rows

(3) rescale rows

(3) $C \text{ Row}_i \rightarrow \text{Row}_i$

Row Reduction

Al 12

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} \uparrow A$$

$$\begin{array}{l} -2R_1 + R_2 \rightarrow R_2 \\ -4R_1 + R_3 \rightarrow R_3 \\ \hline -3R_1 + R_4 \rightarrow R_4 \end{array}$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 3 & 5 & 5 \\ 0 & 4 & 6 & 8 \end{bmatrix}$$

$$\begin{array}{l} -3R_2 + R_3 \rightarrow R_3 \\ \hline -4R_2 + R_4 \rightarrow R_4 \end{array}$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 2 & 4 \end{bmatrix}$$

$$-R_3 + R_4 \rightarrow R_4$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 2 \end{bmatrix} \uparrow U$$

We will see that this implies

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} =$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 4 & 3 & 1 & 0 \\ 6 & 4 & 1 & 1 \end{bmatrix} \uparrow L = \text{lower}$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 2 \end{bmatrix} \uparrow U = \text{upper}$$

How do we use this to solve

$$LUx = A\vec{x} = \vec{b}$$

split into two triangular systems

$$L\vec{y} = \vec{b} \Rightarrow \text{has solution } \vec{y}$$

$$\text{and } U\vec{x} = \vec{y}$$

$$\begin{aligned} \text{Then } A\vec{x} &= LU\vec{x} = L(U\vec{x}) \\ &= L\vec{y} = \vec{b} \end{aligned}$$

so \vec{x} solves the system.

How do we use this to solve $A\vec{x}=\vec{b}$

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Secret is that equations like $L\vec{y}=\vec{b}$ and $U\vec{x}=\vec{y}$ are easy to solve by back and forward substitution.

eg Solve First $L\vec{y}=\vec{b}$ =

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 4 & 3 & 1 & 0 \\ 3 & 4 & 1 & 1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{bmatrix} = \begin{bmatrix} -1 \\ 2 \\ 3 \\ 1 \end{bmatrix}$$

as equations

$$\begin{aligned} y_1 &= -1 \implies y_2 = 2 - 2y_1 = 4 \\ 2y_1 + y_2 &= 2 \implies y_3 = 3 - 4y_1 - 3y_2 = 3 + 4 - 12 = -5 \\ 4y_1 + 3y_2 + y_3 &= 3 \implies y_4 = 1 - 3y_1 - 4y_2 - y_3 = 1 + 3 - 16 + 5 = -7 \end{aligned}$$

FORWARD SUBSTITUTION

Now we solve
 $U\vec{x} = \vec{y}$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} -1 \\ 4 \\ 11 \\ -22 \end{bmatrix}$$

$$\begin{aligned} 2x_1 + x_2 + x_3 &= -1 & \rightarrow x_1 &= \dots \\ x_2 + x_3 + x_4 &= 4 & \rightarrow x_2 &= 4 - x_3 - x_4 \\ 2x_3 + 2x_4 &= 11 & \rightarrow x_3 &= \frac{1}{2} [11 - 2x_4] = \frac{1}{2} [11 + 22] \\ 2x_4 &= -22 & \rightarrow x_4 &= -11 \end{aligned}$$

Back Substitution

Then $LU(\vec{x}) = L\vec{y} = \vec{b}$ so \vec{x} is

Sought solution

This is useful to solve many

equations

$$A\vec{x} = \vec{b}_1$$

$$A\vec{x} = \vec{b}_2$$

$$\vdots$$

$$A\vec{x} = \vec{b}_k$$

many eq.

Compute $A = LU$ once and for all
and then each equation is fast.

• Sometimes you need to permute

rows to compute LU.

• Numerical considerations force

Row and column swapping

(pivots)