

On to chapter 2 - solving equations. (Good time to remind you: READ THE BOOK!)

Talked a lot about subspaces in chapter 1.

Can represent them as span of vectors: pick $\vec{v}_1, \dots, \vec{v}_k$

consider all linear combinations $\{ c_1 \vec{v}_1 + \dots + c_k \vec{v}_k \mid c_i \in \mathbb{R} \}$

Another way to represent subspace:

set $\{ \vec{v} \mid A \cdot \vec{v} = \underline{0} \}$ A : linear transformation
($m \times n$ matrix)

(check that solutions are well behaved under addition, scalar mult. Or think of

$$A = \begin{bmatrix} | & & | \\ A(\vec{e}_1) & \dots & A(\vec{e}_n) \\ | & & | \end{bmatrix}$$

then in review for exam we showed

$\{ \vec{v} \mid A(\vec{e}_i) \cdot \vec{v} = 0 \}$ is a subspace.

the above set is the intersection of these for $i=1, \dots, m$.

Goal: Given ways of moving from one description to the other.

Key tool is row reduction / Gaussian elimination.

Solves more generally equations $A \cdot \vec{x} = \vec{b}$, $\vec{b} \in \mathbb{R}^m$.

Written out $A \cdot \vec{x} = \begin{bmatrix} a_{11} & a_{12} & \dots \\ \vdots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} a_{11}x_1 + a_{12}x_2 + \dots \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + \dots \end{bmatrix}$

want it $= \vec{b} = \begin{bmatrix} b_1 \\ \vdots \end{bmatrix}$ so must agree at each component

so get system of equations

$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + \dots &= b_1 \\ &\vdots \end{aligned}$$

Give an algorithm for solving such a system.

these methods will allow us to

- solve linear / non-linear equations \rightsquigarrow approx. the latter with tangent hyperplanes
- develop a more precise language for subspaces (dimension, etc.)
to say when solutions exist, and if so, how many?
- in Ch. 3, allows us to solve optimization problems in multiple variables.

Given system of equations, like

$$x_1 - x_2 = 4$$

$$x_1 + 2x_2 = 3$$

\leftrightarrow

$$\begin{bmatrix} 1 & -1 \\ 1 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$

\nearrow
subtract top from bottom:

$$3x_2 = -3 \quad \text{so } x_2 = -1$$

(if being more careful to record steps:

$$\text{original system } \Leftrightarrow \text{system } \begin{cases} x_1 - x_2 = 4 \\ 3x_2 = -3 \end{cases}$$

mult. by scalar $\frac{1}{3}$ in second equation:

$$\Leftrightarrow \begin{cases} x_1 - x_2 = 4 \\ x_2 = -1 \end{cases}$$

now easy to see $x_1 = 3$.

(add $x_2 = -1$ back to $x_1 - x_2 = 4$)

Now just axiomatize this process.

Shorthand, act on matrix and on solution vector

$$: \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$

$$\left[\begin{array}{cc|c} 1 & -1 & 4 \\ 1 & 2 & 3 \end{array} \right]$$

$$\downarrow \left[\begin{array}{cc|c} 1 & -1 & 4 \\ 0 & 3 & -3 \end{array} \right]$$

$$\downarrow \left[\begin{array}{cc|c} 1 & -1 & 4 \\ 0 & 1 & -1 \end{array} \right]$$

$$\downarrow \left[\begin{array}{cc|c} 1 & 0 & 3 \\ 0 & 1 & -1 \end{array} \right]$$

Call such a matrix associated to a system $A\underline{x} = \underline{b}$, written

$[A | \underline{b}]$, an augmented matrix. We have 3 basic operations on them (suggested by our previous example)

(1) multiply by a non-zero scalar in row.

(2) Add a multiple of one row to another row.

(3) Exchange rows. \leftarrow not necessary for solving equations, just puts augmented matrix in more prescribed form.

Theorem: These three operations preserve the set of solutions to the associated system of equations.

pf: Clear that (1) and (3) do this. For (2), if

We take
$$\left(\begin{array}{c|c} \sim & \sim \\ \sim & \sim \end{array} \right) \begin{array}{l} r_1 \\ r_2 \end{array} \longrightarrow \left(\begin{array}{c|c} \sim & \sim \\ \sim & \sim \end{array} \right) \begin{array}{l} r_1 \\ c r_1 + r_2 \end{array}$$

c : const. mult.

if \underline{x} is sol'n to first system, then sol'n for second.

And map is reversible. If we let $r_1' = r_1$, $r_2' = c r_1 + r_2$

then map $r_1' \mapsto r_1$, $r_2' \mapsto r_2' - c r_1'$ to go back.

Claim: Any matrix (augmented) can be put into form:

$$\begin{pmatrix} 0 & 0 & 1 & 0 & 0 & 0 & * \\ 0 & 0 & 0 & 1 & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 1 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

(1) every row has first non-zero entry equal to 1 (pivot)

(2) the pivots move down to the right as we go down rows

(3) pivot is only non-zero entry in its column.

(4) rows of zeros, if any, are at bottom

This is called "echelon form" or "row echelon form"

(Theorem 2.1.7 in H-H).

pf of claim: Give an algorithm: ⁽¹⁾ Searching left to right, find first non-zero column. Pick a row with non-zero entry in ~~the~~ this column, move it to the top, scale so the first entry is 1.

Running example:

$$\begin{bmatrix} 5 & 10 & 3 & 2 \\ 2 & 4 & 8 & 2 \\ 1 & 2 & 4 & 1 \end{bmatrix}$$

Pick third row, since we don't have to scale:

$$\begin{bmatrix} 1 & 2 & 4 & 1 \\ 2 & 4 & 8 & 2 \\ 5 & 10 & 3 & 2 \end{bmatrix}$$

(2) Use multiples of new first row to make 0's below the new pivot 1.

$$\begin{bmatrix} 1 & 2 & 4 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & -17 & -3 \end{bmatrix}$$

(3) ~~Choose~~ Find next column with non-zero entries below the first row.

~~Choose~~ Pick a row with non-zero entry below first row in this column. Move it to the next (i.e. second) row.

this column.

Scale the first entry to be 1.

$$\begin{bmatrix} 1 & 2 & 4 & 1 \\ 0 & 0 & 1 & 3/17 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Discuss relation to system of eqns.

(4) Just as in step (2), make all other entries in pivot column = 0.

(5) Repeat until in echelon form. (In our example, done)

$$\begin{bmatrix} 1 & 2 & 0 & 5/17 \\ 0 & 0 & 1 & 3/17 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$