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Jacobs University, Fall 2022

6. Matrices

6.2 Solving systems of linear equations

Topic 6.2.B: Solution space of systems of linear equations

Taking, as an example, the following system

$$\begin{aligned} 2x_1 + 2x_2 &= 4 \\ 4x_1 + 4x_2 &= 8 \end{aligned} \Rightarrow \begin{pmatrix} 2 & 2 \\ 4 & 4 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 4 \\ 8 \end{pmatrix}$$

$$\Rightarrow \left(\begin{array}{cc|c} 2 & 2 & 4 \\ 4 & 4 & 8 \end{array} \right) \xrightarrow[\substack{R_2 - 2R_1 \Rightarrow R_2 \\ R_1/2 \Rightarrow R_1}]{\Rightarrow} \left(\begin{array}{cc|c} 1 & 1 & 2 \\ 0 & 0 & 0 \end{array} \right)$$

focus on the diagonal entries and entries $\neq 0$

Def.: The first non-zero entry of a row is called a pivot provided there is no other pivot in that column.

Ideally, you want to do row operations until all pivots are on the diagonal and are 1.

In general, if not all pivots exist / not all columns have pivots, we expect that a solution only exists for some specific vectors b and the given matrix A (i.e. for some b there may not be a solution).

Sometimes reordering the rows may help to prevent pivots along diagonal to become zero. (more details, e.g., in Numerical Methods).

The matrix above, $\begin{pmatrix} 2 & 2 \\ 4 & 4 \end{pmatrix}$, will have multiple solutions for some b , such as $b = \begin{pmatrix} 4 \\ 8 \end{pmatrix}$, but

no solutions for some other, e.g. $b = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$,

as the augmented system will read
 $\left(\begin{array}{cc|c} 1 & 1 & \frac{1}{2} \\ 0 & 0 & -3 \end{array} \right)$ with the last row a contradiction $0 \cdot x_1 + 0 \cdot x_2 = -3$

We call a system **overdetermined**, when we have more **(linearly independent) equations** than **unknowns**. Such systems do not have an exact solution. It is

E.g.:
$$\begin{pmatrix} 2 & 3 \\ 1 & -1 \\ 6 & 5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

more rows than columns and rows are **linearly independent**, meaning one cannot construct (linearly) one from the other, e.g. $\begin{pmatrix} 2 \\ 3 \end{pmatrix} \neq \lambda \begin{pmatrix} 1 \\ -1 \end{pmatrix} + \mu \begin{pmatrix} 6 \\ 5 \end{pmatrix} \forall \lambda, \mu \in \mathbb{R}$

The solution to just the first two rows is

$$x_1 = \frac{7}{5}, \quad x_2 = -\frac{3}{5}$$

but this does not solve the last row

$$\begin{aligned} 6 \cdot x_1 + 5 \cdot x_2 &= \frac{42}{5} - \frac{15}{5} \\ &= \frac{27}{5} \neq 3 \end{aligned}$$

We call a system **underdetermined**, when we have more **unknowns** than **(linearly independent) equations**. Such systems have multiple solutions. They are **underconstrained**.

E.g.:

$$\begin{pmatrix} 2 & 4 & 2 \\ 1 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix}$$

more columns than rows, i.e. more unknowns than equations. Rows are also linearly independent; if they were linearly dependent, there would be even fewer equations containing additional information.

The solutions are given by: $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{1}{2} \\ 0 \end{pmatrix} + \lambda \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \forall \lambda \in \mathbb{R}$

i.e. the solutions are on a line in 3D.

(see if you can find this as well)

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Def.: Let $A \in M(n \times m)$. Then

$$\text{Ker } A = \{ v \in \mathbb{R}^m : Av = 0 \}$$

is kernel or null space or solution space of the homogeneous equation.

This is a vector space (check yourself)

$$\text{Range } A = \{ Ax : x \in \mathbb{R}^n \}$$

rank A is given by the # of pivots ($\neq 0$) after Gaussian elimination / using row operations

to eliminate all entries except those on the diagonal).

Let $v_1, \dots, v_m \in \mathbb{R}^n$, $b \in \mathbb{R}^n$

We can then write

$$b = x_1 v_1 + \dots + x_m v_m$$

↑ unknown ↙ vector

which is the same as writing

$$Ax = b \quad \text{with} \quad A = \begin{pmatrix} | & & | \\ v_1 & \dots & v_m \\ | & & | \end{pmatrix}, \quad A \in M(n \times m)$$

$\in \mathbb{R}^n$ $\in \mathbb{R}^n$

We can make the following observations:

- Columns without a pivot ($\neq 0$) do not contribute to $\text{Range } A$ (i.e. the respective entries of x for $Ax = b$ do not impact the set of choices for b)

→ to select a basis for the column space (= $\text{Range } A$), use Gaussian elimination to find columns with pivots, then choose original columns of A as basis vectors.

◦ Columns without a pivot contribute to a basis of $\text{Ker } A$ (i.e. the vector space of solutions to $Av = 0$).

◦ The vectors for the span of $\text{Ker } A$ that come from, e.g., Gaussian elimination are linearly independent by construction and are therefore a basis of $\text{Ker } A$.

Rank-nullity theorem

Let $A \in M(n \times m)$, then

$$\underbrace{\dim \text{Ker } A}_{\text{nullity } A} + \underbrace{\dim \text{Range } A}_{\text{rank } A} = m$$

↑
number of
columns of A

(where \dim gives the number of basis vectors of the space)

Columns with pivots contribute to $\dim \text{Range } A$, columns without pivots contribute to $\dim \text{Ker } A$.

Consequences: • $Ax = b$ has a solution for every $b \in \mathbb{R}^n$ iff $\dim \text{Range } A = n$

◦ $Ax = b$ has a unique (i.e. exactly just one) solution if $b \in \text{Range } A$ and $\dim \text{Ker } A = 0$

◦ $Ax = b$ has a unique solution for every $b \in \mathbb{R}^n$ iff $n = m = \text{rank } A$ (i.e. a square matrix $\in M(n \times n)$ and all columns/rows are linearly independent)

In the last case we say A is non-singular or invertible and we write:

$$x = A^{-1}b, \text{ where } A^{-1} \text{ is the inverse of } A$$