

Gaussian elimination revisited:

$$Ax = b \quad \text{with} \quad A = \begin{pmatrix} 2 & -2 & -4 & -1 & 5 \\ 3 & -3 & -6 & 3 & 12 \\ 3 & -3 & -6 & 2 & 11 \end{pmatrix}, \quad b = \begin{pmatrix} 5 \\ 21 \\ 18 \end{pmatrix}$$

augmented matrix

$$\left(\begin{array}{ccccc|c} 2 & -2 & -4 & -1 & 5 & 5 \\ 3 & -3 & -6 & 3 & 12 & 21 \\ 3 & -3 & -6 & 2 & 11 & 18 \end{array} \right) \xrightarrow[\substack{R2 \rightarrow R1 \\ R1 \rightarrow R2}]{\substack{R2 \rightarrow R1 \\ R1 \rightarrow R2}} \left(\begin{array}{ccccc|c} 1 & -1 & -2 & 1 & 4 & 7 \\ 2 & -2 & -4 & -1 & 5 & 5 \\ 3 & -3 & -6 & 2 & 11 & 18 \end{array} \right)$$

$$\begin{array}{l} -2R1 + R2 \rightarrow R2 \\ -3R1 + R3 \rightarrow R3 \end{array} \rightarrow \left(\begin{array}{ccccc|c} 1 & -1 & -2 & 1 & 4 & 7 \\ 0 & 0 & 0 & -3 & -3 & -9 \\ 0 & 0 & 0 & -1 & -1 & -3 \end{array} \right) \xrightarrow[\substack{R2 + R1 \rightarrow R1 \\ R2 \rightarrow R2 \\ R2 + R3 \rightarrow R3}]{\substack{R2 + R1 \rightarrow R1 \\ R2 \rightarrow R2 \\ R2 + R3 \rightarrow R3}} \left(\begin{array}{ccccc|c} 1 & -1 & -2 & 0 & 3 & 4 \\ 0 & 0 & 0 & 1 & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right)$$

$$\left(\begin{array}{ccccc|c} 1 & -1 & -2 & 0 & 3 & 4 \\ 0 & 0 & 0 & 1 & 1 & 3 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right)$$

◆ pivots
◆ "missing pivots"

Solution is

$$x = \begin{pmatrix} 4 \\ 0 \\ 0 \\ 3 \\ 0 \end{pmatrix} + \lambda_1 \begin{pmatrix} -1 \\ -1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + \lambda_2 \begin{pmatrix} -2 \\ 0 \\ -1 \\ 0 \\ 0 \end{pmatrix} + \lambda_3 \begin{pmatrix} 3 \\ 0 \\ 0 \\ 1 \\ -1 \end{pmatrix}$$

In general, if not all columns have pivots, expect that a solution exists only for some vectors b and the given matrix A .

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

• Let $A \in M(n \times m)$. Then

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

"kernel" or "null space" or solution space of homogeneous equation. (It's a vector space!)

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

rank A : # of pivots after Gaussian elimination

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

Goal: write $b = x_1 v_1 + \dots + x_m v_m$

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

Observations: • columns without a pivot do not contribute to $\text{Range } A$.

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

• Columns without a pivot contribute to a basis for $\text{Ker } A$

vector space of solutions to $Av = 0$.

• the vectors for the span of $\text{Ker } A$ that come from above are l.i. by construction, so are a basis.

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

"rank-nullity theorem"

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

• $Ax = b$ has a unique 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

$$n = n = \text{rank } A$$

In that case, we say A is non-singular or invertible and we write

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

$$= A^{-1}(b_1 e_1 + b_2 e_2 + \dots + b_n e_n)$$

$$e_i = (0, \dots, \underset{\substack{\uparrow \\ i}}{1}, \dots)$$

$$= b_1 A^{-1}e_1 + \dots + b_n A^{-1}e_n$$

This gives rise to the following procedure for computing A^{-1} :

$$(A | I) \rightarrow \dots \rightarrow (I | A^{-1})$$

$$| 0 \quad \frac{1}{2} \quad -\frac{1}{2} \quad |$$

E.g. $A = \begin{pmatrix} 1 & 0 & 1 \\ 2 & \frac{1}{2} & 1 \end{pmatrix}$

$$\left(\begin{array}{ccc|ccc} 0 & \frac{1}{2} & -\frac{1}{2} & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 2 & \frac{1}{2} & 1 & 0 & 0 & 1 \end{array} \right) \xrightarrow{\substack{R_2 \leftrightarrow R_1 \\ 2R_1 \rightarrow R_2 \\ -2R_2 + R_3 \rightarrow R_3}} \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 2 & 0 & 0 \\ 0 & \frac{1}{2} & -1 & 0 & -2 & 1 \end{array} \right)$$

$$\xrightarrow{\substack{-\frac{1}{2}R_2 + R_3 \rightarrow R_3 \\ -2R_3 \rightarrow R_3}} \left(\begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & -1 & 2 & 0 & 0 \\ 0 & 0 & -\frac{1}{2} & -1 & -2 & 1 \end{array} \right) \xrightarrow{\substack{R_3 + R_2 \rightarrow R_3 \\ -R_3 + R_1 \rightarrow R_1}} \left(\begin{array}{ccc|ccc} 1 & 0 & 0 & -2 & -3 & 2 \\ 0 & 1 & 0 & 4 & 4 & -2 \\ 0 & 0 & 1 & 2 & 4 & -2 \end{array} \right)$$

$\underbrace{\hspace{10em}}_I \quad \underbrace{\hspace{10em}}_{A^{-1}}$

Check: $\begin{pmatrix} 0 & \frac{1}{2} & -\frac{1}{2} \\ 1 & 0 & 1 \\ 2 & \frac{1}{2} & 1 \end{pmatrix} \begin{pmatrix} -2 & -3 & 2 \\ 4 & 4 & -2 \\ 2 & 4 & -2 \end{pmatrix} = \begin{pmatrix} 0 \cdot (-2) + 4 \cdot \frac{1}{2} - 2 \cdot \frac{1}{2} & \dots & \dots \\ 1 \cdot (-2) + 0 \cdot 4 + 1 \cdot 2 & \dots & \dots \\ \vdots & \dots & \dots \end{pmatrix} = \begin{pmatrix} 1 & \dots & \dots \\ 0 & \dots & \dots \\ \dots & \dots & \dots \end{pmatrix} = I$