

4 (Systems of) linear equations

Definition 1 (linear equation). A linear equation is an equation in which each term is either a constant or the product of a constant and (the first power of) a single variable. If $n \in \mathbb{N}$, $a_1, a_2, \dots, a_n, b \in \mathbb{R}$, and x_1, x_2, \dots, x_n are unknowns or variables, then a linear equation is given by

$$\sum_{k=1}^n a_k x_k = b \Leftrightarrow a_1 x_1 + a_2 x_2 + \dots + a_n x_n = b.$$

The numbers a_1, a_2, \dots, a_n are called coefficients and b is the right-hand side. A solution is an n -tuple $(x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ that fulfills the equation. If $b = 0$, the equation is called homogenous otherwise inhomogenous.

Definition 2 (system of linear equations). Let $m, n \in \mathbb{N}$. A system of linear equations with m equations and n variables is given by

$$\begin{cases} a_{1,1}x_1 + a_{1,2}x_2 + \dots + a_{1,n}x_n & = b_1 \\ a_{2,1}x_1 + a_{2,2}x_2 + \dots + a_{2,n}x_n & = b_2 \\ \vdots & \vdots \\ a_{m,1}x_1 + a_{m,2}x_2 + \dots + a_{m,n}x_n & = b_m \end{cases},$$

where $a_{i,j}, b_i \in \mathbb{R}$ for $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$. A solution is an n -tuple $(x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ that fulfills all m equations of the system simultaneously. If $b_1 = b_2 = \dots = b_m = 0$, the system is called homogenous otherwise inhomogenous.

Problem 3. Given: A system of linear equations.

Goal: Find all possible solutions given as

$$S = \left\{ (x_1, x_2, \dots, x_n)^T \in \mathbb{R}^n : \sum_{k=1}^n a_{i,k} x_k = b_i \text{ for all } i = 1, 2, \dots, m \right\}.$$

Definition 4 (matrix). Let $m, n \in \mathbb{N}$. A scheme

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n} \\ a_{2,1} & a_{2,2} & \dots & a_{2,n} \\ \vdots & & & \\ a_{m,1} & a_{m,2} & \dots & a_{m,n} \end{pmatrix},$$

where $a_{i,j} \in \mathbb{R}$ for $i = 1, 2, \dots, m$ and $j = 1, 2, \dots, n$ is called a matrix, more precisely an $(m \times n)$ -matrix. Furthermore, $(a_{i,1}, a_{i,2}, \dots, a_{i,n})$ are the rows and $(a_{1,j}, a_{2,j}, \dots, a_{m,j})^T$ are the columns of the matrix.

Definition 5 ((extended) matrix of coefficients). If

$$\begin{cases} a_{1,1}x_1 + a_{1,2}x_2 + \dots + a_{1,n}x_n & = b_1 \\ a_{2,1}x_1 + a_{2,2}x_2 + \dots + a_{2,n}x_n & = b_2 \\ \vdots & \vdots \\ a_{m,1}x_1 + a_{m,2}x_2 + \dots + a_{m,n}x_n & = b_m \end{cases}$$

is a system of linear equations, we call

$$A = \begin{pmatrix} a_{1,1} & a_{1,2} & \dots & a_{1,n} \\ a_{2,1} & a_{2,2} & \dots & a_{2,n} \\ \vdots & & & \\ a_{m,1} & a_{m,2} & \dots & a_{m,n} \end{pmatrix}$$

the matrix of coefficients and

$$(A|\mathbf{b}) = \left(\begin{array}{cccc|c} a_{1,1} & a_{1,2} & \dots & a_{1,n} & b_1 \\ a_{2,1} & a_{2,2} & \dots & a_{2,n} & b_2 \\ \vdots & & & & \vdots \\ a_{m,1} & a_{m,2} & \dots & a_{m,n} & b_m \end{array} \right)$$

the extended matrix of coefficients. We write $Ax = \mathbf{b}$ as an abbreviation for the system of linear equations.

Definition 6 (simple form). An $(m \times n)$ -matrix is in simple form if

$$A = \begin{pmatrix} & *_1 & & & & \\ & & *_2 & & & \\ & & & *_3 & & \\ & & & & \dots & \\ & & & & & *_r \end{pmatrix}$$

such that the entries at positions $*_i$ are $\neq 0$ for $i = 1, 2, \dots, r$ and below the line are only 0s.

More precisely, A is in simple form if

(i) there exists a number $r \in \mathbb{N}$ with $0 \leq r \leq m$ such that

- the rows with index $1, 2, \dots, r$ each contain an entry $\neq 0$;
- the rows with index $r + 1, r + 2, \dots, m$ contain only 0s.

(ii) Let $j_i = \min\{j: a_{i,j} \neq 0\}$ for $1 \leq i \leq r$. Obviously $1 \leq j_i \leq n$ and we require $j_1 < j_2 < \dots < j_r$.

Note that $r = 0$ is possible. In this case all entries of A are 0. The entries $a_{1,j_1}, a_{2,j_2}, \dots, a_{r,j_r}$ are called pivots of A .

Proposition 7. If A is a matrix in simple form, then $j_i = i$ for $i = 1, 2, \dots, r$ can be obtained by rearrangement of its columns.

Theorem 8 (solution of linear systems in simple form). Let $Ax = b$ with A in simple form such that the pivots are in the first r columns, i.e.

$$(A|b) = \left(\begin{array}{cccc|cccc} a_{1,1} & & & & b_1 & & & & \\ & a_{2,2} & & & b_2 & & & & \\ & & \dots & & \vdots & & & & \\ & & & a_{r,r} & b_r & & & & \\ & & & & \hline & & & & b_{r+1} & & & & \\ & & & & \vdots & & & & \\ & & & & b_m & & & & \end{array} \right)$$

with $a_{1,1}, a_{2,2}, \dots, a_{r,r} \neq 0$.

- If $b_i \neq 0$ for an index $r + 1 \leq i \leq m$, then $S(A, b) = \emptyset$.
- If $b_i = 0$ for $i = r + 1, r + 2, \dots, m$, then $S(A, b) \neq \emptyset$ and can be computed as follows. Set $k = n - r$, choose $\lambda_1, \lambda_2, \dots, \lambda_k$ as parameters, and set $x_{r+1} = \lambda_1, x_{r+2} = \lambda_2, \dots, x_n = \lambda_k$. To compute x_1, x_2, \dots, x_r begin in row r :

$$a_{r,r}x_r + a_{r,r+1}\lambda_1 + \dots + a_{r,n}\lambda_k = b_r$$

$$\Leftrightarrow x_r = \frac{1}{a_{r,r}} (b_r - a_{r,r+1}\lambda_1 - \dots - a_{r,n}\lambda_k).$$

Plugging this into the $(r - 1)$ st row we can compute x_{r-1} and so on.

Definition 9 (free & bounded variables). *In the situation of Theorem 8, the variables $x_{r+1}, x_{r+2}, \dots, x_n$ are called free, and the variables x_1, x_2, \dots, x_r are called bounded.*

Definition 10 (elementary row transformations). *Row switching (switching all elements of row i with its counterparts in row j), row multiplication (multiplying all elements in row i with a nonzero scalar), and row addition (adding row j multiplied with a nonzero scalar to row i) are called elementary row transformations.*

Theorem 11 (row transformations do not change solutions). *Let $(A|b)$ be an augmented matrix of coefficients and suppose that $(A'|b')$ is obtained from $(A|b)$ by an elementary row transformation. Then $S(A, b) = S(A'|b')$, i.e. $Ax = b$ and $A'x = b'$ have the same set of solutions.*

Theorem 12 (matrices can be transformed into simple form). *For every matrix A there is a matrix B in simple form that can be obtained from A by finitely many elementary row transformations.*

Theorem 13 (GAUSS algorithm). *Given a system $Ax = b$ of linear equations.*

1. *If $A = 0$, STOP.*
2. *Locate the first nonzero column from the left. One row has a nonzero entry in this column. If necessary, swap this line with the first.*
3. *The first nonzero entry of the first row is in the first nonzero column. Using this entry, subtract suitable multiples of the first line from the other lines to generate zero entries in that column. This yields matrix $(A'|b')$.*
4. *If $(A'|b')$ is not in simple form, repeat step 3 with $(A|b) = (A'|b')$. Otherwise compute r and check b' to conclude whether $(A'|b')$ has solutions. If yes, compute the parametrization.*

Theorem 14 (Computation of inverse matrices). *The inverse of a matrix can be computed using the algorithm of GAUSS.*

Theorem 15 (Invertible matrices & linear systems). *Let $A \in \mathbb{R}^{n \times n}$. Then the following are equivalent.*

- (i) *A is invertible.*
- (ii) *For every $\mathbf{b} \in \mathbb{R}^n$: $A\mathbf{x} = \mathbf{b}$ has a unique solution.*
- (iii) *A has rank n .*

Theorem 16. *Let $A \in \mathbb{R}^{m \times n}$ and $r = \text{rank}(A)$. Then*

$$\dim(\text{kernel}(A)) = \dim(S(A|\mathbf{0})) = n - r.$$