

5.4. Systems of linear equations.

System of m equations with n unknowns:

$$(S) \quad \begin{aligned} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n &= b_2 \\ &\vdots \\ a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n &= b_m \end{aligned}$$

Matrix notation

$$\mathbb{A}\mathbf{x} = \mathbf{b},$$

where $\mathbb{A} = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} \in M(m \times n)$, $\mathbf{b} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} \in M(m \times 1)$ and $\mathbf{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in M(n \times 1)$.

Theorem 5.17 (on systems with square matrix). *Let $\mathbb{A} \in M(n \times n)$. Then the following are equivalent.*

- (i) *The matrix \mathbb{A} is invertible.*
- (ii) *The system (S) has for each \mathbf{b} a unique solution.*
- (iii) *The system (S) has for each \mathbf{b} at least one solution.*

Remark. The previous theorem says the following:

- If \mathbb{A} is invertible, then for each \mathbf{b} the system (S) has a unique solution.
- If \mathbb{A} is not invertible, then for some \mathbf{b} the system (S) has no solution.

It can be moreover shown, that, provided \mathbb{A} is not invertible, then for some \mathbf{b} the system (S) has infinitely many solutions ($\mathbf{b} = \mathbf{o}$ works). This follows from the next section.

Theorem 5.18 (Cramer's rule). *Let $\mathbb{A} \in M(n \times n)$ be an invertible matrix, $\mathbf{b} \in M(n \times 1)$, $\mathbf{x} \in M(n \times 1)$, and $\mathbb{A}\mathbf{x} = \mathbf{b}$. Then*

$$x_j = \frac{\begin{vmatrix} a_{11} & \cdots & a_{1,j-1} & b_1 & a_{1,j+1} & \cdots & a_{1n} \\ \vdots & & & \vdots & & & \vdots \\ a_{n1} & \cdots & a_{n,j-1} & b_n & a_{n,j+1} & \cdots & a_{nn} \end{vmatrix}}{\det \mathbb{A}}$$

for $j = 1, \dots, n$.

Definition. The matrix

$$(\mathbb{A}|\mathbf{b}) = \left(\begin{array}{ccc|c} a_{11} & \cdots & a_{1n} & b_1 \\ \vdots & \ddots & \vdots & \vdots \\ a_{m1} & \cdots & a_{mn} & b_m \end{array} \right)$$

is called *augmented matrix of the system (S)*.

Gauss elimination method. Consider the system (S).

- Let T be a transformation of matrices with m rows. Suppose that $\mathbb{A} \xrightarrow{T} \mathbb{A}'$ and $\mathbf{b} \xrightarrow{T} \mathbf{b}'$. Then the system $\mathbb{A}'\mathbf{x} = \mathbf{b}'$ has the same set of solutions as the system (S).
- The augmented matrix of the system (S) can be transformed to a row echelon matrix $(\mathbb{A}'|\mathbf{b}')$. Thus solving the system (S) can be reduced to solving the system $\mathbb{A}'\mathbf{x} = \mathbf{b}'$, which is much simpler.
- One can use only the original version of the transformation, i.e., a finite sequence of elementary **row** transformations.

Theorem 5.19 (solvability of a linear system). *The system (S) has a solution if and only if the matrix has the same rank as the extended matrix of the system.*

Remark. The system (S) has a solution if and only if the vector \mathbf{b} can be expressed as a linear combination of the columns of the matrix \mathbb{A} .

5.5. Matrices and linear mappings.

Definition. We say that a mapping $f: \mathbf{R}^n \rightarrow \mathbf{R}^m$ is *linear* if

- (i) $\forall \mathbf{u}, \mathbf{v} \in \mathbf{R}^n: f(\mathbf{u} + \mathbf{v}) = f(\mathbf{u}) + f(\mathbf{v})$,
- (ii) $\forall \lambda \in \mathbf{R} \forall \mathbf{u} \in \mathbf{R}^n: f(\lambda \mathbf{u}) = \lambda f(\mathbf{u})$.

Definition. Let $i \in \{1, \dots, n\}$. The vector

$$\mathbf{e}^i = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \dots i\text{-th coordinate}$$

is called *i-th canonical vector* of the space \mathbf{R}^n . The set $\{\mathbf{e}^1, \dots, \mathbf{e}^n\}$ of all canonical vectors in \mathbf{R}^n is called *canonical basis of the space \mathbf{R}^n* .

The properties of canonical vectors:

- (i) $\forall \mathbf{x} \in \mathbf{R}^n \exists \lambda_1, \dots, \lambda_n \in \mathbf{R}: \mathbf{x} = \lambda_1 \mathbf{e}^1 + \dots + \lambda_n \mathbf{e}^n$,
- (ii) the vectors $\mathbf{e}^1, \dots, \mathbf{e}^n$ are linearly independent.

Theorem 5.20 (representation of linear mappings). *The mapping $f: \mathbf{R}^n \rightarrow \mathbf{R}^m$ is linear if and only if there exists a matrix $\mathbb{A} \in M(m \times n)$ such that*

$$\forall \mathbf{u} \in \mathbf{R}^n: f(\mathbf{u}) = \mathbb{A}\mathbf{u} = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}.$$

Remark. The matrix \mathbb{A} from the previous theorem is uniquely determined and is called the *representing matrix* of the linear mapping f .

Theorem 5.21 (representing matrix of a composition). *Let $f: \mathbf{R}^n \rightarrow \mathbf{R}^m$ be a linear mapping represented by matrix $\mathbb{A} \in M(m \times n)$ a $g: \mathbf{R}^m \rightarrow \mathbf{R}^k$ be a linear mapping represented by a matrix $\mathbb{B} \in M(k \times m)$. Then the composed mapping $g \circ f: \mathbf{R}^n \rightarrow \mathbf{R}^k$ is linear and is represented by the matrix $\mathbb{B}\mathbb{A}$.*

Theorem 5.22. *Let a mapping $f: \mathbf{R}^n \rightarrow \mathbf{R}^n$ be linear. Then the following are equivalent.*

- (i) *The mapping f is a bijection (i.e., f is an injective mapping \mathbf{R}^n onto \mathbf{R}^n).*
- (ii) *The mapping f is an injective mapping.*
- (iii) *The mapping f is a mapping \mathbf{R}^n onto \mathbf{R}^n .*