### Course 10

### Linear systems of equations



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# Chapter 3. Matrices and Linear Systems

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# Application: simple authentication scheme

Following [Klein], we describe a simple authentication scheme, whose security is related to solving linear systems of equations.

## Linear systems of equations

Throughout this section K will be a field. We use superior indices to denote vectors in  $K^n$  and inferior indices to denote their components. For instance,  $x^0 = (x_1^0, \dots, x_n^0) \in K^n$ .

#### Definition

Consider a *linear system* of m equations with n unknowns  $x_1, \ldots, x_n$ :

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{cases}$$
(5)

where the coefficients are  $a_{ij}, b_i \in K$  (i = 1, ..., m, j = 1, ..., n). If  $b_1 = \cdots = b_m = 0$ , then the system (S) is called homogeneous and is denoted by  $(S_0)$ .

# Matrices of linear systems of equations

#### Definition

The matrix  $A = (a_{ij}) \in M_{m,n}(K)$  is called the matrix of the system (S).

The matrix

$$\bar{A} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{pmatrix}$$

is called the augmented (or extended) matrix of the system (S).

# Matrix form of a linear system of equations

Next we present two equivalent forms of a linear system of equations.

First, denote

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$
 and  $b = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix}$ .

Then the systems (S) and  $(S_0)$  can be written as:

$$A \cdot x = b \,, \tag{S}$$

$$A \cdot x = 0. \tag{S_0}$$

## Linear systems of equations and linear maps

We know that there is a bijective correspondence between K-linear maps and matrices. Thus, since  $A \in M_{m,n}(K)$ , there is  $f_A \in \operatorname{Hom}_K(K^n, K^m)$  such that  $[f_A]_{EE'} = A$ , where E and E' are the canonical bases in  $K^n$  and  $K^m$  respectively.

Denoting  $x = (x_1, \dots, x_n) \in K^n$  and  $b = (b_1, \dots, b_m) \in K^m$ , it follows that

$$[f_A(x)]_{E'} = [f_A]_{EE'} \cdot [x]_E = A \cdot \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ \vdots \\ b_m \end{pmatrix} = [b]_{E'}.$$

Hence  $f_A(x) = b$ . Thus, (S) and  $(S_0)$  can be written as:

$$f_A(x) = b\,, (S)$$

$$f_A(x)=0. (S_0)$$

# Solutions of linear systems of equations

#### Definition

An element  $x^0 \in M_{n1}(K)$   $(x^0 \in K^n)$  is called a:

- (1) (particular) solution of (S) if  $A \cdot x^0 = b$  (or equivalently  $f_A(x^0) = b$ ).
- (2) (particular) solution of  $(S_0)$  if  $A \cdot x^0 = 0$  (or equivalently  $f_A(x^0) = 0$ ).

Denote the sets of solutions of (S) and  $(S_0)$  by

$$S = \{x^0 \in M_{n1}(K) \mid A \cdot x^0 = b\} \quad \text{or} \quad S = \{x^0 \in K^n \mid f_A(x^0) = b\},$$

$$S_0 = \{x^0 \in M_{n1}(K) \mid A \cdot x^0 = 0\} \quad \text{or} \quad S_0 = \{x^0 \in K^n \mid f_A(x^0) = 0\}.$$

## Solutions of the homogeneous system

#### $\mathsf{Theorem}$

The set  $S_0$  of solutions of the system  $(S_0)$  is a subspace of the canonical vector space  $K^n$  over K and

$$\dim S_0 = n - \operatorname{rank}(A).$$

Proof. Since

$$S_0 = \{x^0 \in K^n \mid f_A(x^0) = 0\} = \operatorname{Ker} f_A,$$

we have  $S_0 \leq K^n$ . By the First Dimension Theorem, it follows that

$$\dim S_0 = \dim K^n - \dim(\operatorname{Im} f_A) = n - \operatorname{rank}(f_A) = n - \operatorname{rank}(A),$$

which finishes the proof.



# Solutions of the general system

#### Theorem

If  $x^1 \in S$  is a particular solution of the system (S), then

$$S = x^1 + S_0 = \{x^1 + x^0 \mid x^0 \in S_0\}.$$

*Proof.* Since  $x^1 \in S$ , we have  $Ax^1 = b$ . First, let  $x^2 \in S$ . Then

$$Ax^2=b\Longrightarrow Ax^2=Ax^1\Longrightarrow A(x^2-x^1)=0\Longrightarrow x^2-x^1\in S_0\,,$$

hence  $x^2 \in x^1 + S_0$ .

Conversely, let  $x^2 \in x^1 + S_0$ . There exists  $x^0 \in S_0$  such that  $x^2 = x^1 + x^0$ . Then:

$$Ax^2 = A(x^1 + x^0) = Ax^1 + Ax^0 = b + 0 = b,$$

and consequently  $x^2 \in S$ .

Therefore, 
$$S = x^1 + S_0$$
.



## Compatible systems

#### Definition

The system (S) is called *compatible* (or *consistent*) if it has at least one solution. A compatible system (S) is called *determinate* if it has a unique solution.

The system (S) is compatible if and only if  $\exists x^0 \in K^n$  such that  $f_A(x^0) = b$  if and only if  $b \in \operatorname{Im} f_A$ .

The system  $(S_0)$  is compatible if and only if  $\exists x^0 \in K^n$  such that  $f_A(x^0) = 0$  if and only if  $0 \in \operatorname{Im} f_A$ . But the last condition always holds, since  $\operatorname{Im} f_A$  is a subspace of  $K^m$ . Hence any homogeneous linear system of equations is compatible, having at least the zero (trivial) solution.

# Compatible homogeneous system

#### Theorem

The system ( $S_0$ ) has a non-zero solution if and only if  $\operatorname{rank}(A) < n$ .

Proof. We have:

$$S_0 = \operatorname{Ker} f_A \neq \{0\} \iff \dim S_0 \neq 0 \iff n - \operatorname{rank}(A) \neq 0$$

which is equivalent to rank(A) < n.

### Corollary

Let  $A \in M_n(K)$ . Then

$$S_0 = \{0\} \iff \operatorname{rank}(A) = n \iff \det(A) \neq 0$$
.

## Cramer system I

#### Definition

If  $A \in M_n(K)$  and  $\det(A) \neq 0$ , then the system (S) is called a *Cramer system*.

#### Theorem

A Cramer system Ax = b has a unique solution. More precisely, its unique solution  $(x_1, \ldots, x_n)$  is computed by

$$x_i = \det(A)^{-1} \cdot d_i,$$

where  $d_i$  is the determinant obtained from det(A) by replacing its  $i^{th}$  column by the column b for every  $i \in \{1, ..., n\}$ .

## Cramer system II

*Proof.* The matrix of a Cramer system is an invertible matrix  $A \in M_n(K)$ . Then we deduce that  $x = A^{-1}b$  is the unique solution. Moreover, we have

$$\begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} = A^{-1} \cdot \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix} = \det(A)^{-1} \cdot A^* \cdot \begin{pmatrix} b_1 \\ \vdots \\ b_n \end{pmatrix} = \det(A)^{-1} \cdot \begin{pmatrix} d_1 \\ \vdots \\ d_n \end{pmatrix}.$$

Hence  $x_i = \det(A)^{-1} \cdot d_i$  for every  $i \in \{1, ..., n\}$ .

### Corollary

A homogeneous Cramer system has only the zero solution.

# Kronecker-Capelli Theorem

### Theorem (Kronecker-Capelli Theorem)

The system (S) is compatible if and only if  $rank(\bar{A}) = rank(A)$ .

*Proof.* Let  $(e_1, \ldots, e_n)$  be the canonical basis of the canonical vector space  $K^n$  over K and denote by  $a^1, \ldots, a^n$  the columns of the matrix A. Then we have

(S) is compatible 
$$\iff \exists x^0 \in K^n : f_A(x^0) = b \iff b \in \operatorname{Im} f_A$$
  
 $\iff b \in f_A(\langle e_1, \dots, e_n \rangle) \iff b \in \langle f_A(e_1), \dots, f_A(e_n) \rangle$   
 $\iff b \in \langle a^1, \dots, a^n \rangle \iff \langle a^1, \dots, a^n, b \rangle = \langle a^1, \dots, a^n \rangle$   
 $\iff \dim \langle a^1, \dots, a^n, b \rangle = \dim \langle a^1, \dots, a^n \rangle \iff \operatorname{rank}(\bar{A}) = \operatorname{rank}(A),$ 

which proves the result.

### Rouché Theorem I

#### Definition

A minor  $d_p$  of the matrix A is called a *principal determinant* if  $d_p \neq 0$  and  $d_p$  has the order rank(A).

We call characteristic determinants associated to a principal determinant  $d_p$  of A the minors of the augmented matrix  $\bar{A}$  obtained by completing the matrix of  $d_p$  with a column containing the corresponding constants  $b_i$  and a row containing the corresponding elements of a row of  $\bar{A}$ .

#### Theorem (Rouché Theorem)

The system (S) is compatible if and only if all the characteristic determinants associated to a principal determinant are zero.

### Gauss method

In this section we briefly present a very useful practical method to solve linear systems of equations, called the *Gauss method* (or *Gaussian elimination*).

In the sequel, suppose that  $m \le n$ . In fact, this is the interesting case.

The **Gauss method** consists of the following steps:

- **1** Write the augmented matrix  $\bar{A}$  of the system (S).
- ② Apply elementary operations on rows for  $\bar{A}$  to get to an echelon form A'.
- ① Use the Kronecker-Capelli Theorem to decide if the system is compatible or not.
- If compatible, write and solve the system corresponding to the echelon form, starting with the last equation.

#### Remarks

- (1) Actually, the Gauss method simulates working with equations. When we apply an elementary operation on the rows of  $\bar{A}$ , say multiply a row by a scalar and add it to another row, in fact we multiply an equation by a scalar and add it to another equation.
- (2) The initial system and the system corresponding to the echelon form are equivalent, that is, they have the same solutions. The last system can be easily solved, starting with the last equation.
- (3) The Gauss method includes checking compatibility, done by the Kronecker-Capelli Theorem.
- (4) If the system is compatible, we have a principal determinant of order  $r = \operatorname{rank}(\bar{A}) = \operatorname{rank}(A)$  and it is possible to continue the procedure on the matrix A' to get to a diagonal form having r elements on the principal diagonal and all the other elements zero. Then, when writing the equivalent system, in fact we directly get the solution. This completion is called the *Gauss-Jordan method*.

## Examples I

(a) Consider the system

$$\begin{cases} x+y-z=2\\ 3x+2y-2z=6\\ -x+y+z=0 \end{cases}$$

with real coefficients. It is determinate compatible with the solution x = 2, y = 1, z = 1 [...].

(b) Consider the system

$$\begin{cases} x + y + z = 0 \\ x + 4y + 10z = 3 \\ 2x + 3y + 5z = 1 \end{cases}$$

with real coefficients. It is non-determinate compatible with the solution  $x=2z-1,\ y=1-3z,\ z\in\mathbb{R}$  [...].

# Examples II

(c) Consider the system

$$\begin{cases} x + y + z = 3 \\ x - y + z = 1 \\ -2x + y - 2z = -3 \\ x + z = 4 \end{cases}$$

with real coefficients. It is not compatible.

## Computational cost of the Gauss method I

Following [Robbiano], let us analyze how many operations are required to solve a linear system of equations Ax = b with  $A \in M_n(K)$  invertible by the Gauss method. We assume that the operations of interchanging rows are negligible.

Let us first compute the cost of the reduction to a triangular echelon form having all elements on the principal diagonal equal to 1. We may assume that  $a_{11} \neq 0$ . We reduce  $a_{11}$  to 1 by dividing the first row of A by  $a_{11}$ . Then we produce zeros on the first column under the (1,1)-entry. For each of the n-1 rows we need n multiplications and n additions. Adding the corresponding operations for b, we have 1 more division, n-1 more multiplications and n-1 more additions.

After finishing working with the first row, we move on to the second row, and so on til we get the required triangular form with elements 1 on the principal diagonal.

## Computational cost of the Gauss method II

Counting up the operations we have

- $n + (n-1) + \cdots + 1$  divisions on A and n divisions on b;
- $n(n-1) + (n-1)(n-2) + \cdots + 2 \cdot 1$  multiplications on A and  $(n-1) + \cdots + 1$  multiplications on b;
- $n(n-1) + (n-1)(n-2) + \cdots + 2 \cdot 1$  additions on A and  $(n-1) + \cdots + 1$  additions on b.

So far we have

- $\frac{n(n+1)}{2} + n$  divisions;
- $\frac{n^3-n}{3} + \frac{n(n-1)}{2}$  multiplications;
- $\frac{n^3-n}{3} + \frac{n(n-1)}{2}$  additions.

Now we compute the cost of substitutions in the reduced triangular system. From the last equation we have  $x_n$ . For the substitution on the previous but last equation to find  $x_{n-1}$  we need 1 multiplication and 1 addition. Continuing the procedure, for the first equation to find  $x_1$  we need n-1 multiplications and n-1 additions.

### Computational cost of the Gauss method III

Counting up the operations, we have

• 
$$(n-1)+\cdots+1=\frac{n(n-1)}{2}$$
 multiplications;

• 
$$(n-1) + \cdots + 1 = \frac{n(n-1)}{2}$$
 additions.

Adding up the numbers of operations from the above two stages, it turns out that one needs:

- $\frac{n^3-n}{3}+n(n-1)$  multiplications;

Hence the order of magnitude is  $\frac{2}{3}n^3$  operations.

## Simple authentication scheme I

Let us consider the following simple authentication scheme from cryptography, following [Klein]. We denote by E the canonical basis of the canonical vector space  $\mathbb{Z}_2^n$  over  $\mathbb{Z}_2$ .

- The password is a vector  $v = (x_1, \dots, x_n) \in \mathbb{Z}_2^n$ .
- As a challenge, Computer sends a random vector  $u = (u_1, \dots, u_n) \in \mathbb{Z}_2^n$ .
- As the response, Human sends back the dot-product vector

$$u \cdot v = u_1 x_1 + \cdots + u_n x_n \in \mathbb{Z}_2.$$

• The challenge-response interaction is repeated until Computer is convinced that Human knows password *v*.

## Simple authentication scheme II

Eve eavesdrops and learns m pairs  $(a_1,b_1),\ldots,(a_m,b_m)$  such that each  $b_i$  is the correct response to challenge  $a_i$ . For every  $i\in\{1,\ldots,m\}$ , denote  $a_i=(a_{i1},\ldots,a_{in})$ . Then the password  $v=(x_1,\ldots,x_n)$  is a solution of the linear system of equations:

$$\begin{cases} a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ \dots \\ a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m \end{cases}$$

Once the rank of the matrix of the system reaches n, the solution is unique, and Eve can use the Gauss method to find it, obtaining the password.