

Chapter 3

Linear Equations

System of Linear Equations

System of m linear equations in n unknowns:

$$a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n = b_2$$

$$\vdots \quad \quad \quad \ddots \quad \quad \quad \vdots \quad \quad \quad \vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$$

$$\underbrace{\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix}}_{\text{coefficient matrix}} \cdot \underbrace{\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}}_{\text{variables}} = \underbrace{\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{pmatrix}}_{\text{vector of constants}}$$

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$$

Matrix Representation

Advantages of matrix representation:

- ▶ Short and compact notation.

Compare

$$\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$$

to

$$\sum_{j=1}^n a_{ij}x_j = b_i, \quad \text{for } i = 1, \dots, m$$

- ▶ We can transform equations by means of matrix algebra.
- ▶ We can use names for parts of the equation, like PRODUCTION VECTOR, DEMAND VECTOR, TECHNOLOGY MATRIX, etc. in the case of a Leontief model.

Leontief Model

Input-output model with

\mathbf{A} ... technology matrix

\mathbf{x} ... production vector

\mathbf{b} ... demand vector

$$\mathbf{x} = \mathbf{Ax} + \mathbf{b}$$

For a given output \mathbf{b} we get the corresponding input \mathbf{x} by

$$\mathbf{x} = \mathbf{Ax} + \mathbf{b} \quad | \quad - \mathbf{Ax}$$

$$\mathbf{x} - \mathbf{Ax} = \mathbf{b}$$

$$(\mathbf{I} - \mathbf{A})\mathbf{x} = \mathbf{b} \quad | \quad (\mathbf{I} - \mathbf{A})^{-1}.$$

$$\mathbf{x} = (\mathbf{I} - \mathbf{A})^{-1}\mathbf{b}$$

Solutions of a System of Linear Equations

Three possibilities:

- ▶ The system of equations has *exactly one* solution.
- ▶ The system of equations is *inconsistent*(not solvable).
- ▶ The system of equations has *infinitely many* solutions.

In **Gaussian elimination** the augmented coefficient matrix (\mathbf{A}, \mathbf{b}) is transformed into **row echelon form**.

Then the solution set is obtained by **back substitution**.

It is not possible to determine the number of solutions from the numbers of equations and unknowns. We have to transform the system first.

Row Echelon Form

In **row echelon form** the number of *leading zeros* strictly increases from one row to the row below.

$$\begin{pmatrix} 1 & a_{12} & a_{13} & a_{14} & a_{15} \\ 0 & 1 & a_{23} & a_{24} & a_{25} \\ 0 & 0 & 0 & 1 & a_{35} \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$

For our purposes it is not required that the first nonzero entries are equal to **1**.

Steps in Gaussian Elimination

We (have to) obtain the row echelon form (only) by means of following transformations *which do not change the set of solutions*:

- ▶ Multiplication of a row by some *nonzero* constant.
- ▶ Addition of the multiple of some row to another row.
- ▶ Exchange of two rows.

Example – Gaussian Elimination

$$\begin{array}{ccc|c} 1.0 & -0.2 & -0.2 & 7.0 \\ -0.4 & 0.8 & -0.1 & 12.5 \\ 0.0 & -0.5 & 0.9 & 16.5 \end{array}$$

We first add 0.4 times the first row to the second row.
We denote this operation by

$$R_2 \leftarrow R_2 + 0.4 \times R_1$$

$$\begin{array}{ccc|c} 1 & -0.20 & -0.20 & 7.0 \\ 0 & 0.72 & -0.18 & 15.3 \\ 0 & -0.50 & 0.90 & 16.5 \end{array}$$

Example – Gaussian Elimination

$$R_3 \leftarrow R_3 + \frac{0.5}{0.72} \times R_2$$

$$\begin{array}{ccc|c} 1 & -0.20 & -0.20 & 7.0 \\ 0 & 0.72 & -0.18 & 15.3 \\ 0 & 0 & 0.775 & 27.125 \end{array}$$

Example – Back Substitution

$$\begin{array}{ccc|c} 1 & -0.20 & -0.20 & 7.0 \\ 0 & 0.72 & -0.18 & 15.3 \\ 0 & 0 & 0.775 & 27.125 \end{array}$$

From the third row we immediately get:

$$0.775 \cdot x_3 = 27.125 \quad \Rightarrow \quad x_3 = 35$$

We obtain the remaining variables x_2 and x_1 by **back substitution**:

$$0.72 \cdot x_2 - 0.18 \cdot 35 = 15.3 \quad \Rightarrow \quad x_2 = 30$$

$$x_1 - 0.2 \cdot 30 - 0.2 \cdot 35 = 7 \quad \Rightarrow \quad x_1 = 20$$

The solution is unique: $\mathbf{x} = (20, 30, 35)^T$

Example 2

Find the solution of equation

$$3x_1 + 4x_2 + 5x_3 = 1$$

$$x_1 + x_2 - x_3 = 2$$

$$5x_1 + 6x_2 + 3x_3 = 4$$

$$\begin{array}{ccc|c} 3 & 4 & 5 & 1 \\ 1 & 1 & -1 & 2 \\ 5 & 6 & 3 & 4 \end{array}$$

$$R_2 \leftarrow 3 \times R_2 - R_1, \quad R_3 \leftarrow 3 \times R_3 - 5 \times R_1$$

$$\begin{array}{ccc|c} 3 & 4 & 5 & 1 \\ 0 & -1 & -8 & 5 \\ 0 & -2 & -16 & 7 \end{array}$$

Example 2

$$R_3 \leftarrow R_3 - 2 \times R_2$$

$$\begin{array}{ccc|c} 3 & 4 & 5 & 1 \\ 0 & -1 & -8 & 5 \\ 0 & 0 & 0 & -3 \end{array}$$

The third row implies $0 = -3$, a **contradiction**.

This system of equations is **inconsistent**; solution set $L = \emptyset$.

Example 3

Find the solution of equation

$$\begin{aligned}2x_1 + 8x_2 + 10x_3 + 10x_4 &= 0 \\x_1 + 5x_2 + 2x_3 + 9x_4 &= 1 \\-3x_1 - 10x_2 - 21x_3 - 6x_4 &= -4\end{aligned}$$

$$\begin{array}{cccc|c}2 & 8 & 10 & 10 & 0 \\1 & 5 & 2 & 9 & 1 \\-3 & -10 & -21 & -6 & -4\end{array}$$

$$R_2 \leftarrow 2 \times R_2 - R_1, \quad R_3 \leftarrow 2 \times R_3 + 3 \times R_1$$

$$\begin{array}{cccc|c}2 & 8 & 10 & 10 & 0 \\0 & 2 & -6 & 8 & 2 \\0 & 4 & -12 & 18 & -8\end{array}$$

Example 3

$$R_3 \leftarrow R_3 - 2 \times R_2$$

$$\begin{array}{cccc|c} 2 & 8 & 10 & 10 & 0 \\ 0 & 2 & -6 & 8 & 2 \\ 0 & 0 & 0 & 2 & -12 \end{array}$$

This equation has **infinitely many** solutions.

This can be seen from the *row echelon form* as there are *more* variables than nonzero rows.

Example 3

The third row immediately implies

$$2 \cdot x_4 = -12 \quad \Rightarrow \quad x_4 = -6$$

Back substitution yields

$$2 \cdot x_2 - 6 \cdot x_3 + 8 \cdot (-6) = 2$$

In this case we use *pseudo solution* $x_3 = \alpha$, $\alpha \in \mathbb{R}$, and get

$$x_2 - 3 \cdot \alpha + 4 \cdot (-6) = 1 \quad \Rightarrow \quad x_2 = 25 + 3\alpha$$

$$2 \cdot x_1 + 8 \cdot (25 + 3 \cdot \alpha) + 10 \cdot \alpha + 10 \cdot (-6) = 0$$

$$\Rightarrow \quad x_1 = -70 - 17 \cdot \alpha$$

Example 3

We obtain a solution for each value of α . Using vector notation we obtain

$$\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -70 - 17 \cdot \alpha \\ 25 + 3\alpha \\ \alpha \\ -6 \end{pmatrix} = \begin{pmatrix} -70 \\ 25 \\ 0 \\ -6 \end{pmatrix} + \alpha \begin{pmatrix} -17 \\ 3 \\ 1 \\ 0 \end{pmatrix}$$

Thus the solution set of this equation is

$$L = \left\{ \mathbf{x} = \begin{pmatrix} -70 \\ 25 \\ 0 \\ -6 \end{pmatrix} + \alpha \begin{pmatrix} -17 \\ 3 \\ 1 \\ 0 \end{pmatrix} \mid \alpha \in \mathbb{R} \right\}$$

Equivalent Representation of Solutions

In Example 3 we also could use $x_2 = \alpha'$ (instead of $x_3 = \alpha$).
Then back substitution yields

$$L' = \left\{ \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -\frac{215}{3} \\ 0 \\ -\frac{25}{3} \\ -6 \end{pmatrix} + \alpha' \begin{pmatrix} -\frac{17}{3} \\ 1 \\ \frac{1}{3} \\ 0 \end{pmatrix} \mid \alpha \in \mathbb{R} \right\}$$

However, these two solution sets are equal, $L' = L!$

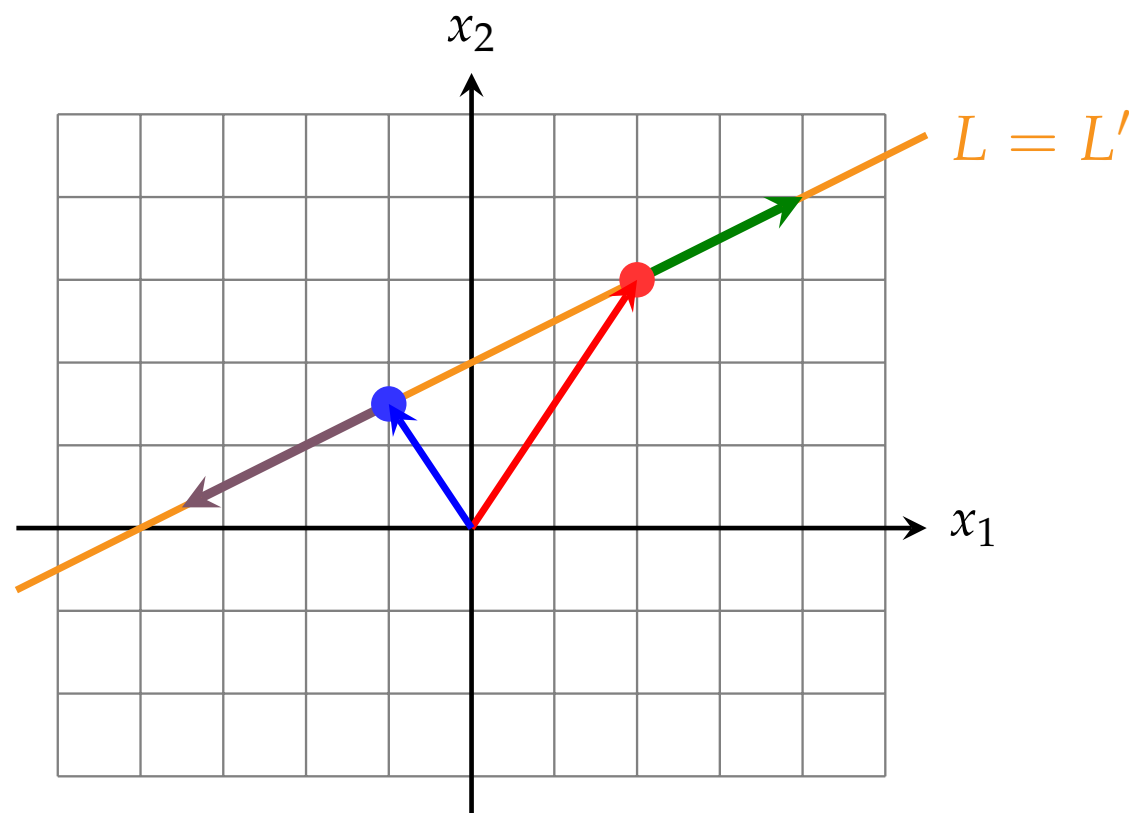
We thus have two different – *but equivalent* – representations of the same set.

The solution set is unique, its representation is not!

Equivalent Representation of Solutions

The set of solution points in Example 3 can be interpreted as a *line* in a (4-dimensional) space.

The representations in L and L' are thus parametric curves in \mathbb{R}^4 with the same image.



A Non-Example

Find the solution of equation

$$\begin{aligned}2x_1 + x_2 &= 1 \\ -2x_1 + 2x_2 - 2x_3 &= 4 \\ 4x_1 + 9x_2 - 3x_3 &= 9\end{aligned}$$

$$\begin{array}{ccc|c}2 & 1 & 0 & 1 \\ -2 & 2 & -2 & 4 \\ 4 & 9 & -3 & 9\end{array}$$

$$R_2 \leftarrow R_2 + R_1, \quad R_3 \leftarrow R_3 - 2 \times R_1$$

$$\begin{array}{ccc|c}2 & 1 & 0 & 1 \\ 0 & 3 & -2 & 5 \\ 0 & 7 & -3 & 7\end{array}$$

A Non-Example

Now one could find $R_3 \leftarrow R_3 - 7 \times R_1$ convenient. However,

$$\begin{array}{ccc|c} 2 & 1 & 0 & 1 \\ 0 & 3 & -2 & 5 \\ -14 & 0 & -3 & 0 \end{array}$$

destroys the already created row echelon form in the first column!

Much better: $R_3 \leftarrow 3 \times R_3 - 7 \times R_2$

$$\begin{array}{ccc|c} 2 & 1 & 0 & 1 \\ 0 & 3 & -2 & 5 \\ 0 & 0 & 5 & -14 \end{array}$$

Reduced Row Echelon Form

In **Gauss-Jordan elimination** the augmented matrix is transformed into **reduced row echelon form**, i.e.,

- ▶ It is in row echelon form.
- ▶ The leading entry in each nonzero row is a **1**.
- ▶ Each column containing a leading **1** has **0**s everywhere else.

$$\left(\begin{array}{ccccc} 1 & 0 & a_{13} & 0 & 0 \\ 0 & 1 & a_{23} & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right)$$

Back substitution is then simpler.

Gauss-Jordan Elimination / Example 3

Find the solution of equation

$$2x_1 + 8x_2 + 10x_3 + 10x_4 = 0$$

$$x_1 + 5x_2 + 2x_3 + 9x_4 = 1$$

$$-3x_1 - 10x_2 - 21x_3 - 6x_4 = -4$$

$$\begin{array}{cccc|c} 2 & 8 & 10 & 10 & 0 \\ 1 & 5 & 2 & 9 & 1 \\ -3 & -10 & -21 & -6 & -4 \end{array}$$

$$R_1 \leftarrow \frac{1}{2} \times R_1, \quad R_2 \leftarrow 2 \times R_2 - R_1, \quad R_3 \leftarrow 2 \times R_3 + 3 \times R_1$$

$$\begin{array}{cccc|c} 1 & 4 & 5 & 5 & 0 \\ 0 & 2 & -6 & 8 & 2 \\ 0 & 4 & -12 & 18 & -8 \end{array}$$

Gauss-Jordan Elimination / Example 3

$$R_1 \leftarrow R_1 - 2 \times R_2, \quad R_2 \leftarrow \frac{1}{2} \times R_2, \quad R_3 \leftarrow R_3 - 2 \times R_2$$

$$\begin{array}{cccc|c} 1 & 0 & 17 & -11 & -4 \\ 0 & 1 & -3 & 4 & 1 \\ 0 & 0 & 0 & 2 & -12 \end{array}$$

$$R_1 \leftarrow R_1 + \frac{11}{2} \times R_3, \quad R_2 \leftarrow R_2 - 2 \times R_3, \quad R_3 \leftarrow \frac{1}{2} \times R_3,$$

$$\begin{array}{cccc|c} 1 & 0 & 17 & 0 & -70 \\ 0 & 1 & -3 & 0 & 25 \\ 0 & 0 & 0 & 1 & -6 \end{array}$$

Gauss-Jordan Elimination / Example 3

$$\begin{array}{cccc|c} 1 & 0 & 17 & 0 & -70 \\ 0 & 1 & -3 & 0 & 25 \\ 0 & 0 & 0 & 1 & -6 \end{array}$$

The third row immediately implies $x_4 = -6$

Set *pseudo solution* $x_3 = \alpha$, $\alpha \in \mathbb{R}$.

Back substitution yields $x_2 = 25 + 3\alpha$

and $x_1 = -70 - 17 \cdot \alpha$

Gauss-Jordan Elimination / Example 3

Thus the solution set of this equation is

$$L = \left\{ \mathbf{x} = \begin{pmatrix} -70 \\ 25 \\ 0 \\ -6 \end{pmatrix} + \alpha \begin{pmatrix} -17 \\ 3 \\ 1 \\ 0 \end{pmatrix} \mid \alpha \in \mathbb{R} \right\}$$

Gauss-Jordan Elimination / Example 3

Compare

$$\begin{array}{cccc|c} 1 & 0 & 17 & 0 & -70 \\ 0 & 1 & -3 & 0 & 25 \\ 0 & 0 & 0 & 1 & -6 \end{array} \quad \text{and} \quad \mathbf{x} = \begin{pmatrix} -70 \\ 25 \\ 0 \\ -6 \end{pmatrix} + \alpha \begin{pmatrix} -17 \\ 3 \\ 1 \\ 0 \end{pmatrix}$$

The positional vector $(-70, 25, 0, -6)^T$ follows from the r.h.s. of the reduced row echelon while the direction vector $(-17, 3, 1, 0)^T$ is given by the column without leading **1**.

Inverse of a Matrix

Computation of the inverse \mathbf{A}^{-1} of matrix \mathbf{A}
by *Gauss-Jordan elimination*:

- (1) Augment matrix \mathbf{A} by the corresponding *identity matrix* to the right.
- (2) Transform the augmented matrix such that the identity matrix appears on the left hand side by means of the transformation steps of Gaussian elimination.
- (3) Either the procedure is successful. Then we obtain the *inverse matrix* \mathbf{A}^{-1} on the right hand side.
- (4) Or the procedure aborts (because we obtain a row of zeros on the l.h.s.). Then the matrix is *singular*.

Example 1

Compute the inverse of

$$\mathbf{A} = \begin{pmatrix} 3 & 2 & 6 \\ 1 & 1 & 3 \\ -3 & -2 & -5 \end{pmatrix}$$

(1) Augment matrix \mathbf{A} :

$$\left(\begin{array}{ccc|ccc} 3 & 2 & 6 & 1 & 0 & 0 \\ 1 & 1 & 3 & 0 & 1 & 0 \\ -3 & -2 & -5 & 0 & 0 & 1 \end{array} \right)$$

Example 1

(2) Transform:

$$R_1 \leftarrow \frac{1}{3} \times R_1, \quad R_2 \leftarrow 3 \times R_2 - R_1, \quad R_3 \leftarrow R_3 + R_1$$

$$\left(\begin{array}{ccc|ccc} 1 & \frac{2}{3} & 2 & \frac{1}{3} & 0 & 0 \\ 0 & 1 & 3 & -1 & 3 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 \end{array} \right)$$

$$R_1 \leftarrow R_1 - \frac{2}{3} \times R_2$$

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

Example 1

$$R_2 \leftarrow R_2 - 3 \times R_3$$

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

(3) Matrix \mathbf{A} is invertible with inverse

$$\mathbf{A}^{-1} = \begin{pmatrix} 1 & -2 & 0 \\ -4 & 3 & -3 \\ 1 & 0 & 1 \end{pmatrix}$$

Example 2

Compute the inverse of

$$\mathbf{A} = \begin{pmatrix} 3 & 1 & 3 \\ 2 & 4 & 1 \\ 5 & 5 & 4 \end{pmatrix}$$

(1) Augment matrix \mathbf{A} :

$$\left(\begin{array}{ccc|ccc} 3 & 1 & 3 & 1 & 0 & 0 \\ 2 & 4 & 1 & 0 & 1 & 0 \\ 5 & 5 & 4 & 0 & 0 & 1 \end{array} \right)$$

Example 2

(2) Transform:

$$R_1 \leftarrow \frac{1}{3} \times R_1, \quad R_2 \leftarrow 3 \times R_2 - 2 \times R_1, \quad R_3 \leftarrow 3 \times R_3 - 5 \times R_1$$

$$\left(\begin{array}{ccc|ccc} 1 & \frac{1}{3} & 1 & \frac{1}{3} & 0 & 0 \\ 0 & 10 & -3 & -2 & 3 & 0 \\ 0 & 10 & -3 & -5 & 0 & 5 \end{array} \right)$$

$$R_1 \leftarrow R_1 - \frac{1}{30} \times R_2, \quad R_2 \leftarrow \frac{1}{10} \times R_2, \quad R_3 \leftarrow R_3 - R_2$$

$$\left(\begin{array}{ccc|ccc} 1 & 0 & \frac{11}{10} & \frac{4}{10} & -\frac{1}{10} & 0 \\ 0 & 1 & -\frac{3}{10} & -\frac{2}{10} & \frac{3}{10} & 0 \\ 0 & 0 & 0 & -3 & -3 & 5 \end{array} \right)$$

(4) Matrix \mathbf{A} is *not* invertible.

Summary

- ▶ system of linear equations
- ▶ Gaussian elimination
- ▶ Gauss-Jordan elimination
- ▶ computation of inverse matrix