

# 37233 Linear Algebra

Subject coordinator and lecturer:

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## **Classes**

- One two-hour lecture (p.w.): Wednesdays 10:00 to 12:00
- One two-hour tutorial/computer laboratory (p.w.)

## **Subject Assessment**

- Tutorial assignments (9 weekly; 40 marks each): 36%
- Written assignment (due in week 8: 8–10 May): 14%
- Final exam (in June; closed book): 50%  
(note: it is required to gain at least 40% of the exam marks)
- To pass the subject: at least 50% for the final combined mark

## **Subject contents**

- Fundamentals of linear algebra
- Applications of linear algebra
- Computational methods

## **Software**

- Wolfram Mathematica

# Why do we need linear algebra?

Linear algebra is one of the fundamental areas of mathematics.

On a practical level, one of the aspects is solving linear equations.

- Linear models in science, engineering, economics, business . . .
- Many systems in real world behave in an (approximately) linear manner over a significant range of parameters
- Genuinely nonlinear problems can be often linearised — that is, approximated by linear systems
- Natural phenomena are often described in terms of partial or ordinary differential equations. Solving these equations requires discretisation. This, in turn, leads to linear systems.

# Applications of linear algebra

- Science
  - Mathematics
  - Astronomy
  - Physics
  - Chemistry
  - Biology
  - Statistics
  - ...
- Engineering (mechanical, electrical, ...)
- Economics and business
- Transport, logistics, ...
- “Big Data” analysis, IT, AI, machine learning, ...

## Revision: Matrices. Determinants.

## Revision: Matrices. Determinants.

For a  $2 \times 2$  matrix

$$\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

the determinant is

$$\det \mathbf{A} \equiv \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

Properties (these properties hold for any  $n \times n$  matrix)

$$\begin{vmatrix} a & b \\ c+e & d+f \end{vmatrix} = \begin{vmatrix} a & b \\ c & d \end{vmatrix} + \begin{vmatrix} a & b \\ e & f \end{vmatrix}$$

$$\begin{vmatrix} ka & b \\ kc & d \end{vmatrix} = k \begin{vmatrix} a & b \\ c & d \end{vmatrix}$$

$$\begin{vmatrix} b & a \\ d & c \end{vmatrix} = - \begin{vmatrix} a & b \\ c & d \end{vmatrix}$$

# Determinants

- Determinant of an identity (unitary) matrix equals to 1

$$\det \mathbf{I} = \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = 1$$

- If two rows of  $\mathbf{A}$  are same, then  $\det \mathbf{A} = 0$

$$\begin{vmatrix} a & b \\ a & b \end{vmatrix} = ab - ab = 0$$

- An elementary row operation of addition of a multiple of one row to another row leaves the determinant unchanged

$$\begin{vmatrix} a & b \\ c & d \end{vmatrix} = \begin{vmatrix} a & b \\ c \pm ka & d \pm kb \end{vmatrix}$$

- Determinant of a matrix with a zero row equals to zero

$$\begin{vmatrix} 0 & 0 \\ a & b \end{vmatrix} = 0$$



# Determinants

**Definition:** Let  $\mathbf{A}$  be  $n \times n$  matrix and  $a_{ij}$  an element of  $\mathbf{A}$ . The **cofactor** of  $a_{ij}$  is the  $(n - 1) \times (n - 1)$  determinant  $A_{ij}$  obtained by:

1. deleting the  $i$ -th row and  $j$ -th column of  $\mathbf{A}$  and
2. multiplying the resulting matrix determinant by  $(-1)^{(i+j)}$ .

The determinant of  $\mathbf{A}$  can be found by *expansion* along row  $i$ :

$$\det \mathbf{A} = a_{i1}A_{i1} + a_{i2}A_{i2} + \cdots + a_{in}A_{in}$$

For example, we can take an expansion along the first row:

$$\det \mathbf{A} = a_{11}A_{11} + a_{12}A_{12} + \cdots + a_{1n}A_{1n}$$

# Determinants

For example, for a  $3 \times 3$  matrix, its determinant can be found as:

$$\begin{aligned}\det \mathbf{A} &= \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = a_{11}A_{11} + a_{12}A_{12} + a_{13}A_{13} \\ &= a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}\end{aligned}$$

Alternatively, expansion along the second row yields

$$\begin{aligned}\det \mathbf{A} &= a_{21}A_{21} + a_{22}A_{22} + a_{23}A_{23} \\ &= -a_{21} \begin{vmatrix} a_{12} & a_{13} \\ a_{32} & a_{33} \end{vmatrix} + a_{22} \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} - a_{23} \begin{vmatrix} a_{11} & a_{12} \\ a_{31} & a_{32} \end{vmatrix}\end{aligned}$$

# Linear algebra for solving linear equations

# Linear equations

- Linear equation in variables  $x_1, x_2, \dots, x_n$  is an equation

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = b, \quad (1)$$

where  $b$  and  $a_1, a_2, \dots, a_n$  are real or complex numbers.  
(subscript  $n$  is an integer, a “counter” for the variables)

- For example, equations

$$4x_1 - 5x_2 + 2 = x_1, \quad x_2 = 2(\sqrt{6} - x_1) + x_3$$

are linear because they can be arranged in the form (1)

$$3x_1 - 5x_2 = -2, \quad 2x_1 + x_2 - x_3 = 2\sqrt{6}$$

- The following examples are not linear equations:

$$4x_1 - 5x_2 + 2 = x_1 \sin x_2, \quad x_2 = 2\sqrt{x_1} - 6$$

# Systems of linear equations

- A system of linear equations (a linear system) is a collection of one or more linear equations involving the same variables

$$\begin{cases} 2x_1 - x_2 + 1.5x_3 = 8, \\ x_1 - x_3 = -7. \end{cases}$$

In general, a system of  $m$  equations for  $n$  variables

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

- The solution of the system is a list of variables  $x_1, x_2, \dots, x_n$  that makes *each* equation a true statement.

# Linear equations: graphical representation

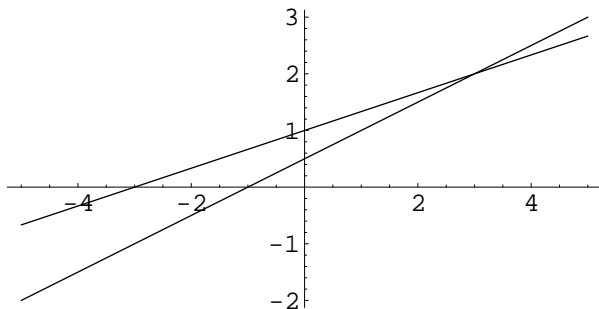
Consider a system of linear equations

$$\begin{cases} x_1 - 2x_2 = -1, \\ -x_1 + 3x_2 = 3. \end{cases}$$

It has a unique solution  $x_1 = 3$  and  $x_2 = 2$ .

This may be represented graphically; in “Mathematica” type

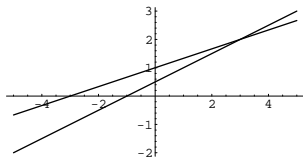
```
ContourPlot[{x1-2x2==-1,-x1+3x2==3},{x1,0,6},{x2,0,4},Axes->True]
```



# Linear equations: graphical representation

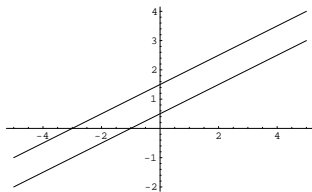
System 1: Unique solution

$$\begin{cases} x_1 - 2x_2 = -1, \\ -x_1 + 3x_2 = 3. \end{cases}$$



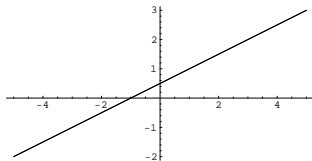
System 2: No solutions

$$\begin{cases} x_1 - 2x_2 = -1, \\ -x_1 + 2x_2 = 3. \end{cases}$$



System 3: Infinitely many solutions

$$\begin{cases} x_1 - 2x_2 = -1, \\ -x_1 + 2x_2 = 1. \end{cases}$$



# Linear systems of equations

- A linear system may have
  - exactly one solution
  - no solutions
  - infinitely many solutions
- For 2 equations in 2 unknowns, easy to assess and visualise. Much harder or impossible in higher dimensions.
- We need a general tool to determine whether a system has a solution, and if so, whether the solution is unique.



# Matrix representation of a linear system

$$\left\{ \begin{array}{cccccc} a_{11}x_1 & + & a_{12}x_2 & + & \dots & + & a_{1n}x_n & = & b_1, \\ a_{21}x_1 & + & a_{22}x_2 & + & \dots & + & a_{2n}x_n & = & b_2, \\ \vdots & & \vdots & & \dots & & \vdots & & \vdots \\ a_{m1}x_1 & + & a_{m2}x_2 & + & \dots & + & a_{mn}x_n & = & b_m. \end{array} \right.$$

Set of coefficients  $a_{ij}$  forms an  $m \times n$  matrix  $\mathbf{A}$  of the system:

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \dots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix},$$

The element  $a_{ij}$  located in  $i$ -th row and  $j$ -th column of  $\mathbf{A}$ .

# Matrix representation of a linear system

The entire system of linear equations

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

can be compactly written using the notations

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \dots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}.$$

in a short form:  $\mathbf{Ax} = \mathbf{b}$ .

# Augmented matrix of a linear system

For a linear system with  $m$  equations and  $n$  unknowns

$$\left\{ \begin{array}{cccccc} a_{11}x_1 & + & a_{12}x_2 & + & \dots & + & a_{1n}x_n & = & b_1, \\ a_{21}x_1 & + & a_{22}x_2 & + & \dots & + & a_{2n}x_n & = & b_2, \\ \vdots & & \vdots & & \dots & & \vdots & & \vdots \\ a_{m1}x_1 & + & a_{m2}x_2 & + & \dots & + & a_{mn}x_n & = & b_m. \end{array} \right.$$

an augmented matrix of the system is

$$\left[ \begin{array}{ccccc} a_{11} & a_{12} & \dots & a_{1n} & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & b_2 \\ \vdots & \vdots & \dots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & b_m \end{array} \right]$$

# Gaussian reduction / elimination

In order to solve a linear system using matrix representation, we need to reduce the augmented matrix to an *echelon form*.

This process is called Gauss – Jordan elimination, achieved with a series of *row operations*.

Row operations that can be used:

- Swapping two rows
- Multiplying a row by a constant
- Adding a multiple of one row to another

# Row reduction and echelon form

The first non-zero element in a row is called the **leading element**

The matrix in **echelon form** has the following properties:

- All non-zero rows are above any rows of all zeros.
- Each leading entry of a row is in a column to the right of the leading entry of the row above it.
- All entries in a column below a leading entry are zeros.

$$\begin{bmatrix} \blacksquare & * & * & * \\ 0 & \blacksquare & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

(where  $\blacksquare$  is a non-zero number, and  $*$  is any number)

# Reduced echelon form (REF)

The next step is to obtain a reduced echelon form (REF):

$$\begin{bmatrix} \blacksquare & * & * & * \\ 0 & \blacksquare & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 0 & * & * \\ 0 & 1 & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

**REF** matrix, in addition to **EF** form, has the properties that

- the leading entry in each non-zero row is 1
- the leading 1 is the only non-zero entry in its column

( $\blacksquare$  is a non-zero number,  $*$  is any number)

# Matrices in EF and REF forms

Scheme of an EF form

$$\begin{bmatrix} \blacksquare & * & * & * \\ 0 & \blacksquare & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{or} \quad \begin{bmatrix} 0 & \blacksquare & * & * & * & * & * & * & * & * \\ 0 & 0 & 0 & \blacksquare & * & * & * & * & * & * \\ 0 & 0 & 0 & 0 & \blacksquare & * & * & * & * & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \blacksquare & * \end{bmatrix}$$

Scheme of a REF form

$$\begin{bmatrix} 1 & 0 & * & * \\ 0 & 1 & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{or} \quad \begin{bmatrix} 0 & 1 & * & 0 & 0 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 1 & 0 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 0 & 1 & 0 & * & * & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & * \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & * \end{bmatrix}$$

( $\blacksquare$  is a non-zero number,  $*$  is any number)

# Matrices in EF and REF forms

- The echelon form of a matrix (**EF**) is not unique, however reduced echelon form (**REF**) is unique.
- **Theorem:** Each matrix is row-equivalent to one and only one matrix in reduced echelon form (REF).
- A **pivot position** corresponds to the leading (non-zero) entry. A **pivot column** is a column that contains a pivot position.

$$\begin{bmatrix} \blacksquare & * & * & * \\ 0 & \blacksquare & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \qquad \begin{bmatrix} 1 & 0 & * & * \\ 0 & 1 & * & * \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$



## Row reduction to EF (example)

Finding solutions of a linear system using Gaussian reduction:

$$\begin{cases} 3x_1 & - & 7x_2 & + & 8x_3 & - & 5x_4 & + & 8x_5 & = & 9 \\ 3x_1 & - & 9x_2 & + & 12x_3 & - & 9x_4 & + & 6x_5 & = & 15 \end{cases}$$

Write the augmented matrix of this system:

$$\left[ \begin{array}{ccccc|c} 0 & 3 & -6 & 6 & 4 & -5 \\ 3 & -7 & 8 & -5 & 8 & 9 \\ 3 & -9 & 12 & -9 & 6 & 15 \end{array} \right]$$

Follow a step-wise procedure, using elementary row operations

- Swapping two rows
- Multiplying a row by a constant
- Adding a multiple of one row to another

## Row reduction to EF (example)

Starting matrix: 
$$\begin{bmatrix} 0 & 3 & -6 & 6 & 4 & -5 \\ 3 & -7 & 8 & -5 & 8 & 9 \\ 3 & -9 & 12 & -9 & 6 & 15 \end{bmatrix}$$

- **Step 1:** Begin with the first left non-zero column.  
Swap the rows to bring any zeros of the first column down.  
Select nonzero entry in the pivot column as a pivot.

$$\begin{bmatrix} \color{red}{3} & -9 & 12 & -9 & 6 & 15 \\ 3 & -7 & 8 & -5 & 8 & 9 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

- **Step 2:** Use row replacement operation to create zeros in all positions below the pivot (here, use  $R_2 \rightarrow R_2 - R_1$ ).

$$\begin{bmatrix} \color{red}{3} & -9 & 12 & -9 & 6 & 15 \\ \color{blue}{0} & 2 & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

## Row reduction to EF (example)

- **Step 3.** Cover the row containing the pivot (and any rows above it). Apply steps 1–2 to the remaining sub-matrix. Repeat until there are no more non-zero rows to modify.

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & \textcolor{red}{2} & -4 & 4 & 2 & -6 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

(now, we divide the second row by 2)

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & \textcolor{red}{1} & -2 & 2 & 1 & -3 \\ 0 & 3 & -6 & 6 & 4 & -5 \end{bmatrix}$$

(now, we use  $R_3 \rightarrow R_3 - 3R_2$ )

$$\begin{bmatrix} 3 & -9 & 12 & -9 & 6 & 15 \\ 0 & \textcolor{red}{1} & -2 & 2 & 1 & -3 \\ 0 & \textcolor{blue}{0} & 0 & 0 & 1 & 4 \end{bmatrix}$$

## Row reduction from EF to REF (example)

We have achieved an EF:

$$\begin{bmatrix} \boxed{3} & -9 & 12 & -9 & 6 & 15 \\ 0 & \boxed{1} & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & \boxed{1} & 4 \end{bmatrix}$$

This is the end of the “forward” phase (down and to the right).

From here, we will work “backward” (to the left and up).

- **Step 4** : If a pivot is not 1, make it 1 by a scaling operation.

Here, row operation  $R_1 \rightarrow R_1/3$  leads to

$$\begin{bmatrix} \textcolor{red}{1} & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

## Row reduction from EF to REF (example)

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 1 & -3 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

- **Step 5** : Beginning with the rightmost pivot and working upward and to the left, create zeros above each pivot.

Here,  $R_2 \rightarrow R_2 - R_3$  leads to

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 2 & 5 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

and  $R_1 \rightarrow R_1 - 2R_3$  leads to

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 0 & -3 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

## The REF form (“backward phase”)

- Done with the lowest-right pivot; address the next left-up

$$\begin{bmatrix} 1 & -3 & 4 & -3 & 0 & -3 \\ 0 & \color{red}{1} & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix}$$

by making  $R_1 \rightarrow R_1 + 3R_2$ :

$$\begin{bmatrix} \color{red}{1} & \color{blue}{0} & -2 & 3 & 0 & -24 \\ \color{blue}{0} & \color{red}{1} & -2 & 2 & 0 & -7 \\ \color{blue}{0} & \color{blue}{0} & 0 & 0 & \color{red}{1} & 4 \end{bmatrix}$$

This finally brings the matrix to the REF form.

The resulting matrix corresponds to an equivalent system

$$x_1 - 2x_3 + 3x_4 = -24$$

$$x_2 - 2x_3 + 2x_4 = -7$$

$$x_5 = 4$$

## REF form and system solution

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 0 & -24 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix} \quad \begin{array}{rcl} x_1 - 2x_3 + 3x_4 & = & -24 \\ x_2 - 2x_3 + 2x_4 & = & -7 \\ x_5 & = & 4 \end{array}$$

- Note there are 3 equations for 5 variables.
- Variables with pivots are called **basic variables**:  $x_1, x_2, x_5$ .
- Variables without pivots are called **free variables**:  $x_3, x_4$ .
- In the final solution basic variables  $x_1, x_2, x_5$  must be expressed in terms of free variables  $x_3, x_4$ .

$$\begin{array}{rcl} \text{The solution is:} & x_1 & = -24 + 2x_3 - 3x_4 \\ & x_2 & = -7 + 2x_3 - 2x_4 \\ & x_5 & = 4 \end{array}$$

## Various representations

$$\begin{bmatrix} 1 & 0 & -2 & 3 & 0 & -24 \\ 0 & 1 & -2 & 2 & 0 & -7 \\ 0 & 0 & 0 & 0 & 1 & 4 \end{bmatrix} \Leftrightarrow \begin{array}{rcl} x_1 & = & -24 + 2x_3 - 3x_4 \\ x_2 & = & -7 + 2x_3 - 2x_4 \\ x_5 & = & 4 \end{array}$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} = \begin{bmatrix} -24 + 2x_3 - 3x_4 \\ -7 + 2x_3 - 2x_4 \\ x_3 \\ x_4 \\ 4 \end{bmatrix} = \begin{bmatrix} -24 \\ -7 \\ 0 \\ 0 \\ 4 \end{bmatrix} + x_3 \begin{bmatrix} 2 \\ 2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -3 \\ -2 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$



## Example 2

Second example to solve:

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & 2x_2 & - & 8x_3 & = & 8 \\ -4x_1 & + & 5x_2 & + & 9x_3 & = & -9 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 2 & -8 & 8 \\ -4 & 5 & 9 & -9 \end{array} \right]$$

To bring this to EF we eliminate  $x_1$  in equation 3:

$$\text{Eq3} + 4 * \text{Eq1} \rightarrow \text{Eq3} \quad \text{or} \quad \text{R3} + 4 * \text{R1} \rightarrow \text{R3}$$

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & 2x_2 & - & 8x_3 & = & 8 \\ & & -3x_2 & + & 13x_3 & = & -9 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 2 & -8 & 8 \\ 0 & -3 & 13 & -9 \end{array} \right]$$

## Example 2

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & 2x_2 & - & 8x_3 & = & 8 \\ & & -3x_2 & + & 13x_3 & = & -9 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 2 & -8 & 8 \\ 0 & -3 & 13 & -9 \end{array} \right]$$

Next, we eliminate  $x_2$  in equation 3. But first, factor R2.

$$\text{Eq2} \rightarrow \frac{1}{2} \times \text{Eq2} \quad \text{or} \quad \text{R2} \rightarrow \frac{1}{2} \times \text{R2}$$

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & x_2 & - & 4x_3 & = & 4 \\ & & -3x_2 & + & 13x_3 & = & -9 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 1 & -4 & 4 \\ 0 & -3 & 13 & -9 \end{array} \right]$$

Next,  $\text{Eq3} + 3 * \text{Eq2} \rightarrow \text{Eq3}$  or  $\text{R3} + 3 * \text{R2} \rightarrow \text{R3}$

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & x_2 & - & 4x_3 & = & 4 \\ & & & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

## Example 2

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & x_2 & - & 4x_3 & = & 4 \\ & & & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

The matrix of the system is now in an echelon form.

Now we continue towards REF, or we can also solve it directly:

Solving directly:

$$\text{Eq3} \Rightarrow x_3 = 3$$

$$\text{Eq2} \Rightarrow x_2 = 4x_3 + 4 = 4 \times 3 + 4 = 16$$

$$\text{Eq1} \Rightarrow x_1 = 2x_2 - x_3 + 0 = 2 \times 16 - 3 = 29$$

This is called **backward substitution**.

## Example 2

Continue with the elimination into REF

$$\begin{array}{rrcr} x_1 & - & 2x_2 & + & x_3 & = & 0 \\ & & x_2 & - & 4x_3 & = & 4 \\ & & & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & -2 & 1 & 0 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

Elimination above the pivots (yes, may be done also in this order):

$$\text{Eq1} + 2 * \text{Eq2} \rightarrow \text{Eq1} \quad \text{or} \quad \text{R1} + 2 * \text{R2} \rightarrow \text{R1}$$

$$\begin{array}{rrcr} x_1 & & - & 7x_3 & = & 8 \\ & x_2 & - & 4x_3 & = & 4 \\ & & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & 0 & -7 & 8 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

## Example 2

$$\begin{array}{rclcl} x_1 & & - & 7x_3 & = & 8 \\ & x_2 & & - & 4x_3 & = & 4 \\ & & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & 0 & -7 & 8 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

$$\text{Eq1} + 7 * \text{Eq3} \rightarrow \text{Eq1} \quad \text{or} \quad \text{R1} + 7 * \text{R3} \rightarrow \text{R1}$$

$$\text{Eq2} + 4 * \text{Eq3} \rightarrow \text{Eq2} \quad \text{or} \quad \text{R2} + 4 * \text{R3} \rightarrow \text{R2}$$

$$\begin{array}{rclcl} x_1 & & & = & 29 \\ & x_2 & & = & 16 \\ & & x_3 & = & 3 \end{array} \quad \left[ \begin{array}{cccc} 1 & 0 & 0 & 29 \\ 0 & 1 & 0 & 16 \\ 0 & 0 & 1 & 3 \end{array} \right]$$

## Example with no solutions

Third example

$$\begin{array}{rrcr} & x_2 & - & 4x_3 & = & 8 \\ 2x_1 & - & 3x_2 & + & 2x_3 & = & 1 \\ 5x_1 & - & 8x_2 & + & 7x_3 & = & 1 \end{array} \quad \left[ \begin{array}{cccc} 0 & 1 & -4 & 8 \\ 2 & -3 & 2 & 1 \\ 5 & -8 & 7 & 1 \end{array} \right]$$

Upon doing  $R_1 \leftrightarrow R_2$  and  $R_3 \rightarrow (2R_3 - 5R_2) + R_1$  we get

$$\begin{array}{rrcr} 2x_1 & - & 3x_2 & + & 2x_3 & = & 1 \\ & x_2 & - & 4x_3 & = & 8 \\ & & & 0 & = & 5 \end{array} \quad \left[ \begin{array}{cccc} 2 & -3 & 2 & 1 \\ 0 & 1 & -4 & 8 \\ 0 & 0 & 0 & 5 \end{array} \right]$$

The **inconsistency**  $0 = 5$  implies that this system does not have any solutions.

## Example with infinitely many solutions

Fourth example

$$\begin{array}{rrcr} x_2 & - & 4x_3 & = & 6 \\ 2x_1 & - & 3x_2 & + & 2x_3 & = & 1 \\ 5x_1 & - & 8x_2 & + & 7x_3 & = & -1/2 \end{array} \quad \left[ \begin{array}{cccc} 0 & 1 & -4 & 6 \\ 2 & -3 & 2 & 1 \\ 5 & -8 & 7 & -1/2 \end{array} \right]$$

Doing again  $R_1 \leftrightarrow R_2$  and  $R_3 \rightarrow (2R_3 - 5R_2) + R_1$  yields

$$\begin{array}{rrcr} 2x_1 & - & 3x_2 & + & 2x_3 & = & 1 \\ & & x_2 & - & 4x_3 & = & 6 \\ & & & & 0 & = & 0 \end{array} \quad \left[ \begin{array}{cccc} 2 & -3 & 2 & 1 \\ 0 & 1 & -4 & 6 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

3 equations in 3 unknowns  $\rightarrow$  2 equations in 3 unknowns  
 $\Rightarrow$  only two independent equations

No contradiction, but no unique solution (infinitely many)

# Summary

Case 1: Consistent system, unique solution:

$$\begin{bmatrix} 1 & -2 & 1 & 0 \\ 0 & 2 & -8 & 8 \\ -4 & 5 & 9 & -9 \end{bmatrix} \Leftrightarrow \begin{bmatrix} 1 & -2 & 1 & 0 \\ 0 & 1 & -4 & 4 \\ 0 & 0 & 1 & 3 \end{bmatrix}$$

Case 2: Inconsistent system, no solution:

$$\begin{bmatrix} 0 & 1 & -4 & 8 \\ 2 & -3 & 2 & 1 \\ 5 & -8 & 7 & 1 \end{bmatrix} \Leftrightarrow \begin{bmatrix} 2 & -3 & 2 & 1 \\ 0 & 1 & -4 & 8 \\ 0 & 0 & 0 & 5 \end{bmatrix}$$

Case 3: Consistent system with infinitely many solutions:

$$\begin{bmatrix} 0 & 1 & -4 & 6 \\ 2 & -3 & 2 & 1 \\ 5 & -8 & 7 & -1/2 \end{bmatrix} \Leftrightarrow \begin{bmatrix} 2 & -3 & 2 & 1 \\ 0 & 1 & -4 & 6 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

1. System of equations is **consistent** if the solution is unique or there are infinitely many solutions.
2. System of equations is **inconsistent** if it has no solutions.



# Applications of linear systems

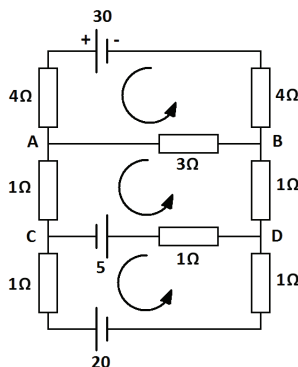
- Linear systems naturally arise in network analysis
- Network is a set of branches through which something "flows"
  - Electrical wires (electricity flow)
  - Economic linkages (money flow)
  - Pipes through which oil, gas or water flows
  - Fibres through which information flows (Internet)
- Branches meet at nodes or junctions
- A numerical measure is the rate of flow through a branch
- Analysis of networks is based on linear systems

## Example: Electric circuits

Voltage drop across a resistor is given by Ohm's law  $V = RI$

Kirchhoff's Voltage Law: *The algebraic sum of the  $IR$  voltage drops in one direction around a loop equals the algebraic sum of the voltage sources in the same direction around the loop.*

$$\sum_{i=1}^N R_i I_i = \sum_{i=1}^M V_i$$



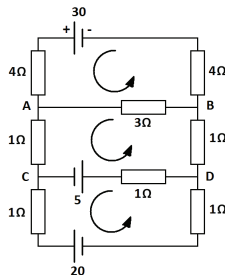
## Example: Electric circuits

$$\text{Loop 1: } 4I_1 + 3I_1 - 3I_2 + 4I_1 = 30$$

$$\text{Loop 2: } -3I_1 + 3I_2 + I_2 + I_2 - I_3 + I_2 = 5$$

$$\text{Loop 3: } -I_2 + I_3 + I_3 + I_3 = -5 - 20$$

$$\begin{cases} 11I_1 & - & 3I_2 & & = & 30 \\ -3I_1 & + & 6I_2 & - & I_3 & = & 5 \\ & - & I_2 & + & 3I_3 & = & -25 \end{cases}$$

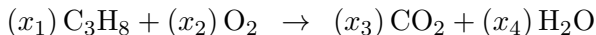


The loop currents are  $I_1 = 3$  A,  $I_2 = 1$  A and  $I_3 = -8$  A

- The total current in the branch AB is  $I_1 - I_2 = 3 - 1 = 2$  A
- The current in branch CD is  $I_2 - I_3 = 9$  A

## Example: Balancing chemical equations

Chemical equations describe the quantities of substances consumed and produced by chemical reactions:



Balancing requires finding amounts  $x_1, x_2, x_3, x_4$  such that the total amounts of carbon C, hydrogen H, and oxygen O atoms on the left match the corresponding numbers on the right.

$$\text{C}_3\text{H}_8 : \begin{bmatrix} 3 \\ 8 \\ 0 \end{bmatrix} \quad \text{O}_2 : \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} \quad \text{CO}_2 : \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} \quad \text{H}_2\text{O} : \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix}$$

$$x_1 \begin{bmatrix} 3 \\ 8 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 0 \\ -2 \end{bmatrix} + x_4 \begin{bmatrix} 0 \\ -2 \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

The solution of this system is  $x_1 = 1$ ,  $x_2 = 5$ ,  $x_3 = 3$ ,  $x_4 = 4$ .

## Revision: Inverse of a matrix

**Definition:** An  $n \times n$  matrix  $\mathbf{A}$  is said to be invertible if there is an  $n \times n$  matrix  $\mathbf{C}$  such that

$$\mathbf{CA} = \mathbf{I} \quad \text{and} \quad \mathbf{AC} = \mathbf{I},$$

where  $\mathbf{I}$  is a unitary  $n \times n$  matrix. Then  $\mathbf{C}$  is the inverse of  $\mathbf{A}$ .

$\mathbf{C}$  is uniquely determined by  $\mathbf{A}$ :

Indeed, suppose  $\mathbf{B}$  is another inverse of  $\mathbf{A}$ . Then

$$\mathbf{B} = \mathbf{BI} = \mathbf{B}(\mathbf{AC}) = (\mathbf{BA})\mathbf{C} = \mathbf{IC} = \mathbf{C}.$$

The inverse of  $\mathbf{A}$  is denoted by  $\mathbf{A}^{-1}$ :

$$\mathbf{A}^{-1}\mathbf{A} = \mathbf{I} \quad \text{and} \quad \mathbf{AA}^{-1} = \mathbf{I}.$$

A non-invertible matrix is called a **singular** matrix.

An invertible matrix is called a **non-singular** matrix.

# Inverse of a matrix

Example:

$$\mathbf{A} = \begin{bmatrix} 2 & 5 \\ -3 & -7 \end{bmatrix}, \quad \mathbf{C} = \begin{bmatrix} -7 & -5 \\ 3 & 2 \end{bmatrix}$$

Check

$$\mathbf{AC} = \begin{bmatrix} 2 & 5 \\ -3 & -7 \end{bmatrix} \begin{bmatrix} -7 & -5 \\ 3 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\mathbf{CA} = \begin{bmatrix} -7 & -5 \\ 3 & 2 \end{bmatrix} \begin{bmatrix} 2 & 5 \\ -3 & -7 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

**Theorem:**

$$\text{If } \mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \quad \text{and} \quad \det \mathbf{A} = ad - bc \neq 0$$

$$\text{then } \mathbf{A} \text{ is invertible and } \mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

# Inverse of a matrix

**Theorem:** A matrix is invertible if and only if  $\det \mathbf{A} \neq 0$ .

So if  $\det \mathbf{A} = 0$ , then  $\mathbf{A}$  is not invertible (singular).

**Theorem:**

a) If  $\mathbf{A}$  is invertible, then  $\mathbf{A}^{-1}$  is invertible, and  $(\mathbf{A}^{-1})^{-1} = \mathbf{A}$ .

b) If  $\mathbf{A}$  and  $\mathbf{B}$  are invertible, then  $(\mathbf{AB})^{-1} = \mathbf{B}^{-1}\mathbf{A}^{-1}$ .

**Theorem:**

If  $\mathbf{A}$  is an invertible  $n \times n$  matrix then  $\forall \mathbf{b} \in \mathbb{R}^n$ ,  
the equation  $\mathbf{Ax} = \mathbf{b}$  has the unique solution  $\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$ .

## $\mathbf{A}^{-1}$ by Gaussian elimination

**Theorem:** An  $n \times n$  matrix  $\mathbf{A}$  is invertible if and only if it is row-equivalent to identity matrix  $\mathbf{I}$ , and a sequence of elementary row operations that reduces  $\mathbf{A}$  to  $\mathbf{I}$ , transforms  $\mathbf{I}$  into  $\mathbf{A}^{-1}$ .

This gives an easy algorithm for calculating  $\mathbf{A}^{-1}$ :

Row reduce the augmented matrix  $[\mathbf{A} \mid \mathbf{I}]$ .

If  $\mathbf{A}$  is row-reduced to  $\mathbf{I}$ , then  $[\mathbf{A} \mid \mathbf{I}]$  is row-reduced to  $[\mathbf{I} \mid \mathbf{A}^{-1}]$ .

Otherwise  $\mathbf{A}$  does not have an inverse.

In practice  $\mathbf{A}^{-1}$  is seldom computed directly ( $2n^3$  operations).

Row reduction is faster and often more accurate.



## $\mathbf{A}^{-1}$ by Gaussian elimination

Example: Find inverse of a matrix, if it exists

$$\mathbf{A} = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 0 & 3 \\ 4 & -3 & 8 \end{bmatrix}$$

We form the extended augmented matrix

$$\begin{bmatrix} 0 & 1 & 2 & 1 & 0 & 0 \\ 1 & 0 & 3 & 0 & 1 & 0 \\ 4 & -3 & 8 & 0 & 0 & 1 \end{bmatrix}$$

$$R_1 \leftrightarrow R_2$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 4 & -3 & 8 & 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 4 & -3 & 8 & 0 & 0 & 1 \end{bmatrix}$$

$$R_3 \rightarrow R_3 - 4R_1$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & -3 & -4 & 0 & -4 & 1 \end{bmatrix}$$

$$R_3 \rightarrow R_3 + 3R_2$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 2 & 3 & -4 & 1 \end{bmatrix}$$

$$R_3 \rightarrow R_3/2$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 1 & 3/2 & -2 & 1/2 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 1 & 3/2 & -2 & 1/2 \end{bmatrix}$$

$$R_2 \rightarrow R_2 - 2R_3$$

$$\begin{bmatrix} 1 & 0 & 3 & 0 & 1 & 0 \\ 0 & 1 & 0 & -2 & 4 & -1 \\ 0 & 0 & 1 & 3/2 & -2 & 1/2 \end{bmatrix}$$

$$R_1 \rightarrow R_1 - 3R_3$$

$$\begin{bmatrix} 1 & 0 & 0 & -9/2 & 7 & -3/2 \\ 0 & 1 & 0 & -2 & 4 & -1 \\ 0 & 0 & 1 & 3/2 & -2 & 1/2 \end{bmatrix}$$

$$\text{Thus } \mathbf{A} \text{ is invertible and } \mathbf{A}^{-1} = \begin{bmatrix} -9/2 & 7 & -3/2 \\ -2 & 4 & -1 \\ 3/2 & -2 & 1/2 \end{bmatrix}$$

See you next Wednesday

C B 0 5 C . 0 1 . 0 3 1