

# Matrix Calculations: Solutions of Systems of Linear Equations

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# Outline

Review: pivots and Echelon form

Application: Data fitting

Vectors and solutions

Homogeneous systems

Non-homogeneous systems



# Pivots

- A **pivot** is the first non-zero entry of a row:

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

- If a row is all zeros, it **has no pivot**:

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

We call this a *zero row*.



# Echelon form

A matrix is in **Echelon form** if:

- ① All of the rows with pivots occur before zero rows, and
- ② Pivots always occur to the right of previous pivots

$$\left( \begin{array}{cccc|c} \boxed{3} & 2 & 5 & -5 & 1 \\ 0 & 0 & \boxed{2} & 1 & -2 \\ 0 & 0 & 0 & \boxed{-2} & 2 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \quad \checkmark$$

$$\left( \begin{array}{cccc|c} \boxed{3} & 2 & 5 & -5 & 1 \\ 0 & 0 & \boxed{2} & 1 & -2 \\ \color{red}{0} & \color{red}{0} & \color{red}{0} & \color{red}{0} & \color{red}{0} \\ 0 & 0 & 0 & \boxed{-2} & 2 \end{array} \right) \quad \text{☠}$$

$$\left( \begin{array}{cccc|c} \boxed{3} & 2 & 5 & -5 & 1 \\ 0 & 0 & \boxed{4} & -2 & 2 \\ 0 & \boxed{2} & 0 & 1 & -2 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \quad \text{☠}$$

$$\left( \begin{array}{cccc|c} \boxed{3} & 2 & 5 & -5 & 1 \\ 0 & 0 & \boxed{4} & -2 & 2 \\ 0 & 0 & \boxed{2} & 1 & -2 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right) \quad \text{☠}$$

# Points and polynomials

Here's a really useful thing about polynomials:

## Theorem

*For any  $n$  points in a plane, there exists a unique polynomial of degree  $n - 1$  which hits them all.*

*That is: given points  $(x_1, y_1), \dots, (x_n, y_n)$ , there is precisely one 'polynomial' function of the form:*

$$f(x) = a_0 + a_1x + a_2x^2 + a_3x^3 + \dots + a_{n-1}x^{n-1}$$

*with  $f(x_i) = y_i$  for all  $i \leq n$ .*

NB. No two points should be on the same vertical line!

- The **data fitting** problem is: given the points  $(x_i, y_i)$  obtained from some experiment, find the  $a_0, \dots, a_{n-1}$
- This can be done with what we have seen so far!

## Data fitting example

- Suppose we have 3 points  $(1, 6)$ ,  $(2, 3)$  and  $(3, 2)$
- we wish to find  $f(x) = a_0 + a_1x + a_2x^2$  that hits them all
- The requirements  $f(1) = 6$ ,  $f(2) = 3$  and  $f(3) = 2$  yield:

$$a_0 + a_1 \cdot 1 + a_2 \cdot 1^2 = 6$$

$$a_0 + a_1 \cdot 2 + a_2 \cdot 2^2 = 3$$

$$a_0 + a_1 \cdot 3 + a_2 \cdot 3^2 = 2$$

- The augmented matrix and its Echelon form are:

$$\left( \begin{array}{ccc|c} 1 & 1 & 1 & 6 \\ 1 & 2 & 4 & 3 \\ 1 & 3 & 9 & 2 \end{array} \right) \quad \text{and} \quad \left( \begin{array}{ccc|c} 1 & 1 & 1 & 6 \\ 0 & 1 & 3 & -3 \\ 0 & 0 & 2 & 2 \end{array} \right)$$

- Its solution is  $a_2 = 1$ ,  $a_1 = -6$  en  $a_0 = 11$ , ie.  $(11, -6, 1)$
- and so the required function is  $f(x) = 11 - 6x + x^2$ . ✓

## Unique solutions

From the first lecture:

### Theorem

A system of equations in  $n$  variables has a *unique solution* if and only if its Echelon form has  $n$  pivots.

### Example ( $\boxed{\phantom{x}}$ denotes a pivot)

$$\begin{array}{rcl} x_1 + x_2 & = & 3 \\ x_1 - x_2 & = & 1 \end{array} \quad \text{gives} \quad \left( \begin{array}{cc|c} 1 & 1 & 3 \\ 1 & -1 & 1 \end{array} \right) \quad \text{and} \quad \left( \begin{array}{cc|c} \boxed{1} & 1 & 3 \\ 0 & \boxed{1} & 1 \end{array} \right)$$

(using transformations  $R_2 := R_2 - R_1$  and  $R_2 := -\frac{1}{2}R_2$ )

**Question:** What if there are more solutions? Can we describe them in a generic way?

# General solutions

## The Goal:

- Describe the **space of solutions** of a system of equations.
- In general, there can be infinitely many solutions, but only a few are actually 'different enough' to matter. These are called **basic solutions**.
- Using the basic solutions, we can write down a formula which gives us any solution: the **general solution**.

## Example (General solution for one equation)

$$2x_1 - x_2 = 3 \quad \text{gives} \quad x_2 = 2x_1 - 3$$

So a general solution (for any  $c$ ) is:

$$x_1 := c \quad x_2 := 2c - 3$$



## A new tool: vectors

- A vector is a list of numbers.
- We can write it like this:  $(x_1, x_2, \dots, x_n)$
- ...or as a matrix with just one column:

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$$

(which is sometimes called a ‘column vector’).



## A new tool: vectors

- Vectors are useful for lots of stuff. In this lecture, we'll use them to hold **solutions**.
- Since variable names don't matter, we can write this:

$$x_1 := 2 \quad x_2 := -1 \quad x_3 := 0$$

- ...more compactly as this:

$$\begin{pmatrix} 2 \\ -1 \\ 0 \end{pmatrix}$$

- ...or even more compactly as this:  $(2, -1, 0)$ .

# Linear combinations

- We can multiply a vector by a number to get a new vector:

$$c \cdot \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} cx_1 \\ cx_2 \\ \vdots \\ cx_n \end{pmatrix}$$

This is called **scalar multiplication**.

- ...and we can add vectors together:

$$\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} + \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} := \begin{pmatrix} x_1 + y_1 \\ x_2 + y_2 \\ \vdots \\ x_n + y_n \end{pmatrix}$$

as long as they are the **same length**.

# Linear combinations

Mixing these two things together gives us a **linear combination** of vectors:

$$c \cdot \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} + d \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} + \dots = \begin{pmatrix} cx_1 + dy_1 + \dots \\ cx_2 + dy_2 + \dots \\ \vdots \\ cx_n + dy_n + \dots \end{pmatrix}$$

A set of vectors  $V_1, V_2, \dots, V_k$  is called **linearly independent** if no vector can be written as a linear combination of the others.

## Aside: checking linear independence

Equivalently:

### Definition

Vectors  $V_1, \dots, V_n$  are called **linearly independent** if for all scalars  $a_1, \dots, a_n \in \mathbb{R}$  one has:

$$a_1 \cdot V_1 + \dots + a_n \cdot V_n = 0 \text{ implies } a_1 = a_2 = \dots = a_n = 0$$

### Example

The 3 vectors  $(1, 0, 0)$ ,  $(0, 1, 0)$ ,  $(0, 0, 1)$  are linearly independent, since if

$$a_1 \cdot (1, 0, 0) + a_2 \cdot (0, 1, 0) + a_3 \cdot (0, 0, 1) = (0, 0, 0)$$

then, using the computation from the previous slide,

$$(a_1, a_2, a_3) = (0, 0, 0), \text{ so that } a_1 = a_2 = a_3 = 0$$

## Proving (in)dependence via equation solving I

- Investigate (in)dependence of  $\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ ,  $\begin{pmatrix} 2 \\ -1 \\ 4 \end{pmatrix}$ , and  $\begin{pmatrix} 0 \\ 5 \\ 2 \end{pmatrix}$
- Thus we ask: are there any non-zero  $a_1, a_2, a_3 \in \mathbb{R}$  with:

$$a_1 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + a_2 \begin{pmatrix} 2 \\ -1 \\ 4 \end{pmatrix} + a_3 \begin{pmatrix} 0 \\ 5 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

- If there is a non-zero solution, the vectors are **dependent**, and if  $a_1 = a_2 = a_3 = 0$  is the only solution, they are **independent**

## Proving (in)dependence via equation solving II

- Our question involves the systems of equations / matrix:

$$\begin{cases} a_1 + 2a_2 = 0 \\ 2a_1 - a_2 + 5a_3 = 0 \\ 3a_1 + 4a_2 + 2a_3 = 0 \end{cases} \quad \text{corresponding to} \quad \left( \begin{array}{ccc|c} 1 & 2 & 0 & 0 \\ 2 & -1 & 5 & 0 \\ 3 & 4 & 2 & 0 \end{array} \right) \quad \left( \begin{array}{c} 1 \\ 2 \\ 3 \end{array} \right)$$

- This one has **non-zero** solutions, for example  
 $a_1 = -2, a_2 = 1, a_3 = 1$   
(compute and check for yourself!)

- Thus the original vectors are **dependent**. Explicitly:

$$-2 \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} + \begin{pmatrix} 2 \\ -1 \\ 4 \end{pmatrix} + \begin{pmatrix} 0 \\ 5 \\ 2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

## Proving (in)dependence via equation solving III

- Same (in)dependence question for:  $\begin{pmatrix} 1 \\ 2 \\ -3 \end{pmatrix}, \begin{pmatrix} -2 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix}$
- With corresponding matrix:

$$\begin{pmatrix} 1 & -2 & 1 \\ 2 & 1 & -1 \\ -3 & 1 & -2 \end{pmatrix} \quad \text{reducing to} \quad \begin{pmatrix} 5 & 0 & -1 \\ 0 & 5 & -3 \\ 0 & 0 & -4 \end{pmatrix}$$

- Thus the only solution is  $a_1 = a_2 = a_3 = 0$ . The vectors are **independent!**



## Linear combinations of solutions

- It is **not** the case in general that linear combinations of solutions give solutions. For example, consider:

$$\begin{cases} x_1 + 2x_2 + x_3 = 0 \\ x_2 + x_4 = \mathbf{2} \end{cases} \quad \Leftrightarrow \quad \left( \begin{array}{cccc|c} 1 & 2 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & \mathbf{2} \end{array} \right)$$

- This has as solutions:

$$V_1 = \begin{pmatrix} -2 \\ 2 \\ -2 \\ 0 \end{pmatrix}, V_2 = \begin{pmatrix} -1 \\ 1 \\ -1 \\ 1 \end{pmatrix} \text{ but } \mathbf{not} \quad V_1 + V_2 = \begin{pmatrix} -3 \\ 3 \\ -3 \\ 1 \end{pmatrix}, 3V_1, \dots$$

- The problem is this system of equations is not **homogeneous**, because the **2** on the right-hand-side (RHS) of the second equation.

# Homogeneous systems of equations

## Definition

A system of equations is called **homogeneous** if it has **zeros** on the RHS of every equation. Otherwise it is called **non-homogeneous**.

- We can always squash a non-homogeneous system to a homogeneous one:

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

- The solutions will change!
- ...but they are still related. We'll see how that works soon.

# Homogeneous and non-homogeneous, illustration

## Example

A non-homogeneous system:  $\begin{cases} x_1 + x_2 = 3 \\ x_1 - x_2 = 0 \end{cases}$

can be made homogeneous, namely as:  $\begin{cases} x_1 + x_2 = 0 \\ x_1 - x_2 = 0 \end{cases}$

In terms of **matrices**, this means going from:

$$\left( \begin{array}{cc|c} 1 & 1 & 3 \\ 1 & -1 & 0 \end{array} \right) \text{ to } \left( \begin{array}{cc|c} 1 & 1 & 0 \\ 1 & -1 & 0 \end{array} \right) \text{ i.e. to } \left( \begin{array}{cc} 1 & 1 \\ 1 & -1 \end{array} \right)$$

## Zero solution, in homogeneous case

### Lemma

Each homogeneous equation has  $(0, \dots, 0)$  as solution.

**Proof:** A homogeneous system looks like this

$$\begin{aligned} a_{11}x_1 + \dots + a_{1n}x_n &= 0 \\ &\vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n &= 0 \end{aligned}$$

Consider the equation at row  $i$ :

$$a_{i1}x_1 + \dots + a_{in}x_n = 0$$

Clearly it has as **solution**  $x_1 = x_2 = \dots = x_n = 0$ .

This holds for each row  $i$ .



# Linear combinations of solutions

## Theorem

The *set of solutions* of a *homogeneous* system is closed under linear combinations (i.e. addition and scalar multiplication of vectors).  
(i.e. *the solutions form a linear subspace*)

...which means:

- if  $(s_1, s_2, \dots, s_n)$  and  $(t_1, t_2, \dots, t_n)$  are solutions, then so is:  
 $(s_1 + t_1, s_2 + t_2, \dots, s_n + t_n)$ , and
- if  $(s_1, s_2, \dots, s_n)$  is a solution, then so is  $(c \cdot s_1, c \cdot s_2, \dots, c \cdot s_n)$

## Example

- Consider the homogeneous system 
$$\begin{cases} 3x_1 + 2x_2 - x_3 = 0 \\ x_1 - x_2 = 0 \end{cases}$$

- A solution is  $x_1 = 1, x_2 = 1, x_3 = 5$ , written as vector  $(x_1, x_2, x_3) = (1, 1, 5)$

- Another solution is  $(2, 2, 10)$

- Addition** yields another solution:

$$(1, 1, 5) + (2, 2, 10) = (1 + 2, 1 + 2, 5 + 10) = (3, 3, 15).$$

- Scalar multiplication** also gives solutions:

$$-1 \cdot (1, 1, 5) = (-1 \cdot 1, -1 \cdot 1, -1 \cdot 5) = (-1, -1, -5)$$

$$100 \cdot (2, 2, 10) = (100 \cdot 2, 100 \cdot 2, 100 \cdot 10) = (200, 200, 1000)$$

$$c \cdot (1, 1, 5) = (c \cdot 1, c \cdot 1, c \cdot 5) = (c, c, 5c)$$

(is a solution for every  $c$ )

## Proof of closure under addition

- Consider an equation  $a_1x_1 + \cdots + a_nx_n = 0$
- Assume two solutions  $(s_1, \dots, s_n)$  and  $(t_1, \dots, t_n)$
- Then  $(s_1 + t_1, \dots, s_n + t_n)$  is also a solution since:

$$\begin{aligned} & a_1(s_1 + t_1) + \cdots + a_n(s_n + t_n) \\ &= (a_1s_1 + a_1t_1) + \cdots + (a_ns_n + a_nt_n) \\ &= (a_1s_1 + \cdots + a_ns_n) + (a_1t_1 + \cdots + a_nt_n) \\ &= 0 + 0 \quad \text{since the } s_i \text{ and } t_i \text{ are solutions} \\ &= 0. \end{aligned}$$

- **Exercise:** do a similar proof of closure under scalar multiplication

# General solution of a homogeneous system

## Theorem

Every solution to a homogeneous system arises from a *general solution* of the form:

$$(s_1, \dots, s_n) = c_1(v_{11}, \dots, v_{1n}) + \dots + c_k(v_{k1}, \dots, v_{kn})$$

for some numbers  $c_1, \dots, c_k \in \mathbb{R}$ .

We call this a **parametrization** of our solution space. It means:

- 1 There is a fixed set of vectors (called **basic solutions**):

$$V_1 = (v_{11}, \dots, v_{1n}), \quad \dots, \quad V_k = (v_{k1}, \dots, v_{kn})$$

- 2 such that every solution  $S$  is a linear combination of  $V_1, \dots, V_k$ .
- 3 That is, there exist  $c_1, \dots, c_k \in \mathbb{R}$  such that

$$S = c_1 V_1 + \dots c_k V_k$$



## Basic solutions of a homogeneous system

### Theorem

*Suppose a homogeneous system of equations in  $n$  variables has  $p \leq n$  pivots. Then there are  $n - p$  **basic solutions**  $V_1, \dots, V_{n-p}$ . This means that the general solution  $S$  can be written as a parametrization:*

$$S = c_1 V_1 + \dots + c_{n-p} V_{n-p}.$$

*Moreover, for any solution  $S$ , the scalars  $c_1, \dots, c_{n-p}$  are unique.*

## Finding basic solutions

- We have two kinds of variables, **pivot variables** and non-pivot, or **free variables**, depending on whether their column has a pivot:

$$\begin{array}{ccccc} x_1 & x_2 & x_3 & x_4 & x_5 \\ \left( \begin{array}{ccccc} \boxed{1} & 0 & 1 & 4 & 1 \\ 0 & 0 & \boxed{1} & 2 & 0 \end{array} \right) \end{array}$$

- The Echelon form lets us (easily) write pivot variables in terms of non-pivot variables, e.g.:

$$\begin{cases} x_1 = -x_3 - 4x_4 - x_5 \\ x_3 = -2x_4 \end{cases} \Rightarrow \begin{cases} x_1 = -2x_4 - x_5 \\ x_3 = -2x_4 \end{cases}$$

- We can find a (non-zero) **basic solution** by setting exactly **one** free variable to 1 and the rest to 0.

## Finding basic solutions

$$\begin{matrix} x_1 & x_2 & x_3 & x_4 & x_5 \\ \left( \begin{array}{ccccc} \boxed{1} & 0 & 1 & 4 & 1 \\ 0 & 0 & \boxed{1} & 2 & 0 \end{array} \right) \Rightarrow \begin{cases} x_1 = -2x_4 - x_5 \\ x_3 = -2x_4 \end{cases} \end{matrix}$$

5 variables and 2 pivots gives us  $5 - 2 = 3$  basic solutions:

$$\begin{matrix} x_2 := 1 & x_2 := 0 & x_2 := 0 \\ x_4 := 0 & x_4 := 1 & x_4 := 0 \\ x_5 := 0 & x_5 := 0 & x_5 := 1 \end{matrix}$$

$$\begin{pmatrix} -2x_4 - x_5 \\ x_2 \\ -2x_4 \\ x_4 \\ x_5 \end{pmatrix} \rightsquigarrow \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -2 \\ 0 \\ -2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

## Generic Solution

Now, any solution to the system is obtainable as a linear combination of basic solutions:

$$x_2 \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} -2 \\ 0 \\ -2 \\ 1 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -2x_4 - x_5 \\ x_2 \\ -2x_4 \\ x_4 \\ x_5 \end{pmatrix}$$

Picking solutions this way guarantees **linear independence**.

## Finding basic solutions: technique 2

- Keep all columns with a pivot,
- One-by-one, keep only the  $i$ -th non-pivot columns (while removing the others), and find a (non-zero) solution
- (this is like setting all the other free variables to zero)
- Add 0's to each solution to account for the columns (i.e. free variables) we removed

## General solution and basic solutions, example

- For the matrix:  $\begin{pmatrix} \boxed{1} & 1 & 0 & 4 \\ 0 & 0 & \boxed{2} & 2 \end{pmatrix}$
- There are 4 columns (variables) and 2 pivots, so  $4 - 2 = 2$  basic solutions
- First keep only the first non-pivot column:

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix} \quad \text{with chosen solution} \quad (x_1, x_2, x_3) = (1, -1, 0)$$

- Next keep only the second non-pivot column:

$$\begin{pmatrix} 1 & 0 & 4 \\ 0 & 2 & 2 \end{pmatrix} \quad \text{with chosen solution} \quad (x_1, x_3, x_4) = (4, 1, -1)$$

- The general 4-variable solution is now obtained as:  
$$c_1 \cdot (1, -1, 0, 0) + c_2 \cdot (4, 0, 1, -1)$$

## General solutions example, check

We double-check that we have solutions:

$$\begin{aligned}c_1 \cdot (4, 0, 1, -1) + c_2 \cdot (1, -1, 0, 0) \\&= (4 \cdot c_1, 0, 1 \cdot c_1, -1 \cdot c_1) + (1 \cdot c_2, -1 \cdot c_2, 0, 0) \\&= (4c_1 + c_2, -c_2, c_1, -c_1)\end{aligned}$$

is a solution of:

$$\begin{pmatrix} 1 & 1 & 0 & 4 \\ 0 & 0 & 2 & 2 \end{pmatrix} \quad \text{i.e. of} \quad \begin{cases} x_1 + x_2 + 4x_4 = 0 \\ 2x_3 + 2x_4 = 0 \end{cases}$$

Just fill in  $x_1 = 4c_1 + c_2$ ,  $x_2 = -c_2$ ,  $x_3 = c_1$ ,  $x_4 = -c_1$

$$\begin{aligned}(4c_1 + c_2) - c_2 + 4 \cdot -c_1 &= 0 \\ 2c_1 - 2c_1 &= 0\end{aligned} \quad \checkmark$$

## Summary of homogeneous systems

Given a homogeneous system in  $n$  variables:

- A **basic solution** is a **non-zero** solution of the system.
- If there are  $n$  pivots in its echelon form, there is no basic solution (but only the  $(0, \dots, 0)$  solution).
- Basic solutions are not unique. For instance, if  $V_1$  and  $V_2$  give basic solutions, so do  $V_1 + V_2$  and  $V_1 - V_2$ .
- If there are  $p < n$  pivots in its Echelon form, it has  $n - p$  **linearly independent** basic solutions.



## Non-homogeneous case: subtracting solutions

### Theorem

The *difference* of two solutions of a **non-homogeneous** system is a solution for the associated **homogeneous** system.

More explicitly: given two solutions  $(s_1, \dots, s_n)$  and  $(t_1, \dots, t_n)$  of an **non-homogeneous** system, the difference  $(s_1 - t_1, \dots, s_n - t_n)$  is a solution of the associated **homogeneous** system.

**Proof:** Let  $a_1x_1 + \dots + a_nx_n = b$  be the equation. Then:

$$\begin{aligned} & a_1(s_1 - t_1) + \dots + a_n(s_n - t_n) \\ &= (a_1s_1 - a_1t_1) + \dots + (a_ns_n - a_nt_n) \\ &= (a_1s_1 + \dots + a_ns_n) - (a_1t_1 + \dots + a_nt_n) \\ &= b - b \quad \text{since the } s_i \text{ and } t_i \text{ are solutions} \\ &= 0. \end{aligned}$$



# General solution for non-homogeneous systems

## Theorem

Assume a non-homogeneous system has a solution given by the vector  $P$ , which we call a *particular solution*.

Then any other solution  $S$  of the non-homogeneous system can be written as

$$S = P + H$$

where  $H$  is a solution of the associated homogeneous system.

**Proof:** Let  $S$  be a solution of the non-homogeneous system. Then  $H = S - P$  is a solution of the associated homogeneous system. Hence we can write  $S$  as  $P + H$ , for  $H$  some solution of the associated homogeneous system.



## Example: solutions of a non-homogeneous system

- Consider the non-homogeneous system 
$$\begin{cases} x + y + 2z = 9 \\ y - 3z = 4 \end{cases}$$
- with solutions:  $(0, 7, 1)$  and  $(5, 4, 0)$
- We can write  $(0, 7, 1)$  as:  $(5, 4, 0) + (-5, 3, 1)$
- where:
  - $P = (5, 4, 0)$  is a **particular solution** (of the original system)
  - $(-5, 3, 1)$  is a solution of the associated **homogeneous** system:
$$\begin{cases} x + y + 2z = 0 \\ y - 3z = 0 \end{cases}$$
- Similarly,  $(10, 1, -1)$  is a solution of the non-homogeneous system and

$$(10, 1, -1) = (5, 4, 0) + (5, -3, -1)$$

- where:
  - $(5, -3, -1)$  is a solution of the associated **homogeneous** system.

# General solution for non-homogeneous systems, concretely

## Theorem

The general solution of a non-homogeneous system of equations in  $n$  variables is given by a *parametrization* as follows:

$$(s_1, \dots, s_n) = (p_1, \dots, p_n) + c_1(v_{11}, \dots, v_{1n}) + \dots + c_k(v_{k1}, \dots, v_{kn})$$

for  $c_1, \dots, c_k \in \mathbb{R}$ ,

where

- $(p_1, \dots, p_n)$  is a *particular solution*
- $(v_{11}, \dots, v_{1n}), \dots, (v_{k1}, \dots, v_{kn})$  are *basic solutions* of the associated homogeneous system.
- So  $c_1(v_{11}, \dots, v_{1n}) + \dots + c_k(v_{k1}, \dots, v_{kn})$  is a general solution for the associated homogeneous system.

## Elaborated example, part I

- Consider the **non-homogeneous** system of equations given by the augmented matrix in echelon form:

$$\left( \begin{array}{ccccc|c} \boxed{1} & 1 & 1 & 1 & 1 & 3 \\ 0 & 0 & \boxed{1} & 2 & 3 & 1 \\ 0 & 0 & 0 & 0 & \boxed{1} & 4 \end{array} \right)$$

- It has 5 variables, 3 pivots, and thus  $5 - 3 = 2$  basic solutions
- To find a **particular solution**, remove the non-pivot columns, and (uniquely!) solve the resulting system:

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

- This has  $(10, -11, 4)$  as solution; the original 5-variable system then has particular solution  $(10, 0, -11, 0, 4)$ .

## Elaborated example, part II

- Consider the **associated homogeneous** system of equations:

$$\begin{pmatrix} \boxed{1} & 1 & 1 & 1 & 1 \\ 0 & 0 & \boxed{1} & 2 & 3 \\ 0 & 0 & 0 & 0 & \boxed{1} \end{pmatrix}$$

- The two **basic solutions** are found by removing each of the two non-pivot columns separately, and choosing solutions:

$$\begin{pmatrix} \boxed{1} & 1 & 1 & 1 \\ 0 & \boxed{1} & 2 & 3 \\ 0 & 0 & 0 & \boxed{1} \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} \boxed{1} & 1 & 1 & 1 \\ 0 & 0 & \boxed{1} & 3 \\ 0 & 0 & 0 & \boxed{1} \end{pmatrix}$$

- We can choose:  $(1, -2, 1, 0)$  and  $(-1, 1, 0, 0)$ , giving for the original matrix:  $(1, 0, -2, 1, 0)$  and  $(-1, 1, 0, 0, 0)$ .

## Elaborated example, part III

Wrapping up: **all solutions** of the system

$$\left( \begin{array}{ccccc|c} \boxed{1} & 1 & 1 & 1 & 1 & 3 \\ 0 & 0 & \boxed{1} & 2 & 3 & 1 \\ 0 & 0 & 0 & 0 & \boxed{1} & 4 \end{array} \right)$$

are of the form:

$$\underbrace{(10, 0, -11, 0, 4)}_{\text{particular sol.}} + \underbrace{c_1(1, 0, -2, 1, 0) + c_2(-1, 1, 0, 0, 0)}_{\text{two basic solutions}}.$$

This is the **general solution** of the non-homogeneous system.