



Matrix Calculation: Solutions of Systems of Linear Equations

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Outline

Solutions and solvability

Vectors and linear combinations

Homogeneous systems





Solutions

When we look for solutions to a system, there are 3 possibilities:

- ① A system of equations has a *single, unique solution*, e.g.

$$x_1 + x_2 = 3$$

$$x_1 - x_2 = 1$$

(unique solution: $x_1 = 2, x_2 = 1$)

- ② A system has *many solutions*, e.g.

$$x_1 - 2x_2 = 1$$

$$-2x_1 + 4x_2 = -2$$

(we have a solution whenever: $x_1 = 1 + 2x_2$)

- ③ A system has *no solutions*.

$$3x_1 - 2x_2 = 1$$

$$6x_1 - 4x_2 = 6$$

(the transformation $E_2 := E_2 - 2E_1$ yields $0 = 4$.)





Solutions, geometrically

Consider systems of only two variables x, y . A linear equation $ax + by = c$ then describes a line in the plane.

For 2 such equations/lines, there are **three** possibilities:

- 1 the lines intersect in a **unique point**, which is the solution to both equations
- 2 the lines are **parallel**, in which case there are no joint solutions
- 3 the lines **coincide**, giving many joint solutions.



Echelon form

We can tell the difference in these 3 cases by writing the augmented matrix and transforming to Echelon form.

Recall: 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

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



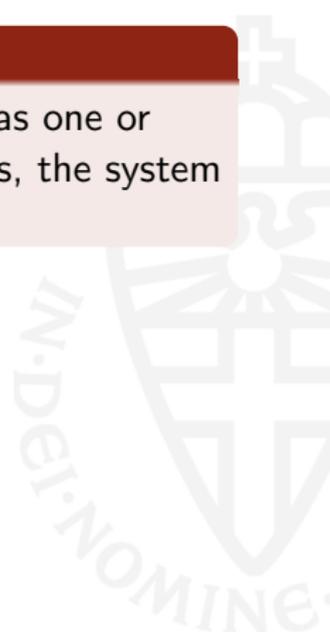
(In)consistent systems

Definition

A system of equations is **consistent** (*oplosbaar*) if it has one or more solutions. Otherwise, when there are no solutions, the system is called **inconsistent**

Thus, for a system of equations:

nr. of solutions	terminology
0	inconsistent
≥ 1 (one or many)	consistent





Inconsistency and echelon forms

Theorem

A system of equations is *inconsistent* (non-solvable) if and only if in the echelon form of its augmented matrix there is a row with:

- only zeros before the bar |
- a non-zero after the bar |,

as in: $0\ 0\ \cdots\ 0\ |\ c$, where $c \neq 0$.

Example

$$\begin{aligned} 3x_1 - 2x_2 &= 1 \\ 6x_1 - 4x_2 &= 6 \end{aligned} \quad \text{gives} \quad \left(\begin{array}{cc|c} 3 & -2 & 1 \\ 6 & -4 & 6 \end{array} \right) \quad \text{and} \quad \left(\begin{array}{cc|c} 3 & -2 & 1 \\ 0 & 0 & 4 \end{array} \right)$$

(using the transformation $R_2 := R_2 - 2R_1$)



Unique solutions

Theorem

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

Proof. (n pivots \implies unique soln., on board)

In summary: A system with n variables has an augmented matrix with n columns before the line. Its Echelon form has n pivots, so there must be *exactly* one pivot in each column. The last pivot uniquely fixes x_n . Then, since x_n is fixed, the second to last pivot uniquely fixes x_{n-1} and so on.



Unique solutions: earlier example

equations

$$\begin{aligned}2x_2 + x_3 &= -2 \\3x_1 + 5x_2 - 5x_3 &= 1 \\2x_1 + 4x_2 - 2x_3 &= 2\end{aligned}$$

matrix

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

After various transformations leads to

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

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

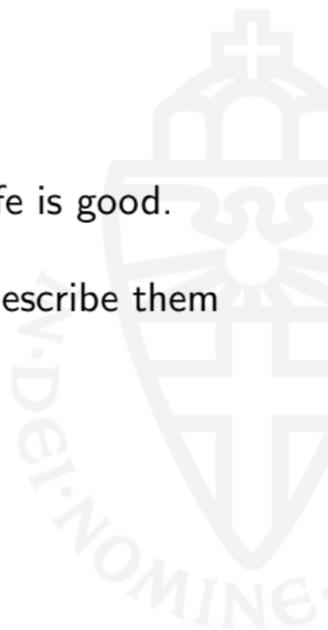
Echelon
formThere are **3 variables** and **3 pivots**, so there is **one unique solution**.



Unique solutions

So, when there are n pivots, there is 1 solution, and life is good.

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





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 $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$ is called **linearly independent** if no vector can be written as a linear combination of the others.



Linear independence

- These vectors:

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \quad \mathbf{v}_3 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

are **NOT** linearly independent, because $\mathbf{v}_3 = \mathbf{v}_1 + \mathbf{v}_2$.

- These vectors:

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \quad \mathbf{v}_3 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$$

are **NOT** linearly independent, because $\mathbf{v}_1 = \mathbf{v}_2 + 2 \cdot \mathbf{v}_3$.



Linear independence

- These vectors:

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \mathbf{v}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

are linearly *independent*. There is no way to write any of them in terms of each other.

- These vectors:

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \mathbf{v}_3 = \begin{pmatrix} 0 \\ 2 \\ 2 \end{pmatrix}$$

are linearly *independent*. There is no way to write any of them in terms of each other.



Linear independence

- These vectors:

$$\mathbf{v}_1 = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \quad \mathbf{v}_2 = \begin{pmatrix} 2 \\ -1 \\ 4 \end{pmatrix} \quad \mathbf{v}_3 = \begin{pmatrix} 0 \\ 5 \\ 2 \end{pmatrix}$$

are... ???

- 'Eyeballing' vectors works sometimes, but we need a better way of checking linear independence!





Checking linear independence

Theorem

Vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly independent if and only if, for all numbers $a_1, \dots, a_n \in \mathbb{R}$ one has:

$$a_1 \cdot \mathbf{v}_1 + \dots + a_n \cdot \mathbf{v}_n = \mathbf{0} \text{ implies } a_1 = a_2 = \dots = a_n = 0$$

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,

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



Checking linear independence

Theorem

Vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly independent if and only if, for all numbers $a_1, \dots, a_n \in \mathbb{R}$ one has:

$$a_1 \cdot \mathbf{v}_1 + \dots + a_n \cdot \mathbf{v}_n = \mathbf{0} \text{ implies } a_1 = a_2 = \dots = a_n = 0$$

Proof. Another way to say the theorem is $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly **dependent** if and only if:

$$a_1 \cdot \mathbf{v}_1 + a_2 \cdot \mathbf{v}_2 + \dots + a_n \cdot \mathbf{v}_n = \mathbf{0}$$

where some a_j are non-zero. If this is true and $a_1 \neq 0$, then:

$$\mathbf{v}_1 = (-a_2/a_1) \cdot \mathbf{v}_2 + \dots + (-a_n/a_1) \cdot \mathbf{v}_n$$

The vectors are **dependent** (also works for any other non-zero a_j).

Exercise: prove the other direction.



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:

$$\begin{cases} a_1 + 2a_2 = 0 \\ 2a_1 - a_2 + 5a_3 = 0 \\ 3a_1 + 4a_2 + 2a_3 = 0 \end{cases} \quad \text{corresponding to} \quad \begin{pmatrix} 1 & 2 & 0 \\ 0 & -1 & 1 \\ 0 & 0 & 0 \end{pmatrix}$$

(in Echelon form)

- This has only 2 pivots, so **multiple solutions**. In particular, it 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:

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



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 independence: summary

To check linear independence of $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$:

- 1 Write the vectors as the *columns of a matrix*
- 2 Convert to Echelon form
- 3 Count the pivots
 - ($\#$ pivots) = ($\#$ columns) means **independent**
 - ($\#$ pivots) < ($\#$ columns) means **dependent**
- 4 Non-zero solutions show linear dependence explicitly, e.g.

$$\mathbf{v}_1 - 2\mathbf{v}_2 + \mathbf{v}_3 = \mathbf{0} \quad \implies \quad \mathbf{v}_1 = 2\mathbf{v}_2 - \mathbf{v}_3$$



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:

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



Linear combinations of solutions

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

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

- This has as solutions:

$$\mathbf{v}_1 = \begin{pmatrix} -2 \\ 2 \\ -2 \\ 0 \end{pmatrix}, \mathbf{v}_2 = \begin{pmatrix} -1 \\ 1 \\ -1 \\ 1 \end{pmatrix} \text{ but not } \mathbf{v}_1 + \mathbf{v}_2 = \begin{pmatrix} -3 \\ 3 \\ -3 \\ 1 \end{pmatrix}, 3 \cdot \mathbf{v}_1, \dots$$

- The problem is this system of equations is not **homogeneous**, because the 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.



Zero solution, in homogeneous case

Lemma

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

Proof: A homogeneous system looks like this

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

Consider the equation at row i :

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

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

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).

...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:

$$\begin{aligned} -1 \cdot (1, 1, 5) &= (-1 \cdot 1, -1 \cdot 1, -1 \cdot 5) = (-1, -1, -5) \\ 100 \cdot (2, 2, 10) &= (100 \cdot 2, 100 \cdot 2, 100 \cdot 10) = (200, 200, 1000) \\ c \cdot (1, 1, 5) &= (c \cdot 1, c \cdot 1, c \cdot 5) = (c, c, 5c) \\ &\text{(is a solution for every } c) \end{aligned}$$



Proof of closure under addition

- Consider an equation $a_1x_1 + \dots + a_nx_n = 0$
- 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:

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

- **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**):

$$\mathbf{v}_1 = (v_{11}, \dots, v_{1n}), \quad \dots, \quad \mathbf{v}_k = (v_{k1}, \dots, v_{kn})$$

- 2 such that every solution \mathbf{s} is a linear combination of $\mathbf{v}_1, \dots, \mathbf{v}_k$.
- 3 That is, there exist $c_1, \dots, c_k \in \mathbb{R}$ such that

$$\mathbf{s} = c_1 \mathbf{v}_1 + \dots + c_k \mathbf{v}_k$$



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** $\mathbf{v}_1, \dots, \mathbf{v}_{n-p}$.

This means that the general solution \mathbf{s} can be written as a parametrization:

$$\mathbf{s} = c_1 \mathbf{v}_1 + \dots + c_{n-p} \mathbf{v}_{n-p}.$$

Moreover, for any solution \mathbf{s} , the scalars c_1, \dots, c_{n-p} are unique.

$$(p = n) \Leftrightarrow (\text{no basic solns.}) \Leftrightarrow (\mathbf{0} \text{ is the unique soln.})$$



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

$$\begin{pmatrix} \boxed{1} & 0 & 1 & 4 & 1 \\ 0 & 0 & \boxed{1} & 2 & 0 \end{pmatrix} \Rightarrow \begin{cases} x_1 = -2x_4 - x_5 \\ x_3 = -2x_4 \end{cases}$$

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

$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \begin{pmatrix} -2x_4 - x_5 \\ x_2 \\ -2x_4 \\ x_4 \\ x_5 \end{pmatrix} \sim \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -2 \\ 0 \\ -2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$x_2 := 1 \quad x_2 := 0 \quad x_2 := 0$
 $x_4 := 0 \quad x_4 := 1 \quad x_4 := 0$
 $x_5 := 0 \quad x_5 := 0 \quad x_5 := 1$



General Solution

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

$$a \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + b \begin{pmatrix} -2 \\ 0 \\ -2 \\ 1 \\ 0 \end{pmatrix} + c \begin{pmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -2b - c \\ a \\ -2b \\ b \\ c \end{pmatrix}$$

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



General Solution

Since the variable names don't matter, we could use instead:

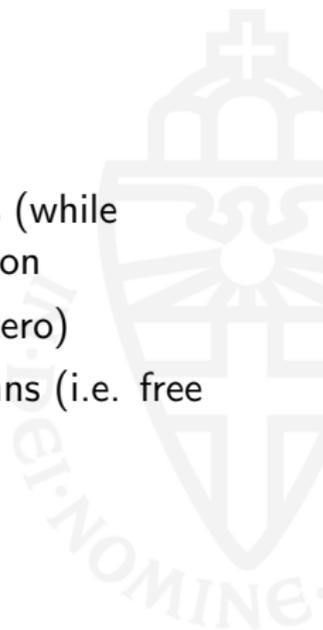
$$x_2 \begin{pmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} -2 \\ 0 \\ -2 \\ 1 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} -1 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} -2x_4 - x_5 \\ x_2 \\ -2x_4 \\ x_4 \\ x_5 \end{pmatrix}$$

...which gives us the vector from 2 slides ago.



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:

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

- Next keep only the second non-pivot column:

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

- 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 any vector:

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

gives 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, so only $\mathbf{0} = (0, \dots, 0)$ is a solution.
- Basic solutions are not unique. For instance, if \mathbf{v}_1 and \mathbf{v}_2 give basic solutions, so do $\mathbf{v}_1 + \mathbf{v}_2$, $\mathbf{v}_1 - \mathbf{v}_2$, and any other linear combination.
- If there are $p < n$ pivots in its Echelon form, it has $n - p$ **linearly independent** basic solutions.