

# MATH1231 Algebra, 2017

## Chapter 7: Linear maps

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## Chapter overview

- $f : \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $f\begin{pmatrix} x \\ y \end{pmatrix} = 2x - 3y$  is example of a linear function,  $g\begin{pmatrix} x \\ y \end{pmatrix} = x^2 - 5y$  is not.
- In this chapter, study more generally **linear transformations**  $T : \mathbb{R}^m \rightarrow \mathbb{R}^n$ .
- Even more gen, study linear  $T : V \rightarrow W$  where  $V, W =$  vector spaces  $/\mathbb{F}$ . Recall  $V$  is the **domain** of  $T$  &  $W$  the **codomain** of  $T$ .
- We'll generalise systems of linear equations  $A\mathbf{x} = \mathbf{b}$  to “linear equations” of form  $T\mathbf{x} = \mathbf{b}$  where  $\mathbf{b} \in W, \mathbf{x} \in V$ .

Often abbrev  $T(\mathbf{x}) = T\mathbf{x}$ .

# Addition Condition

To define linear map, first consider

## Addition Condition.

We say  $T : V \rightarrow W$  satisfies the **addition condition**, if

$$T(\mathbf{v} + \mathbf{v}') = T(\mathbf{v}) + T(\mathbf{v}') \text{ for all } \mathbf{v}, \mathbf{v}' \in V.$$

**E.g.**  $T : \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $T\begin{pmatrix} x \\ y \end{pmatrix} = 2x - 3y$  satisfies the addn condn since given  $\begin{pmatrix} x \\ y \end{pmatrix}, \begin{pmatrix} x' \\ y' \end{pmatrix} \in \mathbb{R}^2$

$$\begin{aligned} T\left(\begin{pmatrix} x \\ y \end{pmatrix} + \begin{pmatrix} x' \\ y' \end{pmatrix}\right) &= \\ T\begin{pmatrix} x \\ y \end{pmatrix} + T\begin{pmatrix} x' \\ y' \end{pmatrix} &= \end{aligned}$$

**Warning** The  $+$  on the two sides of the equation are different!

# Scalar Multiplication Condition

## Scalar Multiplication Condition.

We say  $T : V \rightarrow W$  satisfies the **scalar multiplication condition**, if

$$T(\lambda \mathbf{v}) = \lambda T(\mathbf{v}) \text{ for all } \lambda \in \mathbb{F} \text{ and } \mathbf{v} \in V.$$

**E.g.**  $T : \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $T \begin{pmatrix} x \\ y \end{pmatrix} = 2x - 3y$  satisfies the scalar multn condn since given  $\begin{pmatrix} x \\ y \end{pmatrix} \in \mathbb{R}^2, \lambda \in \mathbb{R}$

$$T \left( \lambda \begin{pmatrix} x \\ y \end{pmatrix} \right) =$$

$$\lambda T \begin{pmatrix} x \\ y \end{pmatrix} =$$

**Warning** The scalar multn on the two sides of the eqn are different!

## Definition

Let  $V, W =$  vector spaces  $/\mathbb{F}$ . A function  $T : V \rightarrow W$  is called a **linear map** or a **linear transformation** if following both hold.

**Addition Condition.**

$$T(\mathbf{v} + \mathbf{v}') = T(\mathbf{v}) + T(\mathbf{v}') \text{ for all } \mathbf{v}, \mathbf{v}' \in V, \text{ and}$$

**Scalar Multiplication Condition.**

$$T(\lambda \mathbf{v}) = \lambda T(\mathbf{v}) \text{ for all } \lambda \in \mathbb{F} \text{ and } \mathbf{v} \in V.$$

**E.g.**  $T : \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $T\begin{pmatrix} x \\ y \end{pmatrix} = 2x - 3y$  is linear.

## Sample question: showing a function is linear.

### Example

Show that the function  $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$  defined by

$$T(\mathbf{x}) = \begin{pmatrix} 4x_2 - 3x_3 \\ x_1 + 2x_2 \end{pmatrix} \quad \text{for } \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \in \mathbb{R}^3$$

is a linear map.

### Solution

## Solution (Continued)

## Solution (Continued)



## Linear maps preserve zero.

### Proposition.

If  $T : V \rightarrow W$  is a linear map, then  $T(\mathbf{0}) = \mathbf{0}$ .

**Proof.**  $T(\mathbf{0}) = T(0\mathbf{0}) = 0T(\mathbf{0}) = \mathbf{0}$ .

### Example

Show that the function  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  defined by  $T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_2 - 2 \\ x_1 \end{pmatrix}$  is not linear.

### Soln

## Another non-linear example

### Example

Show that the function  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by

$$T \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 + x_2 \\ x_2^2 \end{pmatrix}$$

is not linear.

### Solution

*Checking  $T(\mathbf{0}) = \mathbf{0}$  here tells you nothing about linearity.*

*Suffice check that  $T(\lambda\mathbf{v}) = \lambda T\mathbf{v}$  fails for single choice of pair  $\lambda, \mathbf{v}$ .*

# Alternate characterisation of linearity

We can combine the addn condn & scalar multn condn into one!

## Theorem

*The function  $T : V \rightarrow W$  is a linear map iff for all  $\lambda \in \mathbb{F}$  and  $\mathbf{v}_1, \mathbf{v}_2 \in V$*

$$T(\lambda\mathbf{v}_1 + \mathbf{v}_2) = \lambda T(\mathbf{v}_1) + T(\mathbf{v}_2).$$

**Remark** This means that a linear map  $T$  has the special property that it sends the line  $\mathbf{x} = \mathbf{a} + \lambda\mathbf{v}$  to the line  $\mathbf{x} = T\mathbf{a} + \lambda T\mathbf{v}$  or point  $T\mathbf{a}$  if  $T\mathbf{v} = \mathbf{0}$ .

**E.g.** Differentiation is a linear map. More precisely, we define  $T : \mathbb{P} \rightarrow \mathbb{P}$  by  $Tp = \frac{dp}{dx}$ . Then for  $p, q \in \mathbb{P}, \lambda \in \mathbb{R}$

$$T(\lambda p + q) =$$

## Theorem

If  $T : V \rightarrow W$  is linear map,  $\mathbf{v}_1, \dots, \mathbf{v}_n \in V$  &  $\lambda_1, \dots, \lambda_n$  are scalars, then

$$T(\lambda_1 \mathbf{v}_1 + \dots + \lambda_n \mathbf{v}_n) = \lambda_1 T(\mathbf{v}_1) + \dots + \lambda_n T(\mathbf{v}_n).$$

## Example

If  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  a function such that

$$T \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \quad T \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \quad T \begin{pmatrix} 3 \\ 1 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \end{pmatrix}.$$

Show that  $T$  is not linear.

## Solution

## Solution (Continued)

## Example

Given that  $T$  is a linear map and

$$T \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad T \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \quad T \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \end{pmatrix}.$$

Find  $T \begin{pmatrix} x \\ y \\ z \end{pmatrix}$ .

## Solution

# Linear maps are determined by the values on a spanning set

Previous eg illustrates the following general result.

## Theorem

*Let  $T : V \rightarrow W$  be linear &  $V = \text{span}(\mathbf{v}_1, \dots, \mathbf{v}_m)$ . Then  $T$  is completely determined by the  $m$  values  $T\mathbf{v}_1, \dots, T\mathbf{v}_m$ .*

Compare with the following:

An affine linear function  $f(x) = mx + b$  is determined by two of its values  $f(x_1), f(x_2)$ , since its graph is a line which is determined by two points.

# Matrices define Linear Maps

## Theorem

For each  $m \times n$  matrix  $A$ , the function  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , defined by

$$T_A(\mathbf{x}) = A\mathbf{x} \quad \text{for } \mathbf{x} \in \mathbb{R}^n,$$

is a linear map called the **associated linear map**.

Proof.





## Example of reflection

### Example

Let  $A = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$ , describe the associated linear map  $T_A$  geometrically as a mapping  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$ .

### Solution

# Matrix Representation Theorem

Conversely, given linear  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , we can find an  $m \times n$  matrix  $A$  such that  $T(\mathbf{x}) = A\mathbf{x}$  for all  $\mathbf{x} \in \mathbb{R}^n$ . In this case, we say  $A$  is a matrix **representing**  $T$ .

## Example

Given that  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  defined by  $T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x + 2y \\ 2x - y \\ y \end{pmatrix}$  is linear. Find the matrix  $A$  representing  $T$ .

## Solution

# Formula for representing matrix

## Theorem

Let  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear map and let the vectors  $\mathbf{e}_j$  for  $1 \leq j \leq n$  be the standard basis vectors for  $\mathbb{R}^n$ . Then the  $m \times n$  matrix

$$A = (T\mathbf{e}_1 | T\mathbf{e}_2 | \dots | T\mathbf{e}_n)$$

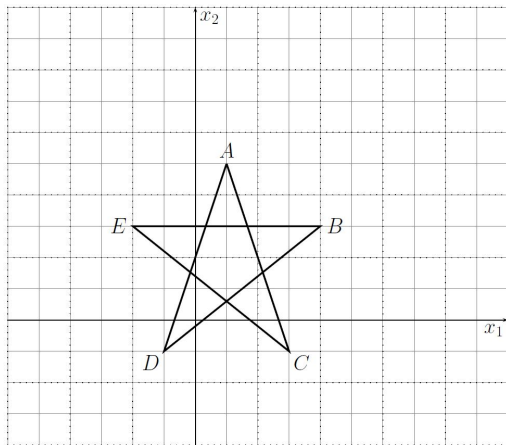
has the property that

$$T(\mathbf{x}) = A\mathbf{x} \text{ for all } \mathbf{x} \in \mathbb{R}^n.$$

**E.g.** In the example of the previous slide,

$T\mathbf{e}_1 = \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix}$ ,  $T\mathbf{e}_2 = \begin{pmatrix} 2 \\ -1 \\ 1 \end{pmatrix}$  so the representing matrix is

# Stretching and Compressing



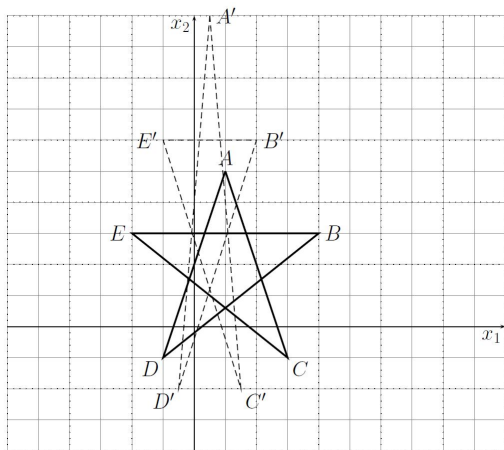
A 5-point star with vertices  
 $A(1, 5)$ ,  $B(4, 3)$ ,  $C(3, -1)$ ,  $D(-1, -1)$  and  $E(-2, 3)$ .

## Example

Find and draw the image of the 5-point star under the linear map  $T_M$  defined by the matrix  $M = \begin{pmatrix} 0.5 & 0 \\ 0 & 2 \end{pmatrix}$ .

## Solution

## Solution (Continued)



## Rotation about $\mathbf{0}$ is linear

Consider the map  $R_\alpha$ , which rotates the  $\mathbb{R}^2$  plane through an angle  $\alpha$  anticlockwise about the origin.

One can show geometrically that  $R_\alpha$  is a linear map see Section 7.3. example 3.

### Example

Find the matrix  $A$  representing  $R_\alpha$ .

### Solution

## Projection onto $\mathbf{b}$ is linear

Recall given given  $\mathbf{b} \in \mathbb{R}^n$  we have a projection map  $\text{proj}_{\mathbf{b}} : \mathbb{R}^n \rightarrow \mathbb{R}^n$  which sends  $\mathbf{x} \mapsto \text{proj}_{\mathbf{b}}\mathbf{x}$ .

### Proposition

$$\text{proj}_{\mathbf{b}}\mathbf{x} = \frac{1}{|\mathbf{b}|^2}\mathbf{b}\mathbf{b}^T\mathbf{x}$$

Hence  $\text{proj}_{\mathbf{b}}$  is linear being the linear map associated to the matrix

$$A = \frac{1}{|\mathbf{b}|^2}\mathbf{b}\mathbf{b}^T.$$

### Proof.

Note

$$A\mathbf{x} = \frac{1}{|\mathbf{b}|^2}\mathbf{b}\mathbf{b}^T\mathbf{x} = \frac{1}{|\mathbf{b}|^2}\mathbf{b}(\mathbf{b} \cdot \mathbf{x}) = \frac{\mathbf{b} \cdot \mathbf{x}}{|\mathbf{b}|^2}\mathbf{b} = \text{proj}_{\mathbf{b}}\mathbf{x}$$

from the formula for  $\text{proj}_{\mathbf{b}}\mathbf{x}$  given in MATH1131 □



# Sample projection

## Example

Let  $\mathbf{b} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $T = \text{proj}_{\mathbf{b}} : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ .

i) Find the matrix  $A$  representing  $T$ . ii) Check your answer by computing the linear map associated to the matrix  $A$  you found.

## Solution

# Kernels of linear maps

Let  $T : V \rightarrow W$  be a linear map.

## Proposition-Definition

The **kernel** of  $T$  (written  $\ker(T)$  or  $\ker T$ ) is the set,

$$\ker(T) = \{\mathbf{v} \in V \mid T(\mathbf{v}) = \mathbf{0}\} \subseteq V.$$

Let  $A \in M_{mn}(\mathbb{R})$  &  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be the assoc linear map. We define

$$\ker A = \ker T_A = \{\mathbf{v} \in \mathbb{R}^n : A\mathbf{v} = \mathbf{0}\}.$$

$\ker T$  is a subspace of  $V$ .

**E.g.** Is  $\begin{pmatrix} 1 \\ 2 \end{pmatrix}$  in  $\ker(2 \ -1)$ ?

**E.g.** Consider the differentiation map  $T : \mathbb{P} \rightarrow \mathbb{P}, (Tp)(x) = p'(x)$ .  
 $\ker T = \{p \in \mathbb{P} \mid \frac{dp}{dx} = 0\} = \mathbb{P}_0$  the subspace of all constant polynomials.

## Proof that kernels are subspaces

Let  $T : V \rightarrow W$  be a linear map. We prove that  $\ker T$  is a subspace of  $V$  by checking closure axioms.

Proof.



# Images of linear maps

Let  $T : V \rightarrow W$  be a linear map.

## Proposition-Definition

The **image** of  $T$  is the set of all function values of  $T$ , that is,

$$\text{im}(T) = \{\mathbf{w} \in W : \mathbf{w} = T(\mathbf{v}) \text{ for some } \mathbf{v} \in V\} \subseteq W.$$

Let  $A \in M_{mn}(\mathbb{R})$  &  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$  be the assoc linear map. We define

$$\text{im } A = \text{im } T_A = \{\mathbf{b} \in \mathbb{R}^m : \mathbf{b} = A\mathbf{x} \text{ for some } \mathbf{x} \in \mathbb{R}^n\} = \text{col}(A).$$

$\text{im } T$  is a subspace of  $W$ .

**Remark** Proof omitted, but note we already know  $\text{col}(A)$  is a subspace as it is the span of the columns of  $A$ .

# Verifying whether or not vectors lie in the image

## Example

Let  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 2 \end{pmatrix}$ ,  $\mathbf{b} = \begin{pmatrix} 3 \\ 1 \\ 3 \end{pmatrix}$ . Is  $\mathbf{b} \in \text{im } A$ ?

## Solution

*The question amounts to asking: Can we write  $\mathbf{b} = A\mathbf{x}$  for some  $\mathbf{x} \in \mathbb{R}^2$ ?  
i.e. Can we solve  $A\mathbf{x} = \mathbf{b}$ .*

# Finding Bases for Kernels & Images.

## Example

Let  $\begin{pmatrix} 1 & 2 & 3 & 1 \\ 2 & 4 & 7 & 1 \\ 1 & 2 & 2 & 2 \end{pmatrix}$ . Find bases for  $\ker(A)$  and  $\text{im}(A) = \text{col}(A)$ .

## Solution

$$\begin{pmatrix} 1 & 2 & 3 & 1 \\ 2 & 4 & 7 & 1 \\ 1 & 2 & 2 & 2 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 3 & 1 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & -1 & 1 \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 2 & 3 & 1 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{pmatrix} = U$$

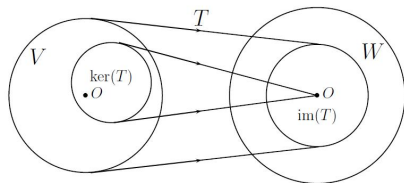
*The row echelon form  $U$  has first & third columns leading.*

## Solution (Continued)

## Solution (Continued)



# Why study kernels & image?



$\text{im } T$  tells you about existence of solutions.  $T\mathbf{x} = \mathbf{b}$  has a solution iff  $\mathbf{b} \in \text{im } T$ .

$\ker T$  tells you about uniqueness of solutions.  $T\mathbf{x} = \mathbf{0}$  has unique solution  $\mathbf{x} = \mathbf{0}$  iff  $\ker T = \mathbf{0}$ . We'll see later, that it also tells you about solutions to  $T\mathbf{x} = \mathbf{b}$ .

# Rank and Nullity

Let  $T : V \rightarrow W$  be a linear map &  $A$  a matrix with associated linear map  $T_A$ .

## Definition

- The **nullity** of  $T$  is  $\text{nullity}(T) = \dim \ker(T)$ .
- The **nullity** of  $A$  is  $\text{nullity}(A) = \text{nullity}(T_A) = \dim \ker(A)$ .
- The **rank** of  $T$  is  $\text{rank}(T) = \dim \text{im}(T)$ .
- The **rank** of  $A$  is  $\text{rank}(A) = \text{rank}(T_A) = \dim \text{im}(A)$ .

## Rank, Nullity example

Example (Continued from the example on p.30)

Let  $A = \begin{pmatrix} 1 & 2 & 3 & 1 \\ 2 & 4 & 7 & 1 \\ 1 & 2 & 2 & 2 \end{pmatrix}$ . Find  $\text{nullity}(A)$  and  $\text{rank}(A)$ .

Solution

*A basis for  $\text{im } A$  was*

*Recall a basis for  $\text{ker } A$  had*

Note that the basis vectors for  $\text{im } A$  corresponded to the leading columns of the row-echelon form  $U$  whilst the basis vectors for  $\text{ker } A$  corresponded to the non-leading columns.

# Rank & Nullity from the row-echelon form

The previous examples illustrates

## Key Lemma

Let  $A$  be an  $m \times n$  matrix which reduces to a row-echelon form  $U$ .

- 1 nullity( $A$ ) = no. parameters in the general soln to  $A\mathbf{x} = \mathbf{0}$   
= the number of non-leading columns of  $U$ .
- 2 rank( $A$ ) = the maximal no. independent columns of  $A$   
= the number of leading columns of  $U$ .

# Rank-Nullity Theorem

Our key lemma gives

## Theorem (Rank-nullity Theorem for Matrices)

*If  $A$  is an  $m \times n$  matrix, then*

$$\text{rank}(A) + \text{nullity}(A) = n.$$

Proof.



This can be used to prove more generally,

## Theorem (Rank-nullity Theorem for Linear Maps)

*Let  $T : V \rightarrow W$  be a linear map with  $V$  finite dimensional. Then*

$$\text{rank}(T) + \text{nullity}(T) = \dim(V).$$

# Example of rank-nullity theorem

## Example

Let  $\mathbf{b} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  and  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the projection map  $\text{proj}_{\mathbf{b}}$ .  
Verify the rank-nullity theorem in this case.

## Solution

# Nature of solutions to $A\mathbf{x} = \mathbf{b}$

Our key lemma also gives

## Theorem

The equation  $A\mathbf{x} = \mathbf{b}$  has:

- ① *no solution if  $\text{rank}(A) \neq \text{rank}([A|\mathbf{b}])$ , and*
- ② *at least one solution if  $\text{rank}(A) = \text{rank}([A|\mathbf{b}])$ . Further,*
  - i) *if  $\text{nullity}(A) = 0$  the solution is unique, whereas,*
  - ii) *if  $\text{nullity}(A) = \nu > 0$ , then the general solution is of the form*

$$\mathbf{x} = \mathbf{x}_p + \lambda_1 \mathbf{k}_1 + \cdots + \lambda_\nu \mathbf{k}_\nu \quad \text{for } \lambda_1, \dots, \lambda_\nu \in \mathbb{R},$$

where  $\mathbf{x}_p$  is any solution of  $A\mathbf{x} = \mathbf{b}$ , and where  $\{\mathbf{k}_1, \dots, \mathbf{k}_\nu\}$  is a basis for  $\ker(A)$ .

# A theoretical application of rank-nullity theorem

## Example

Prove that if  $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  is linear, then the following are equivalent.

- a) For all  $\mathbf{b} \in \mathbb{R}^n$ , there is at least one solution to  $T\mathbf{x} = \mathbf{b}$ .
- b) For all  $\mathbf{b} \in \mathbb{R}^n$ , there is at most one solution to  $T\mathbf{x} = \mathbf{b}$

## Solution



# Linear Differential Equations

**Q** In what sense are second order linear differential equations linear?

**A** They involve the linear map  $T : \mathcal{C}^2[\mathbb{R}] \rightarrow \mathcal{C}[\mathbb{R}]$ , where  $\mathcal{C}^2[\mathbb{R}]$  is the vector space of all  $\mathbb{R}$ -valued functions with continuous second derivatives and  $\mathcal{C}[\mathbb{R}]$  is the vector space of all continuous  $\mathbb{R}$ -valued functions—

$$T(y) = a \frac{d^2y}{dx^2} + b \frac{dy}{dx} + cy, \quad \text{where } a, b, c \in \mathbb{R}.$$

In this case,  $\ker(T)$  is the solution set of the ODE —

$$a \frac{d^2y}{dx^2} + b \frac{dy}{dx} + cy = 0, \quad \text{where } a, b, c \in \mathbb{R}.$$

Hence the homogeneous solution is a vector space. Furthermore, it is of dimension 2 i.e.  $\text{nullity}(T) = 2$ . We can also apply similar ideas for the solution to  $A\mathbf{x} = \mathbf{b}$  to get the solution to

$$a \frac{d^2y}{dx^2} + b \frac{dy}{dx} + cy = f(x), \quad \text{where } a, b, c \in \mathbb{R}.$$

## Example involving polynomials

To study boundary value problems, it's useful to study linear maps such as the one below.

### Example

The function  $T : \mathbb{P}_2 \longrightarrow \mathbb{R}^2$  is defined by  $Tp = \begin{pmatrix} p(1) \\ p'(1) \end{pmatrix}$

- Prove that  $T$  is linear.
- Find  $\ker(T)$ .
- Use the rank-nullity theorem to find  $\text{im}(T)$ .

### Solution

## Solution (Continued)