

# Session 2: Linear transformations and matrices

Optimization and Computational Linear Algebra for Data Science

# Contents

1. Recap of the videos
2. Operation on matrices
3. Kernel and Image
4. Why do we care about all these things ?

Solving linear systems

# Linear maps & matrices

# Two sides of the same coin

## Linear map

$$L : \mathbb{R}^m \rightarrow \mathbb{R}^n$$

## Matrix

$$L \in \mathbb{R}^{n \times m}$$

# Two sides of the same coin

## Linear map

$$L : \mathbb{R}^m \rightarrow \mathbb{R}^n$$

## Matrix

$$L \in \mathbb{R}^{n \times m}$$

# Rotations in $\mathbb{R}^2$

Let  $\theta \in \mathbb{R}$ . The rotation  $R_\theta : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  of angle  $\theta$  about the origin is linear.

**Exercise:** what is the canonical matrix of  $R_\theta$ ?

# Operations on matrices

# Addition and scalar multiplication

Sum of two matrices of the **same** dimensions:

$$\begin{pmatrix} a_{1,1} & \cdots & a_{1,m} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,m} \end{pmatrix} + \begin{pmatrix} b_{1,1} & \cdots & b_{1,m} \\ \vdots & \ddots & \vdots \\ b_{n,1} & \cdots & b_{n,m} \end{pmatrix} = \begin{pmatrix} a_{1,1} + b_{1,1} & \cdots & a_{1,m} + b_{1,m} \\ \vdots & \ddots & \vdots \\ a_{n,1} + b_{n,1} & \cdots & a_{n,m} + b_{n,m} \end{pmatrix}$$

Multiplication by a scalar  $\lambda$ :

$$\lambda \begin{pmatrix} a_{1,1} & \cdots & a_{1,m} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,m} \end{pmatrix} = \begin{pmatrix} \lambda a_{1,1} & \cdots & \lambda a_{1,m} \\ \vdots & \ddots & \vdots \\ \lambda a_{n,1} & \cdots & \lambda a_{n,m} \end{pmatrix}$$



# Product of two matrices

## Warning:

$$\begin{pmatrix} a_{1,1} & \cdots & a_{1,m} \\ \vdots & \ddots & \vdots \\ a_{n,1} & \cdots & a_{n,m} \end{pmatrix} \times \begin{pmatrix} b_{1,1} & \cdots & b_{1,m} \\ \vdots & \ddots & \vdots \\ b_{n,1} & \cdots & b_{n,m} \end{pmatrix} \neq \begin{pmatrix} a_{1,1} \times b_{1,1} & \cdots & a_{1,m} \times b_{1,m} \\ \vdots & \ddots & \vdots \\ a_{n,1} \times b_{n,1} & \cdots & a_{n,m} \times b_{n,m} \end{pmatrix}$$

# Matrix product

Let  $L \in \mathbb{R}^{n \times m}$  and  $M \in \mathbb{R}^{m \times k}$ .

## Definition (Matrix product)

The matrix product  $LM$  is the  $n \times k$  matrix of the linear map  $L \circ M$ .

# Rotations in $\mathbb{R}^2$

The  $R_a$  and  $R_b$  denote respectively the matrix of the rotation of angle  $a$  and  $b$  about the origin, in  $\mathbb{R}^2$ .

**Exercise:** Compute the product  $R_a R_b$ .

# Matrix product properties

# Kernel and image

# Definitions

Let  $L : \mathbb{R}^m \rightarrow \mathbb{R}^n$  be a linear transformation .

## Definition (Kernel)

The kernel  $\text{Ker}(L)$  (or nullspace) of  $L$  is defined as the set of all vectors  $v \in \mathbb{R}^m$  such that  $L(v) = 0$ , i.e.

$$\text{Ker}(L) \stackrel{\text{def}}{=} \{v \in \mathbb{R}^m \mid L(v) = 0\}.$$

## Definition (Image)

The image  $\text{Im}(L)$  (or column space) of  $L$  is defined as the set of all vectors  $u \in \mathbb{R}^n$  such that there exists  $v \in \mathbb{R}^m$  such that  $L(v) = u$ .

**Remark:**  $\text{Im}(L)$  is also the Span of the columns of the matrix representation of  $L$ .

# Picture

# Example: orthogonal projection

Consider  $L : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  to be the orthogonal projection onto the  $x$ -axis.



# **Why do we care about this ?**

# Linear systems

# Questions?