# Lecture 2.1: Linear transformations

Optimization and Computational Linear Algebra for Data Science

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#### **Contents**

1. Definition of a linear transformation

2. Properties of linear transformations

## **Definition**

Definition

## **Examples**

You already know some linear transformations from high-school!

Symmetry	Rotation
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#### **Definition**

Symmetries (about a line passing through the origin) and rotations (about the origin) are mappings

$$L: \mathbb{R}^2 \to \mathbb{R}^2$$

$$v \mapsto L(v),$$

that are "linear":

#### **Definition**

A function  $L: \mathbb{R}^m \to \mathbb{R}^n$  is linear if

- 1. for all  $v, w \in \mathbb{R}^m$  we have L(v+w) = L(v) + L(w) and
- 2. for all  $v \in \mathbb{R}^m$  and all  $\alpha \in \mathbb{R}$  we have  $L(\alpha v) = \alpha L(v)$ .

Definition

#### An example

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### An example of a non-linear map

The function

 $F: \mathbb{R} \to \mathbb{R}$ 

is **not** linear.

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# **Properties**

## **Composition of linear maps**

#### Proposition

If  $L:\mathbb{R}^m\to\mathbb{R}^n$  and  $M:\mathbb{R}^n\to\mathbb{R}^k$  are both linear, then the composite function

$$M \circ L : \mathbb{R}^m \to \mathbb{R}^k$$
 $v \mapsto M(L(v))$ 

is also linear.

Proof.

## **Basic properties**

#### Proposition

If  $L: \mathbb{R}^m \to \mathbb{R}^n$  is linear, then

- L(0) = 0.
- $\qquad \qquad L\Big(\sum_{i=1}^k \alpha_i v_i\Big) = \sum_{i=1}^k \alpha_i L(v_i) \text{, for all } \alpha_i \in \mathbb{R}, v_i \in \mathbb{R}^m.$

Proof.

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