

# Lecture 5.1: Gram-Schmidt algorithm

Optimization and Computational Linear Algebra for Data Science

# Purpose of the algorithm

The Gram-Schmidt process takes as

- **Input:** a *linearly independent* family  $(x_1, \dots, x_k)$  of  $\mathbb{R}^n$ .
- **Output:** an *orthonormal basis*  $(v_1, \dots, v_k)$  of  $\text{Span}(x_1, \dots, x_k)$ .

## Consequence

Every subspace of  $\mathbb{R}^n$  admits an orthonormal basis.

# Gram-Schmidt algorithm

The Gram-Schmidt process constructs  $v_1, v_2, \dots, v_k$  in this order, such that for all  $i \in \{1, \dots, k\}$ :

$$\mathcal{H}_i : \begin{cases} (v_1, \dots, v_i) \text{ is an orthonormal family} \\ \text{Span}(v_1, \dots, v_i) = \text{Span}(x_1, \dots, x_i). \end{cases}$$

# Iterative construction of the $v_i$ 's

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