

# Lecture 3.2: Some properties of the rank

Optimization and Computational Linear Algebra for Data Science

# The rank-nullity theorem

## Theorem

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

$$\text{rank}(L) + \dim(\text{Ker}(L)) = m.$$

# Some inequalities

## Proposition

Let  $A \in \mathbb{R}^{n \times m}$  and  $B \in \mathbb{R}^{m \times k}$ . Then the following holds

1.  $\text{rank}(A) \leq \min(n, m)$ .
2.  $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$ .

**Proof.**



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