

# Optimization and Computational Linear Algebra for Data Science

## Common mistakes

**Warning:** the statements in red below are all wrong!

**Notations.** Let  $x \in \mathbb{R}^n$  and  $A, B \in \mathbb{R}^{n \times n}$ .

$$Ax = \text{Im}(A), \quad \text{Im}(A) + \text{Ker}(A) = n, \quad \text{Ker}(A) = 0.$$

The correct formulation would be  $Ax \in \text{Im}(A)$ ,  $\dim(\text{Im}(A)) + \dim(\text{Ker}(A)) = n$ ,  $\text{Ker}(A) = \{0\}$ .

**Matrix multiplication #1.**

$$\text{If } A = \begin{pmatrix} a_{1,1} & \cdots & a_{1,m} \\ \vdots & & \vdots \\ a_{n,1} & \cdots & a_{n,m} \end{pmatrix}, \text{ then } A^2 = \begin{pmatrix} a_{1,1}^2 & \cdots & a_{1,m}^2 \\ \vdots & & \vdots \\ a_{n,1}^2 & \cdots & a_{n,m}^2 \end{pmatrix}.$$

**Matrix multiplication #2.** Let  $A, B \in \mathbb{R}^{n \times n}$ .

$$\text{If } AB = 0 \text{ then } A = 0 \text{ or } B = 0.$$

This property (which is true when  $A, B$  are numbers) does not hold for matrices. Compute for instance

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}.$$

**Simplifications.** Let  $A \in \mathbb{R}^{k \times n}$  and  $x, u \in \mathbb{R}^n$ . Suppose that  $Ax = Au$  and that  $A \neq 0$  ( $A$  is not the all-zero matrix).

$$\text{Since } Ax = Au \text{ one can simplify by } A: \cancel{A}x = \cancel{A}u \text{ to get } x = u.$$

This is false when  $\text{Ker}(A) \neq \{0\}$ : take  $u = 0$  and  $x$  a non-zero vector in  $\text{Ker}(A)$ . If  $\text{Ker}(A) = \{0\}$ , the right way to justify that  $x = u$  is the following.  $Ax = Au$  implies that  $A(x - u) = 0$ . Hence  $x - u \in \text{Ker}(A) = \{0\}$ :  $x = u$ .

In the case where  $k = n$  and  $A$  invertible, then one could simply multiply the equation  $Ax = Au$  by  $A^{-1}$  on both sides to get that  $x = u$ .

**Expanding the Euclidean norm.** Let  $u, v \in \mathbb{R}^n$

$$\|u + v\|^2 = \|u\|^2 + 2\|u\|\|v\| + \|v\|^2,$$

or

$$\|u + v\|^2 = u^2 + 2uv + v^2.$$

The first formula is not correct, while the second has no meaning (what does  $u^2$  or  $uv$  mean when  $u, v$  are vectors?). The right formula is

$$\|u + v\|^2 = \|u\|^2 + 2\langle u, v \rangle + \|v\|^2,$$

where  $\langle u, v \rangle$  is the dot product between  $u$  and  $v$ .

**Invertible matrices.** Let  $A$  be a  $n \times m$  matrix such that  $\text{Ker}(A) = \{0\}$ .

Since  $\text{Ker}(A) = \{0\}$ ,  $A$  is invertible and  $\text{Im}(A) = \mathbb{R}^n$ .

This is only true in the case where  $n = m$ . Otherwise this makes no sense: an invertible matrix is **by definition a square matrix**. Moreover by the rank-nullity theorem  $\dim(\text{Im}(A)) = m - 0 = m$ , which can be strictly less than  $n$ .

**Matrix products.** Let  $M$  be a symmetric matrix. By the spectral theorem there exists a diagonal matrix  $D$  and an orthogonal matrix  $P$  such that  $M = PDP^T$ . Then we have

$$M = PDP^T = PP^T D,$$

and

$$M^{2019} = P^{2019} D^{2019} (P^T)^{2019}.$$

The first formula is false since for matrices  $A, B$ ,  $AB \neq BA$  in general. For the second,  $(AB)^k = ABAB \dots AB \neq A^k B^k$  in general, since  $AB$  may be different from  $BA$ .

