

Lecture 5.2: Orthogonal matrices

Optimization and Computational Linear Algebra for Data Science

Orthogonal matrices

Definition

A matrix $A \in \mathbb{R}^{n \times n}$ is called an orthogonal matrix if its columns are an orthonormal family.

Example:

$$Id_n = \begin{pmatrix} 1 & 0 & \cdots & 0 \\ 0 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & 1 \end{pmatrix}$$

is an orthogonal matrix

$$R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$$

matrix of the rotation of angle θ

is orthogonal.

A proposition

Proposition

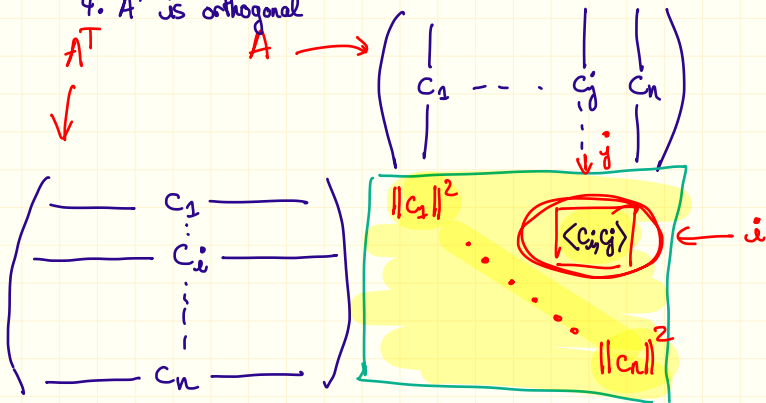
Let $A \in \mathbb{R}^{n \times n}$. The following points are equivalent:

1. A is orthogonal.

2. $A^T A = \text{Id}_n$. \Leftrightarrow A is invertible and $A^{-1} = A^T$

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4. A^T is orthogonal



Orthogonal matrices & norm

Proposition

Let $A \in \mathbb{R}^{n \times n}$ be an orthogonal matrix. Then A preserves the dot product in the sense that for all $x, y \in \mathbb{R}^n$,

$$\langle \underline{Ax}, \underline{Ay} \rangle = \langle \underline{x}, \underline{y} \rangle. \quad \parallel$$

In particular if we take $x = y$ we see that A preserves the Euclidean norm: $\| \underline{Ax} \| = \| \underline{x} \|$. \parallel

Proof: $\langle Ax, Ay \rangle = (Ax)^T (Ay) = x^T \underbrace{A^T A}_= Id_n y = x^T y = \langle x, y \rangle \quad \square$

Orthogonal matrices = { rotations
symmetries of \mathbb{R}^n