

## Série 14

**Keywords:** Singular value decomposition.

**Reminder: Singular values.**

The singular values of  $A \in M_{m \times n}$  are defined by  $\sigma_i = \sqrt{\lambda_i}$  where  $\lambda_i$  are the eigenvalues (always positive) of  $A^T A$ . We always have

$$\sigma_i = \|A\vec{v}_i\|, \quad \text{where } \vec{v}_i \text{ is a unit eigenvector of } A^T A.$$

**Reminder: Singular value decomposition.**

A matrix  $A \in M_{m \times n}$  of rank  $k$  can always be written as  $A = U\Sigma V^T$  with

- $\Sigma \in M_{m \times n}$  is the matrix of the singular values of  $A$ :

$$\Sigma = \begin{pmatrix} D & 0 \\ 0 & 0 \end{pmatrix} \in M_{m \times n}, \quad \text{with} \quad D = \begin{pmatrix} \sigma_1 & & 0 \\ & \ddots & \\ 0 & & \sigma_k \end{pmatrix}, \quad \sigma_1 \geq \dots \geq \sigma_k > 0,$$

where  $\sigma_i$  are the **non-zero** singular values of  $A$ .

- $V = (\vec{v}_1 \ \dots \ \vec{v}_n) \in M_{n \times n}$  is the **orthogonal** matrix of the eigenvectors of  $A^T A$ , ordered according to **the decreasing order of its eigenvalues**.
- $U = (A\vec{v}_1 \ \dots \ A\vec{v}_k \ \vec{u}_{k+1} \ \dots \ \vec{u}_m) \in M_{m \times m}$  is the **orthogonal** matrix of the image of the eigenvectors of  $A^T A$ , completed, if necessary (when  $k < m$ ), to a basis of  $\mathbb{R}^m$  with unit vectors.

**Question 1** Let  $A$  be a matrix of size  $m \times n$ .

- a) Show that  $\text{Ker}A = \text{Ker}(A^T A)$ .
- b) Show that  $A^T A$  is invertible if and only if the columns of  $A$  are linearly independent.

**Question 2** Find the singular value decomposition of the matrices

$$A = \begin{pmatrix} 2 & 0 \\ 0 & 0 \end{pmatrix}, \quad B = \begin{pmatrix} -3 & 1 \\ 6 & -2 \\ 6 & -2 \end{pmatrix} \quad C = \begin{pmatrix} 2 & 0 \\ 1 & 1 \\ -1 & 1 \end{pmatrix}.$$

**Question 3** Let  $A$  be a matrix, and let  $\vec{w}_1, \vec{w}_2$  be two eigenvectors of the matrix  $A^T A$ , such that

$$\vec{w}_1 = \begin{pmatrix} 1 \\ -1 \\ 0 \\ 0 \end{pmatrix}, \quad \vec{w}_2 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \end{pmatrix}, \quad A\vec{w}_1 = \begin{pmatrix} 2 \\ -1 \end{pmatrix}, \quad A\vec{w}_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}.$$

Use this information to find matrices  $U, \Sigma$ , and  $V$  such that  $A$  has a singular value decomposition of the form

$$A = U\Sigma V^T.$$

*Suggested approach:*

- First deduce the sizes of the matrices  $A$ ,  $U$ ,  $\Sigma$ , and  $V$ ;
- Normalize the vectors  $\vec{w}_1$  and  $\vec{w}_2$  to obtain  $\vec{v}_1$  and  $\vec{v}_2$ ;
- Compute  $A\vec{v}_1$  and  $A\vec{v}_2$ ;
- Compute the singular values and define  $\Sigma$ ;
- Complete  $\vec{v}_1$  and  $\vec{v}_2$  to form a basis of  $\mathbb{R}^4$  and ensure it is an orthonormal basis using the orthogonal complement method;
- Define  $V$  using  $\vec{v}_1, \vec{v}_2, \vec{v}_3, \vec{v}_4$ ;
- Normalize  $A\vec{v}_1$  and  $A\vec{v}_2$  and use them to define  $U$ .