

The SVD of Rank-1-Matrices

Let $u \in \mathbb{R}^m \setminus \{0\}$ and $v \in \mathbb{R}^n \setminus \{0\}$ be nonzero vectors and define $A := uv^T$. Find a reduced SVD of A and shortly explain why $\text{rank}(A) = 1$.

Solution:

We will in all detail derive the full and reduced SVD by following our recipe (your answer may be shorter!):

We have

$$A^T A = \|u\|^2 v v^T.$$

Following the recipe for computing the SVD of A we determine the eigenpairs of $A^T A$. We find

$$A^T A v = \|u\|^2 \|v\|^2 v,$$

which implies that v is an eigenvector to the positive (note: $u, v \neq 0$) eigenvalue $\|u\|^2 \|v\|^2$.

- Since $A^T A$ is symmetric we know all eigenvectors are mutually orthogonal.
- However, any vector x orthogonal to v is eigenvector to the eigenvalue 0 , since

$$A^T A x = \|u\|^2 \underbrace{v^T x}_{=0} = 0 \cdot x.$$

Therefore the eigenvalues of $A^T A$ are given by

$$\lambda_1 = \|u\|^2 \|v\|^2, \lambda_2 = \dots = \lambda_n = 0,$$

with precisely $r = 1$ positive one ($\Rightarrow \text{rank}(A) = 1$). Thus we find the singular values and right-singular vector by

$$v_1 := \frac{v}{\|v\|}, \sigma_1 = \|u\| \|v\| > 0.$$

Extend v_1 to the orthogonal matrix $V = \begin{pmatrix} | & & \\ v_1 & \dots & \\ | \end{pmatrix} \in \mathbb{R}^{n \times n}$ with orthonormal columns, where $v_2, \dots, v_n \in \ker(A)$.

Also set

$$\Sigma = \text{diag}(\sigma_1, 0, \dots, 0) \in \mathbb{R}^{m \times n}.$$

- Following the recipe, the corresponding left-singular vector is given by

$$u_1 := \frac{A v_1}{\sigma_1} = \frac{1}{\|u\| \|v\|} u v^T \frac{v}{\|v\|} = \frac{u}{\|u\|} \underbrace{\frac{v^T v}{\|v\|^2}}_{=1} = \frac{u}{\|u\|}.$$

Extend u_1 to the orthogonal matrix $U = \begin{pmatrix} | & & \\ u_1 & \dots & \\ | \end{pmatrix} \in \mathbb{R}^{m \times m}$ with orthonormal columns, where $u_2, \dots, u_m \in \ker(A^T)$.

- All in all we then obtain the full and reduced (as sum of rank-1 matrices) SVD

$$\Rightarrow A = U \Sigma V^T = \|u\| \|v\| u_1 v_1^T.$$