

### Calculating the Singular Value Decomposition

Suppose we have a matrix  $A$  of dimension  $m \times n$  ( $n > m$ ) with rank  $r$ .

We can find the singular value decomposition (SVD)

$$A = \sum_{i=1}^r \sigma_i \vec{u}_i \vec{v}_i^T$$

with the following steps.

1. Find the eigenvalues  $\lambda_i$  of  $A^T A$  and order them such that  $\lambda_1 \geq \dots \geq \lambda_r > 0$  and  $\lambda_{r+1} = \dots = \lambda_n = 0$ .
2. Find the orthonormal eigenvectors of  $A^T A$ , so that

$$A^T A \vec{v}_i = \lambda_i \vec{v}_i, \quad i = 1, \dots, r$$

Note that the vectors must be orthonormal, that is  $\vec{v}_i^T \vec{v}_i = 1$  and  $\vec{v}_i^T \vec{v}_j = 0$  for  $i \neq j$ .

3. Let  $\sigma_i = \sqrt{\lambda_i}$  and set

$$\vec{u}_i = \frac{A \vec{v}_i}{\sigma_i}, \quad i = 1, \dots, r$$

Note: We will see later that real symmetric matrices  $Q = Q^T$  have real eigenvalues and a set of real, orthonormal eigenvectors. Moreover if we can write  $Q = R^T R$ , the eigenvalues are non-negative.

## 1 SVD Example

Define the matrix

$$A = \begin{bmatrix} 1 & -1 \\ -2 & 2 \\ 2 & -2 \end{bmatrix}.$$

a) Find the SVD of  $A$ .

b) Find the rank of  $A$ .

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c) Find a basis for the nullspace of  $A$ .

d) Find a basis for the range (or column space) of  $A$ .

- e) Create the SVD of  $A^T$ . What are the relationships between the answers to (a)-(d) for  $A$  and for  $A^T$ ?