

Worksheet 24 (April 9)

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1 Review

DEFINITIONS

- symmetric matrices, orthogonally diagonalizable;
- singular value (of non-square matrices).

METHODS AND IDEAS

Theorem 1. (Properties of Symmetric Matrices)

Let A be a symmetric $n \times n$ matrix, then:

- (a) all eigenvalues of A are real;
- (b) eigenspaces of different eigenvalues of A are orthogonal;
- (c) there is a basis of \mathbb{R}^n consisting of orthonormal eigenvectors of A (which means A is orthogonally diagonalizable).

In fact the converse of (c) is also true: if a square matrix is orthogonally diagonalizable, it must be symmetric, too.

Method 1. (Algorithm of Singular Value Decomposition)

Assume that we are given an $m \times n$ matrix A with $m > n$.

- 1 Find eigenvalues of the symmetric matrix $A^T A$, which are all real and non-negative. Denote them by $\lambda_1, \dots, \lambda_n$. Assume that the first r of them are strictly positive and the rest are zero.
- 2 Find orthonormal eigenvectors of $A^T A$ corresponding to the strictly positive eigenvalues $\lambda_1, \dots, \lambda_r$. Denote them by $\mathbf{v}_1, \dots, \mathbf{v}_r$.
- 3 Let $\sigma_i = \sqrt{\lambda_i}$ for $i = 1, \dots, r$. These are the **singular values** of A .
- 4 Let $\mathbf{u}_i = \frac{1}{\sigma_i} A \mathbf{v}_i$ for $i = 1, \dots, r$. Then,

$$A = \sigma_1 \mathbf{u}_1 \mathbf{v}_1^T + \dots + \sigma_r \mathbf{u}_r \mathbf{v}_r^T.$$

This is the **singular value decomposition (SVD)** of A .

Note that in this case $\text{rank } A = r$. Moreover, $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$ is an orthonormal basis of $\text{Row}(A)$, while $\{\mathbf{u}_1, \dots, \mathbf{u}_r\}$ is an orthonormal basis of $\text{Col}(A)$.

2 Problems

Example 1. Diagonalize the matrix below with orthogonal matrices.

$$S = \begin{pmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}.$$

Example 2. Let us still consider the matrix S from the previous example. Is there a vector $\mathbf{x} \in \mathbb{R}^3$ such that $\mathbf{x}^T S \mathbf{x} < 0$?

Example 3. Find SVD of the matrix

$$A = \begin{pmatrix} 1 & 3 \\ 2 & 2 \\ 3 & 1 \end{pmatrix}.$$