

Here we are only interested in real matrices, though the results have easy generalizations to complex matrices.

a) A symmetric matrix has real eigenvalues, and the eigenvectors can be taken to be an orthonormal basis.

b) If $\{n_j\}$, $1 \leq j \leq N$, is an orthonormal basis of \mathbb{R}^N , then any vector Y can be written in the form

$$Y = \sum_{j=1}^N y_j n_j,$$

where

$$y_j = \langle n_j, Y \rangle$$

\langle, \rangle = scalar product.

c) Let A be a real-diagonalizable square matrix. Let R_j , $1 \leq j \leq N$, be a set of N linearly independent right (column) eigenvectors of A

$$A R_j = \lambda_j R_j,$$

where the λ 's are the eigenvalues [they need not be distinct]. Then a set of N linearly left (row) eigenvectors of A

$$L_j A = \lambda_j L_j$$

can be selected such that

$$L_n^T R_m = \delta_{n,m},$$

where $\delta_{n,m}$ is the Kronecker delta. Then any (column) vector Y can be written in the form

$$Y = \sum_{j=1}^N y_j R_j,$$

where

$$y_j = L_j^T Y.$$

This formula generalizes the one in (a-b), which applies for symmetric matrices.

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