

# Introduction to Numerical Analysis

## Day 2: Linear algebra and matrix decomposition

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- 3 Cholesky decomposition

# Matrix decomposition

# Matrix decomposition

A **matrix decomposition** or **matrix factorization** is a factorization of a matrix into a product of matrices and there are different matrix decompositions.

# LU decomposition

# LU decomposition

The LU decomposition is a simplification of the Gaussian elimination. This type of decomposition is convenient when we want to calculate the inverse of a matrix or when we want to solve a linear system of equations.

## Definition

A matrix  $\mathbf{A}$  of dimension  $m \times n$  can be written as

$$\mathbf{A} = \mathbf{L}\mathbf{U} \quad (1)$$

where  $\mathbf{L}$  is a lower triangular matrix  $m \times m$  and  $\mathbf{U}$  is an upper triangular matrix  $m \times n$ .

# LU decomposition

Thus, if we have a matrix  $\mathbf{A}$  of dimension  $3 \times 3$ , it can be expressed as:

$$\mathbf{A} = \mathbf{L}\mathbf{U} \quad (2)$$

that is the same as:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix} \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix} \quad (3)$$



# LU decomposition

See the example developed by hand and the R code for LU decomposition.

# Cholesky decomposition

# Cholesky decomposition

The Cholesky decomposition of a matrix provides an expression such that  $\mathbf{A} = \mathbf{L}\mathbf{L}^T$  where  $\mathbf{L}^T$  is transpose of the matrix  $\mathbf{L}$ .

Cholesky decomposition is faster than the  $\mathbf{LU}$  decomposition, but it is more limited since the Cholesky decomposition only can be used on **symmetric positive definite matrices**.

# Cholesky decomposition

## Symmetric positive definite matrices

- Symmetric matrices are matrices that are symmetric about the main diagonal, that is:

$$\forall i \text{ and } j, a_{ij} = a_{ji}, \text{ for a matrix } \mathbf{A}.$$

- Positive definite means that each of the pivot entries is positive.

Besides, for a positive definite matrix, the relationship  $\mathbf{xAx} > 0$  for all vectors,  $\mathbf{x}$ .

# Cholesky decomposition

So, for a matrix  $\mathbf{A}$  of dimension  $3 \times 3$ , the Cholesky decomposition can be expressed as:

$$\mathbf{A} = \mathbf{L}\mathbf{L}^T \quad (4)$$

that is the same as:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = \begin{bmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{bmatrix} \begin{bmatrix} l_{11} & l_{21} & l_{31} \\ 0 & l_{22} & l_{23} \\ 0 & 0 & l_{33} \end{bmatrix} \quad (5)$$

# Cholesky decomposition

See the example developed by hand and the R code for Cholesky decomposition.

See you next class!...



Howard, J. P. (2017). Computational Methods for Numerical Analysis with R. CRC Press.