

## **LU Decomposition**

### **Theory**

#### **LU DECOMPOSITION METHOD**

##### **OVERVIEW:**

LU Decomposition factors a square matrix A into the product of a lower triangular matrix L and an upper triangular matrix U, such that  $A = LU$ . This decomposition is particularly useful for solving multiple systems with the same coefficient matrix.

##### **MATHEMATICAL FOUNDATION:**

For matrix A:  $A = LU$

Where:

- L is a lower triangular matrix (elements above diagonal are 0)
- U is an upper triangular matrix (elements below diagonal are 0)

To solve  $Ax = b$ :

1. Decompose:  $A = LU$
2. Solve  $LY = B$  (forward substitution)
3. Solve  $UX = Y$  (back substitution)

##### **DOOLITTLE'S METHOD:**

This is the most common LU decomposition where:

- L has 1's on the diagonal
- U has the actual pivot values on diagonal

Algorithm: For  $k = 1$  to  $n$ : For  $j = k$  to  $n$ :  $u[k][j] = a[k][j] - \sum(l[k][p] * u[p][j])$  for  $p = 1$  to  $k-1$

For  $i = k+1$  to  $n$ :  $l[i][k] = (a[i][k] - \sum(l[i][p] * u[p][k])) / u[k][k]$  for  $p = 1$  to  $k-1$

##### **CROUT'S METHOD:**

Alternative where:

- U has 1's on the diagonal
- L has the actual pivot values on diagonal

##### **FORWARD SUBSTITUTION ( $Ly = b$ ):**

For  $i = 1$  to  $n$ :  $y[i] = (b[i] - \sum(l[i][j] * y[j])) / l[i][i]$  for  $j = 1$  to  $i-1$

#### **BACK SUBSTITUTION ( $Ux = y$ ):**

For  $i = n$  down to 1:  $x[i] = (y[i] - \sum(u[i][j] * x[j])) / u[i][i]$  for  $j = i+1$  to  $n$

#### **TIME COMPLEXITY:**

- Decomposition:  $O(n^3)$
- Each solve:  $O(n^2)$
- Efficient for multiple right-hand sides

#### **ADVANTAGES:**

- Efficient for multiple systems with same  $A$
- Decomposition done once, reused multiple times
- Can compute determinant easily:  $\det(A) = \text{product of } U \text{ diagonal elements}$
- Numerically stable with partial pivoting

#### **DISADVANTAGES:**

- Requires additional storage for  $L$  and  $U$
- Not all matrices have LU decomposition
- May need pivoting for numerical stability