

LU Decompositions

We seek a factorization of a square matrix A into the product of two matrices which yields an efficient method for solving the system $A \mathbf{x} = \mathbf{b}_k$ where A is the coefficient matrix, \mathbf{x} is our variable vector and \mathbf{b}_k is a constant vector for $k = 1, 2, \dots, n$. The factorization of A into the product of two matrices is closely related to Gaussian elimination.

Definition

1. A square matrix $B = (b_{ij})$ is said to be *lower triangular* if $b_{ij} = 0$ for all $i < j$.
2. A square matrix $B = (b_{ij})$ is said to be *unit lower triangular* if it is lower triangular and each $b_{ii} = 1$.
3. A square matrix $B = (b_{ij})$ is said to be *upper triangular* if $b_{ij} = 0$ for all $t > j$.

Examples

1. The following are all *lower triangular* matrices:

$$\begin{pmatrix} 2 & 0 \\ -1 & 7 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 5 & 3 & 1 \end{pmatrix}, \begin{pmatrix} 3 & 0 & 0 & 0 \\ 9 & 7 & 0 & 0 \\ -5 & -8 & 0 & 0 \\ 3 & 8 & -7 & 8 \end{pmatrix}$$

2. The following are all *unit lower triangular* matrices:

$$\begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 5 & 3 & 1 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 \\ 6 & 1 & 0 \\ -7 & 0 & 1 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 0 & 6 & 0 & 1 \end{pmatrix}$$

3. The following are all *upper triangular* matrices:

$$\begin{pmatrix} 5 & 6 \\ 0 & 9 \end{pmatrix}, \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \begin{pmatrix} 10 & 3 & 0 & 6 \\ 0 & 4 & 3 & -4 \\ 0 & 0 & -3 & 6 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

We note that the $n \times n$ identity matrix I_n is only square matrix that is both unit lower triangular and upper triangular.

Example

Let

$$A = \begin{pmatrix} 2 & 2 & 1 \\ 2 & 3 & -2 \\ 4 & 1 & -2 \end{pmatrix}.$$

For elementary matrices (See solv_lin_equ2.pdf) $E_{21} = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$, $E_{31} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -2 & 0 & 1 \end{pmatrix}$,

and $E_{32} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 3 & 1 \end{pmatrix}$ we find that

$$E_{32}(E_{31}(E_{21}A)) = \begin{pmatrix} 2 & 2 & 1 \\ 0 & 1 & -3 \\ 0 & 0 & -13 \end{pmatrix} \equiv U.$$

Now, if $S \equiv E_{32} E_{31} E_{21}$, then direct computation yields

$$S = \begin{pmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ -5 & 3 & 1 \end{pmatrix}$$

and

$$S^{-1} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & -3 & 1 \end{pmatrix} \equiv L.$$

It follows that $SA = U$ and, hence, that $A = LU$ where L is unit lower triangular and U is upper triangular. That is,

$$A = \begin{pmatrix} 2 & 2 & 1 \\ 2 & 3 & -2 \\ 4 & 1 & -2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & -3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 2 & 1 \\ 0 & 1 & -3 \\ 0 & 0 & -13 \end{pmatrix} = LU.$$

Observe the key fact that the unit lower triangular matrix L “contains” the essential data of the three elementary matrices E_{21} , E_{31} , and E_{32} .

Definition

We say that the $n \times n$ matrix A has an *LU decomposition* if $A = LU$ where L is unit lower triangular and U is upper triangular.

We also call the *LU decomposition* an *LU factorization*.

Example

1. $A = \begin{pmatrix} 2 & 2 & 1 \\ 2 & 3 & -2 \\ 4 & 1 & -2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 2 & -3 & 1 \end{pmatrix} \begin{pmatrix} 2 & 2 & 1 \\ 0 & 1 & -3 \\ 0 & 0 & -13 \end{pmatrix} = LU$ and so

$$\mathbf{A} = \begin{pmatrix} 2 & 2 & 1 \\ 2 & 3 & -2 \\ 4 & 1 & -2 \end{pmatrix} \text{ has an } LU \text{ decomposition.}$$

2. The matrix $\mathbf{B} = \begin{pmatrix} 20 & 5 & 9 \\ 16 & 4 & 7 \\ 4 & 1 & 3 \end{pmatrix}$ has more than one *LU decomposition*. Two such LU

factorizations are

$$\begin{pmatrix} 1 & 0 & 0 \\ 4/5 & 1 & 0 \\ 1/5 & 4 & 1 \end{pmatrix} \begin{pmatrix} 20 & 5 & 9 \\ 0 & 0 & -1/5 \\ 0 & 0 & 2 \end{pmatrix}$$

and

$$\begin{pmatrix} 1 & 0 & 0 \\ 4/5 & 1 & 0 \\ 1/5 & -6 & 1 \end{pmatrix} \begin{pmatrix} 20 & 5 & 9 \\ 0 & 0 & -1/5 \\ 0 & 0 & 0 \end{pmatrix}.$$

Theorem

1. The product of two lower triangular matrices is a lower triangular matrix.
2. The product of two unit lower triangular matrices is a unit lower triangular matrix.
3. Every unit lower triangular matrix is nonsingular and its inverse is also a unit lower triangular matrix.

Proof

2. Suppose \mathbf{M} and \mathbf{N} are unit lower triangular matrices. Let $\mathbf{L} = \mathbf{M}\mathbf{N}$. We must show that $l_{ij} = 0$ for all $i < j$ and $l_{ii} = 1$ for each i . Let $i < j$ and consider:

$$l_{ij} = \sum_{k=1}^n m_{ik} n_{kj} = \sum_{k=1}^{i-1} m_{ik} n_{kj} + \sum_{k=j}^n m_{ik} n_{kj} = \sum_{k=1}^{i-1} m_{ik} 0 + \sum_{k=j}^n 0 n_{kj} = 0.$$

Consider

$$\begin{aligned}
 l_{ii} &= \sum_{k=1}^n m_{ik} n_{ki} \\
 &= \sum_{k=1}^{i-1} m_{ik} n_{ki} + m_{ii} n_{ii} + \sum_{k=i+1}^n m_{ik} n_{ki} \\
 &= \sum_{k=1}^{i-1} m_{ik} 0 + 1(1) + \sum_{k=i+1}^n 0 n_{kj} \\
 &= 1
 \end{aligned}$$

3. Let $\mathbf{L} = (l_{ij})$ be an $n \times n$ unit lower triangular matrix. Now, define the $n \times n$ elementary matrix $E_{pq} = (e_{ij}^{pq})$ where

$$e_{ij}^{pq} = \begin{cases} 1 & \text{if } i = j \\ l_{pq} & \text{if } p = i \text{ \& } q = j. \\ 0 & \text{otherwise} \end{cases}$$

Then one can show that

$$\mathbf{L} = (E_{2,1} E_{3,1} E_{4,1} \dots E_{n,1}) (E_{3,2} E_{4,2} E_{5,2} \dots E_{n,2}) \dots (E_{n-1,n-2} E_{n,n-2}) (E_{n,n-1}).$$

(We note the order of the products within a given set of parenthesis above doesn't impact the product but the order of the groups of products does matter. We also note that each group in some sense yields a column of \mathbf{L} .) Hence, \mathbf{L} is the product of elementary matrices and so it is nonsingular. It is now a relatively easy exercise to show that the inverse is unit lower triangular. (Hint: What is the inverse of $E_{pq} = (e_{ij}^{pq})$?)

Theorem

A square matrix A has an *LU decomposition* iff no row interchanges are required in the Gaussian

reduction of A to upper triangular form. If A is nonsingular, the decomposition is unique.

We leave the above as an exercise.

Example

Find the LU decomposition for the matrix

$$C = \begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix}.$$

Solution

We “build” the unit lower triangular matrix L as we use Gaussian elimination to transform C into an upper triangular matrix U .

$$\begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix} \rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ ? & ? & 1 & 0 \\ ? & ? & ? & 1 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 3 & -2 & 4 \\ -6 & -18 & -1 & 1 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & ? & 1 & 0 \\ ? & ? & ? & 1 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 3 & -2 & 4 \\ 0 & 6 & 2 & 4 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & ? & 1 & 0 \\ -3 & ? & ? & 1 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 6 & 2 & 4 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & ? & ? & 1 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 4 & -2 \end{pmatrix}, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & -2 & ? & 1 \end{pmatrix}$$

$$\rightarrow \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix} = U, \quad L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & -2 & -4 & 1 \end{pmatrix}$$

So, we have that

$$C = \begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & -2 & -4 & 1 \end{pmatrix} \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix} = L U$$

Problem

Find the *LU decomposition* for the following matrices.

1. $\begin{pmatrix} 2 & 4 \\ 6 & 3 \end{pmatrix}$

$$2. \quad \begin{pmatrix} 2 & 1 & 3 \\ -2 & 5 & 1 \\ 4 & 2 & 4 \end{pmatrix}$$

Key Questions:

1. How does one use the *LU decomposition* to solve the system $\mathbf{A} \mathbf{x} = \mathbf{b}$?
2. Why does one care about this method of solution?

Consider the system $\mathbf{A} \mathbf{x} = \mathbf{b}$ and suppose \mathbf{A} has an *LU decomposition* given by $\mathbf{A} = \mathbf{L} \mathbf{U}$. Then

$$\begin{aligned} \mathbf{A} \mathbf{x} = \mathbf{b} & \quad \Upsilon \quad \mathbf{L} \mathbf{U} \mathbf{x} = \mathbf{b} \\ & \quad \Upsilon \quad \mathbf{L} \mathbf{y} = \mathbf{b} \quad \text{where } \mathbf{y} = \mathbf{U} \mathbf{x} \end{aligned}$$

Now, first solve the lower triangular system $\mathbf{L} \mathbf{y} = \mathbf{b}$ for \mathbf{y} using *forward substitution* and then solve the upper triangular system $\mathbf{U} \mathbf{x} = \mathbf{y}$ for \mathbf{x} using *backward substitution*. The solution vector \mathbf{x} then solves the original system $\mathbf{A} \mathbf{x} = \mathbf{b}$.

Example

Solve the system $\mathbf{A} \mathbf{x} = \mathbf{b}$ using the *LU decompositions* where

$$\mathbf{A} = \begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix} \quad \text{and} \quad \mathbf{b} = \begin{pmatrix} 1 \\ 0 \\ -2 \\ 1 \end{pmatrix}.$$

Solution

From a previous example we know that

$$\begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & -2 & -4 & 1 \end{pmatrix} \begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix}.$$

So, we first solve the lower triangular system $\mathbf{L} \mathbf{y} = \mathbf{b}$ using forward substitution:

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ -1 & -1 & 1 & 0 \\ -3 & -2 & -4 & 1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -2 \\ 1 \end{pmatrix}$$

Here we have that

$$y_1 = 1 \quad \Upsilon \quad 2y_1 + y_2 = 0 \quad \Upsilon \quad y_2 = -2$$

$$y_1 = 1 \ \& \ y_2 = -2 \quad \Upsilon \quad -y_1 - y_2 + y_3 = -2 \quad y_3 = -3$$

$$y_1 = 1 \ \& \ y_2 = -2 \ \& \ y_3 = -3 \quad \Upsilon \quad -3y_1 - 2y_2 - 4y_3 + y_4 = 1 \quad \Upsilon \quad y_4 = -12$$

We now solve the upper triangular system

$$\begin{pmatrix} 2 & 8 & 1 & 1 \\ 0 & -3 & 1 & -3 \\ 0 & 0 & -1 & 1 \\ 0 & 0 & 0 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 1 \\ -2 \\ -3 \\ -12 \end{pmatrix}$$

We now have that

$$2x_4 = -12 \quad \Upsilon \quad x_4 = -6$$

$$x_4 = -6 \quad \Upsilon \quad -x_3 + x_4 = -3 \quad \Upsilon \quad x_3 = -3$$

$$x_3 = -3 \text{ \& } x_4 = -6 \quad \Upsilon \quad -3x_2 + x_3 - 3x_4 = -2 \quad \Upsilon \quad x_2 = \frac{17}{3}$$

$$x_2 = \frac{17}{3} \text{ \& } x_3 = -3 \text{ \& } x_4 = -6 \quad \Upsilon \quad 2x_1 + 8x_2 + x_3 + x_4 = 1 \quad \Upsilon \quad x_1 = -\frac{53}{3}$$

Hence,
$$\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} -53/3 \\ 17/3 \\ -3 \\ -6 \end{pmatrix}$$
 solves the system

$$\begin{pmatrix} 2 & 8 & 1 & 1 \\ 4 & 13 & 3 & -1 \\ -2 & -5 & -3 & 3 \\ -6 & -18 & -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -2 \\ 1 \end{pmatrix}.$$

Now for the second question, why does one care about using the *LU decomposition* method to solve the system $\mathbf{A} \mathbf{x} = \mathbf{b}$ where the coefficient matrix \mathbf{A} is an $n \times n$ square matrix? Solving

$\mathbf{A} \mathbf{x} = \mathbf{b}$ by Gaussian elimination requires roughly $\frac{n^3}{3}$ operations and solving $\mathbf{A} \mathbf{x} = \mathbf{b}$ via *LU*

decompositions requires roughly $\frac{n^3}{3}$ operations. So, for a single system there is no computational

savings using the *LU decomposition*. However, if we must solve the systems $\mathbf{A} \mathbf{x} = \mathbf{b}_k$ where \mathbf{b}_k

is a constant vector for $k = 1, 2, \dots, n$, then it requires roughly $\frac{n^3}{3}$ operations to solve

$\mathbf{A} \mathbf{x} = \mathbf{b}_1$ via *LU decomposition* but roughly only n^2 operations to solve the remaining $\mathbf{A} \mathbf{x} = \mathbf{b}_k$

for $k = 2, 3, \dots, n$ using *LU decomposition* since we have the unit lower triangular matrix L and the upper triangular matrix U from solving $A \mathbf{x} = \mathbf{b}_1$.

The table below shows approximately the number of operations required to solve $A \mathbf{x} = \mathbf{b}_1$ and $A \mathbf{x} = \mathbf{b}_2$ via two Gaussian eliminations and via two *LU decompositions* where A is of size $n \times n$.

n	<i>Gaussian Elimination</i>	<i>LU Decompositions</i>
10	700	400
20	5300	3100
30	18000	9900
40	42700	22900
50	83300	44200