SI231b: Matrix Computations

Lecture 7: LU Factorization with Pivoting

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LU Factorization with Pivoting

$$\begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \\ 0 & 0 & 0 & 0 & \times & \times \\ 0 & 0 & 0 & 0 & 0 & \times & \times \\ \end{bmatrix} \xrightarrow{\text{pivoting}} \begin{bmatrix} a_{11}^{(0)} & \times & \times & \cdots & \times & \times \\ 0 & a_{22}^{(1)} & \times & \cdots & \times & \times \\ 0 & 0 & \times & \cdots & \times & \times \\ \vdots & \vdots & \vdots & \cdots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & 0 & \times & \times \\ 0 & 0 & 0 & \times & \times & \times \\ 0 & 0 & 0 & 0 & \times & \times \\ 0 & 0 & 0 & 0 & 0 & \times & \times \\ \end{bmatrix}$$

partial pivoting

- finding $p = \arg\max_{k < i < n} \left| a_{ik}^{(k-1)} \right|$
- let $a_{kk}^{(k-1)} = a_{nk}^{(k-1)}$ (row exchange)

complete pivoting

- finding $[p_r, p_c] = \arg\max_{k < i, j < n} \left| a_{ii}^{(k-1)} \right|$
- let $a_{\iota\iota}^{(k-1)}=a_{D_{\iota}D_{\iota}}^{(k-1)}$ (row and column exchange)



LU Factorization with Partial Pivoting

Step *k* of LU factorization

- 1. row exchange: $\tilde{\mathbf{A}}^{(k-1)} = \mathbf{P}_k \mathbf{A}^{(k-1)}$
- 2. Gaussian elimination: $\mathbf{A}^{(k)} = \mathbf{M}_k \tilde{\mathbf{A}}^{(k-1)}$

In general, the procedure follows

$$\mathbf{M}_{n-1}\mathbf{P}_{n-1}\mathbf{M}_{n-2}\mathbf{P}_{n-2}\cdots\mathbf{M}_{1}\mathbf{P}_{1}\mathbf{A}=\mathbf{U}.$$

Denote

$$\begin{split} \tilde{\mathbf{M}}_{n-1} &= \mathbf{M}_{n-1}, \\ \tilde{\mathbf{M}}_{n-2} &= \mathbf{P}_{n-1} \mathbf{M}_{n-2} \mathbf{P}_{n-1}^T, \\ \vdots &= & \vdots \\ \tilde{\mathbf{M}}_k &= \mathbf{P}_{n-1} \mathbf{P}_{n-2} \cdots \mathbf{P}_{k+1} \mathbf{M}_k \mathbf{P}_{k+1}^T \cdots \mathbf{P}_{n-2}^T \mathbf{P}_{n-1}^T. \end{split}$$

Note: $\tilde{\mathbf{M}}_k$ has the same structure with \mathbf{M}_k (recall the structure of \mathbf{M}_k)

LU Factorization with Partial Pivoting

Following the aforementioned procedure,

where

$$PA = LU$$

- ▶ $P = P_{n-1}P_{n-2}\cdots P_1$ is again a permutation matrix (why?)
- ightharpoonup $\mathbf{L}=\left(ilde{\mathbf{M}}_{n-1} ilde{\mathbf{M}}_{n-2}\cdots ilde{\mathbf{M}}_1
 ight)^{-1}$ is a lower-triangular matrix with unit diagonals
- sometimes called LUP factorization
- ▶ always exists for any square A, no matter A is nonsingular or not¹

Another Interpretation

- 1. permute the rows of A according to P
- 2. compute the LU factorization without pivoting to PA

Note: LU factorization with partial pivoting is not carried out in this way, since **P** is unknown in advance.

A Simple Example

$$\mathbf{A} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Step 1, 1st row \longleftrightarrow 3rd row of **A**, then perform Gaussian elimination

$$\tilde{\mathbf{A}}^{(0)} = \mathbf{P}_1 \mathbf{A} = \begin{bmatrix} & & 1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

$$\boldsymbol{A}^{(1)} = \boldsymbol{M}_1 \tilde{\boldsymbol{A}}^{(0)} = \begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix}$$

A Simple Example

Step 2: 2nd row \longleftrightarrow 4th row of $\mathbf{A}^{(1)}$, then repeat Gaussian elimination

$$\tilde{\textbf{A}}^{(1)} = \textbf{P}_2 \textbf{A}^{(1)} = \begin{bmatrix} 1 & & & \\ & & & 1 \\ & & 1 & \\ & 1 & & \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix}$$

$$\boldsymbol{A}^{(2)} = \boldsymbol{M}_2 \tilde{\boldsymbol{A}}^{(1)} = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & \frac{3}{7} & 1 & \\ & \frac{2}{7} & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix}$$

Now, it's your turn to give P_3 , M_3 and the final P, L, and U

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A Simple Example

$$\begin{bmatrix}
 & 1 & 1 \\
 & & 1 \\
 & & 1 \\
 & 1 & \\
 & 1 & \\
 & 1 & \\
 & 1 & \\
 & 2 & 1 & 1 & 0 \\
 & 4 & 3 & 3 & 1 \\
 & 8 & 7 & 9 & 5 \\
 & 6 & 7 & 9 & 8
\end{bmatrix} = \begin{bmatrix}
 & 1 & & & & \\
 & \frac{3}{4} & 1 & & & \\
 & \frac{1}{2} & -\frac{2}{7} & 1 & & \\
 & \frac{1}{4} & -\frac{3}{7} & \frac{1}{3} & 1
\end{bmatrix} \begin{bmatrix}
 & 8 & 7 & 9 & 5 \\
 & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\
 & & -\frac{6}{7} & -\frac{2}{7} \\
 & & & \frac{2}{3}
\end{bmatrix}$$

In practice, the permutation matrix P

- ▶ is not represented explicitly as a matrix or the product of permutation matrices
- ▶ an equivalent effect can be achieved via a permutation vector

Note: $|\ell_{ij}| \leq 1$ for $i \geq j$

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Computations of the L Matrix

Following the aforementioned procedure,

where

$$PA = LU$$

- ▶ $P = P_{n-1}P_{n-2}\cdots P_1$ is again a permutation matrix (why?)
- ightharpoonup $m L = \left(\tilde{M}_{n-1} \tilde{M}_{n-2} \cdots \tilde{M}_1
 ight)^{-1}$ is a lower-triangular matrix with unit diagonals

$$\tilde{\mathbf{M}}_{k} = \tilde{\mathbf{P}}_{k+1} \mathbf{M}_{k} \tilde{\mathbf{P}}_{k+1}^{T} = \mathbf{I} - \tilde{\mathbf{P}}_{k+1} \tau^{(k)} \mathbf{e}_{k}^{T} \qquad why?$$

Here
$$\tilde{\mathbf{P}}_{k+1} = \mathbf{P}_{n-1}\mathbf{P}_{n-2}\cdots\mathbf{P}_{k+1}$$

Then, we obtain

$$\begin{split} \tilde{\mathbf{M}}_k^{-1} &= \mathbf{I} + \tilde{\mathbf{P}}_{k+1} \boldsymbol{\tau}^{(k)} \mathbf{e}_k^T \\ \mathbf{L} &= \mathbf{I} + \sum_{i=1}^{n-1} \tilde{\mathbf{P}}_{k+1} \boldsymbol{\tau}^{(k)} \mathbf{e}_k^T \end{split}$$

An Alternative Approach for LU Factorization with Partial Pivoting

For $\mathbf{A} \in \mathbb{R}^{n \times n}$, and a permutation matrix \mathbf{P}_1

$$\mathbf{P}_{1}\mathbf{A} = \begin{bmatrix} \begin{array}{c|c} a_{11}^{(0)} & \mathbf{v}^{T} \\ \hline \mathbf{u} & \mathbf{A}_{1}' \end{array} \end{bmatrix} = \underbrace{\begin{bmatrix} \begin{array}{c|c} 1 & 0 \\ \hline 1/a_{11}^{(0)}\mathbf{u} & \mathbf{I}_{n-1} \end{array} \end{bmatrix}}_{\mathbf{L}_{1}} \underbrace{\begin{bmatrix} \begin{array}{c|c} a_{11}^{(0)} & \mathbf{v}^{T} \\ \hline 0 & \mathbf{A}_{1}' - 1/a_{11}^{(0)}\mathbf{u}\mathbf{v}^{T} \end{bmatrix}}_{\mathbf{U}_{1}}$$

Then repeat the above procedure to ${f A}_1' - 1/a_{11}^{(0)} {f u} {f v}^T$, i.e.,

$$\begin{split} \mathbf{P}_{2}' \left(\mathbf{A}_{1}' - 1/a_{11}^{(0)} \mathbf{u} \mathbf{v}^{T} \right) &= \left[\begin{array}{c|c} a_{22}^{(1)} & \mathbf{w}^{T} \\ \hline \mathbf{s} & \mathbf{A}_{2}' \end{array} \right] \\ &= \left[\begin{array}{c|c} 1 & 0 \\ \hline 1/a_{22}^{(1)} \mathbf{s} & \mathbf{I}_{n-2} \end{array} \right] \left[\begin{array}{c|c} a_{22}^{(1)} & \mathbf{w}^{T} \\ \hline 0 & \mathbf{A}_{2}' - 1/a_{22}^{(1)} \mathbf{s} \mathbf{w}^{T} \end{array} \right] \end{split}$$

Denote
$$\mathbf{P}_2 = \begin{bmatrix} 1 & & \\ & \mathbf{P}_2' \end{bmatrix}$$
 , we obtain (next page)

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An Alternative Approach for LU Factorization with Partial Pivoting

$$\mathbf{P}_{2}\mathbf{P}_{1}\mathbf{A} = \underbrace{\left[\begin{array}{ccc} 1 & & & \\ & 1 & \\ \frac{1}{a_{11}^{(0)}}\mathbf{P}_{2}^{\prime}\mathbf{u} & \frac{1}{a_{22}^{(1)}}\mathbf{s} & \mathbf{I}_{n-2} \end{array}\right]}_{\mathbf{L}_{2}} \underbrace{\left[\begin{array}{ccc} a_{11}^{(0)} & & \mathbf{v}^{T} \\ & a_{22}^{(1)} & & \mathbf{w}^{T} \\ & & \mathbf{A}_{2}^{\prime} - \frac{1}{a_{22}^{(1)}}\mathbf{s}\mathbf{w}^{T} \end{array}\right]}_{\mathbf{U}_{2}}$$

- ▶ following the above notations, $\mathbf{L} = \mathbf{L}_{n-1}$, $\mathbf{U} = \mathbf{U}_{n-1}$
- ▶ P_k only acts on the first (k-1) columns of L_k
- ▶ algorithm style, suitable for computer implementation

Remark:

- Gaussian elimination tells why you can perform an LU factorization, and when does it exist
- ▶ the recursive approach tells how you can compute the LU factorization on a modern computer

Example

Please compute an LU factorization with partial pivoting using the method introduced in the last page for

$$\begin{bmatrix} 4 & 2 & 3 \\ -3 & 1 & 4 \\ 2 & 4 & 5 \end{bmatrix}$$

LU Factorization with Complete Pivoting

LU with complete pivoting:

In matrix form, at each stage before Gaussian elimination

- ightharpoonup permutation of rows with \mathbf{P}_k on the left
- **Permutation** of columns with \mathbf{Q}_k on the right

$$\mathbf{M}_{n-1}\mathbf{P}_{n-1}\mathbf{M}_{n-2}\mathbf{P}_{n-2}\cdots\mathbf{M}_{1}\mathbf{P}_{1}\mathbf{A}\mathbf{Q}_{1}\mathbf{Q}_{2}\cdots\mathbf{Q}_{n-1}=\mathbf{U}.$$

Ву

- ▶ using the same definition of L, P with LU factorization with partial pivoting,
- ightharpoonup denoting $\mathbf{Q} = \mathbf{Q}_1 \mathbf{Q}_2 \cdots \mathbf{Q}_{n-1}$,

the LU factorization with complete pivoting can be represented by

$$PAQ = LU$$

Too computationally expensive, why?

Computational Complexity of LU Factorization

LU Factorization without Pivoting:

```
\begin{array}{l} {\tt U} = {\tt A, \ L} = {\tt I;} \\ {\tt for \ k} = {\tt 1: \ n-1} \\ \\ & {\tt for \ j} = {\tt k+1: \ n} \\ \\ & \ell_{jk} = u_{jk}/u_{kk} \\ \\ & u_{j,k:n} = u_{j,k:n} - \ell_{jk}u_{k,k:n} \\ \\ & {\tt end} \\ \\ {\tt end} \\ \\ {\tt U} = {\tt triu}({\tt U}) \end{array}
```

Operations count:

 $ightharpoonup \mathcal{O}\left(\frac{2}{3}n^3\right)$ flops

Please give your own explanation

Computational Complexity of LU Factorization with Partial Pivoting

LU Factorization with Partial Pivoting:

```
U = A, L = I, P = I;
for k = 1 : n-1
        select i \geq k to maximize |u_{ik}|
        u_{k,k,m} \leftrightarrow u_{i,k,m} (exchange of rows)
        \ell_{k,1:k-1} \leftrightarrow \ell_{i,1:k-1}
        p_{k,:} \leftrightarrow p_{i,:}
        for j = k+1 : n
               \ell_{ik} = u_{ik}/u_{kk}
               u_{i,k:n} = u_{i,k:n} - \ell_{ik} u_{k,k:n}
        end
end
U = triu(U)
```

Operations count:

 \triangleright $\mathcal{O}\left(\frac{2}{3}n^3\right)$ flops, flops count of partial pivoting?

