

Lecture 12

Gaussian Elimination and LU Factorization

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MATH 562 Numerical Analysis II

Outline

① Gaussian Elimination and LU Factorization

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Gaussian Elimination and LU Factorization

- Gaussian elimination can be viewed as “triangular triangularization” of nonsingular $\mathbf{A} \in \mathbb{C}^{m \times m}$

$$\underbrace{\mathbf{L}_{m-1} \cdots \mathbf{L}_2 \mathbf{L}_1}_{\mathbf{L}^{-1}} \mathbf{A} = \mathbf{U}$$

analogous to Householder QR factorization of matrix $\mathbf{A} \in \mathbb{C}^{m \times m}$

$$\underbrace{\mathbf{Q}_n \cdots \mathbf{Q}_2 \mathbf{Q}_1}_{\mathbf{Q}^*} \mathbf{A} = \mathbf{R}$$

- Example of LU factorization of 4×4 matrix \mathbf{A}

$$\begin{array}{ccc} \xrightarrow{\mathbf{L}_1} & \xrightarrow{\mathbf{L}_2} & \xrightarrow{\mathbf{L}_3} \\ \underbrace{\begin{bmatrix} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & \times & \times & \times \end{bmatrix}}_{\mathbf{L}_1 \mathbf{A}} & \underbrace{\begin{bmatrix} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & 0 & \times & \times \\ 0 & 0 & \times & \times \end{bmatrix}}_{\mathbf{L}_2 \mathbf{L}_1 \mathbf{A}} & \underbrace{\begin{bmatrix} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & 0 & \times & \times \\ 0 & 0 & 0 & \times \end{bmatrix}}_{\mathbf{L}_3 \mathbf{L}_2 \mathbf{L}_1 \mathbf{A}} \end{array}$$

What is Matrices \mathbf{L}_k

- At step k , eliminate entries below a_{kk} : let \mathbf{x}_k be k th column of $\mathbf{L}_{k-1} \cdots \mathbf{L}_1 \mathbf{A}$,
 $\mathbf{x}_k = [x_{1,k}, x_{2,k}, \dots, x_{k,k}, x_{k+1,k}, \dots, x_{m,k}]^T$
 $\mathbf{L}_k \mathbf{x}_k = [x_{1,k}, x_{2,k}, \dots, x_{k,k}, 0, \dots, 0]^T$
- The multiplier $l_{jk} = x_{jk}/x_{kk}$ appear in \mathbf{L}_k

$$\mathbf{L}_k = \begin{bmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & -l_{k+1,k} & 1 & & \\ & & \vdots & & \ddots & \\ & & -l_{m,k} & & & 1 \end{bmatrix}$$

- Let $\mathbf{l}_k = [0, \dots, 0, l_{k+1,k}, \dots, l_{m,k}]^T$ and $\mathbf{e}_k = [0, \dots, 0, 1, \dots, 0]^T$, then $\mathbf{L}_k = \mathbf{I} - \mathbf{l}_k \mathbf{e}_k^*$.

Forming \mathbf{L}

- Luckily, the \mathbf{L} matrix contains the multipliers $l_{jk} = x_{jk}/x_{kk}$

$$\mathbf{L} = \mathbf{L}_1^{-1} \mathbf{L}_2^{-1} \cdots \mathbf{L}_{m-1}^{-1} = \begin{bmatrix} 1 & & & & \\ l_{21} & 1 & & & \\ l_{31} & l_{32} & 1 & & \\ \vdots & \vdots & \ddots & \ddots & \\ l_{m1} & l_{m2} & \cdots & l_{m,m-1} & 1 \end{bmatrix}$$

and is said to be a unit lower triangular matrix

- First, $\mathbf{L}_k^{-1} = \mathbf{I} + \mathbf{l}_k \mathbf{e}_k^*$, because $\mathbf{e}_k^* \mathbf{l}_k = 0$ and $(\mathbf{I} - \mathbf{l}_k \mathbf{e}_k^*)(\mathbf{I} + \mathbf{l}_k \mathbf{e}_k^*) = \mathbf{I} - \mathbf{l}_k \mathbf{e}_k^* \mathbf{l}_k \mathbf{e}_k^* = \mathbf{I}$
- Second, $\mathbf{L}_1^{-1} \mathbf{L}_2^{-1} \cdots \mathbf{L}_{k+1}^{-1} = \mathbf{I} + \sum_{j=1}^{k+1} \mathbf{l}_j \mathbf{e}_j^*$, since (prove by induction) $(\mathbf{I} + \sum_{j=1}^k \mathbf{l}_j \mathbf{e}_j^*)(\mathbf{I} + \mathbf{l}_{k+1} \mathbf{e}_{k+1}^*) = \mathbf{I} + \sum_{j=1}^{k+1} \mathbf{l}_j \mathbf{e}_j^* + \sum_{j=1}^k \mathbf{l}_j (\mathbf{e}_j^* \mathbf{l}_{k+1}) \mathbf{e}_{k+1}^*$ where $\mathbf{e}_j^* \mathbf{l}_{k+1} = 0$ for $j < k+1$
- In other words, \mathbf{L} is “union” of $\mathbf{L}_1^{-1}, \mathbf{L}_2^{-1}, \dots, \mathbf{L}_{m-1}^{-1}$

Gaussian Elimination without Pivoting

- Factorize $\mathbf{A} \in \mathbb{C}^{m \times m}$ into $\mathbf{A} = \mathbf{L}\mathbf{U}$

Gaussian elimination without pivoting

$$\mathbf{U} = \mathbf{A}, \mathbf{L} = \mathbf{I}$$

for $k = 1$ to $m - 1$

for $j = k + 1$ to m

$$l_{jk} = u_{jk}/u_{kk}$$

$$u_{j,k:m} = u_{j,k:m} - l_{jk}u_{k,k:m}$$

- Flop count $\sim \sum_{k=1}^m 2(m-k)(m-k) \sim 2 \sum_{k=1}^m k^2 \sim 2m^3/3$
- In practice, \mathbf{L} often overwrites lower-triangular part of \mathbf{A} and \mathbf{U} overwrites upper-triangular part of \mathbf{A}
- Question: What if u_{kk} is 0? Answer: The algorithm would break.

Partial Pivoting

- At step k , we divide by u_{kk} , which would break if u_{kk} is 0 (or close to 0), which can happen even if \mathbf{A} is nonsingular

$$\begin{bmatrix} \times & \times & \times & \times & \times \\ 0 & x_{kk} & \times & \times & \times \\ 0 & \times & \times & \times & \times \\ 0 & \times & \times & \times & \times \\ 0 & \times & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times & \times \\ 0 & x_{kk} & \times & \times & \times \\ 0 & 0 & \times & \times & \times \\ 0 & 0 & \times & \times & \times \\ 0 & 0 & \times & \times & \times \end{bmatrix} \rightarrow$$

- However, any nonzero entry in k th column below diagonal can also be used as pivot (In general, we take nonzero entry with largest absolute value)

$$\begin{bmatrix} \times & \times & \times & \times & \times \\ 0 & \times & \times & \times & \times \\ 0 & \times & \times & \times & \times \\ 0 & x_{ik} & \times & \times & \times \\ 0 & \times & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times & \times \\ 0 & 0 & \times & \times & \times \\ 0 & 0 & \times & \times & \times \\ 0 & x_{ik} & \times & \times & \times \\ 0 & 0 & \times & \times & \times \end{bmatrix}$$

More on Partial Pivoting

- k th step of Gaussian elimination of partial pivoting

$$\begin{array}{ccc}
 \left[\begin{array}{cccc} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & x_{ik} & \times & \times \\ 0 & \times & \times & \times \end{array} \right] & \xrightarrow{\mathbf{P}_k} & \left[\begin{array}{cccc} \times & \times & \times & \times \\ 0 & x_{kk} & * & * \\ 0 & * & * & * \\ 0 & \times & \times & \times \end{array} \right] & \xrightarrow{\mathbf{L}_k} & \left[\begin{array}{cccc} \times & \times & \times & \times \\ 0 & x_{kk} & \times & \times \\ 0 & 0 & * & * \\ 0 & 0 & * & * \end{array} \right] \\
 \text{Pivot selection} & & \text{Row Interchange} & & \text{Elimination}
 \end{array}$$

and we interchange row i with row k

- In terms of matrices, it becomes $\underbrace{\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_2\mathbf{P}_2\mathbf{L}_1\mathbf{P}_1}_{\mathbf{L}^{-1}\mathbf{P}}\mathbf{A} = \mathbf{U}$
- $\mathbf{P} = \mathbf{P}_{m-1}\cdots\mathbf{P}_2\mathbf{P}_1$ and $\mathbf{L} = (\mathbf{L}'_{m-1}\cdots\mathbf{L}'_2\mathbf{L}'_1)^{-1}$, where $\mathbf{L}'_k = \mathbf{P}_{m-1}\cdots\mathbf{P}_{k+1}\mathbf{L}_k\mathbf{P}_{k+1}^{-1}\cdots\mathbf{P}_{m-1}^{-1}$
- It is easy to verify that $\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_2\mathbf{P}_2\mathbf{L}_1\mathbf{P}_1 = (\mathbf{L}'_{m-1}\cdots\mathbf{L}'_2\mathbf{L}'_1)(\mathbf{P}_{m-1}\cdots\mathbf{P}_2\mathbf{P}_1)$
- $\mathbf{L}'_k = \mathbf{I} - \mathbf{P}_{m-1}\cdots\mathbf{P}_{k+1}\mathbf{l}_k\mathbf{e}_k^*$ and \mathbf{L} is “union” of $(\mathbf{L}'_k)^{-1} = \mathbf{I} + \mathbf{P}_{m-1}\cdots\mathbf{P}_{k+1}\mathbf{l}_k\mathbf{e}_k^*$

Algorithm of Gaussian Elimination with Partial Pivoting

- Factorize $\mathbf{A} \in \mathbb{C}^{m \times m}$ into $\mathbf{PA} = \mathbf{LU}$

Gaussian elimination with partial pivoting

$$\mathbf{U} = \mathbf{A}, \mathbf{L} = \mathbf{I}, \mathbf{P} = \mathbf{I}$$

for $k = 1$ to $m - 1$

$$i \leftarrow \operatorname{argmax}_{i \geq k} |u_{ij}|$$

$$\mathbf{u}_{k,k:m} \leftrightarrow \mathbf{u}_{i,k:m}$$

$$\mathbf{l}_{k,1:k-1} \leftrightarrow \mathbf{l}_{i,1:k-1}$$

$$\mathbf{p}_{k,:} \leftrightarrow \mathbf{p}_{i,:}$$

for $j = k + 1$ to m

$$l_{jk} = u_{jk}/u_{kk}$$

$$u_{j,k:m} = u_{j,k:m} - l_{jk}u_{k,k:m}$$

- Question: What if u_{kk} is 0?
- Flot count $\sim \sum_{k=1}^m 2(m-k)(m-k) \sim 2 \sum_{k=1}^m k^2 \sim 2m^3/3$, same as without pivoting.

An Alternative Implementation

- In practice, **L** and **U** overwrite **A** and **P** is represented by a vector

Gaussian elimination with partial pivoting (alternative)

```
p = [1, 2, ..., m];  
for k = 1 to m - 1  
    i ←  $\operatorname{argmax}_{i \geq k} |u_{ij}|$   
    ak,1:m ↔ ai,1:m  
    pk ↔ pi  
    ak+1:m,k ← ak+1:m,k / ak,k  
    Ak+1:m,k+1:m ← Ak+1:m,k+1:m - ak+1:m,k × ak,k+1:m
```

- Using LU factorization to solve **Ax = b**:
 - **PA = LU** (LU factorization with partial pivoting)
 - **Ly = Pb** (Forward substitution)
 - **Ux = y** (Back substitution)

Complete Pivoting

- More generally, we can use any nonzero entry
- In theory, any nonzero entry $(i, j), i \geq k, j \geq k$

$$\begin{bmatrix} \times & \times & \times & \times \\ 0 & \times & \times & \times \\ 0 & \times & x_{ij} & \times \\ 0 & \times & \times & \times \end{bmatrix} \rightarrow \begin{bmatrix} \times & \times & \times & \times \\ 0 & * & 0 & * \\ 0 & \times & x_{ij} & \times \\ 0 & * & 0 & * \end{bmatrix}$$

and we then permute row i with row k , column j with column k

- In matrix operations, it can be expressed as

$$\underbrace{\mathbf{L}_{m-1}\mathbf{P}_{m-1}\cdots\mathbf{L}_2\mathbf{P}_2\mathbf{L}_1\mathbf{P}_1}_{\mathbf{L}^{-1}\mathbf{P}}\underbrace{\mathbf{A}\mathbf{Q}_1\mathbf{Q}_2\cdots\mathbf{Q}_{m-1}}_{\mathbf{Q}}=\mathbf{U}$$

- Therefore, $\mathbf{PAQ} = \mathbf{LU}$ where $\mathbf{P} = \mathbf{P}_{m-1}\cdots\mathbf{P}_2\mathbf{P}_1$ and $\mathbf{L} = (\mathbf{L}'_{m-1}\cdots\mathbf{L}'_2\mathbf{L}'_1)^{-1}$
- However, complete pivoting is typically not used in practice because it increases cost in search of pivot and complexity of implementation