

Chapter 13 Gaussian Elimination (III)

Bunch-Parlett diagonal pivoting

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Reference: James R. *Bunch* and Linda *Kaufman*, Some Stable Methods for Calculating Inertia and Solving Symmetric Linear Systems, *Mathematics of Computation*, volume 31, number 137, January 1977, page 163-179

OutLine

- Preliminary
 - 1x1, 2x2 pivoting in Gaussian Elimination
 - real symmetric indefinite matrix
- Symmetric permutation
- LDL' decomposition (diagonal pivoting)
- Example of complete diagonal pivoting
- Algorithm of complete diagonal pivoting

Pivoting in Traditional Gaussian Elimination

consider $A = A^{(1)} = \left(\begin{array}{c|cccc} 6 & -2 & 2 & 4 \\ \hline 12 & -8 & 6 & 10 \\ 3 & -13 & 9 & 3 \\ -6 & 4 & 1 & -18 \end{array} \right) \equiv \left(\begin{array}{c|c} E & d^T \\ \hline c & B \end{array} \right) \quad E = a_{11}, \quad c = \begin{pmatrix} 12 \\ 3 \\ -6 \end{pmatrix}, \quad d = \begin{pmatrix} -2 \\ 2 \\ 4 \end{pmatrix}$

then one step LU -factorization leads to $\left(\begin{array}{c|c} a_{11} & d^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} 1 & \\ \hline c/a_{11} & I \end{array} \right) \left(\begin{array}{c|c} a_{11} & d^T \\ \hline & A^{(2)} \end{array} \right) = L^{(1)}U^{(1)}$

where $L^{(1)} \equiv \left(\begin{array}{c|ccc} 1 & & & \\ \hline 2 & 1 & & \\ 0.5 & & 1 & \\ -1 & & & 1 \end{array} \right) = \left(\begin{array}{c|c} 1 & 0 \\ \hline c/a_{11} & I_{3 \times 3} \end{array} \right) \quad \text{and} \quad U^{(1)} = \left(\begin{array}{c|cccc} a_{11} & d^T & & \\ \hline & & B & \end{array} \right) = \left(\begin{array}{c|cccc} 6 & -2 & 2 & 4 \\ \hline 0 & -4 & 2 & 2 \\ 0 & -12 & 8 & 1 \\ 0 & 2 & 3 & -14 \end{array} \right)$

It is easy to show $A^{(2)} \equiv B - \frac{cd^T}{a_{11}} = \begin{pmatrix} -4 & 2 & 2 \\ -12 & 8 & 1 \\ 2 & 3 & -14 \end{pmatrix}$ from Gaussian Elimination

We use principal submatrix $E = a_{11}$ as pivoting, called 1×1 pivoting since $E \in R^{1 \times 1}$

Question: can we use 2×2 or higher $s \times s$ pivoting, i.e. $E \in R^{s \times s}$

2x2 pivoting in block Gaussian Elimination

$$A = \left(\begin{array}{c|c} \overbrace{\begin{matrix} E & d^T \end{matrix}}^{s} \\ \hline \underbrace{\begin{matrix} c & B \end{matrix}}_{m-s} \end{array} \right)$$

If principal sub-matrix E is non-singular,

then one-step GE leads to
$$\left(\begin{array}{c|c} E & d^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) \left(\begin{array}{c|c} E & d^T \\ \hline & A^{(s+1)} \end{array} \right) \quad A^{(s+1)} = B - cE^{-1}d^T$$

Example 1: 2×2 pivoting

$$A = A^{(1)} = \left(\begin{array}{cc|cc} 6 & -2 & 2 & 4 \\ 12 & -8 & 6 & 10 \\ \hline 3 & -13 & 9 & 3 \\ -6 & 4 & 1 & -18 \end{array} \right) \equiv \left(\begin{array}{c|c} E & d^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) \left(\begin{array}{c|c} E & d^T \\ \hline & A^{(3)} \end{array} \right) \quad |A|_{\infty} = 18$$

where
$$\left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) = \left(\begin{array}{cc|c} 1 & & \\ \hline & 1 & \\ -5.5 & 3 & 1 \\ \hline 0 & -0.5 & 1 \end{array} \right) \quad \text{and} \quad \left(\begin{array}{c|c} E & d^T \\ \hline & A^{(3)} \end{array} \right) = \left(\begin{array}{cc|cc} 6 & -2 & 2 & 4 \\ 12 & -8 & 6 & 10 \\ \hline & & 2 & -5 \\ & & 4 & -13 \end{array} \right)$$

$|L|_{\infty} = 5.5$ $|U|_{\infty} = 13$

3x3 pivoting in block Gaussian Elimination

Example 2: 3x3 pivoting

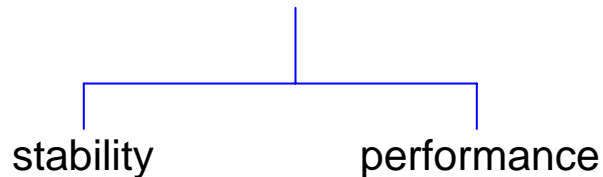
$$A = A^{(1)} = \left(\begin{array}{ccc|c} 6 & -2 & 2 & 4 \\ 12 & -8 & 6 & 10 \\ 3 & -13 & 9 & 3 \\ \hline -6 & 4 & 1 & -18 \end{array} \right) \equiv \left(\begin{array}{c|c} E & d^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) \left(\begin{array}{c|c} E & d^T \\ \hline & A^{(4)} \end{array} \right) \quad |A|_{\infty} = 18$$

where

$$\left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) = \left(\begin{array}{ccc|c} 1 & & & \\ & 1 & & \\ \hline & & 1 & \\ 11 & -6.5 & 2 & 1 \end{array} \right) \quad \text{and} \quad \left(\begin{array}{c|c} E & d^T \\ \hline & A^{(3)} \end{array} \right) = \left(\begin{array}{ccc|c} 6 & -2 & 2 & 4 \\ 12 & -8 & 6 & 10 \\ 3 & -13 & 9 & 3 \\ \hline & & & -3 \end{array} \right)$$

$$|L|_{\infty} = 11 \qquad |U|_{\infty} = 13$$

Question: what is criterion to choose 1x1, 2x2 or 3x3 pivoting?



Question: what is the **cost** to estimate the criterion ?

Real symmetry indefinite matrix A

$$A = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} \text{ is symmetric implies } E = E^T \text{ and } B = B^T$$

From linear algebra, if A is real symmetric, then A has real spectrum decomposition (譜分解)

$$AV_j = \lambda_j V_j \quad \lambda_j \in \mathbb{R} \text{ and } V_i^T V_j = \delta_i^j = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases} \quad \forall j=1,2,\dots,n$$

or write in matrix form $AV = V\Lambda$ $V = (V_1 | V_2 | \dots | V_n)$ and $\Lambda = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n) = \begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}$

without loss of generality, we can rearrange eigen-values to be increasing

$$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$$

Definition: suppose matrix A is real symmetric, then we say “ A is indefinite” if there exists an eigen-value of A less than zero. If A is nonsingular, then

$$\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_k < 0 < \lambda_{k+1} \leq \dots \leq \lambda_n$$

Thereafter we focus on LU -decomposition of symmetry indefinite matrix A

LU-factorization for real symmetry indefinite matrix A

LU – factorization $A = \left(\begin{array}{c|c} E & c^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) \left(\begin{array}{c|c} E & c^T \\ \hline - & B - cE^{-1}c^T \end{array} \right)$

LDL^T – factorization $A = \left(\begin{array}{c|c} E & c^T \\ \hline c & B \end{array} \right) = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right) \left(\begin{array}{c|c} E & \\ \hline - & B - cE^{-1}c^T \end{array} \right) \left(\begin{array}{c|c} I & E^{-1}c^T \\ \hline & I \end{array} \right)$

where $L = \left(\begin{array}{c|c} I & \\ \hline cE^{-1} & I \end{array} \right)$ and $L^T = \left(\begin{array}{c|c} I & E^{-T}c^T \\ \hline & I \end{array} \right) = \left(\begin{array}{c|c} I & E^{-1}c^T \\ \hline & I \end{array} \right)$

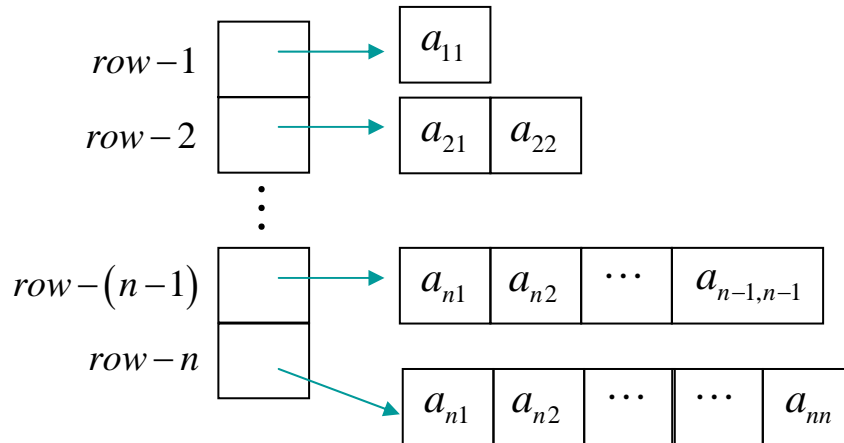
Question: why not LU -decomposition?

A is real symmetric, we only store lower triangle part of A , say

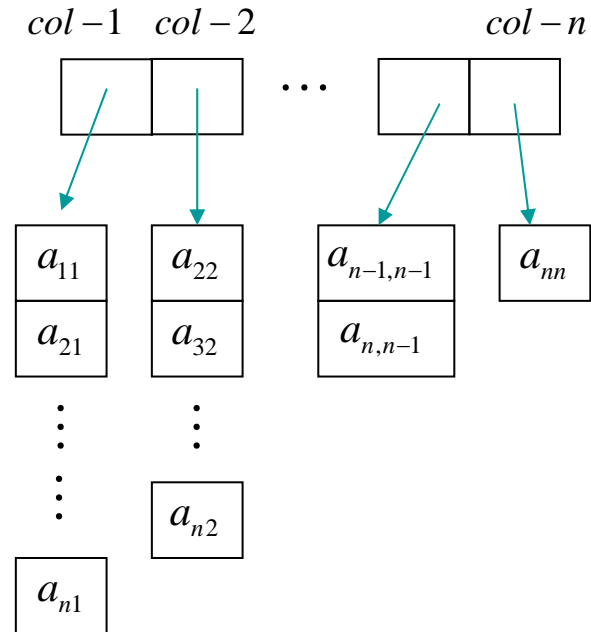
$$\text{mem}(A) = \begin{pmatrix} a_{11} & & & \\ a_{21} & a_{22} & & \\ \vdots & \vdots & \ddots & \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix}$$

Pointer array to store lower triangular part of A

row-major based



col-major based



Question: what is the cost to fetch one element of A ? That is, operation count for $A(i, j)$

Question: can you find another representation for lower part of matrix A ?

Question: if one uses row-major to store lower part of matrix A , then how to fetch a column of A efficiently?

OutLine

- Preliminary
- **Symmetric permutation**
 - change diagonal element to (1,1) position
 - change off-diagonal element to (2,1) position
 - implementation of symmetric permutation
- LDL' decomposition (diagonal pivoting)
- Example of complete diagonal pivoting
- Algorithm of complete diagonal pivoting

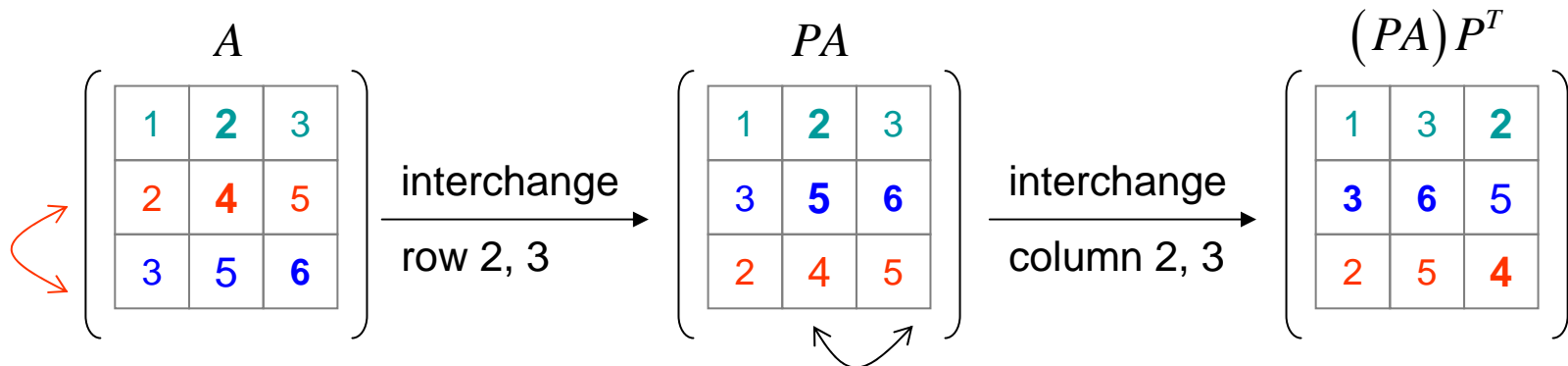
How to do permutation on real symmetry indefinite matrix A

Let $A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix}$ Define permutation matrix $P = (1,3,2) \equiv \begin{pmatrix} e_1^T \\ e_3^T \\ e_2^T \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$

$$Px = \begin{pmatrix} e_1^T \\ e_3^T \\ e_2^T \end{pmatrix} x = \begin{pmatrix} e_1^T x \\ e_3^T x \\ e_2^T x \end{pmatrix} = \begin{pmatrix} x_1 \\ x_3 \\ x_2 \end{pmatrix} \xrightarrow{\text{interchange row 2, 3}} PA = \begin{pmatrix} 1 & 2 & 3 \\ 3 & 5 & 6 \\ 2 & 4 & 5 \end{pmatrix}$$

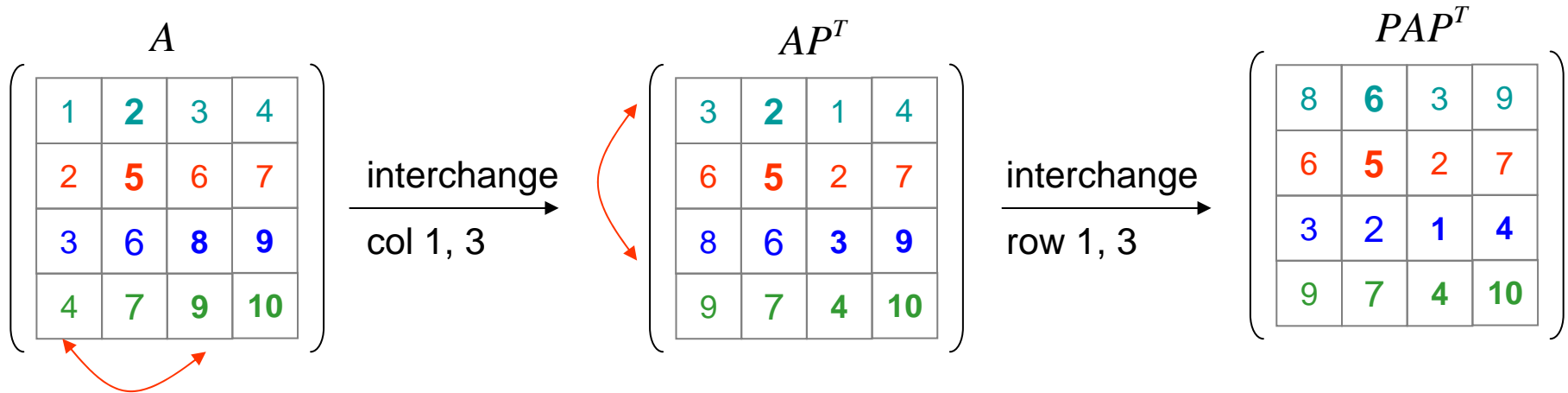
$$x^T P^T = (Px)^T = \begin{pmatrix} x_1 \\ x_3 \\ x_2 \end{pmatrix}^T = (x_1 \quad x_3 \quad x_2) \xrightarrow{\text{interchange column 2, 3}} AP^T = \begin{pmatrix} 1 & 3 & 2 \\ 2 & 5 & 4 \\ 3 & 6 & 5 \end{pmatrix}$$

Symmetric permutation: $A \rightarrow PAP^T$

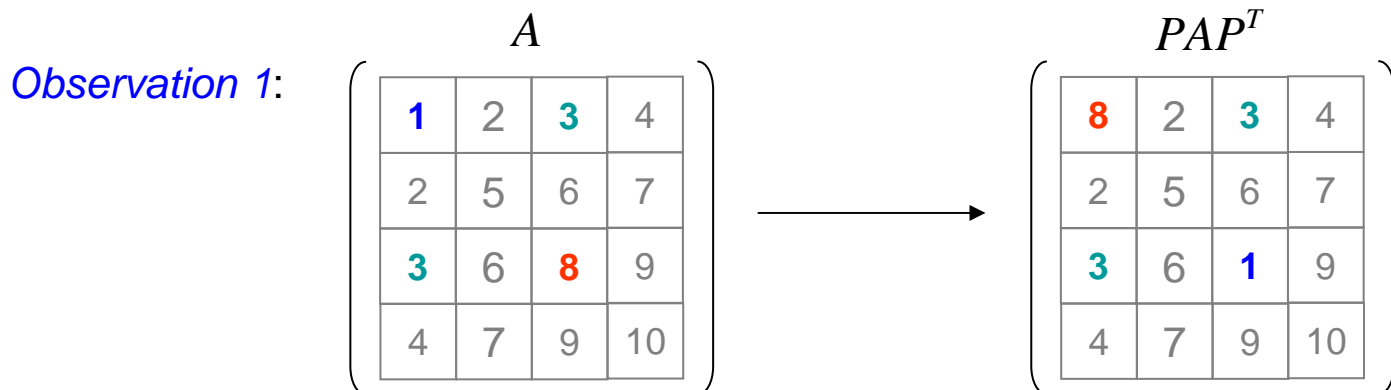


Change diagonal element to (1,1) position [1]

suppose we want to change $a_{33} = 8$ to position (1,1), then consider permutation $P = (3,2,1,4)$ and do symmetric permutation on A , say PAP^T



Question: we only store lower triangle part of matrix A , above permutation does not work, how to modify it?



Change diagonal element to (1, 1) position [2]

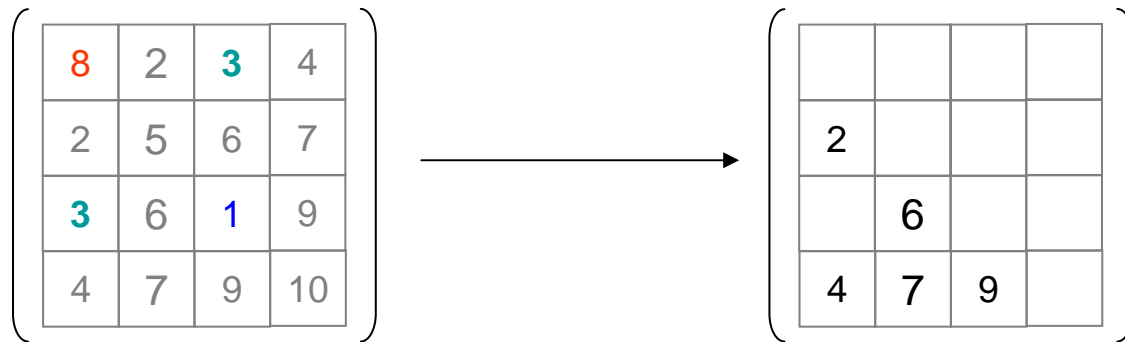
$$a_{11} \xrightarrow{\text{col-1} \rightarrow \text{col-3}} a_{13} \xrightarrow{\text{row-1} \rightarrow \text{row-3}} a_{33}$$

$$a_{33} \xrightarrow{\text{col-3} \rightarrow \text{col-1}} a_{31} \xrightarrow{\text{row-3} \rightarrow \text{row-1}} a_{11}$$

$$a_{13} \xrightarrow{\text{col-3} \rightarrow \text{col-1}} a_{11} \xrightarrow{\text{row-1} \rightarrow \text{row-3}} a_{31}$$

$$a_{31} \xrightarrow{\text{col-1} \rightarrow \text{col-3}} a_{33} \xrightarrow{\text{row-3} \rightarrow \text{row-1}} a_{13}$$

Observation 2: we only need to update lower triangle part of A (diagonal term is excluded).



diagonal term $a_{kk} (k \neq 1, 3) \xrightarrow{\text{col-1} \leftrightarrow \text{col-3}} a_{kk} \xrightarrow{\text{row-1} \leftrightarrow \text{row-3}} a_{kk}$ does not changed

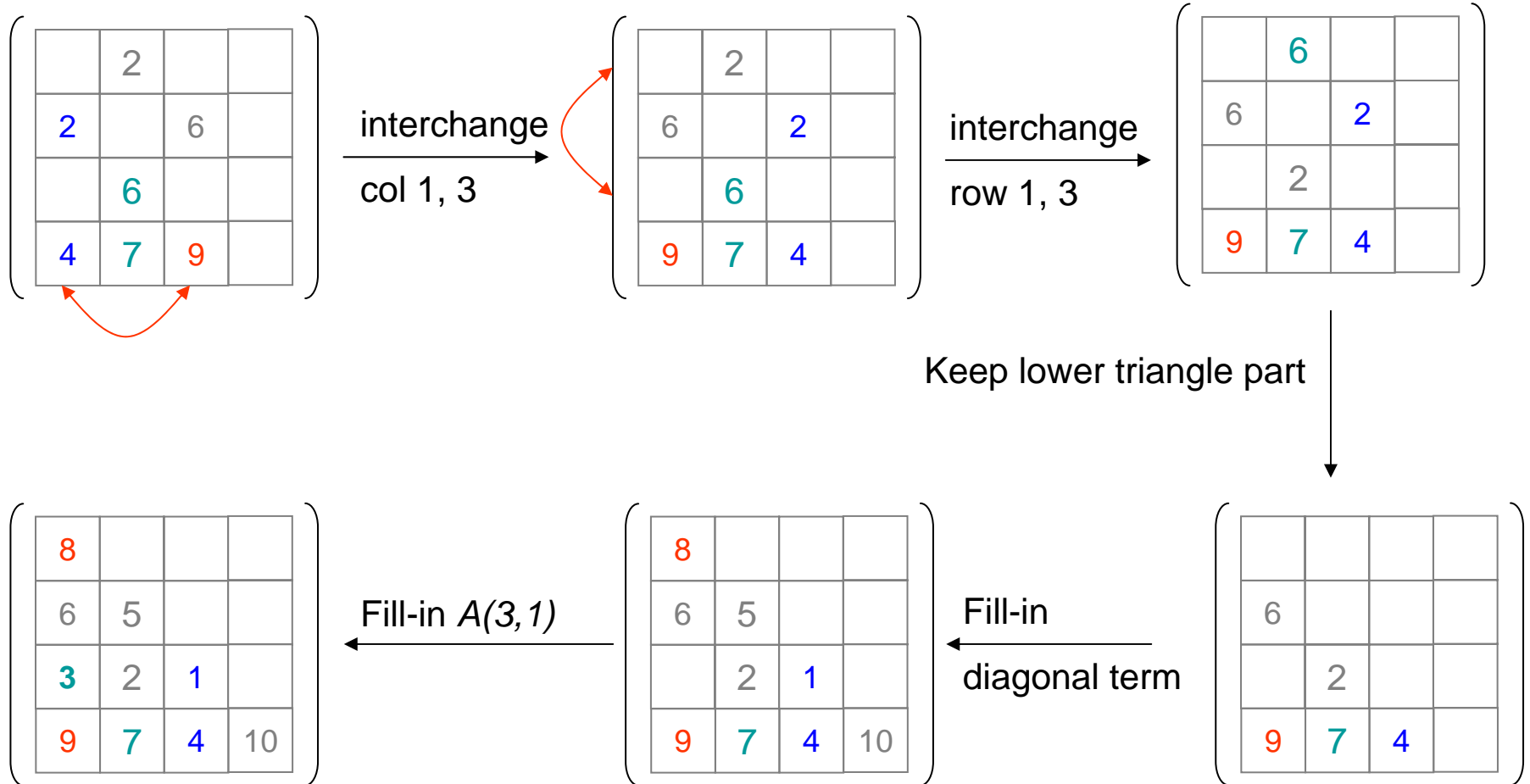
$a_{11} \xrightarrow{\text{col-1} \leftrightarrow \text{col-3}, \text{row-1} \leftrightarrow \text{row-3}} a_{33}$
 $a_{33} \xrightarrow{\text{col-1} \leftrightarrow \text{col-3}, \text{row-1} \leftrightarrow \text{row-3}} a_{11}$

} we have changed

$a_{13} = a_{31}$ does not changed

Change diagonal element to (1,1) position

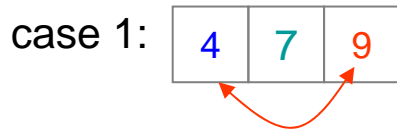
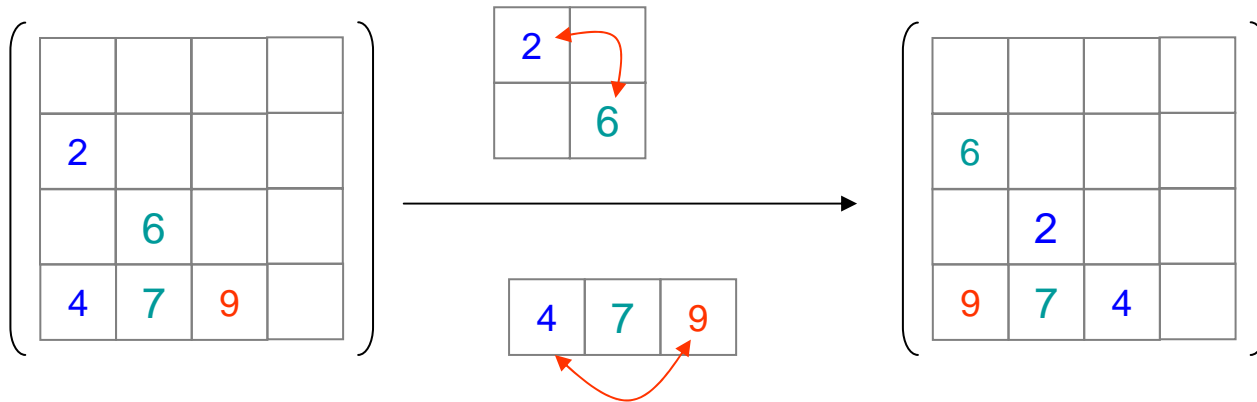
[3]



Question: any compact form to represent above interchange?

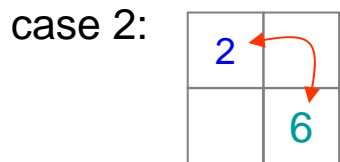
Change diagonal element to (1,1) position

[4]



$$a_{41} (i \neq 1, 3) \xrightarrow{col-1 \leftrightarrow col-3} a_{43} \xrightarrow{row-1 \leftrightarrow row-3} a_{43}$$

$$a_{43} (i \neq 1, 3) \xrightarrow{col-1 \leftrightarrow col-3} a_{41} \xrightarrow{row-1 \leftrightarrow row-3} a_{41}$$



$$a_{21} (i \neq 1, 3) \xrightarrow{col-1 \leftrightarrow col-3} a_{23} \xrightarrow{row-1 \leftrightarrow row-3} a_{23} (\text{upper triangle}) = a_{32}$$

$$a_{32} (j \neq 1, 3) \xrightarrow{col-1 \leftrightarrow col-3} a_{32} \xrightarrow{row-1 \leftrightarrow row-3} a_{12} (\text{upper triangle}) = a_{21}$$

Change diagonal element to (1, 1) position [5]

For general real symmetric matrix A with dimension n , suppose we want to change $a_{kk} \rightarrow a_{11}$

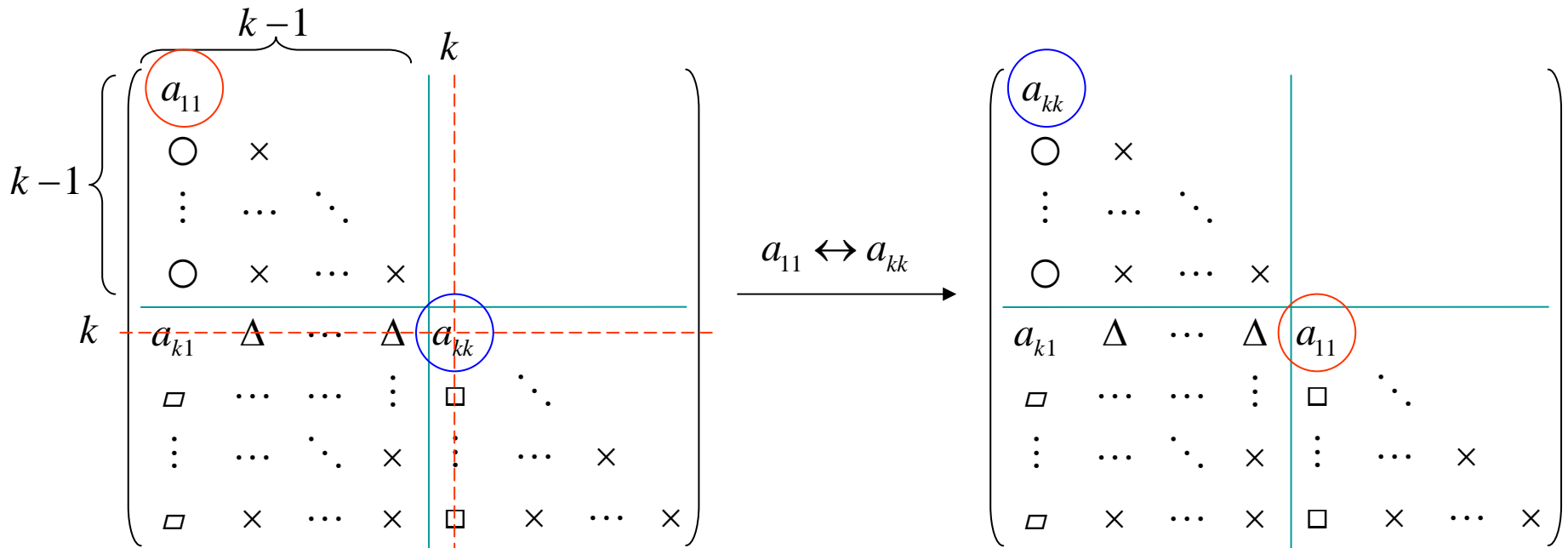
$$P = (k, 2, 3, \dots, k-1, 1, k+1, n) \quad \text{interchanges rows } 1, k \text{ or columns } 1, k$$

\uparrow
 1-th

\uparrow
 $k\text{-th}$

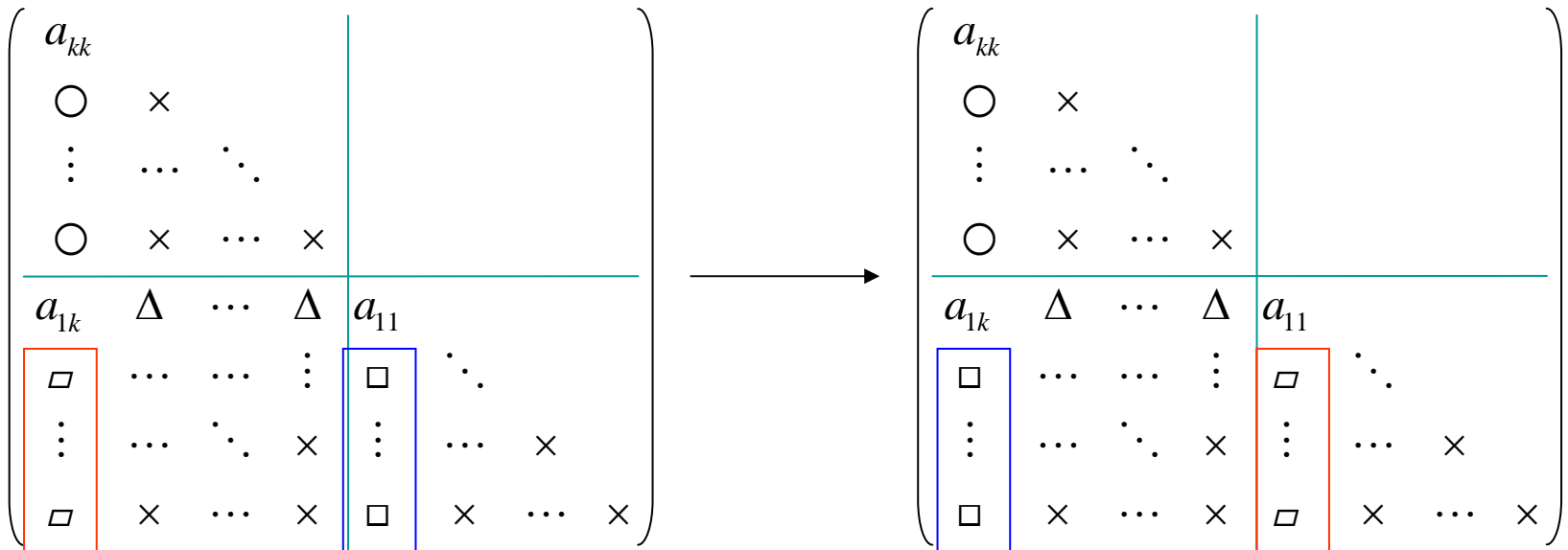
Step 1: interchange position $(1, 1)$ and (k, k) , i.e. $a_{11} \leftrightarrow a_{kk}$

keep position $(k, 1)$ and $(1, k)$, i.e. $a_{1k} = a_{k1}$ does not change



Change diagonal element to (1, 1) position [6]

Step 2: interchange column 1 and k below row $k+1$, i.e $A(k+1:n,1) \leftrightarrow A(k+1:n,k)$



$$A(i,1)(i > k) \xrightarrow{col-1 \leftrightarrow col-k} A(i,k) \xrightarrow{row-1 \leftrightarrow row-k} A(i,k)$$

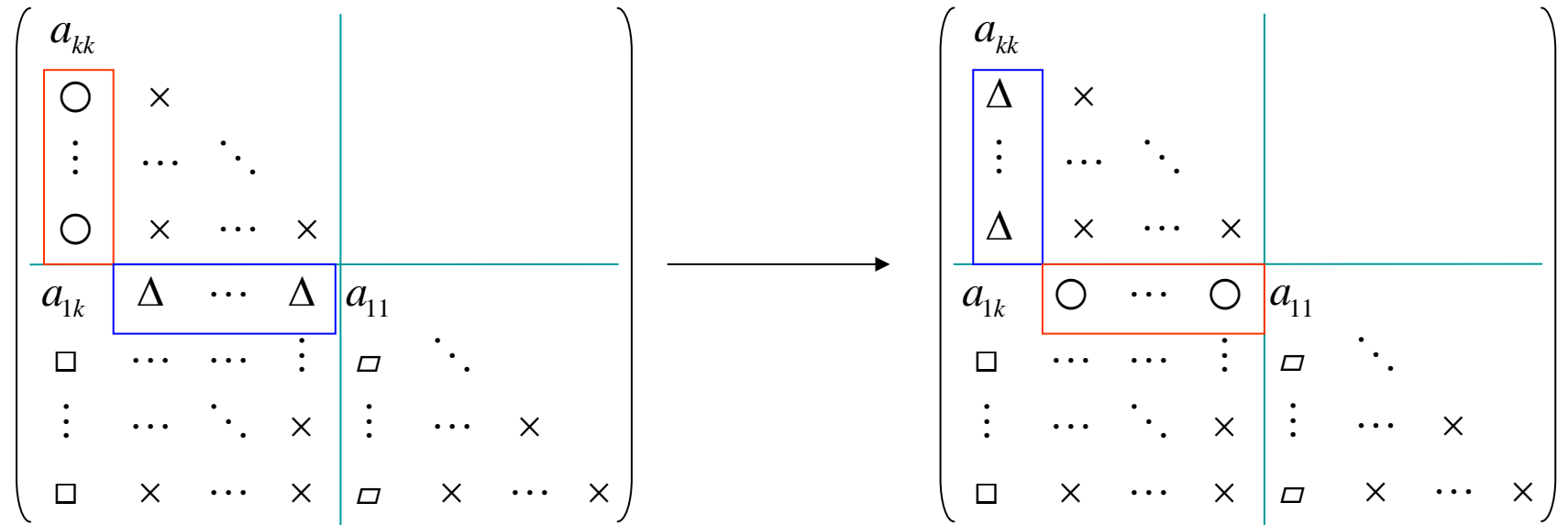
$$A(i,k)(i > k) \xrightarrow{col-1 \leftrightarrow col-k} A(i,1) \xrightarrow{row-1 \leftrightarrow row-k} A(i,1)$$

Therefore $A(i,1) \leftrightarrow A(i,k) \quad \forall k+1 \leq i \leq n$

Change diagonal element to (1, 1) position [7]

Step 3: interchange column 1 (from row-2 to row-(k-1)) and row k (from col-2 to col-(k-1))

$$A(2:k-1, 1) \leftrightarrow A(k, 2:k-1)$$



$$A(i, 1) (2 \leq i < k) \xrightarrow{col-1 \leftrightarrow col-k} A(i, k) \xrightarrow{row-1 \leftrightarrow row-k} A(i, k) (\text{upper triangle}) = A(k, i)$$

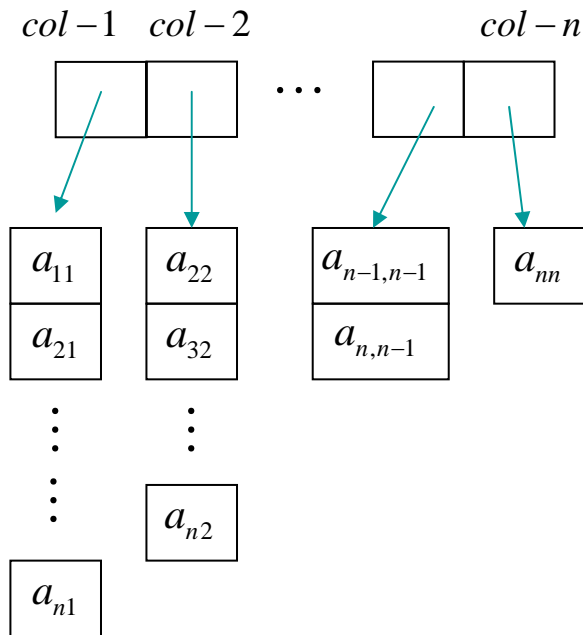
$$A(k, i) (2 \leq i < k) \xrightarrow{col-1 \leftrightarrow col-k} A(k, i) \xrightarrow{row-1 \leftrightarrow row-k} A(1, i) (\text{upper triangle}) = A(i, 1)$$

Therefore $A(i, 1) \leftrightarrow A(k, i) \quad \forall 2 \leq i \leq k-1$

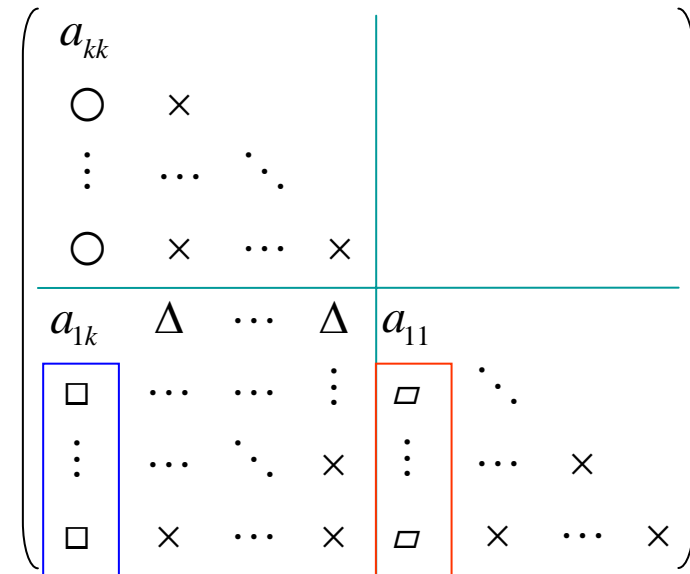
Implementation of symmetric permutation: swapping overhead [1]

Suppose we use col-major based pointer array to store real symmetric matrix A

$$\text{mem}(A) = \begin{pmatrix} a_{11} & & & & \\ a_{21} & a_{22} & & & \\ \vdots & \vdots & \ddots & & \\ a_{n1} & a_{n2} & \cdots & a_{nn} & \end{pmatrix}$$



Step 2: interchanging column 1 and k is O.K since memory block is contiguous.

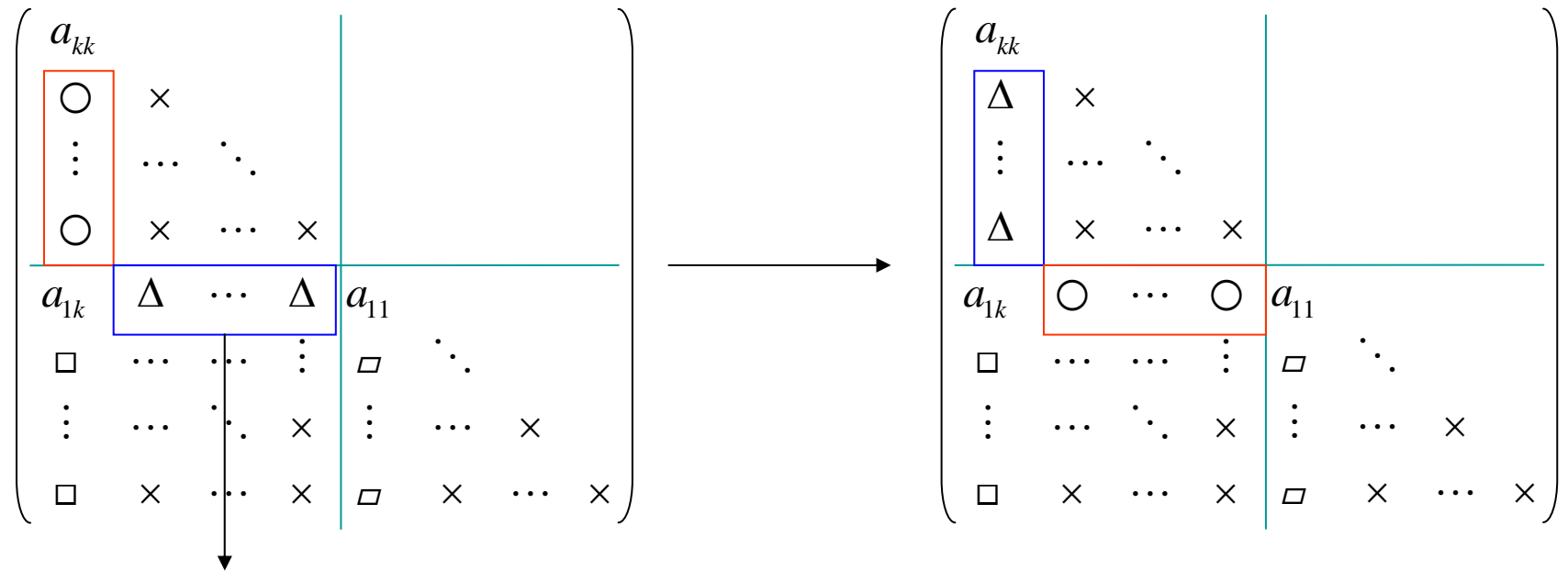


Contiguous memory block

Implementation of symmetric permutation: swapping overhead [2]

Step 3: interchanging column 1 and row k , i.e. $A(2:k-1,1) \leftrightarrow A(k,2:k-1)$

is **NOT** efficient since $A(k,2:k-1)$ is **NOT** contiguous.



NOT contiguous, swapping is very slow.

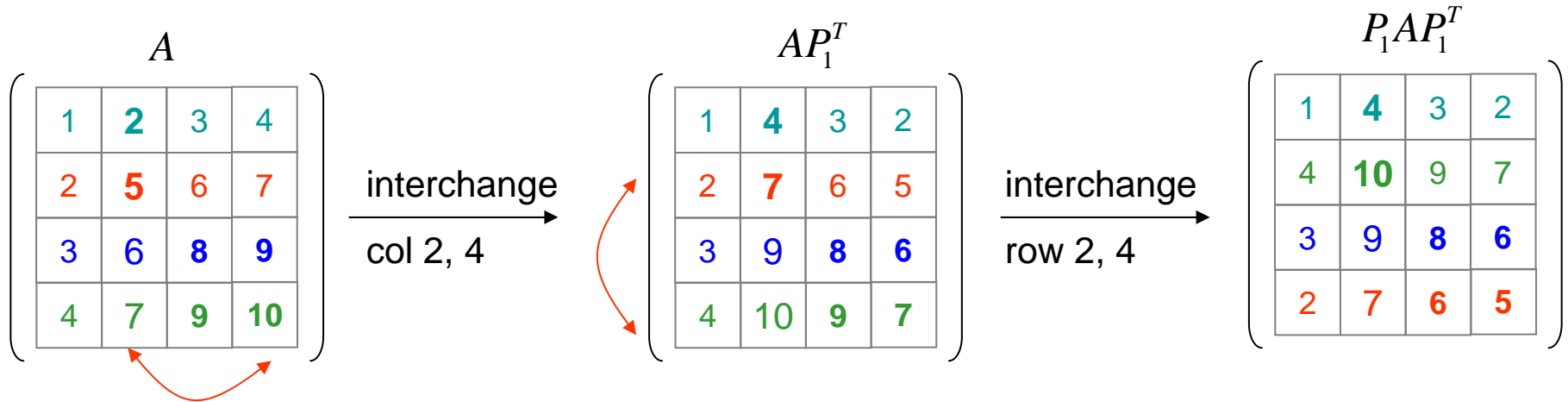
Observation: the situation is the same even we choose row-major based pointer array as storage strategy.

Solution: in order to avoid high swapping overhead, we should avoid permutation.

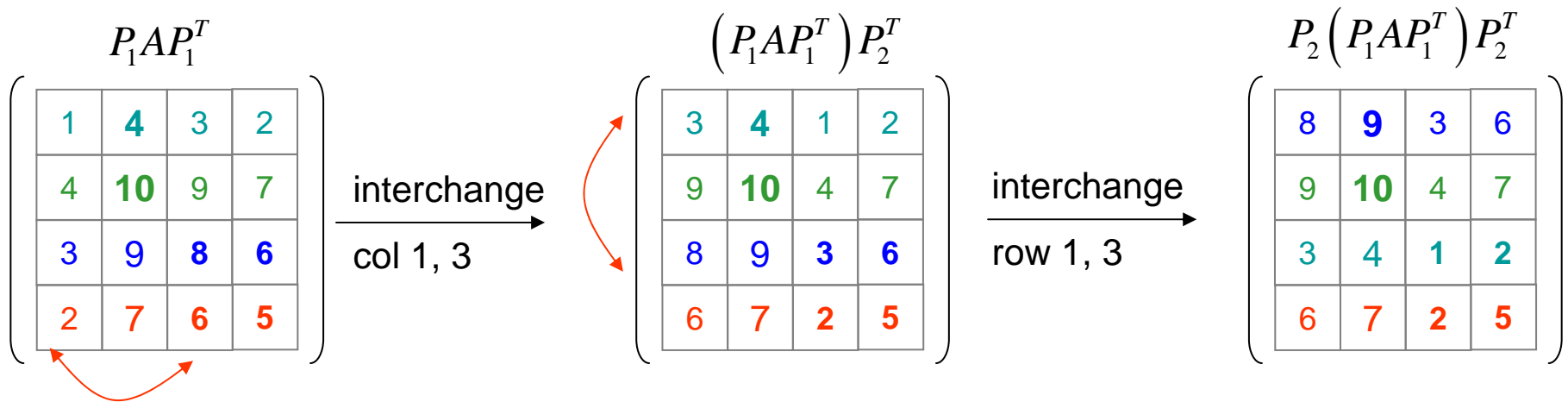
Change off-diagonal element to (2,1) position [1]

suppose we want to change $a_{43} = 9$ to position (2,1), then consider first permutation

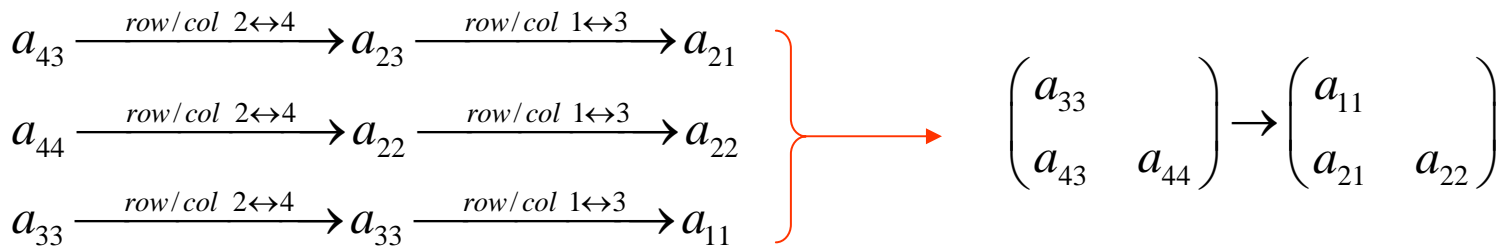
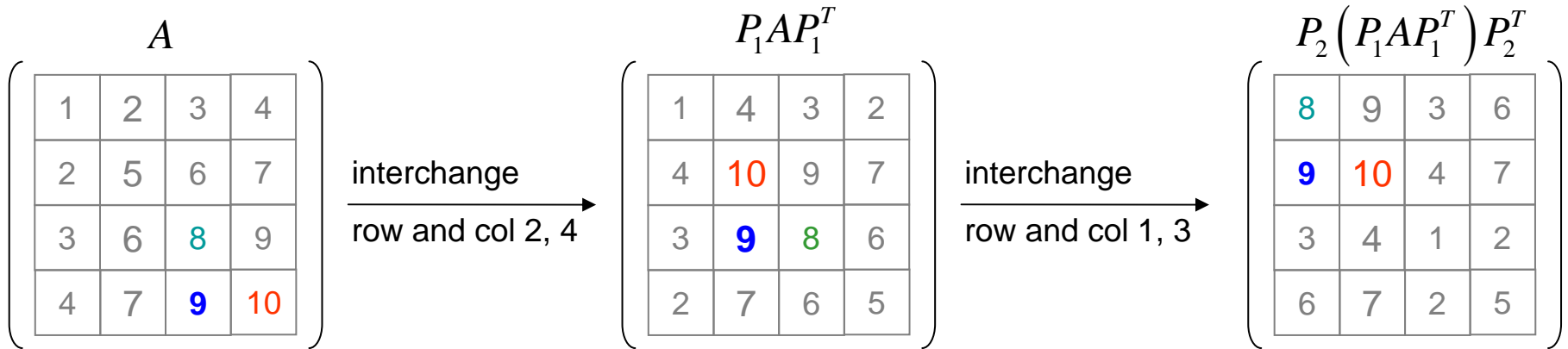
$P_1 = (1, 4, 3, 2)$ which change first index from 4 to 2



and second permutation $P_2 = (3, 2, 1, 4)$ which change second index from 3 to 1



Change off-diagonal element to (2, 1) position [2]



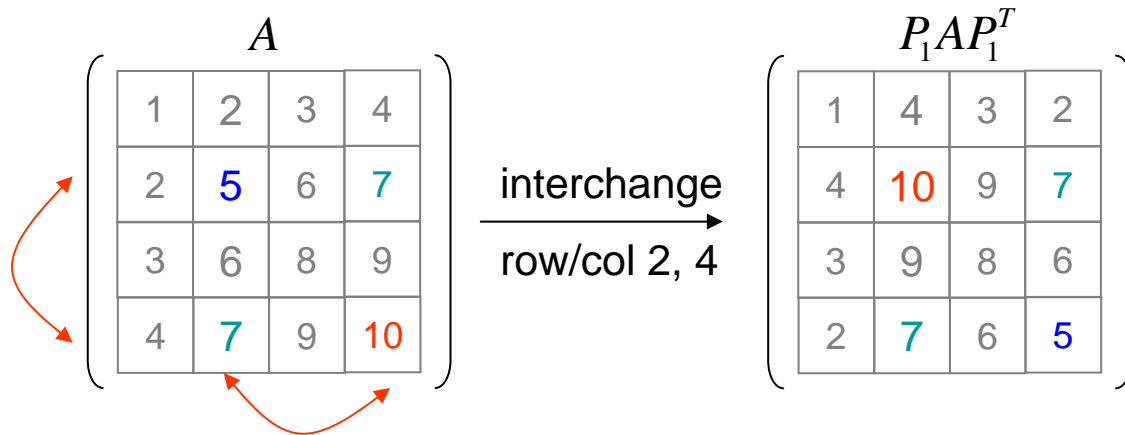
Observation: two permutation matrices can be computed easily since $2 \leftrightarrow 4$ and $1 \leftrightarrow 3$ are independent.

$$P_1 = (1, 4, 3, 2) \xrightarrow{P_2 = (3, 2, 1, 4)} P \equiv P_2 P_1 = (3, 4, 1, 2)$$

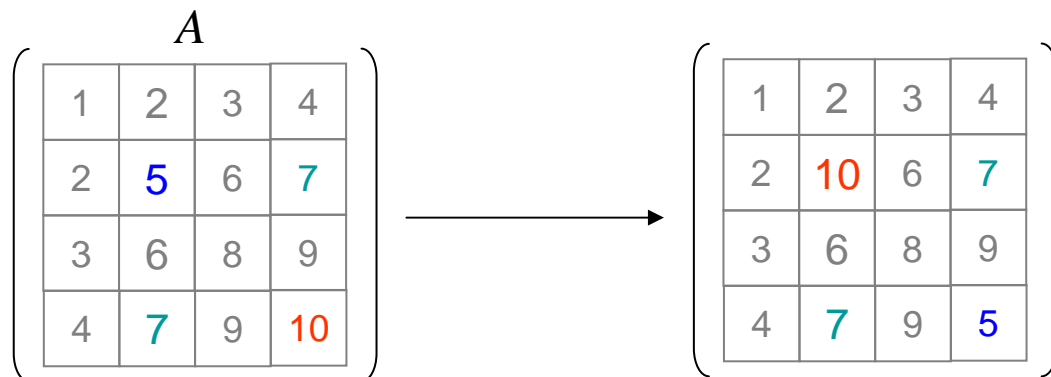
Change off-diagonal element to (2, 1) position [3]

HOW to transform $a_{43} = 9$ to position (2, 1), under only lower triangle part of A is stored?

Observation 1: $a_{22} \leftrightarrow a_{44}$ and $a_{42} = a_{24}$ does not changed



Step 1: interchange position (4,4) and (2,2), $a_{22} \leftrightarrow a_{44}$



Change off-diagonal element to (2,1) position [4]

Observation 2: we only need to update lower triangle part of A (diagonal term is excluded).

$$\begin{pmatrix} 1 & 2 & 3 & 4 \\ 2 & 10 & 6 & 7 \\ 3 & 6 & 8 & 9 \\ 4 & 7 & 9 & 5 \end{pmatrix} \longrightarrow \begin{pmatrix} & & & \\ 2 & & & \\ 3 & 6 & & \\ 4 & & 9 & \end{pmatrix}$$

Step 2: interchange row 2 and 4 below col 2, i.e. $A(2,1:1) \leftrightarrow A(4,1:1)$

$$\begin{pmatrix} & & & \\ & 2 & & \\ 3 & 6 & & \\ 4 & & 9 & \end{pmatrix} \xrightarrow[\text{row 2, 4}]{\text{interchange}} \begin{pmatrix} & & & \\ 4 & & 9 & \\ 3 & 6 & & \\ 2 & & 6 & \end{pmatrix} \xrightarrow[\text{col 2, 4}]{\text{interchange}} \begin{pmatrix} & & & \\ 4 & & 9 & \\ 3 & 9 & & 6 \\ 2 & & 6 & \end{pmatrix}$$

NOTE: Interchanging column 2, 4 does not change position (2,1) and (4,1), it suffices to interchange position (2,1) and (4,1) first.

Change off-diagonal element to (2, 1) position [5]

$$\begin{pmatrix} & & & \\ & & & \\ 2 & & & \\ 3 & 6 & & \\ 4 & & 9 & \end{pmatrix} \xrightarrow{A(2,1:1) \leftrightarrow A(4,1:1)} \begin{pmatrix} & & & \\ & & & \\ 4 & & & \\ 3 & 6 & & \\ 2 & & 9 & \end{pmatrix}$$

Step 3: interchange column 2 (from row-3 to row-3) and row 4 (from col-3 to col-3)

$$A(3:3, 2) \leftrightarrow A(4, 3:3)$$

$$\begin{pmatrix} & & & \\ & & & \\ 4 & & & \\ 3 & 6 & & \\ 2 & & 9 & \end{pmatrix} \xrightarrow{A(3:3, 2) \leftrightarrow A(4, 3:3)} \begin{pmatrix} & & & \\ & & & \\ 4 & & & \\ 3 & 9 & & \\ 2 & & 6 & \end{pmatrix}$$

$$a_{32} \xrightarrow{col-2 \leftrightarrow col-4} a_{34} \xrightarrow{row-2 \leftrightarrow row-4} a_{34} \text{ (upper triangle)} = a_{43}$$

$$a_{43} \xrightarrow{col-2 \leftrightarrow col-4} a_{43} \xrightarrow{row-2 \leftrightarrow row-4} a_{23} \text{ (upper triangle)} = a_{32}$$

Change off-diagonal element to (2,1) position [6]

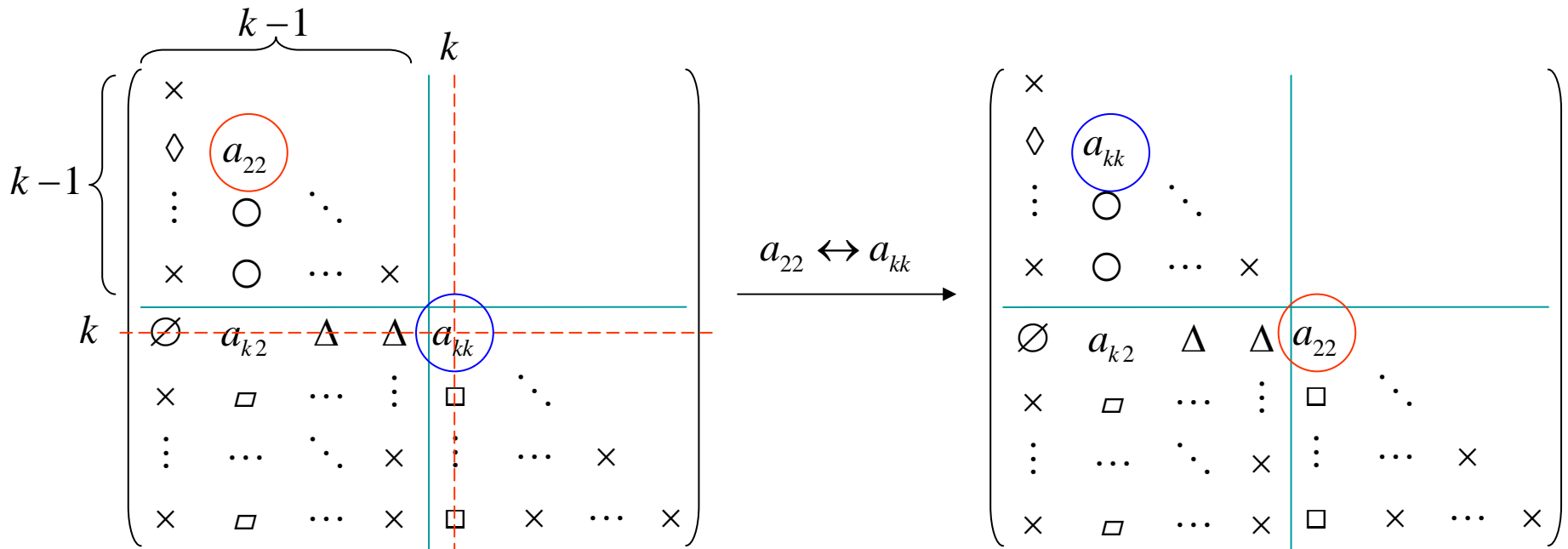
For general real symmetric matrix A with dimension n , suppose we want to change $a_{kk} \rightarrow a_{22}$

$$P = (1, \underset{\substack{\uparrow \\ 2\text{-th}}}{k}, 3, \dots, k-1, 2, \underset{\substack{\uparrow \\ k\text{-th}}}{k+1}, n)$$

interchanges rows $2, k$ or columns $2, k$

Step 1: interchange position $(2,2)$ and (k,k) , i.e. $a_{22} \leftrightarrow a_{kk}$

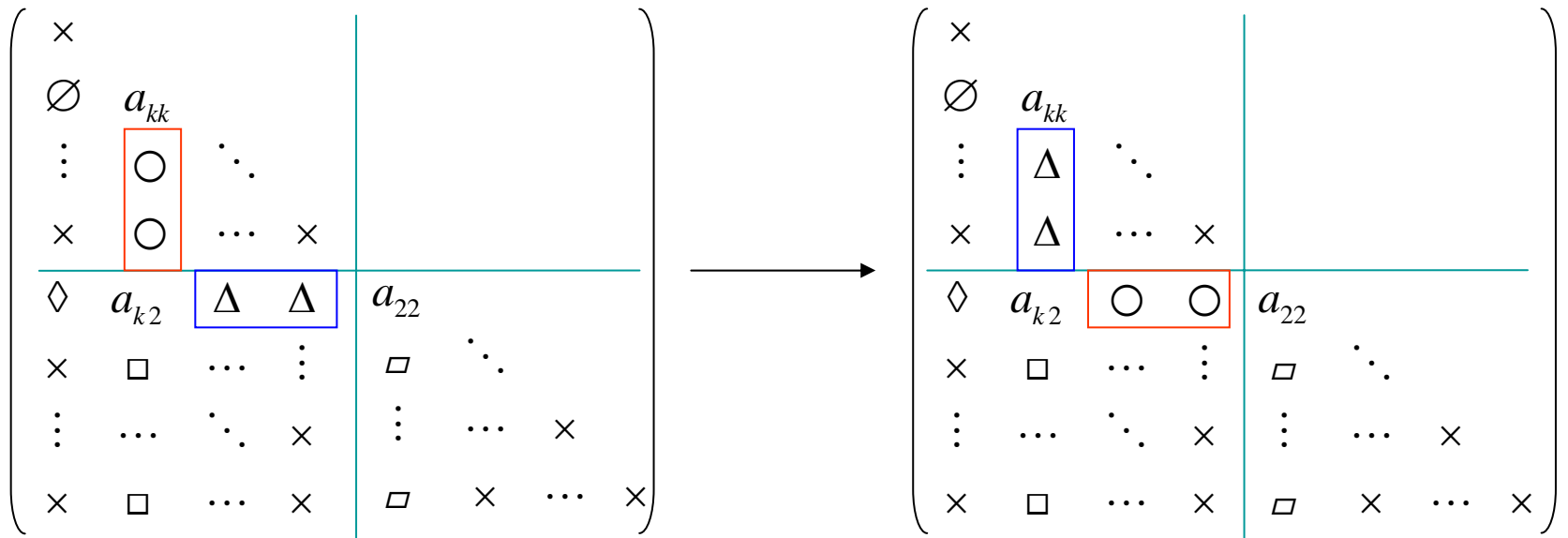
keep position $(k,2)$ and $(2,k)$, i.e. $a_{2k} = a_{k2}$ does not changed



Change off-diagonal element to (2, 1) position [9]

Step 4: interchange column 2 (from row-3 to row-(k-1)) and row k (from col-3 to col-(k-1))

$$A(3:k-1, 2) \leftrightarrow A(k, 3:k-1)$$



$$A(i, 2) (3 \leq i < k) \xrightarrow{col-2 \leftrightarrow col-k} A(i, k) \xrightarrow{row-2 \leftrightarrow row-k} A(i, k) (\text{upper triangle}) = A(k, i)$$

$$A(k, i) (3 \leq i < k) \xrightarrow{col-2 \leftrightarrow col-k} A(k, i) \xrightarrow{row-2 \leftrightarrow row-k} A(2, i) (\text{upper triangle}) = A(i, 2)$$

Therefore $A(i, 2) \leftrightarrow A(k, i) \quad \forall 3 \leq i \leq k-1$

OutLine

- Preliminary
- Symmetric permutation
- LDL' decomposition (diagonal pivoting)
 - growth rate of 1x1, 2x2 pivoting
 - criterion for pivot strategy
 - comparison to complete pivoting in GE
- Example of complete diagonal pivoting
- Algorithm of complete diagonal pivoting

LDL' decomposition: 1x1, 2x2 pivoting

- diagonal pivoting method with **complete** pivoting:
[Bunch-Parlett](#), "Direct methods for solving symmetric indefinite systems of linear equations," SIAM J. Numer. Anal., v. 8, 1971, pp. 639-655
- diagonal pivoting method with **partial** pivoting:
[Bunch-Kaufman](#), "Some Stable Methods for Calculating Inertia and Solving Symmetric Linear Systems," Mathematics of Computation, volume 31, number 137, January 1977, page 163-179

$$A \equiv \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & A^{(n-s)} \end{pmatrix} \begin{pmatrix} I & E^{-1}c^T \\ & I \end{pmatrix} \quad E \text{ is of order } s$$

Define $L = \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix}$ and constant $\begin{cases} \mu_0 = \max_{1 \leq i, j \leq n} |A_{ij}| = \text{maximal element of matrix } A \\ \mu_1 = \max_{1 \leq i \leq n} |A_{ii}| = \text{maximal diagonal element of matrix } A \\ \nu = |\det E| \end{cases}$

Case 1: $s = 1$

$$A = \left(\begin{array}{c|ccc} a_{11} & a_{21} & \cdots & a_{n1} \\ \hline a_{21} & a_{22} & \cdots & \vdots \\ \vdots & \cdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{array} \right) = L \tilde{A}^{(n-1)} L^T \quad \text{where} \quad \tilde{A}^{(n-1)} = \begin{pmatrix} a_{11} & \\ & A^{(n-1)} \end{pmatrix} \quad E = a_{11} \quad \text{with} \quad \nu = |\det E| = |a_{11}| > 0$$

$$c = \begin{pmatrix} a_{21} \\ a_{31} \\ \vdots \\ a_{n1} \end{pmatrix} \quad \text{with} \quad c_j = a_{j+1,1} \quad \text{and} \quad B = \begin{pmatrix} a_{22} & & & \\ a_{23} & a_{33} & & \\ \vdots & \cdots & \ddots & \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix} = B^T \quad \text{with} \quad B_{ij} = a_{i+1, j+1}$$

Growth rate in LDL' decomposition

[2]

$$cE^{-1} = \frac{1}{a_{11}} \begin{pmatrix} a_{21} \\ a_{31} \\ \vdots \\ a_{n1} \end{pmatrix} \rightarrow |cE^{-1}|_{\infty} \leq \max_{2 \leq j \leq n} \left| \frac{a_{j1}}{a_{11}} \right| \leq \frac{1}{\nu} \max_{2 \leq j \leq n} |a_{j1}| \leq \frac{\mu_0}{\nu}$$

Therefore $|L|_{\infty} = \max\left(1, |cE^{-1}|_{\infty}\right) \leq \max\left(1, \frac{\mu_0}{\nu}\right)$

Observation: in $PA=LU$, we choose $|a_{11}| = \max |A(:,1)|$ such that $|L|_{\infty} \leq 1$

$$A^{(n-1)} = B - cE^{-1}c^T = B - \frac{cc^T}{a_{11}} \quad A_{ij}^{(n-1)} = B_{ij} - \frac{c_i c_j}{a_{11}} = a_{i+1,j+1} - \frac{a_{i+1,1} a_{j+1,1}}{a_{11}}$$

$$|A_{ij}^{(n-1)}| \leq |a_{i+1,j+1}| + \frac{|a_{i+1,1}| \cdot |a_{j+1,1}|}{|a_{11}|} \leq \mu_0 + \frac{\mu_0^2}{\nu} \quad \text{implies} \quad |A^{(n-1)}|_{\infty} \leq \mu_0 + \frac{\mu_0^2}{\nu}$$

Therefore $|\tilde{A}^{(n-1)}|_{\infty} \leq \max\left(|a_{11}|, |A^{(n-1)}|_{\infty}\right) \leq \max\left(\nu, \mu_0 + \frac{\mu_0^2}{\nu}\right) \leq \mu_0 + \frac{\mu_0^2}{\nu}$

Observation: large element growth will not occur for a 1×1 pivot if ν is large relative to μ_0

Case 2: $s = 2$

$$A = \left(\begin{array}{cc|ccc} a_{11} & a_{21} & \cdots & a_{n1} & \\ a_{21} & a_{22} & \cdots & \vdots & \\ \vdots & \cdots & \ddots & \vdots & \\ a_{n1} & a_{n2} & \cdots & a_{nn} & \end{array} \right) = L\tilde{A}^{(n-2)}L^T \quad \text{where} \quad \tilde{A}^{(n-2)} = \begin{pmatrix} E & \\ & A^{(n-2)} \end{pmatrix}, \quad L = \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix} \quad \text{and} \quad A^{(n-2)} = B - cE^{-1}c^T$$

$$E = \begin{pmatrix} a_{11} & a_{21} \\ a_{21} & a_{22} \end{pmatrix} \quad \text{with} \quad \nu = |\det E| = |a_{11}a_{22} - a_{21}^2| > 0 \quad \text{and} \quad E^{-1} = \frac{1}{\det E} \begin{pmatrix} a_{22} & -a_{21} \\ -a_{21} & a_{11} \end{pmatrix}$$

$$c = \begin{pmatrix} a_{31} & a_{32} \\ a_{41} & a_{42} \\ \vdots & \vdots \\ a_{n1} & a_{n2} \end{pmatrix} \quad \text{with} \quad c_{ij} = a_{i+2,j} \quad \text{and} \quad B = \begin{pmatrix} a_{33} & & \\ a_{43} & a_{44} & \\ \vdots & \cdots & \ddots \\ a_{n3} & a_{n4} & \cdots & a_{nn} \end{pmatrix} = B^T \quad \text{with} \quad B_{ij} = a_{i+2,j+2}$$

$$(cE^{-1})_{ij} = c(i,:)E^{-1}(:,j) \quad \text{implies} \quad \begin{cases} (cE^{-1})_{i1} = (c_{i1} \quad c_{i2}) \frac{1}{\det E} \begin{pmatrix} a_{22} \\ -a_{21} \end{pmatrix} \\ (cE^{-1})_{i2} = (c_{i1} \quad c_{i2}) \frac{1}{\det E} \begin{pmatrix} -a_{21} \\ a_{11} \end{pmatrix} \end{cases}$$

Growth rate in LDL' decomposition

[4]

for $2 < i$

$$\left\{ \begin{array}{l} L_{i1} = (cE^{-1})_{i-1,1} = (a_{i1} \quad a_{i2}) \frac{1}{\det E} \begin{pmatrix} a_{22} \\ -a_{21} \end{pmatrix} = \frac{a_{i1}a_{22} - a_{i2}a_{21}}{\det E} \\ L_{i2} = (cE^{-1})_{i-1,2} = (a_{i1} \quad a_{i2}) \frac{1}{\det E} \begin{pmatrix} -a_{21} \\ a_{11} \end{pmatrix} = \frac{-a_{i1}a_{21} + a_{i2}a_{11}}{\det E} \end{array} \right.$$

$$\left\{ \begin{array}{l} |L_{i1}| \leq \frac{|a_{i1}||a_{22}| + |a_{i2}||a_{21}|}{|\det E|} \leq \frac{\mu_0\mu_1 + \mu_0^2}{\nu} = \frac{\mu_0}{\nu}(\mu_0 + \mu_1) \\ |L_{i2}| \leq \frac{|a_{i1}||a_{21}| + |a_{i2}||a_{11}|}{|\det E|} \leq \frac{\mu_0\mu_0 + \mu_0\mu_1}{\nu} = \frac{\mu_0}{\nu}(\mu_0 + \mu_1) \end{array} \right.$$

Therefore $\boxed{|L|_{\infty} = \max\left(1, |cE^{-1}|_{\infty}\right) \leq \max\left(1, \frac{\mu_0}{\nu}(\mu_0 + \mu_1)\right)}$

$$\begin{aligned}
 A_{ij}^{(n-2)} &= B_{ij} - (cE^{-1})(i,:)c^T(:,j) = B_{ij} - c(i,:)E^{-1}c(j,:) \\
 &= a_{i+2,j+2} - \begin{pmatrix} a_{i+2,1} & a_{i+2,2} \end{pmatrix} \frac{1}{\det E} \begin{pmatrix} a_{22} & -a_{21} \\ -a_{21} & a_{11} \end{pmatrix} \begin{pmatrix} a_{j+2,1} \\ a_{j+2,2} \end{pmatrix} \\
 &= a_{i+2,j+2} - \frac{1}{\det E} \begin{pmatrix} a_{i+2,1}a_{22} - a_{i+2,2}a_{21} & -a_{i+2,1}a_{21} + a_{i+2,2}a_{11} \end{pmatrix} \begin{pmatrix} a_{j+2,1} \\ a_{j+2,2} \end{pmatrix} \\
 &= a_{i+2,j+2} - \frac{(a_{i+2,1}a_{22} - a_{i+2,2}a_{21})a_{j+2,1} + (-a_{i+2,1}a_{21} + a_{i+2,2}a_{11})a_{j+2,2}}{\det E}
 \end{aligned}$$

$$\begin{aligned}
 |A_{ij}^{(n-2)}| &\leq |a_{i+2,j+2}| + \frac{(|a_{i+2,1}||a_{22}| + |a_{i+2,2}||a_{21}|)|a_{j+2,1}| + (|a_{i+2,1}||a_{21}| + |a_{i+2,2}||a_{11}|)|a_{j+2,2}|}{|\det E|} \\
 &\leq \mu_0 + \frac{(\mu_0\mu_1 + \mu_0\mu_0)\mu_0 + (\mu_0\mu_0 + \mu_0\mu_1)\mu_0}{\nu} = \mu_0 \left[1 + \frac{2\mu_0(\mu_0 + \mu_1)}{\nu} \right]
 \end{aligned}$$

Therefore

$$\left| \tilde{A}^{(n-2)} \right|_{\infty} \leq \max \left(|E|_{\infty}, |A^{(n-2)}|_{\infty} \right) \leq \max \left(\mu_0, \mu_0 \left[1 + \frac{2\mu_0(\mu_0 + \mu_1)}{\nu} \right] \right) = \mu_0 \left[1 + \frac{2\mu_0(\mu_0 + \mu_1)}{\nu} \right]$$

Criterion for pivot strategy [1]

Fix a number $0 < \alpha < 1$ (later on, we will determine optimal value of α)

$$A = A^{(n)} = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & A^{(n-s)} \end{pmatrix} L^T$$

$$\mu_0 = \max_{1 \leq i, j \leq n} |A_{ij}| = \text{maximal element of matrix } A$$

$$\mu_1 = \max_{1 \leq i \leq n} |A_{ii}| = \text{maximal diagonal element of matrix } A$$

$$\nu = |\det E|$$

Case 1: $\mu_1 \geq \alpha \mu_0$

suppose $\mu_1 = |A_{kk}| = \max_{1 \leq i \leq n} |A_{ii}|$ we interchange 1st and k -th rows and columns by introducing permutation matrix $P_1 = (k, 2, 3, \dots, k-1, 1, k+1, \dots, n)$ and do symmetric permutation

$$\hat{A}^{(n)} = P_1 A^{(n)} P_1^T$$

$$A^{(n)} = \begin{pmatrix} a_{11} & \cdots & a_{k1} & \times \\ \vdots & \ddots & \vdots & \times \\ a_{k1} & \cdots & a_{kk} & \times \\ \times & \times & \times & \times \end{pmatrix} \longrightarrow \tilde{A}^{(n)} = \begin{pmatrix} a_{kk} & \cdots & a_{k1} & \times \\ \vdots & \ddots & \vdots & \times \\ a_{k1} & \cdots & a_{11} & \times \\ \times & \times & \times & \times \end{pmatrix}$$

Then do LDL^T on $\hat{A}^{(n)}$ with 1×1 pivot, written as $\hat{A}^{(n)} = \begin{pmatrix} \hat{a}_{11} & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ c/\hat{a}_{11} & I \end{pmatrix} \begin{pmatrix} \hat{a}_{11} & \\ & A^{(n-1)} \end{pmatrix} L_1^T$

Criterion for pivot strategy [2]

Recall for 1×1 pivot, $A = L\tilde{A}^{(n-1)}L^T$, we have growth rate

$$\boxed{|L|_\infty \leq \max\left(1, \frac{\mu_0}{\nu}\right)} \quad \text{and} \quad \boxed{|\tilde{A}^{(n-1)}|_\infty \leq \mu_0 + \frac{\mu_0^2}{\nu}}$$

Now for $\hat{A}^{(n)} = \begin{pmatrix} \hat{a}_{11} & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ c/\hat{a}_{11} & I \end{pmatrix} \begin{pmatrix} \hat{a}_{11} & \\ & A^{(n-1)} \end{pmatrix} L_1^T$, $\nu = |\hat{a}_{11}| = |a_{kk}| = \mu_1 \geq \alpha\mu_0$

$$\boxed{|L_1|_\infty \leq \max\left(1, \frac{\mu_0}{\nu}\right) \leq \max\left(1, \frac{1}{\alpha}\right) = \frac{1}{\alpha}} \quad \text{and} \quad \boxed{|\tilde{A}^{(n-1)}|_\infty \leq \mu_0 + \frac{\mu_0^2}{\nu} \leq \left(1 + \frac{1}{\alpha}\right)\mu_0}$$

Case 2: $\mu_1 < \alpha\mu_0$

suppose $\mu_0 = |A_{rq}| = \max_{1 \leq i, j \leq n} |A_{ij}|$, $r > q$, we interchange q -th and 1 st rows and columns,

and then r -th and 2 nd rows and columns by introducing permutation matrix

$$P_1 = (1, r, 3, \dots, r-1, 2, r+1, \dots, n) \cdot (q, 2, 3, \dots, q-1, 1, q+1, \dots, n)$$

and do symmetric permutation $\hat{A}^{(n)} = P_1 A^{(n)} P_1^T$

Question: we must change $q \leftrightarrow 1$ first, then change $r \leftrightarrow 2$, why?

Criterion for pivot strategy [3]

$P_1 = (1, r, 3, \dots, r-1, 2, r+1, \dots, n) \cdot (q, 2, 3, \dots, q-1, 1, q+1, \dots, n)$ transforms $A_{rq} \rightarrow A_{21}$

1 $r > q > 3$

$$P_1 = (q, r, 3, \dots, q-1, \underset{\substack{\uparrow \\ q-th}}{1}, q+1, \dots, r-1, \underset{\substack{\uparrow \\ r-th}}{2}, r+1, \dots, n)$$

2 $r > q = 3$

$$P_1 = (3, r, 1, 4, \dots, r-1, \underset{\substack{\uparrow \\ r-th}}{2}, r+1, \dots, n)$$

3 $3 = r > q = 2$ (boundary case)

$$P_1 = (1, 3, 2, 4, \dots, n) \cdot (2, 1, 3, 4, \dots, n) = (2, 3, 1, 4, \dots, n)$$

but $(2, 1, 3, 4, \dots, n) \cdot (1, 3, 2, 4, \dots, n) = (3, 1, 2, 4, \dots, n)$

Criterion for pivot strategy [4]

$$A^{(n)} = \begin{pmatrix} a_{11} & a_{21} & \cdots & a_{q1} & \times & \times & \times \\ a_{21} & a_{22} & \cdots & \vdots & \times & a_{r2} & \times \\ \vdots & \cdots & \ddots & \vdots & \times & \times & \times \\ a_{q1} & \cdots & \cdots & a_{qq} & \cdots & a_{rq} & \times \\ \times & \times & \times & \vdots & \ddots & \vdots & \times \\ \times & a_{r2} & \times & a_{rq} & \cdots & a_{rr} & \times \\ \times & \times & \times & \times & \times & \times & \times \end{pmatrix} \xrightarrow[r \leftrightarrow 2]{q \leftrightarrow 1} \hat{A}^{(n)} = \begin{pmatrix} a_{qq} & a_{rq} & \cdots & a_{q1} & \times & \times & \times \\ a_{rq} & a_{rr} & \cdots & \vdots & \times & a_{r2} & \times \\ \vdots & \cdots & \ddots & \vdots & \times & \times & \times \\ a_{q1} & \cdots & \cdots & a_{11} & \cdots & a_{21} & \times \\ \times & \times & \times & \vdots & \ddots & \vdots & \times \\ \times & a_{r2} & \times & a_{21} & \cdots & a_{22} & \times \\ \times & \times & \times & \times & \times & \times & \times \end{pmatrix}$$

2x2 pivot

Then do LDL^T on $\hat{A}^{(n)}$ with 2×2 pivot, written as $\hat{A}^{(n)} = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & A^{(n-2)} \end{pmatrix} L_1^T$

Recall for 2×2 pivot, $A = L\tilde{A}^{(n-2)}L^T$, we have growth rate

$$|L|_\infty \leq \max \left(1, \frac{\mu_0}{\nu} (\mu_0 + \mu_1) \right) \quad \text{and} \quad |\tilde{A}^{(n-2)}|_\infty \leq \mu_0 \left[1 + \frac{2\mu_0 (\mu_0 + \mu_1)}{\nu} \right]$$

Criterion for pivot strategy [5]

$$\text{Now for } \hat{A}^{(n)} = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & A^{(n-2)} \end{pmatrix} L_1^T$$

$$\nu = |\det E| = |a_{qq}a_{rr} - a_{rq}^2| = a_{rq}^2 - a_{qq}a_{rr} \geq \mu_0^2 - \mu_1^2 > (1 - \alpha^2)\mu_0^2$$

$$|L_1|_\infty \leq \max\left(1, \frac{\mu_0}{\nu}(\mu_0 + \mu_1)\right) \leq \max\left(1, \frac{1}{1 - \alpha}\right) = \frac{1}{1 - \alpha}$$

$$|\tilde{A}^{(n-2)}|_\infty \leq \mu_0 \left[1 + \frac{2\mu_0(\mu_0 + \mu_1)}{\nu}\right] \leq \left(1 + \frac{2}{1 - \alpha}\right)\mu_0$$

To sum up

Case 1: $\mu_1 \geq \alpha\mu_0$

$$\hat{A}^{(n)} = \begin{pmatrix} \hat{a}_{11} & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ c/\hat{a}_{11} & I \end{pmatrix} \begin{pmatrix} \hat{a}_{11} & \\ & A^{(n-1)} \end{pmatrix} L_1^T,$$

$$|L_1|_\infty \leq \frac{1}{\alpha}$$

and

$$|\tilde{A}^{(n-1)}|_\infty \leq \left(1 + \frac{1}{\alpha}\right)\mu_0$$

Case 2: $\mu_1 < \alpha\mu_0$

$$\hat{A}^{(n)} = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & A^{(n-2)} \end{pmatrix} L_1^T$$

$$|L_1|_\infty \leq \frac{1}{1 - \alpha}$$

and

$$|\tilde{A}^{(n-2)}|_\infty \leq \left(1 + \frac{2}{1 - \alpha}\right)\mu_0$$

Question: How to choose value of $0 < \alpha < 1$

Criterion for pivot strategy [6]

worst case analysis : choose $0 < \alpha < 1$ such that

$$\left[\text{growth rate of } 1 \times 1 \text{ pivot} + 1 \times 1 \text{ pivot} \right] \approx \left[\text{growth rate of } 2 \times 2 \text{ pivot} \right]$$

or equivalently $\left[\text{growth rate of } 1 \times 1 \text{ pivot} \right] \approx \sqrt{\left[\text{growth rate of } 2 \times 2 \text{ pivot} \right]}$

Define $B(\alpha) = \max\left(1 + \frac{1}{\alpha}, \sqrt{1 + \frac{2}{1-\alpha}}\right)$

$$\min_{0 < \alpha < 1} B(\alpha) = B(\alpha_0) = \frac{1 + \sqrt{17}}{2} \approx 2.5616 < 2.57$$

where $\alpha_0 = \frac{1 + \sqrt{17}}{8} \approx 0.6404$ satisfies $1 + \frac{1}{\alpha_0} = \sqrt{1 + \frac{2}{1-\alpha_0}}$

Exercise: verify $\min_{0 < \alpha < 1} B(\alpha) = B(\alpha_0) = \frac{1 + \sqrt{17}}{2}$ and $\alpha_0 = \frac{1 + \sqrt{17}}{8}$

Diagonal pivoting versus complete pivoting of Gaussian Elimination [1]

- 1 We must search for the largest element in each reduced matrix, this is a complete pivoting strategy analogous to Gaussian Elimination with complete pivoting

$$\mu_0 = \max_{1 \leq i, j \leq n} |A_{ij}| = \text{maximal element of matrix } A$$

$$\mu_1 = \max_{1 \leq i \leq n} |A_{ii}| = \text{maximal diagonal element of matrix } A$$

complete pivoting of *GE* (*Gaussian Elimination*)

(1) find a $p, q \in \{k, k+1, \dots, n\}$ such that $|A_{pq}^{(k)}| = \max |A^{(k)}(k:n, k:n)|$

(2) swap row $A^{(k)}(k, 1:n)$ and row $A^{(k)}(p, 1:n)$

(2) swap column $A^{(k)}(1:n, k)$ and column $A^{(k)}(1:n, q)$

then $PA\Pi = LU$ with $|L|_{\infty} \leq 1$

Diagonal pivoting versus complete pivoting of Gaussian Elimination [2]

2 growth rate

diagonal pivoting $PAP^T = LDL^T$

$$|L|_{\infty} \leq \max\left(\frac{1}{\alpha_0}, \frac{1}{1-\alpha_0}\right) = \max(1.56, 2.78) = 2.78$$

$$|D|_{\infty} \leq B(\alpha_0)^{n-1} \mu_0 \leq 2.57^{n-1} \mu_0$$

complete pivoting in GE $PA\Pi = LU$

$$|L|_{\infty} \leq 1$$

$$|U|_{\infty} \leq ?$$

Remark: Bunch [1] proves that the element growth in the diagonal pivoting method

with complete pivoting is bounded by $3n \cdot f(n)$ in comparison with $\sqrt{n} \cdot f(n)$

for Gaussian Elimination with complete pivoting, where

$$f(n) = \sqrt{\prod_{k=2}^n k^{1/(k-1)}} < 1.8 \cdot n^{\frac{1}{4} \log n}$$

[1] J.R. Bunch, "Analysis of the diagonal pivoting method", SIAM J. Numer. Anal., v. 8, 1971, pp. 656-680

OutLine

- Preliminary
- Symmetric permutation
- LDL' decomposition (diagonal pivoting)
- **Example of complete diagonal pivoting**
- Algorithm of complete diagonal pivoting

Example (complete pivoting) [1]

$$A^{(1)} = A = \begin{pmatrix} 6 & 12 & 3 & -6 \\ 12 & -8 & -13 & 4 \\ 3 & -13 & -7 & 1 \\ -6 & 4 & 1 & 6 \end{pmatrix}$$

$$\mu_0 = \max_{1 \leq i, j \leq n} |A_{ij}| = |A_{32}| = 13$$

$$\mu_1 = \max_{1 \leq i \leq n} |A_{ii}| = |A_{22}| = 8 \quad \alpha_0 \approx 0.6404$$

$\mu_1 < \alpha_0 \mu_0 = 8.33$ we choose 2×2 pivot

swap row/column $2 \leftrightarrow 1$ and $3 \leftrightarrow 2$ by permutation matrix

$$P_1 = (2, 3, 1, 4)$$

$$\begin{pmatrix} 6 & 12 & 3 & -6 \\ 12 & -8 & -13 & 4 \\ 3 & -13 & -7 & 1 \\ -6 & 4 & 1 & 6 \end{pmatrix}$$

$2 \leftrightarrow 1$

$$\begin{pmatrix} -8 & -13 & -7 & 1 \\ 12 & 6 & 3 & -6 \\ -13 & 3 & -7 & 1 \\ 4 & -6 & 1 & 6 \end{pmatrix}$$

$3 \leftrightarrow 2$

$$\tilde{A}^{(1)} = P_1 A^{(1)} P_1^T = \begin{pmatrix} -8 & -13 & 12 & 4 \\ -13 & -7 & 3 & 1 \\ 12 & 3 & 6 & -6 \\ 4 & 1 & -6 & 6 \end{pmatrix}$$

$$\begin{pmatrix} E & c^T \\ c & B \end{pmatrix}$$

Example (complete pivoting) [2]

$$P_1 A^{(1)} P_1^T = \tilde{A}^{(1)} = \begin{pmatrix} E & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix} \begin{pmatrix} E & \\ & B - cE^{-1}c^T \end{pmatrix} \begin{pmatrix} I & \\ cE^{-1} & I \end{pmatrix}^T = L^{(1)} A^{(3)} \left(L^{(1)}\right)^T$$

$$\tilde{A}^{(1)} = P_1 A^{(1)} P_1^T = \begin{pmatrix} \begin{array}{|c|c|c|c|} \hline -8 & -13 & 12 & 4 \\ \hline -13 & -7 & 3 & 1 \\ \hline 12 & 3 & 6 & -6 \\ \hline 4 & 1 & -6 & 6 \\ \hline \end{array} & = & \begin{pmatrix} \begin{array}{|c|c|c|c|} \hline 1 & & & \\ \hline & 1 & & \\ \hline 0.3982 & -1.1681 & 1 & \\ \hline 0.1327 & -0.3894 & & 1 \\ \hline \end{array} & \begin{pmatrix} \begin{array}{|c|c|c|c|} \hline -8 & -13 & & \\ \hline -13 & -7 & & \\ \hline & & 4.7257 & -6.4248 \\ \hline & & -6.4248 & 5.8584 \\ \hline \end{array} & \left(L^{(1)}\right)^T$$

Recursively do the same procedure for $A^{(3)} (3:4, 3:4)$

$$\mu_0 = \max_{3 \leq i, j \leq 4} |A_{ij}^{(3)}| = |A_{43}^{(3)}| = 6.4248$$

$$\mu_1 = \max_{3 \leq i \leq 4} |A_{ii}^{(3)}| = |A_{44}^{(3)}| = 5.8584$$

$$\alpha_0 \approx 0.6404$$

$\mu_1 > \alpha_0 \mu_0 = 4.1144$ we choose 1×1 pivot

swap row/column $3 \leftrightarrow 4$ by permutation matrix $P_3 = (1, 2, 4, 3)$ such that $A_{44}^{(3)} \rightarrow A_{33}^{(3)}$

$$P_3 \tilde{A}^{(1)} P_3^T = \left[P_3 L^{(1)} P_3^T \right] \left(P_3 A^{(3)} P_3^T \right) \left[P_3 \left(L^{(1)}\right)^T P_3^T \right] = \tilde{L}^{(1)} \tilde{A}^{(3)} \left(\tilde{L}^{(1)}\right)^T$$

Example (complete pivoting) [3]

Do symmetry permutation for $A^{(3)}$ and $L^{(1)}$

$$\begin{array}{c}
 A^{(3)} \\
 \left(\begin{array}{cc|cc}
 -8 & -13 & & \\
 -13 & -7 & & \\
 \hline
 & & 4.7257 & -6.4248 \\
 & & -6.4248 & 5.8584
 \end{array} \right)
 \xrightarrow{P_3 = (1, 2, 4, 3)}
 \begin{array}{c}
 \tilde{A}^{(3)} = P_3 A^{(3)} P_3^T \\
 \left(\begin{array}{cc|cc}
 -8 & -13 & & \\
 -13 & -7 & & \\
 \hline
 & & 5.8584 & -6.4248 \\
 & & -6.4248 & 4.7257
 \end{array} \right)
 \end{array}
 \end{array}$$

$$\begin{array}{c}
 L^{(1)} \\
 \left(\begin{array}{cc|cc}
 1 & & & \\
 & 1 & & \\
 \hline
 0.3982 & -1.1681 & 1 & \\
 0.1327 & -0.3894 & & 1
 \end{array} \right)
 \xrightarrow{P_3 = (1, 2, 4, 3)}
 \begin{array}{c}
 \tilde{L}^{(1)} = P_3 L^{(1)} P_3^T \\
 \left(\begin{array}{cc|cc}
 1 & & & \\
 & 1 & & \\
 \hline
 0.1327 & -0.3894 & 1 & \\
 0.3982 & -1.1681 & & 1
 \end{array} \right)
 \end{array}
 \end{array}$$

we do 1×1 pivot on $\tilde{A}^{(3)}(3:4, 3:4) = \begin{pmatrix} 5.8584 & -6.4248 \\ -6.4248 & 4.7257 \end{pmatrix}$

Example (complete pivoting) [4]

$$\tilde{A}^{(3)}(3:4,3:4) = \begin{pmatrix} a_{33}^{(3)} & c^T \\ c & B \end{pmatrix} = \begin{pmatrix} 1 & \\ c/a_{33}^{(3)} & I \end{pmatrix} \begin{pmatrix} a_{33}^{(3)} & \\ & B - cc^T/a_{33}^{(3)} \end{pmatrix} \begin{pmatrix} I & \\ c/a_{33}^{(3)} & I \end{pmatrix}^T$$

$$\begin{pmatrix} \boxed{\begin{array}{c|c} 5.8584 & -6.4248 \\ \hline -6.4248 & 4.7257 \end{array}} \end{pmatrix} = \begin{pmatrix} \boxed{\begin{array}{c|c} 1 & \\ \hline -1.0967 & 1 \end{array}} \end{pmatrix} \begin{pmatrix} \boxed{\begin{array}{c|c} 5.8584 & \\ \hline & -2.3202 \end{array}} \end{pmatrix} \begin{pmatrix} \boxed{\begin{array}{c|c} 1 & -1.0967 \\ \hline & 1 \end{array}} \end{pmatrix}$$

Or write in original matrix form

$$\tilde{A}^{(3)} = P_3 A^{(3)} P_3^T = \begin{pmatrix} \boxed{\begin{array}{cc|cc} -8 & -13 & & \\ -13 & -7 & & \\ \hline & & 5.8584 & -6.4248 \\ & & -6.4248 & 4.7257 \end{array}} \end{pmatrix} = \begin{pmatrix} \boxed{\begin{array}{cc|cc} 1 & & & \\ & 1 & & \\ \hline & & 1 & \\ & & -1.0967 & 1 \end{array}} \end{pmatrix} \begin{pmatrix} \boxed{\begin{array}{cc|cc} -8 & -13 & & \\ -13 & -7 & & \\ \hline & & 5.8584 & \\ & & & -2.3202 \end{array}} \end{pmatrix} \begin{pmatrix} (L^{(3)})^T \end{pmatrix}$$

$$P_3 P_1 A^{(1)} P_1^T P_3^T = P_3 \tilde{A}^{(1)} P_3^T = \tilde{L}^{(1)} \tilde{A}^{(3)} (\tilde{L}^{(1)})^T = \tilde{L}^{(1)} L^{(3)} A^{(4)} (L^{(3)})^T (\tilde{L}^{(1)})^T$$

Example (complete pivoting) [5]

$$P_3 P_1 = (1, 2, 4, 3) \cdot (2, 3, 1, 4) = (2, 3, 4, 1)$$

$$\tilde{L}^{(1)} L^{(3)} = \begin{pmatrix} 1 & & & \\ & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix}$$

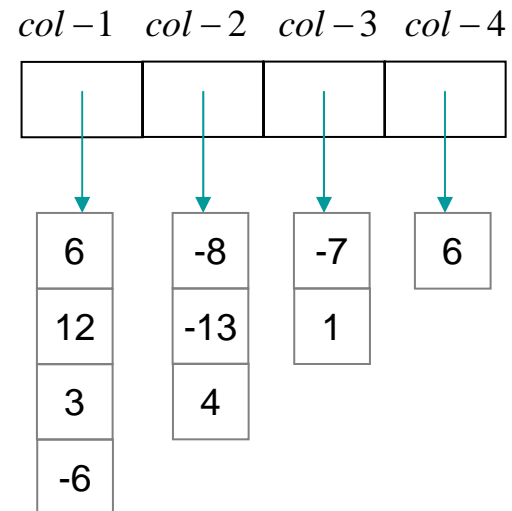
$$A^{(4)} = \begin{pmatrix} -8 & -13 & & \\ -13 & -7 & & \\ & & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

In practice, we have two key issues

- 1 we only store lower part of matrix A

$$A = \begin{pmatrix} 6 & & & \\ 12 & -8 & & \\ 3 & -13 & -7 & \\ -6 & 4 & 1 & 6 \end{pmatrix}$$

col-major based



Example (complete pivoting) [6]

2 we store decomposition $\tilde{L}^{(1)}L^{(3)}$ and $A^{(4)}$ into original A

$$\tilde{L}^{(1)}L^{(3)} = \begin{pmatrix} 1 & & & \\ & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix} + A^{(4)} = \begin{pmatrix} -8 & & & \\ -13 & -7 & & \\ & & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

$$\text{mem}(\tilde{L}^{(1)}L^{(3)}) + \text{mem}(A^{(4)}) = \begin{pmatrix} -8 & & & \\ -13 & -7 & & \\ 0.1327 & -0.3984 & 5.8584 & \\ 0.3982 & -1.1681 & -1.0967 & -2.3202 \end{pmatrix}$$

Question: How can you judge correct decomposition $PAP^T = LDL^T$

from $\text{mem}(\tilde{L}^{(1)}L^{(3)}) + \text{mem}(A^{(4)})$

Example (complete pivoting) [7]

Case 1: four 1x1 pivot

$$L = \begin{pmatrix} 1 & & & \\ 13 & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix}$$

$$D = \begin{pmatrix} -8 & & & \\ & -7 & & \\ & & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

Case 2: two 2x2 pivot

$$L = \begin{pmatrix} 1 & & & \\ & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & & 1 \end{pmatrix}$$

$$D = \begin{pmatrix} -8 & & & \\ -13 & -7 & & \\ & & 5.8584 & \\ & & -1.0967 & -2.3202 \end{pmatrix}$$

Example (complete pivoting) [8]

Case 3: 1x1 pivot + 2x2 pivot + 1x1 pivot

$$L = \begin{pmatrix} 1 & & & \\ -13 & 1 & & \\ 0.1327 & & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix}$$

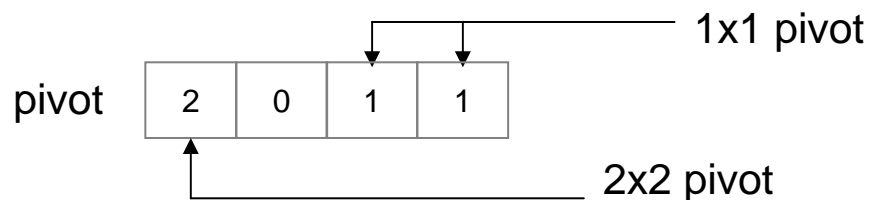
$$D = \begin{pmatrix} -8 & & & \\ & -7 & & \\ & -0.3984 & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

Case 4: 2x2 pivot + 1x1 pivot + 1x1 pivot

$$L = \begin{pmatrix} 1 & & & \\ & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix}$$

$$D = \begin{pmatrix} -8 & & & \\ -13 & -7 & & \\ & & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

Solution: we need an array to record pivot sequence



OutLine

- Preliminary
- Symmetric permutation
- LDL' decomposition (diagonal pivoting)
- Example of complete diagonal pivoting
- **Algorithm of complete diagonal pivoting**

Algorithm ($PAP' = LDL'$) [1]

Given symmetric indefinite matrix $A \in R^{n \times n}$, construct initial lower triangle matrix $L = I$

use permutation vector P to record permutation matrix $P^{(k)}$

let $A^{(1)} := A$, $L^{(0)} = I$, $P^{(0)} = (1, 2, 3, \dots, n)$ and $pivot = zero(n)$, $\alpha = \frac{1 + \sqrt{17}}{8} \approx 0.6404$

$k = 1$

while $k \leq (n-1)$

we have compute $P^{(k-1)} A (P^{(k-1)})^T = L^{(k-1)} A^{(k)} (L^{(k-1)})^T$

$$A^{(k)} = \left(\begin{array}{ccc|ccc} D_1 & 0 & \dots & 0 & \dots & 0 \\ 0 & \ddots & & & & \\ \vdots & & D_s & \dots & \dots & 0 \\ \hline \vdots & & 0 & a_{k,k}^{(k)} & \dots & a_{n,k}^{(k)} \\ \vdots & & \vdots & \vdots & & \vdots \\ 0 & \dots & 0 & a_{n,k}^{(k)} & \dots & a_{nn}^{(k)} \end{array} \right)$$

update original matrix A , where $D_i : 1 \times 1$ or 2×2

$$L^{(k-1)} = \left(\begin{array}{c|c} \overbrace{\left(\begin{array}{c|c} W & O \\ \hline M & I \end{array} \right)}^{k-1} \\ \hline \end{array} \right) \} k-1$$

combines all lower triangle matrix and store in L

Algorithm ($PAP' = LDL'$) [2]

1 compute $\mu_0 = \max_{k \leq i, j \leq n} |A_{ij}| = |A_{\eta\eta}|$ and $\mu_1 = \max_{k \leq i \leq n} |A_{ii}| = A_{\eta\eta}$

Case 1: $\mu_1 \geq \alpha\mu_0$

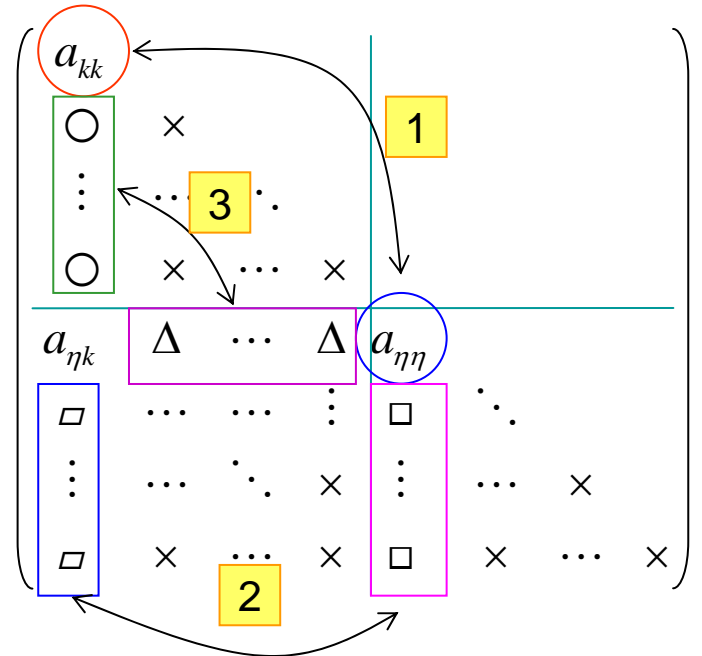
define permutation $P_k = (1, 2, 3, \dots, k-1, \eta, k+1, \dots, \eta-1, k, \eta+1, \dots, n)$ to do symmetric permutation

2 $P(k) \leftrightarrow P(\eta)$

To compute $\tilde{A}^{(k)} = P_k A^{(k)} P_k^T$, we only update lower triangle of $A^{(k)}$

- 3
- 1 $A(k, k) \leftrightarrow A(\eta, \eta)$
 - 2 $A(\eta+1:n, k) \leftrightarrow A(\eta+1:n, \eta)$
 - 3 $A(k+1:\eta-1, k) \leftrightarrow A(\eta, k+1:\eta-1)$

then $\tilde{A}^{(k)} = \left(\begin{array}{c|cc} D_{k-1} & & \\ \hline & a_{kk}^{(k)} & c^T \\ & c & B \end{array} \right), a_{kk}^{(k)} := a_{\eta\eta}$



Algorithm ($PAP' = LDL'$) [3]

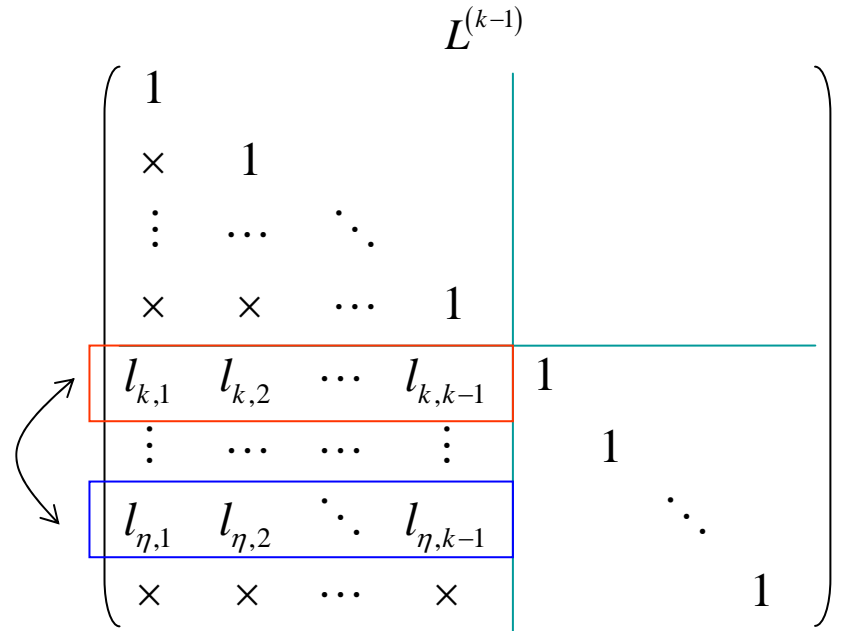
To compute $\tilde{L}^{(k-1)} = P_k L^{(k-1)} P_k^T$

4 We only update lower triangle matrix L

$$L(k, 1:k-1) \leftrightarrow L(\eta, 1:k-1)$$

then

$$P^{(k)} A (P^{(k)})^T = \tilde{L}^{(k-1)} \tilde{A}^{(k)} (\tilde{L}^{(k-1)})^T$$



5 do 1x1 pivot : $\tilde{A}^{(k)} = \begin{pmatrix} D_{k-1} & & & \\ & a_{kk}^{(k)} & c^T & \\ & c & B & \end{pmatrix} = \begin{pmatrix} I & & & \\ & 1 & & \\ & c/a_{kk}^{(k)} & & I \end{pmatrix} \begin{pmatrix} D_{k-1} & & & \\ & a_{kk}^{(k)} & & \\ & & B - cc^T/a_{kk}^{(k)} & \end{pmatrix} (L^{(k)})^T$

$$\begin{cases} L(k+1:n, k) \leftarrow c / a_{kk}^{(k)} \\ A(k+1:n, k+1:n) \leftarrow B - cc^T / a_{kk}^{(k)} \end{cases} \quad \text{then} \quad P^{(k)} A (P^{(k)})^T = \tilde{L}^{(k-1)} L^{(k)} A^{(k+1)} L^{(k)} (\tilde{L}^{(k-1)})^T$$

6 $k \leftarrow k+1$ and $\text{pivot}(k) = 1$

Algorithm ($PAP' = LDL'$) [4]

Case 2: $\mu_1 < \alpha\mu_0$

define permutation $P_k = (1:k, r, k+2, \dots, r-1, k+1, r+1, \dots, n) \cdot (1:k-1, q, k+1, 3, \dots, q-1, k, q+1, \dots, n)$

to interchange q -th and k -th rows and columns, and then r -th and $(k+1)$ -th rows and columns

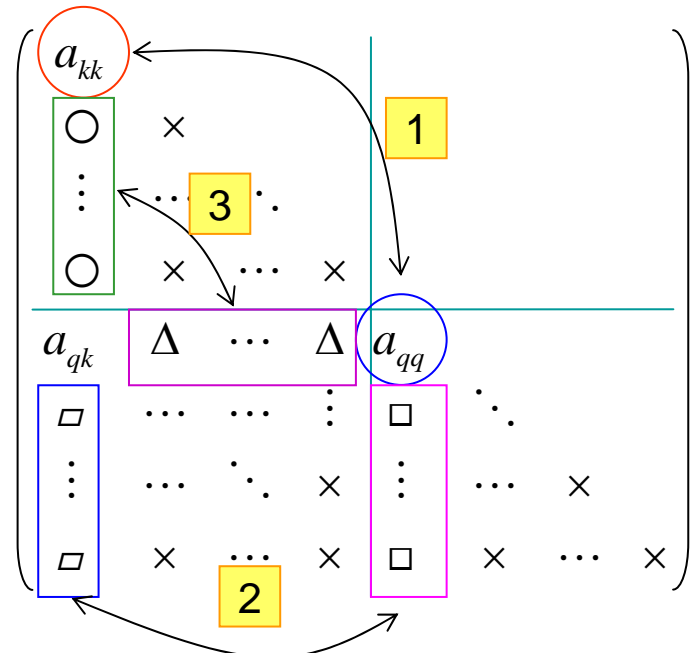
7 $P(k) \leftrightarrow P(q)$ and then $P(k+1) \leftrightarrow P(r)$

To compute $\tilde{A}^{(k)} = P_k A^{(k)} P_k^T$, we only update lower triangle of $A^{(k)}$

(1) do interchange row/col $k \leftrightarrow q$

- 8
- 1 $A(k, k) \leftrightarrow A(q, q)$
 - 2 $A(q+1:n, k) \leftrightarrow A(q+1:n, q)$
 - 3 $A(k+1:q-1, k) \leftrightarrow A(q, k+1:q-1)$

then $A^{(k)} \rightarrow \left(\begin{array}{c|cc} D_k & & \\ \hline & a_{qq}^{(k)} & c^T \\ & c & B \end{array} \right)$



Algorithm (PAP' = LDL') [5]

9 (2) do interchange row/col $k+1 \leftrightarrow r$

1 $A(k+1, k+1) \leftrightarrow A(r, r)$

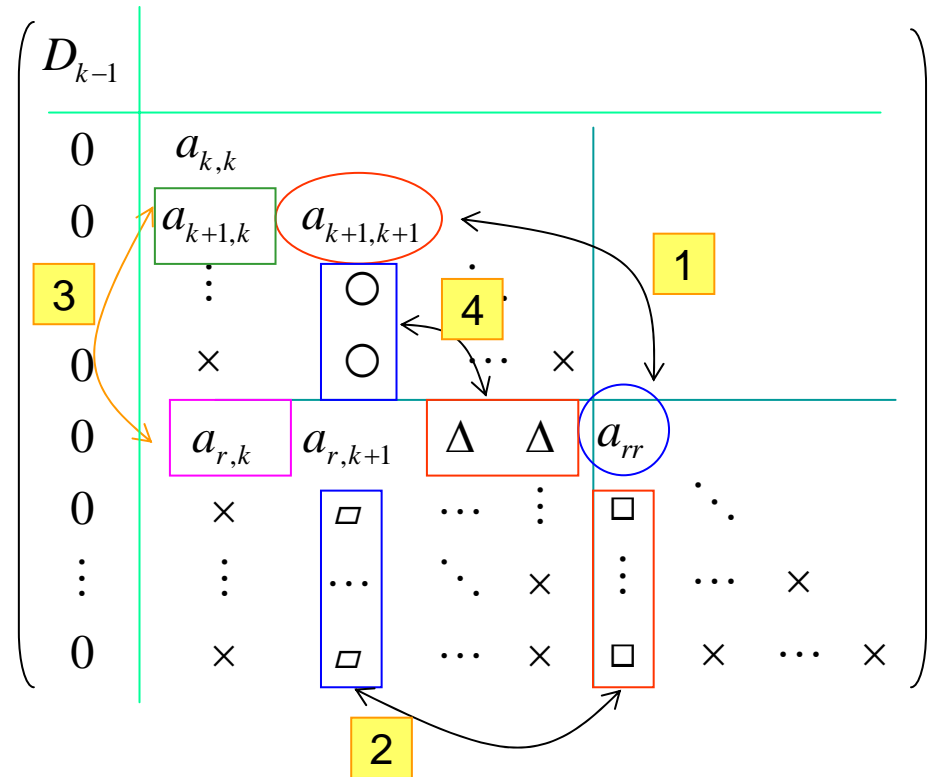
2 $A(r+1:n, k+1) \leftrightarrow A(r+1:n, r)$

3 $A(k+1, k) \leftrightarrow A(r, k)$

4 $A(k+2:r-1: k+1) \leftrightarrow A(r, k+2:r-1)$

then $\tilde{A}^{(k)} = \left(\begin{array}{c|cc} D_{k-1} & & \\ \hline & E & c^T \\ & c & B \end{array} \right)$

where $E = \begin{pmatrix} a_{qq}^{(k)} & a_{rq}^{(k)} \\ a_{rq}^{(k)} & a_{rr}^{(k)} \end{pmatrix}$



Algorithm ($PAP' = LDL'$) [7]

12 do 2x2 pivot : $\tilde{A}^{(k)} = \left(\begin{array}{c|cc} D_{k-1} & & \\ \hline & E & c^T \\ \hline & c & B \end{array} \right) = \left(\begin{array}{c|cc} I & & \\ \hline & I & \\ \hline & cE^{-1} & I \end{array} \right) \left(\begin{array}{c|cc} D_{k-1} & & \\ \hline & E & \\ \hline & & B - cE^{-1}c^T \end{array} \right) (L^{(k)})^T$

$\begin{cases} L(k+2:n, k:k+1) \leftarrow cE^{-1} \\ A(k+2:n, k+2:n) \leftarrow cE^{-1}c^T \end{cases}$ then $P^{(k)} A (P^{(k)})^T = \tilde{L}^{(k-1)} L^{(k)} \boxed{A^{(k+2)}} L^{(k)} (\tilde{L}^{(k-1)})^T$

13 $k \leftarrow k + 2$ and $\text{pivot}(k) = 2$

$A^{(k+2)} = \left(\begin{array}{c|cc} D_{k-1} & & \\ \hline & D_k & \\ \hline & & a_{k+2,k+2}^{(k+2)} \quad \times \\ & & \times \quad \times \end{array} \right)$

$\text{pivot}(k:k+1) = [2, 0]$ means $A(k:k+1, k:k+1) = D_k = \begin{pmatrix} d_{k,k} & d_{k+1,k} \\ d_{k+1,k} & d_{k+1,k+1} \end{pmatrix}$ is 2x2 pivot

endwhile

Question: recursion implementation

normal {

- Initialization
 - check algorithm holds for $k=1$
- Recursion formula
 - check algorithm holds for k or $k+1$, if $k-1$ is true
- Termination condition
 - check algorithm holds for $k=n-1$

abnormal {

- Breakdown of algorithm
 - check which condition $PAP'=LDL'$ fails

Extension of algorithm {

- No extension: algorithm works only for square, symmetric indefinite matrix.

MATLAB implementation [1]

Given a (full) symmetric indefinite matrix A , compute factorization $PAP^T = LDL^T$

return four quantities $P, D, L, pivot$

Remark: try to neglect upper triangle part of A in MATLAB implementation

Example :

$$A = \begin{pmatrix} 6 & 12 & 3 & -6 \\ 12 & -8 & -13 & 4 \\ 3 & -13 & -7 & 1 \\ -6 & 4 & 1 & 6 \end{pmatrix}$$

$$\alpha = \frac{1 + \sqrt{17}}{8}$$

return :

$$P = \begin{pmatrix} 2 & 3 & 4 & 1 \end{pmatrix}$$

$$pivot = \begin{pmatrix} 2 & 0 & 1 & 1 \end{pmatrix}$$

$$L = \begin{pmatrix} 1 & & & \\ & 1 & & \\ 0.1327 & -0.3984 & 1 & \\ 0.3982 & -1.1681 & -1.0967 & 1 \end{pmatrix}$$

$$D = \begin{pmatrix} -8 & & & \\ -13 & -7 & & \\ & & 5.8584 & \\ & & & -2.3202 \end{pmatrix}$$

MATLAB implementation [2]

When factorization is complete, $PAP^T = LDL^T$, you need to provide linear solver

$$Ax = b \longrightarrow (PAP^T)(Px) = Pb \longrightarrow LDL^T(Px) = Pb$$

1 define $z = DL^T(Px)$, then $Lz = Pb \longrightarrow z = L^{-1}(Pb)$ by forward substitution

2 define $y = L^T(Px)$, then $Dy = z \longrightarrow y = D^{-1}z$ by diagonal block inversion

$$\text{Example: } D = \begin{pmatrix} a_1 & & & & \\ & a_2 & a_3 & & \\ & a_3 & a_4 & & \\ & & & & \\ & & & & a_5 \end{pmatrix}, \text{ then } D^{-1} = \begin{pmatrix} 1/a_1 & & & & \\ & \frac{1}{a_2 a_4 - a_3^2} \begin{pmatrix} a_4 & -a_3 \\ -a_3 & a_2 \end{pmatrix} & & & \\ & & & & \\ & & & & \\ & & & & 1/a_5 \end{pmatrix}$$

3 define $w = Px$, then $L^T w = y \longrightarrow w = (L^T)^{-1} y$ by backward substitution

However you cannot transpose L explicitly in MATLAB

4 $x = P^T w$, you must scan w once to determine x