

# A Remark on the Inverse of Principal Matrices by Implicit LU Factorization\*

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**Abstract** We prove the inverting formula of the principal submatrix on a symmetric strongly nonsingular matrix by the implicit LU factorization algorithm.

**Keywords** inverting matrix, principal submatrix, symmetric strongly nonsingular matrix, implicit LU factorization algorithm.

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The following theorem is stated in [1] concerning the inverse of the principal submatrix of a given matrix  $A$  but with an incorrect proof.

**Theorem**<sup>[1]</sup> Let  $p_1, p_2, \dots, p_k$  be the first  $k$  ( $k \leq n$ ) search vectors generated by the implicit LU factorization algorithm on the  $n$  by  $n$  symmetric strongly nonsingular matrix  $A$ . Then

$$\sum_{i=1}^k \frac{p_i p_i^T}{a_i^T p_i} = \begin{bmatrix} (A^{k,k})^{-1} & 0 \\ 0 & 0 \end{bmatrix}, \quad (1)$$

where  $A^{k,k}$  is the  $k$ th principal submatrix of  $A$ .

In this note we prove this theorem.

**Proof** Denote  $A = (a_{ij}) = (a_1, a_2, \dots, a_n)^T$ ,  $A^i = (a_1, a_2, \dots, a_i)$ . We proceed by induction. For  $k=1$ , from the implicit LU factorization algorithm<sup>[1]</sup>,  $p_1 = e_1 \in R^n$ . It follows that

$$\frac{p_1 p_1^T}{a_1^T p_1} = \begin{bmatrix} (a_{11})^{-1} & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} (A^{1,1})^{-1} & 0 \\ 0 & 0 \end{bmatrix},$$

the theorem is true for  $k=1$ .

For  $k=2$ ,  $H_2 = I - a_1 p_1^T / a_1^T p_1 = (h_{21}, e_2, e_3, \dots, e_n)$ , where  $h_{21} = (0, -a_{12}/a_{11}, -a_{13}/a_{11}, \dots, -a_{1n}/a_{11})^T$ , and  $p_2 = H_2^T e_2 = (-a_{21}/a_{11}, 1, 0, \dots, 0)^T$ . Thus  $\sum_{i=1}^2 p_i p_i^T / a_i^T p_i$

is of the following form  $\begin{bmatrix} B_2 & 0 \\ 0 & 0 \end{bmatrix}$ , where  $B_2 \in R^{2 \times 2}$ . It is enough to prove that

$$\sum_{i=1}^k \frac{p_i p_i^T}{a_i^T p_i} \begin{bmatrix} A^{k,k} & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} I_k & 0 \\ 0 & 0 \end{bmatrix} \quad (2)$$

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holds for  $k = 2$ . Since  $\frac{p_1 p_1^T}{a_1^T p_1} \begin{pmatrix} A^{2,2} & 0 \\ 0 & 0 \end{pmatrix} = (e_1, (a_{12}/a_{11})e_1, 0, \dots, 0)$  and  $p_2^T a_1 = e_2^T H_2 a_1 = 0$ , then

$$\frac{p_2 p_2^T}{a_2^T p_2} \begin{pmatrix} A^{2,2} & 0 \\ 0 & 0 \end{pmatrix} = p_2 / a_2^T p_2 (p_2^T A^2, 0) = (0, (-a_{12}/a_{11})e_1 + e_2, 0, \dots, 0). \quad (3)$$

Hence when  $k = 2$ , (2) is true

Assume that the theorem is true for  $i \leq k$ , we prove that it will still be valid for  $i = k + 1$ . Denote  $A^{k+1, k+1} = \begin{pmatrix} A^{k,k} & \bar{a}_{k+1} \\ \bar{a}_{k+1}^T & a_{k+1, k+1} \end{pmatrix}$ , and from assuming the validity for  $i \leq k$ , then

$$\sum_{i=1}^{k+1} \frac{p_i p_i^T}{a_i^T p_i} \begin{pmatrix} A^{k+1, k+1} & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} I_k & 0 \\ 0 & 0 \end{pmatrix} + \sum_{i=1}^k \frac{p_i p_i^T}{a_i^T p_i} \begin{pmatrix} 0 & \bar{a}_{k+1} & 0 \\ \bar{a}_{k+1}^T & a_{k+1, k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix} + \frac{p_{k+1} p_{k+1}^T}{a_{k+1}^T p_{k+1}} \begin{pmatrix} A^{k+1, k+1} & 0 \\ 0 & 0 \end{pmatrix}$$

$I + II + III$ . Since the last  $(n - i)$  ( $i \leq k$ ) components of  $p_i$  are zero, so  $II = \sum_{i=1}^k \frac{p_i p_i^T}{a_i^T p_i} \begin{pmatrix} 0 & \bar{a}_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} (A^{k,k})^{-1} & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & \bar{a}_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 0 & (A^{k,k})^{-1} \bar{a}_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix}$ , where  $(A^{k,k})^{-1} \bar{a}_{k+1} (R^k)$  lies in the  $(k + 1)$ th column. Recall that [1]

$$H_{k+1} = I - \begin{pmatrix} I_k & 0 \\ \bar{A}_k (A^{k,k})^{-1} & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ -\bar{A}_k (A^{k,k})^{-1} & I_{n-k} \end{pmatrix},$$

where  $\bar{A}_k$  is the matrix comprising the last  $n - k$  rows of  $A^k$ . Since  $p_{k+1} = H_{k+1}^T e_{k+1} = (-\bar{a}_{k+1}^T (A^{k,k})^{-1}, 1, 0, \dots, 0)^T$ , thus  $II = \begin{pmatrix} 0 & -p_{k+1} + e_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix}$ . Applying  $p_{k+1}^T a_j = 0$ ,  $j < k + 1$ , then  $III = (0, p_{k+1}, 0)$ , where the vector  $p_{k+1}$  lies in the  $(k + 1)$ th column. From above, we have

$$\sum_{i=1}^{k+1} \frac{p_i p_i^T}{a_i^T p_i} \begin{pmatrix} (A^{k+1, k+1}) & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} I_k & -p_{k+1} + e_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix} + \begin{pmatrix} 0 & p_{k+1} & 0 \\ 0 & 0 & 0 \end{pmatrix} = \begin{pmatrix} I_{k+1} & 0 \\ 0 & 0 \end{pmatrix}.$$

The proof is complete

**Remark** From (3) it may be seen that the statement 'From (6.99) it follows that  $p_i u_i^T$  is a matrix whose elements are all zero, save that on the intersection of the  $i$ th row with the  $i$ th column, which is equal to  $a_i^T p_i$ ' in [1] is wrong.

## References

- [1] J. A. Baffy and E. Spedicato, *ABS Projection Algorithm: Mathematical Techniques for Linear and Nonlinear Equations*, John Wiley & Sons, New York, 1989.