

FAST PARALLEL ALGORITHMS FOR COMPUTING GENERALIZED INVERSES A^+ AND A_{MN}^+

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Abstract

The parallel arithmetic complexities for computing generalized inverse A^+ , computing the minimum-norm least-squares solution of $Ax=b$, computing order $m+n-r$ determinants and finding the characteristic polynomials of order $m+n-r$ matrices are shown to have the same growth rate. Algorithms are given that compute A^+ and A_{MN}^+ in $O(\log r \cdot \log n + \log m)$ and $O(\log^2 n + \log m)$ steps using a number of processors which is a polynomial in m , n and r ($A \in R_r^{m \times n}$, $r = \text{rank } A$).

§ 1. Introduction

Let $I(n)$, $E(n)$, $D(n)$, $P(n)$ denote the parallel arithmetic complexities of inverting order n matrices, solving a system of n linear equations in n unknowns, computing order n determinants and finding the characteristic polynomials of order n matrices respectively. Then Csanky gave an important theoretical result [1]:

Theorem 1. $I(n) = O(f(n)) \Leftrightarrow E(n) = O(f(n)) \Leftrightarrow D(n) = O(f(n)) \Leftrightarrow P(n) = O(f(n))$.

He also gives algorithms that compute these problems in $O(\log^2 n)$ steps using a number of processors which is a polynomial in n (n is the order of the matrix of the problem).

Let $A \in R_r^{m \times n}$, $r = \text{rank } A$. In this paper, we give two parallel algorithms for computing A^+ and A_{MN}^+ respectively. The one for A^+ is based on Decell's method in [2], and the one for A_{MN}^+ is a generalization of Decell's method in [3].

The parallel arithmetic complexities for computing the generalized inverse A^+ , computing the minimum-norm least-squares solution of $Ax=b$, computing order $m+n-r$ determinants and finding the characteristic polynomials of order $m+n-r$ matrices are shown to have the same growth rate.

§ 2. The Parallel Algorithm for Computing A^+

Let $A \in R_r^{m \times n}$. Then there is a unique matrix $X \in R_r^{n \times m}$ satisfying

$$AXA = A, XAX = X, (AX)^T = AX, (XA)^T = XA.$$

This X is called the M-P inverse of A and is denoted by $X = A^+$.

In [2], Decell gave a finite algorithm for computing A^+ . We rewrite it as follows:

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Algorithm 1. (1) Parallely compute $B = A^T A$.

(2) Parallely compute $B^k = (b_{ij}^{(k)})$, $k = 1, 2, \dots, r$.

(3) Let $\lambda_1, \lambda_2, \dots, \lambda_n$ denote the roots of the characteristic polynomial $f(\lambda)$ of B . Let

$$s_k = \sum_{i=1}^n \lambda_i^k, \quad k = 1, 2, \dots, r.$$

Parallely compute

$$s_k = \text{tr}(B^k) = \sum_{i=1}^n b_{ii}^{(k)}, \quad k = 1, 2, \dots, r.$$

(4) Let the characteristic polynomial $f(\lambda)$ of B be

$$f(\lambda) = \det(\lambda I - B) = \lambda^n + c_1 \lambda^{n-1} + \dots + c_n.$$

From the Newton formula

$$s_k + c_1 s_{k-1} + c_2 s_{k-2} + \dots + c_{k-1} s_1 + k c_k = 0, \quad k \leq n$$

we have

$$\begin{bmatrix} 1 & & & & & & \\ s_1 & 2 & & & & & \\ s_2 & s_1 & 3 & & & & \\ \vdots & \vdots & \ddots & \ddots & \ddots & & \\ s_{r-1} & s_{r-2} & \dots & s_1 & r & & \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_r \end{bmatrix} = - \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_r \end{bmatrix}.$$

Parallely compute the solution of the above triangular system.

(5) Parallely compute

$$A^+ = -((A^T A)^{r-1} + c_1 (A^T A)^{r-2} + \dots + c_{r-1} I) A^T / c_r. \tag{2.1}$$

Theorem 2. Let $A \in R_r^{m \times n}$, and $GI(m, n)$ denote the parallel arithmetic complexity for computing the M-P inverse A^+ . Then

$$GI(m, n) = \log r (\log n + 7/2) + (1/2) \log^2 r + 2 \log n + \log m + 4 = O(f(m, n, r))$$

and the number of processors used in the algorithm is

$$cp = \begin{cases} n^3 r / 2, & m < nr / 2, \\ mn^2, & m \geq nr / 2. \end{cases}$$

Proof. (1) Parallel computation of $B = A^T A$ takes $T_1 = \log m + 1$ steps and $cp_1 = mn^2$ processors.

(2) Parallel computation of B^k ($k = 1, 2, \dots, r$) takes $T_2 = \log r (\log n + 1)$ steps and $cp_2 = n^3 r / 2$ processors.

(3) Parallel computation of s_k ($k = 1, 2, \dots, r$) takes $T_3 = \log n$ steps and $cp_3 = rn / 2$ processors.

(4) Parallel computation of c_k ($k = 1, 2, \dots, r$) takes $T_4 = (1/2) \log^2 r + (3/2) \log r$ steps and $cp_4 = O(r^3)$ processors.

(5) Since B^2, \dots, B^{r-1} are already available, parallel computation of A^+ takes $T_5 = \log r + \log n + 3$ steps and $cp_5 = n^2 m$ processors.

Thus

$$cp = \max_{1 \leq i \leq 5} cp_i = \begin{cases} n^3 r / 2, & m < nr / 2, \\ mn^2, & m \geq nr / 2. \end{cases}$$