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FOUR O(n**2) MULTIPLICATION METHODS FOR SPARSE AND DENSE BOOLEAN MATRICES*

CS-80-25

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* This work has been partially supported by the Natural Sciences and Engineering Research Council.

Abstract

In this paper, we present four algorithms to evaluate the product of sparse and dense Boolean matrices. The proposed algorithms determine the Boolean product of two n*n matrices in O(n**2) operations whenever one of the matrices is either sparse or dense. Finally, the multiplication methods for sparse Boolean matrices are extended to work with general sparse matrices with the same time complexity.

<u>Key words and phrases</u>: Boolean matrix multiplication, sparse matrices, dense matrices, bipartite graphs.

1. INTRODUCTION

With the notable exception of the "four Russians' algorithm" [2], Boolean multiplication methods have been usually extensions of general matrix multiplication techniques. Namely, asymptotically fast Boolean algorithms [1,3] have been modeled on the well known asymptotically fast algorithms for general matrix multiplication [6,8]. Because of the versatility of Boolean matrices and of their capability to model different problems, to find efficient and practical algorithms (i.e. algorithms that are simpler to implement than, and/or outperform the asymptotically fast methods for special classes of matrices) becames a relevant problem.

In this paper, we present four efficient and practical multiplication methods particularly suited for sparse and dense Boolean matrices. In fact, with the proposed methods we can evaluate the product of two n*n Boolean matrices in O(n**2) operations whenever one of the two matrices is sparse or dense. In the following sections, we describe the framework and present the algorithms for the multiplication of sparse and dense Boolean matrices. Finally, we show how to modify the multiplication algorithms for sparse Boolean matrices to work with general sparse matrices with the same time complexity. All the proposed algorithms use a data structure similar to the classical structures emploied by the indexing techniques for sparse and dense matrices.

2. THE FRAMEWORK

There is a natural correspondence between Boolean matrices and binary relations on finite sets or, equivalently, finite directed graphs. Thus, we can represent a $n \times n$ Boolean matrix B by the sets R(B,i), $i=1,\ldots,n$, where $R(B,i)=\{j|B(i,j)=1\}$. The sets R(B,i) are called adjacency lists of B. Analogously we can define the predecessor lists P(B,i), $i=1,\ldots,n$, where $P(B,i)=\{k|B(k,i)=1\}$. Let m(B) be the number of non-zero elements in B; then, matrix B is said to be sparse if m(B)=0 (n) [7].

Let us remark that many indexing techniques for sparse matrices use storage schemas very similar to the adjacency or the predecessor lists, e.g. the indexing with row and column designators, with row vector indicators and column designation, with row vector and column index vector, etc. [7]. We will not further specify the implementation characteristics of the information structure used by the proposed algorithms; in fact, whatever of the above data representations is effectively used, this alters the complexity of the matrix multiplication only by a constant factor [4]. Finally, let us remark that the adjacency and the predecessor lists, if not available, can be easily constructed from B in time O(n**2).

Before presenting the algorithms, let us introduce some additional terminology that will be used in the following sections. As mentioned before, we can uniquely associate to each

Boolean matrix B a directed graph G(B) = (V(B), E(B)), where there is a vertex in V(B) for each row and one for each column in B, and there is an edge from the vertex associated with row i to the vertex associated with column j if and only if B(i,j)=1. Let h denote the isomorphic mapping between the set of rows and columns and the set of vertices. It is easy to see that G(B) is a bipartite graph and that $\{VR(B), VC(B)\}$ is the bipartition [10], where VR(B) and VC(B) are the set of the images of the rows and columns of B, respectively. If B(i,j)=1, then h(j) is said to be reachable from h(i) in G(B). Intuitively, R(B,i) represents the set of the vertices reachable from vertex h(i), and P(B,i) represents the set of vertices from which vertex h(i) is reachable.

In this paper we are interested in the evaluation of the product of two $n \times n$ Boolean matrices B=B1*B2. Without loss of generality, let us assume VC(B1)=VR(B2). Obviously, VR(B)=VR(B1) and VC(B)=VC(B2).

3. ALGORITHMS FOR SPARSE BOOLEAN MATRICES

When evaluating the Boolean product B=B1*B2, if B1 or B2 is sparse, then we can take advantage of this fact in the computation of the product matrix B. In this section, we consider two cases, depending on whether B1 or B2 is sparse, and present a simple and efficient algorithm for each case. Obviously, if

both Bl and B2 are sparse, then any of the proposed algorithms can be used.

Case 1 (B2 is sparse)

The algorithm AS to evaluate the product B=B1*B2 when B2 is sparse is described in the appendix.

In order to discuss the correctness and the complexity of procedure AS, let us observe the following relation:

(1) $B(i,j)=1 \iff j \in R(B,i) \iff k \in R(B1,i) j \in R(B2,k) \iff h(j)$ is reachable in G(B2) from at least one vertex h(k) with $k \in R(B1,i)$

where <==> means "if and only if". It is easy to see that, for each ie[1,n], procedure AS performs the equivalent of a depth first search [9] in G(B2) from all vertices h(k) reachable from h(i); thus, procedure AS satisfies relation (1). Let us now analyze the complexity of the algorithm. For a given ie[1,n], the algorithm requires max{n,e(i)} operations,

where
$$e(i) = \sum_{k \in R(B1,i)} |R(B,k)|$$
.

Since $e(i) \le m(B2)$ and since this search is done for all $i \in [1,n]$, in the worst case we need $n \le m(B2)$ operations. That is,

Property 1 Procedure AS determines the Boolean product of a n×n matrix by a n×n sparse matrix in O(n**2) operations.

Let us remark that a procedure similar to AS has been used to produce a fast expected time algorithm for Boolean matrix multiplication [5].

Case 2 (Bl is sparse):

Using the predecessor lists P(Bl,i) and P(B2,i), we can write a procedure SA, similar to AS, to evaluate B when Bl is sparse; the algorithm is described in the appendix.

<u>Property 2</u> Procedure SA determines the Boolean product of a $n \cdot n$ sparse matrix by a $n \cdot n$ matrix in $O(n^**2)$.

To prove the above property, let us observe that, if vertex h(j) is reachable from vertex h(i) in G(B), then h(i) is reachable from h(j) in G(B), where B is the transpose of B. Therefore, relation (1) can be re-expressed as follows:

(2) $B(i,j)=1 \iff i \in P(B,j) \iff node h(i) \text{ is reachable in } G(B)$ from $h(j) \iff node h(i) \text{ is reachable in } G(B) \text{ from at}$ least a node h(k) with $k \in P(B2,j) \iff k \in P(B2,j) \text{ i} \in P(B1,k)$

For each $j \in [1,n]$, procedure SA performs the equivalent of a T depth first search in G(Bl) starting from all vertices reachable in G(B2) from h(j); thus, it obviously satisfies relation (2). The analysis of the complexity of procedure SA follows the same lines as the analysis of procedure AS.

4. ALGORITHMS FOR DENSE BOOLEAN MATRICES

Given a n_*n Boolean matrix B, its complement \overline{B} is the n_*n Boolean matrix defined by

$$B(i,j)=1 <==> B(i,j)=0$$

We shall say that B is (very) <u>dense</u> if its complement is sparse. Since \overline{B} has $m(\overline{B})=n**2-m(B)$ non-zero entries, where m(B) is the number of non-zero elements in B, then we can say that B is dense if $m(\overline{B})=O(n)$.

Let us consider the product B=B1*B2 when B1 or B2 is dense. In order to reduce the complexity of the multiplication, we must take advantage of both the density of one of the matrices, and the fact that we are dealing with Boolean matrices. Namely we determine the zero entries of the product matrix instead of the non-zero entries, and use the sets $R(\overline{B},i)$ and $P(\overline{B},i)$ instead of R(B,i) and P(B,i). Again, we present two algorithms, depending on B1 or B2 being a dense matrix. Obviously, if both B1 and B2 are dense, then any of the proposed algorithms can be used. The algorithms (procedures AD and DA)

are described in the appendix.

Property 3 Procedure AD determines the Boolean product of a n*n matrix by a dense n*n matrix in O(n**2) operations.

Let us first prove that procedure AD computes exactly the product B. To do this, it is sufficient to observe that relation (1) is equivalent to

(3)
$$B(i,j)=0 \iff j \notin R(B,i) \iff (\exists k \in R(B1,i) j \in R(B2,k))$$

 $\iff \forall k \in R(B1,i) j \notin (B2,k) \iff \forall k \in R(B1,i) j \in R(\overline{B2},k)$

That is, we must set B(i,j) to zero if and only if h(j) is not reachable in G(B2) from any of the vertices reachable from h(i) in G(B1). It is easy to see that h(j) is not reachable from any h(k) \in V(B1) if and only if \max k(j)=|R(B1,i)|. To prove the order complexity, let us observe that given i \in [1,n], we search the graph G($\overline{B2}$) starting from all the vertices reachable from h(i) in G(B1). Since the whole search will not take more than m(B2) steps and since we repeat this process for i=1,...,n, then the entire procedure will require at most $n \cdot m(\overline{B2})$ steps. Since B2 is dense, then the above property holds.

Let us now consider the case of Bl being dense.

Property 4 Procedure DA determines the Boolean product of a
 dense Boolean matrix by a Boolean matrix in O(n**2)

The proof follows the lines of the proof of Property 3, and by observing that relation (2) can be re-written as

(4)
$$B(i,j)=0 <==> i \notin P(B,j) <==> j(\exists k \in P(B2,j) i \in P(B1,k))$$

 $<==> \forall k \in P(B2,j) i \notin P(B1,k) <==> \forall k \in P(B2,j) i \in P(B1,k)$

5. EXTENSION TO NON-BOOLEAN MATRICES

The algorithms for the multiplication of sparse Boolean matrices, AS and SA, can be easily modified to work with general sparse matrices with the same time complexity. Unfortunately, this property does not hold for the dense Boolean matrix multiplication methods AD and DA. We will now describe how to extend algorithms AS and SA to work with general matrices.

Let us consider the product M=M1*M2, where M1 and M2 are $n \times n$ matrices and at least one of them is sparse; i.e. it has O(n) non-zero elements. We can easily extend to non-Boolean matrices the terminology introduced in the previous sections. The adjacency list R(M,i) is the set of couples (index,value) defined as follows: $R(M,i) = \{(index,value) \mid M(i,index) = value \neq 0\}$. That is, $p = (j,x) \in R(M,i)$ iff $M(i,j) = x \neq 0$; in this case we will

use the notation <u>value</u> [p] = x and <u>index</u> [p] = j. Analogously, $P(M,j) = \{ (index, value) | M(index, j) = value/0 \}$.

still represent M in graph form: We can G(M) = (V(M), L(M), E(M)) is the directed edge-labelled bipartite graph where V(M)=VR VC is the set of the isomorphic images of the rows and columns of M, L(M) is the set of labels, and E(M) VR VC L(M) is such that there is an edge e=(h(i),h(j),x)from vertex h(i) to vertex h(j) labelled x if and only if M(i,j)=x. The generalization of AS and SA to non Boolean matrices is now straightforward. As usual, we present two algorithms depending on whether Ml or M2 is sparse; the algorithms (procedures GAS and GSA) are described in the appendix.

Let us analyze procedure GAS. When evaluating the entry

(5)
$$M(i,j) = \sum_{k \in [1,n]} Ml(i,k) * M2(k,j).$$

we do not need to consider in the summation the contribution for those k such that M1(i,k)=0 or M2(k,j)=0. Let $Nij=\{k\in [1,n] \mid M1(i,k)\neq 0 \text{ and } M2(k,j)\neq 0\}$; then, relation (5) is equivalent to

(6)
$$M(i,j) = \sum_{k \in Nij} Ml(i,k) * M2(k,j).$$

The set Nij represents exactly those vertices in VC(M1) that are reachable from h(i) in G(M1) and from which h(j) is reachable in G(M2) (recall VC(M1)=VR(M2)). It is easy to see that

$$\sum_{j \in [1,n]} |Nij| \leq m(M2).$$

That is, for a given $i\in[1,n]$, we compute at most m(M2) multiplications and m(M2) additions. Since this is repeated for $i=1,\ldots,n$, and since M2 is sparse (i.e. m(M2)=O(n)), then O(n**2) operations are needed. The analysis of procedure GSA follows in a similar way. This leads to the following

Property 5 The product of two $n \times n$ matrices can be evaluated in O(n**2) whenever one of the two matrices is sparse.

6. CONCLUSIONS

In this paper we have presented four efficient multiplication algorithms for sparse and dense Boolean matrices. The algorithms work in time $O(n^{**}2)$ and use a data structure similar to many classical structures for sparse and dense matrices (in case of dense matrices, we obviously store the <u>zero</u> entries). Extending some of the above algorithms, two $O(n^{**}2)$ multiplication methods for non-Boolean sparse matrices have been presented. Finally, all the proposed algorithms will yield O(n f(n)) time bounds if sparse is defined to mean O(f(n)) non-zero entries.

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APPENDIX

```
/* B2 is dense */
procedure AD
begin
integer mark(n);
for i=1 until n do
      begin
      for j=1 until n do
            begin
            \overline{mark(j)} := 0;
            B(i,j):=1;
            end;
      for all keR(Bl,i) do
            for all j \in R(\overline{B2}, k) do mark(j):=+ 1;
      for j=1 until n do
            if \overline{mark(j)} = \overline{|R(Bl,i)|} then B(i,j) := 0;
      end;
end
```

```
procedure GAS /* M2 is sparse */
begin
for i=l until n do
     begin
     for j=1 until n do M(i,j):=0;
     for all per(Ml,i) do
           \frac{\text{for all q}\in R(M2, index [p])}{M(i, index [q]) := M(i, index [q]) + value [p]} *
                value [q];
     end;
end;
procedure GSA /* Ml is sparse */
begin
for j=l until n do
     begin
     for i=1 until n do M(i,j):=0;
     for all pep(M2,j) do
           for all qCP(MI, index [p]) do
                M(index [q],j):=M(index [q],j)+value
                                                              [p] *
                value [q];
     end;
end;
```