# Context-Free Path Querying by Using Kronecker Product\*

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**Abstract.** Abstact is very abstract. Abstact is very abstract.

**Keywords:** Path querying  $\cdot$  Graph database  $\cdot$  Context-free grammars  $\cdot$  CFPQ  $\cdot$  Kronecker product  $\cdot$  !!! .

### 1 Introduction

CFPQ is popular.

Matrices [?] — algorithm is fast, but grammar size is problem. Moreover, bad for regualr queryes.

<sup>\*</sup> Supported by organization x.

Following contribution.

- 1. !!!
- 2. !!!
- 3. !!!

## 2 Recursive State Machines

Or recursive networks [?] or resursive finite automata [?] or ...

### 3 Kronecker Product

For graphs, for matrices, for FA intersection.

## 4 Kronecker Product Based CFPQ Algorithm

The idea of the algorithm is based on generalisation of the finite-state machine intersection for a recursive automata, created from input grammar, and an input graph. The result of the intersection is evaluated as a tensor product of the corresponding adjacency matrices for automata and graph. To solve reachability problem it is enough to represent intersection result as a Boolean matrix, what simplifies algorithm implementation and allows to express it in terms of basic matrix operations. Listing 1. shows main steps of the solution.

As an input algorithm accepts context-free grammar  $G = (\Sigma, N, P)$  and graph  $\mathcal{G} = (V, E, L)$ . Recursive automata R is created from G. The process of the creation is out of the scope of this article.  $M_1$  and  $M_2$  are the adjacency matrices for automata R and graph  $\mathcal{G}$  correspondingly. Cell values of this matrices could be represented as sets of elements from  $L \cup N \cup \Sigma$ .

As an result the algorithm returns updated matrix  $M_2$  which contains initial graph  $\mathcal{G}$  data and non-terminals from N. If a cell  $M_2[i,j]$  for any valid indices i and j contains symbol  $S \in N$ , therefore, vertex j is reachable from vertex i in grammar G for non-terminal S.

#### Listing 1 Kronecker product based CFPQ

```
1: function ContextFreePathQuerying(G, \mathcal{G})
 2:
         R \leftarrow \text{Recursive automata for } G
 3:
         M_1 \leftarrow \text{Adjacency matrix for } R
         M_2 \leftarrow \text{Adjacency matrix for } \mathcal{G}
 4:
 5:
         while Matrix M_2 is changing do
 6:
             M_3 \leftarrow M_1 \otimes M_2
                                                                           7:
             tC_3 \leftarrow transitiveClosure(M_3)
 8:
             n \leftarrow \text{Matrix } M_3 \text{ dimension}
                                                                           \triangleright Matrix M_3 size = n \times n
9:
             for i \in 0..n - 1 do
10:
                 for j \in 0..n - 1 do
                      if tC_3[i,j] then
11:
12:
                           s \leftarrow \text{initial vertex of the edge } tC_3[i,j]
13:
                           f \leftarrow \text{final vertex of the edge } tC_3[i,j]
                          if hasPathForNonterminals(R, s, f) then
14:
15:
                               x, y \leftarrow getCoordinates(tC_3, i, j)
                               M_2[x,y] \leftarrow M_2[x,y] \cup getNonterminals(R,s,f)
16:
17:
         return M_2
```

#### 4.1 Remarks

- Mentioned above algorithm description does not take into account the use of ε-transitions in the automata R. This transitions might appear in the automata if the grammar allows to derive ε-word for some non-terminal. In this case there is required additional step for matrix  $M_2$  before the while loop is entered. For each  $i ∈ 0..dim(M_2) 1$  symbol ε must be explicitly added for  $M_2[i,i]$  as follows:  $M_2[i,i] ← M_2[i,i] ∪ {ε}$ . Here the rule is implied: each vertex of the graph  $\mathcal G$  is reachable by itself through ε-transition.
- The performance-critical part of the algorithm is transitive closure computation. Generally this step requires  $O(n^3)$  operations and  $O(n^2)$  memory where n is dimension of  $M_3$  what equals  $dim(M_1) \times dim(M_2)$ .

#### 4.2 Example

This section is intended to provide step-by-step demonstration of the proposed algorithm. As an example query consider the following context-free grammar  $G = (\Sigma, N, P)$  for a language  $\{a^n b^n | n \ge 1\}$  where:

```
 \begin{array}{l} - \text{ Set of terminals } \varSigma = \{a,b\}. \\ - \text{ Set of non-terminals } V = \{S\}. \\ - \text{ Set of production rules } P = \{S \rightarrow aSb, S \rightarrow ab\}. \end{array}
```

Since the proposed algorithm processes grammar in form of recursive automata, we first provide automata R in Figure 1. The initial state of the automata is (0), the final state is (3). The notation  $\{S\}$  denotes here that non-terminal S could be derived in automata path from vertex (0) to (3).

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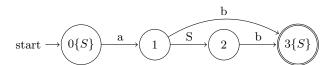


Fig. 1: The recursive automata R of grammar G for example query

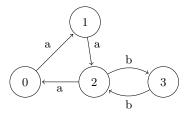


Fig. 2: The input graph  $\mathcal{G}$  for example query

For this example we run query on graph  $\mathcal{G}$  presented in Figure 2. This graph consists of 4 vertices and 5 edges with labels.

Adjacency matrices for automata R and graph  $\mathcal{G}$  are initialised as depicted in Figure 3.

$$M_1 = \begin{pmatrix} \dots \{a\} & \dots \\ \dots \{S\} \{b\} \\ \dots & \{b\} \\ \dots & \dots \end{pmatrix} M_2^0 = \begin{pmatrix} \dots \{a\} & \dots \\ \dots & \dots \{a\} & \dots \\ \{a\} & \dots & \{b\} \\ \dots & \dots & \{b\} & \dots \end{pmatrix}$$

Fig. 3: Adjacency matrices  $M_1$  for R and  $M_2$  for G

Because automata R does not have  $\epsilon$ -transitions and  $\epsilon$ -word is not included in grammar G language, we can skip additional step for matrix  $M_2$  mentioned in section 4.1.

After all the data is initialised in lines 2-4 of the algorithm, it enters while loop and iterates as long as matrix  $M_2$  is changing. We provide step-by-step evaluation of matrices  $M_3$ ,  $tC_3$  and updating of matrix  $M_2$ . All the matrices are denoted with upper index of the current loop iteration. The first loop iteration is indexed as 1.

For the first while loop iteration the tensor product  $M_3^1 = M_1 \otimes M_2^0$  and transitive closure  $tC_3^1$  are evaluated as shown in Figure 4. The dimension n of the matrix  $M_3$  equals 16, and this value is constant in time of the algorithm execution.

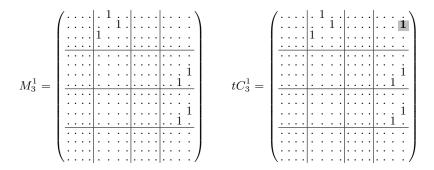


Fig. 4: The first iteration tensor product and transitive closure evaluation for example query

After the transitive closure evaluation matrix  $tC_3^1$  cell (1,15) contains non-zero value. It means that vertex with index 15 is accessible from vertex with index 1 in a graph, represented by adjacency matrix  $M_3^1$ .

Then the algorithm lines 9-16 are executed. In that section algorithm adds non-terminals to the graph matrix  $M_2^1$ . Because this step is additive we are only interested in newly appeared values in matrix  $tC_3^1$  such as value  $tC_3^1[1,15]$ .

For the value  $tC_3^1[1, 15]$ :

- Indices of the automata vertices s=0 and f=3, because value  $tC_3^1[1,15]$  located in upper right matrix block (0,3).
- Indices of the graph vertices x = 1 and y = 3 are evaluated as value  $tC_3^1[1, 15]$  indices relatively to its block (0, 3).
- Function call hasPathForNonterminals() returns **true** since the automata R has path for non-terminal S from vertex 0 to 3.
- Function call getNonterminals() returns  $\{S\}$  since this is the only non-terminal which could be derived in path from vertex 0 to 3.

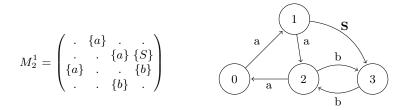


Fig. 5: The updated matrix  $M_2^1$  and graph  $\mathcal{G}$  after first loop iteration for example query

After the first loop iteration matrix symbol S is added to the cell  $M_2^1[1,3]$ . It is relevant data, because initial graph has path  $1 \to 2 \to 3$  which could be derived for S. The updated matrix and graph are depicted in figure 5.

For the second loop iteration matrices  $M_3^2$  and  $tC_3^2$  are evaluated as listed in figure 6. For this iteration in the matrix  $tC_3^2$  appeared new non-zero values in cells with indices [0,11], [0,14] and [5,14]. Because only the cell value with index [0,14] corresponds to the automata path with not empty non-terminal set  $\{S\}$  its data affects adjacency matrix  $M_2$ . The update matrix and graph  $\mathcal{G}$  are depicted in figure 7.

Fig. 6: The second iteration tensor product and transitive closure evaluation for example query

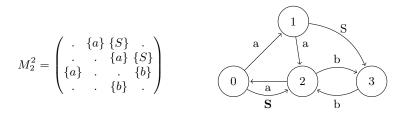


Fig. 7: The updated matrix  $M_2^2$  and graph  $\mathcal{G}$  after second loop iteration for example query

The remaining matrices  $tC_3$  and  $M_2$  for the algorithm main loop execution are listed in the figure 8 and figure 9 correspondingly. For the sake of simplicity evaluated matrices  $M_3$  are not included because its computation is a straightforward process. The last loop iteration is 7. Although the matrix  $M_2^6$  is updated with new non-terminal S for the cell [2,2] after transitive closure evaluation the

new values to the matrix  $M_2$  is not added. Therefore matrix  $M_2$  has stopped changing and the algorithm is successfully finished.

For the example query algorithm takes 7 iterations for the while – loop. The updated graph  $\mathcal{G}$  is depicted in the figure 10.

Fig. 8: Transitive closure for 3-6 loop iterations for example query

## 5 Evaluation

 $RedisGraph + CFPQ\_Data$ 

Cases, when kronecker should be significantly better that matrix. When grammar is big. When query is regular.

# 6 Conclusion

Future research. GraphBLAST. Paths, not just reachability.

$$M_{2}^{3} = \begin{pmatrix} \cdot & \{a\} & \{S\} & \cdot \\ \cdot & \cdot & \{a\} & \{S\} \\ \{a\} & \cdot & \cdot & \{b, S\} \end{pmatrix} M_{2}^{4} = \begin{pmatrix} \cdot & \{a\} & \{S\} & \cdot \\ \cdot & \cdot & \{a, S\} & \{S\} \\ \{a\} & \cdot & \cdot & \{b, S\} \end{pmatrix} M_{2}^{5} = \begin{pmatrix} \cdot & \{a\} & \{S\} & \{S\} \\ \cdot & \cdot & \{b\} & \cdot \end{pmatrix} M_{2}^{5} = \begin{pmatrix} \cdot & \{a\} & \{S\} & \{S\} \\ \cdot & \cdot & \{a, S\} & \{S\} \\ \{a\} & \cdot & \cdot & \{b, S\} \\ \cdot & \cdot & \{b\} & \cdot \end{pmatrix} M_{2}^{6} = \begin{pmatrix} \cdot & \{a\} & \{S\} & \{S\} \\ \{a\} & \cdot & \{S\} & \{S\} \\ \{a\} & \cdot & \{S\} & \{b, S\} \\ \cdot & \cdot & \{b\} & \cdot \end{pmatrix}$$

Fig. 9: The updated matrix  $M_2$  for 3-6 loop iterations for example query

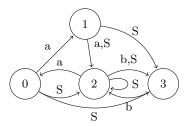


Fig. 10: The result graph  $\mathcal{G}$  for example query