

# Context-Free Path Querying by Kronecker Product<sup>\*</sup>

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**Abstract.** Context-free path queries (CFPQ) extend regular path queries (RPQ) by allowing context-free grammars to be used as constraints for paths. Algorithms for CFPQ are actively developed, but J. Kuijpers et al. have recently concluded, that existing algorithms are not performant enough to be used in real-world applications. Thus the development of new algorithms for CFPQ is justified. In this paper, we provide a new CFPQ algorithm which is based on such linear algebra operations as Kronecker product and transitive closure and handles grammars presented as recursive state machines. Thus, the proposed algorithm can be implemented by using high-performance libraries and modern parallel hardware. Moreover it avoids grammar growth which provides the possibility for queries optimization.

**Keywords:** Context-free path querying · Graph database · Context-free grammars · CFPQ · Kronecker product · Recursive state machines.

## 1 Introduction

Language-constrained path querying [3], and particularly context-free path querying (CFPQ) [13], allows one to express constraints for paths in a graph in terms of context-free grammars. A path in a graph is included into a query result only if the labels along this path form a word which belongs to the language, generated by the query grammar. CFPQ is widely used in bioinformatics [12], graph databases querying [5, 10, 9], and RDF analysis [14].

CFPQ algorithms are actively developed, but still suffer from poor performance [9]. The algorithm proposed by Rustam Azimov [2] is one of the most promising. This algorithm makes it possible to offload computational intensive computations to high-performance libraries for linear algebra, this way one can utilize modern parallel hardware for CFPQ. One disadvantage of this algorithm

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is that a query grammar should be converted to a Chomsky Normal Form (CNF) which significantly increases its size. Performance of the algorithm depends on the grammar size, thus it is desirable to create the algorithm which does not modify the query grammar.

In this work, we propose a new algorithm for CFPQ which can be expressed in terms of matrix operations and does not require grammar transformation. This algorithm can be efficiently implemented on modern parallel hardware and it provides ways to optimize queries. The main contribution of this paper could be summarized as follows.

1. We introduce a new algorithm for CFPQ, which is based on the intersection of recursive state machines and can be expressed in terms of Kronecker product and transitive closure.
2. We provide a step-by-step example of the algorithm.
3. We provide an evaluation of the presented algorithm and its comparison with the matrix-based algorithm. The presented algorithm outperforms the previous matrix-based algorithm in the worst-case scenario, but further optimizations are required to make it applicable for real-world cases.

## 2 Recursive State Machines

In this section, we introduce recursive state machines (RSM) [1]. This kind of computational machine extends the definition of finite state machines and increases the computational capabilities of this formalism.

A recursive state machine  $R$  over a finite alphabet  $\Sigma$  is defined as a tuple of elements  $(M, m, \{C_i\}_{i \in M})$ , where:

- $M$  is a finite set of labels of boxes.
- $m \in M$  is an initial box label.
- Set of *component state machines* or *boxes*, where  $C_i = (\Sigma \cup M, Q_i, q_i^0, F_i, \delta_i)$ :
  - $\Sigma \cup M$  is a set of symbols,  $\Sigma \cap M = \emptyset$
  - $Q_i$  is a finite set of states, where  $Q_i \cap Q_j = \emptyset, \forall i \neq j$
  - $q_i^0$  is an initial state for the component state machine  $C_i$
  - $F_i$  is a set of final states for  $C_i$ , where  $F_i \subseteq Q_i$
  - $\delta_i$  is a transition function for  $C_i$ , where  $\delta_i : Q_i \times (\Sigma \cup M) \rightarrow Q_i$

RSM behaves as a set of finite state machines (or FSM). Each FSM is called a *box* or a *component state machine* [1]. A box works almost the same as a classical FSM, but it also handles additional *recursive calls* and employs an implicit *call stack* to *call* one component from another and then return execution flow back.

According to [1], recursive state machines are equivalent to pushdown systems. Since pushdown systems are capable of accepting context-free languages [7], it is clear that RSMs are equivalent to context-free languages. Thus RSMs suit to encode query grammars. Any CFG can be easily converted to an RSM with one box per nonterminal. The box which corresponds to a nonterminal  $A$  is constructed using the right-hand side of each production for  $A$ . An example of

such RSM  $R$  constructed for the grammar  $G$  with productions  $S \rightarrow aSb \mid ab$  is provided in figure 1.

Since  $R$  is a set of FSMs, it is useful to represent  $R$  as an adjacency matrix for the graph where vertices are states from  $\bigcup_{i \in M} Q_i$  and edges are transitions between  $q_i^a$  and  $q_i^b$  with label  $l \in \Sigma \cup M$ , if  $\delta_i(q_i^a, l) = q_i^b$ . An example of such adjacency matrix  $M_R$  for the machine  $R$  is provided in section 3.1.

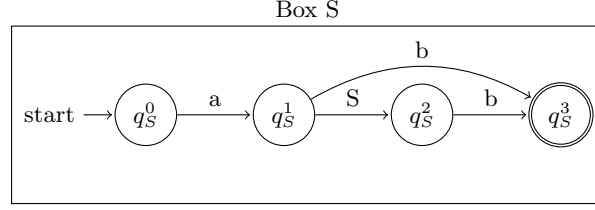


Fig. 1: The recursive state machine  $R$  for grammar  $G$

### 3 Kronecker Product Based CFPQ Algorithm

In this section, we introduce the algorithm for the computation of context-free reachability in a graph  $\mathcal{G}$ . The algorithm determines the existence of a path, which forms a sentence of the language defined by the input RSM  $R$ , between each pair of vertices in the graph  $\mathcal{G}$ . The algorithm is based on the generalization of the FSM intersection for an RSM, and an input graph. Since a graph can be interpreted as a FSM, in which transitions correspond to the labeled edges between vertices of the graph, and an RSM is composed of a set of FSMs, the intersection of such machines can be computed using the classical algorithm for FSM intersection, presented in [7].

The intersection can be computed as a Kronecker product of the corresponding adjacency matrices for an RSM and a graph. Since we are only determining the reachability of vertices, it is enough to represent intersection result as a Boolean matrix. It simplifies the algorithm implementation and allows one to express it in terms of basic matrix operations.

Listing 1 shows main steps of the algorithm. The algorithm accepts context-free grammar  $G = (\Sigma, N, P)$  and graph  $\mathcal{G} = (V, E, L)$  as an input. An RSM  $R$  is created from the grammar  $G$ . Note, that  $R$  must have no  $\varepsilon$ -transitions.  $M_1$  and  $M_2$  are the adjacency matrices for the machine  $R$  and the graph  $\mathcal{G}$  correspondingly.

Then for each vertex  $i$  of the graph  $\mathcal{G}$ , the algorithm adds loops with non-terminals, which allows deriving  $\varepsilon$ -word. Here the following rule is implied: each vertex of the graph is reachable by itself through an  $\varepsilon$ -transition. Since the machine  $R$  does not have any  $\varepsilon$ -transitions, the  $\varepsilon$ -word could be derived only if

a state  $s$  in the box  $B$  of the  $R$  is both initial and final. This data is queried by the *getNonterminals()* function for each state  $s$ .

The algorithm terminates when the matrix  $M_2$  stops changing. Kronecker product of matrices  $M_1$  and  $M_2$  is evaluated for each iteration. The result is stored in  $M_3$  as a Boolean matrix. For the given  $M_3$  a  $C_3$  matrix is evaluated by the *transitiveClosure()* function call. The  $M_3$  could be interpreted as an adjacency matrix for an oriented graph with no labels, used to evaluate transitive closure in terms of classical graph definition of this operation. Then the algorithm iterates over cells of the  $C_3$ . For the pair of indices  $(i, j)$ , it computes  $s$  and  $f$  — the initial and final states in the recursive automata  $R$  which relate to the concrete  $C_3[i, j]$  of the closure matrix. If the given  $s$  and  $f$  belong to the same box  $B$  of  $R$ ,  $s = q_B^0$ , and  $f \in F_B$ , then *getNonterminals()* returns the respective non-terminal. If the condition holds then the algorithm adds the computed non-terminals to the respective cell of the adjacency matrix  $M_2$  of the graph.

The functions *getStates* and *getCoordinates* (see listing 2) are used to map indices between Kronecker product arguments and the result matrix. The Implementation appeals to the blocked structure of the matrix  $C_3$ , where each block corresponds to some automata and graph edge.

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

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**Listing 1** Kronecker product based CFPQ

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1: function CONTEXTFREEPATHQUERYING( $G, \mathcal{G}$ )
2:    $R \leftarrow$  Recursive automata for  $G$ 
3:    $M_1 \leftarrow$  Adjacency matrix for  $R$ 
4:    $M_2 \leftarrow$  Adjacency matrix for  $\mathcal{G}$ 
5:   for  $s \in 0..dim(M_1) - 1$  do
6:     for  $i \in 0..dim(M_2) - 1$  do
7:        $M_2[i, i] \leftarrow M_2[i, i] \cup getNonterminals(R, s, s)$ 
8:   while Matrix  $M_2$  is changing do
9:      $M_3 \leftarrow M_1 \otimes M_2$  ▷ Evaluate Kroncker product
10:     $C_3 \leftarrow transitiveClosure(M_3)$ 
11:     $n \leftarrow dim(M_3)$  ▷ Matrix  $M_3$  size =  $n \times n$ 
12:    for  $i \in 0..n - 1$  do
13:      for  $j \in 0..n - 1$  do
14:        if  $C_3[i, j]$  then
15:           $s, f \leftarrow getStates(C_3, i, j)$ 
16:          if  $getNonterminals(R, s, f) \neq \emptyset$  then
17:             $x, y \leftarrow getCoordinates(C_3, i, j)$ 
18:             $M_2[x, y] \leftarrow M_2[x, y] \cup getNonterminals(R, s, f)$ 
19:  return  $M_2$ 

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**Listing 2** Help functions for Kronecker product based CFPQ

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1: function GETSTATES( $C, i, j$ )
2:    $r \leftarrow \dim(M_1)$   $\triangleright M_1$  is adjacency matrix for automata  $R$ 
3:   return  $\lfloor i/r \rfloor, \lfloor j/r \rfloor$ 
4: function GETCOORDINATES( $C, i, j$ )
5:    $n \leftarrow \dim(M_2)$   $\triangleright M_2$  is adjacency matrix for graph  $\mathcal{G}$ 
6:   return  $i \bmod n, j \bmod n$ 

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**3.1 Example**

This section provides a step-by-step demonstration of the presented algorithm. We consider the theoretical worst case for CFPQ time complexity, proposed by J.Hellings [5] as an example: the graph  $\mathcal{G}$  is presented in Figure 2a and the context-free grammar  $G$  for a language  $\{a^n b^n \mid n \geq 1\}$  is  $S \rightarrow aSb \mid ab$ .

Since the proposed algorithm processes grammar in form of recursive machine, we first provide RSM  $R$  in Figure 1. The initial box of the  $R$  is  $S$ , the initial state  $q_S^0$  is (0), the set of final states  $F_S = \{(3)\}$ .

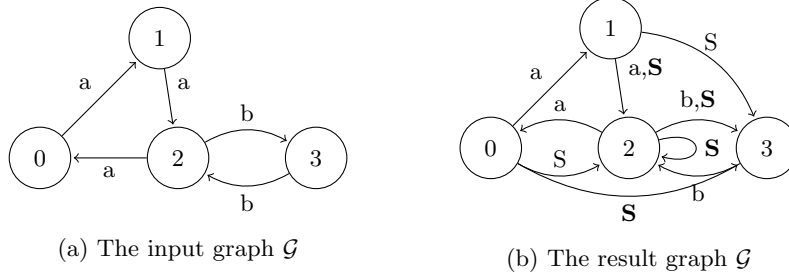


Fig. 2: The input and result graphs for example

Adjacency matrices  $M_1$  and  $M_2$  for automata  $R$  and graph  $\mathcal{G}$  respectively are initialized as follows:

$$M_1 = \begin{pmatrix} \cdot & \cdot & \{a\} & \cdot \\ \cdot & \cdot & \{S\} & \{b\} \\ \cdot & \cdot & \cdot & \{b\} \\ \cdot & \cdot & \cdot & \cdot \end{pmatrix}, \quad M_2^0 = \begin{pmatrix} \cdot & \cdot & \{a\} & \cdot \\ \cdot & \cdot & \cdot & \{a\} \\ \{a\} & \cdot & \cdot & \{b\} \\ \cdot & \cdot & \{b\} & \cdot \end{pmatrix}.$$

After the initialization in lines 2–4, the algorithm handles  $\varepsilon$ -case. Because the machine  $R$  does not have  $\varepsilon$ -transitions and  $\varepsilon$ -word is not included in the language, lines 5–7 of the algorithm do not affect the input data.

Then the algorithm enters the while loop and iterates while matrix  $M_2$  is changing. We provide both the values of the matrices  $M_3$ ,  $C_3$  at each algorithm step as well as how the matrix  $M_2$  is updated. The current loop



For the second iteration matrices  $M_3^2$  and  $C_3^2$  are evaluated as follows:

$$M_3^2 = \begin{pmatrix} \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{pmatrix}, C_3^2 = \begin{pmatrix} \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{pmatrix}.$$

New non-zero values in the matrix  $C_3^2$  have appeared during this iteration 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 the adjacency matrix  $M_2$ . The updated matrix and graph  $\mathcal{G}$  are shown in Figure 4.

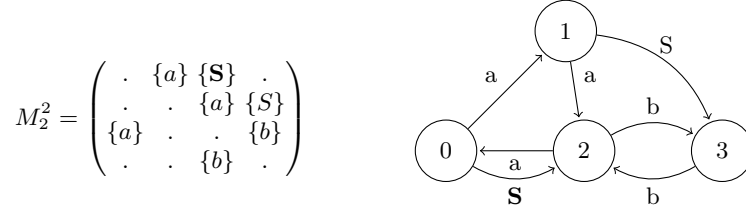


Fig. 4: The updated matrix  $M_2^2$  and graph  $\mathcal{G}$  after second loop iteration for example query

The remaining matrices  $C_3$  and  $M_2$  for the algorithm's main loop execution are listed in the Figure 5 and Figure 6 correspondingly. The evaluated matrices  $M_3$  are not included because its computation is straightforward. The last loop iteration is 7. Although the matrix  $M_2^6$  is updated with the new non-terminal  $S$  for the cell  $[2, 2]$  after the transitive closure evaluation, the new values are not added to the matrix  $M_2$ . Therefore matrix  $M_2$  has stopped changing and the algorithm has finished. The graph  $\mathcal{G}$  with the new edges is presented in the Figure 2b.

## 4 Evaluation

We implement the proposed algorithm by using SuiteSparse<sup>3</sup> [4]: the implementation of GraphBlas API [8]. GraphBlas API specifies a set of linear algebra primitives and operation which allows one to formulate graph algorithms using linear algebra over custom semirings.

<sup>3</sup> SuiteSparse is a sparse matrix software which includes GraphBLAS API implementation. Project web page: <http://faculty.cse.tamu.edu/davis/suitesparse.html>. Access date: 12.03.2020

[illegible]

Fig. 5: Transitive closure for loop iterations 3 – 6 for example query

$$M_2^3 = \begin{pmatrix} \cdot & \{a\} & \{S\} \\ \{a\} & \cdot & \{S\} \\ \cdot & \cdot & \{b\} \end{pmatrix} M_2^4 = \begin{pmatrix} \cdot & \{a\} & \{S\} \\ \cdot & \{a, S\} & \{S\} \\ \{a\} & \cdot & \{b, S\} \end{pmatrix}$$

$$M_2^5 = \begin{pmatrix} \cdot & \{a\} & \{S\} \\ \{a\} & \cdot & \{S\} \\ \cdot & \cdot & \{b\} \end{pmatrix} M_2^6 = \begin{pmatrix} \cdot & \{a\} & \{S\} \\ \cdot & \{a, S\} & \{S\} \\ \{a\} & \cdot & \{b, S\} \end{pmatrix}$$

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



We compare our implementation with results provided in [11], accordingly we use dataset described in this article which consists of **RDF**, **Worst case**, and **Full** subsets. For RDF querying we use same-generatoin query  $G_4$  from [11].

For evaluation, we use a PC with Ubuntu 18.04 installed. It has Intel(R) Core(TM) i7-4790 CPU @ 3.60GHz CPU, DDR4 32 Gb RAM.

The results of the evaluation are summarized in the table 1. Time is measured in seconds,  $t_1$  is an execution time for the proposed algorithm, and  $t_2$  is a time for M4RI-based implementation — the best CPU version form [11]. The result for the algorithm is averaged over 10 runs. We exclude the time required to load data from a file. The time required for data transfer and its conversion is included.

Table 1: Evaluation results

	Graph	#V	#E	$t_1$	$t_2$		Graph	#V	#E	$t_1$	$t_2$
RDF	atm-prim	291	685	0.24	0.02	RDF	core	1323	8684	0.28	0.12
	biomed	341	711	0.24	0.05		wine	733	2450	1.71	0.06
	foaf	256	815	0.07	0.02	Worst case	$WC_1$	64	65	0.03	0.04
	funding	778	1480	0.43	0.07		$WC_2$	128	129	0.16	0.23
	generations	129	351	0.04	0.03		$WC_3$	256	257	0.96	1.99
	people_pets	337	834	0.18	0.03		$WC_4$	512	513	7.14	23.21
	pizza	671	2604	1.14	0.08		$WC_5$	1024	1025	121.99	528.52
	skos	144	323	0.02	0.04	Full	$F_1$	100	100	0.17	0.02
	travel	131	397	0.05	0.05		$F_2$	200	200	1.04	0.03
	unv-bnch	179	413	0.05	0.04		$F_3$	500	500	18.86	0.03
	pathways	6238	37196	4.88	0.18		$F_4$	1000	1000	554.22	0.07

We can see, that while RDF querying time is better for M4RI in general, in some cases execution times are comparable. For example, for graphs *generations*, *travel*, *unv-bnch*, *skos*. For **Full** data set performance of our algorithm is bad because SuiteSparse is based on sparse matrix representation, and in this case matrices density changes aggressively from very sparse to full. At the same time, we can see, that in **Worst case** our algorithms up to 4 times faster than M4RI (graph  $WC_5$ ).

To sum up, our prototype implementation of the described algorithm is not performant enough to be used for real-world applications but it outperforms a matrix-based algorithm on **Worst case** dataset and comparable with it on some graphs from the **RDF** dataset. Thus we can conclude that we should improve our implementation to achieve better performance.

## 5 Conclusion

We presented a new algorithm for CFPQ which is based on Kronecker product and transitive closure. Thus it can be implemented by using high-performance

libraries for linear algebra. Also, our algorithm handles queries represented as recursive state machines, thus it avoids grammar growth.

We implement the proposed algorithm by using SuiteSparse and evaluate it on several graphs and queries. We show that in some cases our algorithm outperforms the matrix-based algorithm, but in the future, we should improve our implementation to be applicable for real-world graphs analysis.

Also in the future, we should investigate such formal properties of the proposed algorithm as time and space complexity. Moreover, we should analyze behavior dependency on query type and its form. Namely, we should analyze regular path queries evaluation and context-free path queries in the form of extended context-free grammars (ECFG) [6]. Utilization of ECFGs may provide a way to optimize queries by minimization of both the right parts of productions and whole result RSM.

Finally, it is necessary to compare our algorithm with the matrix-based one in cases when the size difference between Chomsky Normal Form and ECFG representation of the query is significant.

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