Parsing by matrix multiplication for the string-matching problem*

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Abstract. The abstract should briefly summarize the contents of the paper in 15-250 words.

Keywords: Parsing \cdot Matrix multiplication \cdot Context-free grammars \cdot String-matching \cdot Bioinformatics.

1 Introduction

Since the theory of context-free grammars was developed by Noam Chomsky, its utility has been extensively studied. The classic application of context-free grammars is describing natural and programming languages. Recent research has shown that the theory of formal languages and, in particular, context-free languages can be used in bioinformatics. For example, parsing is applicable to DNA and RNA processing. The secondary structure of the nucleotide sequence contains information about the organism species, therefore, it can be used in solving the recognition and classification problems. Moreover, in terms of parsing, the secondary structure can be described by some context-free grammar.

Such field of application as bioinformatics requires working with large amount of data. That is why it is important not only to develop more efficient recognition or parsing algorithms, but also to adapt the existing ones to new problems that come with the new application areas.

At this moment the asymptotically fastest parsing algorithm that can be applied to any context-free grammar was proposed by Leslie Valiant [1]. The algorithm computes the parsing table for a linear input, where each element of this table is responsible for deriving a particular substring. By offloading the most time-consuming computations on a Boolean matrix multiplication, Valiant has achieved an improvement in time complexity, which is O(BMM(n)log(n)) for an input string of size n, where BMM(n) is the number of operations needed

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to multiply two Boolean matrices of size n n. Nevertheless, this algorithm also has some drawbacks, for example, adaptation to the string-matching problems.

This paper aims to present a modification of Valiants algorithm partially dealing with the problem described above. We divide the parsing table into successively computed layers of disjoint submatrices of the same size through the reorganization of the matrix multiplication order. Thus, we can suspend the modification execution as soon as the layer we need was computed. Moreover, all submatrices in layer can be processed independently and further it allow to use GPGPU (General-Purpose computing on Graphics Processing Units) and parallel computation.

We make the following contribution:

- we introduce a modification of Valiants algorithm;
- we prove the correctness of the modification, and also we prove that time complexity is the same as for the original algorithm;
- we show the applicability of our modification for the string-matching problem.

2 Background

In this section we briefly describe the key definitions and the basic parsing algorithms which are necessary for further understanding of the results obtained in this paper.

2.1 Preliminaries

 Σ is a finite nonempty set called an alphabet, Σ^* is a set of all finite strings over Σ . A grammar is a quadruple (Σ, N, R, S) , where Σ is a finite set of terminals, N is a finite set of nonterminals, R is a finite set of productions of the form $\alpha \to \gamma$, where $\alpha \in V^*NV^*$, $\gamma \in V^*$, $V = \Sigma \cup N$ and $S \in N$ is a start symbol.

Definition 1. Grammar $G = (\Sigma, N, R, S)$ is called context-free, if $\forall r \in R$ are of the form $A \to \beta$, where $A \in N, \beta \in V^+$.

Definition 2. Context-free grammar $G = (\Sigma, N, R, S)$ is said to be in Chomsky normal form if $\forall r \in R$ are of the form:

- $-A \rightarrow BC$,
- $-A \rightarrow a$,
- $-S \rightarrow \epsilon$,

where $A, B, C \in N, a \in \Sigma, \epsilon$ is an empty string.

Definition 3. $L_G(A)$ is language of grammar $G_A = (\Sigma, N, R, A)$, which means all the sentences that can be derived in a finite number of steps from the start symbol A.

2.2 Parsing by matrix multiplication

The main problem of parsing is to verify if the input string belongs to the language of some given grammar L_G . In this section we will describe parsing algorithm, based on matrix multiplication. It has been proposed by Leslie Valiant and is the most asymptotically efficient parsing algorithm, which works for all context-free grammars, although they can be generalized to conjunctive and boolean grammars due to Alexander Okhotin.

The CYK algorithm is a basic parsing algorithm. Its main idea is to construct for an input string $a_1a_2...a_n$ a parsing table T of size $n \times n$, where

$$T_{i,j} = \{A | a_{i+1} ... a_j \in L_G(A)\} \quad \forall i < j$$
 (1)

and $G = (\Sigma, N, R, S)$ is a context-free grammar.

The elements of T are filled successively beginning with

$$T_{i-1,i} = \{A|A - > a_i \in R\} \tag{2}$$

Then.

$$T_{i,j} = f(P_{i,j}) \tag{3}$$

where

$$P_{i,j} = \bigcup_{k=i+1}^{j-1} T_{i,k} \times T_{k,j}$$
 (4)

$$f(P) = \{A | \exists A \to BC \in R : (B, C) \in P\}$$
 (5)

The input string $a_1a_2...a_n$ belongs to L_G if and only if $S \in T_{0,n}$.

The time complexity of this algorithm is $O(n^3)$. Valiant proposed to offload the most intensive computations to the Boolean matrix multiplication. As the most time-consuming is computing $\bigcup_{k=i+1}^{j-1} T_{i,k} \times T_{k,j}$, Valiant rearranged computation of $T_{i,j}$, in order to use multiplication of submatrices of T.

Definition 4. Let $X \in (2^N)^{m \times l}$ and $Y \in (2^N)^{l \times n}$ be two submatrices of parsing table T. Then, $X \times Y = Z$, where $Z \in (2^{N \times N})^{m \times n}$ and $Z_{i,j} = \bigcup_{k=1}^{l} X_{i,k} \times Y_{k,j}$.

In **Algorithm 1** full pseudo-code of Valiant's algorithm written in the terms proposed by Okhotin, is presented. All elements of T and P are initialized by empty sets. Then, the elements of these two table are successively filled by two recursive procedures.

The procedure compute(l, m) constructs the correct values of $T_{i,j} \forall l \leq i < j < m$.

The procedure complete(l, m, l', m') constructs the submatrix $\forall T_{i,j} \ l \leq i < m, \ l' \leq j < m'$. This procedure assumes $T_{i,j} \forall l \leq i < j < m, \ l' \leq i < j < m'$ are already constructed and the current value of $P[i,j] = \{(B,C) | \exists (m \leq k < l') : a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\} \ \forall l \leq i < m, \ l' \leq j < m'$.

Algorithm 1: Parsing by matrix multiplication: Valiant's Version

```
Input: Grammar G = (\Sigma, N, R, S), w = a_1...a_n, n \ge 1, a_i \in \Sigma, where n + 1 — power
                 of two
 1 main():
 2 compute(0, n + 1);
 3 accept if and only if S \in T_{0,n}
 4 compute(l, m): if m - l \ge 4 then
           \begin{array}{c} \hline compute(l,\frac{l+m}{2}); \\ compute(\frac{l+m}{2},m) \end{array}
 6
 7 end
 8 complete(l, \frac{l+m}{2}, \frac{l+m}{2}, m)
 9 complete(l, m, l', m'): if m - l = 4 and m = l' then
10
           T_{l,l+1} = \{A|A \to a_{l+1} \in R\};
11 end
12 else if m - l = 1 and m < l' then
13
           T_{l,l'} = f(P_{l,l'});
14 end
15 else if m-l>1 then
          B = (l, \frac{l+m}{2}, \frac{l'+m'}{2}, m'), B' = (\frac{l+m}{2}, m, l', \frac{l'+m'}{2}),
C = (\frac{l+m}{2}, m, l', \frac{l'+m'}{2}), D = (l, \frac{l+m}{2}, l', \frac{l'+m'}{2}),
D' = (\frac{l+m}{2}, m, \frac{l'+m'}{2}, m'), E = (l, \frac{l+m}{2}, \frac{l'+m'}{2}, m');
           complete(C);
17
           P_D = P_D \cup (T_B \times T_C);
18
           complete(D);
19
20
           P_{D'} = P_{D'} \cup (T_C \times T_{B'});
           complete(D');
21
           P_E = P_E \cup (T_B \times T_{D'});
22
           P_E = P_E \cup (T_D \times T_{B'});
23
           complete(E)
\mathbf{24}
25 end
26 complete(l, \frac{l+m}{2}, \frac{l+m}{2}, m)
```

Then Valiant described that product of multiplying of two submatrices of parsing table T can be provided as $|N|^2$ Boolean matrices (for each pair of nonterminals). Denote matrix corresponding to pair $(B,C) \in N \times N$ as $Z^{(B,C)}$, then $Z^{(B,C)}_{i,j} = 1$ if and only if $(B,C) \in Z_{i,j}$. It should also be noted that $Z^{(B,C)} = X^B \times Y^C$. So, matrix multiplication in **Definition 4** can be replaced by Boolean matrix multiplication, each of which can be computed independently. Following these changes, time complexity of **Algorithm 1** is O(|G|BMM(n)log(n)) for an input string of length n, where BMM(n) is the number of operations needed to multiply two Boolean matrices of size $n \times n$.

3 Modification of Valiant's algorithm

In this section we describe the modification of Valiant's algorithm, which has a number of advantages, such as possibility to broke it down into several subtasks that can be processed independently. Also this version can be simply applied to the string-matching problem, which often arises in text editing, DNA and RNA sequence analysis.

The main change of this modification is to divide the parsing table into layers of disjoint submatrices of the same size. The division, we have made from the reorganization of the matrix multiplication order, is presented in **Figure 1**. The layers are computes successively from the bottom up.

Let us consider the pseudo-code of the modification, which is written in **Algorithm 2**. The procedure main() computes the lowest layer $(T_{l,l+1})$, and then divide the table into layers, described earlier, and computes them through the completeVLayer() call. Thus, main() computes all elements of parsing table T correctly.

Algorithm 2: Parsing by matrix multiplication: Modified Version

```
Input: Grammar G = (\Sigma, N, R, S), w = a_1...a_n, n \ge 1, a_i \in \Sigma, where n + 1 — power
                of two
 1 main():
 2 for l \in \{1, ..., n\} do
          T_{l,l+1} = \{A|A \to a_{l+1} \in R\}
 3
 4 end
 5 for 1 \le i < k do
          layer = constructLayer(i);
 6
           complete V Layer(layer)
 7
 8 end
 9 constructLayer(i):
10 \{B|\exists k \geq 0: B = (k*2^i, (k+1)*2^i, (k+1)*2^i, (k+2)*2^i)\}
11 completeLayer(M):
12 if \forall (l, m, l', m') \in M \quad (m - l = 1) then
          for (l, m, l', m') \in M do
13
               T_{l,l'} = f(P_{l,l'});
14
15
16 end
17 else
          \text{bottomLayer} = \{(\tfrac{l+m}{2}, m, l', \tfrac{l'+m'}{2}) | (l, m, l', m') \in M\};
18
19
          completeLayer(bottomLayer);
          complete VLayer(M)
20
21 end
22 comleteVLayer(M):
26 multiplicationTask1 =
      \{(l, m, l', m'), (l, m, m, 2m - l), (m, 2m - l, l', m') | (l, m, l', m') \in leftSubLayer\} \cup \{(l, m, l', m'), (l, m, m, 2m - l), (m, 2m - l, l', m') | (l, m, l', m') \in leftSubLayer\} \cup \{(l, m, l', m'), (l, m, m, 2m - l), (m, 2m - l, l', m') | (l, m, l', m') \in leftSubLayer\} \cup \{(l, m, l', m'), (l, m, m, 2m - l), (m, 2m - l, l', m') | (l, m, l', m') \in leftSubLayer\} \cup \{(l, m, l', m'), (l, m, m, 2m - l), (m, 2m - l, l', m') | (l, m, l', m') \in leftSubLayer\} \cup \{(l, m, l', m'), (l, m, m, 2m - l), (l, m, l', m'), (l, m, l', m') \in leftSubLayer\}
      \{(l, m, l', m'), (l, m, 2l' - m', l'), (2l' - m', l', l', m') | (l, m, l', m') \in rightSubLayer\};
27 multiplicationTask2 =
      \{(l, m, l', m'), (l, m, m, 2m - l'), (m, 2m - l, l', m') | (l, m, l', m') \in topSubLayer\};
28 multiplicationTask3 =
      \{(l, m, l', m'), (l, m, 2l' - m', l'), (2l' - m', l', l', m') | (l, m, l', m') \in topSubLayer\};
29 performMultiplications(multiplicationTask1);
30 completeLayer(leftSubLayer <math>\cup rightSubLayer);
31 performMultiplications(multiplicationTask2);
32 performMultiplications(multiplication Task3);
33 completeLayer(topSubLayer)
```

Denote some subsidiary functions for matrix m:

```
-bottom(m) = (\frac{l+m}{2}, m, l', \frac{l'+m'}{2}), 
-left(m) = (l, \frac{l+m}{2}, l', \frac{l'+m'}{2}), 
-right(m) = (\frac{l+m}{2}, m, \frac{l'+m'}{2}, m'), 
-top(m) = (l, \frac{l+m}{2}, \frac{l'+m'}{2}, m').
```

The procedure completeVLayer(M) takes an array of disjoint submatrices M. For each $m = (l, m, l', m') \in M$ this procedure computes left(m), right(m), top(m). The procedure assumes that the elements of bottom(m) and all $T_{i,j} \forall l \leq i < j < m, l' \leq i < j < m'$ are already constructed. Also it is assumed that the current value of $P[i,j] = \{(B,C)|\exists (m \leq k < l'): a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\}$ $\forall l \leq i < m, l' \leq j < m'$.

The procedure completeLayer(M) also takes an array of disjoint submatrices M, but unlike the previous one, it computes $T_{i,j} \forall (i,j) \in m$. This procedure, just as in the previous case, assumes that $T_{i,j} \forall l \leq i < j < m, l' \leq i < j < m'$ are already constructed and the current value of $P[i,j] = \{(B,C) | \exists (m \leq k < l') : a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\} \ \forall l \leq i < m, l' \leq j < m'$.

Algorithm 3 describes how the procedure performMultiplication(task), where task is an array of a triple of submatrices, works. It is worth mentioning that, as distinct from the original algorithm, $|tasks| \ge 1$ and all these multiplications can be computed independently.

Algorithm 3:

```
1 performMultiplication(task):
2 for (m, m1, m2) \in M do
```

 $\mathbf{3} \qquad P_m = P_m \cup (T_{m1} \times T_{m2});$

4 end

4 Proof of correctness

In this section ...

Theorem 1. Let M be a submatrix array. Assume that $T[i,j] = \{A|a_{i+1}...a_j \in L(A)\}\ \forall l \leq i < j < m, l' \leq i < j < m' \ and \ P[i,j] = \{(B,C)|\exists (m \leq k < l'): a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\}\ \forall l \leq i < m, l' \leq j < m' \ \forall (l,m,l',m') \in M.$ Then the procedure complete Layer (M), returns correctly computed sets of $T[i,j]\ \forall l \leq i \leq m, l' \leq j \leq m'\ \forall (l,m,l',m') \in M.$

Proof. Induction on m - l. (Hereinafter denoting (l, m, l', m') as a typical example of array M, and all the computations are implemented for all submatrices in M).

The base case: m - l = 1. There is only one element to compute, and P[l, l'] = $\{(B,C)|a_{l+1}...a_{l'}\in L(B)L(C)\}$. Further, algorithm computes $f(P[l,l'])=\{A|a_{l+1}...a_{l'}\in A_{l+1}...a_{l'}\}$ L(A), so T[l, l'] computed correctly.

For the induction step, assume that (11, m1, 12, m2) is correctly computed for m2 - l2 = m1 - l1 > m - l.

Let us consider complete completeLayer(M), where m - l > 1.

Firstly, consider $completeLayer(bottom = \{(\frac{l+m}{2}, m, l', \frac{l'+m'}{2})\})$, as theorem conditions are fulfilled, then this call returns correct sets $T[i,j] \ \forall (i,j) \in bottom$ (hereinafter is means $\forall (i,j) \in m \ \forall m \in bottom$). All submatrices with size m1 – l1 > m-l, all previous layers and also bottom(M) are correct, so, complete VLayer(M)can be called, and multiplication By Task(task1) adds to each $P[i,j] \ \forall (i,j) \in$ $left = \{(\frac{l+m}{2}, m, l', \frac{l'+m'}{2}))\}$ all pairs $\{(B, C) | \exists (\frac{l+m}{2} \leq k < l') : a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\}$ and $\forall (i, j) \in right = \{(\frac{l+m}{2}, m, \frac{l'+m'}{2}, m')\}$ all pairs $\{(B,C)|\exists (m \leq k < \frac{l'+m'}{2}): a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)\}.$ Now all the theorem conditions are fulfilled so, it is possible to call $completeLayer(left \cup right)$, which returns correct sets $T[i,j] \ \forall (i,j) \in (\text{left } \cup right).$

Next, multiplicationByTask(task2) and multiplicationByTask(task3) add to each $P[i,j] \ \forall (i,j) \in top = \{(l,\frac{l+m}{2},\frac{l'+m'}{2},m'))\}$ all pairs $\{(B,C)|\exists (\frac{l+m}{2} \leq k < m')\}$ m) and $(l' \le k < \frac{l'+m'}{2}): a_{i+1}...a_k \in L(B), a_{k+1}...a_j \in L(C)$ }. Now all the theorem conditions are fulfilled so, it is possible to call completeLayer(top), which returns correct sets $T[i,j] \ \forall (i,j) \in top$.

Thus, all $T[i,j] \ \forall (i,j) \in M$ are computed correctly.

Theorem 2. Let M be a submatrix array. Assume that, $T[i,j] = \{A|a_{i+1}...a_j \in$ $L(A)\} \ \forall l \leq i < j < m, l' \leq i < j < m' \ and \ \forall b1 \leq i < b2, \ b3 \leq j < b4, \\ where \ (b1, b2, b3, b4) = (\frac{l+m}{2}, m, l', \frac{l'+m'}{2}), \ also \ P[i, j] = \{(B, C) | \exists (m \leq k < l'): \}$ $a_{i+1}...a_k \in L(B), a_{k+1}...a_i \in L(C)$ $\forall l \leq i < m, l' \leq j < m' \ \forall (l, m, l', m') \in M.$ Then, the procedure complete VLayer(M), returns correctly computed sets of $T[i,j] \ \forall l \le i \le m, l' \le j \le m' \ \forall (l,m,l',m') \in M.$

Proof. The proof is similar to the proof of Theorem 1.

Note 1. Function costructLayer(i) returns $2^{k-i} - 1$ matrices of size 2^i .

Lemma 1.

- $\forall i \in \{1,..,k-1\} \sum |layer| \text{ for the calls of complete VLayer(layer) where}$
- $\forall (l, m, l', m') \in layer \ with \ m l = 2^{k-i} \ is \ exactly \ 2^{2i-1} 2^{i-1};$ $\forall i \in \{1, ..., k-1\} \ products \ of \ submatrices \ of \ size \ 2^{k-i} \times 2^{k-i} \ are \ calculated$ $exactly 2^{2i-1} - 2^i$

Proof. The base case: i = 1. complete VLayer(layer) where $\forall (l, m, l', m') \in layer$ with $m-l=2^{k-1}$ is called only once in the main() and |layer|=1. So, 2^{2i-1} $2^{i-1} = 2^1 - 2^0 = 1$.

For the induction step, assume that $\forall i \in \{1,..,j\} \sum |layer|$ for the calls of complete VLayer (layer) where $\forall (l, m, l', m') \in layer \text{ with } m - l = 2^{k-i} \text{ which is}$ exactly $2^{2i-1} - 2^{i-1}$.

Let us consider i = j + 1.

Firstly, it is the call of complete V Layer(costruct Layer(k - i)), where costruct Layer(i) returns $2^i - 1$ matrices of size 2^i . Secondly, complete V Layer(layer) is called 3 times for the left, right and top submatrices of size $2^{k-(i-1)}$. Finally, complete V Layer(layer) is called 4 times for the bottom, left, right and top submatrices of size $2^{k-(i-2)}$, except $2^{i-2} - 1$ matrices which were already computed.

Then,
$$\sum |layer| = 2^i - 1 + 3 \times (2^{2(i-1)-1} - 2^{(i-1)-1}) + 4 \times (2^{2(i-2)-1} - 2^{(i-2)-1}) - (2^{i-2} - 1) = 2^{2i-1} - 2^{i-1}$$
.

To calculate the number of products of submatrices of size $2^{k-i} \times 2^{k-i}$, we consider the calls of $complete\ V Layer(layer)$ where $\forall (l,m,l',m') \in layer$ with $m-l=2^{k-(i-1)}$, which is $2^{2(i-1)-1}-2^{(i-1)-1}$. During these calls $perform\ Multiplications$ run 3 times, $|multiplication\ Task1| = 2 \times 2^{2(i-1)-1} - 2^{(i-1)-1}$ and $|multiplication\ Task2| = |multiplication\ Task3| = 2^{2(i-1)-1} - 2^{(i-1)-1}$. So, the number of products of submatrices of size $2^{k-i} \times 2^{k-i}$ is $4 \times (2^{2(i-1)-1}-2^{(i-1)-1}) = 2^{2i-1}-2^i$.

Theorem 3. The time complexity of the Algorithm 1 is O(|G|BMM(n)log(n)) for an input string of length n, where G is a context-free grammar in Chomsky normal form, BMM(n) is the number of operations needed to multiply two Boolean matrices of size $n \times n$.

Proof. The proof is almost identical with that of the theorem given by Okhotin [2], because, as shown in the last lemma, the Algorithm 1 has the same number of products of submatrices.

5 Applications

Let 2^k be an input string length. Let s be a substring length. $2^{l'-2} < s \le 2^{l'-1}$ Let $2^{l'}$ be a maximum size of layer to calculate. l' = k - (m-2), (m-2) = k - l'

Let 2 be a maximum size of layer to calculate.
$$t = \kappa - (m-2), (m-2) = \kappa - t$$

$$C \sum_{i=m}^{k} 2^{2i-1} \cdot 2^{\omega(k-i)} \cdot f(2^{k-i}) = C \sum_{i=2}^{l'} 2^{2(i+(k-l'))-1} \cdot 2^{\omega(l'-i)} \cdot f(2^{l'-i}) = C \cdot 2^{\omega l'} \sum_{i=2}^{l'} 2^{(2-\omega)i} \cdot 2^{2(k-l')-1} \cdot f(2^{l'-i}) \le C \cdot 2^{\omega l'} f(2^{l'}) \cdot 2^{2(k-l')-1} \sum_{i=2}^{l'} 2^{(2-\omega)i} = BMM(2^{l'}) \cdot 2^{2(k-l')-1} \sum_{i=2}^{l'} 2^{(2-\omega)i}$$

$$\sum_{i=2}^{l'} 2^{(2-\omega)i} = \text{const}, \text{ if } \omega > 2$$

$$\sum_{i=2}^{l'} 2^{(2-\omega)i} = l'-1, \text{ if } \omega = 2$$

Time complexity for substrings of size s is $O(|G|BMM(2^{l'})(l'-1))$, while time complexity for the full input string is $O(|G|BMM(2^k)(k-1))$.

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