

# Words Distinguished by their Subwords

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# Words distinguished by their subwords

Word  $y$  distinguishes words  $x_1$  and  $x_2$  if  $y$  is a subword of exactly one of  $x_1$  and  $x_2$ .

$y$  is a subword of  $x$  if there exist words  $y_1, y_2, \dots, y_n, x_0, x_1, \dots, x_n$  in  $A^*$ , for some  $n \geq 1$ , such that

$$y = y_1 y_2 \cdots y_n \quad \text{and} \quad x = x_0 y_1 x_1 y_2 x_2 \cdots y_n x_n.$$

Sometimes  $y$  is called “a subsequence” of  $x$ .

There exist a word distinguishing  $x_1$  and  $x_2$  iff  $x_1$  and  $x_2$  are distinct.

We are interested in finding a shortest word distinguishing  $x_1$  and  $x_2$  (if there is one).

Result to be presented here: an  $O(|A|(|x_1| + |x_2|))$  time complexity algorithm.

Main result: an  $O(|A| + |x_1| + |x_2|)$  time complexity algorithm.

# Some examples

$x_1 =$  b c a c a c a d c a b a b a b d a c b a

$x_2 =$  b a c c a d c a b b c b c a b a d a b c

$y_1 =$  d a c b a c

$y_2 =$  d c c c

$y_1$  and  $y_2$  both distinguish  $x_1$  and  $x_2$

$y_2$  is a shortest word that distinguishes  $x_1$  and  $x_2$ :

$x_1$  and  $x_2$  have the same subwords up to length 3

# Some theory

$x_1$  and  $x_2$  are  $m$ -equivalent if they have the same subwords of length up to  $m$ :  $x_1 \equiv x_2 [J_m]$ .

Let  $y$  be a shortest word distinguishing  $x_1 \neq x_2$ . We define:

$$\delta(x_1, x_2) = \begin{cases} \infty & \text{if } x_1 = x_2, \\ |y| - 1 & \text{otherwise.} \end{cases}$$

$\delta(x_1, x_2)$  is the greatest  $m$  for which  $x_1 \equiv x_2 [J_m]$ .

Two one-sided particular cases:  $\delta(ua, u)$  and  $\delta(av, v)$  lead to right and left distinguishers for  $u, v \in A^*$  and  $a \in A$

**A fundamental property:**

$$\delta(uav, uv) = \delta(ua, u) + \delta(av, v).$$



# Computing a shortest distinguisher for $x_1$ and $x_2$

Distinguisher( $x_1, x_2$ )

- 1   ▷ Preparing the leftist data structures
- 2    $Ld_1 \leftarrow \text{LeftDistinguisher}(x_1, x_2)$
- 3    $Ld_2 \leftarrow \text{LeftDistinguisher}(x_2, x_1)$
- 4   ▷ Merging  $x_1$  and  $x_2$  into  $z$
- 5    $(z, Rd, m, jm, psm, (im, am)) \leftarrow \text{Merge}(x_1, x_2)$
- 6   **if**  $m = \infty$  **then**                   ▷  $x_1 = x_2$ , they can not be distinguished
- 7   **else**        $yr \leftarrow \text{Collect}(Rd, (jm - 1, z[jm]))$
- 8        $yl \leftarrow \text{Collect}(Ld_{psm}, (im, am))$
- 9        $y \leftarrow \text{reverse}(yr).z[jm].yl$
- 10    ▷ We just collected a shortest distinguisher in  $y$ ,  $|y| = m + 1$
- 11        $|y| \leftarrow m + 1$
- 12   **return**  $(m, y)$

# Left distinguisher $Ld$ of $x$ with embeddings in $y$

LeftDistinguisher( $x, y$ )

```
1  for  $a \in A$  do                                ▷ Subword automaton of the reverse of  $y$ 
2      Auto[0,  $a$ ]  $\leftarrow$  DeadState
3      Auto[DeadState,  $a$ ]  $\leftarrow$  DeadState
4  for  $i$  from 1 to  $|y|$  do
5      for  $a \in A$  do
6          if  $y[i] = a$  then
7              Auto[ $i, a$ ]  $\leftarrow i - 1$ 
8          else Auto[ $i, a$ ]  $\leftarrow$  Auto[ $i - 1, a$ ]
9  for  $a \in A$  do                                ▷ Left Distinguisher matrix  $Ld$  of  $x$ 
10     Ld[ $|x|, a$ ]  $\leftarrow (0, \text{nil}, |y|)$ 
11  for  $i$  from  $|x|$  to 1 do
12     for  $a \in A$  do
13          $(l, n, s) \leftarrow Ld[i, x[i]]$ 
14         if  $a = x[i]$  or length(Ld[ $i, a$ ])  $> 1 + l$  then
15             Ld[ $i - 1, a$ ]  $\leftarrow (1 + l, (i, x[i]), \text{Auto}[s, x[i]])$ 
16         else Ld[ $i - 1, a$ ]  $\leftarrow$  Ld[ $i, a$ ]
17  return Ld
```

# Merging $x_1 + x_2 \Rightarrow z$ without new short subwords

Merge( $x_1, x_2$ )

```
1   $(k_1, k_2, j, m) \leftarrow (0, 0, 0, \infty)$ 
2  for  $a \in A$  do
3       $\text{Rd}[0, a] \leftarrow (0, \text{nil}, \text{nil})$ 
4  while  $k_1 < |x_1|$  or  $k_2 < |x_2|$  do
5       $j \leftarrow j + 1$ 
6       $(\text{Case}, pz, ps, (i', a')) \leftarrow \text{MergeStep}()$ 
7       $(k_{pz}, z[j]) \leftarrow (k_{pz} + 1, x[pz][k_{pz}])$ 
8      for  $a \in A$  do
9           $(l, n, s) \leftarrow \text{Rd}[j - 1, z[j]]$ 
10         if  $a = z[j]$  or  $\text{length}(\text{Rd}[j - 1, a]) > 1 + l$  then
11              $\text{Rd}[j, a] \leftarrow (1 + l, (j - 1, z[j]), \text{nil})$ 
12         else  $\text{Rd}[j, a] \leftarrow \text{Rd}[j - 1, a]$ 
13         if  $\text{Case} \neq \text{Match}$  then
14              $m' \leftarrow \text{length}(\text{Rd}[j - 1, z[j]]) + \text{length}(\text{Ld}_{ps}[i', a'])$ 
15             if  $m = \infty$  or  $m' < m$  then
16                  $(m, jm, psm, (im, am)) \leftarrow (m', j, ps, (i', a'))$ 
17   $|z| \leftarrow j$ 
18  return  $(z, \text{Rd}, m, jm, psm, (im, am))$ 
```



# Auxiliary procedure MergeStep

MergeStep()

```
1  if  $k_1 < |x_1|$  and  $k_2 < |x_2|$  then
2       $(b_1, b_2) \leftarrow (x_1[k_1 + 1], x_2[k_2 + 1])$ 
3      if  $b_1 = b_2$  then
4           $(pz, k_2, \text{Case}) \leftarrow (1, k_2 + 1, \text{Match})$ 
5      else  $\text{Case} \leftarrow \text{Mismatch}$ 
6           $(pz, ps, (i', a')) \leftarrow \text{Race}(k_1, b_2, k_2, b_1)$ 
7  else  $\text{Case} \leftarrow \text{Singleton}$ 
8      if  $k_1 = |x_1|$  then
9           $pz \leftarrow 2$ 
10     else  $pz \leftarrow 1$ 
11      $\triangleright (i', a')$  is a pointer in  $\text{Ld}_{ps}$ 
12      $(ps, (i', a')) \leftarrow (pz, (|x_{ps}|, z[j]))$ 
13 return  $(\text{Case}, pz, ps, (i', a'))$ 
14  $\triangleright$  If  $\text{Case} = \text{Merge}$  then  $ps$  and  $(i', a')$  are undefined
```

# Confronting $x_1$ and $x_2$ in the case of a mismatch

Race( $k_1, b_2, k_2, b_1$ )

- 1 **if**  $\text{length}(\text{Ld}_1[k_1, b_2]) \leq \text{length}(\text{Ld}_2[k_2, b_1])$  **then**
- 2        $(ps, pl) \leftarrow (1, 2)$
- 3 **else**        $(ps, pl) \leftarrow (2, 1)$
- 4  $\triangleright \text{Ld}_{ps}[k_{ps}, b_{pl}]$  **is the left distinguisher which won the race**
- 5  $(i', a') \leftarrow (k_{ps}, b_{pl})$
- 6 **if**  $\text{next}(\text{Ld}_{ps}[k_{ps}, b_{pl}]) = \text{nil}$  **or**  $\text{suffix}(\text{Ld}_{ps}[k_{ps}, b_{pl}]) \geq 1 + k_{pl}$  **then**
- 7        $pz \leftarrow pl$
- 8 **else**        $pz \leftarrow ps$
- 9 **return**  $(pz, ps, (i', a'))$         $\triangleright (i', a')$  **is a pointer in**  $\text{Ld}_{ps}$

# Collecting the word encoded in a (one sided) distinguisher

Collect( $D, (i, a)$ )

- 1  $\triangleright$  We collect the word  $y$  encoded in  $D[i, a]$ ; with  $|y| = \text{length}(D[i, a])$
- 2  $(j, k) \leftarrow (1, \text{length}(D[i, a]))$
- 3 **while**  $k > 0$  **do**
- 4        $(l, (i, a), s) \leftarrow D[i, a]$         $\triangleright$  We must have  $l = k$
- 5        $y[j] \leftarrow a$
- 6        $(j, k) \leftarrow (j + 1, k - 1)$
- 7  $|y| \leftarrow j - 1$
- 8 **return**  $y$

# The merging of the example words

x\_1= b c a c      a c a d c a b                      a b a b d a c b      a  
      | | | |      | | | | | | | |                      | | | | | | | |      |  
z= b c a c c a c a d c a b b c b c a b a b d a c b c a  
      |      | | | |                      | | | | | | | | | | | |      | |      | |  
x\_2= b      a c c a                      d c a b b c b c a b a      d a      b c  
  
y=                      d c                      c      c

# Part of the $|A| \times |x_1|$ table of left distinguishers of $x_1$

$A$	...	11	12	13	14	15
$a$	...	$(3, (12, a), 13)$	=	$(2, (14, a), 15)$	=	$(1, (16, d), 16)$
$b$	...	=	$(3, (13, b), 11)$	=	$(2, (15, b), 14)$	=
$c$	...	=	=	=	=	=
$d$	...	=	=	=	=	$(1, (16, d), 16)$

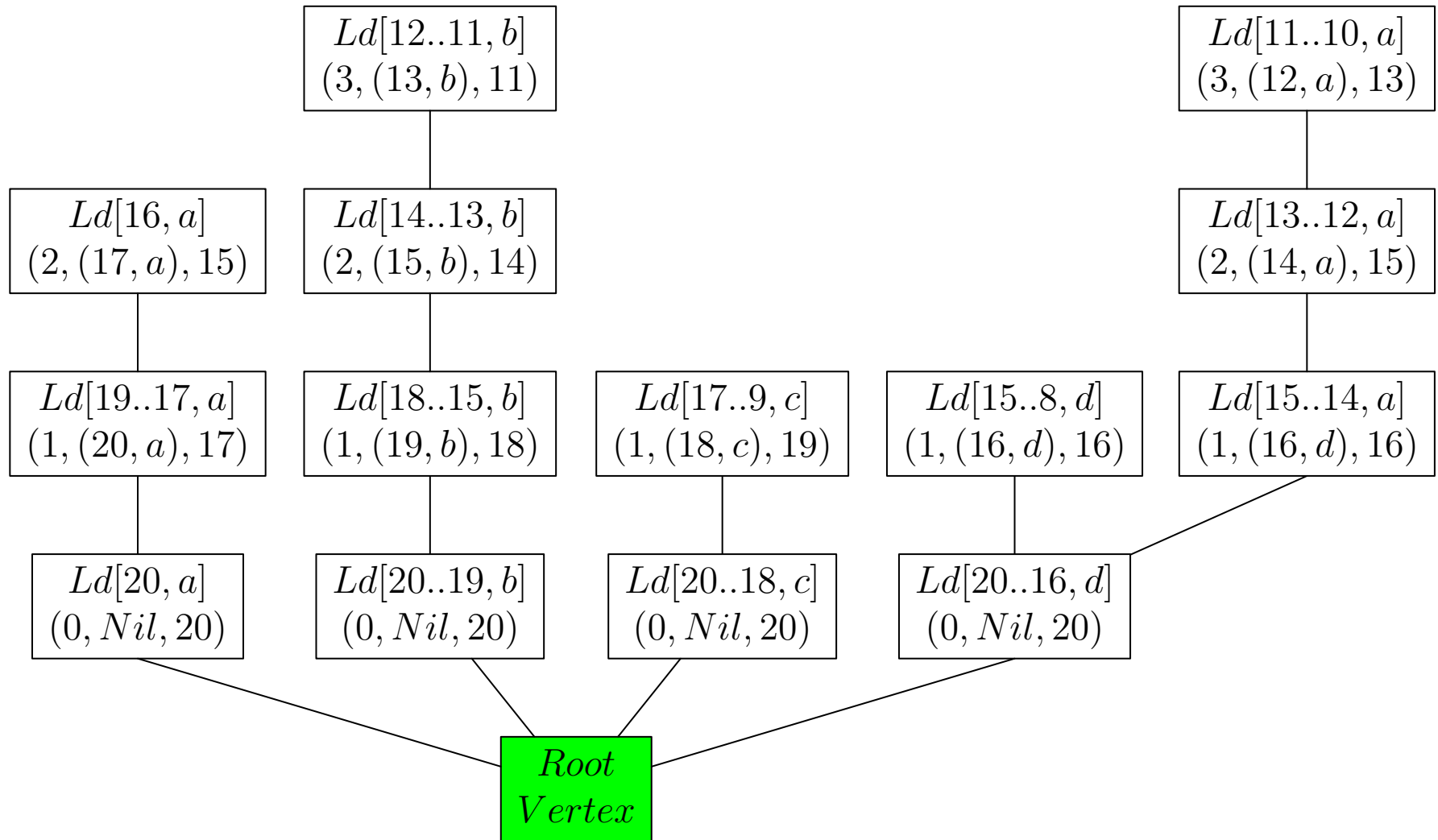
$A$	16	17	18	19	20
$a$	$(2, (17, a), 15)$	=	=	$(1, (20, a), 17)$	$(0, \text{nil}, 20)$
$b$	=	=	$(1, (19, b), 18)$	=	$(0, \text{nil}, 20)$
$c$	=	$(1, (18, c), 19)$	=	=	$(0, \text{nil}, 20)$
$d$	=	=	=	=	$(0, \text{nil}, 20)$

$$\text{Ld}_1[13, a] = (2, (14, a), 15)$$

$$\delta(a \cdot x_1[13..20], x_1[13..20]) = 2 \text{ where } x_1[13..20] = \text{abdacba}$$

word stored in  $\text{Ld}_1[13, a]$  is  $ad$  which is a subword of  $x_2[15..20] = \text{adabc}$

# Part of the left distinguisher tree of $x_1$





**From**  $O(|A|(|x_1| + |x_2|))$  **to**  $O(|A| + |x_1| + |x_2|)$

The  $|A| \times |x|$  matrices are sparse: they have  $O(|A| + |x_1| + |x_2|)$  different elements

We use inversion techniques to obtain the left distinguisher trees

It would be nice to have mechanisms to recover  $\text{Ld}[i, a]$  in  $O(1)$  time (does such a mechanism exist?)

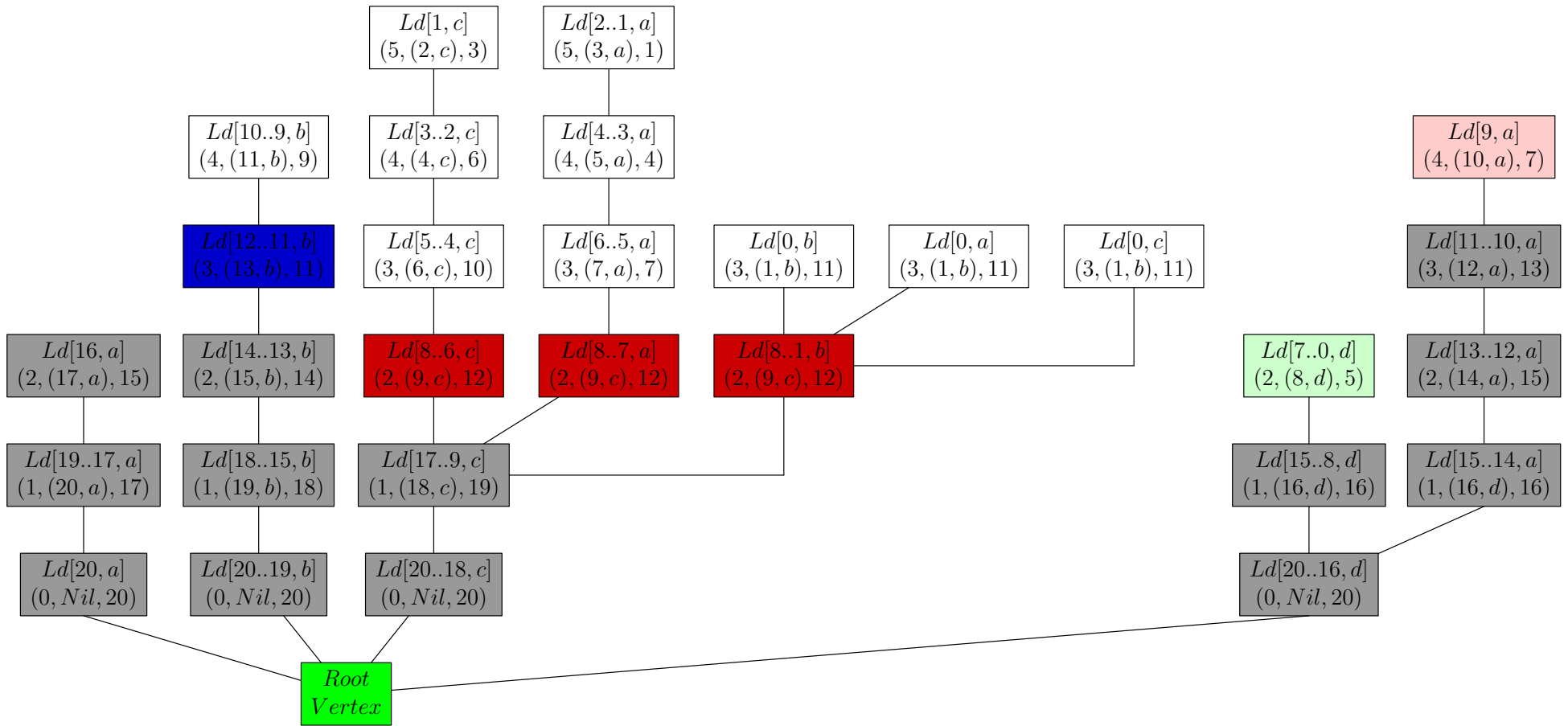
We can survive with an  $O(|A| + |x|)$  **sum** of access times throughout the algorithm (**the right data in the right place at the right time**)

Computing the  $s$ 's in  $\text{Ld}[i, a] = (l, (i', a'), s)$  is the trickiest part





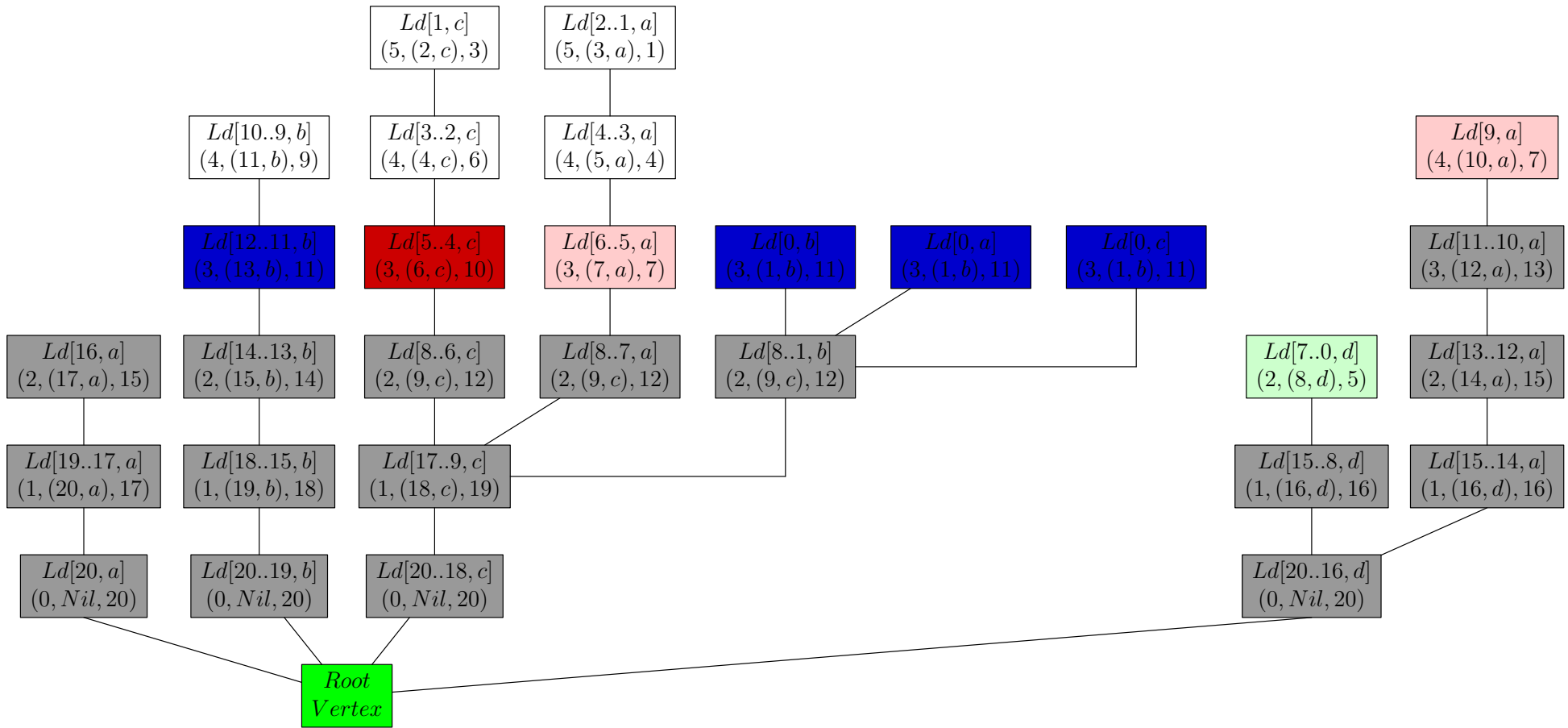
Just entered  $x_2[14] = a$ ,  $s \geq 13$ 's are known



$x_2 = b a c c a d c a b b c b c a b a d a b c$

\*

Just entered  $x_2[13] = c$ ,  $s \geq 12$ 's are known



$x_2 = b a c c a d c a b b c b c a b a d a b c$

\*

# Ours is a least common ancestor (LCA) problem

$$[x_1]_0 = [x_2]_0 = [\lambda]_0 = A^*$$

$$[x_1]_1 = [x_2]_1 = [abcd]_1$$

$$[x_1]_2 = [x_2]_2 = [abcdabcd]_2$$

$$[x_1]_3 = [x_2]_3 = [abcdabcdabc]_3$$

$$[x_1]_4 = [bacacdcababdacba]_4$$

$$[x_2]_4 = [bacacdabcabcdabc]_4$$

$$[x_1]_5 = [bacacacdcabababdacba]_5$$

$$[x_2]_5 = [baccadcabbccaabdabc]_5$$

$$[x_1]_6 = \{ x_1 \}$$

$$[x_2]_6 = [baccadcabbcbcabadabc]_6$$

$$[x_2]_7 = \{ x_2 \}$$

# A family of $O(|A| + |x|)$ algorithms: “the subword calculus”

For every  $x$  there exists a least  $t$ ,  $t \leq |x| + 1$ , such that  $[x]_t$  is a singleton:

$$A^* = [x]_0 \supset [x]_1 \supset [x]_2 \supset \cdots \supset [x]_m \supset \cdots \supset [x]_t = \{x\} = [x]_{t+1} = \cdots$$

For every  $x \neq 1$  there exists a largest  $s$ ,  $s < t$ , such that  $[x]_s$  is idempotent

For every  $m$ ,  $x$  contains a minimal word which is  $m$ -equivalent to it, these subwords can be represented by the following structure:

x_1=	b	c	a	c	a	c	a	d	c	a	b	a	b	a	b	d	a	c	b	a
	3	5	5	4	4	3	3	2	2	5	5	4	4	2	2	1	4	1	1	1
m=3	b					c	a	d	c				a	b	d		c	b	a	

Can this vector of  $m$ 's be computed in  $O(|A| + |x|)$  time?

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