

PSC 2015

# Faster Longest Common Extension on Compressed Strings and Applications

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# Collaborators

This work is a collaboration with:



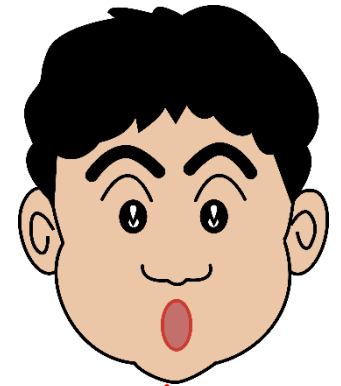
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# Longest common extension (LCE)

**Longest common extension (LCE)** on string  $T$  is a task such that, given two positions  $p$  and  $q$ , compute the length of the longest common substring of  $T$  starting at positions  $p$  and  $q$ .

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$q = 34$



I argue string algorithms at Prague stringology

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$$\mathbf{LCE}(6, 34) = 9$$

# Background & Motivation

- ✓ LCE has numerous applications, e.g., approximate pattern matching, computing palindromes, computing approximate repeats.
- ✓ A string  $T$  of length  $u$  can be preprocessed in  $O(u)$  time and space so that each LCE query can be answered in  $O(1)$  time [Demaine et al.].
- ✓ However, the  $O(u)$  complexity can be prohibitive for large-scaled text.
- ✓ To save preprocessing time and space, we consider **LCE on grammar-compressed text**.

# Straight Line Program (SLP)

## Definition

An SLP is a sequence of  $n$  productions

$$X_1 \rightarrow expr_1, X_2 \rightarrow expr_2, \dots, X_n \rightarrow expr_n$$

- $expr_i = a$  ( $a \in \Sigma$ )
- $expr_i = X_l X_r$  ( $l, r < i$ )

- ✓ An SLP is a CFG in the Chomsky normal form which derives a single string.
- ✓ SLPs model outputs of grammar-based compression algorithms (e.g., Re-pair, LZ78, LZDF, OLCA, etc).



# Straight Line Program (SLP)

$n$  : size (# of productions) of a given SLP  $S$

$h$  : height of the derivation tree of  $S$

$u$  : length of the uncompressed string  $T$   
represented by SLP  $S$

# Example of SLP

SLP  $S$

$X_1 \rightarrow a$

$X_2 \rightarrow b$

$X_3 \rightarrow X_1 X_1$

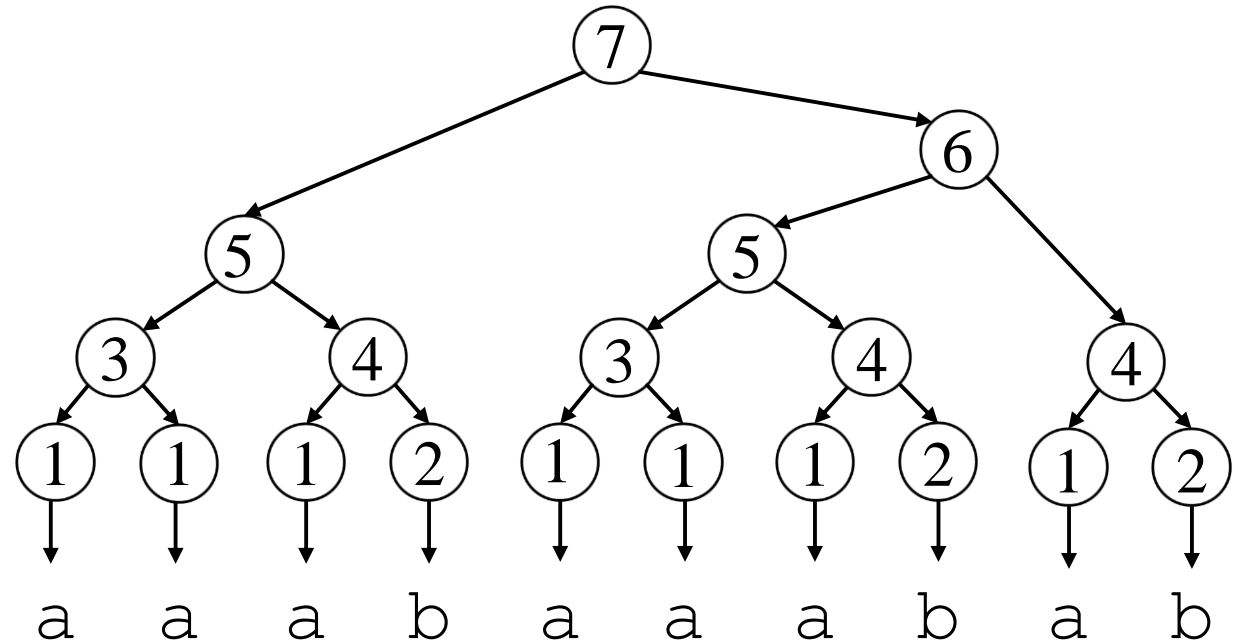
$X_4 \rightarrow X_1 X_2$

$X_5 \rightarrow X_3 X_4$

$X_6 \rightarrow X_5 X_4$

$X_7 \rightarrow X_5 X_6$

Derivation tree of SLP  $S$



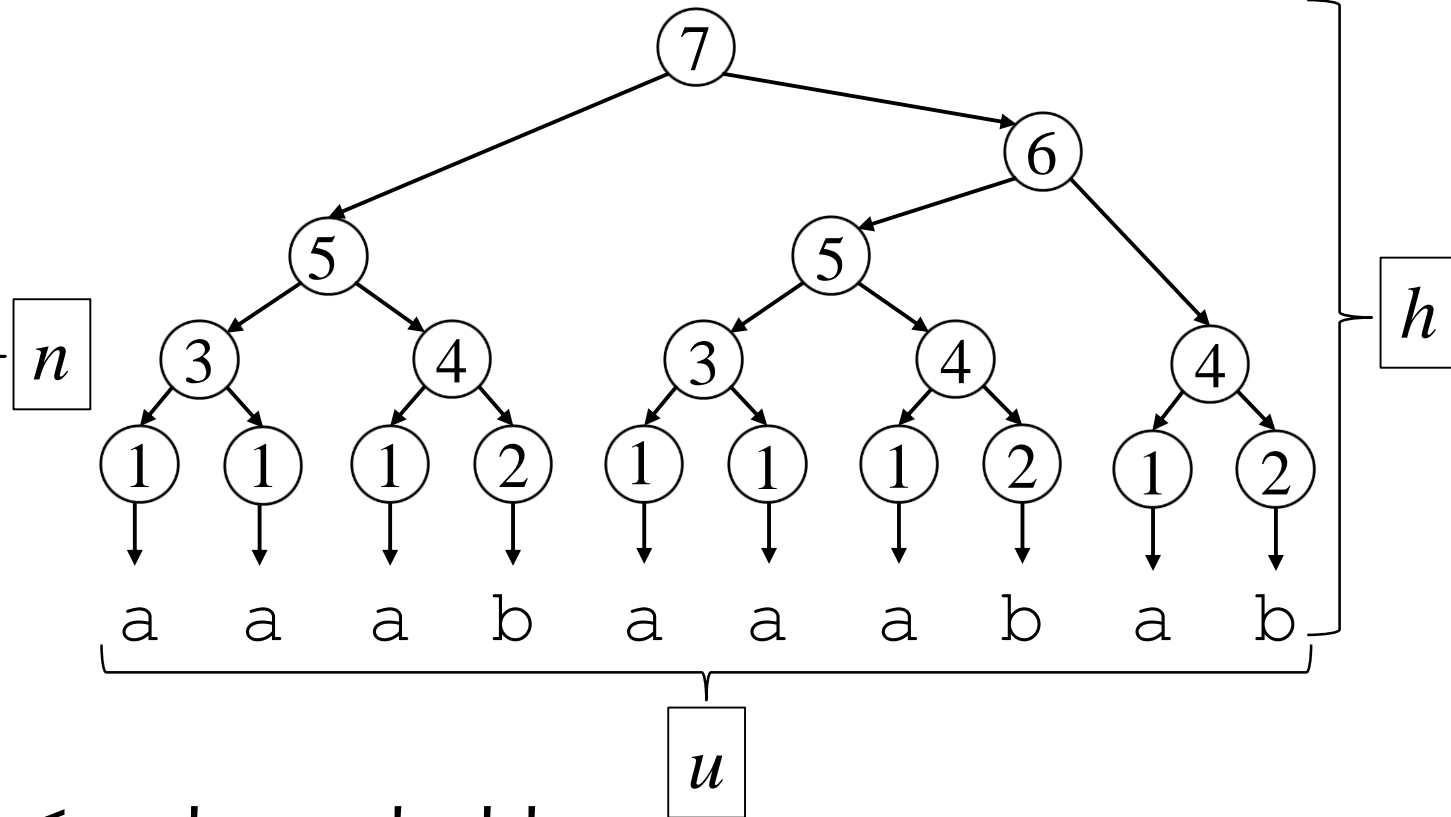


# Example of SLP

SLP  $S$

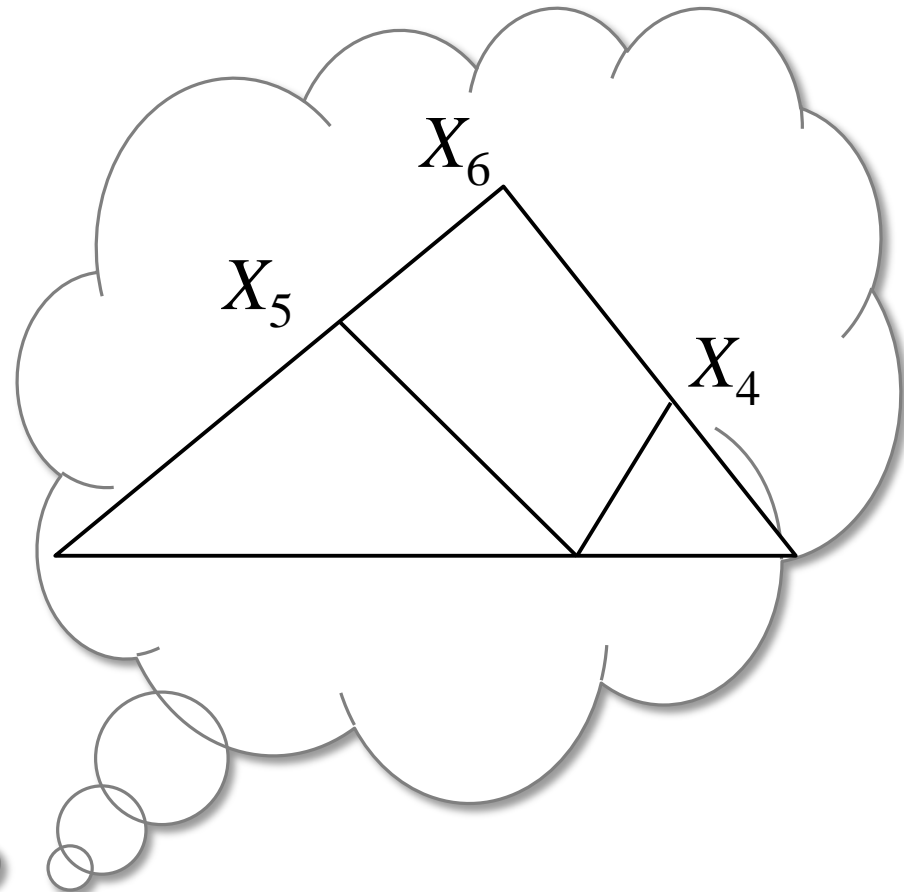
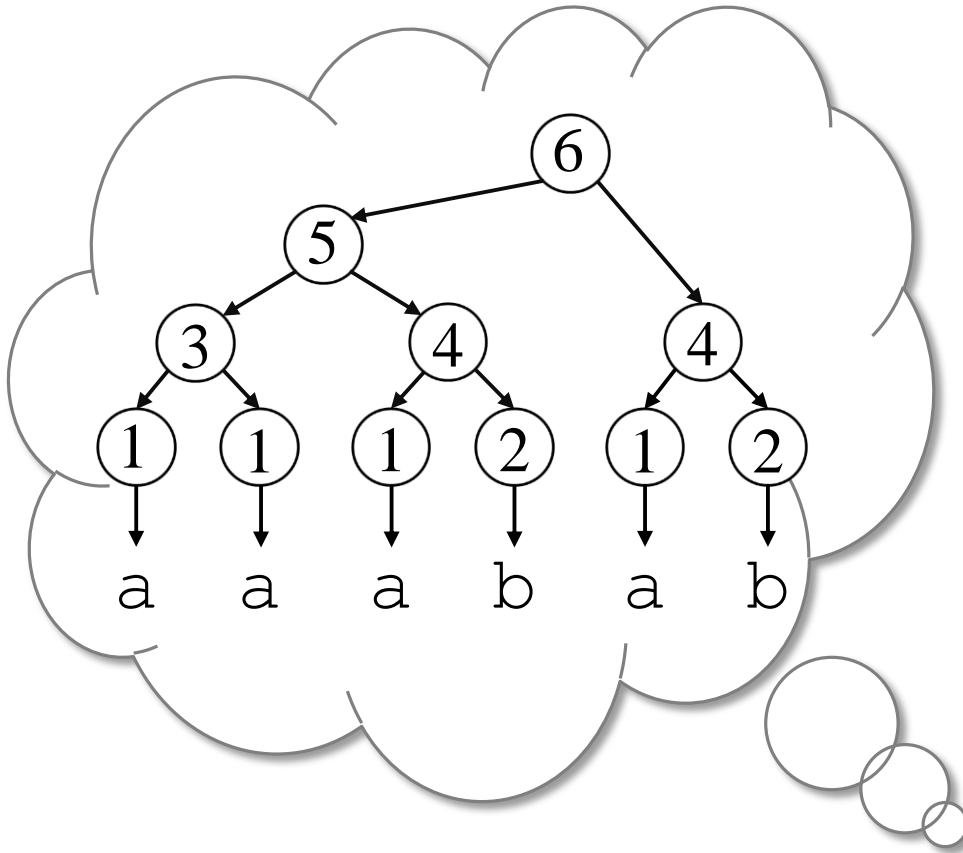
$X_1 \rightarrow a$   
 $X_2 \rightarrow b$   
 $X_3 \rightarrow X_1 X_1$   
 $X_4 \rightarrow X_1 X_2$   
 $X_5 \rightarrow X_3 X_4$   
 $X_6 \rightarrow X_5 X_4$   
 $X_7 \rightarrow X_5 X_6$

Derivation tree of SLP  $S$



- ✓  $\log_2 u \leq h \leq n$  always holds.
- ✓  $u$  can be exponential in  $n$  (e.g. consider string  $a^u$ ).
  - Hence,  $O(\text{poly}(n))$  solutions are of significance.

# Important Remarks

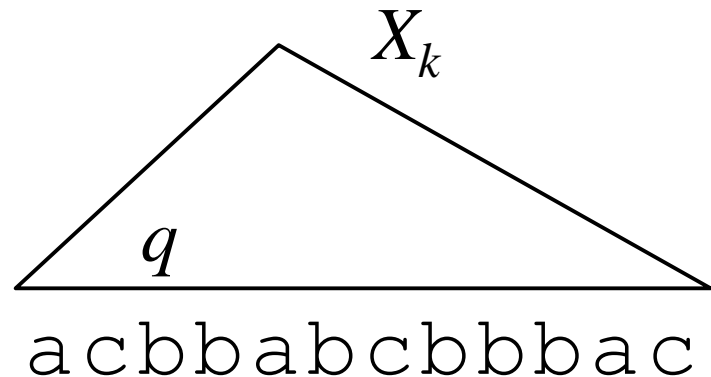
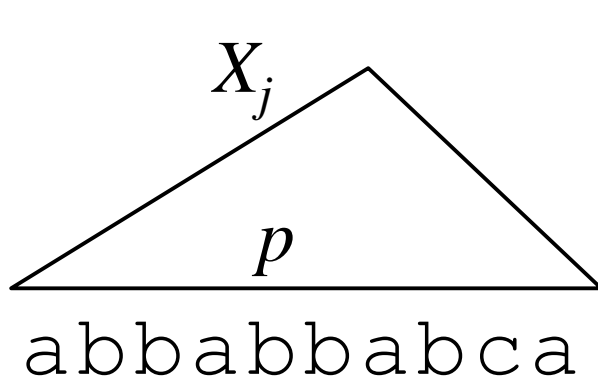


- ✓ Derivation trees are only **imaginary** (used only for explanations) and are never constructed explicitly.

# Longest Common Extension on SLP

## Problem 1 (grammar compressed LCE)

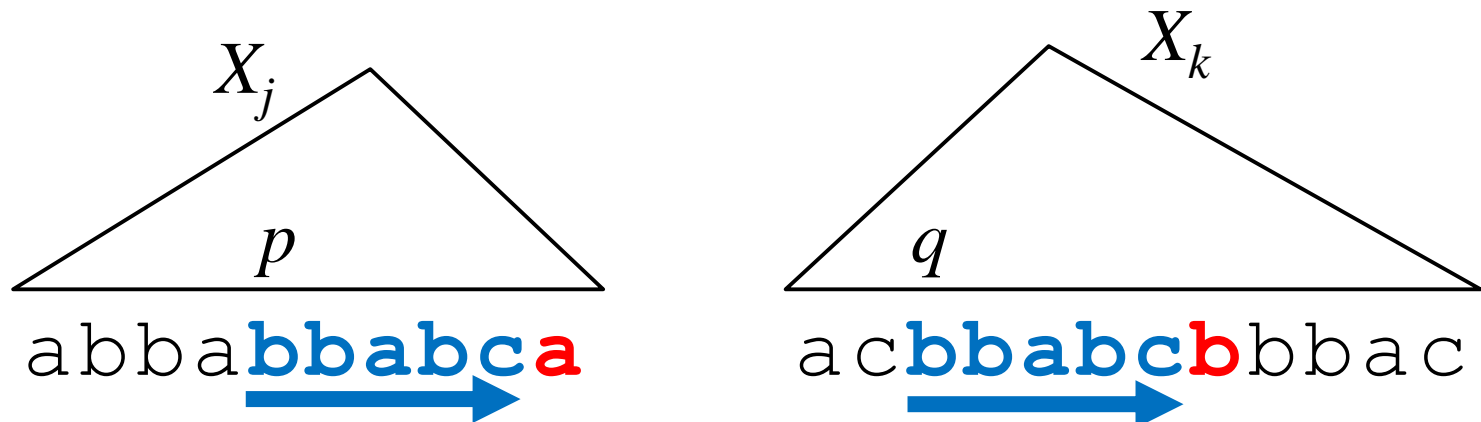
Preprocess an input SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  so that subsequent longest common extension queries  $\mathbf{LCE}(X_j, X_k, p, q)$  can be answered quickly.



# Longest Common Extension on SLP

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Preprocess an input SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  so that subsequent longest common extension queries  $\mathbf{LCE}(X_j, X_k, p, q)$  can be answered quickly.



Query output is LCE length 5

# What is the difficulty?

- ✓ We are not allowed to expand the SLP (compressed text), since this takes  $O(2^n)$  time in the worst case.
- ✓ But we want to know the length of the longest common extension!



# LCE algorithms on SLPs

Algorithms	Query time	Preprocessing time	Space
Folklore	$O(hL)$	$O(n)$	$O(n)$
(extended) Miyazaki et al. '97	$O(hn^2)$	$O(n^4)$	$O(n^2)$
(extended) Lifshits '07	$O(hn^2)$	$O(hn^2)$	$O(n^2)$
I et al. '15	$O(h \log u)$	$O(hn^2)$	$O(n^2)$
Bille et al. '15 (randomized)	$O(\log u + \log^2 L)$	N/A	$O(n)$

$n$ : size of SLP

$u$ : length of uncompressed string  $T$

$h$ : height of SLP derivation tree

$L$ : LCE length (output)

$z$ : size of LZ77 factorization of  $T$

- $\log u \leq h \leq n$
- $L = O(u)$
- $\log^* u = o(\log u)$
- $z \leq n$  (due to Rytter '03)

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This work	$O(\log u + \log^* u \log L)$	$O(n \log \log n \log^* u \log u)$	$O(n + z \log^* u \log u)$

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# Logstar (iterated logarithm)

## Definition

The **logstar** of a positive integer  $u$ , denoted  $\log^* u$ , is the number of times the logarithm function needs to be iteratively applied to  $u$  until the result becomes less than or equal to 1.

- ✓ The logstar is a very slowly growing function, e.g.,  $\log^* 2^{65536} = 5$ .

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$n$ : size of SLP

**Fastest**  
deterministic  
queries

**Fastest**  
preprocessing

**Smallest**  
in many cases

$z$ : size of...ization of  $L$

- $\log u \leq h \leq \dots$
- $z \leq n$  (due to...)

# Our strategy

- ✓ All previous algorithms work on the SLP derivation trees of two query non-terminals.
- ✓ Our new algorithm does NOT work on the SLP derivation trees.
- ✓ Instead, we construct a different tree of logarithmic height, based on
  - locally consistent parsing
  - signature encoding.

# Locally consistent parsing

Lemma 1 [Mehlhorn et al., Alstrup et al.]

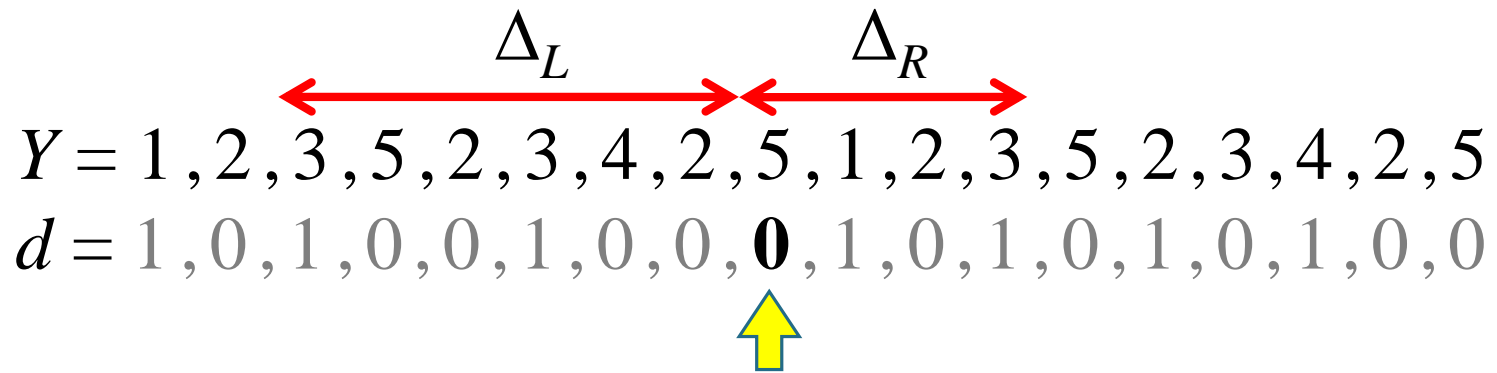
For any integer string  $Y \in \{1..m\}^*$  in which no adjacent elements are equal (i.e.  $Y[i] \neq Y[i+1]$ ), there is a bit string  $d$  of length  $|Y|$  such that

1. no 1's appear consecutively;
2. at most three 0's appear consecutively;
3. each  $d[i]$  is determined locally, i.e., by  $Y[i-\Delta_L \dots i-1]$  and  $Y[i \dots i+\Delta_R]$ , where  $\Delta_L \leq \log^* m + 6$  and  $\Delta_R \leq 4$ ;
4.  $d$  can be computed in  $O(|Y|)$  time.

# Locally consistent parsing

$Y = 1, 2, 3, 5, 2, 3, 4, 2, 5, 1, 2, 3, 5, 2, 3, 4, 2, 5$   
 $d = 1, 0, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0, 1, 0, 0$

# Locally consistent parsing



$$\Delta_L \leq \log^* m + 6$$

$$\Delta_R \leq 4$$



# Locally consistent parsing

$$Y = \boxed{1,2}, \boxed{3,5,2}, \boxed{3,4,2,5}, \boxed{1,2}, \boxed{3,5}, \boxed{2,3}, \boxed{4,2,5}$$
$$d = 1,0,1,0,0,1,0,0,0,1,0,1,0,1,0,1,0,0$$

- ✓ Using the bit string  $d$ , any integer string  $Y$  can be uniquely decomposed in linear time into blocks of length 2-4.

# Signature encoding [Mehlhorn et al. '97]

- ✓ Iteratively apply locally consistent parsing to input string  $T$  until a single integer is obtained.

$T = a b c a c a b b c a b a c c c a$

# Signature encoding [Mehlhorn et al. '97]

- ✓ Iteratively apply locally consistent parsing to input string  $T$  until a single integer is obtained.

Each character is assigned to a unique integer called a signature.

	1	2	3	1	3	1			3	1	2	1				1	
$T =$	a	b	c	a	c	a	b	b	c	a	b	a	c	c	c	a	

# Signature encoding [Mehlhorn et al. '97]

- ✓ Iteratively apply locally consistent parsing to input string  $T$  until a single integer is obtained.

Maximal run of the same signatures is assigned to a new signature.

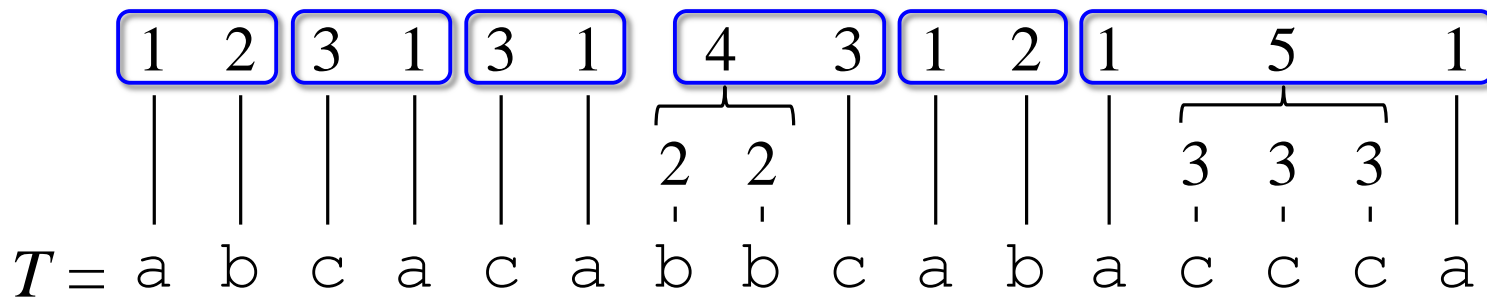
$T =$  a b c a c a b b c a b a c c c a

1	2	3	1	3	1	4	3	1	2	1	5	1			
						┌───┐					┌───┐				
						2	2				3	3	3		
a	b	c	a	c	a	b	b	c	a	b	a	c	c	c	a

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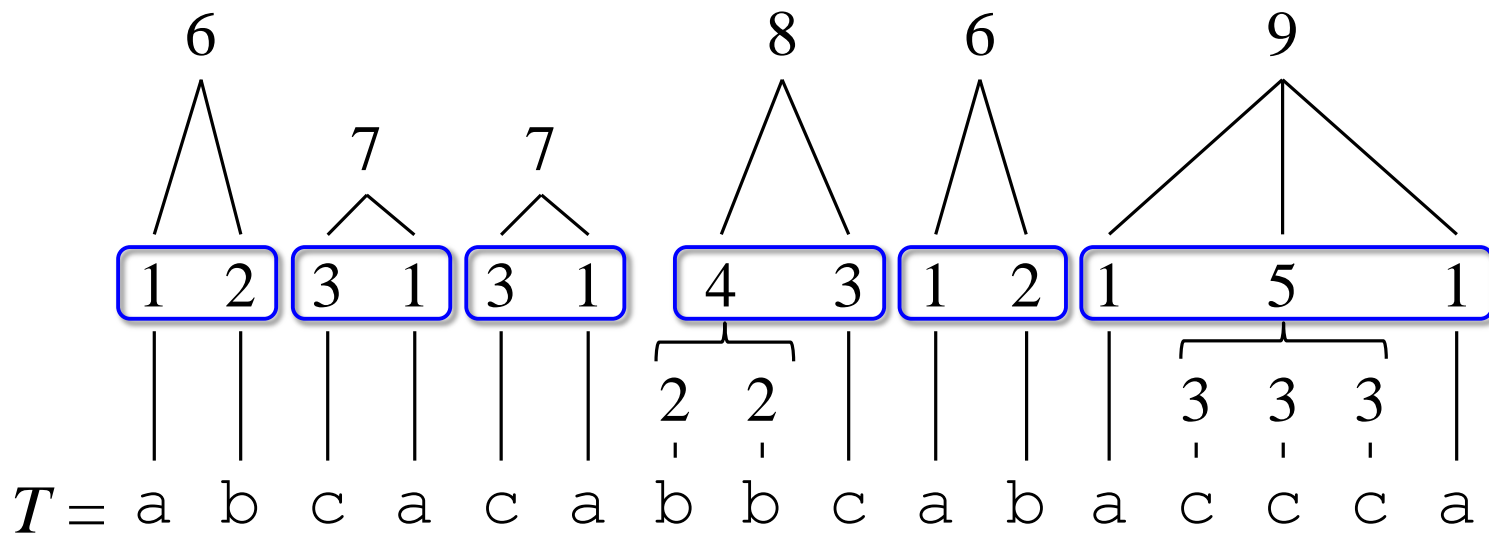
Apply locally consistent parsing to this string.



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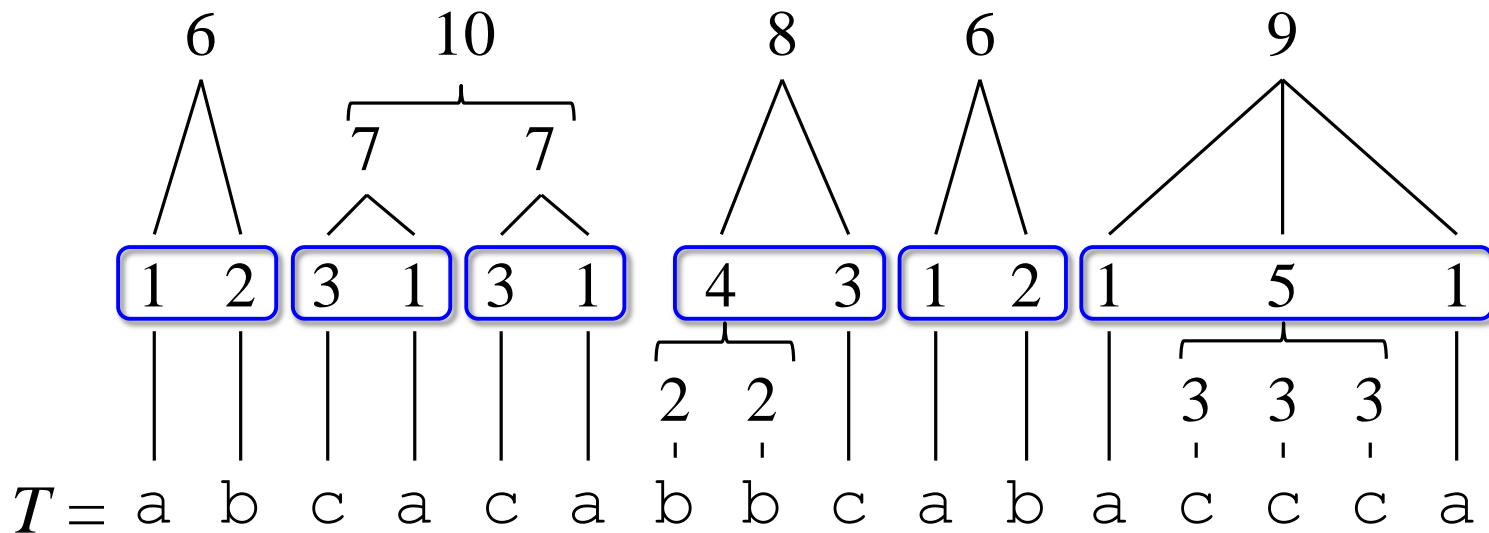
Each block is assigned to a new signature.



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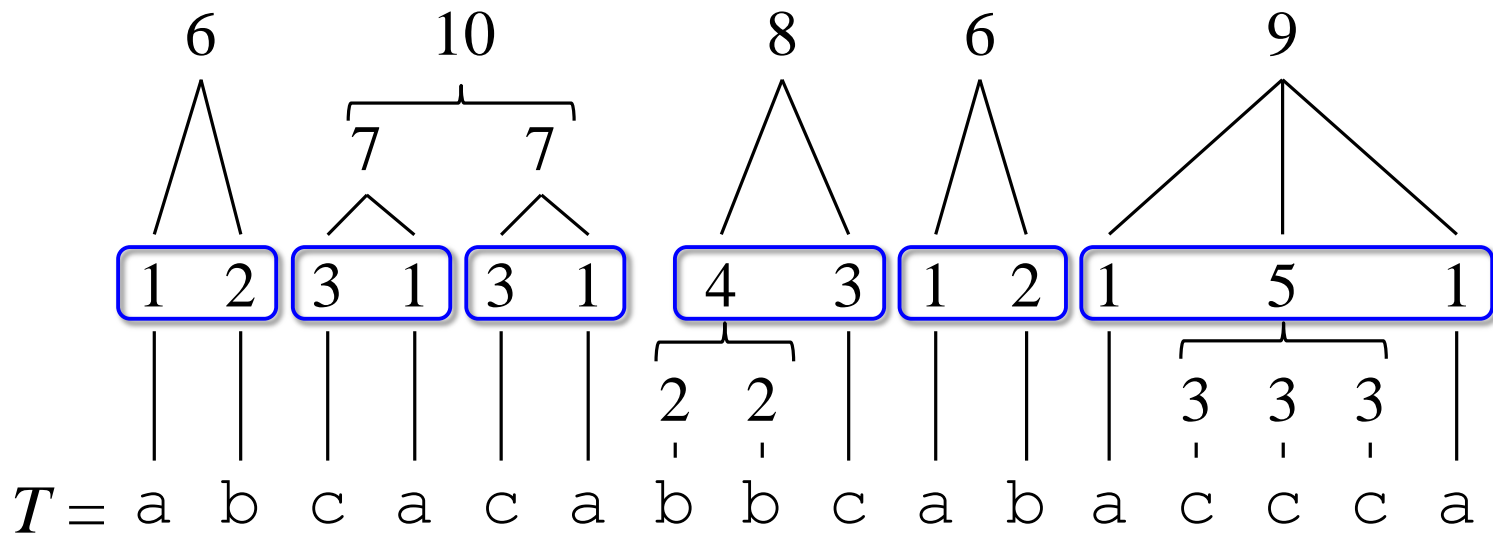
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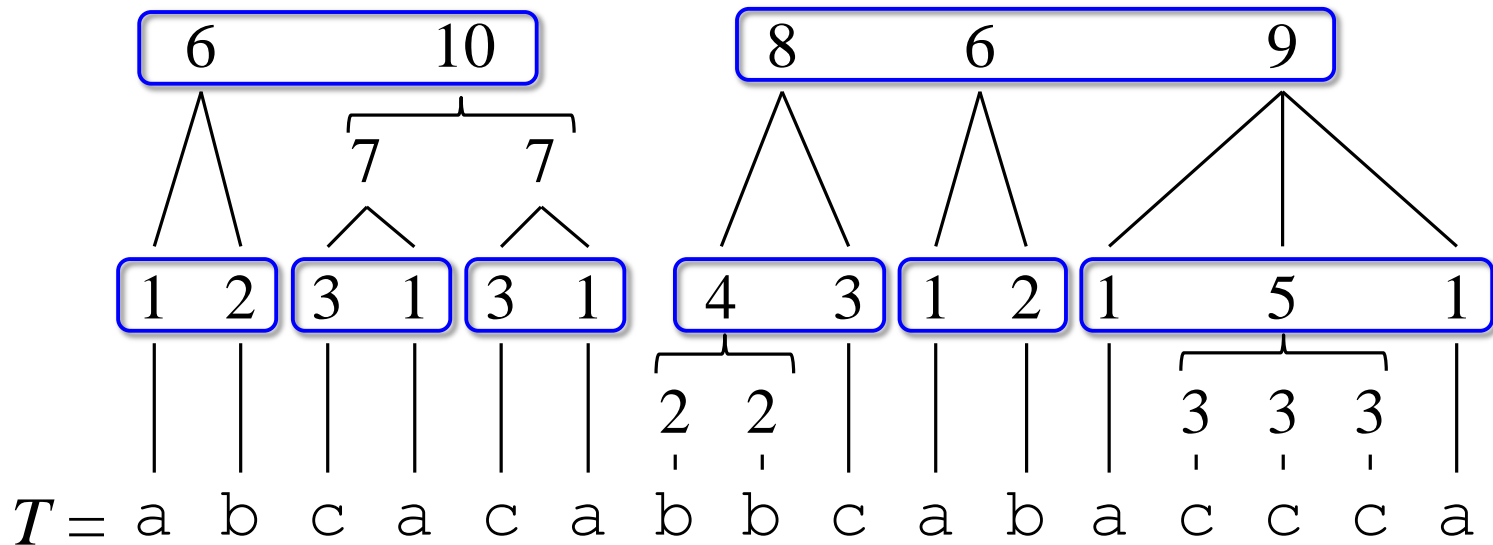
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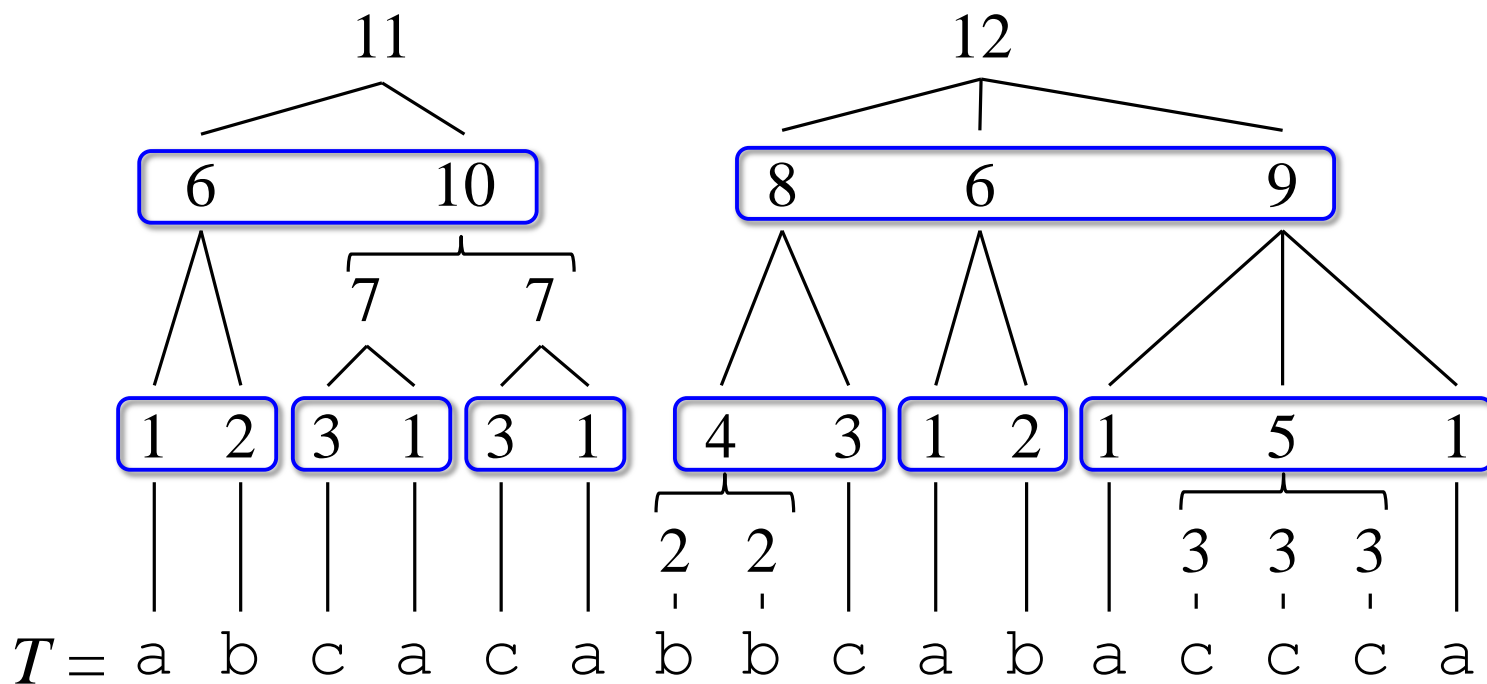
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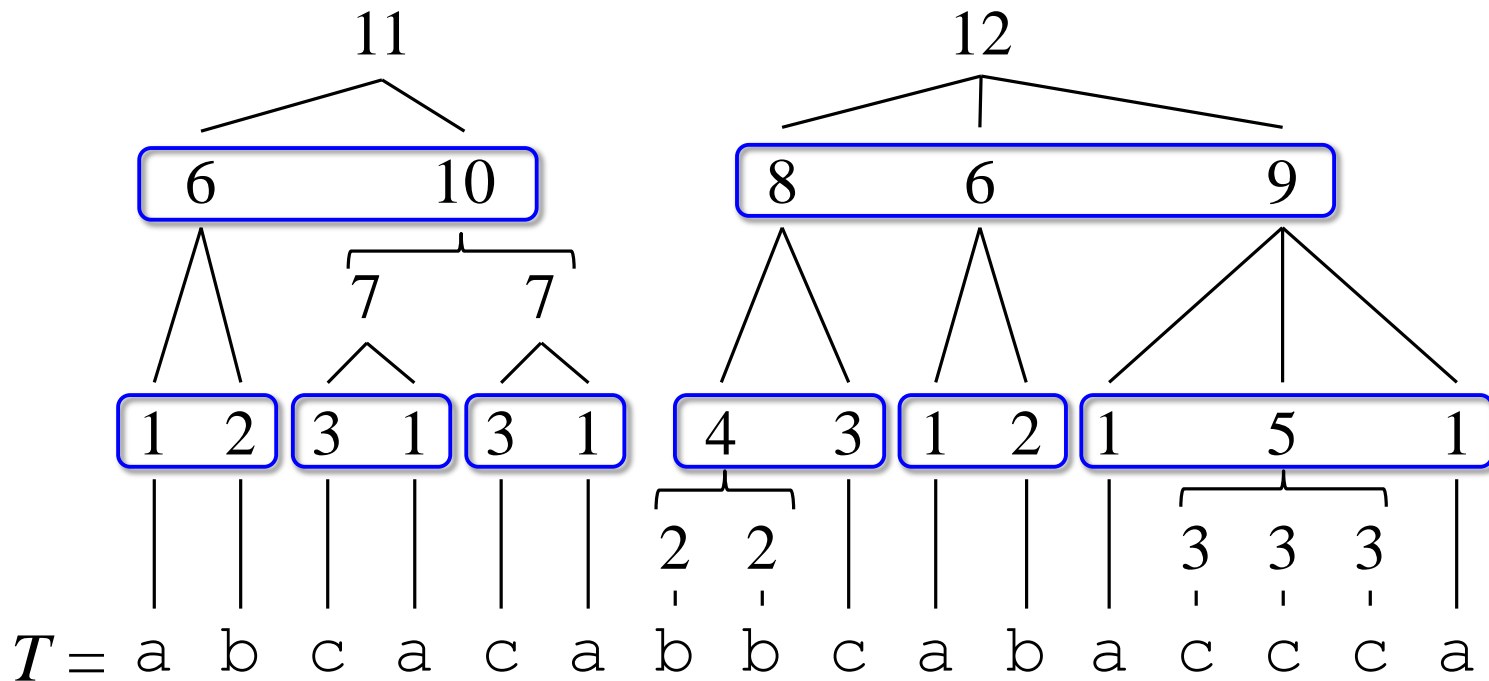
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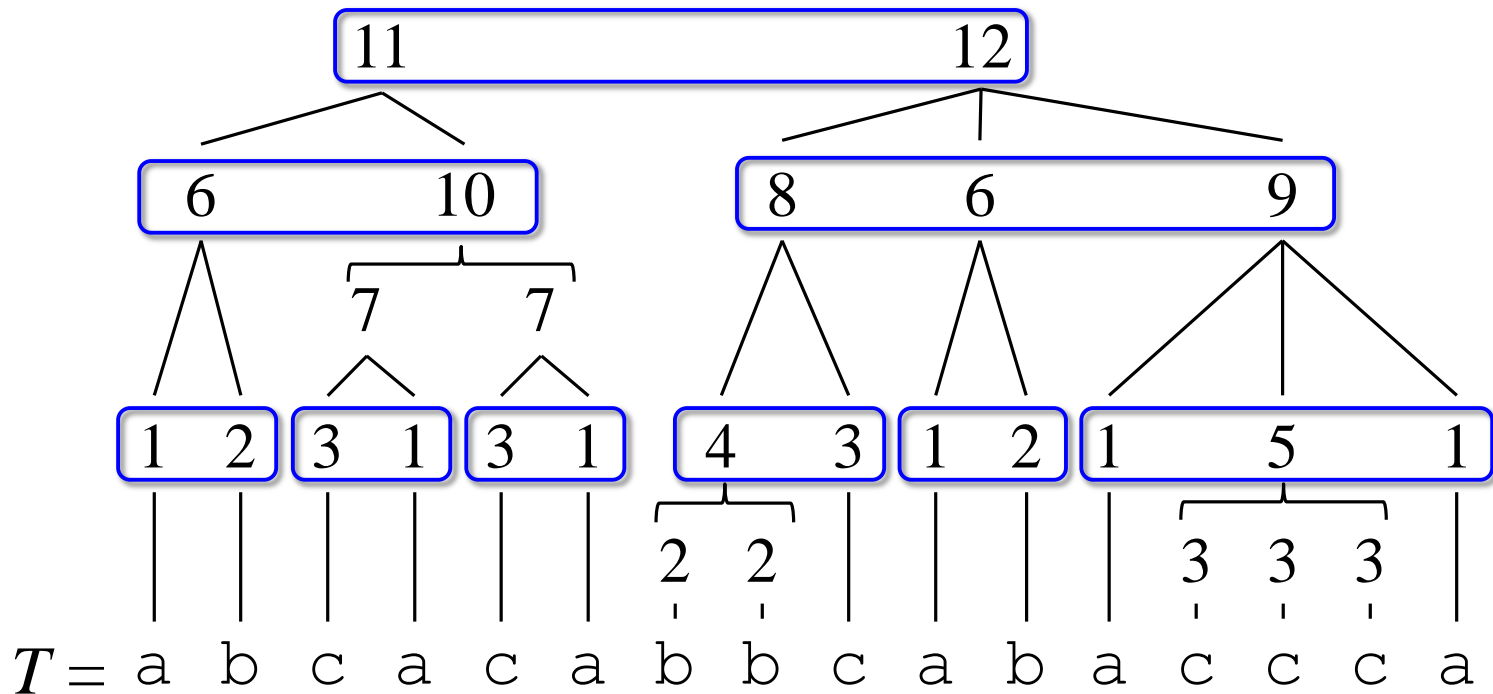
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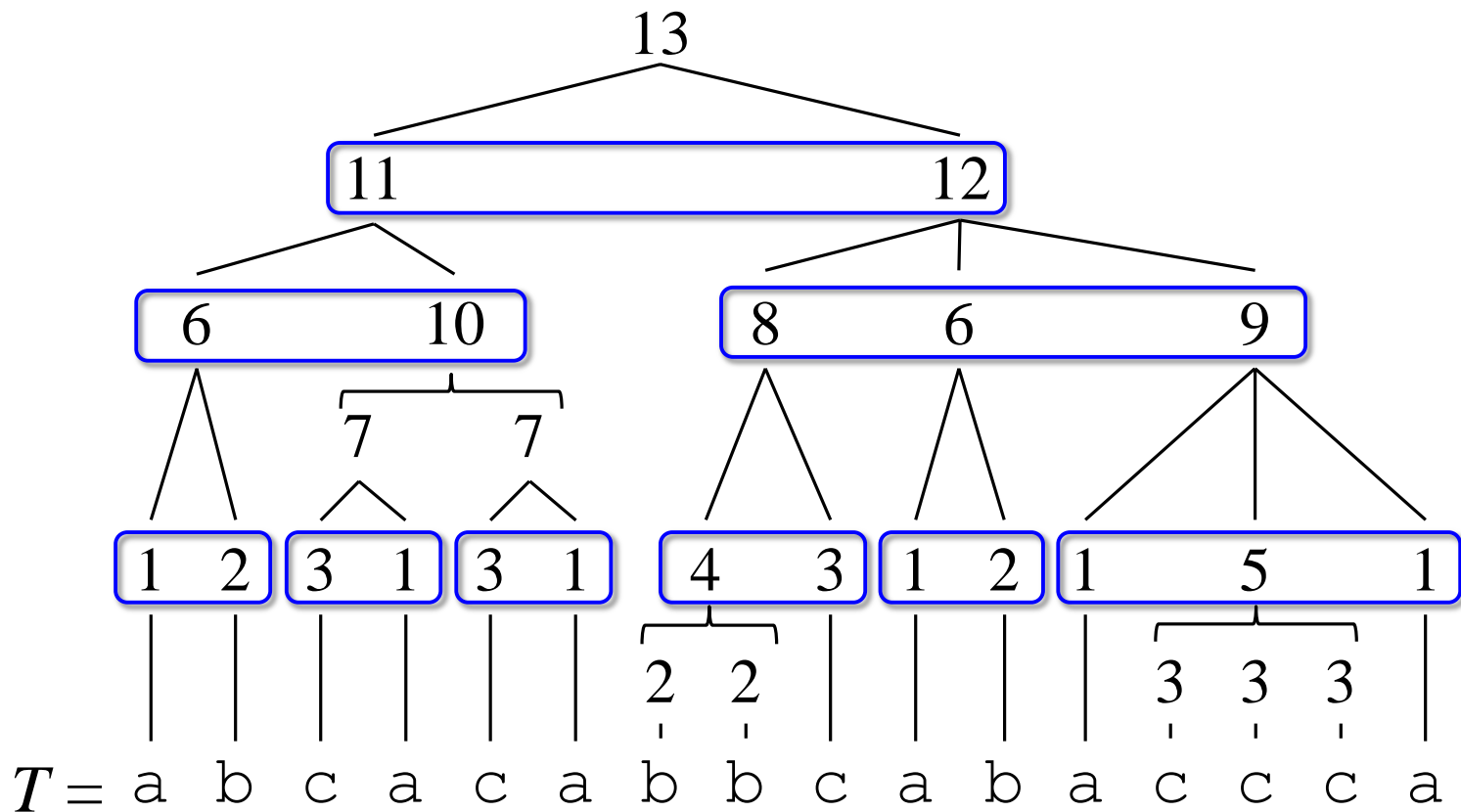
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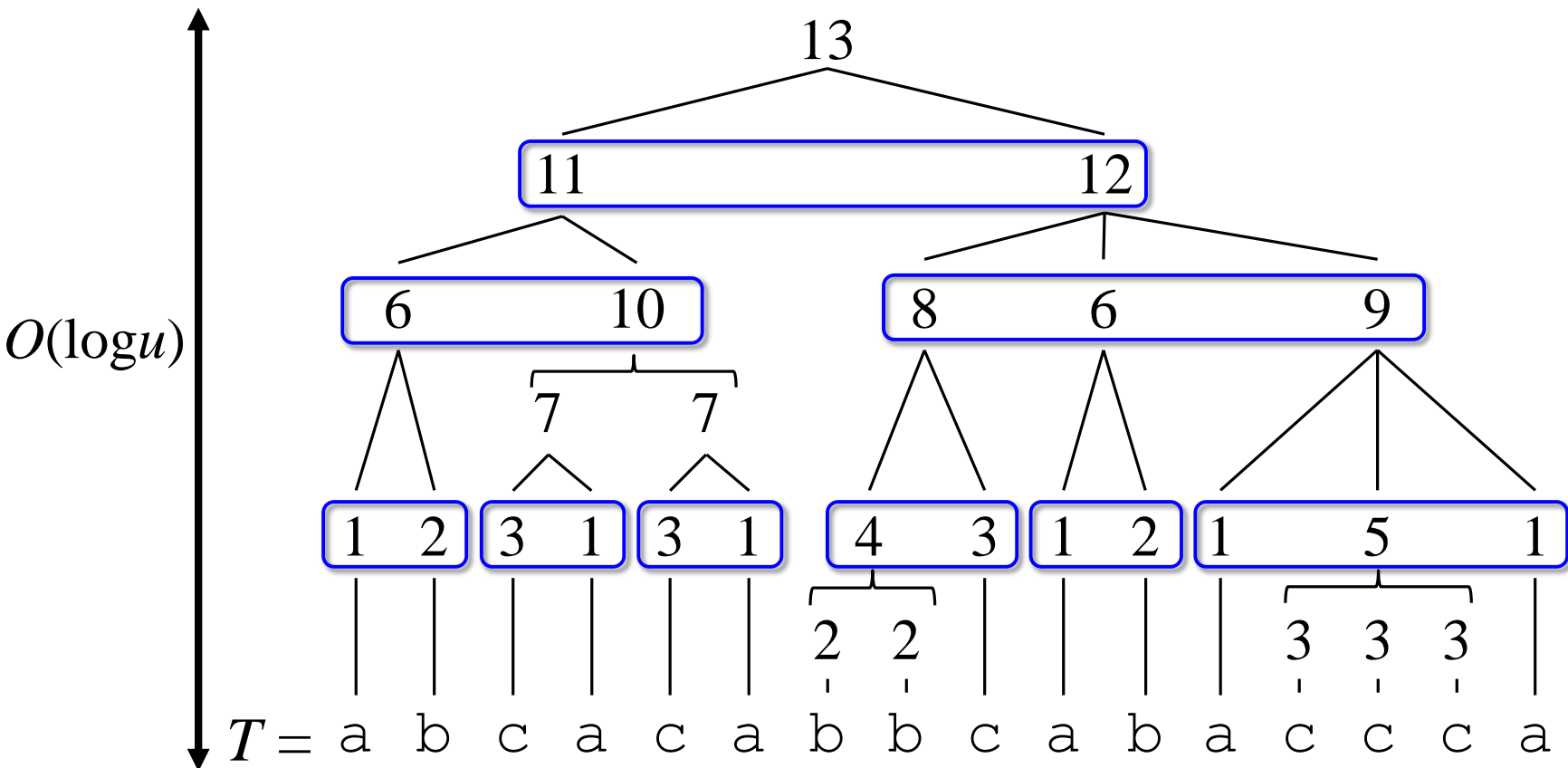
# Signature encoding [Mehlhorn et al. '97]

- ✓ Iteratively apply locally consistent parsing to input string  $T$  until a single integer is obtained.



# Signature encoding [Mehlhorn et al. '97]

- ✓ The height of this tree, called the signature tree, is  $O(\log u)$ , where  $u = |T|$ .

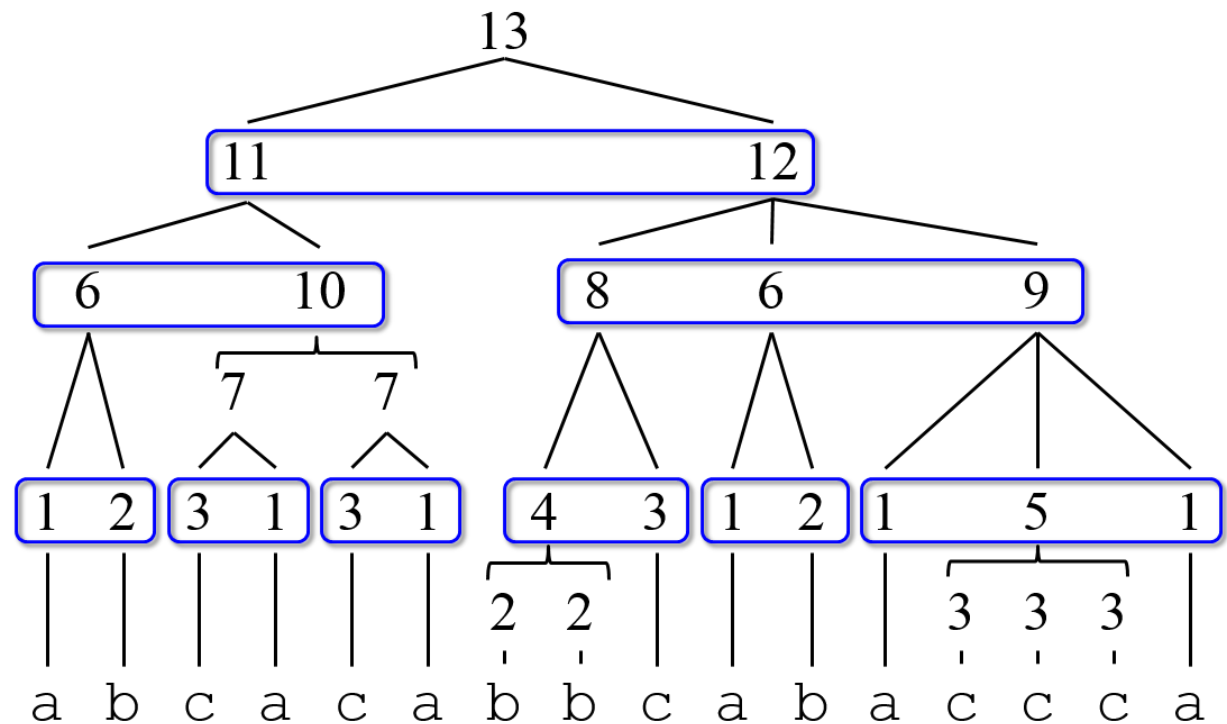


# Signature encoding [Mehlhorn et al. '97]

- ✓ The dictionary  $D_T$  of signatures is the **signature encoding** of input string  $T$ .

$D_T$	$13 \rightarrow 11, 12$
	$12 \rightarrow 8, 6, 9$
	$11 \rightarrow 6, 10$
	$10 \rightarrow 7^2$
	$9 \rightarrow 1, 5, 1$
	$8 \rightarrow 4, 3$
	$7 \rightarrow 3, 1$
	$6 \rightarrow 1, 2$
	$5 \rightarrow 3^3$
	$4 \rightarrow 2^2$
	$3 \rightarrow c$
	$2 \rightarrow b$
	$1 \rightarrow a$

signature tree of  $T$



# Faster LCE algorithm on SLP

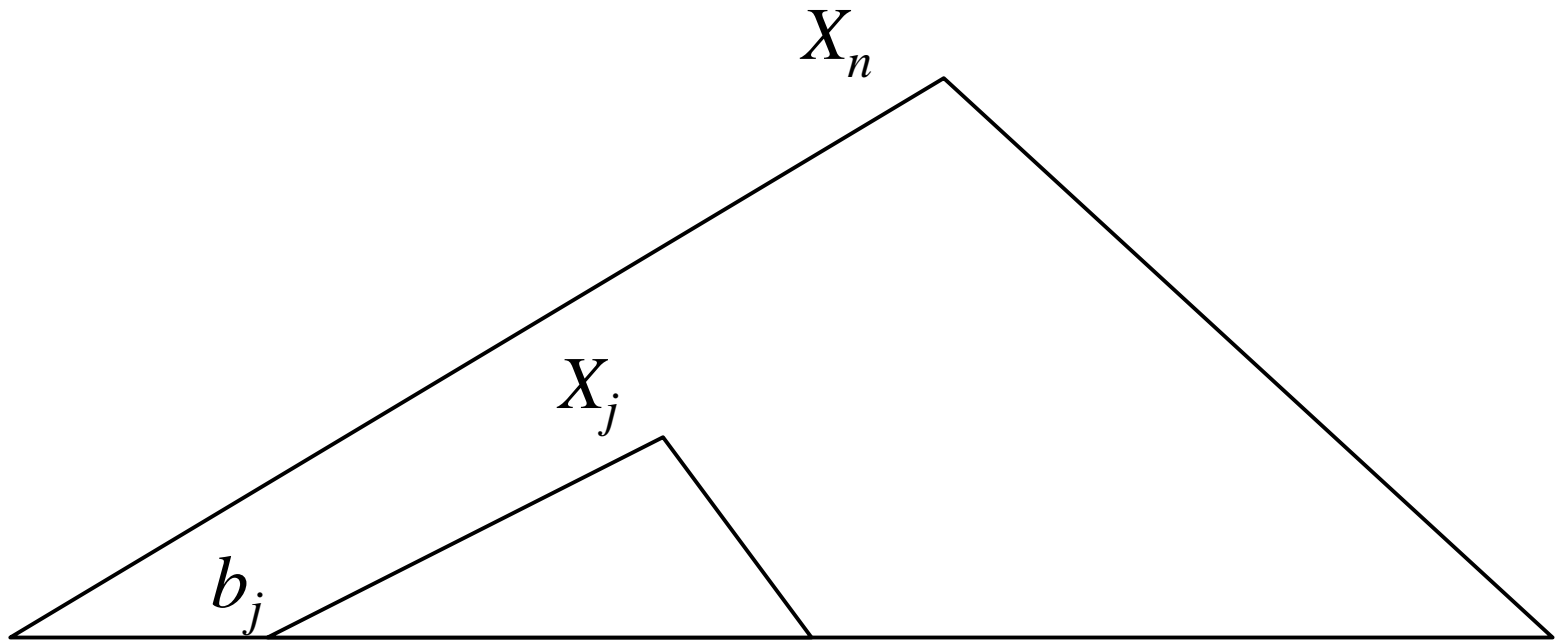
## Lemma 2 (Faster LCE on SLP)

Given the signature encoding  $D_T$  of string  $T$  of length  $u$ , we can compute  $\mathbf{LCE}(X_j, X_k, p, q)$  for any variables  $X_j, X_k$  and positions  $p, q$  in  $O(\log u + \log^* u \log L)$  time, where  $L$  is the answer to the query (LCE length).



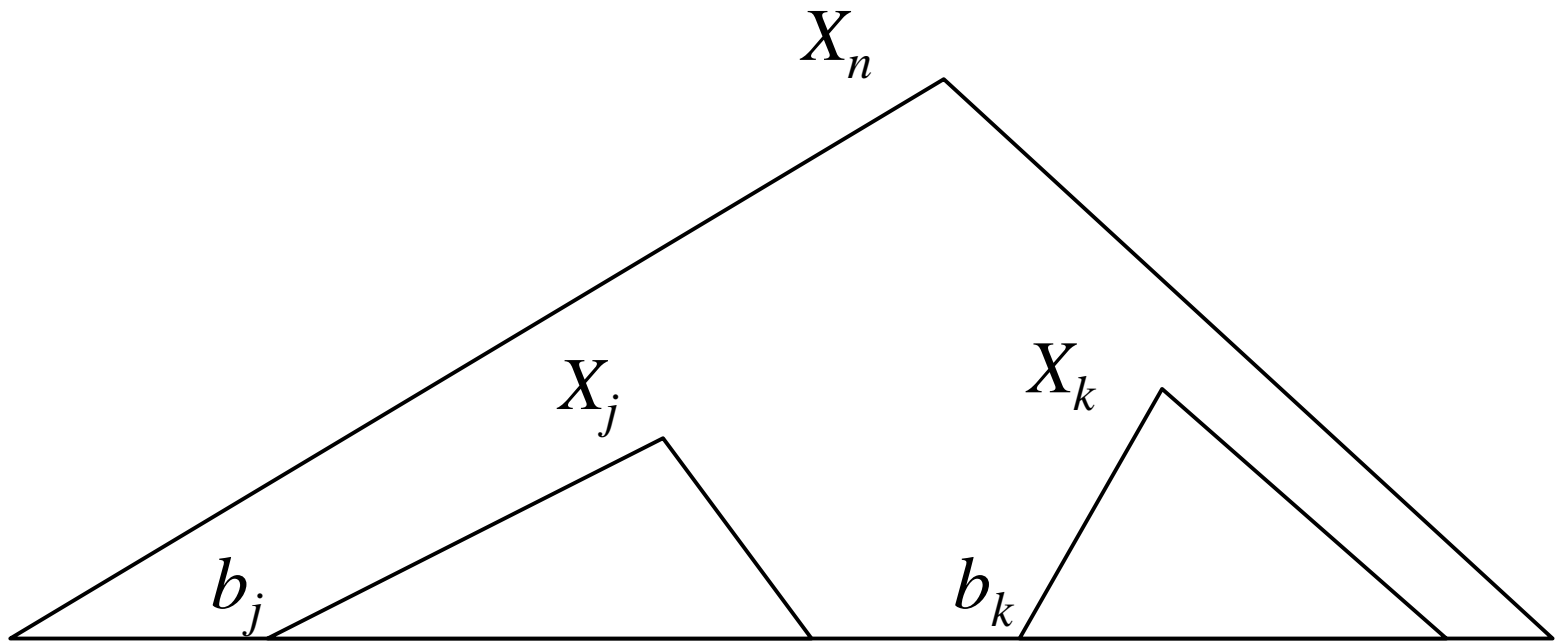
# Faster LCE algorithm on SLP

1. For every non-terminal  $X_j$ , we precompute and store its occurrence  $b_j$  in the derivation tree of  $X_n$ .



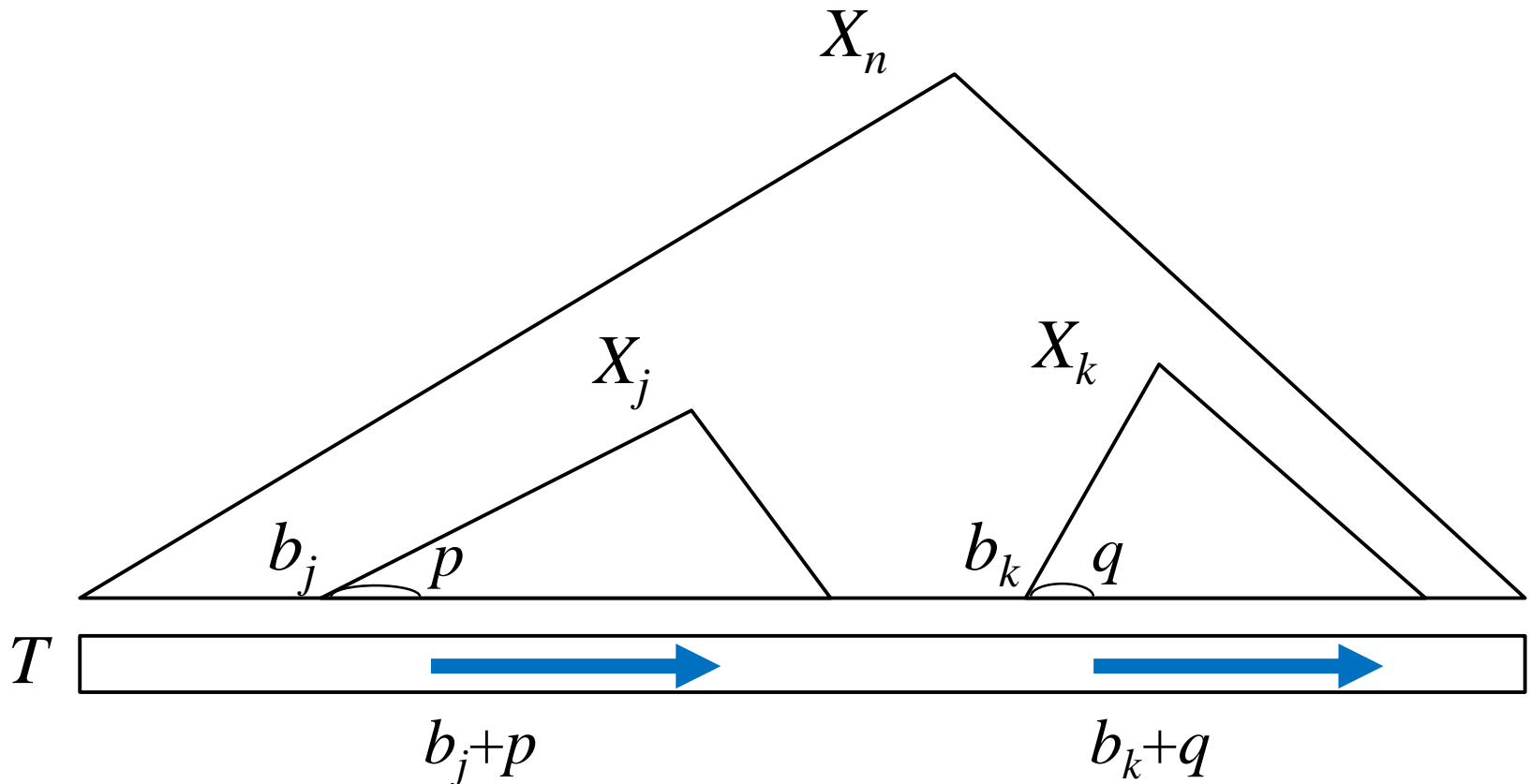
# Faster LCE algorithm on SLP

- Given query variables  $X_j$  and  $X_k$  for LCE, we retrieve  $b_j$  and  $b_k$ .



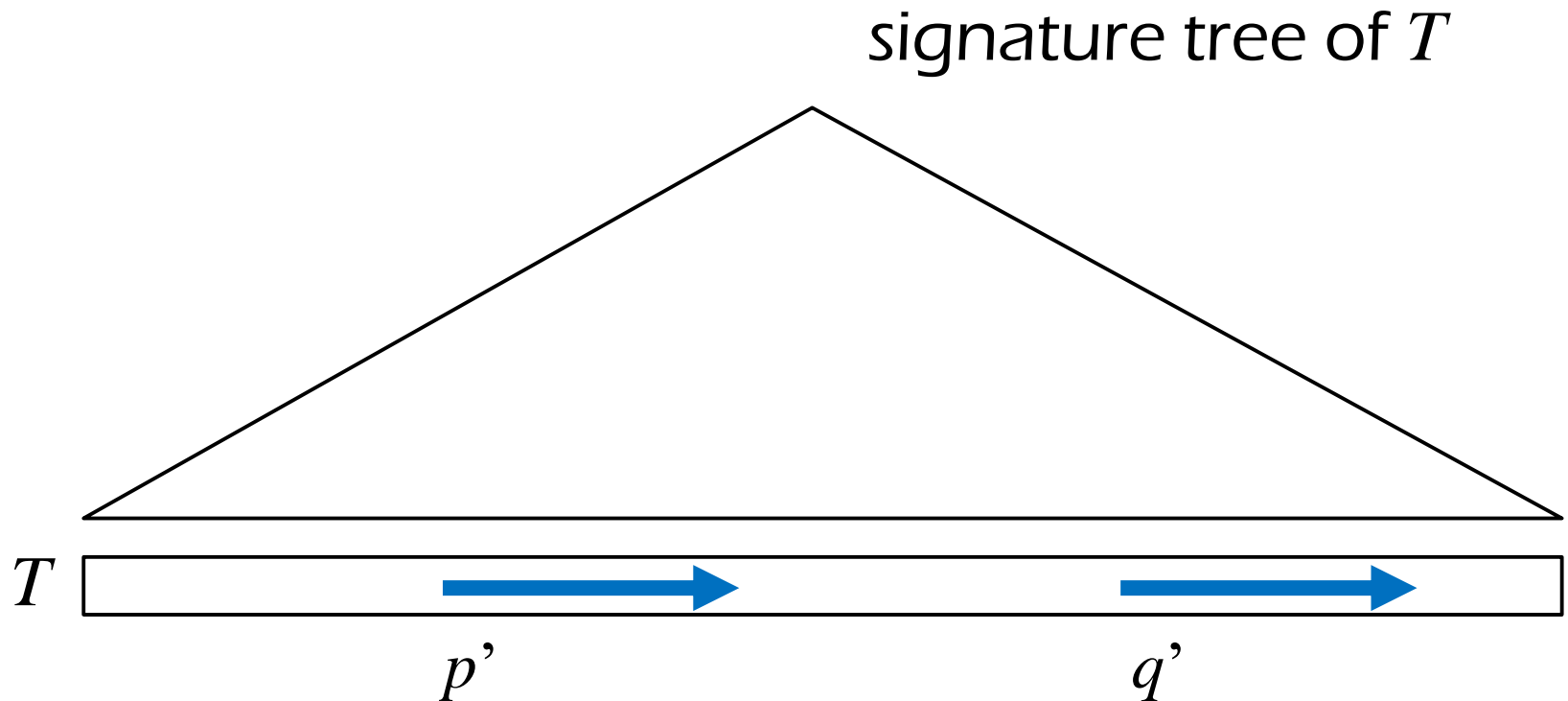
# Faster LCE algorithm on SLP

3. Since the last variable  $X_n$  derives string  $T$ ,  $\mathbf{LCE}(X_j, X_k, p, q)$  reduces to  $\mathbf{LCE}(b_{j+p}, b_{k+q})$  on string  $T$ .



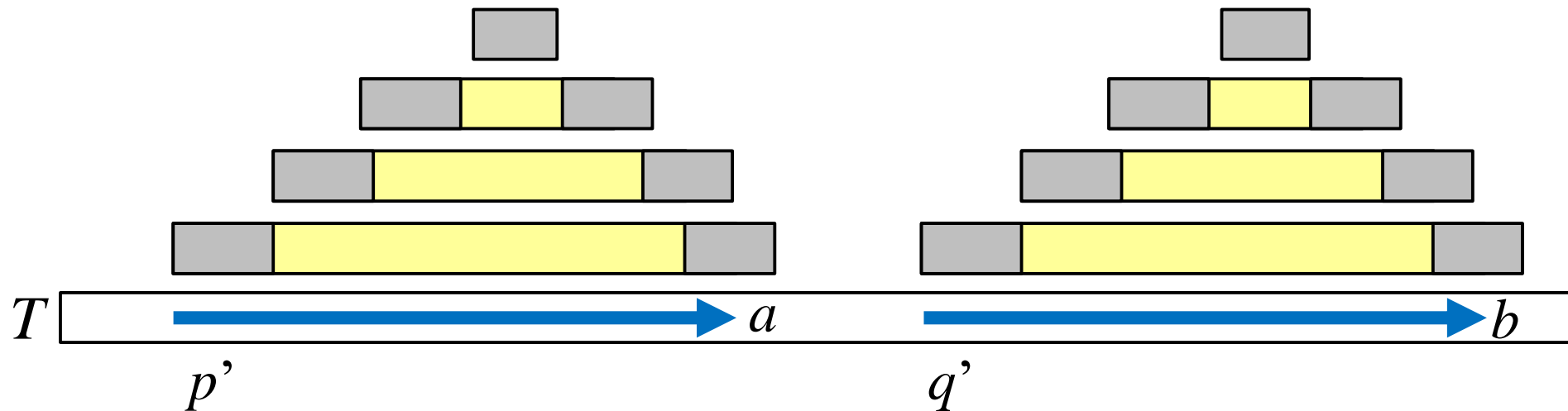
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4. We turn attention to the signature tree of  $T$ , and compute  $\mathbf{LCE}(p', q')$  there, where  $p' = b_j + p$  and  $q' = b_k + q$ .



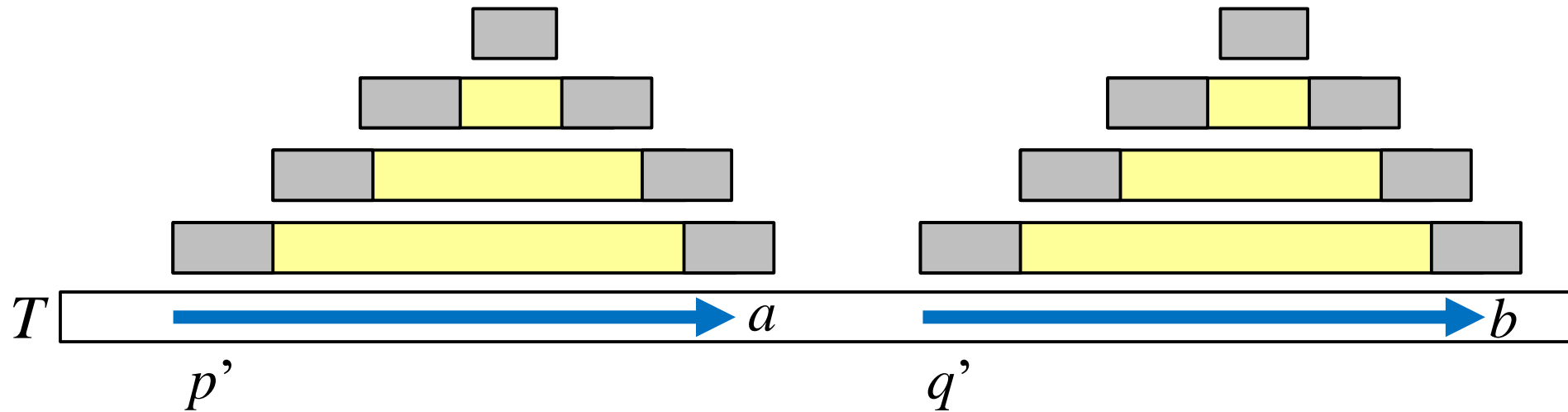
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5. By the property of signature encoding, at each level of the signature tree, there must be a common sequence of signatures for  $\mathbf{LCE}(p', q')$  (yellow parts).



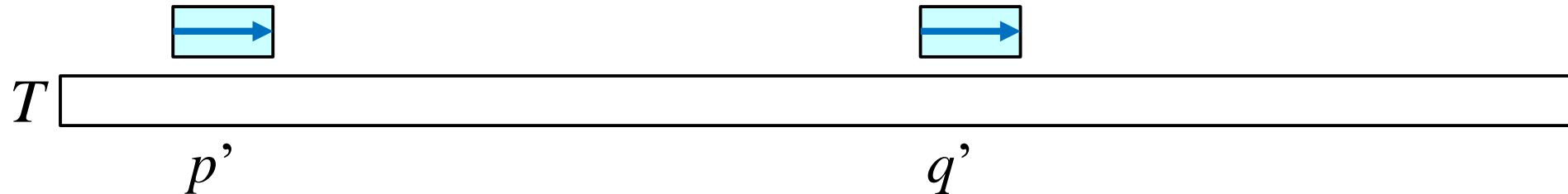
# Faster LCE algorithm on SLP

5. [Cont.] The left boundaries of length  $\Delta_L + O(1)$  may or may not be equal depending on the left contexts at each level, while the right boundaries of length  $\Delta_R + O(1)$  always have a mismatch.



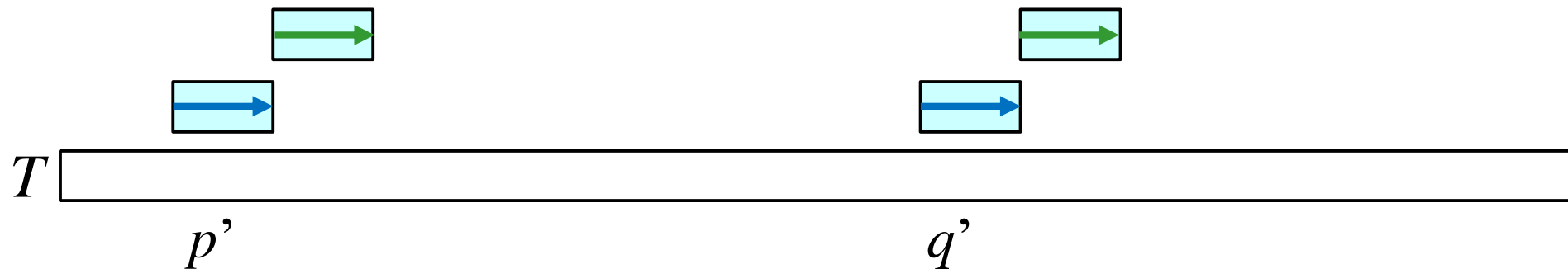
# Faster LCE algorithm on SLP

6. In a bottom-up manner, we re-compute the left boundary signatures of length  $\Delta_L + O(1)$  ignoring their left contexts, and compare them until we find a mismatch.



# Faster LCE algorithm on SLP

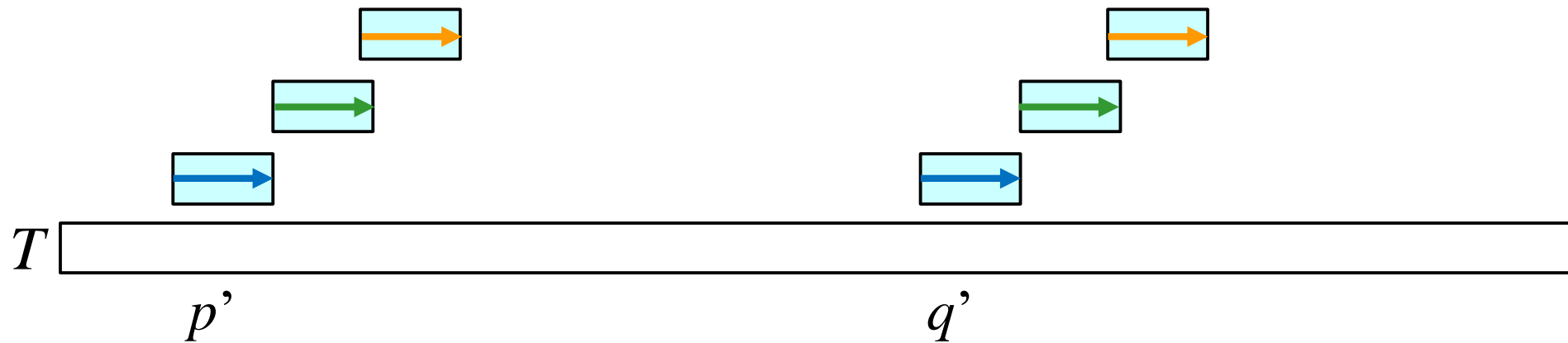
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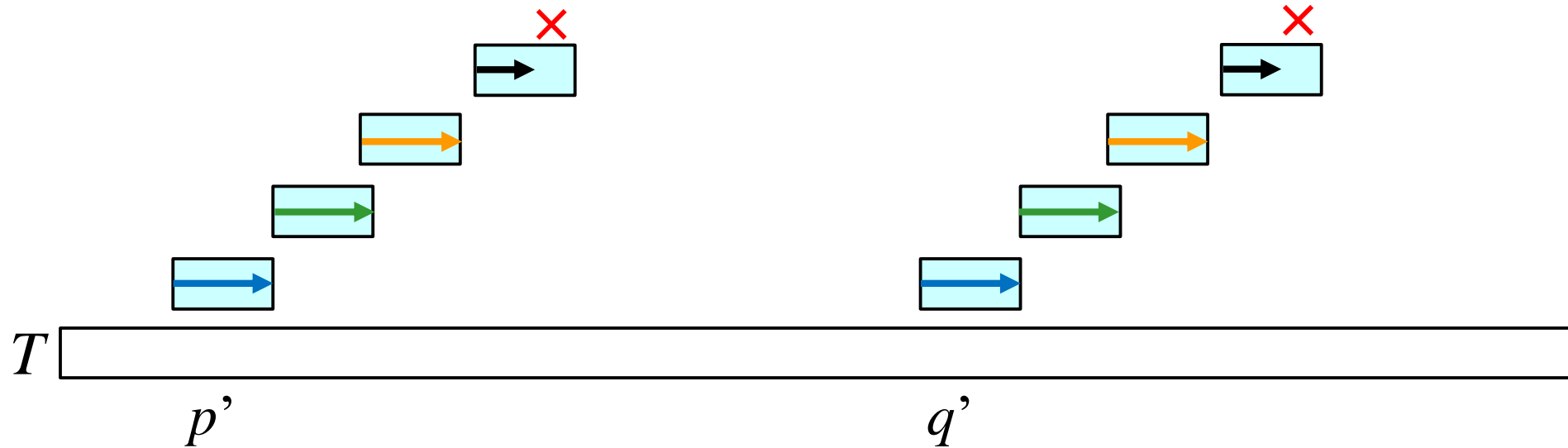
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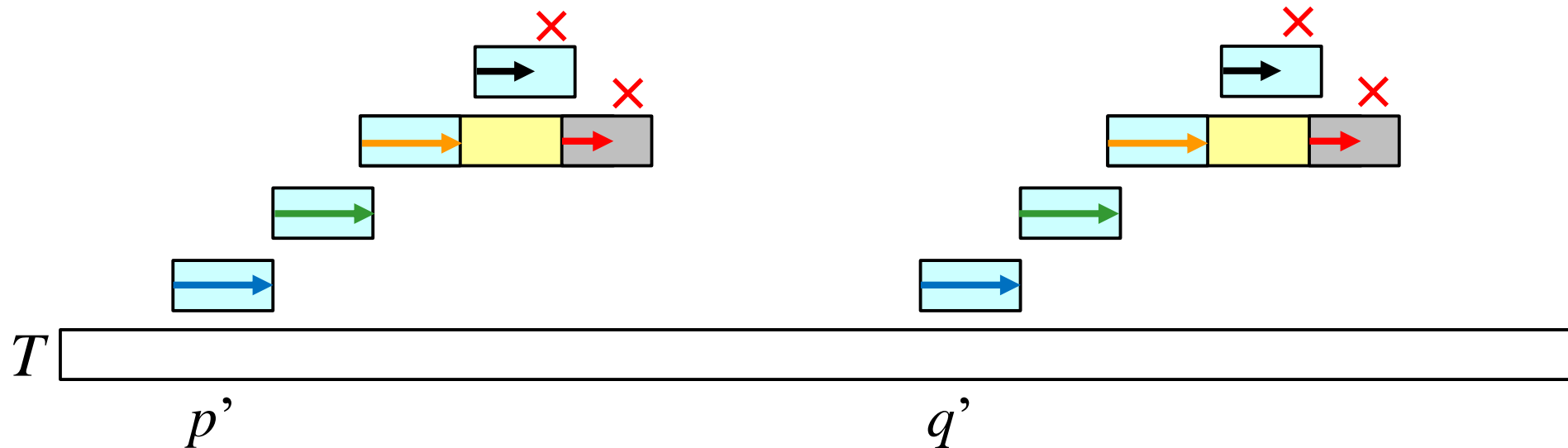
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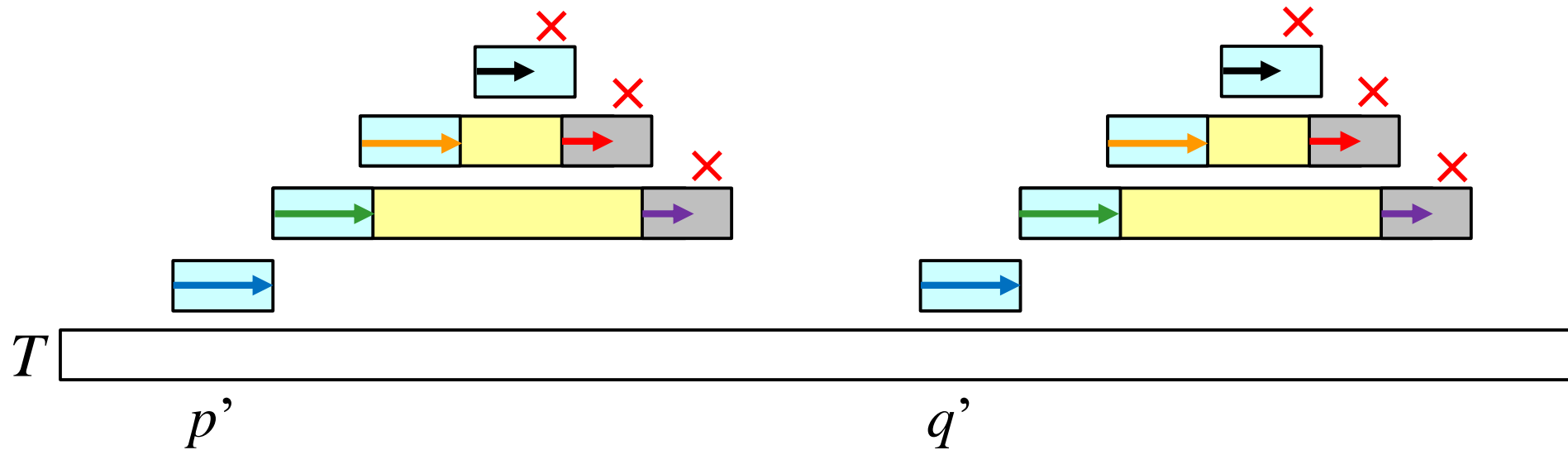
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7. In a top-down manner, we compare the right boundary signatures of length  $\Delta_R + O(1)$  until we find the first mismatch.



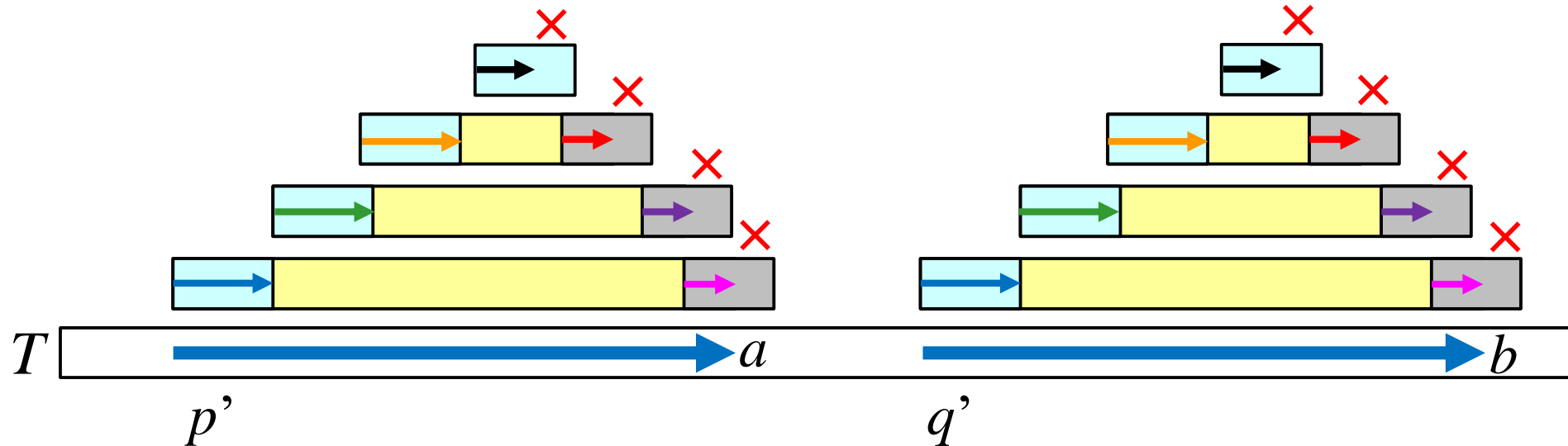
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# Analysis of LCE query time

- ✓ The paths from the root to the  $p$ 'th and  $q$ 'th leaves of the signature tree can be found in  $O(\log u)$  time, since its height is  $O(\log u)$ .
- ✓ The total number of signatures to re-compute and to compare is  $O(\log^* u \log L)$ , since:
  - $\Delta_L \leq \log^* u + 6$  and  $\Delta_R \leq 4$ , and
  - the first mismatch is found at the  $(\log L)$ th level from the bottom.
- ✓ Therefore, LCE query can be answered in  $O(\log u + \log^* u \log L)$  time.

# From SLP to signature encoding

## Lemma 3 (SLP to signature encoding)

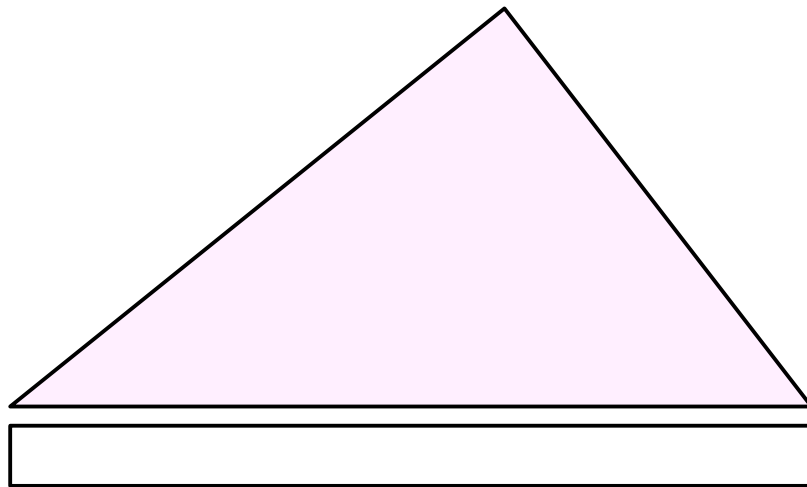
Given an SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  of size  $n$  which derives a string  $T$  of length  $u$ , we can compute the signature encoding of  $T$  in  $O(n \log \log n \log^* u \log u)$  time.

- ✓ In this talk I show a simpler  $O(n \log n \log^* u \log u)$ -time construction.

# From SLP to signature encoding

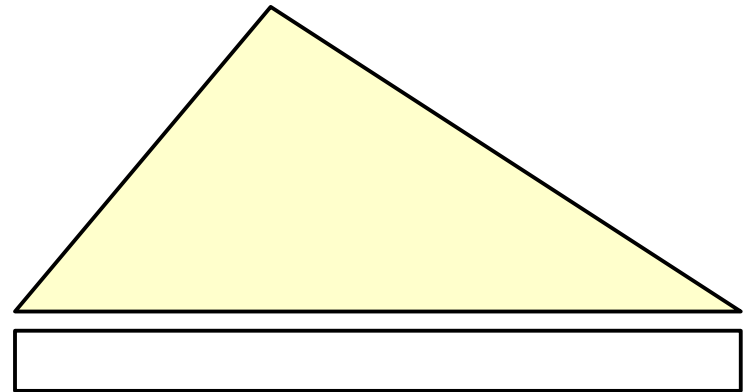
- ✓ Assume that, for a production  $X_i \rightarrow X_l X_r$ , we have computed the signature encodings of the decompressed strings  $val(X_l)$  and  $val(X_r)$ .

signature tree of  $val(X_l)$



$val(X_l)$

signature tree of  $val(X_r)$



$val(X_r)$

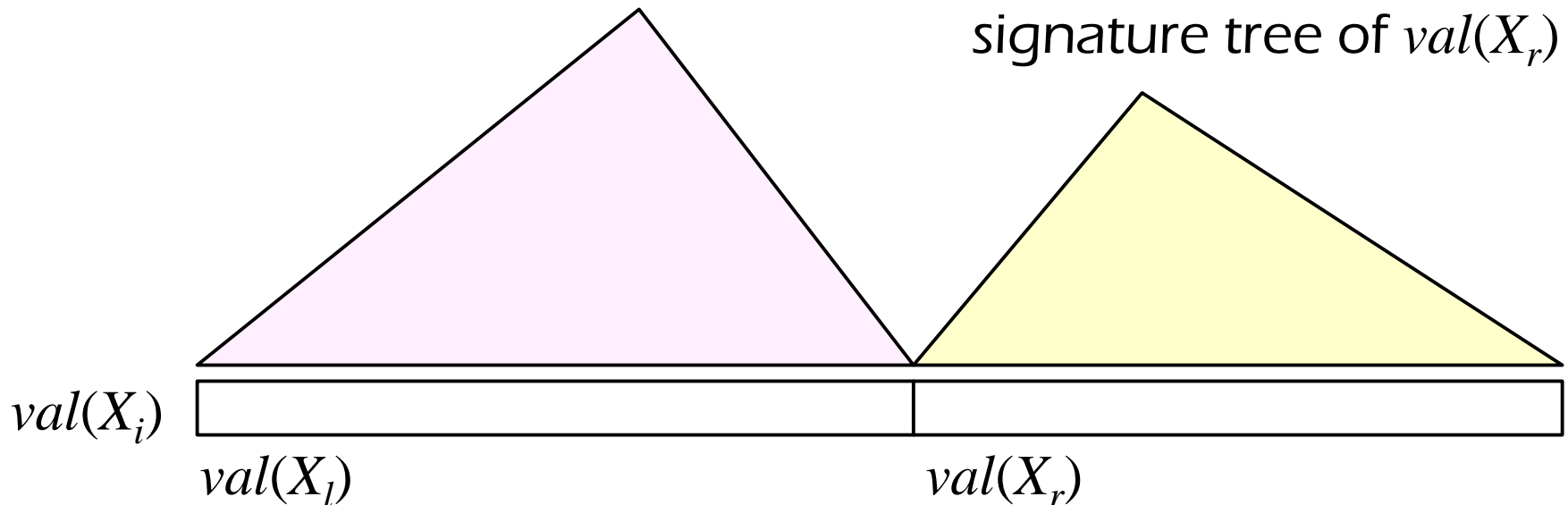


# From SLP to signature encoding

- ✓ By “concatenating” the signature trees of  $val(X_l)$  and  $val(X_r)$ , we obtain the signature tree of  $val(X_i)$ .

signature tree of  $val(X_l)$

signature tree of  $val(X_r)$

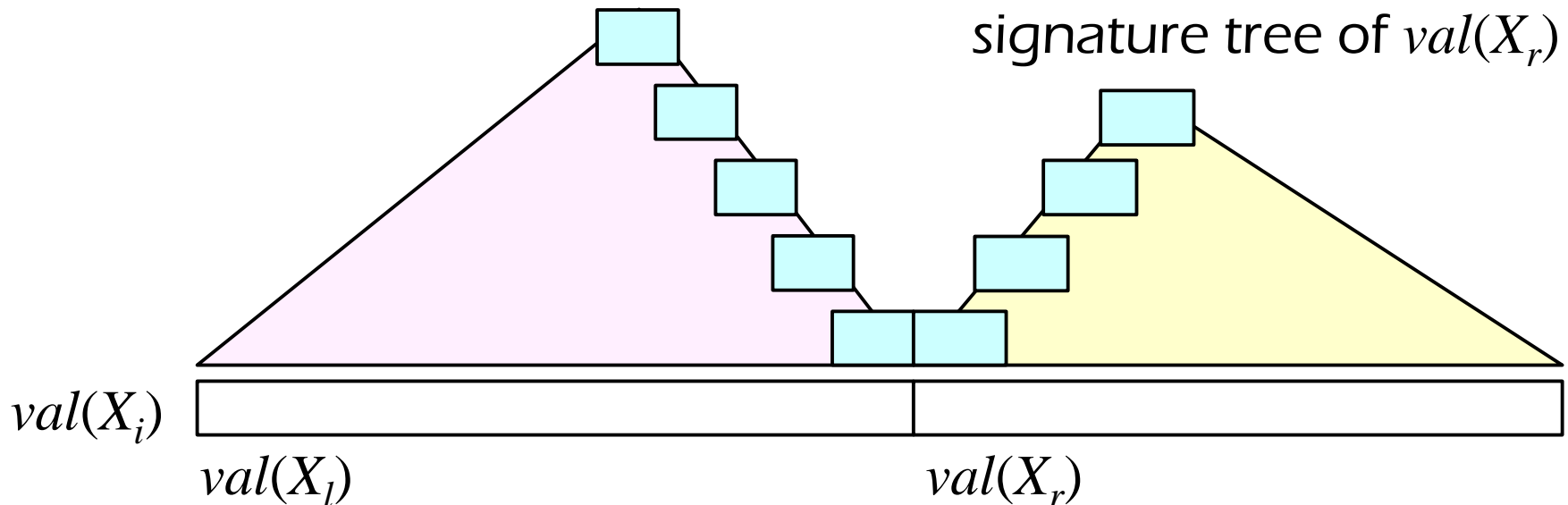


# From SLP to signature encoding

- ✓ In a bottom-up manner, we re-compute the boundary signatures of length  $\Delta_R + O(1)$  and  $\Delta_L + O(1)$  each, and concatenate the new signatures level-wise.

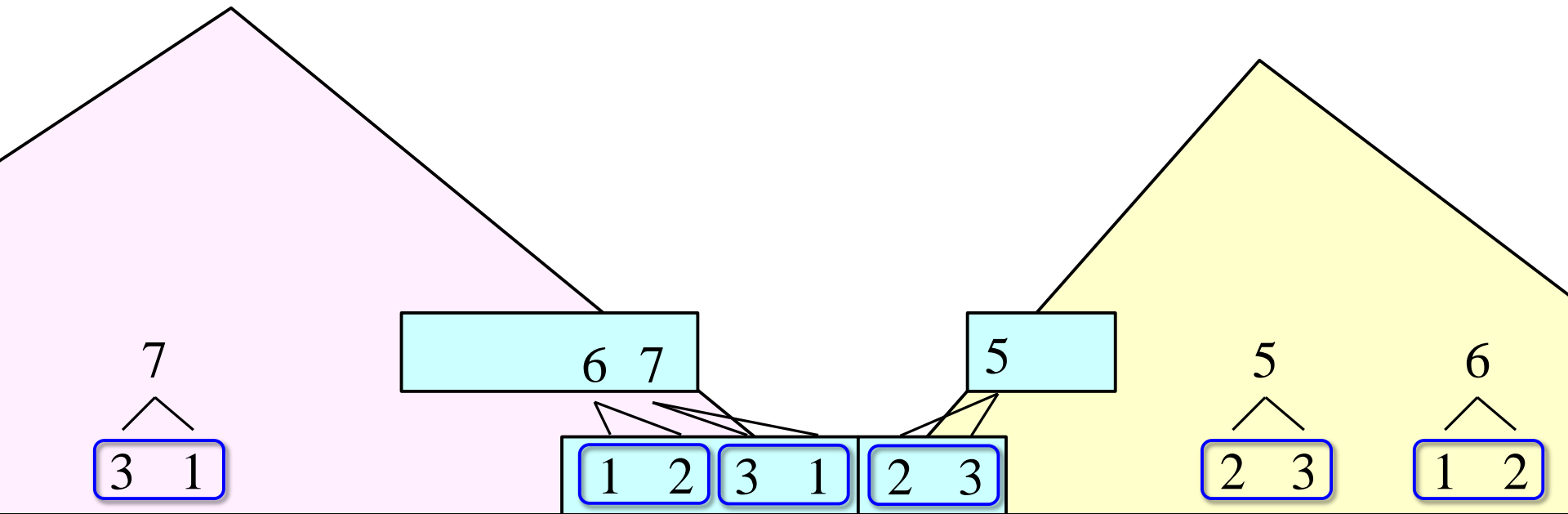
signature tree of  $val(X_l)$

signature tree of  $val(X_r)$



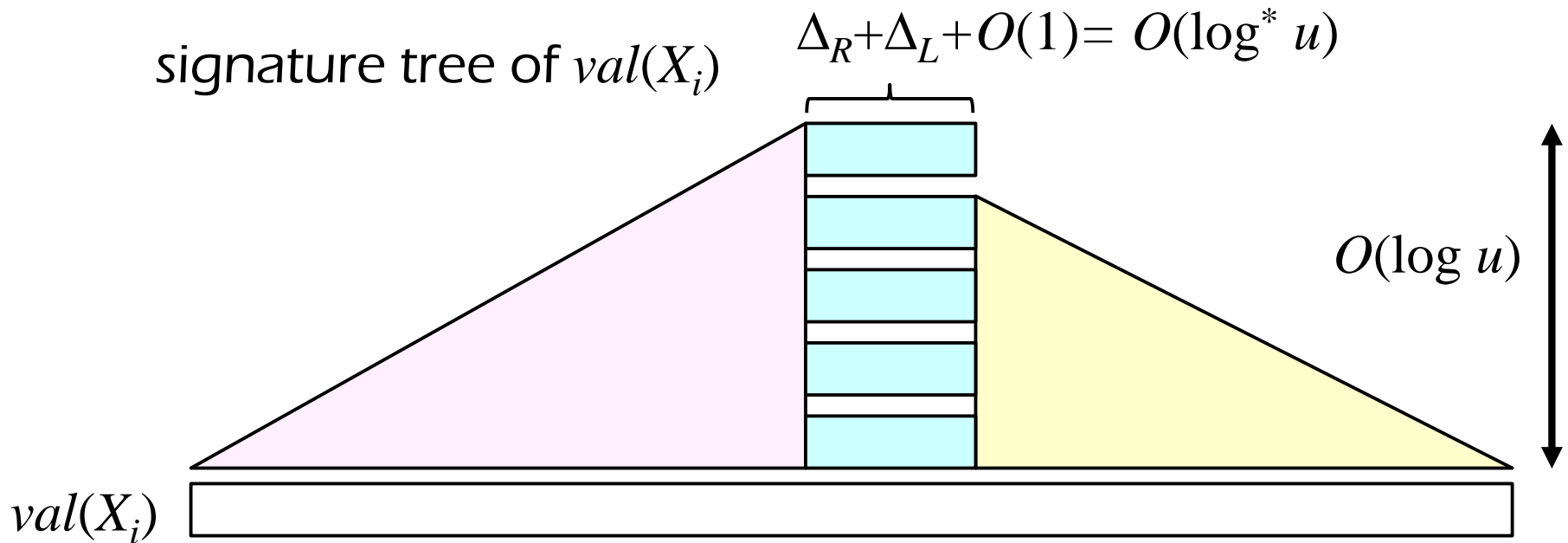
# From SLP to signature encoding

- ✓ If a block of re-computed signatures already exists somewhere else, then we assign the same signature to the block at the next level.  
This is done in  $O(\log n)$  time each, using a BST.



# From SLP to signature encoding

- ✓ Since the height of each signature tree is  $O(\log u)$ , we can compute the signature encoding of  $val(X_i)$  for each  $X_i$  in  $O(\log n \log^* u \log u)$  time.



# How much space?

Lemma 4 [Sahinalp & Vishkin, '95]

The number of signatures involved in the signature encoding of string  $T$  of length  $u$  is  $O(z \log^* u \log u)$ , where  $z$  is the number of factors in the Lempel-Ziv 77 factorization of  $T$ .

- ✓ In our data structure, we need an additive  $n$  term to store beginning positions of occurrences of all non-terminals in the derivation tree of  $X_n$ .

# Main result

## Theorem 1

For any SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  of size  $n$  which represents a string  $T$  of length  $u$ , there exists a data structure which

- supports LCE in  $O(\log u + \log^* u \log L)$  time;
- requires  $O(n + z \log^* u \log u)$  space;
- can be built in  $O(n \log \log n \log^* u \log u)$  time,

where  $L$  is the LCE length and  $z$  is the size of the LZ77 factorization of  $T$ .


# App 1: Finding palindromes

## Problem 2 (finding palindromes on SLP)

Given an SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  representing a string  $T$ , compute a compact representation of all maximal palindromes in  $T$ .

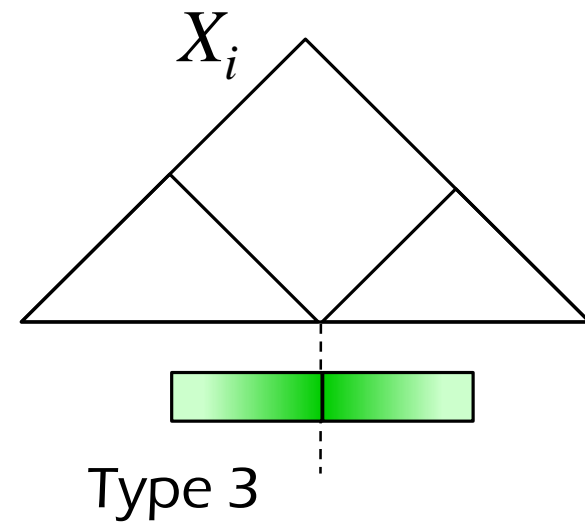
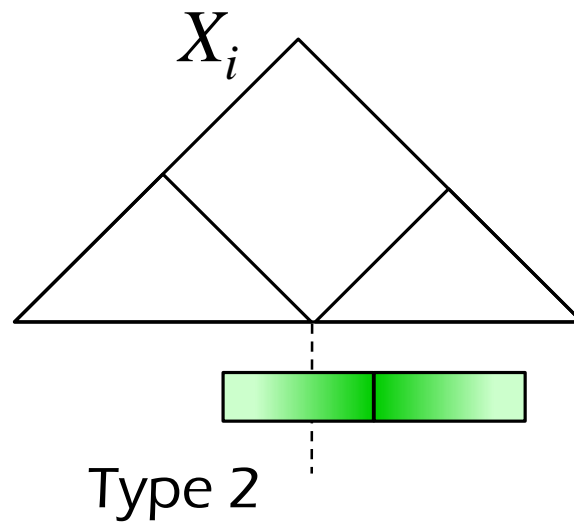
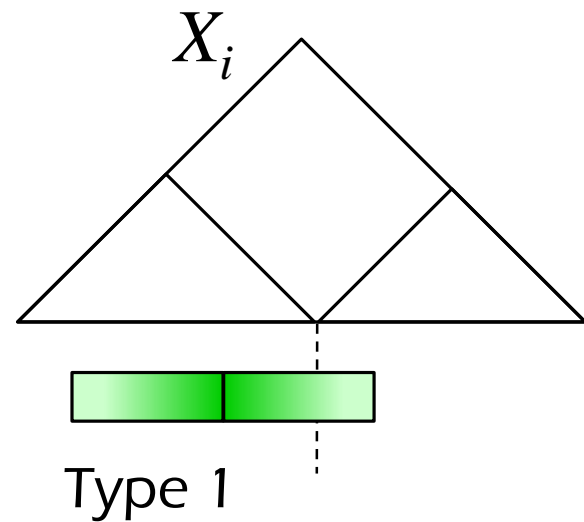
maximal

palindromes

$T =$    $abbbaabbbbabbbaab$

# Stabbed Palindromes

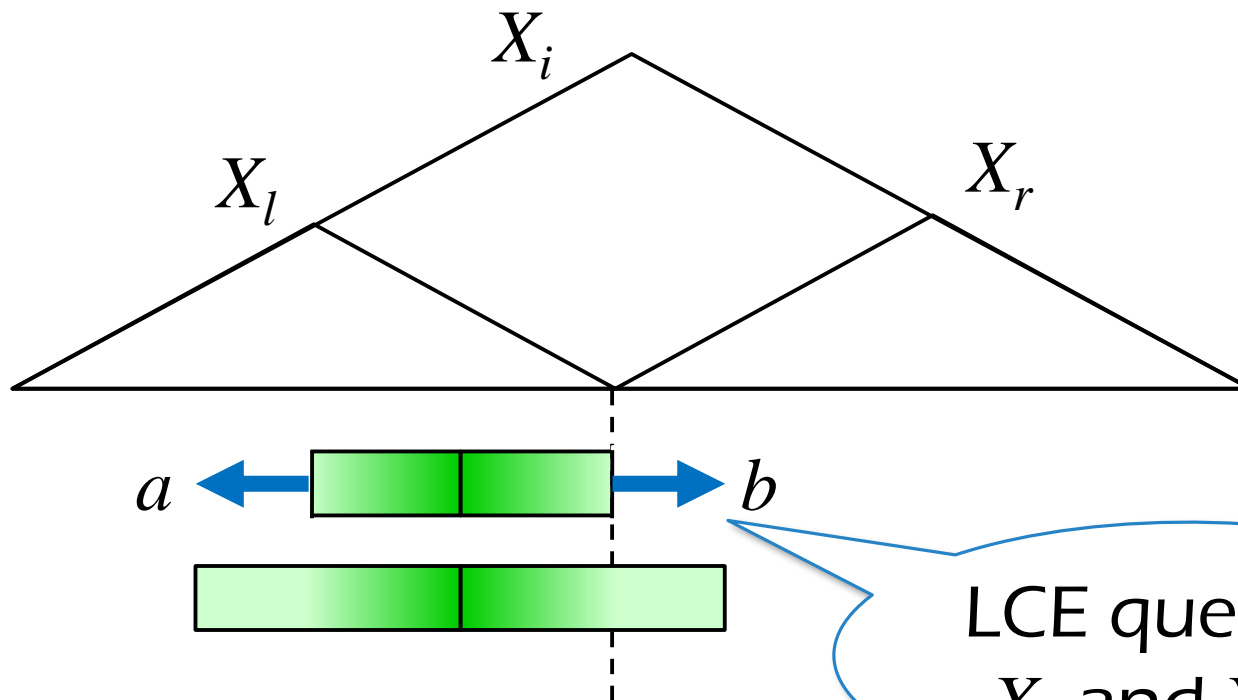
- ✓ For each non-terminal  $X_i$ , there are 3 different types of “stabbed” maximal palindromes.





# Computing Type 1 Palindromes

- ✓ Each Type 1 maximal palindrome of  $X_i$  can be computed by extending the arms of a suffix palindrome of  $X_l$ .



LCE query for  
 $X_r$  and  $X_l^{\text{rev}}$ .

# Suffix Palindromes

Lemma 5 [Apostolico et al., '95]

For any string of length  $k$ , the lengths of its suffix palindromes can be represented by  $O(\log k)$  arithmetic progressions.

- ✓ We can extend the arms of the suffix palindromes belonging to the same arithmetic progression in a batch, using periodicity.

# App 1: Finding Palindromes

## Theorem 2

Given an SLP of size  $n$ , an  $O(n \log u)$ -size representation of all maximal palindromes of string  $T$  can be computed in  $O(n \log^* u \log^2 u)$  time.

With this representation, given an interval  $[i, j]$ , we can decide whether the substring  $T[i..j]$  is a maximal palindrome or not in  $O(\log u)$  time.

# App 2: Comparing Suffixes on SLP

Problem 3 (lexicographical comparison of suffixes)

Preprocess an input SLP representing string  $T$  so that later, any suffixes of the string  $T$  can be lexicographically compared efficiently.

# App 2: Comparing Suffixes on SLP

## Theorem 3

We can preprocess an input SLP of size  $n$  representing string  $T$  of length  $u$  in  $O(n \log \log n \log^* u \log u)$  time such that later, any suffixes of  $T$  can be lexicographically compared in  $O(\log u + \log^* u \log L)$  time, where  $L$  is the length of the LCP of the suffixes.

- ✓ Since the height of the signature tree is  $O(\log u)$ , this theorem is immediate from our LCE data structure.

# App 3: Lyndon factorization on SLP

## Problem 4 (Lyndon factorization on SLP)

Given an SLP  $S = \{X_i \rightarrow expr_i\}_{i=1}^n$  representing a string  $T$ , compute the factor boundaries of the Lyndon factorization of  $T$ .

# Lyndon word

## Definition

A string is said to be a Lyndon word if it is lexicographically smaller than any of its proper cyclic shifts.

For example, “aaaab”, “abc”, “bcbcc” are Lyndon words.

# Lyndon factorization

## Definition

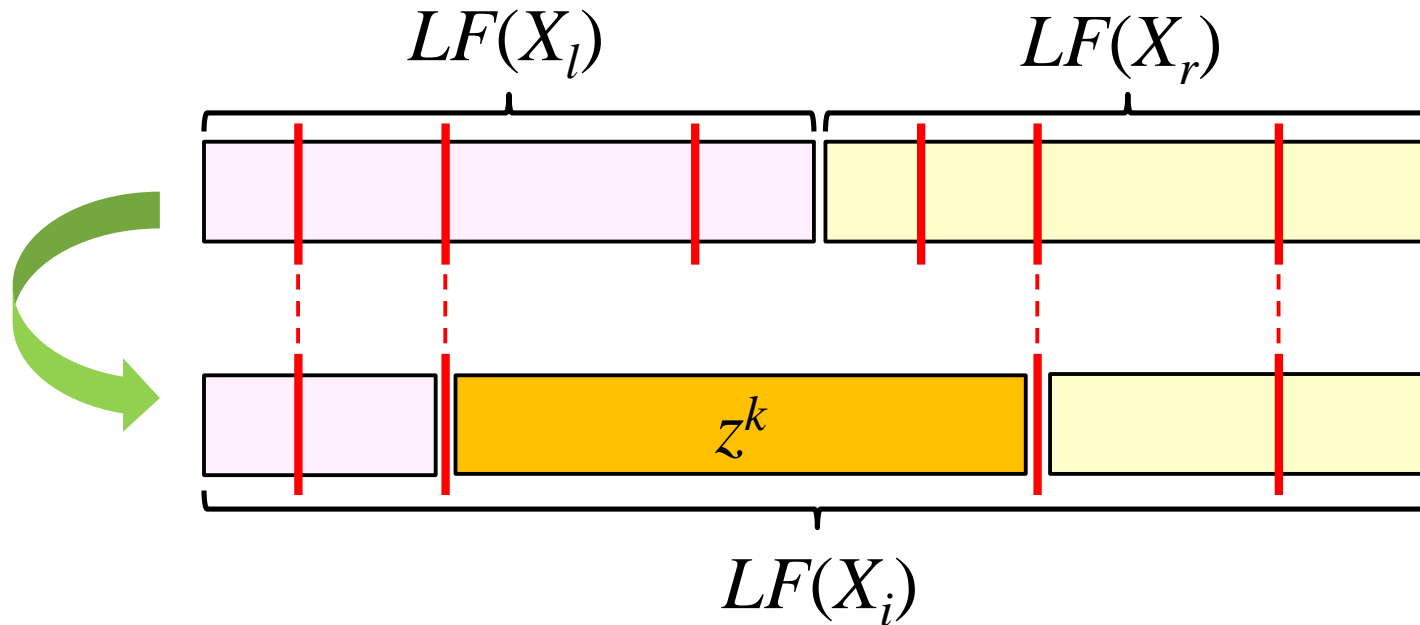
The Lyndon factorization  $LF(T)$  of a string  $T$  is the factorization  $u_1^{p_1}, \dots, u_m^{p_m}$  of  $T$  such that  $u_1, \dots, u_m$  is a sequence of Lyndon words in lexicographical descending order, and  $p_i \geq 1$ .

$T =$  a b c | a b b | a b b | a a b c | a | a | a  
 $u_1$  |  $u_2$  |  $u_2$  |  $u_3$  |  $u_4$  |  $u_4$  |  $u_4$

$LF(T) =$  (abc)<sup>1</sup> | (abb)<sup>2</sup> | (aabc)<sup>1</sup> | (a)<sup>3</sup>  
 $u_1^1$  |  $u_2^2$  |  $u_3^1$  |  $u_4^3$



# Lyndon factorization on SLP



- ✓ I et al. showed an algorithm which computes  $LF(X_i)$  with  $X_i \rightarrow X_l X_r$  in the above manner.
- ✓ The beginning and ending positions of the median Lyndon factor  $z^k$  can be found by a binary search based on lex-comparison of suffixes.

# App 3: Lyndon factorization on SLP

## Theorem 4

Given an SLP of size  $n$  representing string  $T$  of length  $u$ , we can compute the factor boundaries of the Lyndon factorization of  $T$  in  $O(n \log \log n \log^* u \log u)$  time and  $O(n^2 + z \log^* u \log u)$  space.

# Conclusions & further work

- We proposed a new LCE algorithm on SLPs with  $O(\log u + \log^* u \log L)$  query time.
  - ✓ This is the fastest deterministic solution to date.
  - ✓ More details can be found in our arxiv paper: “Dynamic index, LZ factorization, and LCE queries in compressed space”.
- ◆ Lower bound?
- ◆ Other applications?