

# Algorithms for Longest Common Extensions

Jesper Kristensen

DTU Informatics  
Technical University of Denmark

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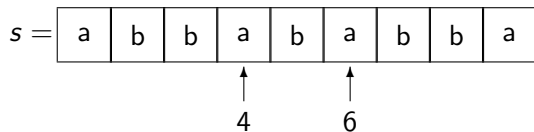
## Summary

# The DIRECTCOMP algorithm

## Input

- ▶  $s = abbababba$
- ▶  $(i, j) = (4, 6)$

## The DIRECTCOMP algorithm

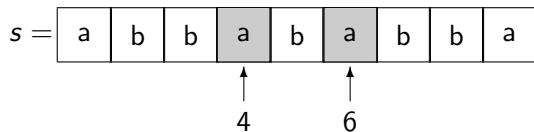


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## The DIRECTCOMP algorithm

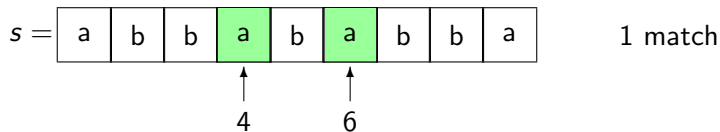


# The DIRECTCOMP algorithm

## Input

- ▶  $s = \text{abbababba}$
- ▶  $(i, j) = (4, 6)$

## The DIRECTCOMP algorithm

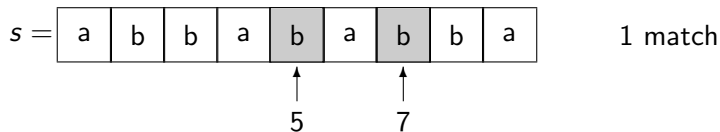


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## Input

- ▶  $s = \text{abbababba}$
- ▶  $(i, j) = (4, 6)$

## The DIRECTCOMP algorithm

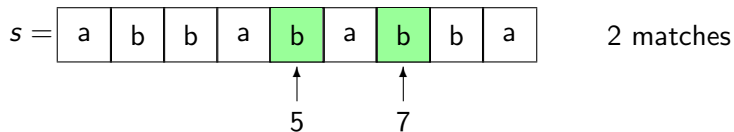


## The DIRECTCOMP algorithm

## Input

- ▶  $s = \text{abbababba}$
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## The DIRECTCOMP algorithm

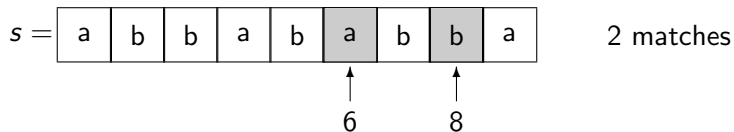


# The DIRECTCOMP algorithm

## Input

- ▶  $s = \text{abbababba}$
- ▶  $(i, j) = (4, 6)$

## The DIRECTCOMP algorithm



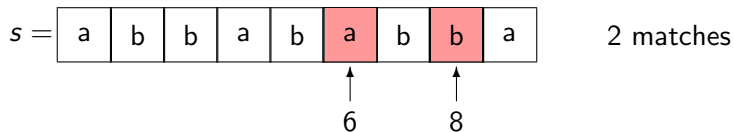


# The DIRECTCOMP algorithm

## Input

- ▶  $s = abbababba$
- ▶  $(i, j) = (4, 6)$

## The DIRECTCOMP algorithm



## Result

$$LCE_s(4, 6) = 2$$

# The LCE Problem

**LCE value**  $LCE_s(i, j)$  is the length of the longest common prefix of the two suffixes of  $s$  starting at index  $i$  and  $j$

**LCE problem** Efficiently query multiple LCE values on a static string  $s$  and varying pairs  $(i, j)$

## Existing Algorithm: DIRECTCOMP

Preprocessing	None
Space	$O(1) +  s $
Query	$O( LCE(i, j) ) = O(n)$
Average query	$O(1)$

For a string length  $n$  and alphabet size  $\sigma$ , the average LCE value over all  $n^\sigma$  strings and  $n^2$  query pairs is  $O(1)$ .

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### References

L. Ilie, G. Navarro, and L. Tinta. The longest common extension problem revisited and applications to approximate string searching. *J. Disc. Alg.*, 8(4):418-428, 2010.

# Existing Algorithms: SUFFIXNCA and LCPRMQ

Two algorithms with best known bounds:

**SUFFIXNCA** Nearest common ancestor queries on a suffix tree

**LCPRMQ** Range minimum queries on a longest common prefix array

Preprocessing	$O(\text{sort}(n, \sigma))$
Space	$O(n)$
Query	$O(1)$
Average query	$O(1)$

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## References

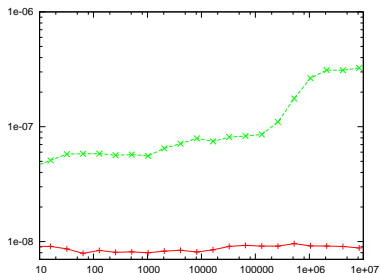
J. Fischer, and V. Heun. Theoretical and Practical Improvements on the RMQ-Problem, with Applications to LCA and LCE. In *Proc. 17th CPM*, pages 36-48, 2006.

D. Harel, R. E. Tarjan. Fast Algorithms for Finding Nearest Common Ancestors. *SIAM J. Comput.*, 13(2):338-355, 1984.

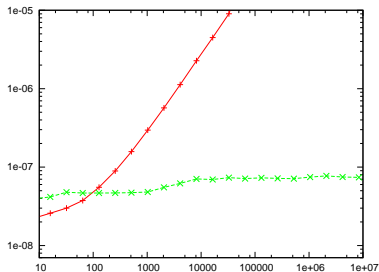
# Existing Algorithms: Practical Results

Query times of **DIRECTCOMP** and **LCPRMQ** by string length

Average case



Worst case



# The FINGERPRINT<sub>k</sub> Algorithm: Data Structure

- ▶ For a string  $s[1..n]$ , the  $t$ -length fingerprints  $F_t[1..n]$  are natural numbers, such that  $F_t[i] = F_t[j]$  if and only if  $s[i..i+t-1] = s[j..j+t-1]$ .
- ▶  $k$  levels,  $1 \leq k \leq \lceil \log n \rceil$
- ▶ For each level,  $\ell = 0..k-1$ :
  - ▶  $t_\ell = \Theta(n^{\ell/k})$ ,  $t_0 = 1$
  - ▶  $H_\ell = F_{t_\ell}$

$H_2[i]$	1	2	3	4	5	6	7	8	9	1	2	3	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$H_1[i]$	1	(2)	3	4	1	(2)	5	6	5	1	(2)	3	4	1	(2)	5	6	5	6	3	4	6	5	6	7	8	9
$s = H_0[i]$	a	(b b a)			a	(b b a)			b	a	(b b a)			a	(b b a)			b	a	b	a	a	b	a	b	a	\$
$i$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27

Space  $O(k \cdot n)$

## The FINGERPRINT<sub>k</sub> Algorithm: Query

1. As long as  $H_\ell[i + v] = H_\ell[j + v]$ , increment  $v$  by  $t_\ell$ , increment  $\ell$  by one, and repeat this step unless and  $\ell = k - 1$ .
2. As long as  $H_\ell[i + v] = H_\ell[j + v]$ , increment  $v$  by  $t_\ell$  and repeat this step.
3. Stop and return  $v$  when  $\ell = 0$ , otherwise decrement  $\ell$  by one and go to step two.

$H_2[i + v]$	1	2	3	4	5	6	7	8	9	1	2	3	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$H_1[i + v]$	1	2	3	4	1	2	5	6	5	1	2	3	4	1	2	5	6	5	6	3	4	6	5	6	7	8	9
$H_0[i + v]$	a	b	b	a	a	b	b	a	b	a	b	b	a	a	b	b	a	b	a	b	a	a	b	a	b	a	\$
$i + v$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27

$H_2[j + v]$	1	2	3	4	5	6	7	8	9	1	2	3	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$H_1[j + v]$	1	2	3	4	1	2	5	6	5	1	2	3	4	1	2	5	6	5	6	3	4	6	5	6	7	8	9
$H_0[j + v]$	a	b	b	a	a	b	b	a	b	a	b	b	a	a	b	b	a	b	a	b	a	a	b	a	b	a	\$
$j + v$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27

$$LCE(3, 12) = 9$$

Query  $O(k \cdot n^{1/k})$

Average query  $O(1)$

# The FINGERPRINT<sub>k</sub> Algorithm: I/O

## ► Original:

- Data structure:  $H_\ell[i] = F_{t_\ell}[i]$
- Size:  $|H_\ell| = n$
- I/O:  $O(k \cdot n^{1/k})$

## ► Cache optimized:

- Data structure:  

$$H_\ell[((i-1) \bmod t_\ell) \cdot \lceil n/t_\ell \rceil + \lfloor (i-1)/t_\ell \rfloor + 1] = F_{t_\ell}[i]$$
- Size:  $|H_\ell| = n + t_\ell$
- I/O:  $O\left(k \cdot \left(\frac{n^{1/k}}{B} + 1\right)\right)$ 
  - Best when  $k$  is small  $\implies n^{1/k}$  is large.

$H_2[i+v]$	1	2	3	4	5	6	7	8	9	1	2	3	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$H_1[i+v]$	1	2	3	4	1	2	5	6	5	1	2	3	4	1	2	5	6	5	6	3	4	6	5	6	7	8	9
$H_0[i+v]$	a	b	b	a	a	b	b	a	b	a	b	b	a	a	b	b	a	b	a	b	a	a	b	a	b	a	\$
$i+v$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27

$H_2[j+v]$	1	2	3	4	5	6	7	8	9	1	2	3	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24
$H_1[j+v]$	1	2	3	4	1	2	5	6	5	1	2	3	4	1	2	5	6	5	6	3	4	6	5	6	7	8	9
$H_0[j+v]$	a	b	b	a	a	b	b	a	b	a	b	b	a	a	b	b	a	b	a	b	a	a	b	a	b	a	\$
$j+v$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27



# The FINGERPRINT<sub>k</sub> Algorithm

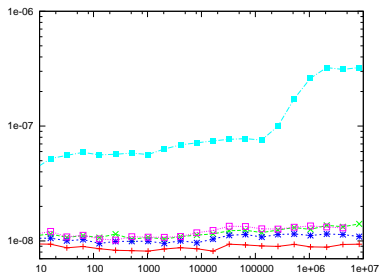
	$1 \leq k \leq \lceil \log n \rceil$
Preprocessing	$O(k \cdot n + \text{sort}(n, \sigma))$
Space	$O(k \cdot n)$
Query	$O(k \cdot n^{1/k})$
Average query	$O(1)$
Query I/O	$O\left(k \cdot \left(\frac{n^{1/k}}{B} + 1\right)\right)$

	$k = 1$	$k = 2$	$k = \lceil \log n \rceil$
Preprocessing	$O(\text{sort}(n, \sigma))$	$O(\text{sort}(n, \sigma))$	$O(n \log n)$
Space	$O(n)$	$O(n)$	$O(n \log n)$
Query	$O(n)$	$O(\sqrt{n})$	$O(\log n)$
Average query	$O(1)$	$O(1)$	$O(1)$
Query I/O	$O\left(\frac{n}{B}\right)$	$O\left(\frac{\sqrt{n}}{B}\right)$	$O(\log n)$

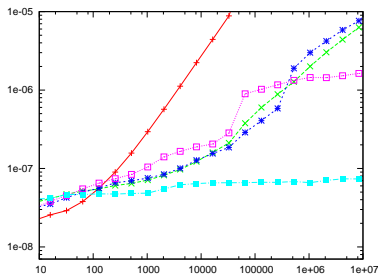
# Practical Results

Query times of **DIRECTCOMP**, **FINGERPRINT<sub>2</sub>** (cache opt.), **FINGERPRINT<sub>3</sub>** (not cache opt.), **FINGERPRINT<sub>[log n]</sub>** (not cache opt.) and **LCPRMQ** by string length

Average case

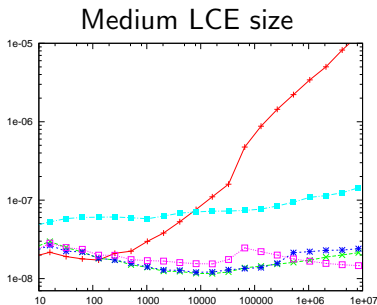


Worst case



# Practical Results

Query times of **DIRECTCOMP**, **FINGERPRINT<sub>2</sub>** (cache opt.), **FINGERPRINT<sub>3</sub>** (not cache opt.), **FINGERPRINT<sub>⌈log n⌋</sub>** (not cache opt.) and **LCPRMQ** by string length



# Cache Optimization, Practical Results

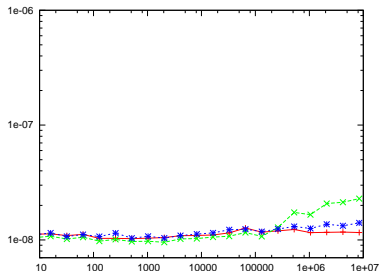
Is I/O optimization good in practice?

- ▶ Pro: better cache efficiency
  - ▶ Best for small  $k$ , no change for  $k = \lceil \log n \rceil$
- ▶ Con: Calculating memory addresses is more complicated
  - ▶  $((i - 1) \bmod t_\ell) \cdot \lceil n/t_\ell \rceil + \lfloor (i - 1)/t_\ell \rfloor + 1$  vs.  $i$

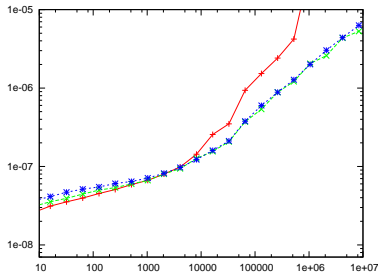
# The FINGERPRINT<sub>k</sub> Algorithm: Practical Results, I/O

Query times of FINGERPRINT<sub>2</sub> **without cache optimization** and with cache optimization using **shift operations** vs. **multiplication, division and modulo**

Average case



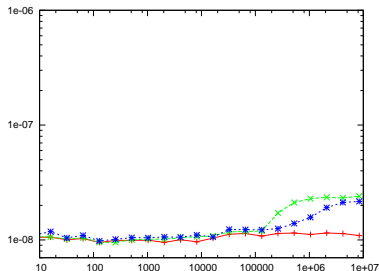
Worst case



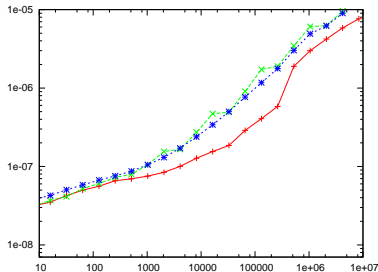
# The FINGERPRINT<sub>k</sub> Algorithm: Practical Results, I/O

Query times of FINGERPRINT<sub>3</sub> **without cache optimization** and with cache optimization using **shift operations** vs. **multiplication, division and modulo**

Average case



Worst case



# LCE on Compressed Strings

## Goal

- ▶ Allow LCE queries without decompressing the string
- ▶ Using Ziv-Lempel compression (LZ)

## How

- ▶ LZ compression represents the string as a tree
- ▶ An LCE query on a LZ compressed string is a number of LCE queries on a tree





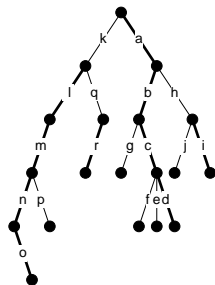
# Constant Time String LCE on Heavy Paths

- ▶ Data structure:

- ▶ Construct a heavy path decomposition
- ▶ For each heavy path, store characters as a substring of  $s$

- ▶ Query:

- ▶ Use constant time string LCE on  $s$  for each heavy path



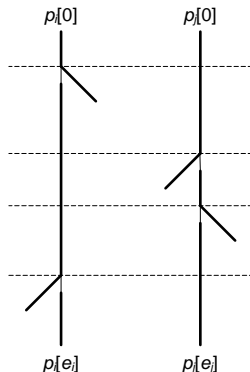
$s =$ 

k	l	m	n	o	q	r	p	a	b	c	d	g	f	e	h	i	j
---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---

Preprocessing	$O(\text{sort}(n, \sigma))$
Space	$O(n)$
Query	$O(\log n)$

# Constant Time String LCE on Heavy Paths

- ▶ How to find the indexes  $(i, j)$  in the string:
  - ▶ Store an index at each node
- ▶ How to know then the heavy path splits from the queried path:
  - ▶ Store a pointer to the end of the heavy path at each node
  - ▶ Find NCA of the end of the heavy path and the end of the queried path
- ▶ How to find a node on the queried path:
  - ▶ Use level ancestor



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## References

Michael A. Bendera, Martín Farach-Colton. The Level Ancestor Problem simplified. Theoretical Computer Science 321, 2004, pages 5-12.

# Constant Time String LCE on Heavy Paths

Data structure:

- ▶ A **heavy path decomposition**
- ▶ A string  $s$  storing labels of each heavy path
  - ▶ Constant time **string LCE** on  $s$
- ▶ **Pointers** from each node to
  - ▶ its index in  $s$
  - ▶ the end of its heavy path
- ▶ Constant time **nearest common ancestor**
- ▶ Constant time **level ancestor**

Preprocessing  $O(\text{sort}(n, \sigma))$

Space  $O(n)$

Query  $O(\log n)$

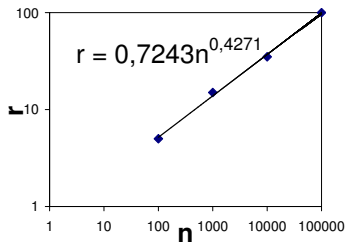
## Summary

	DIRECT- COMP	LCPRMQ / SUFFIXNCA	FINGERPRINT <sub>k</sub>
Preprocessing	$O(1)$	$O(\text{sort}(n, \sigma))$	$O(k \cdot n + \text{sort}(n, \sigma))$
Space	$O(1)$	$O(n)$	$O(k \cdot n)$
Query	$O(n)$	$O(1)$	$O(k \cdot n^{1/k})$
Average query	$O(1)$	$O(1)$	$O(1)$
Query I/O	$O(\frac{n}{B})$	$O(1)$	$O\left(k \cdot \left(\frac{n^{1/k}}{B} + 1\right)\right)$

- ▶ In practice, the FINGERPRINT<sub>k</sub> algorithm is...
  - ▶ ...almost as good as DIRECTCOMP and significantly better than LCPRMQ in average case
  - ▶ ...significantly better than DIRECTCOMP but worse than LCPRMQ in worst case
- ▶ Cache optimization of FINGERPRINT<sub>k</sub> improves query times at  $k = 2$  and worsens query times at  $k \geq 3$
- ▶ We can generalize LCE to use on trees with  $O(\log n)$  query time in linear space

# Kommentarer og rettelser til rapporten

- Hvor kommer  $r = 0.73n^{0.42}$  fra?



- Afsnit 4.5.3 Cache Optimization: FINGERPRINT<sub>3</sub> skal være FINGERPRINT<sub>2</sub>