

Text Indexing

Lecture 12: Longest Common Extensions

Florian Kurpicz

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Recap: Document Listing and Top- k Retrieval

Definition: Document Listing

Given a collection of D documents $\mathcal{D} = \{d_1, d_2, \dots, d_D\}$ containing symbols from an alphabet $\Sigma = [1, \sigma]$ and a pattern $P \in \Sigma^*$, return all $j \in [1, D]$, such that d_j contains P .

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- $d_1 = \text{ATA}$
- $d_2 = \text{TAAA}$
- $d_3 = \text{TATA}$

And for queries:

- $P = \text{TA}$ is contained in d_1 , d_2 , and d_3
- $P = \text{ATA}$ is contained in d_1 and d_3

Recap: Document Listing and Top- k Retrieval

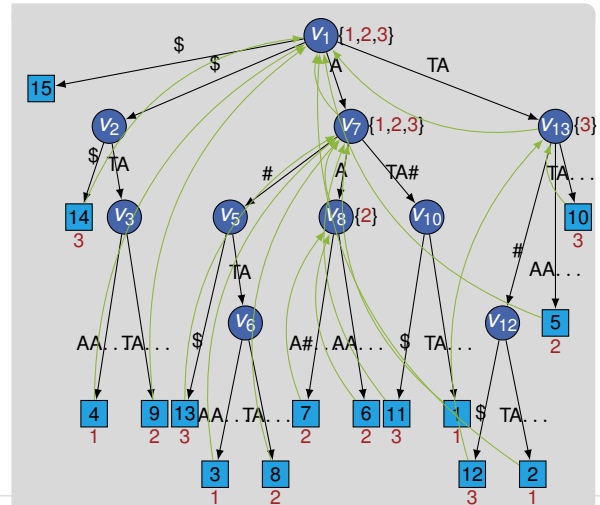
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Recap: Pattern Matching with the LCP-Array (1/3)

- remember how many characters of the pattern and suffix match
- identify how long the prefix of the old and next suffix is
- do so using the LCP-array and
- **range minimum queries** ⓘ detailed introduction in **Advanced Data Structures**

- $lcp(i, j) = \max\{k: T[i..i+k)$
- $lcp(i, j) = T[j..j+k)\} = LCP[RMQ_{LCP}(i+1, j)]$
- RMQs can be answered in $O(1)$ time and
- require $O(n)$ space

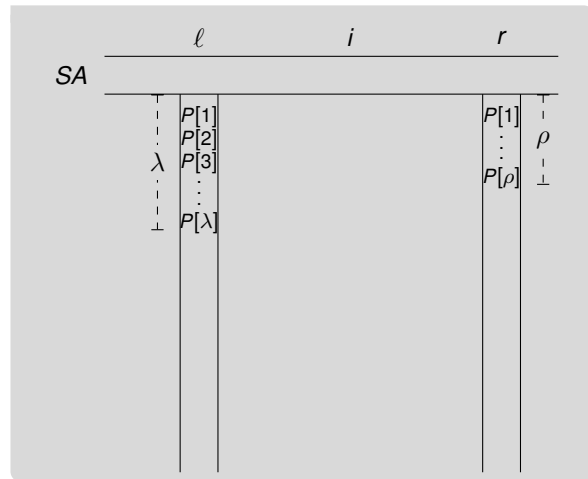
Definition: Range Minimum Queries

Given an array $A[1..m)$, a **range minimum query** for a range $\ell \leq r \in [1, n)$ returns

$$RMQ_A(\ell, r) = \arg \min\{A[k]: k \in [\ell, r)\}$$

Recap: Pattern Matching with the LCP-Array (2/3)

- during binary search matched
 - λ characters with left border ℓ and
 - ρ characters with right border r
 - w.l.o.g. let $\lambda \geq \rho$
-
- middle position i
 - decide if continue in $[\ell, i]$ or $[i, r]$
-
- let $\xi = \text{lcp}(SA[\ell], SA[i])$ $\odot O(1)$ time with RMQs



Old Problem, New Name

Definition: Longest Common Extensions

Given a text T of size n over an alphabet of size σ , construct data structure that answers for $i, j \in [1, n]$

$$\text{lce}_T(i, j) = \max\{\ell \geq 0 : T[i, i + \ell] = T[j, j + \ell]\}$$

- also denoted as $\text{lcp}(i, j)$ ⓘ in this lecture

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										1										2
	1	2	3	4	5	6	7	8	9	0	1	2	3	4	5	6	7	8	9	0
T	A	B	C	D	A	B	C	C	D	B	C	C	B	A	B	C	D	A	D	A

$$\text{lce}_T(1, 14) = 0\ 1\ 2\ 3\ 4\ 5$$

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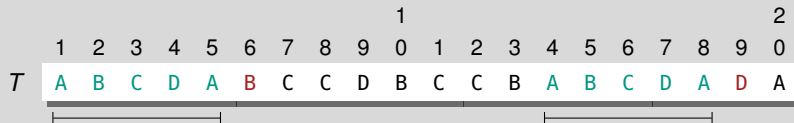
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Applications

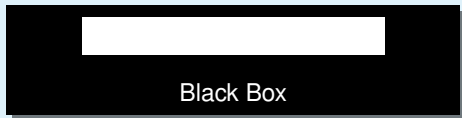
- (sparse) suffix sorting
- approximate pattern matching
- ...



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Sophisticated Black Box (BB)

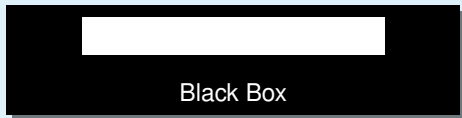
- based on ISA, LCP, and RMQ



- $O(1)$ query time, $\approx 9n$ bytes additional space

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Ultra Naive Scan (UNS)

- compare character by character

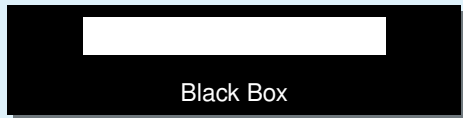


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Practical Algorithms for Longest Common Extensions [IT09]

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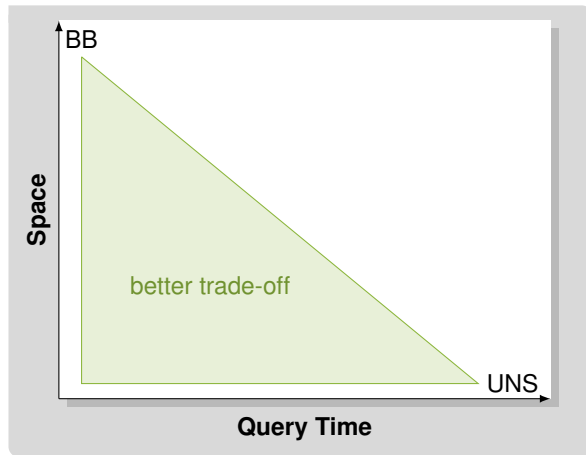
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Monte Carlo and Las Vegas Algorithms

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- some Monte Carlo algorithms can be turned into Las Vegas algorithms
- depends on correctness check
- all Monte Carlo algorithms presented today can be turned into Las Vegas algorithms

Randomized String Matching

- compute s of strings
- application not limited to LCEs

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$$\text{fingerprint}(i, j) = \left(\sum_{k=i}^j T[k] \cdot \sigma^{j-k} \right) \bmod q$$

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- if $T[i..i + \ell] = T[j..j + \ell]$, then

$$\text{fingerprint}(i, i + \ell) = \text{fingerprint}(j, j + \ell)$$

- if $T[i..i + \ell] \neq T[j..j + \ell]$, then

$$\text{Prob}(\text{fingerprint}(i, i + \ell) = \text{fingerprint}(j, j + \ell)) \in O\left(\frac{\ell \lg \sigma}{n^c}\right)$$

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- example on the board 

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$$B[i] = T[(i-1)\tau + 1..i\tau]$$

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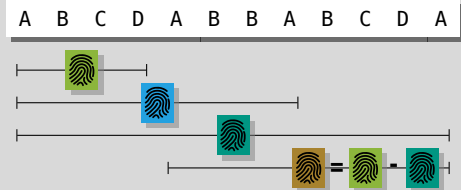
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- choose $\tau \in \Theta(w / \lg \sigma)$ (8 for byte alphabet)
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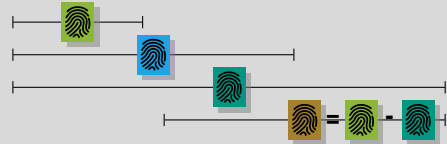


- all parts of text are restorable
- how?

Overwriting the Text with Fingerprints (2/2)

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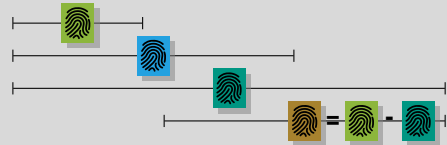
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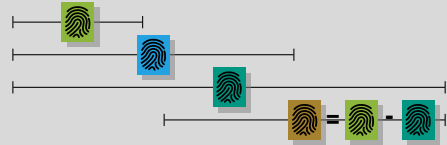
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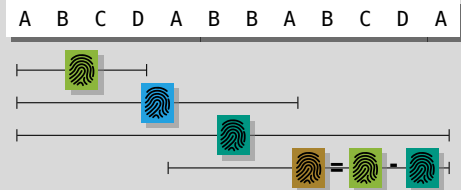
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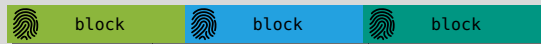
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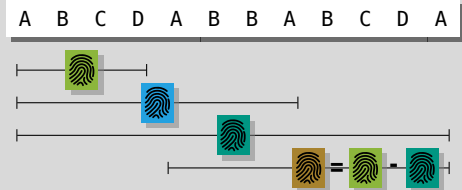


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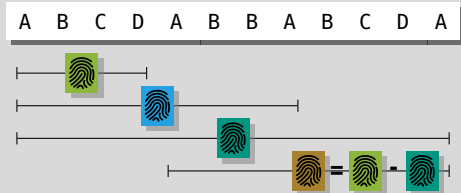


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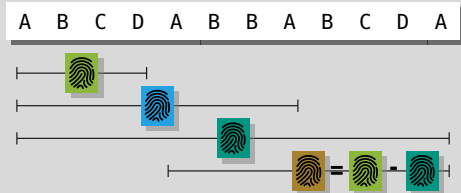


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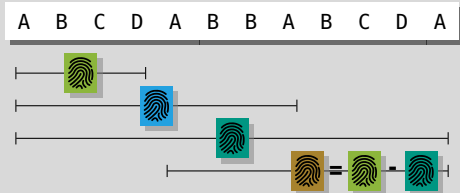
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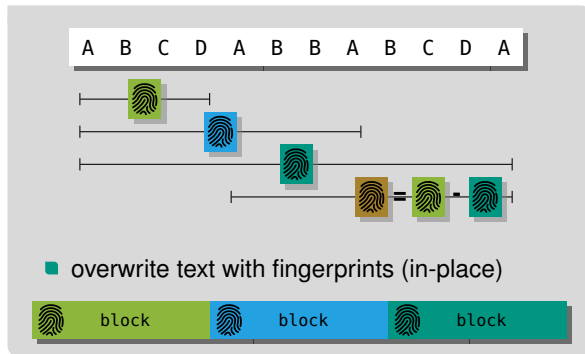


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Answering LCE Queries with Fingerprints

LCEs with Fingerprints

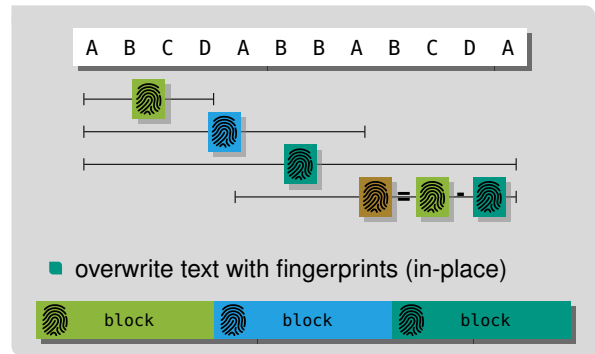
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- exponential search until $\text{fingerprint}(i, i + 2^k) \neq \text{fingerprint}(j, j + 2^k)$
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-
- requires $O(\lg \ell)$ time for LCEs of size ℓ



String Synchronizing Sets (Simplified, 1/2)

Definition: Simplified τ -Synchronizing Sets [KK19]

Given a text T of length n and $0 < \tau \leq n/2$, a **simplified** τ -synchronizing set S of T is

$$S = \{i \in [1, n - 2\tau + 1] : \min\{\text{fingerprint}(j, j + \tau - 1) : j \in [i, i + \tau]\} \in \{\text{fingerprint}(i, i + \tau - 1), \text{fingerprint}(i + \tau, i + 2\tau - 1)\}\}$$

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$$S = \{i \in [1, n - 2\tau + 1] : \min\{\text{fingerprint}(j, j + \tau - 1) : j \in [i, i + \tau]\} \in \{\text{fingerprint}(i, i + \tau - 1), \text{fingerprint}(i + \tau, i + 2\tau - 1)\}\}$$

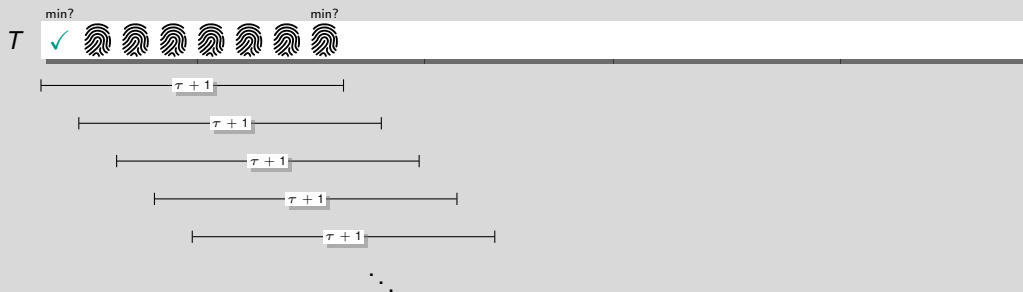


String Synchronizing Sets (Simplified, 1/2)

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String Synchronizing Sets (Simplified, 2/2)

- $|S| = \Theta(n/\tau)$ in practice (on most data sets)
- more complex definition required to obtain this size

Consistency & (Simplified) Density Property

- for all $i, j \in [1, n - 2\tau + 1]$ we have

$$T[i, i+2\tau-1] = T[j, j+2\tau-1] \Rightarrow i \in S \Leftrightarrow j \in S$$

- for any τ consecutive positions there is at least one position in S

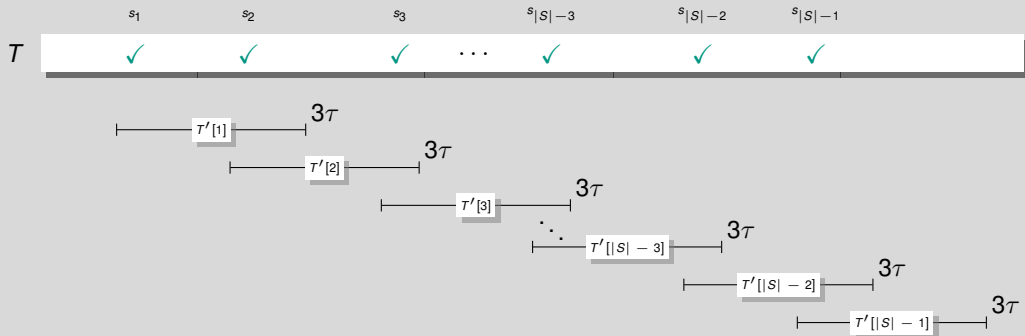
Answering LCE Queries with String Synchronizing Sets (1/2)

Text T' for Positions in S

	s_1	s_2	s_3	\dots	$s_{ S -3}$	$s_{ S -2}$	$s_{ S -1}$
T	✓	✓	✓	\dots	✓	✓	✓

Answering LCE Queries with String Synchronizing Sets (1/2)

Text T' for Positions in S



Answering LCE Queries with String Synchronizing Sets (2/2)

- in practice, we sort the substrings
- characters of T' are the ranks of substrings
- build BB LCE for T' w.r.t. length in T

Answering Queries

- compare naively for 3τ characters
- if equal find successors of i and j in S
- compute LCE of successors in T'

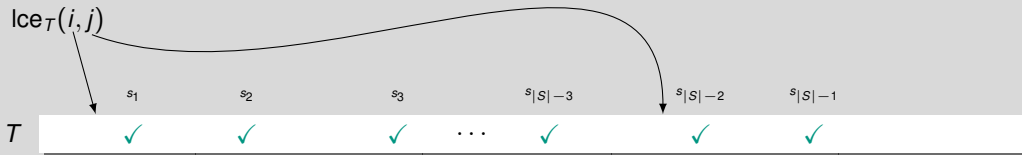


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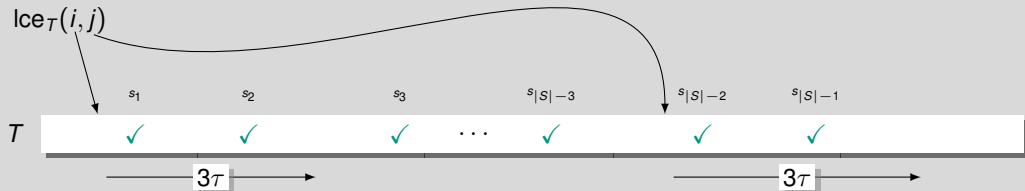


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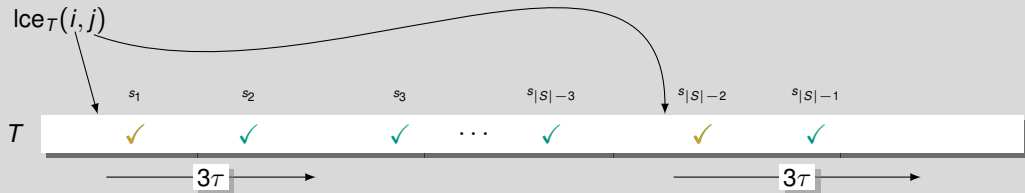


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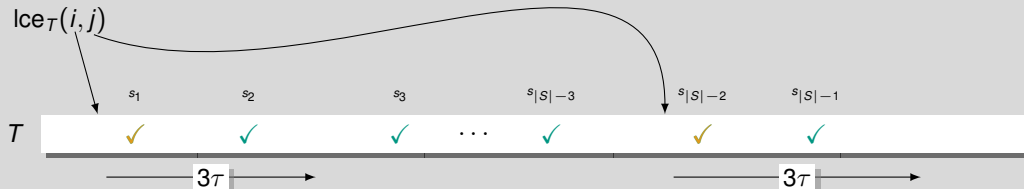


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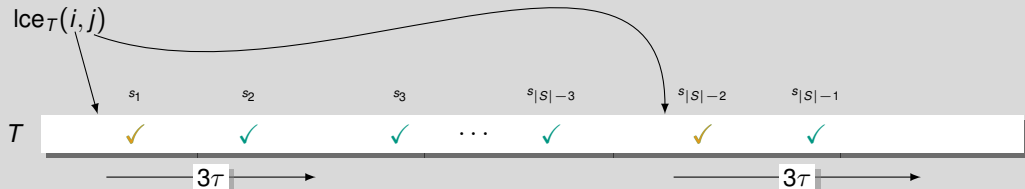
- in this example: $\text{Ice}_T(i, j) = s_1 - i + \text{Ice}_{T'}(1, |S| - 2)$

Answering LCE Queries with String Synchronizing Sets (2/2)

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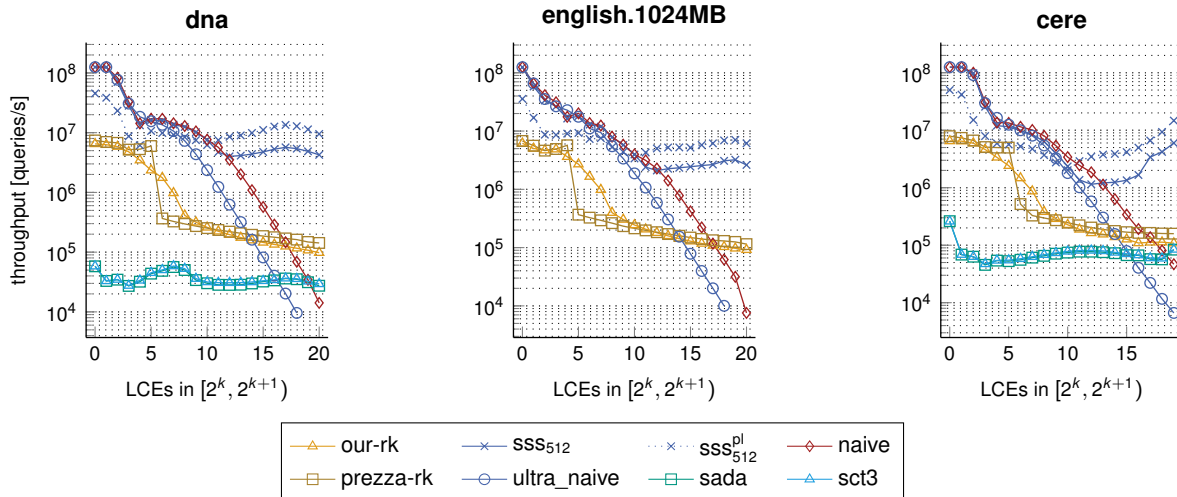
Answering Queries

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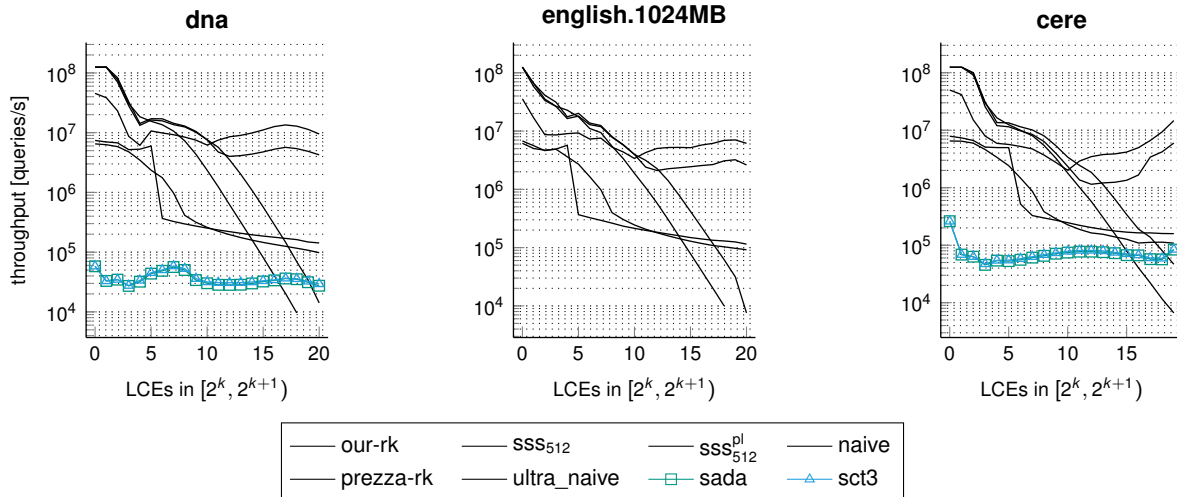


- in this example: $\text{lce}_T(i, j) = s_1 - i + \text{lce}_{T'}(1, |S| - 2)$
- in practice: we have a very fast static successor data structure

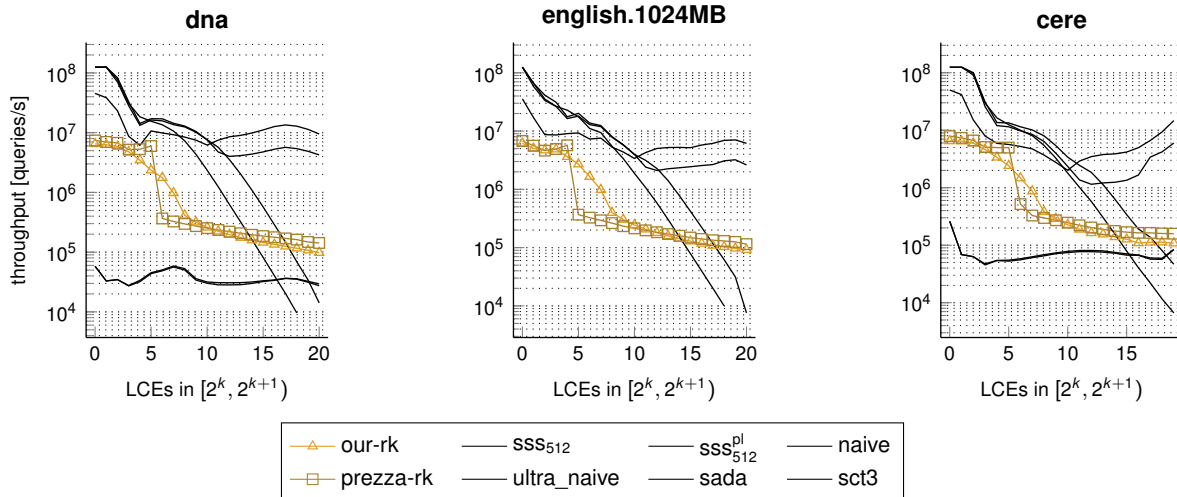
Practical Evaluation [Din+20]



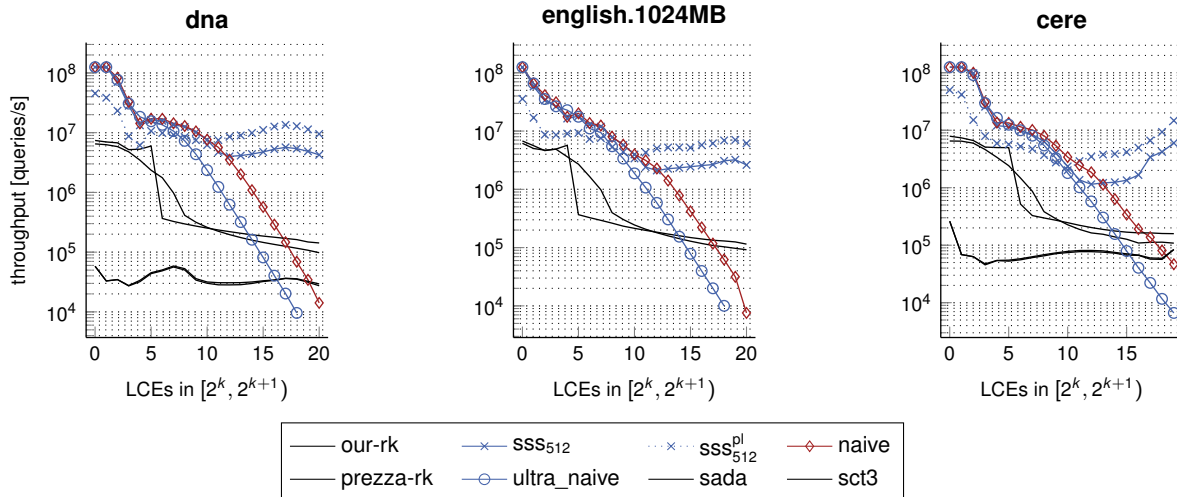
Practical Evaluation [Din+20]



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Practical Evaluation [Din+20]



Conclusion and Outlook

This Lecture

- longest common extension queries
- Karp-Rabin fingerprints
- string synchronizing sets

That's all! We are (mostly)
done.

Conclusion and Outlook

This Lecture

- longest common extension queries
- Karp-Rabin fingerprints
- string synchronizing sets

Next Lecture

- big recap and Q&A

Thats all! We are (mostly)
done.

Anmeldung Projekt & Discussion of the evaluation

Bibliography I

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