

# c-trie++: A Dynamic Trie Tailored for Fast Prefix Searches

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# Problem Definition (1/4)

## 【Problem】 Dynamic Prefix Search

Maintain a data structure for a dynamic set  $S = \{T_1..T_k\}$  of strings that, given a query pattern  $P$ , can compute the pair

- a)  $\max \{l : P[1..l] = T_i[1..l] \text{ for some } i \in [1..k]\}$  and
- b)  $I_P = \{i : T_i[1..l] = P[1..l]\}$  efficiently.

Example:

$$S = \{T_1, T_2, T_3, T_4, T_5\}$$

$$T_1 = \text{idea}$$

$$T_2 = \text{inter}face$$

$$T_3 = \text{inter}net$$

$$T_4 = \text{infinite}$$

$$T_5 = \text{laboratory}$$

$$P = \text{inter}$$

$$\text{output} = (5, \{2, 3\})$$

# Problem Definition (2/4)

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$$T_1 = \text{idea}$$

$$T_2 = \text{interface}$$

$$T_3 = \text{internet}$$

$$T_4 = \text{infinite}$$

$$T_5 = \text{laboratory}$$

$$P = \text{inner}$$

$$\text{output} = (2, \{2, 3, 4\})$$

# Problem Definition (3/4)

## 【Problem】 Dynamic Prefix Search

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Example:

$$S = \{T_1, T_2, T_3, T_4, T_5, \textcolor{red}{T}_6\}$$

$T_1 = \text{idea}$

$T_2 = \text{interface}$

$T_3 = \text{internet}$

$T_4 = \text{infinite}$

$T_5 = \text{laboratory}$

$T_6 = \text{indexing}$

Supports insertion of a string into  $S$ .

$\text{insert}(T_6)$

# Problem Definition (4/4)

## 【Problem】 Dynamic Prefix Search

Maintain a data structure for a dynamic set  $S = \{T_1..T_k\}$  of strings that, given a query pattern  $P$ , can compute the pair

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Example:

$$S = \{T_1, T_2, T_3, T_4, \cancel{T_5}, T_6\}$$

$$T_1 = \text{idea}$$

$$T_2 = \text{interface}$$

$$T_3 = \text{internet}$$

$$T_4 = \text{infinite}$$

~~$$T_5 = \text{laboratory}$$~~

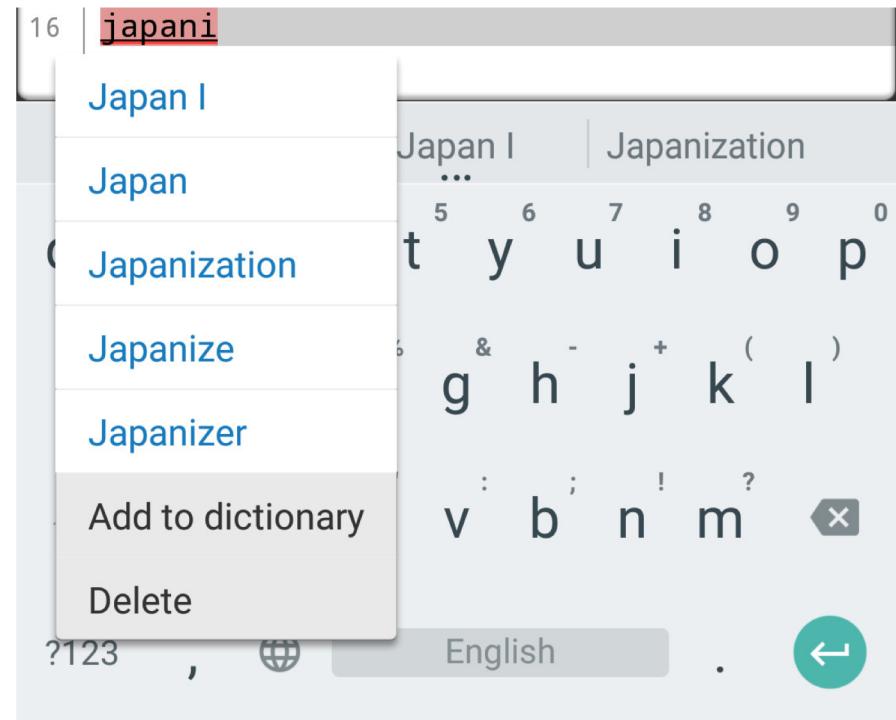
$$T_6 = \text{indexing}$$

Supports deletion of a string from  $S$ .

$\text{delete}(T_5)$

# Introduction (1/3)

- prefix search applications
  - **input method editors**
  - query auto-completion
  - range query filtering



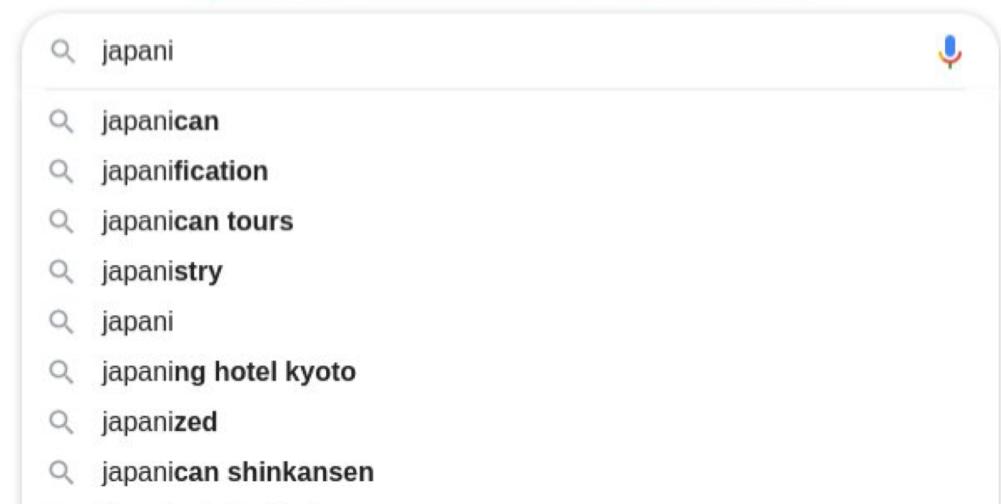
(entering *japani* on an Android phone)

# Introduction (2/3)

## □ prefix search applications

- input method editors
- **query auto-completion**
- range query filtering

(entering `japani` on google)



A screenshot of a Google search suggestions interface. The search term 'japani' is entered in the search bar. Below it, a list of suggested queries is shown:

- ✓ japanican
- ✓ japanification
- ✓ japanican tours
- ✓ japanistry
- ✓ japani
- ✓ japaning hotel kyoto
- ✓ japanized
- ✓ japanican shinkansen

<https://www.google.com/>

# Introduction (3/3)

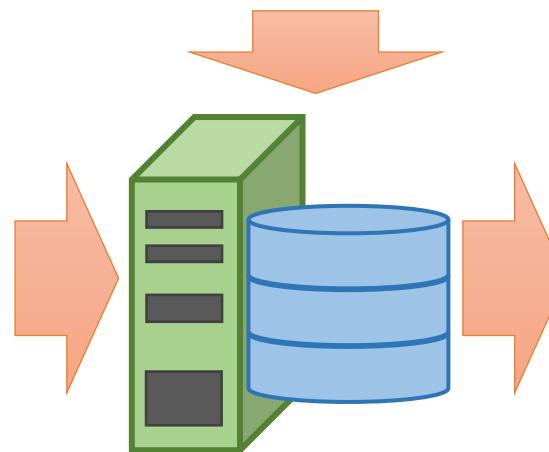
## □ prefix search applications

- input method editors
- query auto-completion
- **range query filtering**

UserQueries

User	Date	Word
661	12/03/2020	refund-ticket
457	11/03/2020	trip-cancellation
139	01/03/2020	corona-virus
:	:	:

```
SELECT User, Word  
FROM UserQueries  
WHERE Word LIKE 'japani%'  
AND ...
```



User	Word
79	japanican
83	japanification
89	japanistry
97	japani

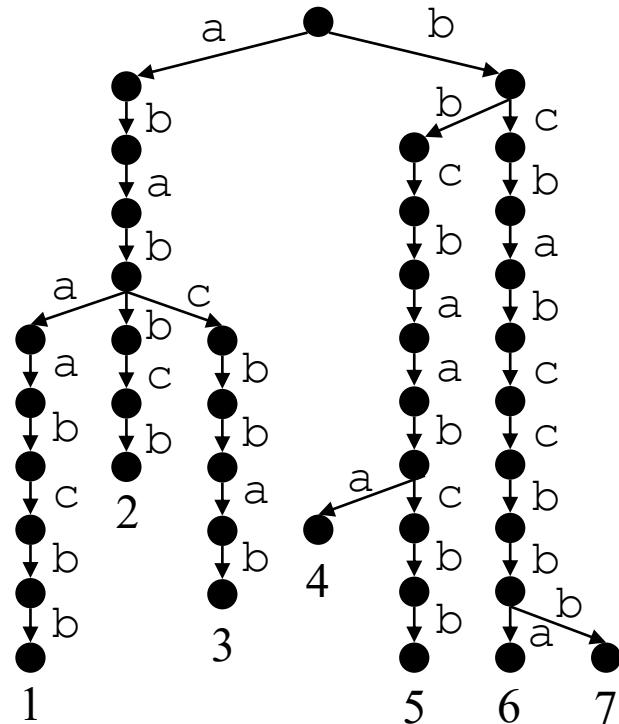
# Compact Trie [Morrison, 1968]

- Trie : represents strings where common prefixes are compressed to a single path (front encoding).
  - Compact Trie : reduces the number of nodes by replacing branchless path segments with a single edge.

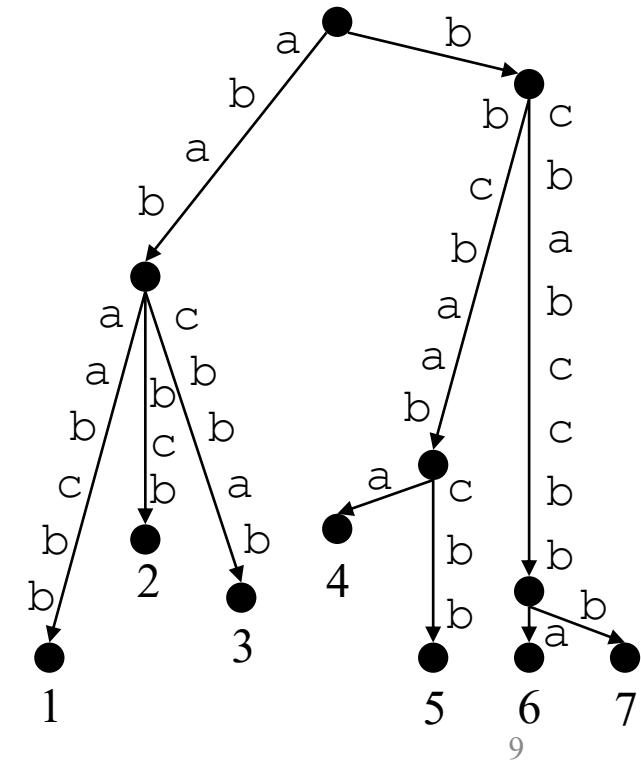
# Strings

$$\begin{aligned}
 T_1 &= ababaabcb \\
 T_2 &= ababbcb \\
 T_3 &= ababcbbab \\
 T_4 &= bbcbaaba \\
 T_5 &= bbcbaabcb \\
 T_6 &= bcbabcccba \\
 T_7 &= bcbabccbbb
 \end{aligned}$$

Trie



## Compact Trie



# Previous Works for Dynamic Prefix Search (1/2)

	Space [bits]	Prefix Search Time in Expectation
Trie	$O( T  \log  T )$	$O(m + occ)$
Compact Trie [Morrison, 1968]	$ T  \log \sigma + \Theta(k \log k)$	$O(m + occ)$

$|T|$  : trie size (front encoding size)

$k = |S|$ ,  $m = |P|$ ,  $\sigma = |\Sigma|$ ,  $\Sigma$  : alphabet,  $occ$  : number of occurrences

# Previous Works for Dynamic Prefix Search (2/2)

	Space [bits]	Prefix Search Expected Time
Trie	$O( T  \log  T )$	$O(m + occ)$
Compact Trie [Morrison, 1968]	$ T  \log \sigma + \Theta(k \log k)$	$O(m + occ)$
Z-Fast Trie [Belazzougui et al., 2010]	$ T  \log \sigma + \Theta(kw)$	$O(m / \alpha + occ + \log (\textcolor{red}{m} \log \sigma))$
Packed C-Trie [Takagi et al., 2017]	$ T  \log \sigma + \Theta(kw)$	$O(m / \alpha + occ + \log \textcolor{red}{w})$

$|T|$  : trie size (front encoding size)

$n = \sum_i |T_i|$ ,  $k = |S|$ ,  $m = |P|$ ,  $\sigma = |\Sigma|$ ,  $\Sigma$  : alphabet,  $occ$  : number of occurrences

$w$  : machine word size ( $w = \Omega(\log n)$ ),  $\alpha = O(w / \log \sigma)$

# Our Contribution for Dynamic Prefix Search

	Space [bits]	Prefix Search Expected Time
Trie	$O( T  \log  T )$	$O(m + occ)$
Compact Trie [Morrison, 1968]	$ T  \log \sigma + \Theta(k \log k)$	$O(m + occ)$
Z-Fast Trie [Belazzougui et al., 2010]	$ T  \log \sigma + \Theta(kw)$	$O(m / \alpha + occ + \log (m \log \sigma))$
Packed C-Trie [Takagi et al., 2017]	$ T  \log \sigma + \Theta(kw)$	$O(m / \alpha + occ + \log w)$
<b>C-Trie++ [Ours]</b>	$ T  \log \sigma + \Theta(kw)$	$O(m / \alpha + occ + \log \min\{\alpha, m\})$

$\alpha < w$  always holds.

$|T|$  : trie size (front encoding size)

$n = \sum_i |T_i|$ ,  $k = |S|$ ,  $m = |P|$ ,  $\sigma = |\Sigma|$ ,  $\Sigma$  : alphabet,  $occ$  : number of occurrences

$w$  : machine word size ( $w = \Omega(\log n)$ ),  $\alpha = O(w / \log \sigma)$

# Word Packing

In the word RAM model with word size  $w$  bits,  
we can compare ( $=, <, >$ ) two  $O(w)$  bits integers in  $O(1)$  time.

⇒ Let  $\alpha = w / \log \sigma$ .

i	d	e	a
01101001	01100100	01100101	01100001

a character uses  $\lceil \log \sigma \rceil$  bits  
⇒ strings of length  $\alpha$  use  $O(w)$  bits

e.g.

64-bits architecture

$\sigma = 256$

$w = 32$

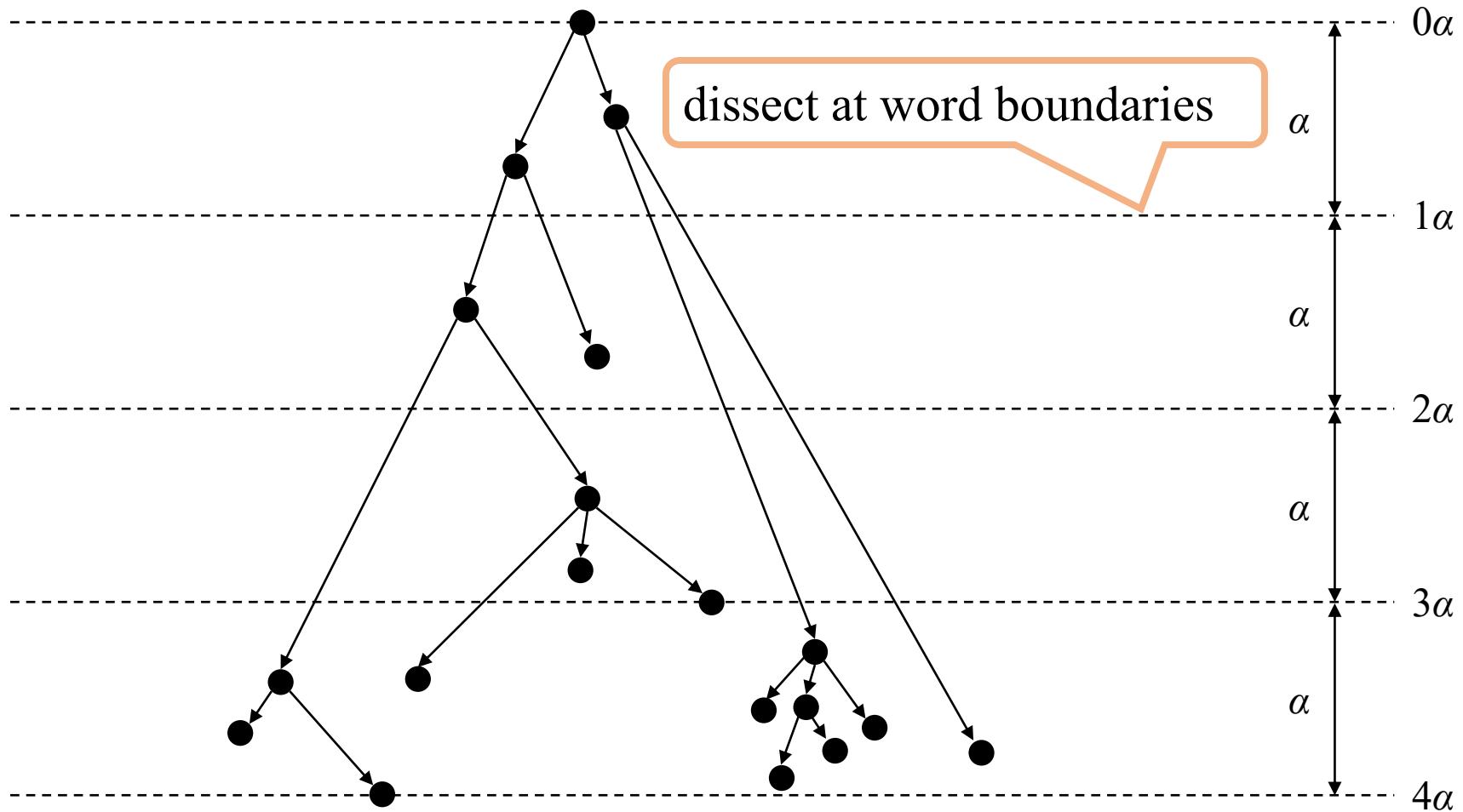
$\alpha = 4$

We can compare two strings of length  $\alpha$  in  $O(1)$  time.

⇒ We can compare two strings of length  $m$  in  $O(m / \alpha)$  time.

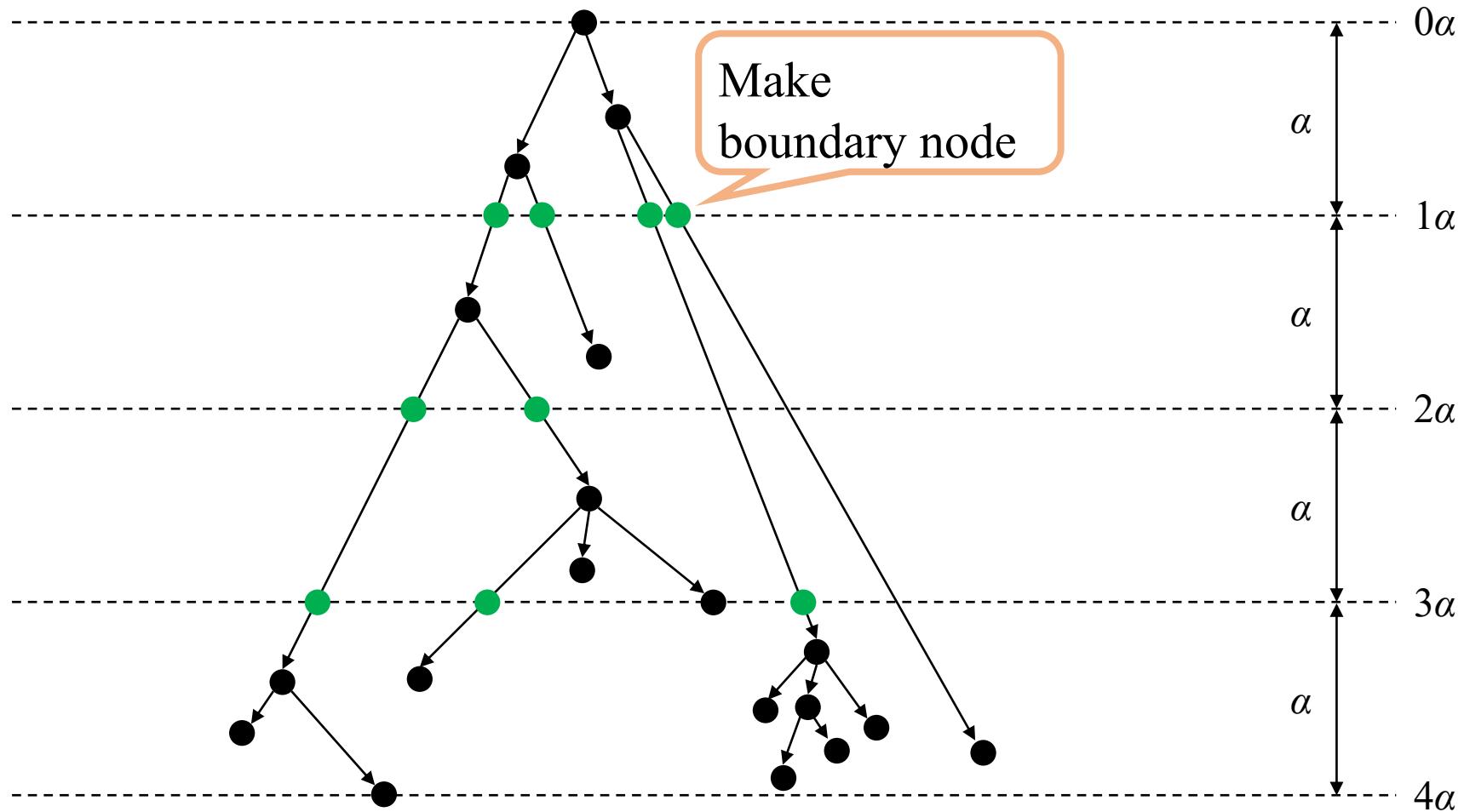
# Introduction of Micro Trie (1/4)

## Compact Trie



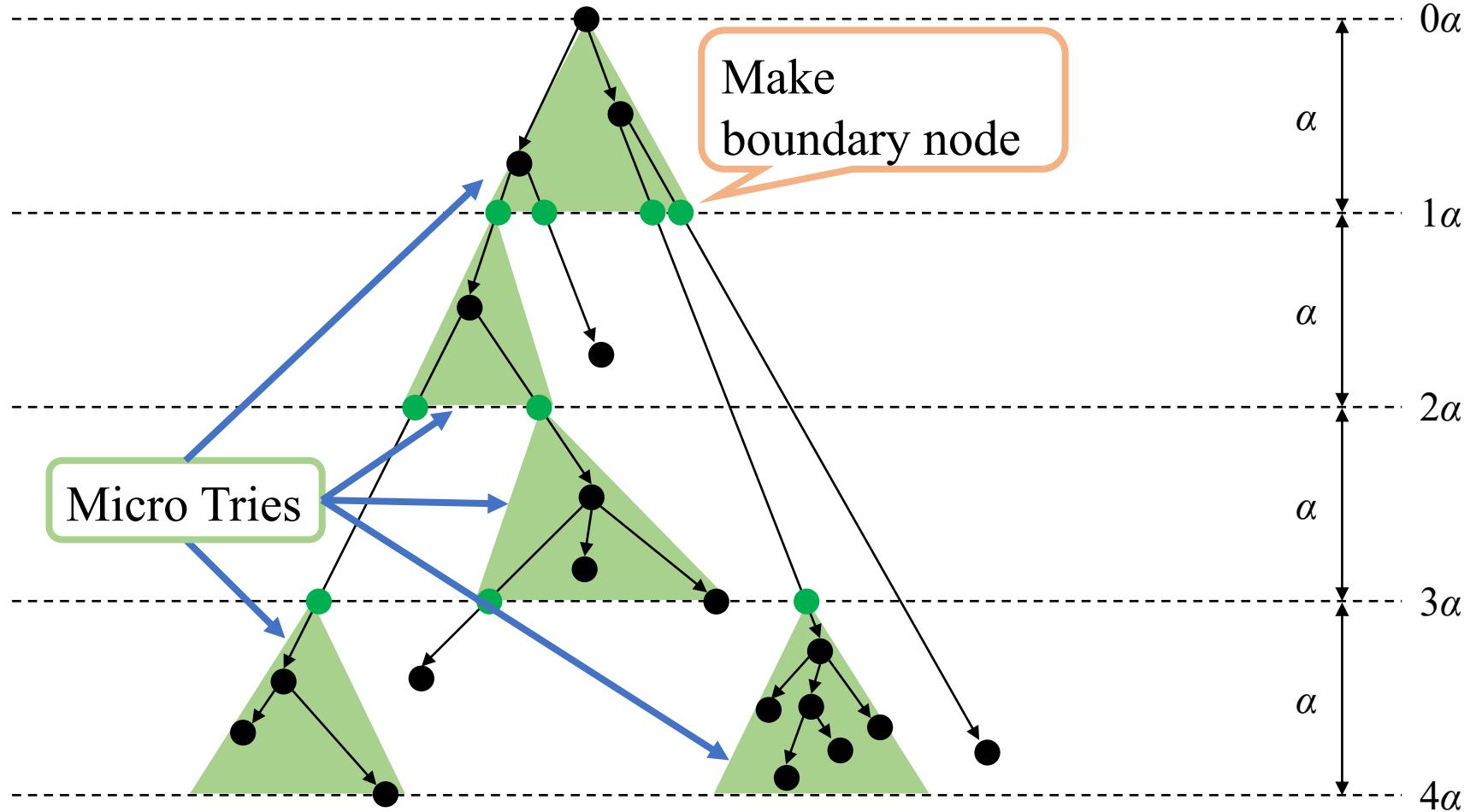
# Introduction of Micro Trie (2/4)

## Compact Trie



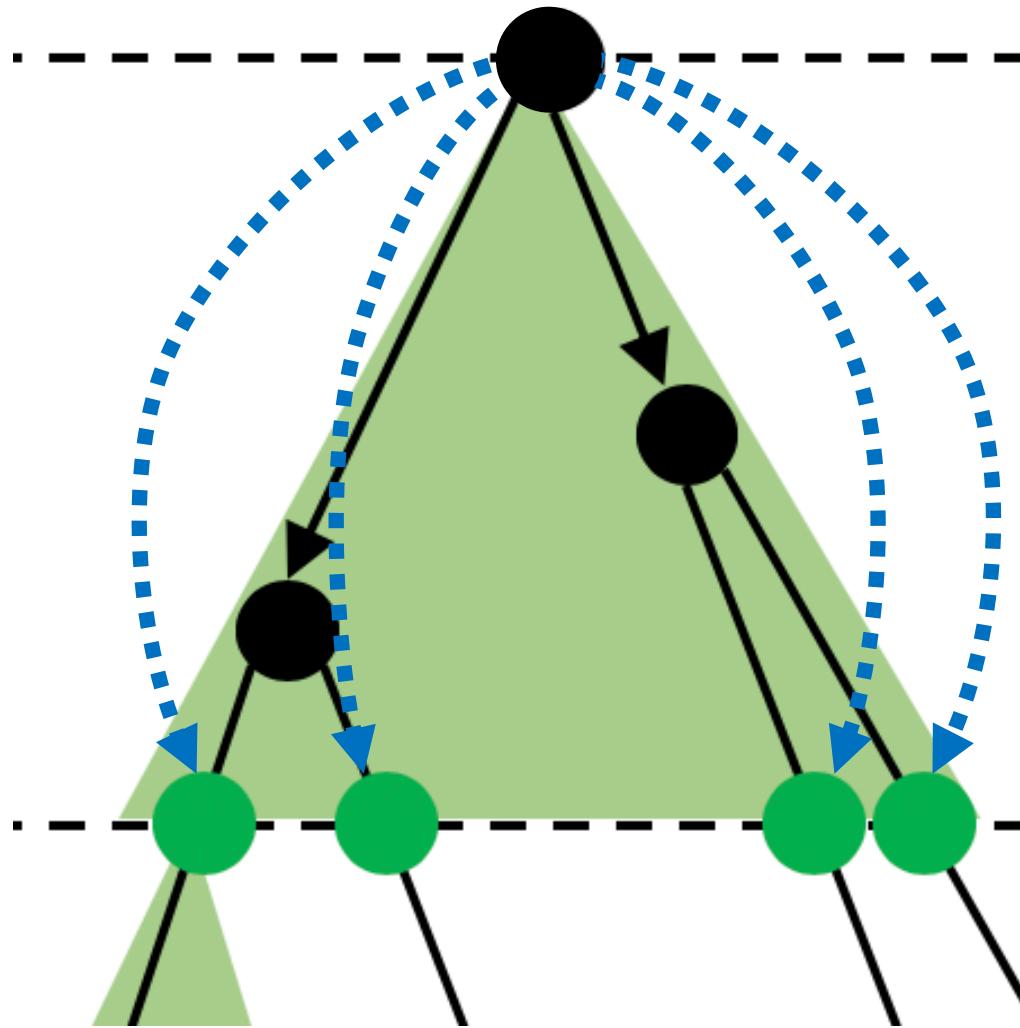
# Introduction of Micro Trie (3/4)

## C-Trie++



# Introduction of Micro Trie (4/4)

Equip each Micro Trie with a Hash Table

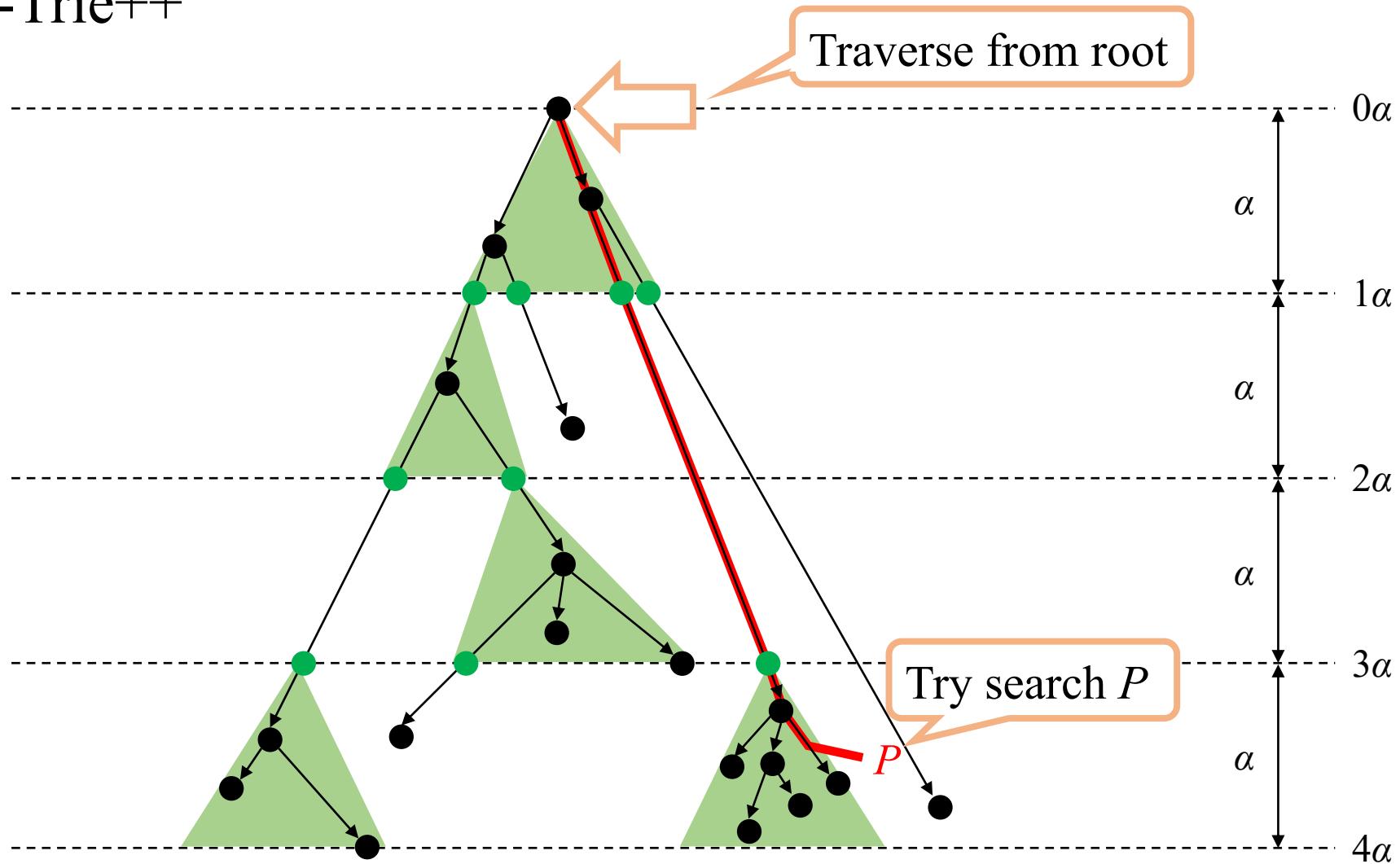


Hash Table:

- jump from root to leaf in  $O(1)$  expected time.
- key: string of length  $\alpha$
- value: leaf node

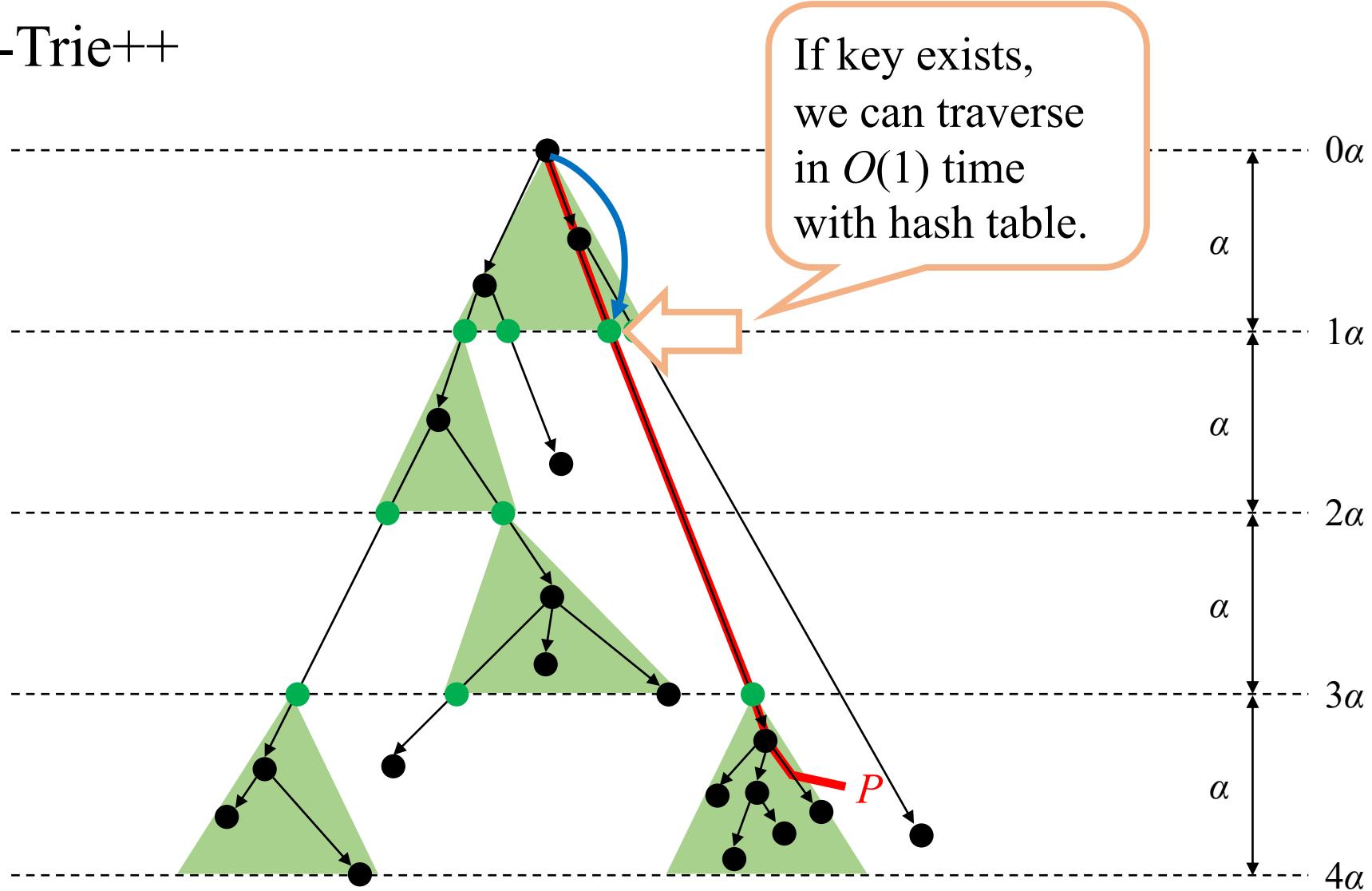
# Trie Traversal (1/5)

## C-Trie++



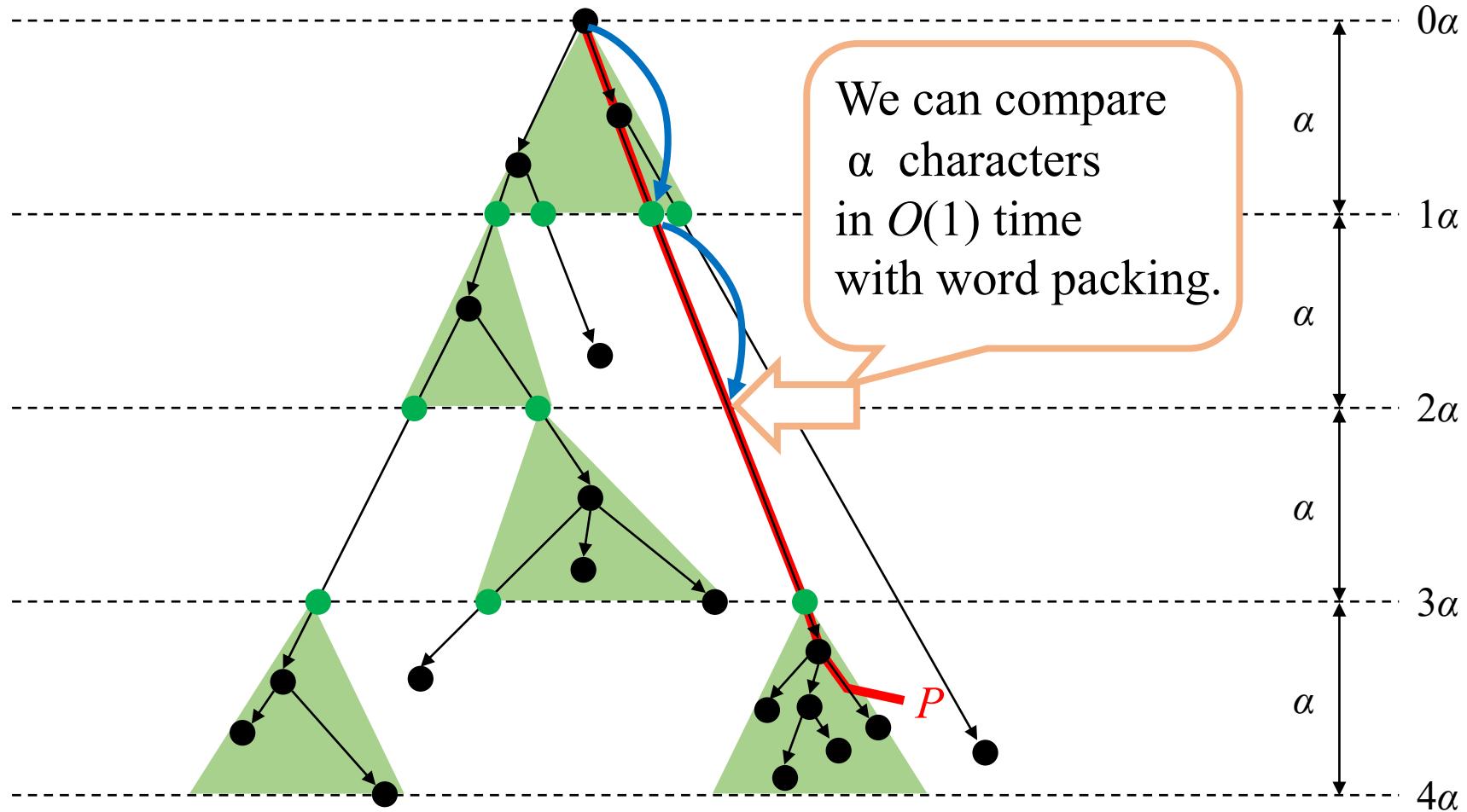
# Trie Traversal (2/5)

## C-Trie++



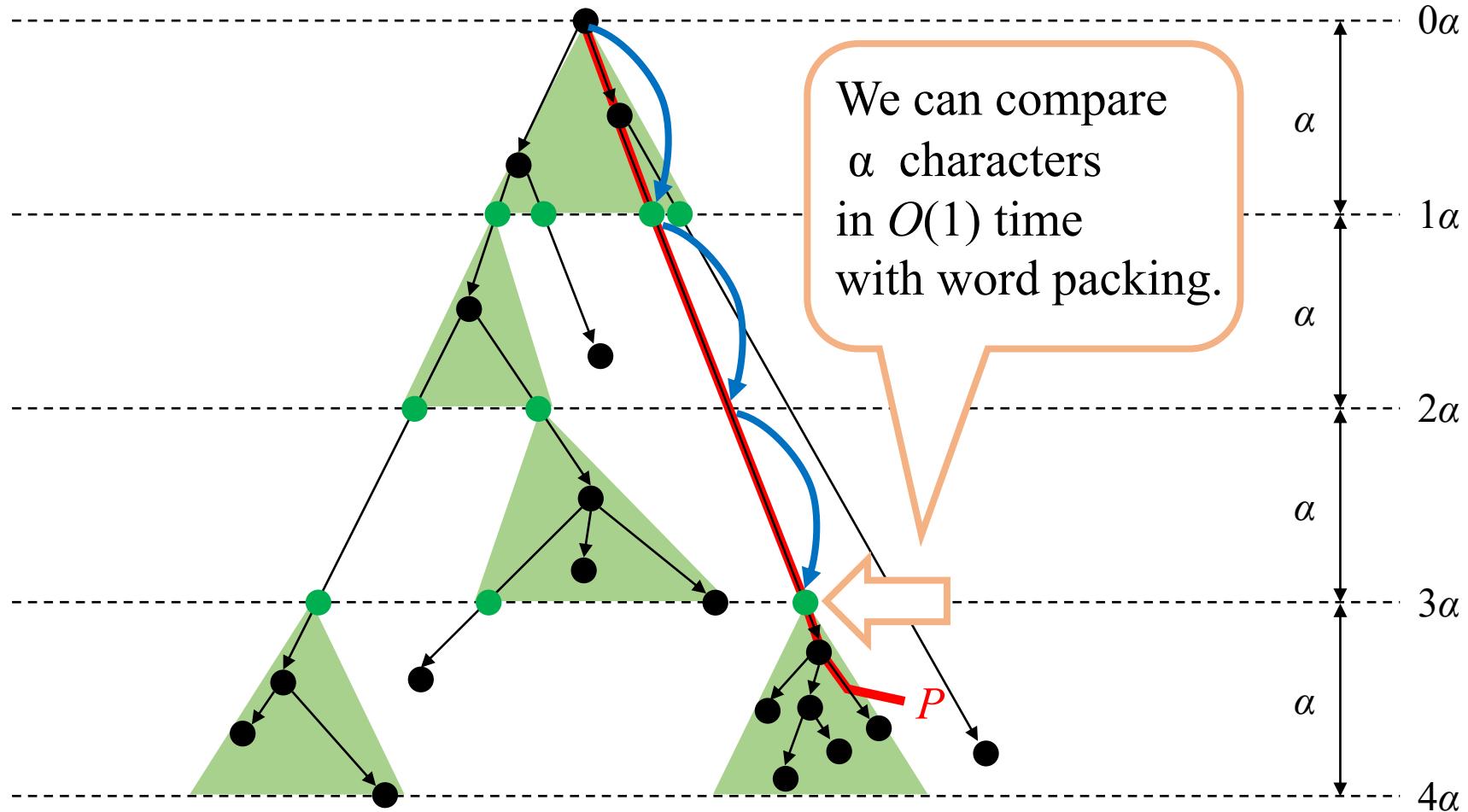
# Trie Traversal (3/5)

## C-Trie++



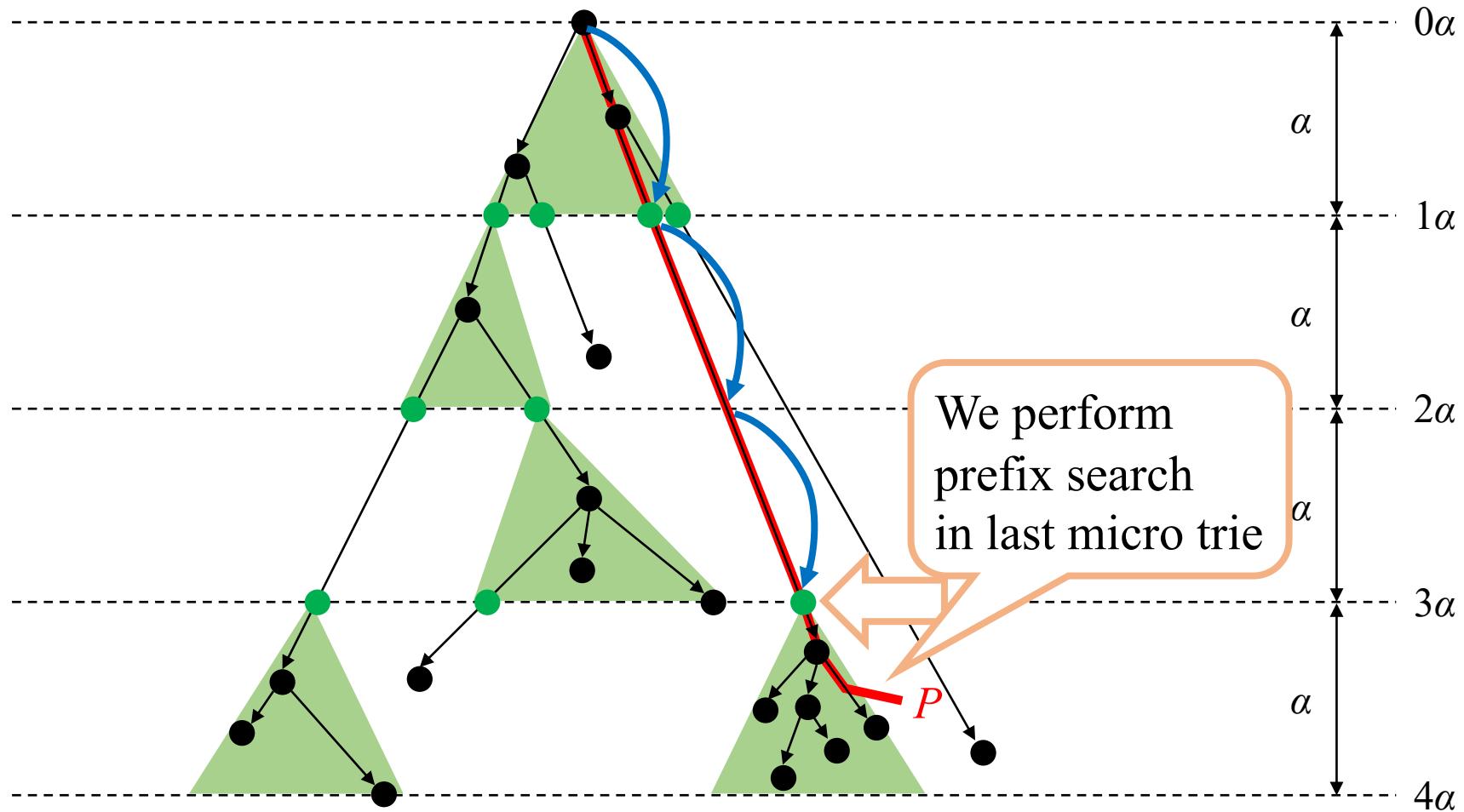
# Trie Traversal (4/5)

## C-Trie++



# Trie Traversal (5/5)

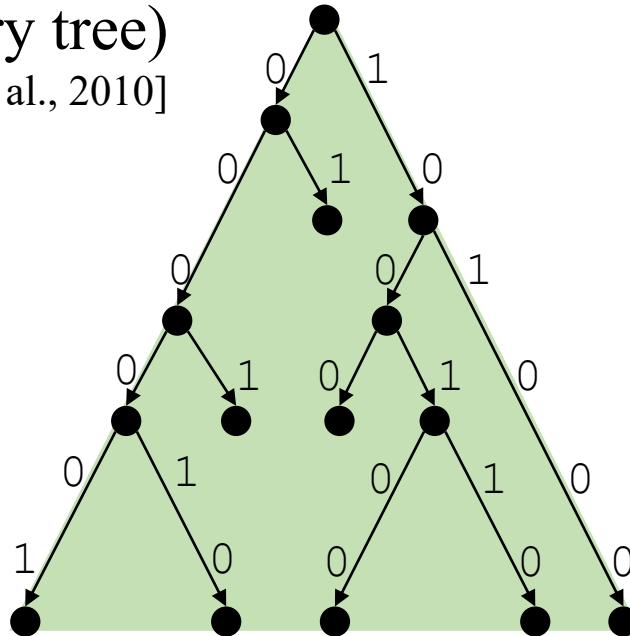
## C-Trie++



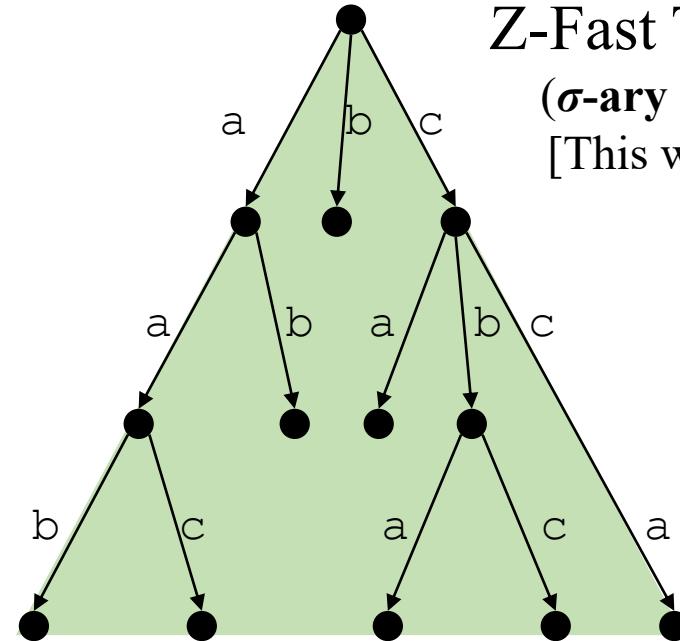
# Dynamic Prefix Search in Micro Tries

- We improve prefix search (expected) time in micro trie
  - Belazzougui et al., 2010 :  $O(\log(m \log \sigma) + occ)$
  - Takagi et al., 2017 :  $O(\log w + occ)$
  - This work :  $O(\log \min\{\alpha, m\} + occ)$

Z-Fast Trie  
(binary tree)  
[Belazzougui et al., 2010]



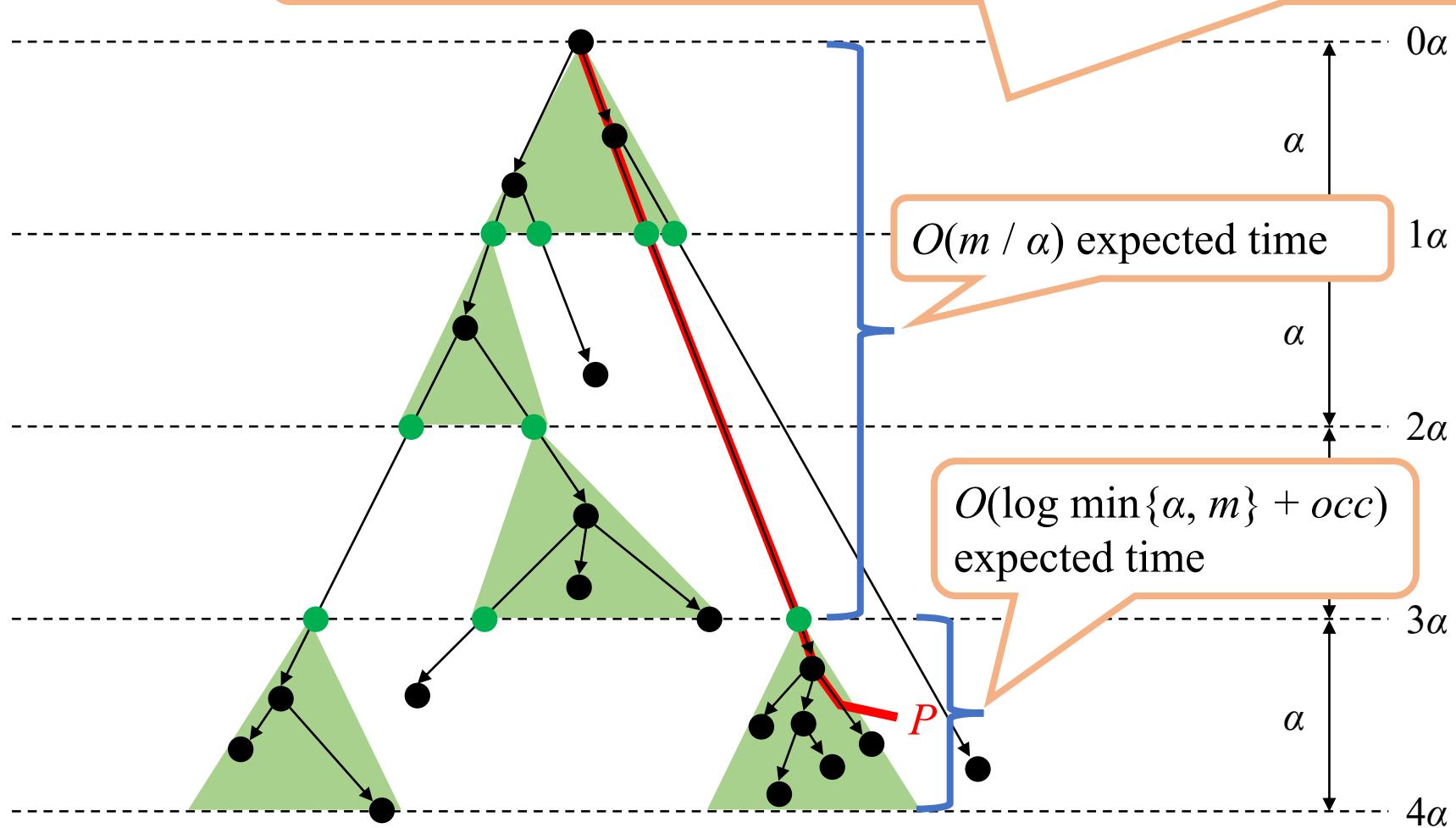
Alphabet Aware  
Z-Fast Trie  
( $\sigma$ -ary tree)  
[This work]



# Prefix Search Time

C-Trie++

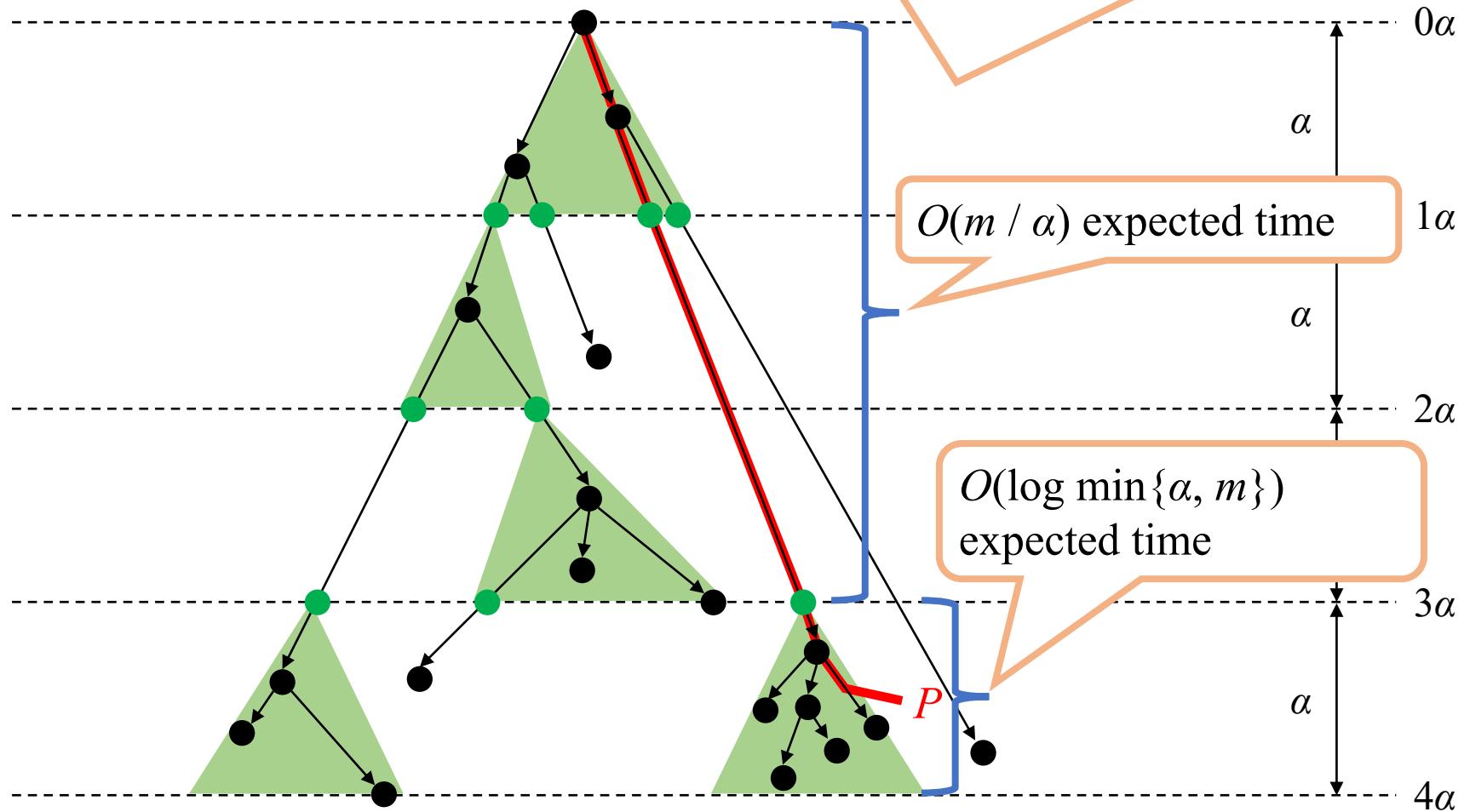
total :  $O(m / \alpha + \log \min\{\alpha, m\} + occ)$  expected time



# Insertion Time

C-Trie++

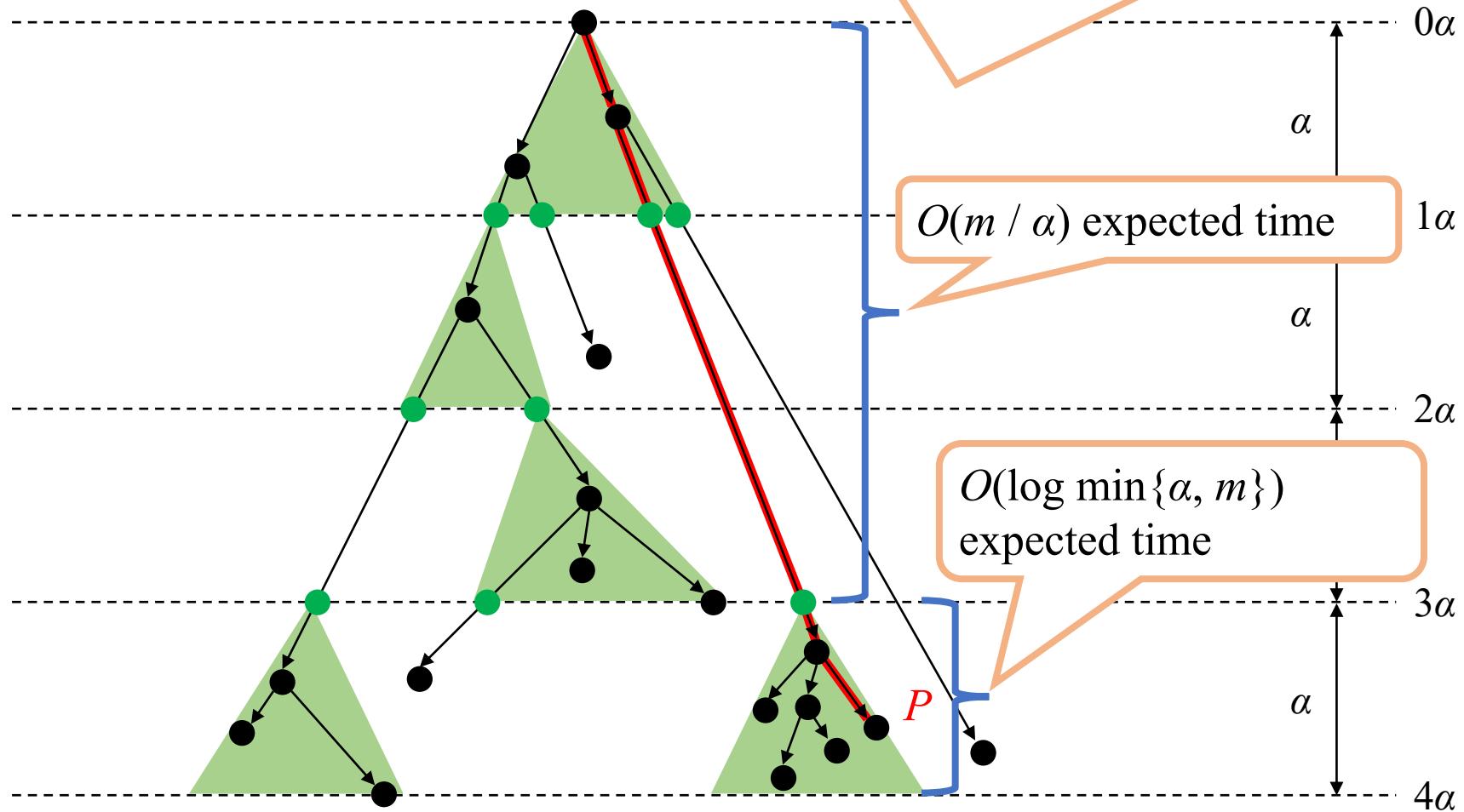
total :  $O(m / \alpha + \log \min\{\alpha, m\})$  expected time



# Deletion Time

C-Trie++

total :  $O(m / \alpha + \log \min\{\alpha, m\})$  expected time



# Experimental Setup

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- ❑ CPU : Intel Xeon X5560 @2.80 GHz
- ❑ Memory : 198GB
- ❑ OS : CentOS 6.10
- ❑ Language : C++
- ❑ Implementations
  - Compact Trie [Takagi et al.]
  - Z-Fast Trie [Ours]
  - Packed C-Trie [Takagi et al.]
  - **C-Trie++** [Ours]

# Datasets

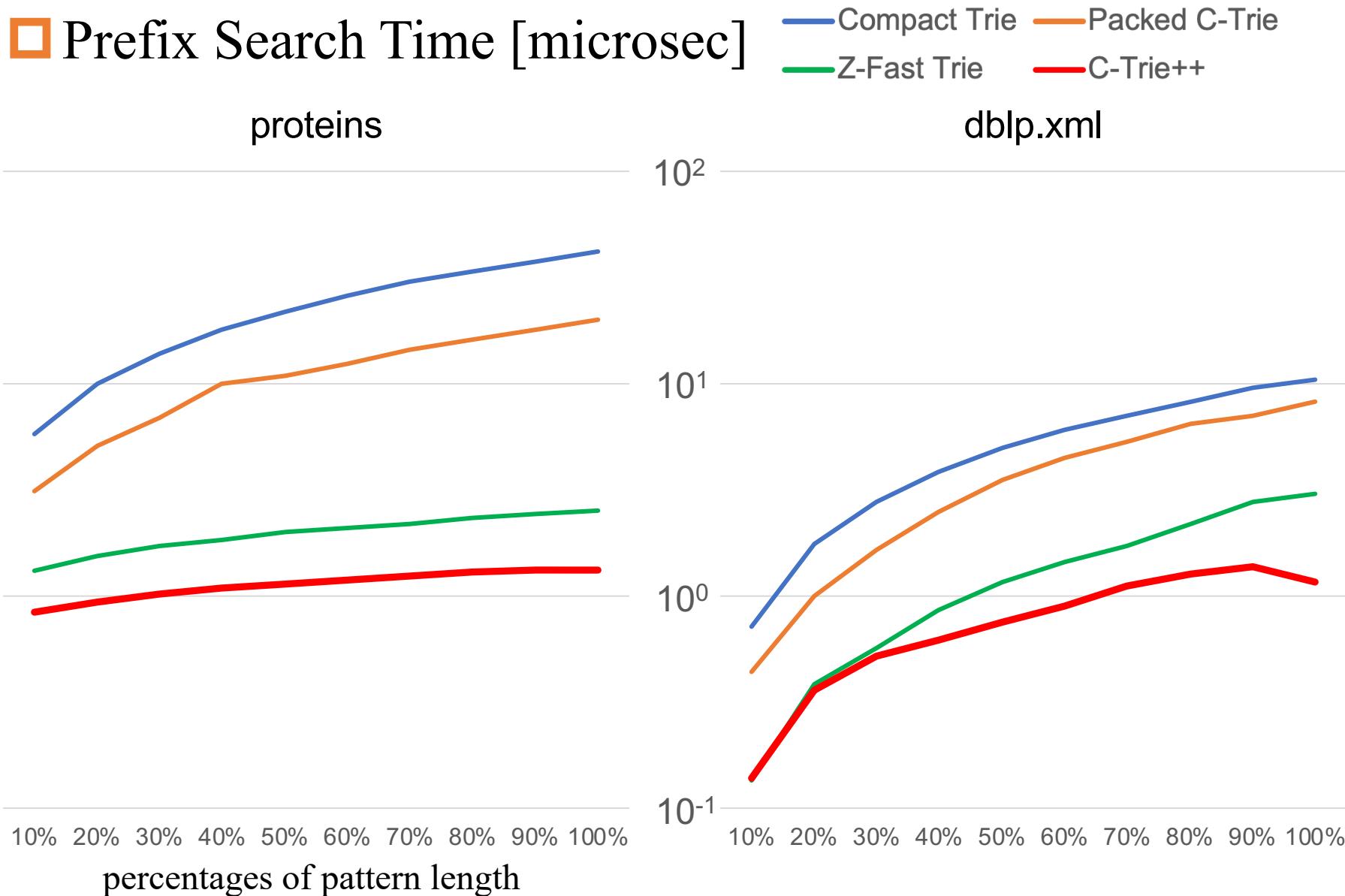
## □ Characteristics

Datasets	size[MB]	$\sigma$	$k[10^3]$	average length	avg. LCP	C-Tries nodes[ $10^3$ ]
proteins	864.14	26	2,982	302.8	38.8	5,778
dblp.xml	164.89	96	2,950	57.6	34.4	5,899

- We split a data set into strings at delimiters such as carriage returns, which form our input set  $S$ .
- In our experiment for prefix search, we took the prefixes of length 10%, 20%, ..., 100% of the strings of  $S$  as patterns, and measured the average query time.

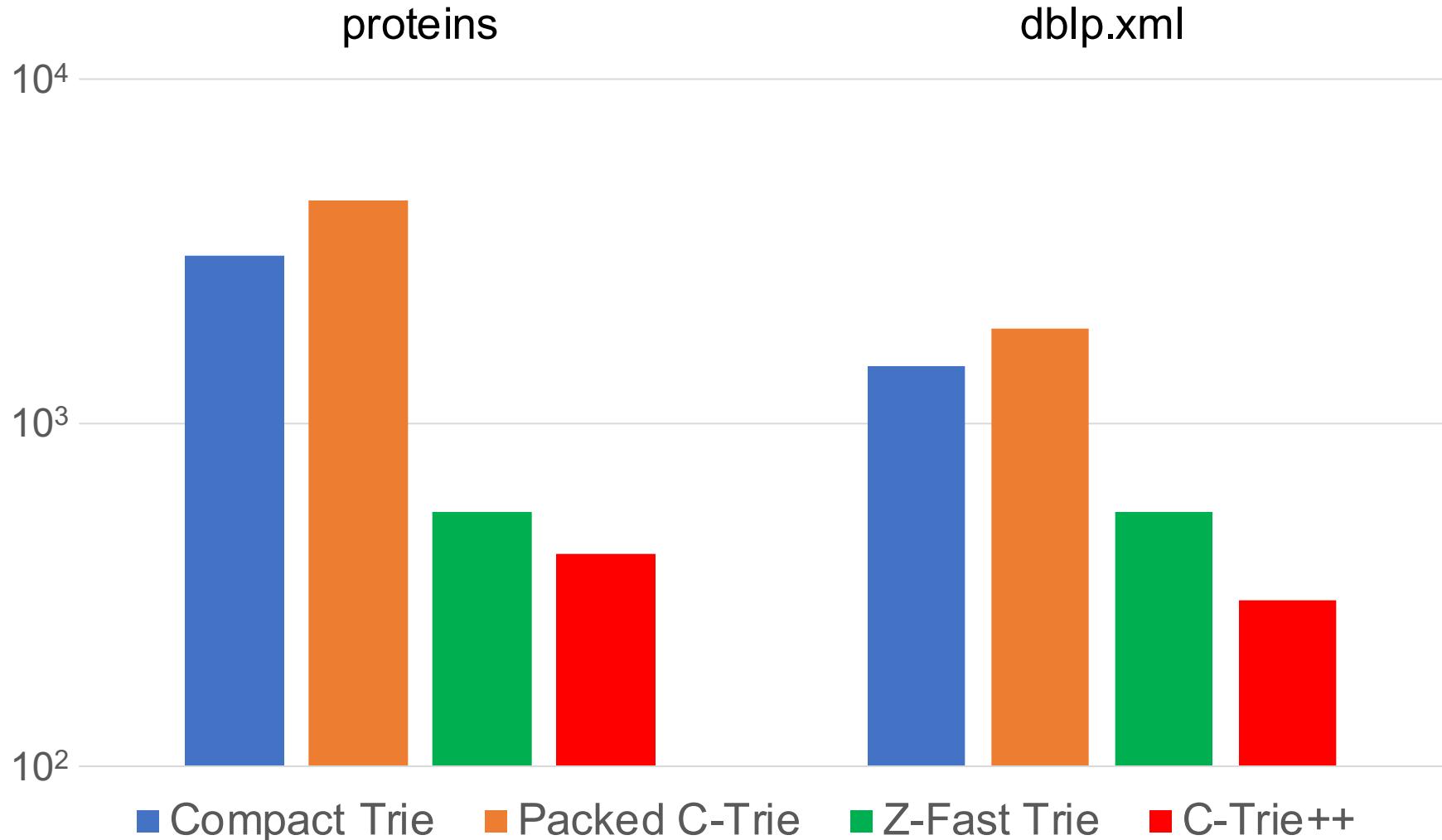
# Experimental Results

## Prefix Search Time [microsec]



# Experimental Results

## Memory Usage [MB]



# Conclusions

## □ Summary

- We proposed c-trie++:
  - ◆ Space:  $|T| \log \sigma + \Theta(kw)$  bits.
  - ◆ Prefix Search :  $O(m / \alpha + \log \min\{\alpha, m\} + occ)$  time.
  - ◆ Insert :  $O(m / \alpha + \log \min\{\alpha, m\})$  time.
  - ◆ Delete :  $O(m / \alpha + \log \min\{\alpha, m\})$  time.
- Our computational experiments support the claim that c-trie++ is the fastest trie for prefix search.

## □ Future Work

- Use SIMD instruction sets that allow larger machine word sizes (here  $w = 64$  bits).