

CS 240 – Data Structures and Data Management

Module 6: Dictionaries for special keys

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Based on lecture notes by many previous cs240 instructors

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Outline

6 Dictionaries for special keys

- Lower bound
- Interpolation Search
- Tries
 - Standard Tries
 - Variations of Tries
 - Compressed Tries
 - Multiway Tries

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Dictionary ADT: Implementations thus far

Realizations we have seen so far:

- **Balanced Binary Search trees** (AVL trees):
 $\Theta(\log n)$ search, insert, and delete (worst-case)
- **Skip lists**:
 $\Theta(\log n)$ search, insert, and delete (expected)
- Various other realizations sometimes faster on insert, but *search* always takes $\Omega(\log n)$ time.

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- Various other realizations sometimes faster on insert, but *search* always takes $\Omega(\log n)$ time.

Question: Can one do better than $\Theta(\log n)$ time for *search*?

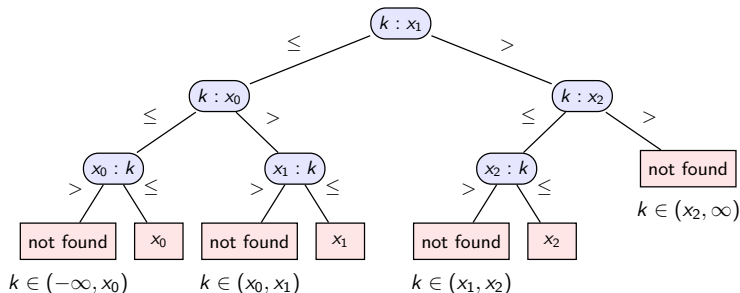
Answer: Yes and no! *It depends on what we allow.*

- No: Comparison-based searching lower bound is $\Omega(\log n)$.
- Yes: Non-comparison-based searching can achieve $o(\log n)$ (under restrictions!).

Lower bound for search

Theorem: Any *comparison-based* algorithm requires in the worst case $\Omega(\log n)$ comparisons to search among n distinct items.

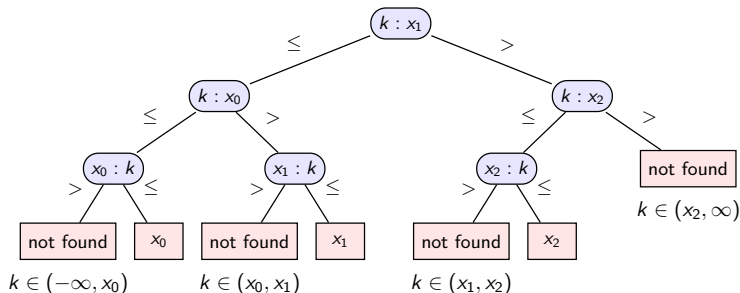
Proof: Via decision tree for items $x_0, \dots, x_{n-1}$ and search for k



Lower bound for search

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Proof: Via decision tree for items $x_0, \dots, x_{n-1}$ and search for k



- How many possible outcomes are there?
- What does that tell us about the height of the decision tree?

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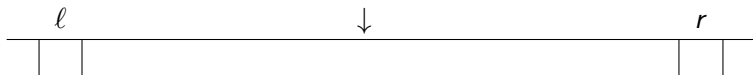
Towards interpolation search

We can match the lower bound asymptotically in a *sorted array*.

binary-search(A, n, k)

1. $\ell \leftarrow 0, r \leftarrow n - 1$
2. **while** ($\ell \leq r$)
3. $m \leftarrow \lfloor \frac{\ell+r}{2} \rfloor$
4. **if** ($A[m]$ equals k) **then return** “found at $A[m]$ ”
5. **else if** ($A[m] < k$) **then** $\ell \leftarrow m + 1$
6. **else** $r \leftarrow m - 1$
7. **return** “not found, but would be between $A[\ell-1]$ and $A[\ell]$ ”

binary-search: Compare at index $\lfloor \frac{\ell+r}{2} \rfloor = \ell + \lceil \frac{1}{2}(r - \ell - 1) \rceil$



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ℓ		$\downarrow$		r	
40				120	

Question: If keys are *numbers*, where would you expect key $k = 100$?

Interpolation search

- Code very similar to binary search, but compare at index

$$\ell + \left\lceil \frac{\overbrace{k - A[\ell]}^{\text{distance from left key}}}{\underbrace{A[r] - A[\ell]}_{\text{distance between left and right keys}}} \cdot \underbrace{(r - \ell - 1)}_{\# \text{ unknown keys in range}} \right\rceil$$

- Need a few extra tests to avoid crash during computation of m .

interpolation-search($A, n \leftarrow A.\text{size}, k$)

1. $\ell \leftarrow 0, r \leftarrow n - 1$
2. **while** ($\ell \leq r$)
3. **if** ($k < A[\ell]$ or $k > A[r]$) **return** “not found”
4. **if** ($k = A[r]$) **then return** “found at $A[r]$ ”
5. $m \leftarrow \ell + \left\lceil \frac{k - A[\ell]}{A[r] - A[\ell]} \cdot (r - \ell - 1) \right\rceil$
6. **if** ($A[m]$ equals k) **then return** “found at $A[m]$ ”
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Interpolation search: Example

0	1	2	3	4	5	6	7	8	9	10	11	12	13
0	10	20	30	40	50	71	110	112	114	116	118	119	120

interpolation-search(A[0..13],14,71):

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- $\ell = 0$, $r = n - 1 = 13$, $m = \ell + \lceil \frac{71-0}{120-0}(13-0-1) \rceil = \ell + 8 = 8$

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- $\ell = 0, r = 7, m = \ell + \lceil \frac{71-0}{110-0}(7-0-1) \rceil = \ell + 4 = 4$

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- $\ell = 5, r = 7, m = \ell + \lceil \frac{71-50}{110-50}(7-5-1) \rceil = \ell + 1 = 6$, found at A[6]

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If instead we had $A[6] = 72$:

- $\ell = 5 = r$, exit at line 3 with “not found”

Interpolation search: Example 2

0	1	2	3	4	5	6	7	8	9	10
0	1	2	3	4	5	6	7	8	9	1500
ℓ	$\uparrow$									r

interpolation-search(A[0..10],10):

- $\ell = 0$, $r = n - 1 = 10$, $m = \ell + \lceil \frac{10-0}{1500-0}(10-0-1) \rceil = \ell + 1 = 1$

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- ... in the worst case this can be very slow ($\Theta(n)$ time)

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- ... in the worst case this can be very slow ($\Theta(n)$ time)

But it works well on average:

- Can show (difficult): $T^{\text{avg}}(n) \leq T^{\text{avg}}(\sqrt{n}) + \Theta(1).$
- This resolves to $T^{\text{avg}}(n) \in O(\log \log n).$

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Review: Words

Scenario: Keys in dictionary are *words*. Need brief review.

Words (= strings): sequences of characters over alphabet Σ

- Typical alphabets: $\{0, 1\}$ ($\rightarrow$ bitstrings), ASCII, $\{C, G, T, A\}$
- Stored in an array: $w[i]$ gets i th character (for $i = 0, 1, \dots$)

0	1	2	3	4
b	e	a	r	\$

- Words have end-sentinel \$ (sometimes not shown)
- $w.size = |w| = \#$ non-sentinel characters: $|bear\$|=4$.

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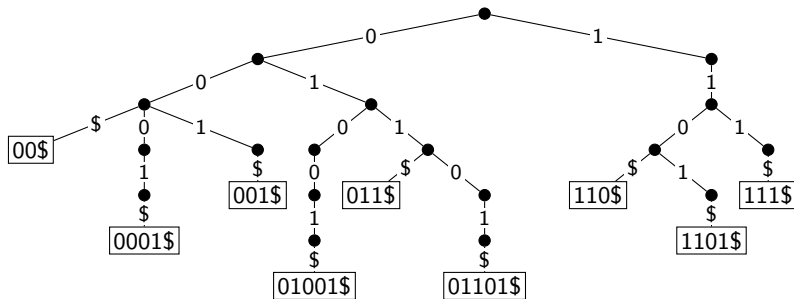
Should know:

- prefix, suffix, substring
- Sort words **lexicographically**: $be\$ <_{\text{lex}} bear\$ <_{\text{lex}} beer\$$
This is different from sorting numbers: $001\$ <_{\text{lex}} 010\$ <_{\text{lex}} 1\$$

Tries: Introduction

Trie (also know as **radix tree**): A dictionary for bitstrings.

- Comes from retrieval, but pronounced “try”
- A tree based on *bitwise comparisons*: Edge labelled with corresponding bit
- Similar to *radix sort*: use individual bits, not the whole key
- Due to end-sentinels, all key-value pairs are at leaves.



Tries: Search

- Follow links that correspond to current bits in w , keeping track of current depth d
- Repeat until no such link or w found at a leaf

Similar as for skip lists, we find search-path P separately first.

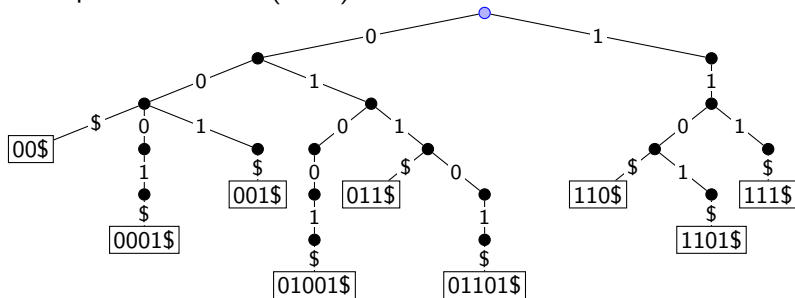
Trie::get-path-to(w)

Output: Stack with all ancestors of where w would be stored

1. $P \leftarrow$ empty stack; $z \leftarrow \text{root}$; $d \leftarrow 0$; $P.\text{push}(z)$
2. **while** $d \leq |w|$
3. **if** z has a child-link labelled with $w[d]$
4. $z \leftarrow$ child at this link; $d++$; $P.\text{push}(z)$
5. **else break**
6. **return** P

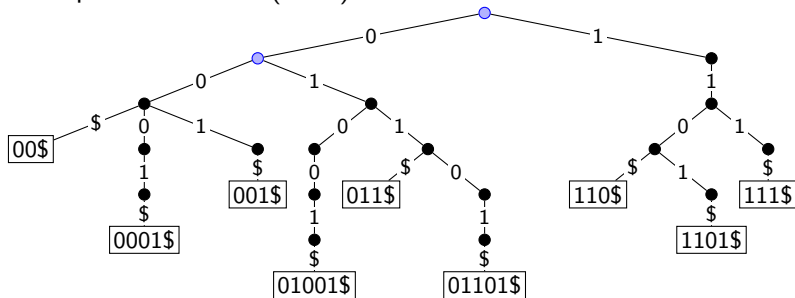
Tries search: Example

Example: Trie::search(011\$)



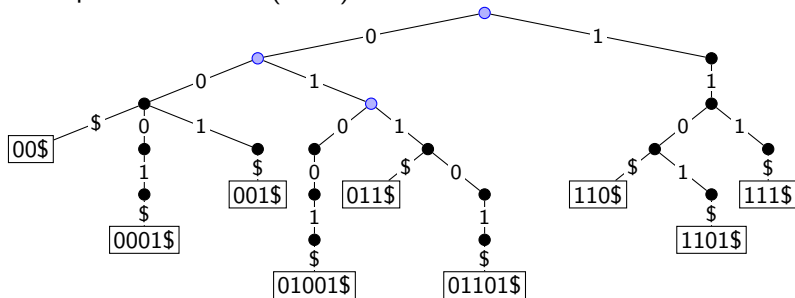
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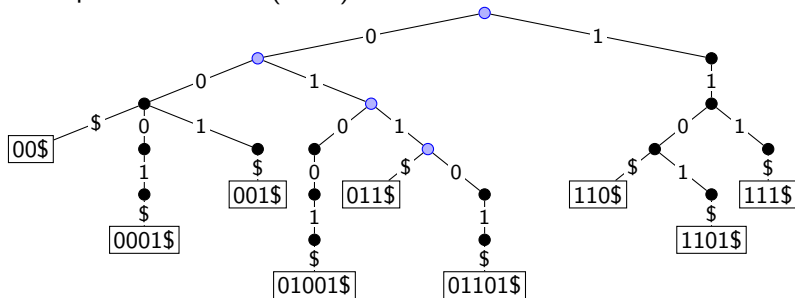
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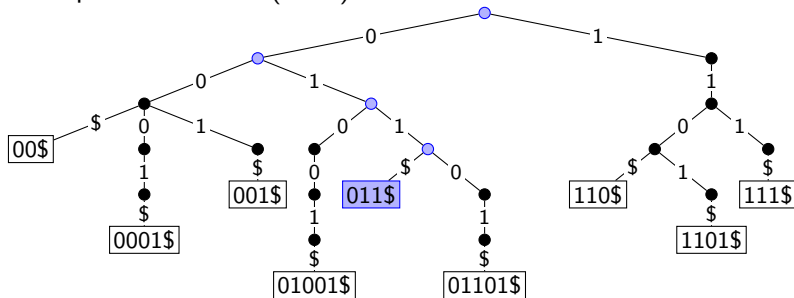
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Example: Trie::search(011\$)



Tries search: Example

Example: Trie::search(011\$)



Trie::search(w)

- ```

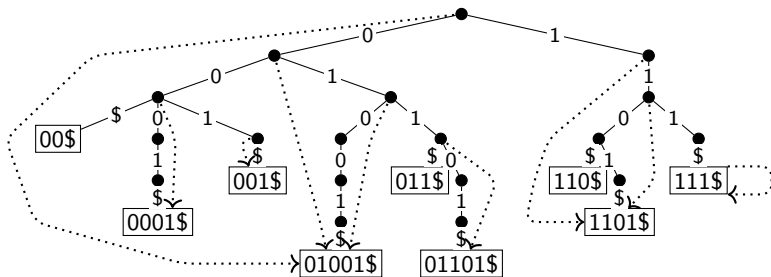
1. $P \leftarrow \text{get-path-to}(w), z \leftarrow P.\text{top}$
2. if (z is not a leaf) then
3. return “not found, would be in sub-trie of z ”
4. return key-value pair at z

```

# Tries: Prefix-search

For later applications of tries, we want another search-operation:

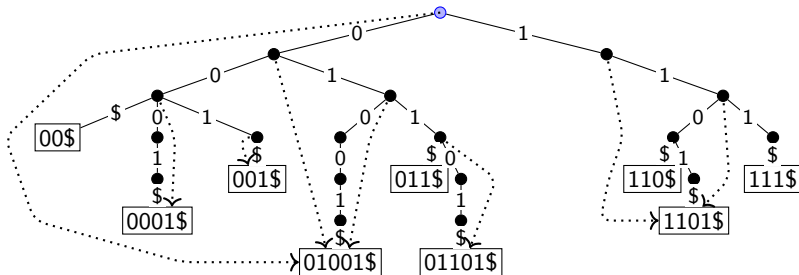
- *prefix-search*( $w$ ): Find **extension**(word for which  $w$  is a prefix).
- Testing whether extension exists is easy (how?)
- To find extension quickly, we need **leaf-references**
  - ▶ Every node  $z$  stores reference  $z.\text{leaf}$  to a leaf in subtree
  - ▶ **Convention**: store leaf with longest word



(not all leaf-references are shown)

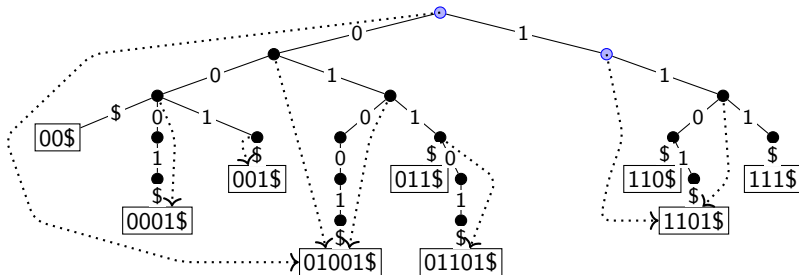
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Example: Trie::prefix-search(11\$)



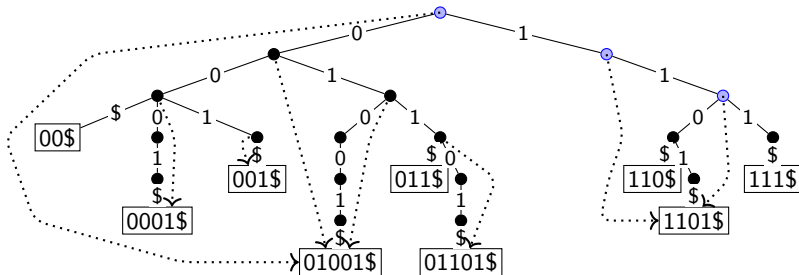
## Tries prefix-search: Example

Example: Trie::prefix-search(11\$)



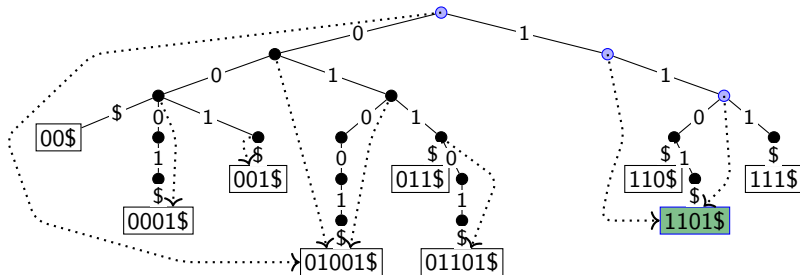
## Tries prefix-search: Example

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Example: Trie::prefix-search(11\$)



*Trie::prefix-search*(w)

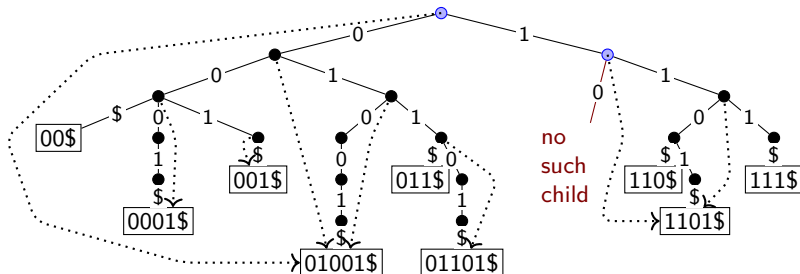
- ```

1.  $P \leftarrow \text{get-path-to}(w[0..|w|-1])$  // ignore end-sentinel
2. if number of nodes on  $P$  is at most  $|w|$ 
3.     return “no extension of  $w$  found”
4. return  $P.\text{top}().\text{leaf}$ 

```

Tries prefix-search: Example

Example: Trie::prefix-search(10\$)



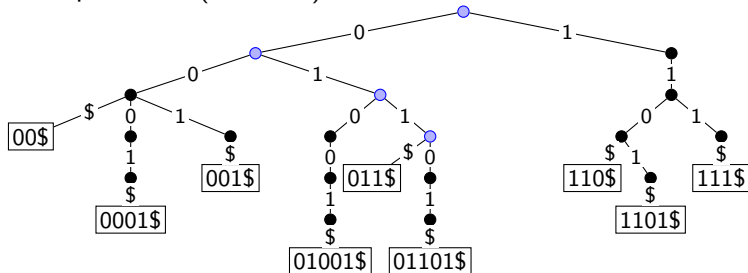
- Word 10\$ has size 2.
- *get-path-to*(10) returns stack with two nodes.
- We need more than $|w|$ nodes on P to have an extension.

Tries: Insert

Trie::insert(w)

- $P \leftarrow \text{get-path-to}(w)$ gives ancestors that exist already,
- Expand the trie from $P.\text{top}()$ by adding necessary nodes that correspond to extra bits of w .
- Update leaf-references (also at P if w is longer than previous leaves)

Example: *insert*(011101\$)

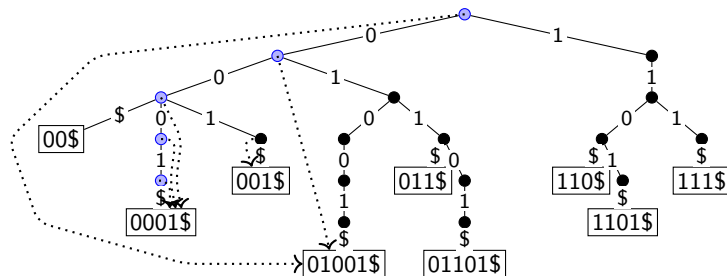


Tries: Delete

Trie::delete(w)

- $P \leftarrow \text{get-path-to}(w)$ gives all ancestors.
- Let ℓ be the leaf where w is stored
- Delete ℓ and nodes on P until ancestor that had two or more children.
- Update leaf-references on rest of P .
(If $z \in P$ referred to ℓ , find new $z.\text{leaf}$ from other children.)

Example: *trie::delete*(0001\$)



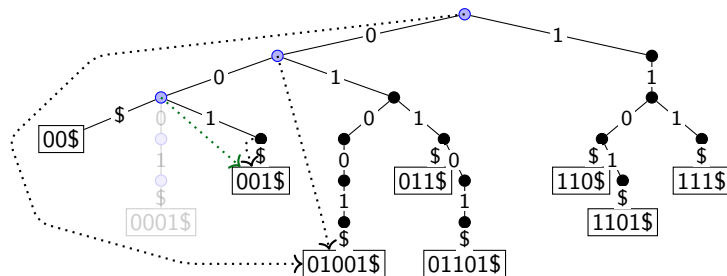
(only some leaf-references are shown)

Tries: Delete

trie::delete(w)

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(If $z \in P$ referred to ℓ , find new $z.\text{leaf}$ from other children.)

Example: *trie::delete*(0001\$)



(only some leaf-references are shown)

Binary tries: Summary

search(w), *prefix-search*(w), *insert*(w), *delete*(w) all take time $\Theta(|w|)$.

- Search-time is *independent* of number n of words stored in the trie!
- Search-time is small for short words.

The trie for a given set of words is unique

(except for order of children and ties among leaf-references)

Binary tries: Summary

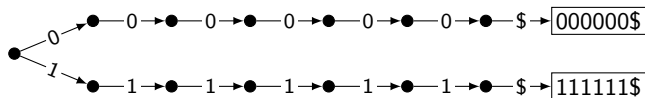
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Disadvantages:

- Tries can be wasteful with respect to space.

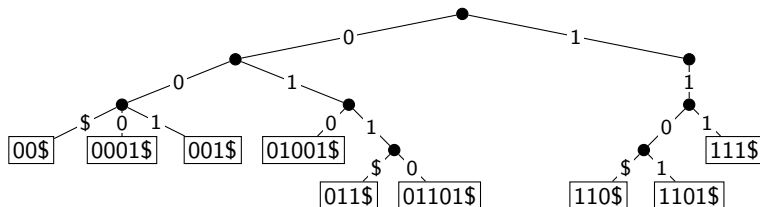


- Worst-case space is $\Theta(n \cdot (\text{maximum length of a word}))$
- What can we do to save space?

Variations of tries: Pruned tries

Pruned Trie: Stop adding nodes to trie as soon as the key is unique.

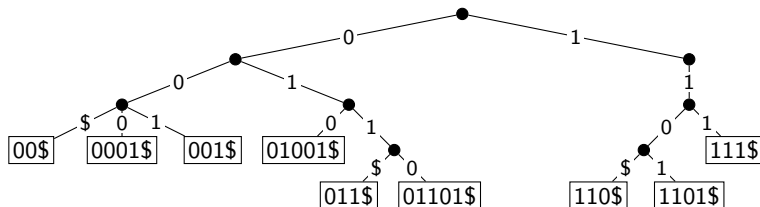
- A node has a child only if it has at least two descendants.
- Saves space if there are only few bitstrings that are long.



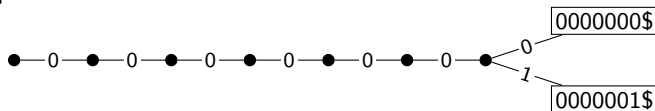
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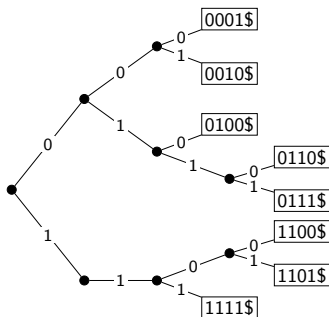
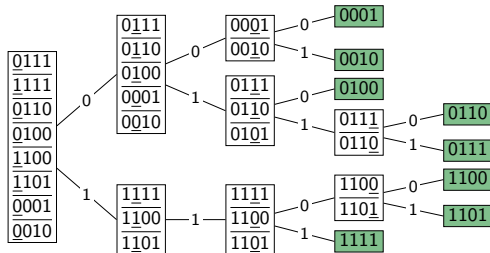
But the space can still be bad.



Pruned tries and MSD-radix sort

We have (implicitly) seen pruned tries before:

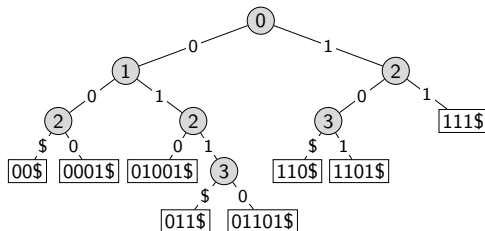
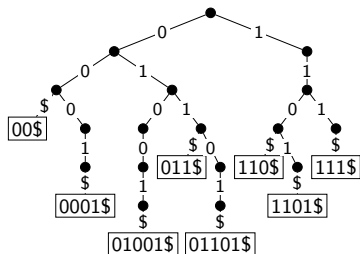
- For equal-length bitstrings:
Pruned trie equals recursion tree of MSD radix-sort.



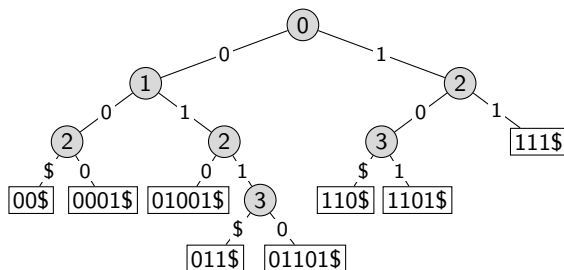
Compressed tries

Another (important!) variation:

- Compress paths of nodes with only one child.
- Each node stores an *index*, corresponding to the level of the node in the uncompressed trie. (On level d , we searched for link with $w[d]$.)

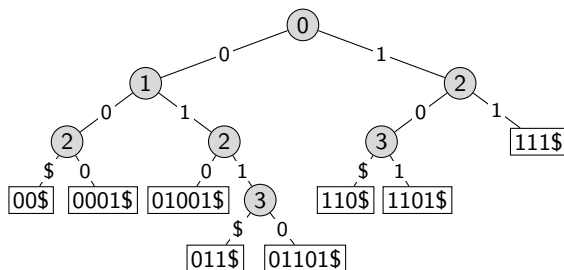


Compressed tries



- **Invariant:** At any node z where $d = z.\text{index}$, *all* words in the subtree rooted at z have the same initial d bits.

Compressed tries



- **Invariant:** At any node z where $d = z.\text{index}$, *all* words in the subtree rooted at z have the same initial d bits.
- **Observe:** Any compressed trie with n words has $O(n)$ nodes.
 - ▶ $\# \text{nodes} = \# \text{leaves} + \# \text{internal node}$.
 - ▶ Every internal node has two or more children.
 - ▶ **Can show:** Therefore more leaves than internal nodes.
 - ▶ So $\# \text{nodes} \leq 2n - 1$.

In particular we use $O(n)$ auxiliary space.

Compressed tries: Search

- As for tries, follow links that corresponds to current bits in w
- Main difference: stored indices say which bits to compare.
- Also: must compare w to word found at the leaf (why?)

CompressedTrie::get-path-to(w)

1. $P \leftarrow$ empty stack; $z \leftarrow \text{root}$; $P.\text{push}(z)$
2. **while** z is not a leaf and ($d \leftarrow z.\text{index} \leq |w|$) **do**
3. **if** (z has a child-link labelled with $w[d]$) **then**
4. $z \leftarrow$ child at this link; $P.\text{push}(z)$
5. **else break**
6. **return** P

CompressedTrie::search(w)

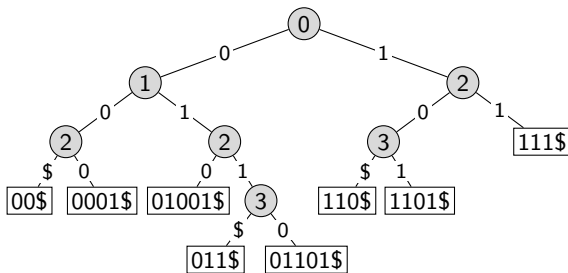
1. $P \leftarrow \text{get-path-to}(w)$, $z \leftarrow P.\text{top}$
2. **if** (z is not a leaf or word stored at z is not w) **then**
3. **return** “not found”
4. **return** key-value pair at z

Compressed tries search: Example

Example 1: CompressedTrie::search(

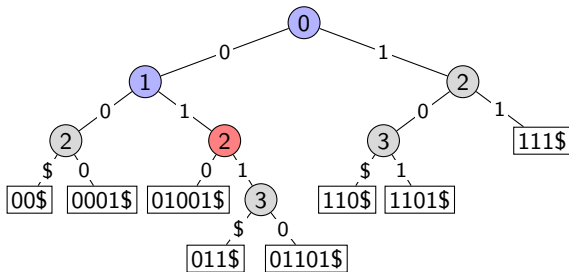
0	1	\$
---	---	----

)



Compressed tries search: Example

Example 1: CompressedTrie::search($\begin{matrix} 0 & 1 & 2 \\ \boxed{0} & \boxed{1} & \boxed{\$} \end{matrix}$) **unsuccessful** (no \$-child)

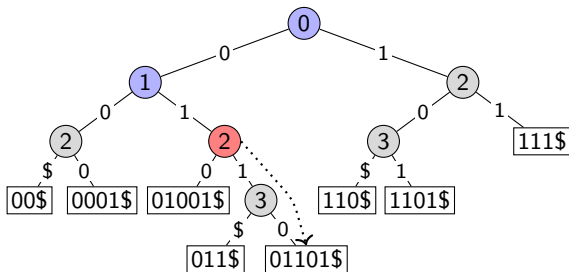


Compressed tries search: Example

Example 1: CompressedTrie::search(

0	1	2
0	1	\$

) **unsuccessful** (no \$-child)



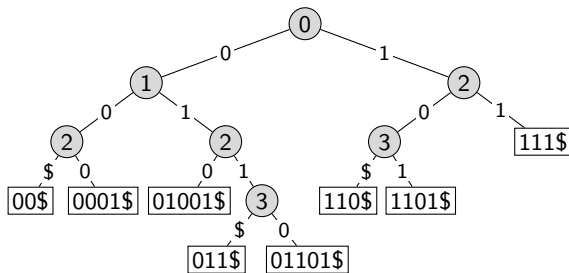
prefix-search(w): Compare w to z.leaf at last visited node z.

Compressed tries search: Example

Example 2: CompressedTrie::search(

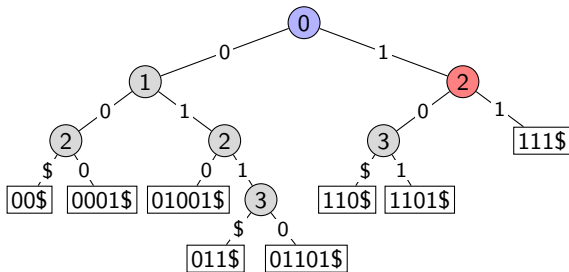
1	\$
---	----

)



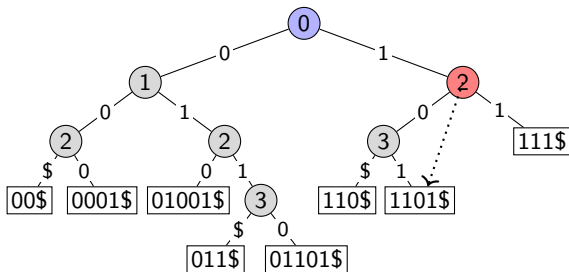
Compressed tries search: Example

Example 2: CompressedTrie::search($\begin{matrix} 0 & 1 \\ \boxed{1} & \boxed{\$} \end{matrix}$) **unsuccessful** (d too big)



Compressed tries search: Example

Example 2: CompressedTrie::search($\overset{0}{\boxed{1}} \overset{1}{\boxed{\$}}$) **unsuccessful** (d too big)



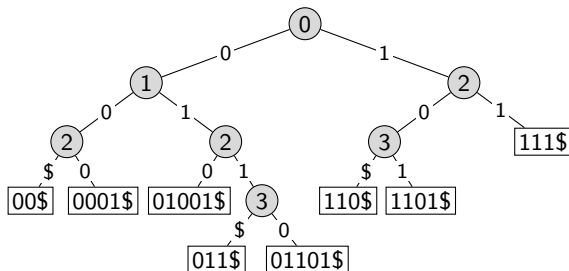
prefix-search(w): Compare w to z . *leaf* at last visited node z .

Compressed tries search: Example

Example 3: CompressedTrie::search(

0	1	2	3
1	0	1	\$

)

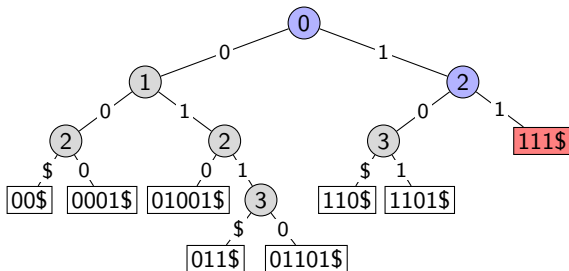


Compressed tries search: Example

Example 3: CompressedTrie::search(

0	1	2	3
1	0	1	\$

) **unsuccessful**
(wrong word at leaf)

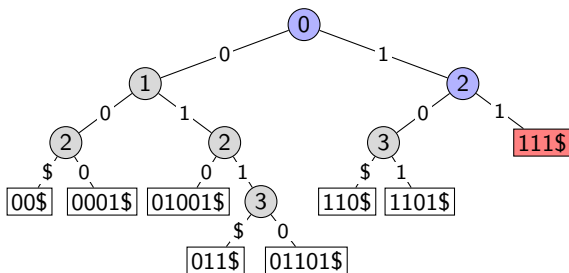


Compressed tries search: Example

Example 3: CompressedTrie::search(

0	1	2	3
1	0	1	\$

) **unsuccessful**
(wrong word at leaf)



prefix-search(w): Compare w to word at reached leaf.

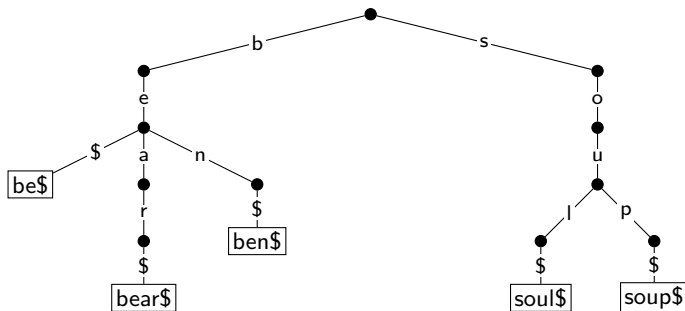
Compressed tries: Summary

- *search*(w) and *prefix-search*(w) are fairly easy.
- *insert*(w) and *delete*(w) are conceptually simple by *uncompressing*:
 - ▶ Search for path P to word w (say we reach node z)
 - ▶ Uncompress this path (using characters of z .*leaf*)
 - ▶ Insert/Delete w as in an uncompressed trie.
 - ▶ Compress path from root to where change happened(Pseudocode gets more complicated and is omitted.)
- All operations take $O(|w|)$ time for a word w .
- Compressed tries use $O(n)$ space (better than other trie-variants)

Overall, code is more complicated, but space-savings are worth it if words are unevenly distributed.

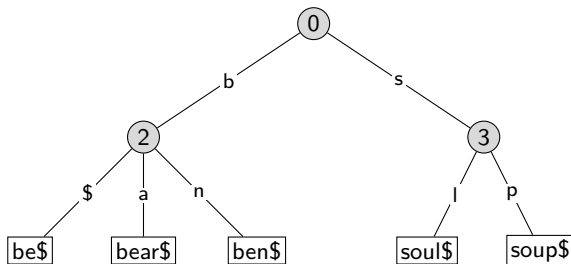
Multiway tries for larger alphabets

- To represent *strings* over any *fixed alphabet* Σ
- Any node will have at most $|\Sigma| + 1$ children (one child for the end-of-word character \$)
- Example: A trie holding strings {bear\$, ben\$, be\$, soul\$, soup\$}



Compressed multiway tries

- **Variation:** Compressed multi-way tries: compress paths as before
- Example: A compressed trie holding strings {bear\$, ben\$, be\$, soul\$, soup\$}

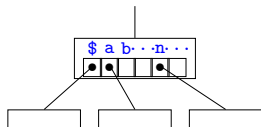


Multiway tries: Summary

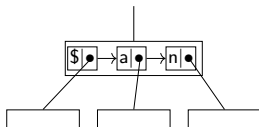
- Operations *search*(w), *prefix-search*(w), *insert*(w) and *delete*(w) are exactly as for tries for bitstrings.
- Run-time $O(|w| \cdot (\text{time to find the appropriate child}))$

Multiway tries: Summary

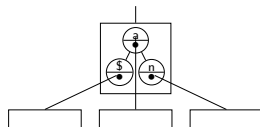
- Operations *search*(w), *prefix-search*(w), *insert*(w) and *delete*(w) are exactly as for tries for bitstrings.
- Run-time $O(|w| \cdot (\text{time to find the appropriate child}))$
- Each node now has up to $|\Sigma| + 1$ children. How should they be stored?



Array?



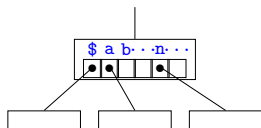
List?



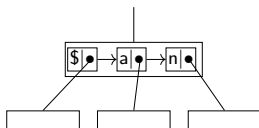
Dictionary?

Multiway tries: Summary

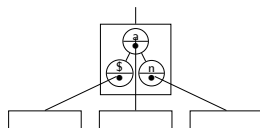
- Operations *search*(w), *prefix-search*(w), *insert*(w) and *delete*(w) are exactly as for tries for bitstrings.
- Run-time $O(|w| \cdot (\text{time to find the appropriate child}))$
- Each node now has up to $|\Sigma| + 1$ children. How should they be stored?



Array?



List?



Dictionary?

- Time/space tradeoff: arrays are fast, lists are space-efficient.
- Dictionary best in theory, not worth it in practice unless $|\Sigma|$ is huge.
- In practice, use *direct addressing* or *hashing* ($\rightarrow$ module 07).