

Representing and Searching Sets of Strings

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- Radix search
- Ternary search tries

Sets of Strings

- Several very important applications

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E.g.,

- ▶ dictionary (of words)
- ▶ symbol table in a compiler
- ▶ all kinds of key-based index
- ▶ ...

- Operations

■ Operations

- ▶ `insert(Key)`
- ▶ `delete(Key)`
- ▶ `search(Key)`
- ▶ `min()`
- ▶ `max()`

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- Operations
 - ▶ insert(Key)
 - ▶ search(Key)
- No *delete* operation
- Built once and searched many times

BINARYSEARCH(A, K)

```
1 first = 1
2 last = length(A)
3 while first ≤ last
4      $x = \lceil (first + last) / 2 \rceil$ 
5     if  $A[x] == K$ 
6         return TRUE
7     elseif first == last
8         return FALSE
9     elseif  $A[x] > K$ 
10        last =  $x - 1$ 
11    else first =  $x + 1$ 
12 return FALSE
```

TREE-SEARCH(T, K)

```
1  $x = T.root$ 
2 while  $x \neq NIL$  and  $K \neq x.key$ 
3     if  $K < x.key$ 
4          $x = x.left$ 
5     else  $x = x.right$ 
6 if  $x \neq NIL$ 
7     return TRUE
8 else return FALSE
```

- Complexity?

Binary Search

K is a string!

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```

■ Complexity?

- ▶ we must account for the *complexity of string comparisons*

String Comparison

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- Assuming a string is an array of bytes, the condition $A[x] == K$ (line 5 of **BINARYSEARCH**) becomes **STRINGEQUALS**($A[x], K$)

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- The complexity of **STRINGEQUALS**(S_1, S_2) is $O(m)$, where m is the max string size
- So, the complexity of **BINARYSEARCH**(A, K) is $O(m \log n)$

What About a Hash Table

CHAINED-HASH-SEARCH(T, K)

```
1  $L = T[h(K)]$   
2 return LIST-SEARCH( $L, K$ )
```

HASH-SEARCH(T, K)

```
1 for  $i = 1$  to  $length(T)$   
2      $j = h(K, i)$   
3     if  $T[j] == K$   
4         return TRUE  
5     if  $T[j] == NIL$   
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■ Complexity?

- ▶ here, too, we must account for the string comparisons
- ▶ and for the hash functions

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 - ▶ problem is, **STRINGEQUALS**($A[x], K$) is likely to go through the same prefix of K many times

- So, since $m = \Theta(\log N)$, and **BINARYSEARCH**(A, K) uses $\Theta(\log N)$ comparisons each one running in $O(m)$:

$$T(N, m) = O(\log^2 N)$$

A New Data Structure

- Idea: a data structure where *common prefixes* are shared

- Data structure useful for information retrieval (pronounced “try” to distinguish it from a tree...)

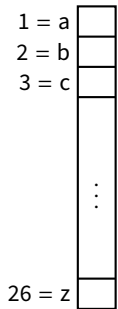
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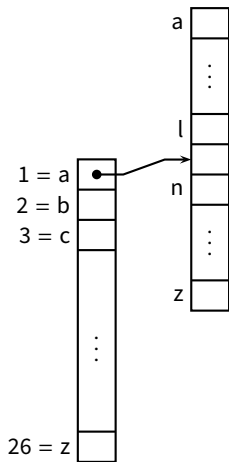
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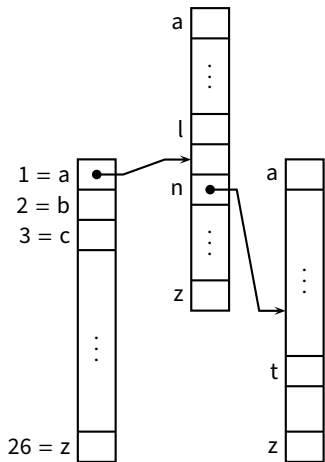
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 - ▶ one way would be to hold $|\Sigma|$ links
 - ▶ one for each character of the given alphabet Σ



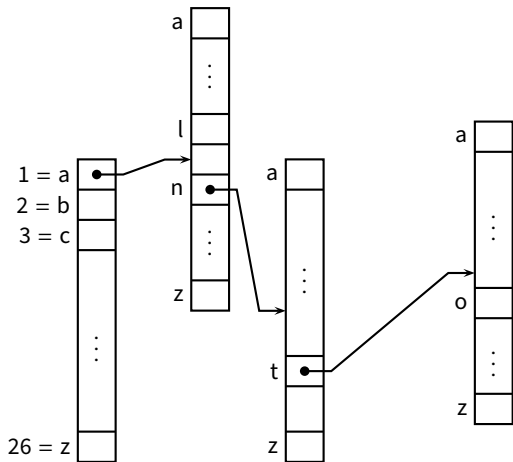
Radix Trie



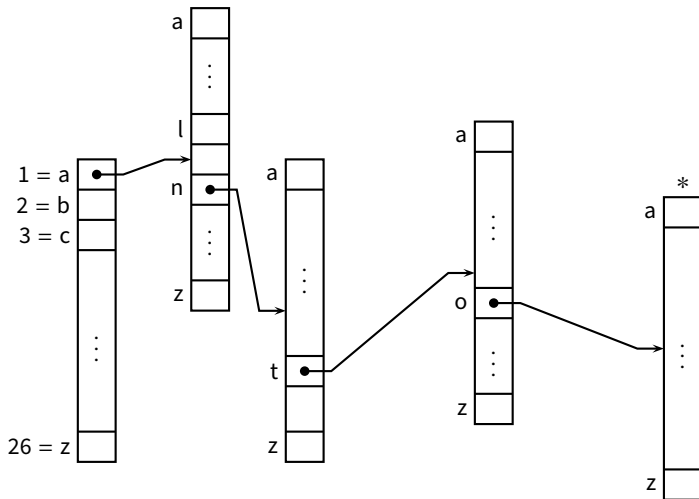
Radix Trie



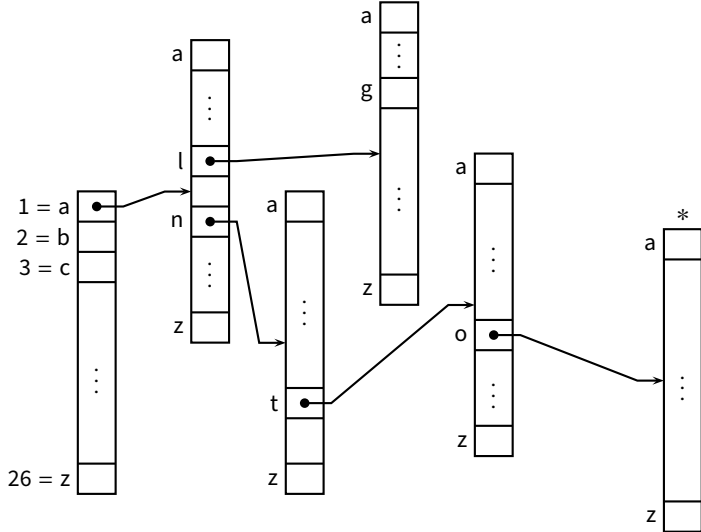
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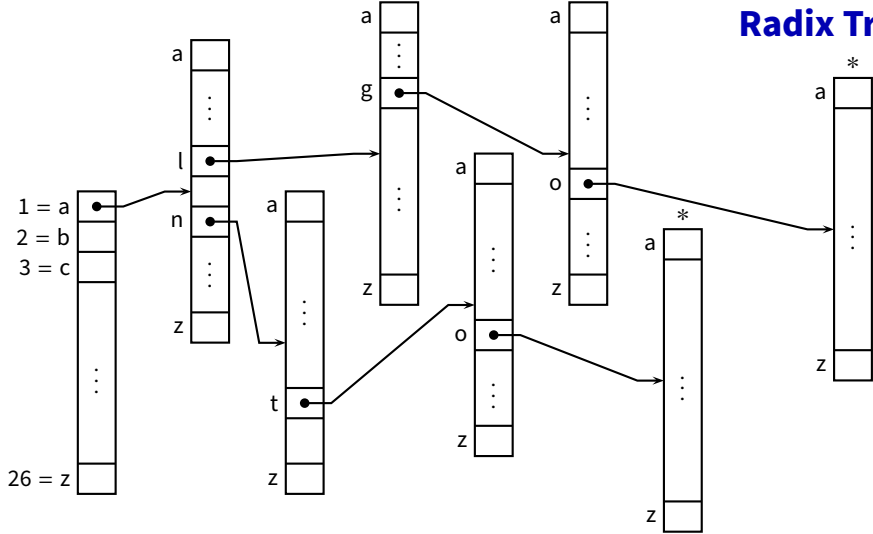
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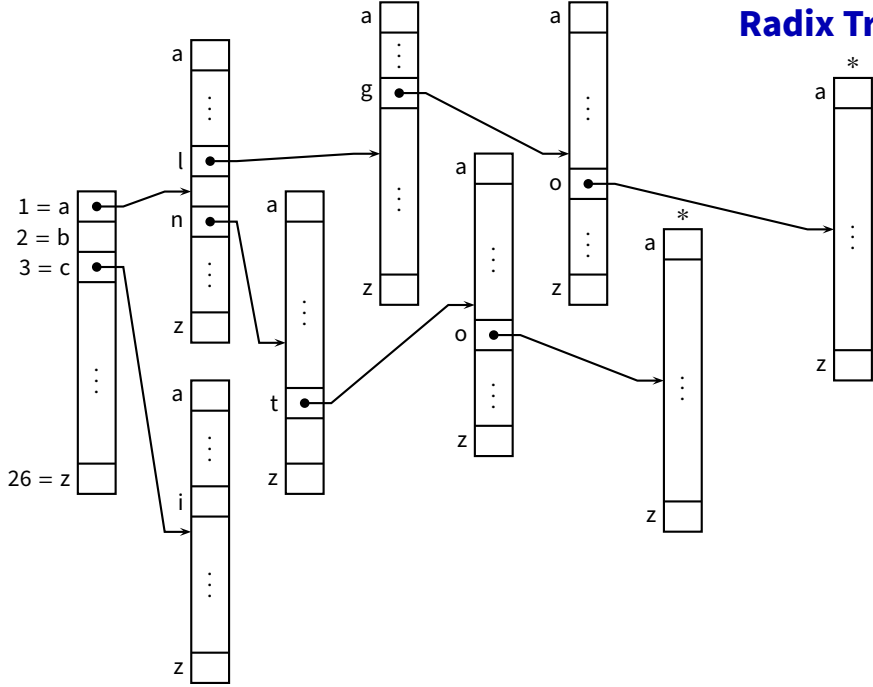
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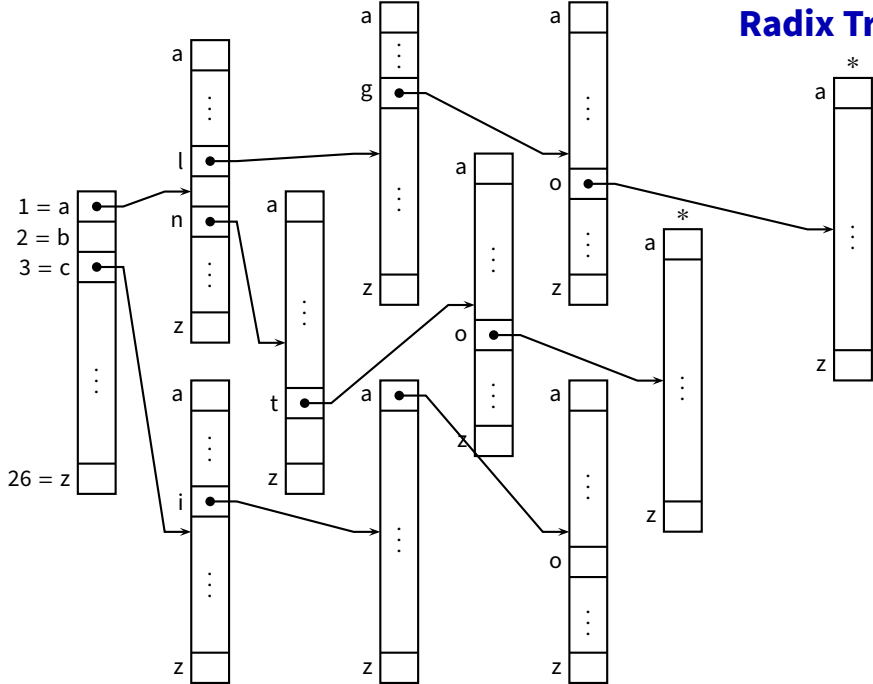
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- Every element x has an array of links $x.links$
 - ▶ e.g., in “radix-256,” an element represents a *byte* in a string (of bytes)
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```
RADIXSEARCH( $Root, K$ )  
1   $n = Root$   
2  for  $i = 1$  to  $length(K)$   
3      if  $n.links[K[i]] == NIL$   
4          return FALSE  
5      else  $n = n.links[K[i]]$   
6  return  $n.value$ 
```

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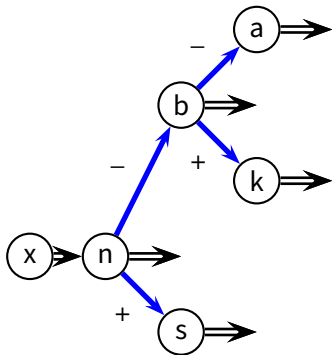
- ▶ a better characterization (*Exercise*: figure this out!):

$$S(N, m) = \Theta \left(|\Sigma| \left[\frac{N-1}{|\Sigma|-1} + N \left(m - \frac{\log N}{\log |\Sigma|} \right) \right] \right)$$

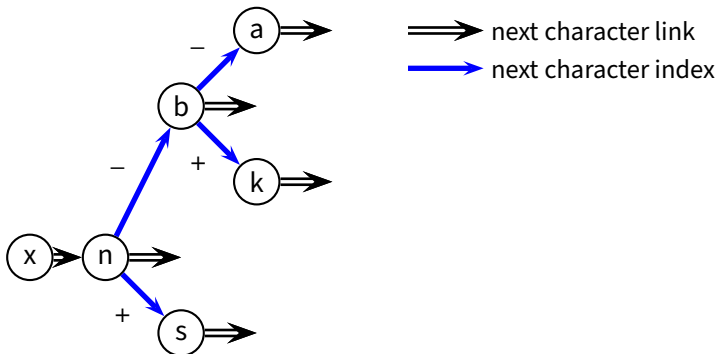
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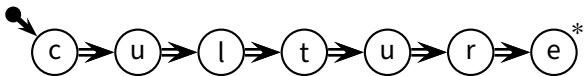
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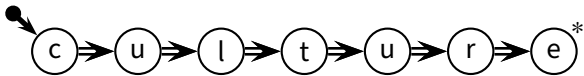
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- A node *n* has three links
 - ▶ *n.lower* links to a node representing a “lower” character at the same position
 - ▶ *n.higher* links to a node representing a “higher” character at the same position
 - ▶ *n.equal* links to a node representing a character in the next position

“culture”

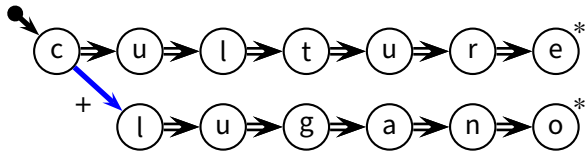
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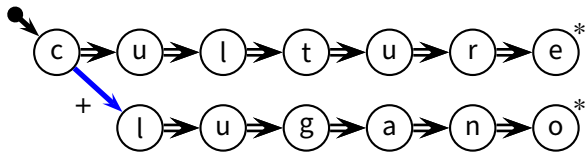
“lugano”



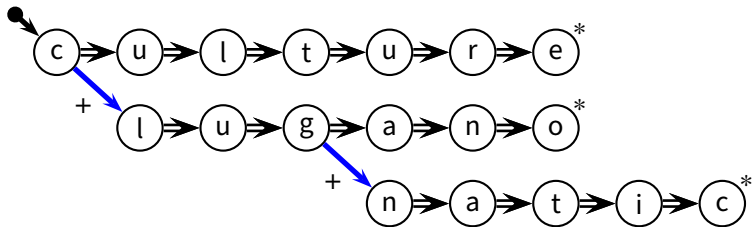
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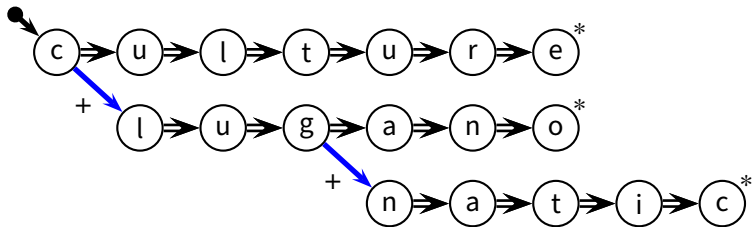
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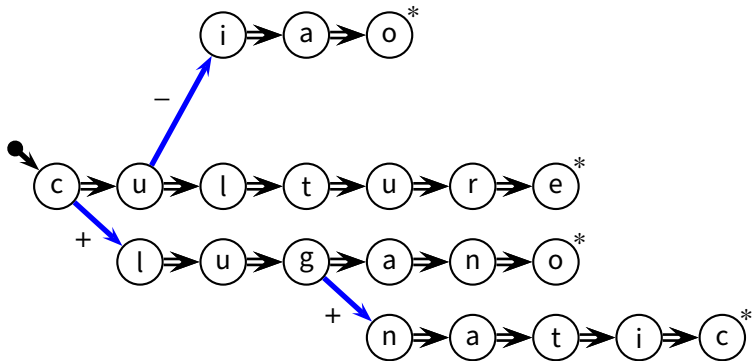
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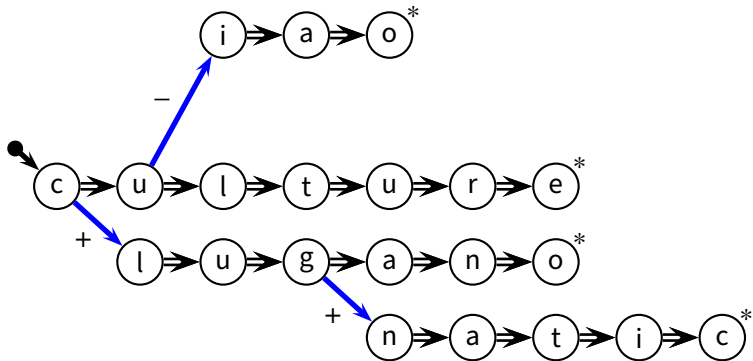
“ciao”



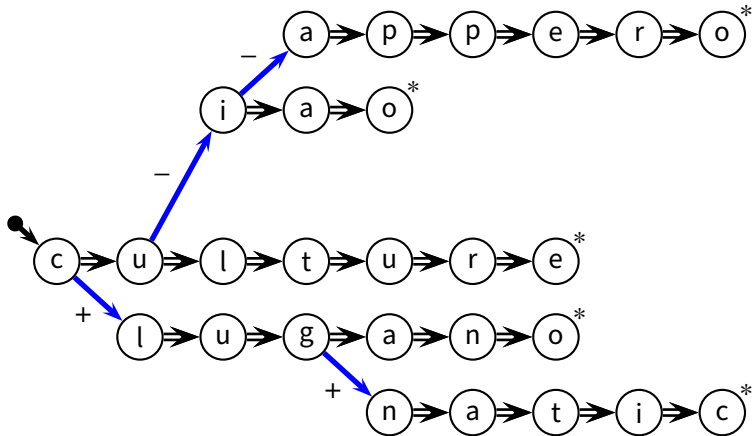
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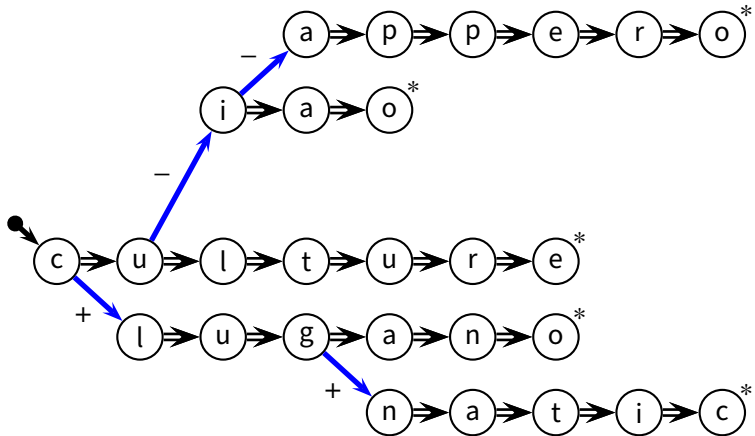
“cappero”



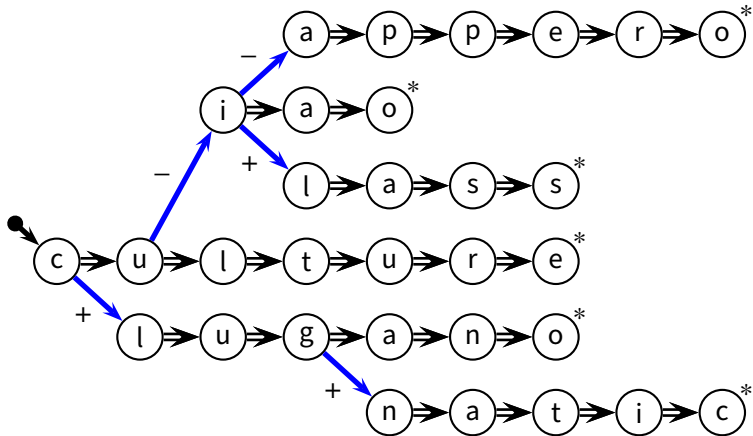
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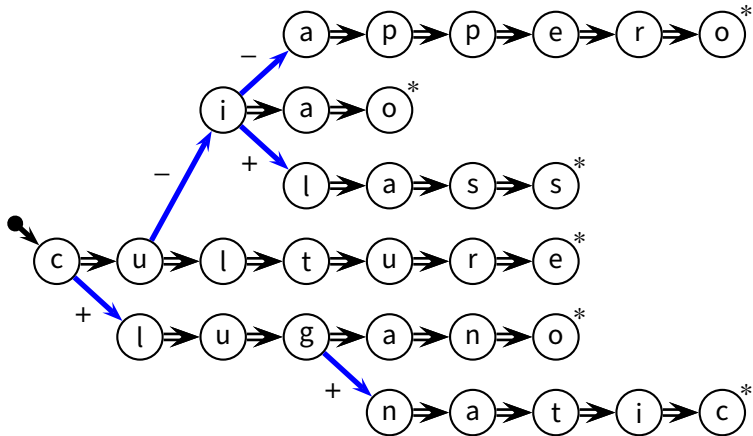
“class”



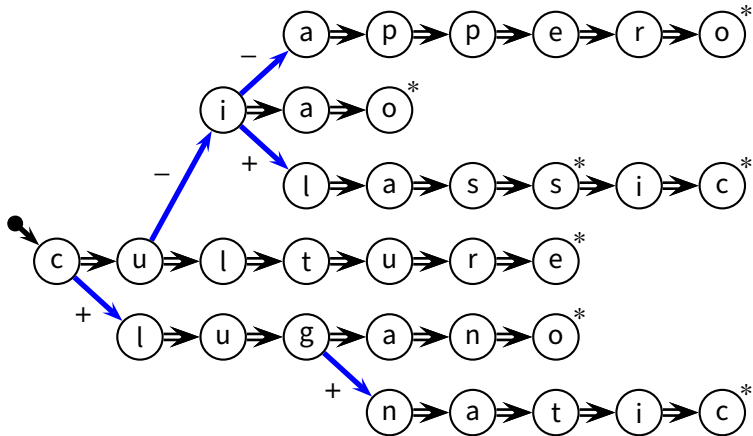
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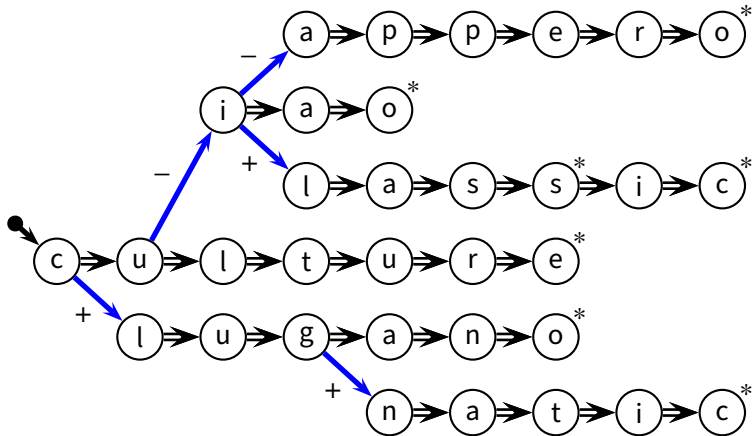
“classic”



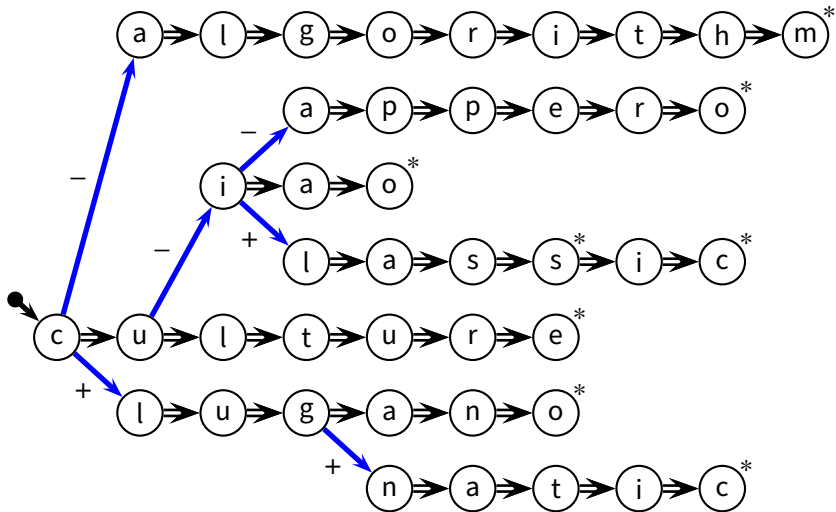
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“algorithm”

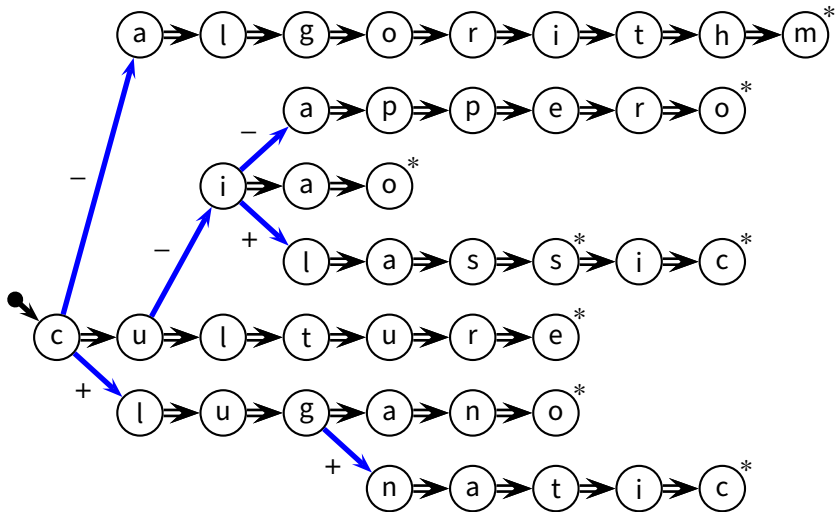


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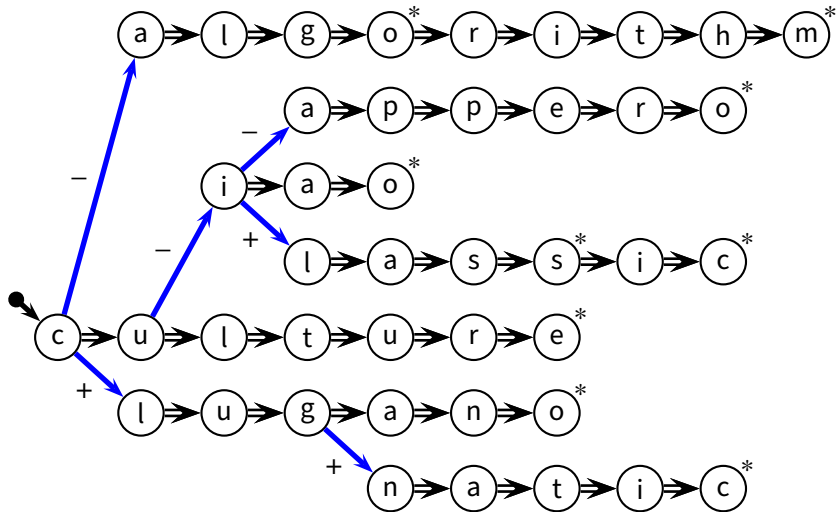
Example

“algo”



Example

“algo”



TSTSEARCH(*T*, *K*)

```
1  for i = 1 to |K|
2      if i > 1
3          T = T.equal
4      while T ≠ NIL and K[i] ≠ T.character
5          if K[i] < T.character
6              T = T.lower
7          else T = T.higher
8      if T == NIL
9          return FALSE
10 return n.value
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- Is it correct?

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- Is it correct? Not completely! (**Exercise:** fix it.)
- Complexity?

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- Is it correct? Not completely! (**Exercise:** fix it.)
- Complexity? Non-trivial...

- Recursion starts with $root = \mathbf{TSTINSERT}(root, K, 1)$

```
TSTINSERT(T, K, i)
```

```
1  if T == NIL
2      T = NEWNODE(K[i])
3  if K[i] < T.character
4      T.lower = TSTINSERT(T.lower, K, i)
5  elseif K[i] > T.character
6      T.higher = TSTINSERT(T.higher, K, i)
7  elseif K[i] == T.character
8      if i < |K|
9          T.equal = TSTINSERT(T.equal, K, i + 1)
10     else T.value = TRUE
11 return T
```