

Tries

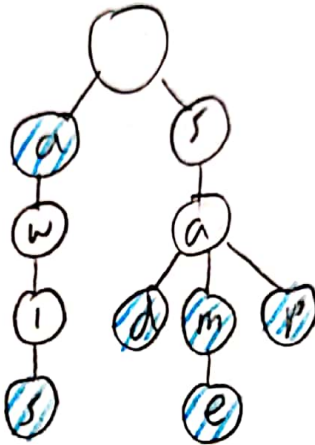
Basics

- good for tracking strings
- Same as tree but every node represents a char
- Root node is a sentinel node
- A word represents sum of chars from sentinel node to a "marked node" (end of word)

Example:

Contains("a") = true
 Contains("aw/s") = true
 Contains("sad") = true
 Contains("sam") = true
 Contains("same") = true
 Contains("sap") = true
 Contains("sa") = false
 Contains("sax") = false

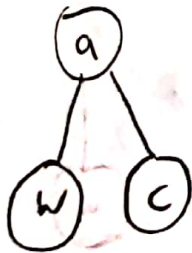
"hit"
 "miss"



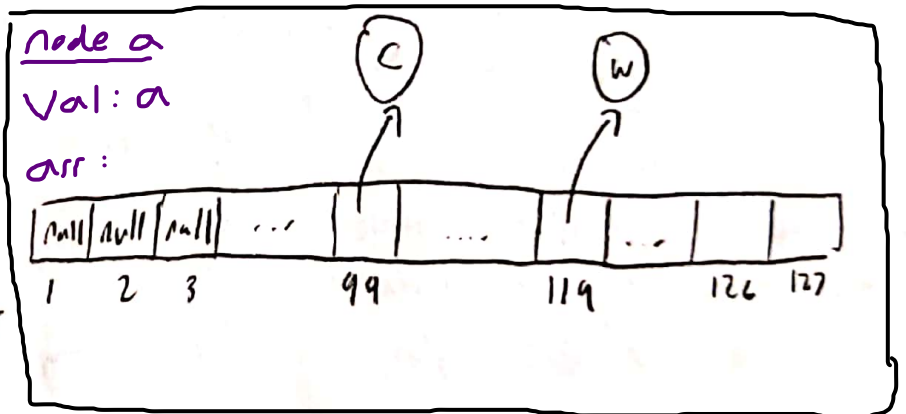
/// = marked
 ↑
 indicates ending of word

Representing node's children: Data Indexed Array Implementation

- create a size 128 array w/ the i th index representing a child node with the i th ASCII char (128 ASCII chars)



$w = 119^{\text{th}}$ ASCII char
 $c = 99^{\text{th}}$ ASCII char



Runtime:

Function
add(String word)

Cost
 $\Theta(\text{word.length}()) \approx \Theta(1)$

contains(String word)

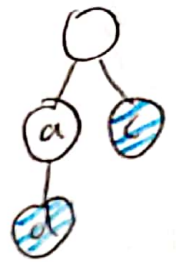
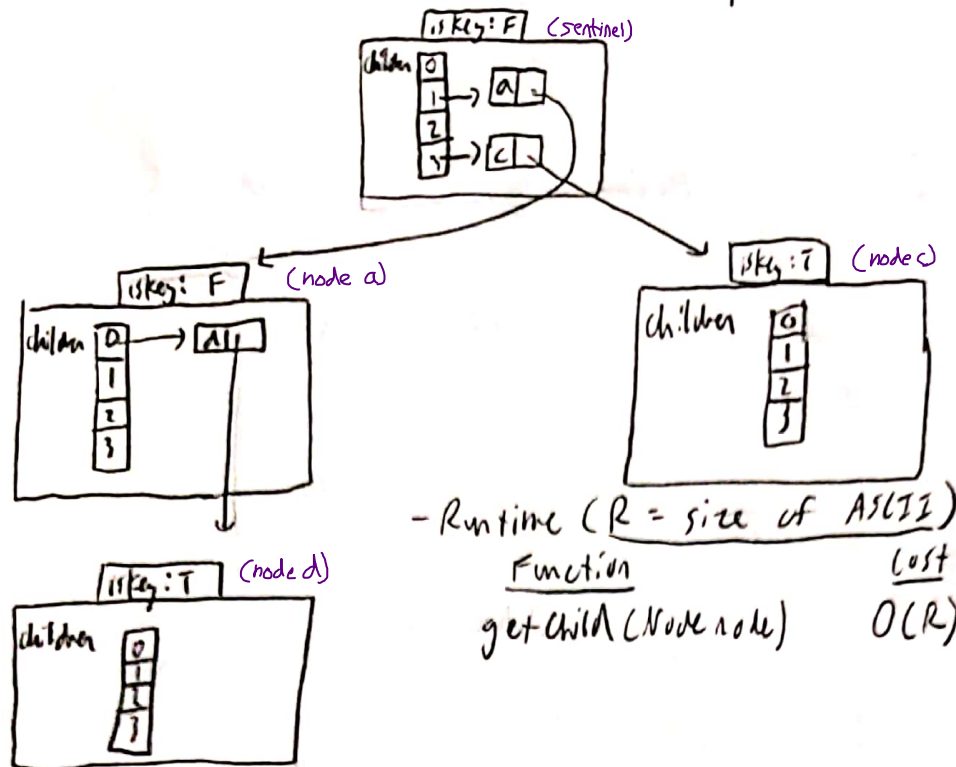
$\Theta(\text{word.length}()) \approx \Theta(1)$

getChild(Node node)

$\Theta(1)$

- Downside: many wasted nulls

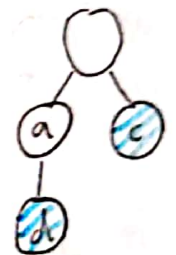
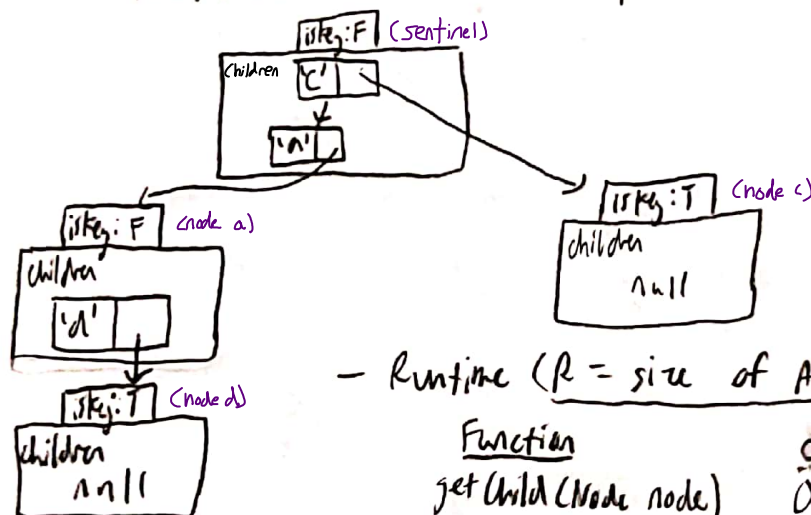
Representing children 2: Hash-Table Based Trie Implementation



- Runtime ($R = \text{size of ASCII}$)

Function: $\text{getChild}(\text{Node node})$ Cost: $O(R) \approx \Theta(1)$, since R fixed

Representing children 3: BST/Treemap Based Trie Implementation



- Runtime ($R = \text{size of ASCII}$)

Function: $\text{getChild}(\text{Node node})$ Cost: $O(\log R) \approx \Theta(1)$, since R fixed

Data Index Array vs. Hash-table vs. BST

- Data Index Array uses more space
- Hash-Table and BST slightly slower for finding children ($O(R)$ and $O(\log R)$)