

CSCI 104

Tries

Mark Redekopp

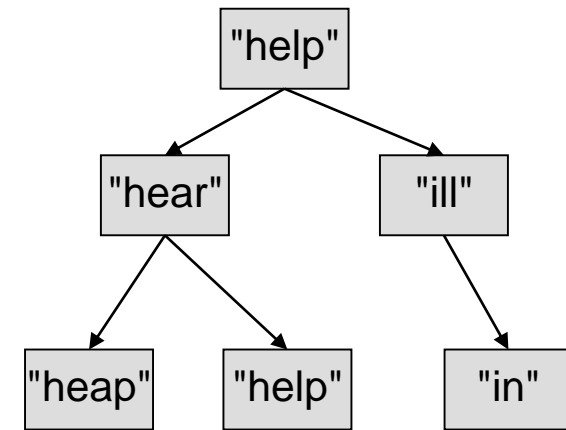
David Kempe

Sandra Batista

TRIES

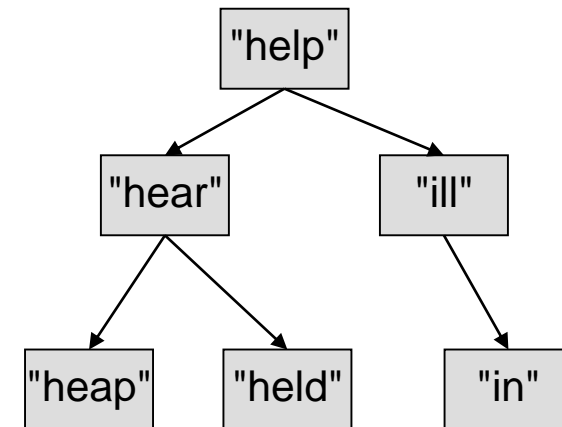
Review of Set/Map Again

- Recall the operations a set or map performs...
 - Insert(key)
 - Remove(key)
 - find(key) : bool/iterator/pointer
 - Get(key) : value **[Map only]**
- We can implement a set or map using a binary search tree
 - Search = $O(\text{_____})$
- But what work do we have to do at each node?
 - Compare (i.e. string compare)
 - How much does that cost?
 - Int = $O(1)$
 - String = $O(k)$ where k is length of the string
 - Thus, search costs $O(\text{_____})$



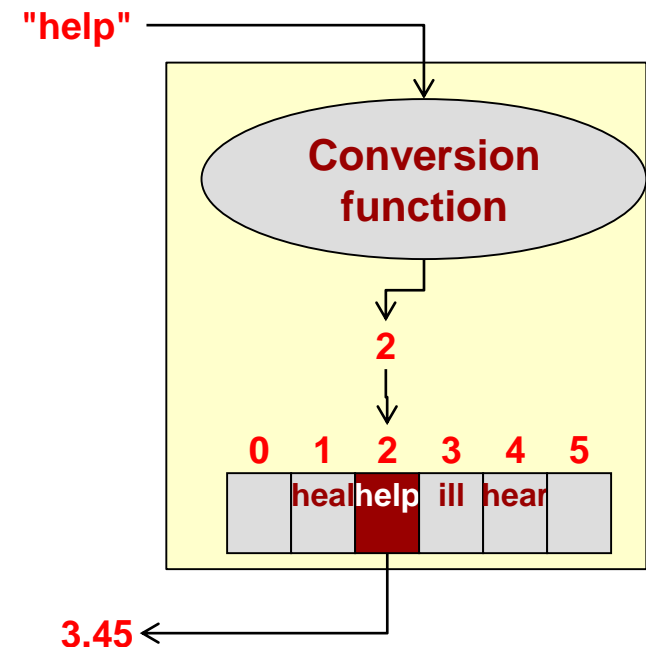
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 - Insert(key)
 - Remove(key)
 - find(key) : bool/iterator/pointer
 - Get(key) : value **[Map only]**
- We can implement a set or map using a binary search tree
 - Search = $O(\log(n))$
- But what work do we have to do at each node?
 - Compare (i.e. string compare)
 - How much does that cost?
 - Int = $O(1)$
 - String = $O(k)$ where k is length of the string
 - Thus, search costs $O(k * \log(n))$



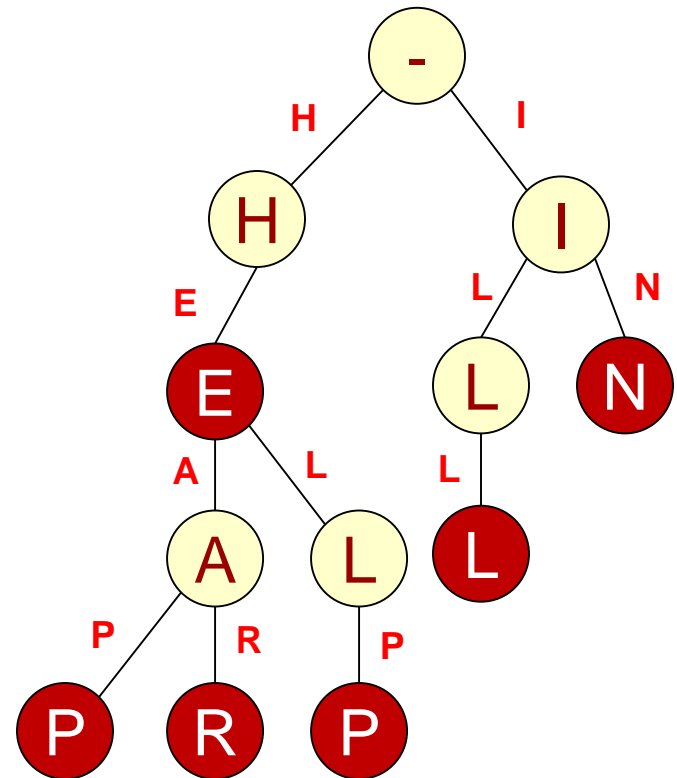
Review of Set/Map Again

- We can implement a set or map using a hash table
 - Search = $O(1)$
- But what work do we have to do once we hash?
 - Compare (i.e. string compare)
 - How much does that cost?
 - Int = $O(1)$
 - String = $O(k)$ where k is length of the string
 - Thus, search costs $O(k)$



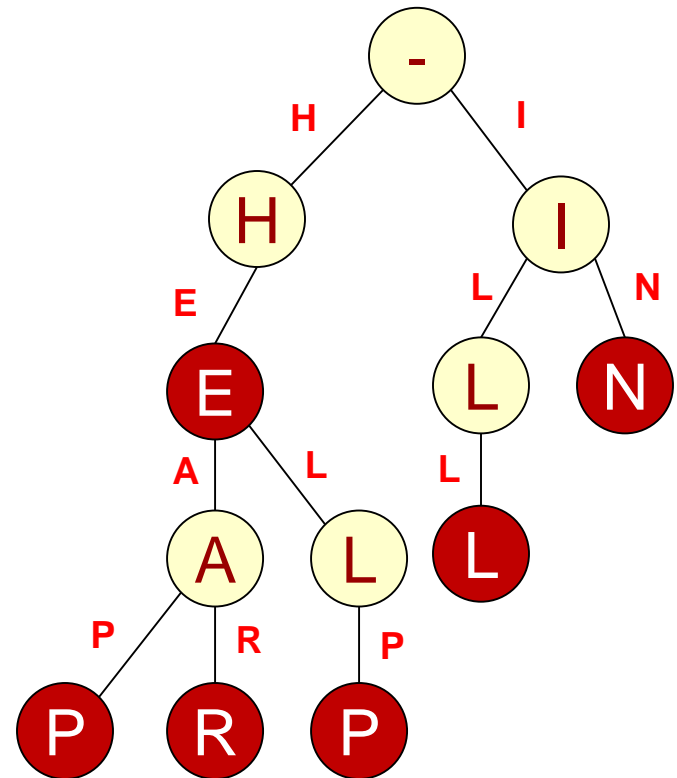
Tries

- Assuming unique keys, can we still achieve $O(k)$ search but not have collisions?
 - $O(k)$ means the time to compare is *independent* of how many keys (i.e. n) are being stored and only depends on the length of the key
- Trie(s) (often pronounced "try" or "tries") allow $O(k)$ retrieval
 - Sometimes referred to as a radix tree or prefix tree
- Consider a trie for the keys
 - "HE", "HEAP", "HEAR", "HELP", "ILL", "IN"



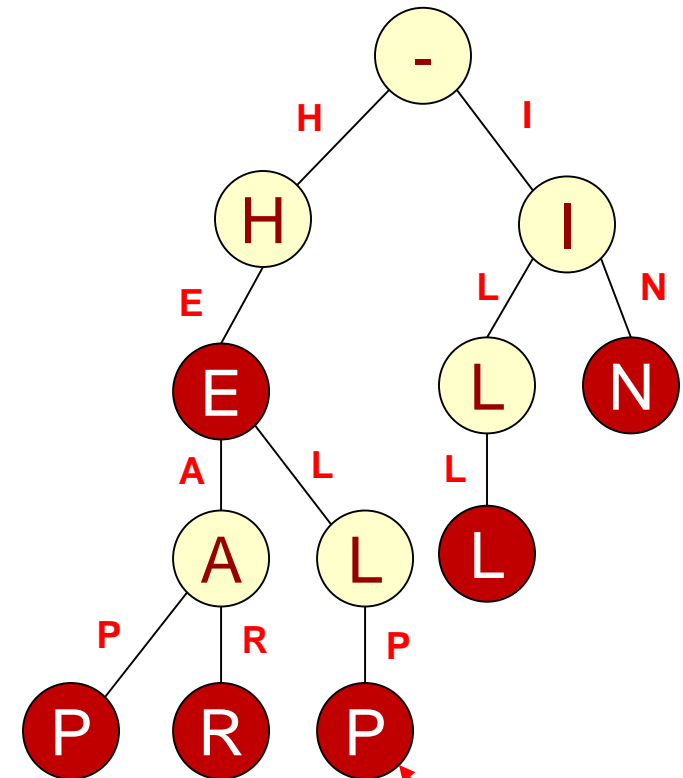
Tries

- Rather than each node storing a full key value, each node represents a prefix of the key
- Highlighted nodes indicate terminal locations
 - For a map we could store the associated value of the key at that terminal location
- Notice we "share" paths for keys that have a common prefix
- To search for a key, start at the root consuming one unit (bit, char, etc.) of the key at a time
 - If you end at a terminal node, SUCCESS
 - If you end at a non-terminal node, FAILURE



Tries

- To search for a key, start at the root consuming one unit (bit, char, etc.) of the key at a time
 - If you end at a terminal node, SUCCESS
 - If you end at a non-terminal node, FAILURE
- Examples:
 - Search for "He"
 - Search for "Help"
 - Search for "Head"
- Search takes $O(k)$ where k = length of key
 - Notice this is the same as a hash table



For a map, a "value" type could be stored for each terminal node

Your Turn

- Construct a trie to store the set of words
 - Ten
 - Tent
 - Then
 - Tense
 - Tens
 - Tenth

Application: IP Lookups

- Network routers form the backbone of the Internet
- Incoming packets contain a destination IP address (128.125.73.60)
- Routers contain a "routing table" mapping some prefix of destination IP address to output port
 - 128.125.x.x => Output port C
 - 128.209.32.x => Output port B
 - 128.x.x.x => Output port D
 - 132.x.x.x => Output port A
- Keys = Match the longest prefix
 - Keys are unique
- Value = Output port

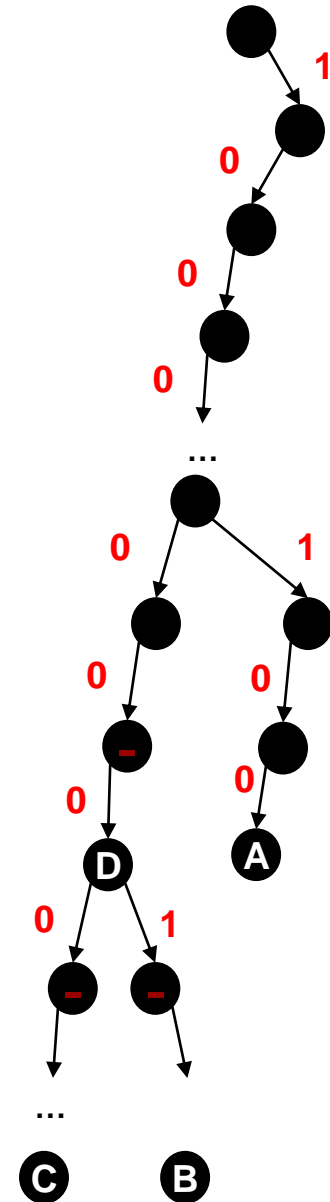


Octet 1	Octet 2	Octet 3	Port
10000000	01111101		C
10000000	11010001	00100000	B
10000000			D
10000100			A

IP Lookup Trie

- A binary trie implies that the
 - Left child is for bit '0'
 - Right child is for bit '1'
- Routing Table:
 - 128.125.x.x => Output port C
 - 128.209.32.x => Output port B
 - 128.209.44.x => Output port D
 - 132.x.x.x => Output port A

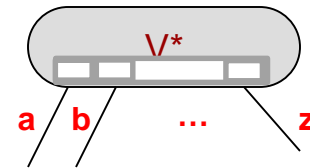
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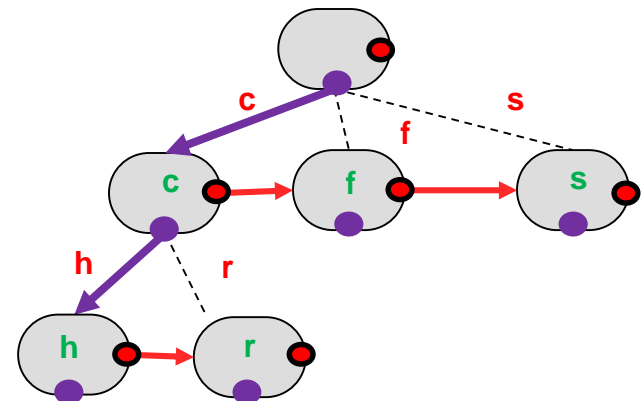
Structure of Trie Nodes

- What do we need to store in each node?
- Depends on how "dense" or "sparse" the tree is?
- Dense (most characters used) or small size of alphabet of possible key characters
 - Array of child pointers
 - One for each possible character in the alphabet
- Sparse
 - (Linked) List of children
 - Node needs to store _____

```
template < class V >
struct TrieNode{
    V* value; // NULL if non-terminal
    TrieNode<V>* children[26];
};
```



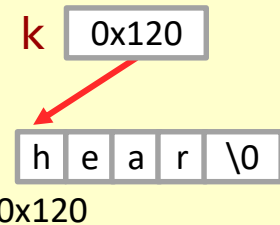
```
template < class V >
struct TrieNode{
    char key;
    V* value;
    TrieNode<V>* next; // sibling
    TrieNode<V>* children; // head ptr
};
```



Search

- Search consumes one character at a time until
 - The end of the search key
 - If value pointer exists, then the key is present in the map
 - Or no child pointer exists in the TrieNode
- Insert
 - Search until key is consumed but trie path already exists
 - Set v pointer to value
 - Search until trie path is NULL, extend path adding new TrieNodes and then add value at terminal

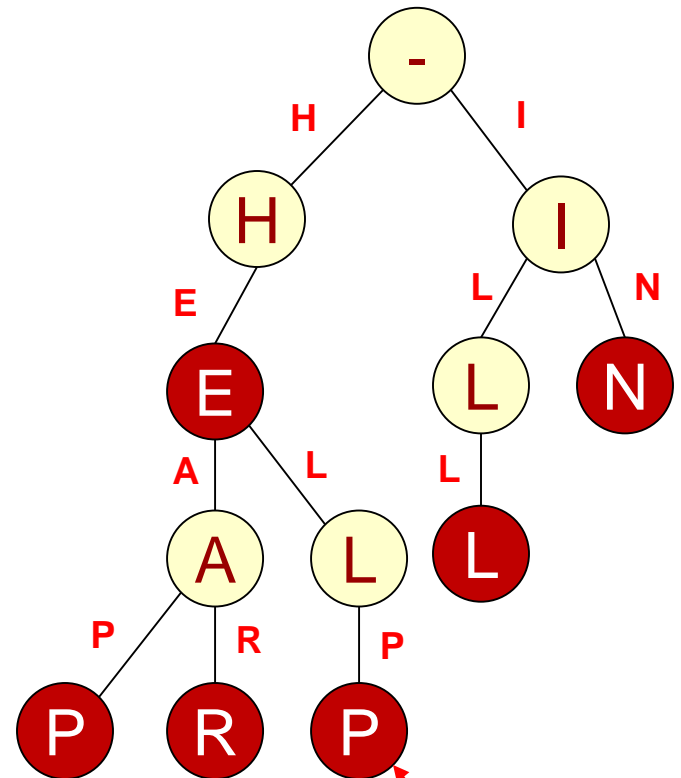
```
V* search(char* k, TrieNode<V>* node)
{
  while(*k != '\0' && node != NULL){
    node = node->children[*k - 'a'];
    k++;
  }
  if(node) return node->v;
  else return NULL;
}
```



```
void insert(char* k, Value& v)
{
  TrieNode<V>* node = root;
  while(*k != '\0' && node != NULL){
    node = node->children[*k - 'a']; k++;
  }
  if(node){
    node->v = new Value(v);
  }
  else {
    // create new nodes in trie
    // to extend path
    // updating root if trie is empty
  }
}
```

Thinking Exercise: Removal

- How would removal of a key work in a trie and what are the cases you'd have to worry about?
 - Does removal of a key always mean removal of a node?
 - If we do remove a node, would it only be one node in the trie?



A "value" type could be stored for each non-terminal node

SUFFIX TREES (TRIES)

Prefix Trees (Tries) Review

- What problem does a prefix tree solve
 - Lookups of keys (and possible associated values)
- A prefix tree helps us match 1-of-n keys
 - "He"
 - "Help"
 - "Hear"
 - "Heap"
 - "In"
 - "Ill"
- Here is a slightly different problem:
 - Given a large text string, T, can we find certain substrings or answer other queries about patterns in T
 - A suffix tree (trie) can help here

Suffix Trie Slides

- <http://www.cs.cmu.edu/~ckingsf/bioinfo-lectures/suffixtrees.pdf>

Suffix Trie Wrap-Up

- How many nodes can a suffix trie have for text, T , with length $|T|$?
 - $|T|^2$
 - Can we do better?
- Can compress paths without branches into a single node
- Do we need a suffix trie to find substrings or answer certain queries?
 - We could just take a string and search it for a certain query, q
 - But it would be slow $\Rightarrow O(|T|)$ and not $O(|q|)$

What Have We Learned

- [Key Point]: Think about all the data structures we've been learning?
 - There is almost always a trade-off of memory vs. speed
 - i.e. Space vs. time
 - Most data structures just exploit different points on that time-space tradeoff continuum
 - Think about searches in your project...Do we need a map?
 - No, we could just search all items each time a keyword is provided
 - But think how slow that would be
 - So we build a data structure (i.e. a map) that replicates data and takes a lot of memory space...
 - ...so that we can find data faster