

CSCI 104

Tries

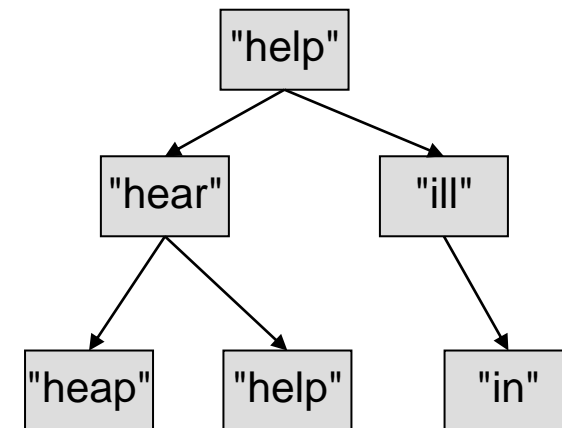
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TRIES

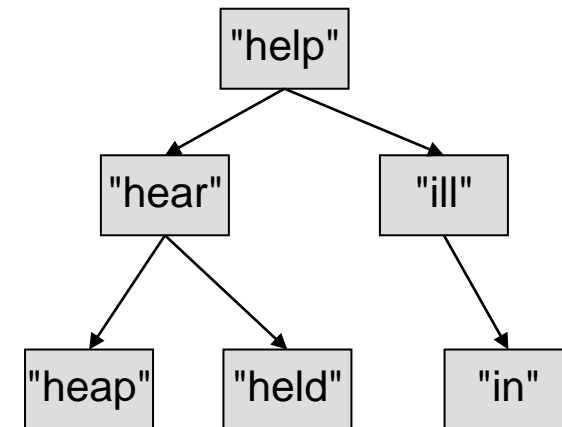
Review of Set/Map Again

- 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 $O(\text{_____})$



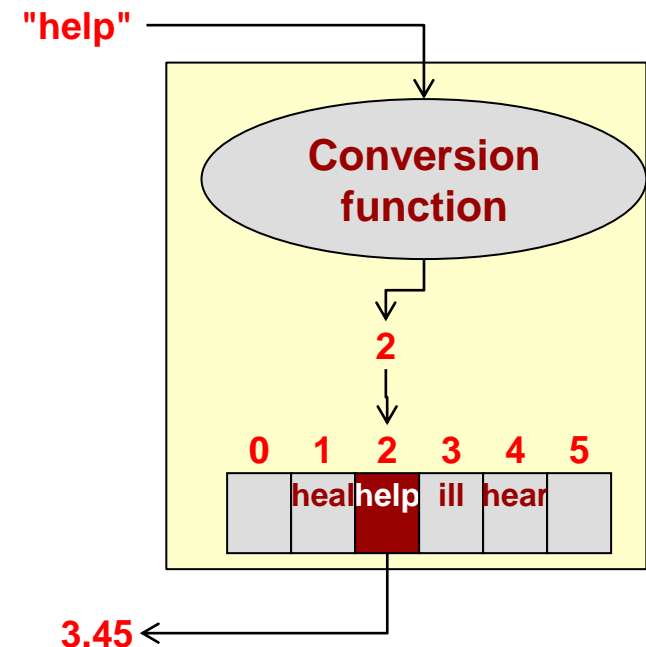
Review of Set/Map Again

- 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 = $O(\log(n))$
- 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 $O(k * \log(n))$



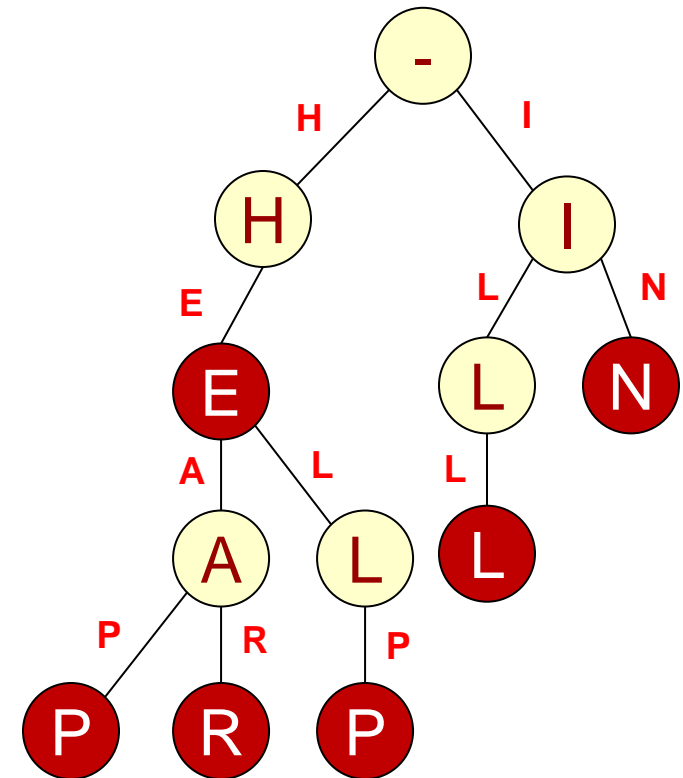
Review of Set/Map Again

- 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 $O(k)$



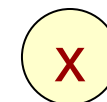
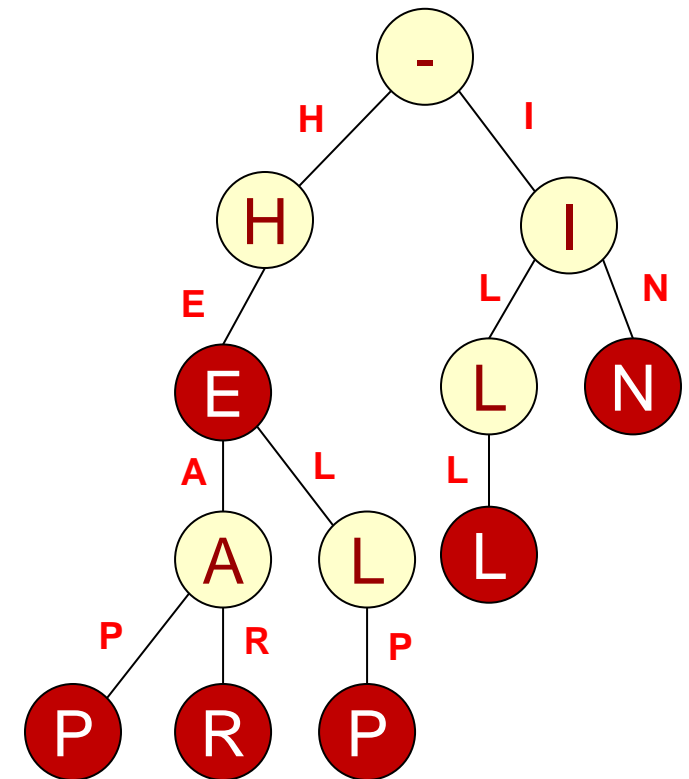
Tries

- 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 $O(k)$ (i.e. constant time) retrieval
 - Sometimes referred to as a radix tree or **prefix tree**
- Consider a trie (prefix tree) 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
- A key is represented by a path through the tree
 - 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



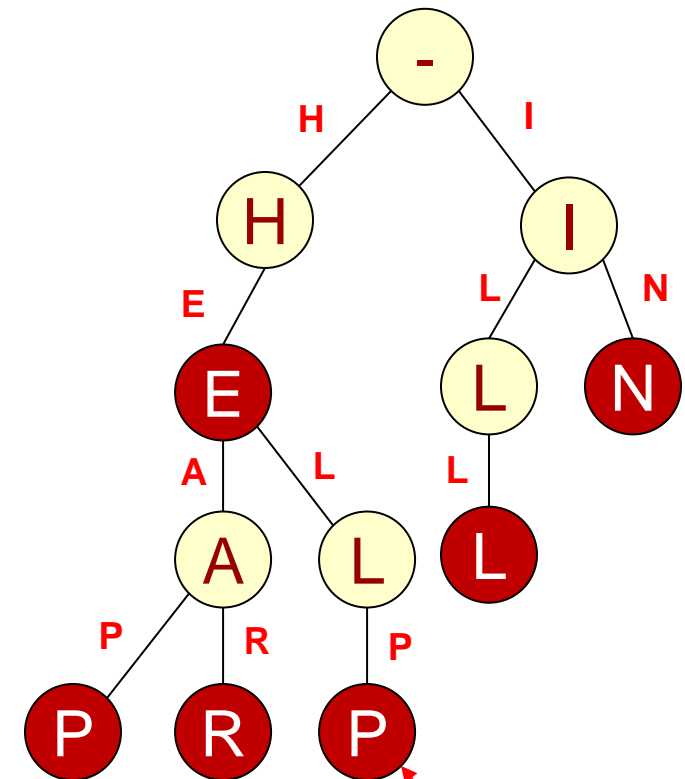
Internal node
(prefix is not a key)



Terminal node
(prefix is an inserted key)

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



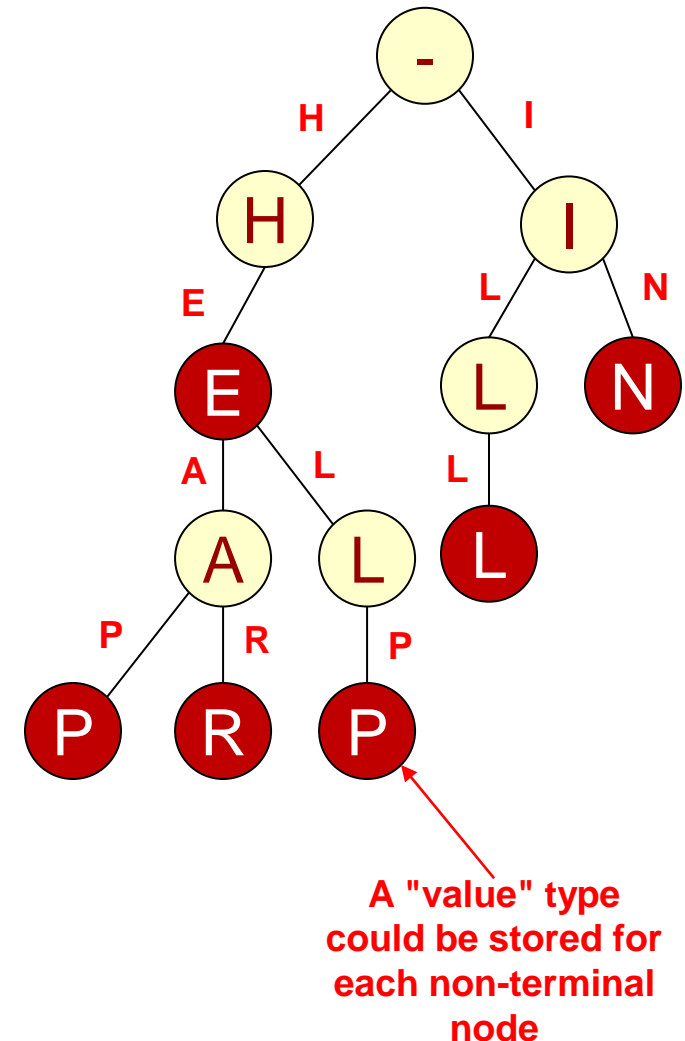
For a map, a
"value" type could
be stored for each
terminal node

Your Turn

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

Thinking Exercise: Removal

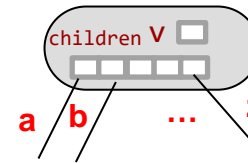
- How would removal of a key work in a trie and what are the cases you'd have to worry about?
 - Does removal of a key always mean removal of a node?
 - If we do remove a node, would it only be one node in the trie?



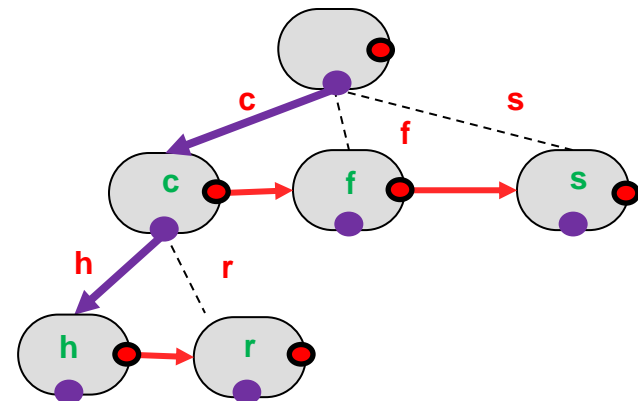
Structure of Trie Nodes

- 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

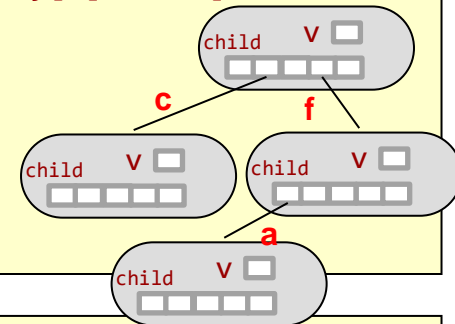
f a i r key

- 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(string key)
{
    TrieNode<V>* node = root;
    size_t i=0;
    while(i != key.size() && node != NULL){
        node = node->children[key[i] - 'a'];
        i++;
    }
    if(node) return node->v;
    else return NULL;
}

```



```

void insert(string key, const V& v) {
    TrieNode<V>* node = root;
    size_t i=0;
    while(i != key.size() && node != NULL){
        node = node->children[key[i] - 'a']; i++;
    }
    if(node){
        node->v = new V(v);
    }
    else {
        // create new nodes in trie to extend path
        // updating root if trie is empty
    }
}

```

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

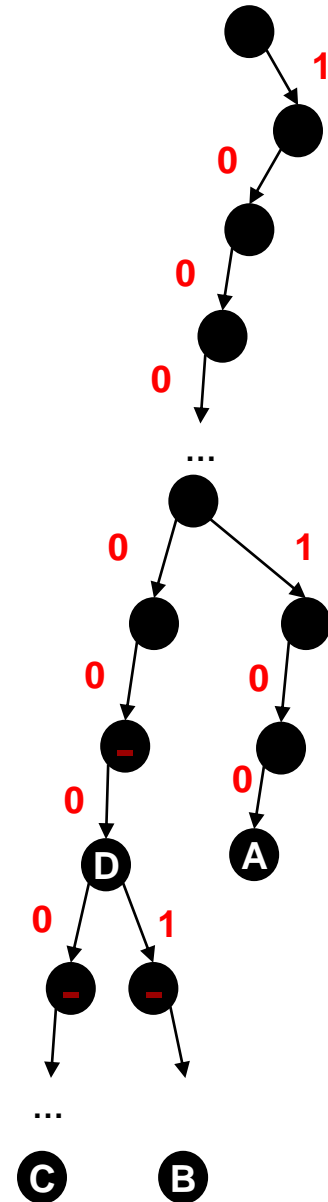


| Octet 1 | Octet 2 | Octet 3 | Port |
|----------|----------|----------|------|
| 10000000 | 01111101 | | C |
| 10000000 | 11010001 | 00100000 | B |
| 10000000 | | | D |
| 10000100 | | | A |

IP Lookup Trie

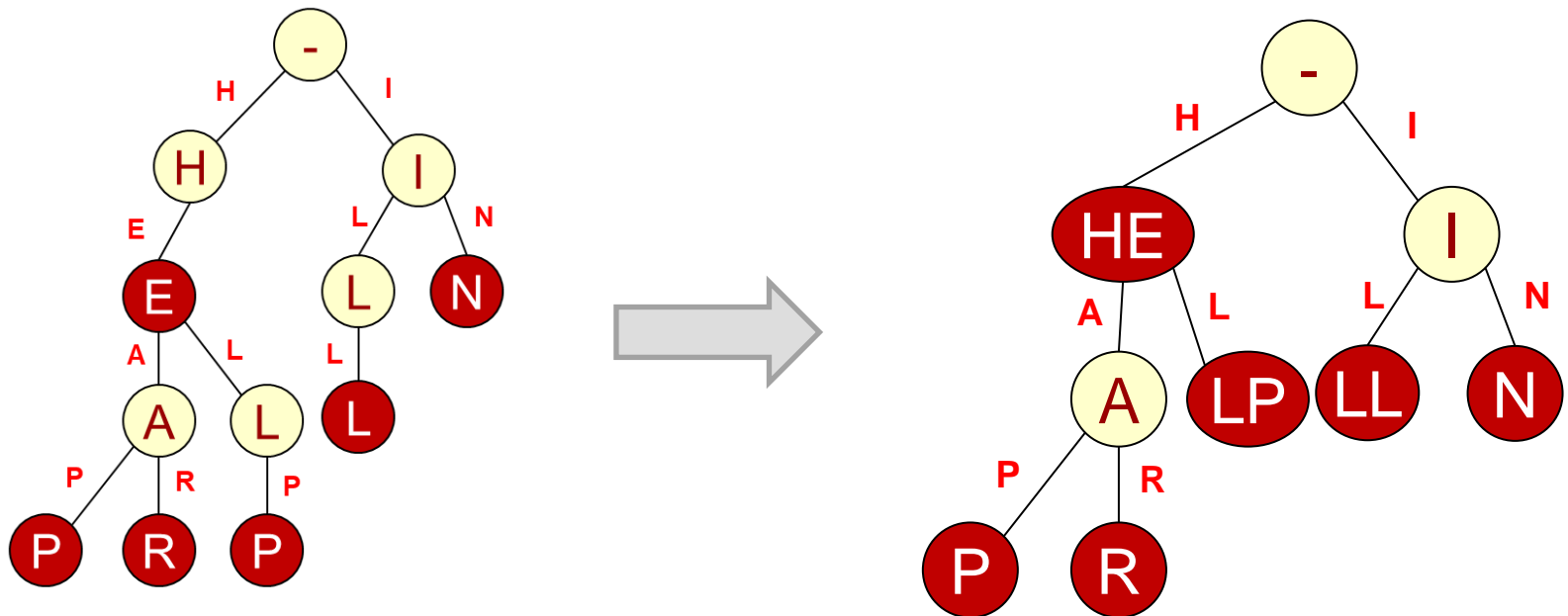
- A binary trie implies that the
 - Left child is for bit '0'
 - Right child is for bit '1'
- Routing Table:
 - 128.125.x.x => Output port C
 - 128.209.32.x => Output port B
 - 128.209.44.x => Output port D
 - 132.x.x.x => Output port A

| Octet 1 | Octet 2 | Octet 3 | Port |
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| 10000000 | 01111101 | | C |
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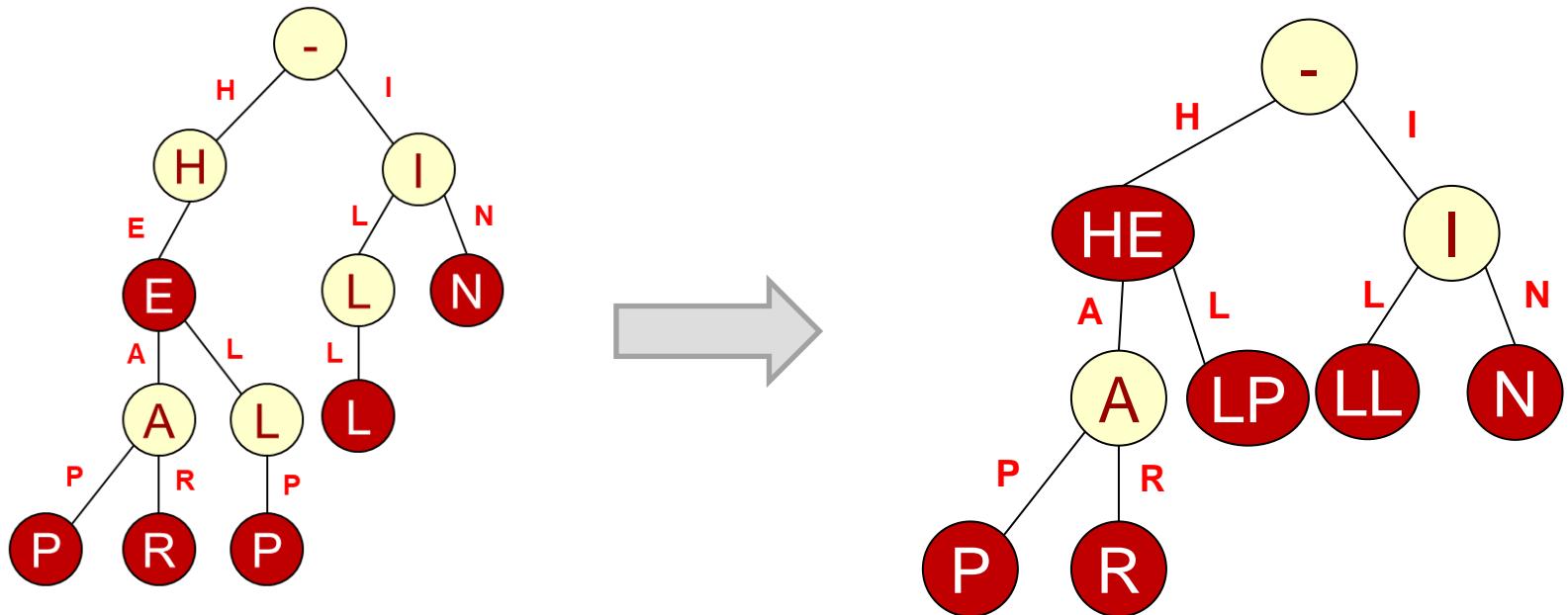
Compressed Prefix Tree

- We can reduce the number of nodes and thus storage, by storing substrings in each node
 - If a node has only one child, combine with child storing concatenation of characters
 - <https://www.cs.usfca.edu/~galles/visualization/RadixTree.html>



Compressed Prefix Tree

- Walk key string based on the length of the substring in the current node and then use the next key string character to choose the child node
- Key is not present if key string characters are exhausted before substring in node or no corresponding child entry
- Examples: 'H', 'HERD'



Practice

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

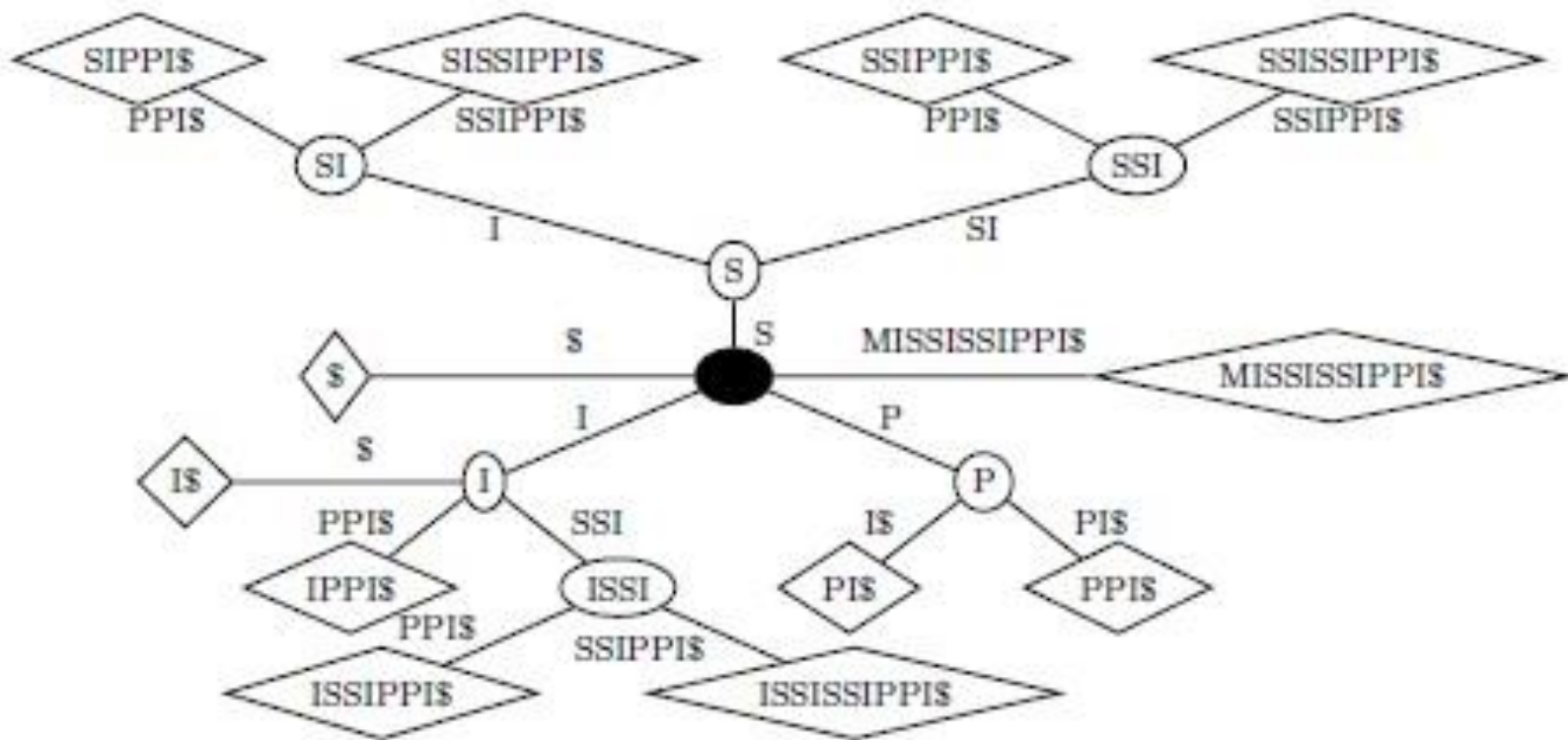
Prefix Trees (Tries) Review

- 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 TREES

Suffix Trees

A **suffix tree** of a string W is a compressed trie consisting of all possible suffixes of W .



Are 'issip' or 'sipi' substrings?

Suffix Trees

- When W has n characters (indexed 0 to n), the suffix tree has:
 - n leaves, each one representing a single suffix
 $W[i: (n - 1)], 0 \leq i \leq (n - 1)$
 - Every non-leaf node has at least two children
 - Each edge is labelled with a substring of W
 - If e and e' are edges out of the same node, then their labels start with different letters.
 - For any root-leaf path, the concatenation of their edge labels is equal to $W[i: (n - 1)]$
 - $< n$ internal nodes
 - $O(n)$ total nodes
- There is an algorithm (Ukkonen's Algorithm) which can build a suffix tree in linear time.

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 (Space vs. time)
 - Most data structures just exploit different points on that time-space tradeoff continuum
 - Often we build a data structure that replicates data and takes a lot of memory space...
 - ...so that we can find data faster

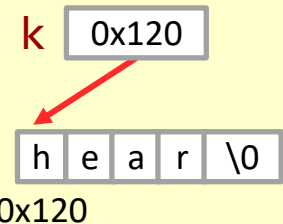
Suffix Trie Slides

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

Search (Using C-Strings / Char arrays)

- 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
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    while(*k != '\0' && node != NULL){
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    }
    if(node){
        node->v = new Value(v);
    }
    else {
        // create new nodes in trie
        // to extend path
        // updating root if trie is empty
    }
}
```