

Trie Time Complexity

- Let N be the number of keys in the trie
- Let L_{key} be the length of a key

Let's analyze contains...

```
contains(T,key):  
O(1)  if T is null:  
      return false (for Set) or null (for Map)  
  
O(1)  if key is empty string:  
      return T.terminates (for Set) or T.value (for Map)  
  
      return contains(T.children.get(key[0]),rest(key))  
                    O(Lkey) or O(1)    O(Lkey) or O(1)
```

Independent of $N!$

Constant time!

With proper optimization, each instance is $O(1)$. How many times does it run in the worst case? $O(L_{\text{key}})$.

Trie Time Complexity, Summary

With a max key length, worst case time complexities:

	Contains	Insert	Delete
BST	$O(N)$	$O(N)$	$O(N)$
AVL	$O(\log N)$	$O(\log N)$	$O(\log N)$
Trie	$O(1)$	$O(1)$	$O(1)$

Time Complexity, a Fraud Revealed?

So why ever use an AVL tree over a trie?

- If B is the size of the alphabet, and L is the max length of a key, there are B^L unique keys.
 - A maximum size before the tree is full!
- Asymptotic runtime complexity studies behavior as $N \rightarrow \infty$.
- For an arbitrary N , we need $L=O(\log_B N)$
 - Now $O(\log N)$, not $O(1)$ recursive calls
- Aside: a similar result is true for Radix Sort:
 - Runtime is $O(NL)$
 - Don't want each key to appear on average ∞ times?
then it's really $O(N \log N)$

Now that you know the truth...

We'll pretend we don't.

Trie operations? $O(1)$

Radix sort? $O(N)$

Tries and radix sort can be good options when the keys are convenient and naturally bounded.

They are not magic bullets.

Asymptotic analysis doesn't tell the full story.

