

Hash Tables

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CSC220 Programming II – Spring 2019



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- ▶ Balanced binary search tree (DLLTree): add and remove in $O(\log n)$ amortised time and find in $O(\log n)$ worst case time.



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- ▶ When you look up a name, it is nice to be able to go forward or back a few names in case you misspelled it.
- ▶ If you want Milenkovic in a hash table, you better not look for Milenkovich because it will be far away.



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- ▶ We will look at the one for String:

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public int hashCode()
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Returns a hash code for this string. The hash code for a String object is computed as

$$s[0]*31^{(n-1)} + s[1]*31^{(n-2)} + \dots + s[n-1]$$

using int arithmetic, where $s[i]$ is the i th character of the string, n is the length of the string, and $^$ indicates exponentiation. (The hash value of the empty string is zero.)



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WHAT???



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String name = "Milenkovic";  
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An int can only hold integers in the range from -2147483648 to 2147483647.



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```
int hashIndex (String name, int m) {  
    int code = name.hashCode();  
    int index = code % m;  
    if (index < 0)  
        index += m;  
    return index;  
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- ▶ But it is also -3 remainder 2 . Check $-3*7+2=-19$.



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- ▶ Entries in the same list in the first table will be in different lists in the second table.



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- ▶ No more than $m/2$ show up on any particular day,
- ▶ but collisions are still certain.

So if you arrive and your space is filled,

- ▶ you are supposed to keep driving until you find an empty one and park there.
- ▶ If you reach the end of the lot, you go back to the beginning.

Here is one problem with this scheme.

- ▶ If I need to find your car, I go to your assigned space.
- ▶ If I don't find you there, I keep driving until I get to your car
- ▶ or an empty space.
- ▶ The latter tells me you are not there.



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- ▶ You have to keep $n \leq m/2$.



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- ▶ Item cycles through the array until it finds an empty spot.
- ▶ Removed items need to leave a “traffic cone”.
- ▶ Needs to rehash and possibly reallocate when less than $m/2$ spaces are empty.

