

Using a heap as a priority queue

- To remove highest priority item from heap:
`remove root; /* O(1) complexity */`
`Move last item to root, then ...`
`heapify in reverse; /* O(log n) complexity */`
– So overall complexity is $O(\log n)$
- Also $O(\log n)$ for insert function
- Compare to other priority queue strategies
– Sorted list: insert – $O(n)$; remove – $O(1)$
– Unsorted array: insert – $O(1)$; remove – $O(n)$
- Choose heap strategy if n is expected to be large

A table ADT

- Declare Table type (define in implementation)
`typedef struct TableTag Table;`
– Also define a KeyType, and maybe a DataType (or just use void *)
- Let user define initial size of table
`Table *createTable(int startingSize);`
- Can put/get/update/remove info associated with unique key
`int put(KeyType key, void *info, Table *table);`
`void* get(KeyType key, Table *table);`
`int update(KeyType key, void *info, Table *table);`
`int remove(KeyType key, Table *table);`
– Functions return false if unsuccessful (except get returns NULL)
- Can print all info, usually in key order
`void printAll(Table *table);`

Table implementation options

- Many possibilities – depends on application
– And how much trouble efficiency is worth
- Option 1: use a BST
– To put: insertTree using key for ordering
– To update: deleteTree, then insertTree
– To getAll: use in-order traversal
- Option 2: sorted array with binary searching (later)
- Option 3: implement as a “hash table”
– Hashing – general technique works great with tables

Hashing ideas and concepts

- Idea: transform arbitrary key domain (e.g., strings) into “dense integer range” – then use result as index to table
– `index = hash(key); /* function returns int */`
- Collisions: `hash(k1) == hash(k2), k1 != k2`
– Usually impossible to avoid (“perfect hashing”), so must have a way to handle collisions
– e.g., probe for empty slot if using “open addressing” -
`while (!empty(index)) index = probe(key);`
- Concept: insertion/searching is quick – but only until the table starts to get filled up
– Then collisions start happening too often!

Open address hashing

– & implementing basic table ADT

- Define structs for table items and whole table of items
`typedef struct`
`{ KeyType key; void *info; } TableItem;`
`typedef struct`
`{ int size; int n; TableItem *items; } Table;`
– size is size of array; n is the number of items in the table
– Constructor allocates memory for array of items, and initializes all items to “empty” key
- The put function uses `hash(key)` and `probe(key)` to find empty slot for new item
– Resizes array (and rehashes existing items) whenever table “load factor” reaches 50 percent (rule of thumb for open addressing)

Open address hashing (cont.)

- get & update functions use `hash(key)` and `probe(key)` in *exact same sequence* as put – to find existing info
- remove is more complicated
– Cannot just remove an item – future probes for get and update might terminate prematurely at empty slot
– Inefficient technique rehashes all items
– Alternative technique uses “deleted” key markers
• But problem with that is table fills up prematurely
- printAll in key order – must sort first!
– So $O(n \log n)$ at best!

Resolving collisions

- Simplest open address approach is linear probing
 - If $(\text{index} = \text{hash}(\text{key}))$ is not empty, try $\text{index}+1$, then $\text{index}+2$, ..., until empty slot
 - In other words, searching for first “open address”
 - Biggest problem: it leads to “primary clusters”
- Quadratic probing – varies probe, like 1, 3, 6, ...
 - Leads to “secondary clusters” but not as quickly
- Double hashing – probe (key) varies by key
 - Best open addressing approach for avoiding clusters
- Or a completely different approach: “chaining”

Chaining

- Table is an array of pointers to lists:


```
typedef struct TableTag
    { int size; int n;
      ListPointer *lists; };
```
- Constructor allocates memory for array, and creates an empty list for each element of the array
- put function uses $\text{hash}(\text{key})$ and appends to end of list
 - Clustering not a problem, but long lists can be, so rule of thumb is resize when load factor approaches 80%
- remove function is easier now – just delete from list
- But lots more overhead than open addressing
 - Must store node pointers as well as key and info
 - Use list function calls instead of direct array access

Recursive binary searching

- Start with a sorted array: $a[0..n-1]$
 - Useful item in a is $\text{struct}\{\text{key}, \text{info}\}$ ItemType;
- Binary searching algorithm is naturally recursive:

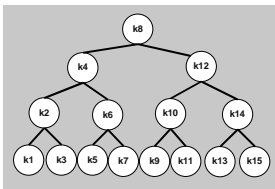

```
int bsearch(KeyType key, ItemType a[],
            int left, int right) {
    /* first call is for left=0, and right=n-1 */
    int middle = (left + right) / 2;
    if (key == a[middle].key) return middle; /* success */
    if (left > right) return -1; /* unsuccessful */
    if (key > a[middle].key) /* search one half or the other */
        return bsearch(key, a, middle+1, right);
    else return bsearch(key, a, left, middle-1);
}
```
- Iterative version is a little trickier (but not too hard)

Iterative binary searching

- ```
int bsearch(KeyType key, ItemType a[], int n) {
 int low = 0, high = n-1, middle;
 while (low <= high) {
 middle = (low + high) / 2;
 if (key == a[middle].key)
 return middle; /* success */
 if (key > a[middle].key) low = middle + 1;
 else high = middle - 1;
 }
 return -1; /* unsuccessful */
}
```
- Both versions are same complexity class
    - But recursive version has more overhead, so actually runs a bit slower than iterative version
    - Interpolation search, by the way, is in a faster class

## Complexity of binary search

- Say array has 15 elements,  $k_1..k_{15}$ ;  $a[0..14]$ 
  - If key is at  $k_8$  ( $a[7]$ ) then found by 1 comparison
  - If key is at  $k_4$  or  $k_{12}$ , takes 3 comparisons ...
- i.e., it's just like searching a BST



- Problem size is halved at each step
  - So complexity class is  $O(\log n)$
- Interpolation search reduces more quickly
  - Class is  $O(\log \log n)$

## Compare 3 table implementations

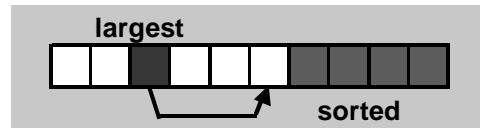
| Table operation   | Hash table    | BST         | Sorted array |
|-------------------|---------------|-------------|--------------|
| create            | $O(n)$        | $O(1)$      | $O(n)$       |
| find, get, update | $O(1)$        | $O(\log n)$ | $O(\log n)$  |
| put               | $O(1)$        | $O(\log n)$ | $O(n)$       |
| remove            | $O(1)$        | $O(\log n)$ | $O(n)$       |
| printAll          | $O(n \log n)$ | $O(n)$      | $O(n)$       |

- Conclusion – depends on table purpose &  $n$  size
  - Hash table wins for most applications if  $n$  is large
  - BST wins if expect to printAll frequently
  - Sorted array might win for small  $n$  – to minimize overhead/work

## Sorting

- Probably *the* most expensive common operation
- Problem: arrange  $a[0..n-1]$  by some ordering
  - e.g., in ascending order:  $a[i-1] \leq a[i]$ ,  $0 < i < n$
- Two general types of strategies
  - Comparison-based sorting – includes most strategies
    - Apply to any comparable data – (key, info) pairs
    - Lots of simple, inefficient algorithms
    - Some not-so-simple, but more efficient algorithms
  - Address calculation sorting – rarely used in practice
    - Must be tailored to fit the data – not all data are suitable
    - Won't cover in CS 12 – see proxmap and radix sorts in sec. 13.6

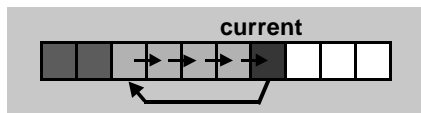
## Selection sort



- Idea: build sorted sequence at end of array
- At each step:
  - Find *largest* value in not-yet-sorted portion
  - Exchange this value with the one at end of unsorted portion (now beginning of sorted portion)
- Complexity is  $O(n^2)$  – but simple to program
  - Also – best way to find  $k^{\text{th}}$  largest, or top  $k$  values

## Insertion sort

- Generally “better” than other simple algorithms
- Inserts one element into sorted part of array
  - Must move other elements to make room for it



- Complexity is  $O(n^2)$  (code)
  - But runs faster than selection sort and others in class
  - Really quick on *nearly sorted* array
- Often used to supplement more sophisticated sorts

## Divide & conquer strategies

- Idea: (1) divide array in two; (2) sort each part; (3) combine two parts to overall solution
- e.g., mergeSort
 

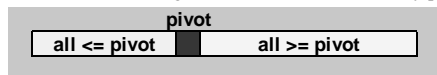
```
if (more than one item in array):
 divide array into left half and right half;
 mergeSort(left half); mergeSort(right half);
 merge(left half and right half together);
```

  - Requires helper method to merge two halves
  - Complexity is  $O(n \log n)$
  - The best sort for large files (too big for memory)
- But for most problems, quickSort is a better divide & conquer strategy

## Quick sort

- Basic quicksort algorithm is recursive
 

```
if (there is something to sort)
{
 partition array elements;
 sort left part; sort right part;
} /* It's the utility partition function that does all the work! */
```
- Partition idea: arrange elements around an arbitrary pivot



```
scan from (i = left) until a[i] >= pivot;
scan from (j = right) until a[j] <= pivot;
swap a[i], a[j];
continue both scans until i > j; (code)
```

## Quick sort (cont.)

- Complexity is  $O(n \log n)$  on average
  - Fastest comparison-based sorting algorithm
  - But overkill, and not-so-fast with small arrays
    - One frequently-used optimization applies insertion sort for partitions smaller than than 10 or so
- But worst case is  $O(n^2)$ !
  - Just like BST worst case – sorted order can be bad
    - Especially if first or last is chosen as pivot – middle is better
- By the way – see `qsort` in `<stdlib.h>` (code)
  - Also by the way – see  $O(n)$  address calculation sorts if really fast sorting is required for an application