Priority queues and binary heaps

Data Structures and Algorithms for Computational Linguistics III (ISCL-BA-07)

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Priority queue ADT

- A priority queue is a collection, an abstract data type, that stores items
- The items in a priority queue are *key–value* pairs
- The key determines the priority of the item, while the value is the actual data of interest
- The interface of a priority queue is similar to a standard queue
- Instead of the first item entered into the queue, the item with the highest priority (minimum or maximum key value) is removed from the priority queue
- Priority queues have many applications ranging from data compression to discrete optimization
- We will see their application to sorting (this lecture) and searching on graphs (later)

Priority queues

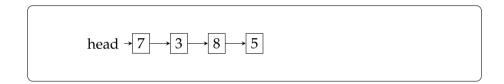
Key operations

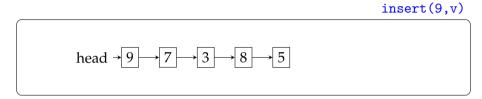
- insert(k, v) Similar to enqueue(v), inserts the value v with priority k into the queue
 - remove() Similar to dequeue(), removes and returns the item with highest priority
 - This operation is often called remove_min() or remove_max()
 depending on minimum or maximum key value is considered
 having the highest priority

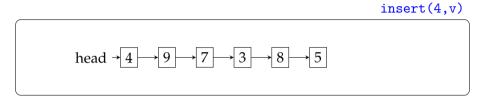
Priority queues

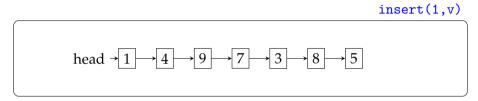
Example operations

Operation	Return value	Priority queue
insert(5, a)		{(5,a)}
insert(9, c)		$\{(5,a), (9,c)\}$
insert(3, b)		$\{(5,a), (9,c), (3,b)\}$
insert(7, d)		$\{(5,a), (9,c), (3,b), (7,d)\}$
remove()	С	$\{(5,a), (3,b), (7,d)\}$
remove()	d	$\{(5,a), (3,b)\}$
remove()	a	$\{(3,b)\}$
remove()	b	{}



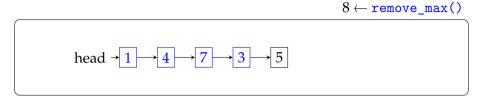




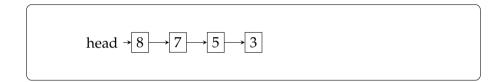


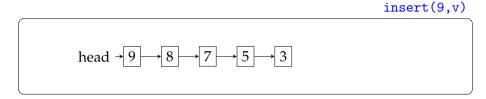
unsorted list

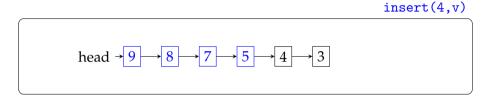
 $9 \leftarrow \text{remove_max()}$ $\text{head} \rightarrow 1 \rightarrow 4 \rightarrow 7 \rightarrow 3 \rightarrow 8 \rightarrow 5$

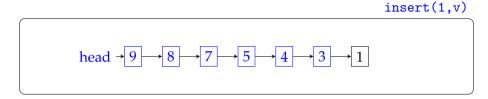


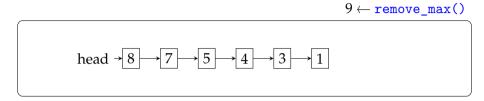
- Insert: O(1)
- Remove: O(n)

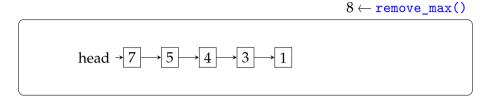






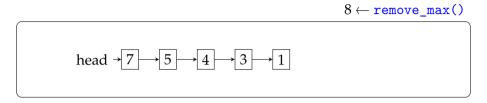






- Insert: O(n)
- Remove: O(1)

sorted list

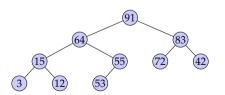


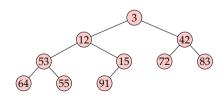
- Insert: O(n)
- Remove: O(1)

We can do better on average (coming soon).

Binary heaps

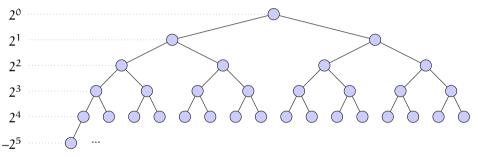
- A binary heap is a binary tree where the nodes store items with an ordering relation. A binary heap has two properties:
 - 1. Shape: a binary heap is a complete binary tree
 - all levels of the tree, except possibly the last one, are full
 - all empty slots (if any) are to the right of the filled nodes at the lowest level
 - 2. Heap order:
 - max-heap Parents' keys are larger than children's keys
 - min-heap Parents' keys are smaller than children's keys



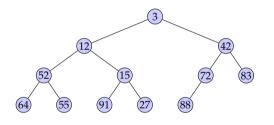


Height of a binary heap

Height of a binary heap is [log n]

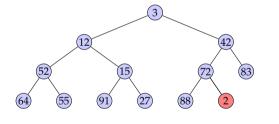


- At least 2^h nodes $\Rightarrow h \le \log n$
- At most $2^{h+1} 1$ nodes $\Rightarrow h \ge \log(n+1) 1$



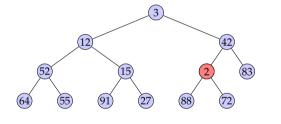
- Add the new element to the fist available slot
- "Bubble up" until the heap property is satisfied
- At most $h = \log n$ comparisons/swaps

8 / 22

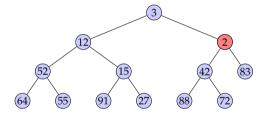


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8 / 22

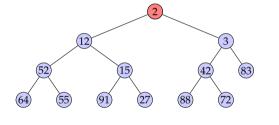


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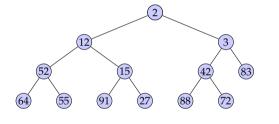
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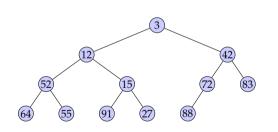


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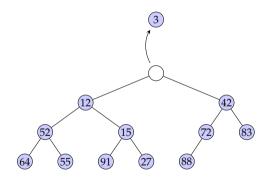
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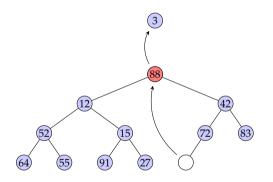
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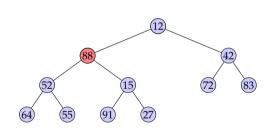
- The item to be removed is at the root
- We replace root with the element at the last slot
- "Bubble down" until the heap property is satisfied



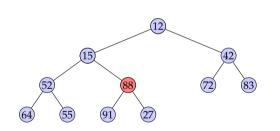
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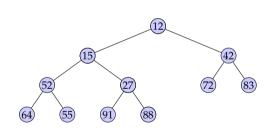
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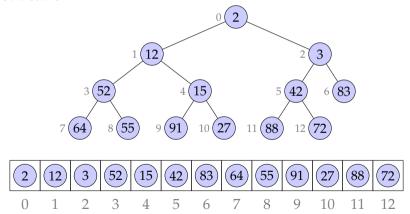
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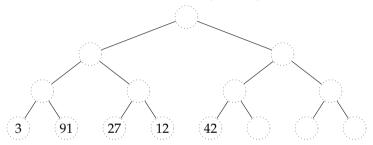
Array based implementation of heaps

 As any complete binary tree, heaps can be stored efficiently using an array data structure

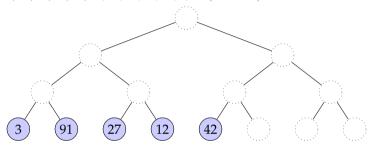


- For n items, we can construct a heap by inserting each key to the heap in $O(n \log n)$ time
- If we have the complete list, there is a bottom-up procedure that runs in O(n) time
 - 1. First fill the leaf nodes, single-node trees satisfy the heap property
 - $h = |\log n|$
 - we have $2^h 1$ internal nodes
 - $n (2^h 1)$ leaf nodes
 - 2. Fill the next level, "bubble down" if necessary
 - 3. Repeat 2 until all elements are inserted, and heap property is satisfied

demonstration with: 3, 91, 27, 12, 42, 88, 72, 52, 15, 64, 2, 83 (12 items)

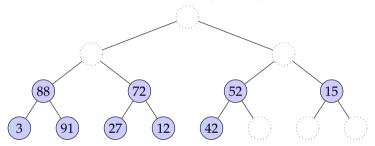


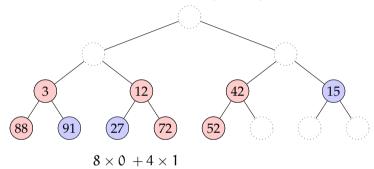
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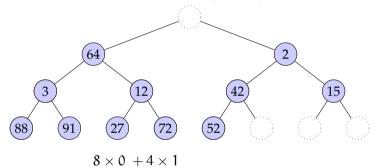


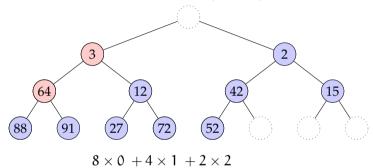
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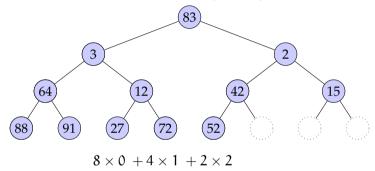




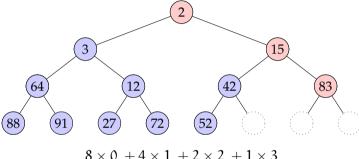




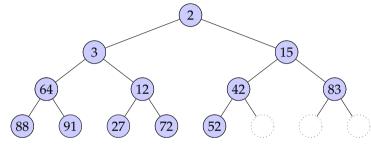
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$$8 \times 0 + 4 \times 1 + 2 \times 2 + 1 \times 3$$



$$8 \times 0 + 4 \times 1 + 2 \times 2 + 1 \times 3$$

$$T(n) = \sum_{i=0}^{h} i \times 2^{h-i} = \sum_{i=0}^{h} i \times \frac{2^{h}}{2^{i}} = 2^{h} \sum_{i=0}^{h} \frac{i}{2^{i}} = \frac{n+1}{2} \underbrace{\sum_{i=0}^{h} \frac{i}{2^{i}}}_{constant} = O(n)$$

```
Implementation insert() remove()
Unsorted list
```

Implementation	insert()	remove()
Unsorted list Sorted list	O(1)	O(n)

Implementation	insert()	remove()
Unsorted list Sorted list Binary heap	O(1) O(n)	O(n) O(1)

Implementation	insert()	remove()
Unsorted list	O(1)	O(n)
Sorted list	O(n)	O(1)
Binary heap	$O(\log \mathfrak{n})$	$O(\log \mathfrak{n})$

- Some improvements are possible, such as
 - d-ary heaps: $O(\log_d n)$ insert, $O(d \log_d n)$ remove
 - Fibonacci heaps: O(1) insert, $O(\log n)$ remove

Python standard heap implementation

- Python standard heapq module allows maintaining a list (array) based heap
 - The heappush (h, e) insert e into heap h
 - The heappop(h) return the minimum value from heap h
 - The heapify(h) construct a heap from given list heappush(h)

Sorting with priority queues

- Inserting the items in a priority queue and removing them effectively sorts the given array
- There is an interesting connection with this approach and some sorting algorithms
 - If we use a sorted list, the algorithm is equivalent to the insertion sort $O(n^2)$
 - If we use a unsorted list, the algorithm is equivalent to the selection sort $O(n^2)$
 - If use a binary heap, we get an $O(n \log n)$ algorithm (heap sort)

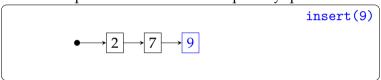
Step 1: insert the items to a priority queue



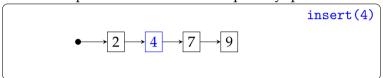
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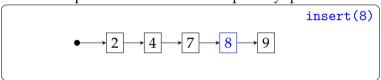
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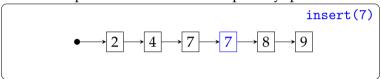
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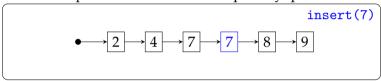


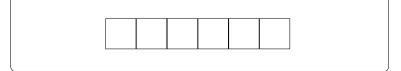
Step 1: insert the items to a priority queue



priority queues implemented with sorted lists – sorting: 7, 2, 9, 4, 8, 7

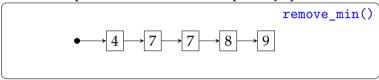
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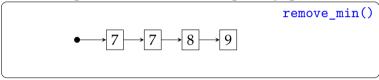
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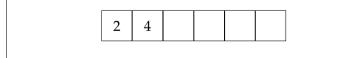




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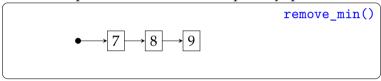
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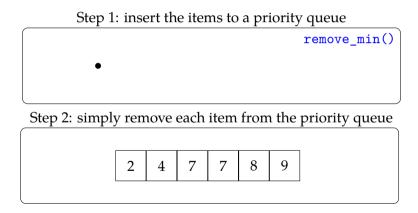


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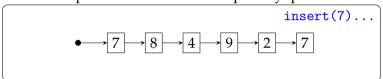
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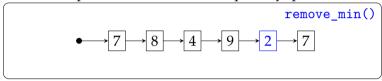


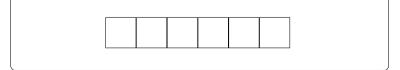
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priority queues implemented with unsorted lists – sorting: 7, 2, 9, 4, 8, 7

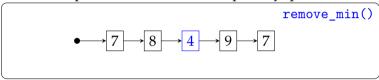
Step 1: insert the items to a priority queue





priority queues implemented with unsorted lists – sorting: 7, 2, 9, 4, 8, 7

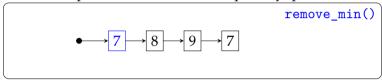
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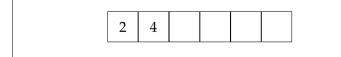




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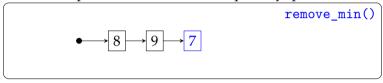
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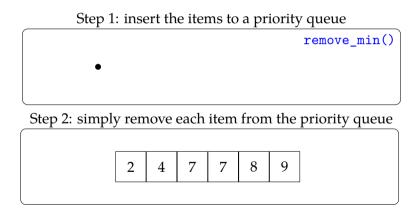


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Step 1: insert the items to a priority queue







Sorting with heaps

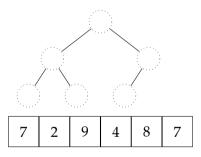
a first attempt

- The idea is simple: as before, insert all items to the heap
- Remove them in order
- Complexity of $O(n \log n)$
- However,
 - not stable
 - not in-place: needs O(n) extra space (we can fix this)

```
def heap_sort(seq):
  heap = []
  for item in seq:
    heappush(item)
  for i in range(len(seq)):
    seq[i] = heappop(heap)
```

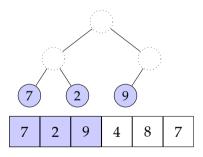
In-place heap sort

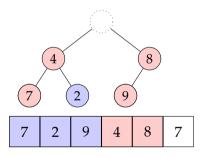
step 1: bottom-up heap construction—sorting: 7, 2, 9, 4, 8, 7

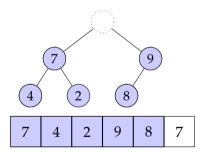


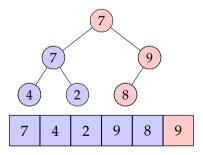
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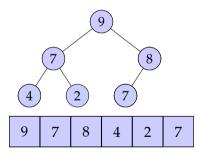
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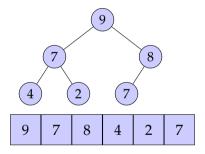


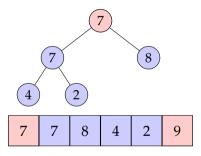


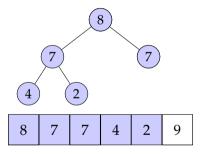


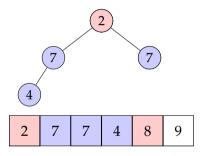


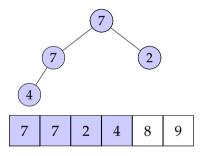


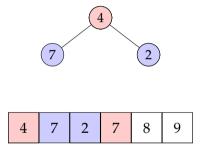


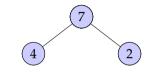


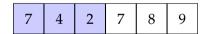


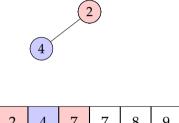




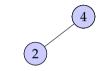








step 2: iteratively remove the maximum element, place it at the end



4 2 7 7 8 9

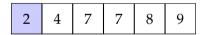
step 2: iteratively remove the maximum element, place it at the end

(2)

2 4 7 7 8 9

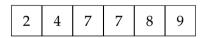
step 2: iteratively remove the maximum element, place it at the end





Heap construction: $O(n) + n \times remove_min()$: $O(n \log n) = O(n \log n)$

step 2: iteratively remove the maximum element, place it at the end



Heap construction: $O(n) + n \times remove_min()$: $O(n \log n) = O(n \log n)$

A summary of sorting algorithms so far

Algorithm	worst	average	best	memory	in-place	stable
Bubble sort	n^2	n^2	n	1	yes	yes
Selection sort	n^2	n^2	n^2	1	yes	no
Insertion sort	n^2	n^2	n	1	yes	yes
Merge sort	$n \log n$	$n \log n$	$n \log n$	n	no	yes
Quicksort	n^2	$n \log n$	$n \log n$	$\log n$	yes	no
Bucket sort	n^2	n^2/k	n	kn	no	yes
Heap sort	$n \log n$	$n \log n$	n	1	yes	no
Timsort	$n \log n$	$n \log n$	n	n	no	yes

A summary of sorting algorithms so far

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Insertion sort	n^2	n^2	n	1	yes	yes
Merge sort	$n \log n$	$n \log n$	$n \log n$	n	no	yes
Quicksort	n^2	$n \log n$	$n \log n$	$\log \mathfrak{n}$	yes	no
Bucket sort	n^2	n^2/k	n	kn	no	yes
Heap sort	$n \log n$	$n \log n$	n	1	yes	no
Timsort	$n \log n$	$n \log n$	n	n	no	yes
?	$n \log n$	$n \log n$	n	1	yes	yes

Summary

- A priority queue is a useful ADT for many purposes
- Binary heaps implement priority queues efficiently
- Heap sort is an efficient algorithm based on priority queue implementation with heaps (Goodrich, Tamassia, and Goldwasser 2013, ch. 9)

Next:

- Graphs
- Reading: Goodrich, Tamassia, and Goldwasser (2013, chapter 14)

Acknowledgments, credits, references



Goodrich, Michael T., Roberto Tamassia, and Michael H. Goldwasser (2013). *Data Structures and Algorithms in Python*. John Wiley & Sons, Incorporated. ISBN: 9781118476734.

A.3

A.4

A.5