



Design and Analysis
of Algorithms I

Data Structures

Heaps and Their Applications

Heap: Supported Operations

- a container for objects that have keys
 - employer records, network edges, events, etc.

INSERT: add a new object to a heap.

Running time: $O(\log n)$

EXTRACT-MIN: remove an object in heap with a minimum key value. [ties broken arbitrarily]

Running time: $O(\log n)$ $\Sigma n = \#$ of objects in heap

Also: HEAPIFY (n batched Inserts) in $O(n)$ time, DELETE ($O(\log n)$ time)

(equally well,
EXTRACT
MAX)

Application: Sorting

Canonical use of heap: fast way to do repeated minimum computations.

Example: SelectionSort $\approx O(n)$ linear scans, $O(n^2)$ runtime on array of length n .

HeapSort: ① insert all n array elements into a heap
② Extract-Min to pluck out elements in sorted order

Running time = $2n$ heap operations = $O(n \log n)$ time.

\Rightarrow optimal for a "comparison-based" sorting algorithm!

Application: Event Manager

"Priority queue" - synonym for a heap.

Example: Simulation (e.g., for a video game)

- objects = event records [action/update to occur at given time in the future]
- key = time event scheduled to occur
- Extract-Min \Rightarrow yields the next scheduled event

Application: Median Maintenance

I give you: a sequence x_1, \dots, x_n of numbers, one-by-one.

You tell me: at each time step i , the median of $\{x_1, \dots, x_i\}$.

Constraint: use $O(\log i)$ time at each step i .

Solution: maintain heaps H_{low} : supports EXTRACT MAX
 H_{high} : supports EXTRACT MIN

Key idea: maintain invariant that $\approx i/2$ smallest (largest) elements in H_{low} (H_{high})

You CHECK: ① can maintain invariant with $O(\log i)$ work

② given invariant, can compute median in $O(\log i)$ work

Application: Speeding Up Dijkstra

Dijkstra's Shortest-Path Algorithm

- naive implementation \Rightarrow runtime = $\Theta(nm)$
- with heaps \Rightarrow run time = $O(m \log n)$

