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1. Introduction

A heap, also known as a priority queue, is a specialized data structure that allows efficient insertion, deletion, and retrieval of elements with the highest (or lowest) priority. It is commonly used to solve problems that involve finding the minimum or maximum element repeatedly.

A heap is typically implemented as a binary tree, where each node satisfies the heap property. In a min heap, for example, the value of each node is smaller than or equal to the values of its children. In a max heap, the value of each node is greater than or equal to the values of its children.

The key operations supported by a heap are:

1. Insertion: Adding an element to the heap. The new element is placed in the appropriate position to maintain the heap property.
2. Deletion: Removing the element with the highest (or lowest) priority from the heap. The root node is typically removed, and the heap is adjusted to ensure the heap property is maintained.
3. Retrieval: Accessing the element with the highest (or lowest) priority without removing it from the heap. In a max heap, the maximum element is always the root node, and in a min heap, the minimum element is the root node.

Heaps are commonly used in various algorithms and data structures, such as Dijkstra's algorithm for finding the shortest path, Prim's algorithm for finding minimum spanning trees, and heap sort for sorting elements efficiently.

The time complexity of basic heap operations depends on the implementation. In a binary heap, insertion and deletion operations have a time complexity of $O(\log n)$, where n is the number of elements in the heap. Retrieval of the highest (or lowest) priority element can be done in constant time, $O(1)$.

2. Algorithm

Here is a high-level algorithm for a heap (priority queue):

1. Initialize an empty heap data structure.
2. Implement the necessary operations: insertion, deletion, and retrieval.
 - Insertion:
 - Add the new element to the bottom of the heap.
 - Compare the new element with its parent and swap if necessary to maintain the heap property (e.g., in a min heap, the parent should have a smaller value than its children).
 - Repeat the comparison and swapping process with the element's parent until the heap property is satisfied.
 - Deletion:
 - Remove the element at the root (which has the highest or lowest priority depending on the heap type).
 - Replace the root with the last element in the heap.
 - Compare the new root with its children and swap it with the smaller or larger child (depending on the heap type) to maintain the heap property.
 - Repeat the comparison and swapping process with the element's new children until the heap property is satisfied.
 - Retrieval:
 - Return the value of the root element, which represents the highest or lowest priority depending on the heap type.
3. Optionally, implement additional operations based on your requirements, such as updating an element's priority or checking if the heap is empty.
4. Analyze the time complexity of the implemented operations.
 - Insertion and deletion operations typically have a time complexity of $O(\log n)$, where n is the number of elements in the heap.
 - Retrieval of the highest or lowest priority element can be done in constant time, $O(1)$.
5. Use the heap/priority queue in your desired application or algorithm.

3. Example Usage

Here's an example of how you can use the `heapq` module in Python to work with heaps (priority queues):

```

import heapq

# Create an empty heap
heap = []

# Insert elements into the heap
heapq.heappush(heap, 5)
heapq.heappush(heap, 2)
heapq.heappush(heap, 10)
heapq.heappush(heap, 1)

# Retrieve the element with the highest priority (minimum value)
min_value = heapq.heappop(heap)
print(min_value) # Output: 1

# Insert another element into the heap
heapq.heappush(heap, 3)

# Retrieve the element with the highest priority again
min_value = heapq.heappop(heap)
print(min_value) # Output: 2

# Convert a list into a heap
list_values = [7, 4, 9, 6]
heapq.heapify(list_values)

# Retrieve the element with the highest priority from the list-converted heap
min_value = heapq.heappop(list_values)
print(min_value) # Output: 4

```

In this example, the `heapq` module is used to create and manipulate a heap. The `heappush` function is used to insert elements into the heap, and the `heappop` function is used to retrieve the element with the highest priority (minimum value in this case).

You can also convert an existing list into a heap using the `heapify` function from the `heapq` module. This allows you to easily convert a list into a heap and perform heap operations on it.

Note that `heapq` implements a min heap by default. If you want to create a max heap instead, you can invert the sign of the elements or use a tuple with a custom key that represents the priority.

The `heapq` module provides a convenient way to work with heaps (priority queues) in Python, making it easier to handle elements based on their priority.

✓ 4. Advantages

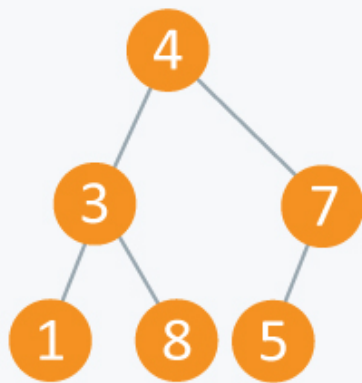
Using a heap (priority queue) data structure offers several advantages in various scenarios:

1. **Efficient Priority-based Operations:** Heaps allow efficient insertion, deletion, and retrieval of elements with the highest (or lowest) priority. These operations have a time complexity of $O(\log n)$, making heaps suitable for applications that require efficient prioritization.
2. **Dynamic Ordering:** Heaps maintain a dynamic ordering of elements based on their priority. This is particularly useful when the relative priorities of elements change frequently. The heap automatically adjusts its structure to ensure the highest (or lowest) priority element is readily accessible.
3. **Space Efficiency:** Heaps typically use an array-based or tree-based representation, resulting in efficient space utilization. They do not require additional pointers or metadata for maintaining the priority order, making them compact and memory-efficient data structures.
4. **Versatility:** Heaps have a variety of applications and can be used to solve different types of problems. They are commonly used in algorithms such as Dijkstra's algorithm, Prim's algorithm, and Huffman coding. Heaps also form the basis for sorting algorithms like heap sort.
5. **Simplicity:** The basic heap operations (insertion, deletion, retrieval) are relatively simple to implement and understand. Libraries or modules, such as Python's `heapq`, provide ready-to-use implementations, reducing the need for manual implementation.
6. **Partially Sorted Output:** Heap operations can provide a partially sorted output. For instance, by repeatedly extracting the minimum element from a min heap, you can obtain the elements in increasing order. This property can be advantageous in certain scenarios, such as finding the top k elements in a large dataset.
7. **Priority Updates:** Heaps allow efficient priority updates for elements already present in the heap. If the priority of an element changes, you can update it in $O(\log n)$ time complexity, which is faster than removing and re-inserting the element.

These advantages make heaps a valuable tool for optimizing algorithms, solving problems with priority-based requirements, and efficiently managing elements based on their ordering.

Arr		4	3	7	1	8	5
	0	1	2	3	4	5	6

Initial Elements



Max Heap

