

In general, I use two heaps to implement the median-heap data structure.

One is a min-heap to store the element on the right of the median, say, `right_heap`.

The other is a max-heap to store the element on the left of the median, and the median itself, say `left_heap`.

If there are odd elements, $\text{size}(\text{left_heap}) == \text{size}(\text{right_heap}) + 1$, else even elements, $\text{size}(\text{left_heap}) == \text{size}(\text{right_heap})$.

The median is always the peek element of the `left_heap`.

Firstly, the median-heap is initialized with empty `left_heap` and `right_heap`:

```
class MedianHeap:
    def __init__(self):
        self.left_heap = []
        self.right_heap = []
```

For Build(S), I will apply 3 helper functions:

1. `find_median(arr)`: given an unsorted array, it can output the lower median of the array in $O(n)$ time by a median of median algorithm.
2. `make_min_heap(arr)`: this function works like `heapq.heapify` in Python, it turns an array into a min-heap in $O(n)$ time.
3. `make_max_heap(arr)`: this function will make an array into a max-heap in $O(n)$ time.

The implementations of `make_min_heap` and `make_max_heap` are similar.

Let the last level, say the h^{th} level has m arbitrary elements, then we add $m/2$ arbitrary elements to the $(h - 1)^{\text{th}}$ level, and make adjustments so that each $(h-1)^{\text{th}}$ node with two children is a heap(max or min). For each adjustment, only at most 1 operation is needed, to switch the upper node down or not depending on comparison. For example, if a min-heap is needed but the upper node is larger than lower node, just switch the upper node with the smaller lower node and a min-heap is made. Thus, for the $(h - 1)^{\text{th}}$ level, at most $h/2 * 1$ operations are needed to build a heap.

For the $(h - 2)^{\text{th}}$ level, # operations = $m/2^2 * 2$, ..., for the $(h - i)^{\text{th}}$ level, # operations = $m/2^i * i$.

$$\# \text{ total operations} = \frac{m}{2} \times 1 + \frac{m}{2^2} \times 2 + \dots + \frac{m}{2^h} \times h,$$

$$h = \log n,$$

$$\frac{m}{2^h} = 1 \cdot m = 2^h = n.$$

$$S = m \left(\frac{1}{2} + \frac{1}{2^2} \times 2 + \dots + \frac{1}{2^h} \times h \right)$$

$$\frac{1}{2}S = m \left(\frac{1}{2^2} + \dots + \frac{1}{2^h} \times (h-1) + \frac{1}{2^{h+1}} \times h \right)$$

$$S - \frac{1}{2}S = m \left(\frac{1}{2} + \frac{1}{2^2} + \dots + \frac{1}{2^h} - \frac{h}{2^{h+1}} \right)$$

$$= m \left(\frac{\frac{1}{2} \left(1 - \frac{1}{2^h} \right)}{1 - \frac{1}{2}} - \frac{h}{2^{h+1}} \right)$$

$$= m \left(1 - \frac{1}{2^h} - \frac{h}{2^{h+1}} \right).$$

$$S = 2m \left(1 - \frac{1}{2^h} - \frac{h}{2^{h+1}} \right).$$

$$m = n, \quad h = \log n, \quad 2^h = n.$$

$$\therefore S = 2n \left(1 - \frac{1}{n} - \frac{\log n}{2n} \right) = O(n).$$

With the clarification above, the Build(S) function is like this:

1. Find the median by find_median in O(n) time.
2. Get the half_size of the array by dividing in O(1) time.
3. Loop the elements in the array, compared to the median and add to left_heap or right_heap depending on 2 conditions in O(n) time:

If size of left_heap > half_size + 1, elements must be added to the right_heap;

If size of `right_heap` > `half_size`, elements must be added to the `left_heap`;

If both sizes are suited, elements less than or equal to the median are added to the `left_heap`, else to the `right_heap`.

4. Make the `left_heap` a max-heap and the `right_heap` a min-heap in $O(n)$ time.

```
def build(self, arr):
    # step 1: find the median of the given unsorted array,  $O(n)$  time.
    median = find_median(arr)
    # step 2: get the half_size of the array,  $O(1)$  time.
    half_size = len(arr) // 2
    # step 3: loop the arr,
    # if the element is less than or equal to the median, add it to self.left_heap
    # if the element is larger than median, add it to self.right_heap
    # if the size of the right_heap is already larger than half_size, add elements to the left_heap
    # if the size of the left_heap is larger than half_size + 1, add elements to the right_heap
    #  $O(n)$  time
    for element in arr:
        if element <= median:
            if len(self.left_heap) > half_size + 1:
                self.right_heap.append(element)
            else:
                self.left_heap.append(element)
        else:
            if len(self.right_heap) > half_size:
                self.left_heap.append(element)
            else:
                self.right_heap.append(element)
    # step 4: make self.left a max-heap, and make self.right a min-heap
    #  $O(n)$  time
    make_min_heap(self.right_heap)
    make_max_heap(self.left_heap)
```

The 4 steps in total take $O(n)$ time.

For `Insert(x)`, I will apply 4 helper functions:

1. `max_heap_insert(heap, x)`, this function takes in a max-heap and an element x , and insert x to the heap.

This function takes $O(\log n)$ time, because x is added to the end/right-most of the heap, and swapped up if it is larger than its parent to maintain a max-heap. As the height of heap is $\log n$, the swap-up operation will take at most $\log n$ times, thus the time complexity of inserting is $O(\log n)$.

2. `max_heap_pop(heap)`, this function takes in a max-heap, pop the root of the heap and maintain as a max-heap.

This function takes $O(\log n)$ time, because we firstly swap the root with the end/right-most of the heap, drop the end of the heap, i.e. the root value in $O(1)$ time, and swap the new root value down if it is smaller than its child to maintain a max-heap. As the height of heap is $\log n$, the swap-down operation will take at most $\log n$ times, thus the time complexity of popping is $O(\log n)$.

3. `min_heap_insert(heap, x)`, this function takes in a min-heap and an element x , and insert x to the heap in $O(\log n)$ time. The implementation is similar to `max_heap_insert`.

4. `min_heap_pop(heap)`, this function takes in a min-heap, pop the root of the heap and maintain as a min-heap in $O(\log n)$ time. The implementation is similar to `max_heap_pop`.

With the clarification above, the **Insert(x) function** is like this:

1. Get the current median in $O(1)$ time by peek function.
2. Compare the x with the `curr_median`:

If $x > \text{curr_median}$, add x to the `right_heap` in $O(\log n)$ time by `min_heap_insert` function in $O(\log n)$ time, and balance the `left_heap` and the `right_heap` in $O(\log n)$ time.

For balance, if $\text{size of right_heap} > \text{size of left_heap}$, we pop out the minimum of `right_heap` by `min_heap_pop` function in $O(\log n)$ time and insert this element into the `left_heap` by `max_heap_insert` function in $O(\log n)$ time.

Else, add x to the `left_heap` by `max_heap_insert` function in $O(\log n)$ time, and balance the two heaps in $O(\log n)$ time.

For balance, if $\text{size of left_heap} > \text{size of right_heap} + 1$, we pop out the maximum of `left_heap` by `max_heap_pop` function in $O(\log n)$ time and insert this element into the `right_heap` by `min_heap_insert` function in $O(\log n)$ time.

Above steps take $O(\log n)$ time.

```
def insert(self, x):
    # step 1: get the current median in O(1) time
    curr_median = self.peak()
    # step 2: compare x with current median, add it into left_heap or right_heap,
    # and balance the left_heap and right_heap in O(logn) time.
    if x > curr_median:
        min_heap_insert(self.right_heap, x)
        # balance
        if len(self.right_heap) > len(self.left_heap):
            element = min_heap_pop(self.right_heap)
            max_heap_insert(self.left_heap, element)
    else: # x <= curr_median
        max_heap_insert(self.left_heap, x)
        # balance
        if len(self.left_heap) - len(self.right_heap) > 1:
            element = max_heap_pop(self.left_heap)
            min_heap_insert(self.right_heap, element)
```

For **Peek()**, the root of left_heap is just the median we want, we can return it in $O(1)$ time.

```
def peek(self):  
    """  
    ~~~~~  
    return the root of the left_heap, i.e. the max-heap  
    ~~~~~  
    """  
    return self.left_heap[0]
```

For **Extract()**:

1. The root of left_heap is the median, we pop it in $O(\log n)$ time by max_heap_pop function.
2. Balance the left_heap and right_heap if size of right_heap > size of left_heap by pop the minimum from right_heap in $O(\log n)$ time and insert it into left_heap in $O(\log n)$ time.

Such steps takes $O(\log n)$ time in total.

```
def extract(self):  
    # step 1: pop the median, i.e. the root of the left_heap in  $O(\log n)$  time.  
    median = max_heap_pop(self.left_heap)  
    # balance in  $O(\log n)$  time.  
    if len(self.right_heap) > len(self.left_heap):  
        element = min_heap_pop(self.right_heap)  
        max_heap_insert(self.left_heap, element)  
    return median
```