Homework 5: Binary Heap

Code due: Tue., Nov. 12 at 11:59 PM (via GSC)
Self-eval due: Thu., Nov. 14 at 11:59 PM (on GSC)

You may work on your own or with one (1) partner.

In this assignment, you will implement a fixed-size binary heap. The structure of the heap is already defined for you in \[\text{binheap.rkt}\] The elements of the heap are stored in a DSSL2 vector. Each heap also contains a comparison function for ordering the elements of the heap, so that your implementation can support heaps of integers, heaps of strings, heaps of whatits, heaps of sporkles, etc.

Interface overview

Your BinHeap class must implement the priority queue abstract data type, as specified by this DSSL2 interface:

\[
\text{interface PRIORITY\_QUEUE}[X]:
\begin{align*}
\text{def } & \text{len}(\text{self}) \rightarrow \text{nat} \? \\
\text{def } & \text{find\_min}(\text{self}) \rightarrow X \\
\text{def } & \text{remove\_min}(\text{self}) \rightarrow \text{NoneC} \\
\text{def } & \text{insert}(\text{self}, \text{element}: X) \rightarrow \text{NoneC}
\end{align*}
\]

That is, the PRIORITY\_QUEUE interface is parameterized by an element type \(X\), and declares four operations:

- The \text{len} method returns the number of elements in the priority queue.
- The \text{find\_min} method returns the “smallest”\(^2\) element of the priority queue, or calls \text{error} if it is empty. (If there’s more than one element tied for smallest then it doesn’t matter which of those is returned, so long as \text{find\_min} is deterministic.)

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2. Through, when we say “smallest” we mean “smallest according to the ordering function the heap was created with,” which may well implement some other order entirely. When the ordering function is greater-than then “smallest” actually means largest.
• The `remove_min` method removes the smallest element of the priority queue, or calls `error` if it is empty. (If there is more than one element tied for smallest, then `remove_min` must be consistent with `find_min`, in the sense that it removes the same element that `find_min` would return.)

• The `insert` method takes a new element and adds it to the priority queue if there’s sufficient space for an additional element; if there isn’t room then it calls `error` instead.

DSSL2 interfaces specify how objects can be used, but not how they are constructed. To understand that, we must describe the constructor of the `BinHeap` class. Like the `PRIORITY_QUEUE` interface that it implements, the `BinHeap` class is parameterized by an element type `X`; its constructor takes two parameters, `capacity` and `lt?`, as seen here:

```python
class BinHeap[X] (PRIORITY_QUEUE):
    # ... fields omitted ...

def __init__(self, capacity, lt?):
    # ... initialization omitted ...

    # ... lots more ...
```

The value given for `capacity` must be a natural number, which will be the size of the vector used to store the elements. Thus, `capacity` determines the number of elements that can be stored before the `insert` method signals an error.

The purpose of the `lt?` parameter is to allow clients of the `BinHeap` class to specify what elements they want considered “least,” that is, what ordering the heap should implement. A client does so by providing a function for the heap to use for comparing elements when restoring its invariant. If the client wants a min-heap then it will pass an actual less-than function for `lt?`, and that will cause the heap to return its smallest elements first; but if the client wants a max-heap then it can pass a greater-than function as `lt?` and it will get a largest-first heap. If a client wants some other order, or if it wants a bespoke element type, then it can provide a comparison function for that.

In general, the value given for `lt?` must be a binary predicate for comparing two elements of type `X`. In particular, given two elements (i.e., `X`s) `a` and
b, \( \text{lt}(a, b) \) must return \textbf{True} if \( a \) is to be considered less than \( b \) for the purposes of ordering the heap, and it must return \textbf{False} if \( a \) is to be considered greater than or equal to \( b \). The \textbf{BinHeap} class stores this function in a field and uses it to compare elements and maintain its invariant.

Thus, when constructing a heap, we need to provide an heap ordering predicate appropriate for the intended element type and order. For example, we could construct a heap for holding up to 20 integers like this:

```plaintext
let h = BinHeap(20, lambda a, b: a < b)
```

The \texttt{lambda} given as the second argument is a two argument function that compares its arguments using DSSL2’s built-in \texttt{<} operator. Since that operator works on strings as well, we can construct a heap of strings, ordered lexicographically, the same way.

But what if we wanted to construct a heap of vertex–distance pairs, as used in Dijkstra’s algorithm? Then we’d need to pass \texttt{BinHeap} a function for comparing such things:

```plaintext
struct VertexWithDistance:
    let vertex: nat?
    let distance: num?

let h = BinHeap(n, lambda a, b: a.distance < b.distance)
```

And as discussed above, supplying a heap ordering predicate also lets us make a heap that returns the largest element first instead of the smallest.

```plaintext
let h = BinHeap(10, lambda a, b: a > b)
```

In the literature, not everyone agrees on whether a “heap,” order unspecified, is smallest-first or largest-first. It can be confusing. The heap you are implementing can be used as either a smallest-first “min-heap” or a largest-first “max-heap,” depending on the predicate given for \texttt{lt?}. When you are implementing it, however, you will do best if you 1) think of it as a min-heap, where \texttt{lt?} means “less than,” and 2) never assume you know how to compare elements any other way, because the client gets to determine the element type and order.
Representation overview

A binary heap is a complete binary tree satisfying the heap property. Completeness means that each level is full except the last one, which is filled from the left; the heap property means that each node’s value is less than or equal to its children’s values, according to the heap ordering predicate (lt?).

Here is an example of a binary heap of integers drawn as a tree:

```
    2
   / \
  3   6
 /   /\n7   5  8
```

In practice, a binary heap is rarely represented as an actual tree made of links and nodes. Instead, the binary tree completeness property means that the elements can be stored in level order in an array, which is more space- and cache-efficient. In particular, in this assignment we will represent a heap as a class with three fields. Here is the same heap as above in our actual heap representation:

```
data 6
len 2
lt? λ
```

The first field, _data, stores a reference to the vector holding the elements. In order to allow the heap to grow, we do not necessarily use the full capacity of the vector; instead we store the actual number of elements in _len, the second field. The remainder of the vector is then filled with some dummy value, such as None, depending on the language.

The third field, _lt?, stores the heap ordering predicate that was given to the BinHeap constructor. This is necessary because your heap must work with any element type whatsoever, not just with numbers or strings. If elements were only numbers or strings, you could just compare them using DSSL2’s built-in < operator, but that won’t work for, say, bank account structs, employees, or weighted graph edges.

When restoring the invariant after insertion or removal, you have to compare elements. In particular, at each step, you will be considering some elements
parent and child, where parent is directly above child in the heap. The invariant says that parent must be less than or equal to child. So, the invariant is broken if child is less than parent. Assuming we are in a method where the BinHeap object is named self, the invariant is broken if self._lt?(child, parent).

**Algorithms**

Heap insertion and deletion both involve breaking and then restoring the heap invariant. Each requires a different algorithm for restoring the invariant.

**Insertion**

To insert into the heap, first we place the new element at the next available position in the vector, which is self._len, and we increment self._len. Then we must restore the invariant. Since the new element is at the bottom of the tree, it can only possibly move up; hence the algorithm for restoring the invariant after an insertion is often known as **bubbling up**.

To bubble up the new element, we compare it to its parent using the heap ordering predicate. If the invariant is broken—that is, if the new element is less than its parent—then we swap the new element with its parent, and continue checking upward at the new element’s new position, until it rises high enough that either the invariant holds between the new element and its parent, or the new element has no parent.

**Removal**

The element to remove is always at the root of the tree (index 0 in the vector), but the binary tree completeness invariant requires that the heap shrink at the end (index self._len - 1 in the vector). So to remove an element, first we move the last element to the root position, overwriting the old copy with None (which aids in debugging) and decrementing the size. Then we must restore the invariant, because the element that we moved to the root position might be in conflict with one or both of its children, if it has any. From the root of the tree, the element will only have to move down; hence the algorithm
for restoring the invariant after a remove is sometimes known as percolating down. We begin percolating down at the root.

To percolate down an element, we need to find out whether it has a child smaller than itself; and if it has two children, we need to find the smaller of the two. This can be a bit tricky, since an element may have two children, a left child only, or no children. (Completeness forbids a right child without a left.) If the invariant is broken between the element and a child—that is, if the child is less than the element—then we swap the element with the smallest of its children, and continue checking downward at the element’s new position, until the element sinks far enough that the invariant holds between the element and any children.

**Heap sort**

Given a vector and an ordering predicate, it’s possible to permute the vector into heap order in $O(n)$ time in bottom-up fashion: Iterate through the elements of the vector in reverse and percolate down each of them in turn. After that, it is possible to permute the heap-ordered vector into a sorted vector in $O(n \log n)$ time, by repeated percolating down, as follows. Think of the vector as partitioned into an (initially empty) sorted section at one end, and the remaining heap-ordered section. Move an element from the heap-ordered section to the sorted section by simulating a removal from the heap-ordered section; this causes the “heap” to shrink and leaves the removed element in the correct position in the sorted section. Repeat until all elements are in the sorted section.

Thus, the ideas behind a binary heap can be used to sort a vector in-place, with no auxiliary storage, in $O(n \log n)$ time. That said, you do not need to implement a fully in-place heap sort as described above for this course, since your `BinHeap` class likely isn’t designed to be used in that way. (How could you reuse most of the code?) Instead, the `heap_sort` function you implement

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3It probably isn’t immediately clear why this `build-heap` procedure isn’t $O(n \log n)$. We’re doing $n$ percolate-downs on a log-height tree, right? Wrong! Consider that for the second half of the vector, there are no children, so there’s no percolation to do; for the previous quarter, percolation involves at most one step, and for the previous eighth of the vector, at most two. Overall, only $\frac{n}{2^r}$ percolations can take $k$ or more steps, and if you add that up, you find that the number of log-time percolations is sub-linear.
should sort a vector of \( n \) elements using \( \mathcal{O}(n) \) auxiliary space in the form of a BinHeap that it creates and uses as internal scratch space. The new heap will need to hold all the elements of the vector at once, so that determines its size. Iterate through the vector, inserting each element into the heap. Then iterate over the indices of the vector in order, overwriting each element in the vector with the next smallest element from the heap. This algorithm will terminate with a sorted vector and an empty heap.

Your task

In \texttt{binheap.rkt} I’ve supplied a skeleton of a class BinHeap that implements the \texttt{PRIORITY_QUEUE} interface. The class already defines the three fields \_data, \_len, and \_lt? as described previously, as well as the \texttt{__init__} and \texttt{len} methods. I’ve provided stubs for the three additional required methods and four suggested helper methods. After the class and some tests, you will find stubs for three further suggested helper functions, and finally a stub for the \texttt{heap_sort} function that you must complete.

The four suggested helper methods and three suggested helper functions all have names beginning with underscore, which makes them private to the class and to the file, respectively. I’ve provided stubs for them in order to help you factor your program effectively, but you needn’t use them, and can safely delete them if you prefer doing things a different way.

The BinHeap class

First, you should complete the definition of the BinHeap class, by implementing the three unimplemented methods from the interface:

1. Complete the definition of method \texttt{find_min}. \( \mathcal{O}(1) \) time
2. Complete the definition of method \texttt{remove_min}. \( \mathcal{O}(\log n) \) time
3. Complete the definition of method \texttt{insert}. \( \mathcal{O}(\log n) \) time

Note the required asymptotic time complexities of the operations.

I have provided two small tests for the BinHeap class, but you will need many more.
The heap_sort function

Second, you must implement sorting, using your BinHeap class and the algorithm described above.

4. Complete the definition of the heap_sort function. \( O(n \log n) \) time

Note the required asymptotic time complexity of the function.

I have provided three small tests for heap_sort, but you almost certainly should write more. It’s relatively easy to come up with vectors to sort, and that’s a low-effort, high-impact way to exercise your whole heap implementation.

Deliverables

- The provided file named binheap.rkt containing:
  - the completed BinHeap class, with your definitions for the three methods insert, find_min, and remove_min;
  - the heap_sort function fully defined; and
  - sufficient tests to cover all cases.

4[https://bit.ly/2NQz98K]