

Lab07-Amortized Analysis

CS214-Algorithm and Complexity, Xiaofeng Gao & Lei Wang, Spring 2021.

* If there is any problem, please contact TA Yihao Xie.

* Name:BeichenYu Student ID:519030910245 Email:polarisybc@sjtu.edu.cn

1. Suppose we perform a sequence of n operations on a data structure in which the i th operation costs i if i is an exact power of 2, and 1 otherwise. Use an accounting method to determine the amortized cost per operation.

Solution. Assume a cost of \$3 for each operation. The operation from (2^i+1) th to $(2^{i+1}-1)$ th costs \$1 actually, so we have a total credit of $\$2 \times ((2^{i+1}-1) - (2^i+1) + 1) = \$2^{i+1} - 2$. Together with the \$3 of the 2^{i+1} th operation, we have a total of $\$2^{i+1} + 1$ to pay for the actual cost of $\$2^{i+1}$. Thus, for any operation we have nonnegative credit. So the amortized cost per operation is $O(1)$.

□

2. Consider an ordinary **binary min-heap** data structure with n elements supporting the instructions INSERT and EXTRACT-MIN in $O(\log n)$ worst-case time. Give a potential function Φ such that the amortized cost of INSERT is $O(\log n)$ and the amortized cost of EXTRACT-MIN is $O(1)$, and show that it works.

Solution. Let D_i be the heap after the i -th operation. Assume that there are n_i elements in D_i . Assume each INSERT and EXTRACT-MIN operation takes $k \log n$ time at most.

Define that $\Phi(D_i) = kn_i \log n_i$ if $n_i > 0$, and if the heap is empty, $\Phi(D_i) = 0$.

If the i -th operation is an INSERT, we have $n_i = n_{i-1} + 1$.

If the i -th operation inserts into an empty heap, we have $n_{i-1} = 0$ and $n_i = 1$, so we have:

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq k \log 1 + k \log 1 - 0 \\ &= 0\end{aligned}$$

If the i -th operation inserts into a non-empty heap, we have:

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq k \log n_i + kn_i \log n_i - k(n_{i-1}) \log n_{i-1} \\ &= k \log n_i + kn_i \log n_i - k(n_i - 1) \log(n_i - 1) \\ &= k \log n_i + kn_i \log n_i - kn_i \log(n_i - 1) + k \log(n_i - 1) \\ &\leq 2k \log n_i + kn_i \log \frac{n_i}{n_i - 1} \\ &\leq 2k \log n_i + kn_i \left(\frac{n_i}{n_i - 1} - 1 \right) \\ &\leq 2k \log n_i + k \frac{n_i}{n_i - 1} \\ &\leq 2k \log n_i + 2k \\ &= O(\log n_i)\end{aligned}$$

If the i -th operation is an EXTRACT-MIN, we have $n_i = n_{i-1} - 1$.

If before the i -th operation, there is only one element in the heap, we have:

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq k \log 1 + 0 - k \log 1 \\ &= 0\end{aligned}$$

If before the i -th operation, there are more than one elements in the heap, we have:

$$\begin{aligned}\hat{c}_i &= c_i + \Phi(D_i) - \Phi(D_{i-1}) \\ &\leq k \log n_{i-1} + kn_i \log n_i - k(n_{i-1}) \log n_{i-1} \\ &= k \log n_{i-1} + k(n_{i-1} - 1) \log(n_{i-1} - 1) - kn_{i-1} \log n_{i-1} \\ &= k \log \frac{n_{i-1}}{n_{i-1} - 1} + kn_{i-1} \log \frac{n_{i-1} - 1}{n_{i-1}} \\ &\leq k \log \frac{n_{i-1}}{n_{i-1} - 1} \\ &\leq k \log 2 \\ &= O(1)\end{aligned}$$

□

3. Assume we have a set of arrays A_0, A_1, A_2, \dots , where the i^{th} array A_i has a length of 2^i . Whenever an element is inserted into the arrays, we always intend to insert it into A_0 . If A_0 is full then we pop the element in A_0 off and insert it with the new element into A_1 . (Thus, if A_i is already full, we recursively pop all its members off and insert them with the elements popped from A_0, \dots, A_{i-1} and the new element into A_{i+1} until we find an empty array to store the elements.) An illustrative example is shown in Figure 1. Inserting or popping an element take $O(1)$ time.

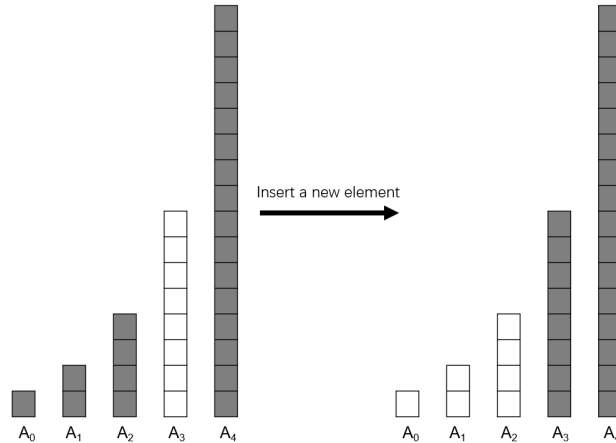


Figure 1: An example of making room for one new element in the set of arrays.

- In the worst case, how long does it take to add a new element into the set of arrays containing n elements?
- Prove that the amortized cost of adding an element is $O(\log n)$ by *Aggregation Analysis*.
- If each array A_i is required to be sorted but elements in different arrays have no relationship with each other, how long does it take in the worst case to search an element in the arrays containing n elements?

- (d) What is the amortized cost of adding an element in the case of (c) if the comparison between two elements also takes $O(1)$ time?

Solution. (a) In the worst case, every nonempty array is full. To add a new element, we have to move all the elements to the next array. As there are already n elements, to move all these elements, we need $O(n)$ time.

- (b) First, to push a new element into the arrays, we need $O(1)$ time.

Second, to move the i^{th} array to the next one, we need to pop and push for $O(2^{i-1})$ times, so the time complexity is $2O(2^{i-1}) = O(2^i)$.

Third, in the first n operations, the times of moving the i^{th} array is:

$$\left\lceil \frac{n - \sum_{k=0}^{i-1} 2^k}{2^{i-1}} \right\rceil = \left\lceil \frac{n - 2^i + 1}{2^{i-1}} \right\rceil = \left\lceil \frac{n+1}{2^{i-1}} \right\rceil - 2$$

Therefore, the time of first n pushing operations is:

$$\begin{aligned} T(n) &= n + \sum_{k=1}^{\lfloor \log_2(n+1) \rfloor} 2^k \left(\left\lceil \frac{n+1}{2^{k-1}} \right\rceil - 2 \right) \\ &= n + \sum_{k=1}^{\lfloor \log_2(n+1) \rfloor} (2[n+1] - 2^{k+1}) \\ &= n + 2[n+1]\lfloor \log_2(n+1) \rfloor - 4(2^{\lfloor \log_2(n+1) \rfloor} - 1) \\ &= O(n \log n) \end{aligned}$$

So the amortized cost of adding an element is:

$$O\left(\frac{n \log n}{n}\right) = O(\log n)$$

- (c) In the i^{th} array A_i with 2^{i-1} elements, using the binary search, the worst case is to find the needed element in the last turn or the needed element is not in that array. That take $\log_2 2^{i-1} = i - 1$ time. So in the worst case in total, it takes:

$$\begin{aligned} T_{worst} &= \sum_{k=1}^{\lfloor \log_2(n+1) \rfloor} (k - 1) \\ &= \frac{\lfloor \log_2(n+1) \rfloor (\lfloor \log_2(n+1) \rfloor - 1)}{2} \\ &= O(\log^2 n) \end{aligned}$$

- (d) Compare with (b), to push a new element into the arrays, we need still $O(1)$ time. When moving the full arrays, we need to make a merge sort. If we merge two arrays with size m and n , the time complexity is $O(m + n)$. So we can merge the arrays from small to big. So merge A_1, A_2, \dots, A_i the time complexity is at most 2^i . So moving the i^{th} array needs 2^{i+1} time.

So we have:

$$\begin{aligned} T(n) &= n + \sum_{k=1}^{\lfloor \log_2(n+1) \rfloor} 2^{k+1} \left(\left\lceil \frac{n+1}{2^{k-1}} \right\rceil - 2 \right) \\ &= n + \sum_{k=1}^{\lfloor \log_2(n+1) \rfloor} (4[n+1] - 2^{k+2}) \\ &= n + 4[n+1]\lfloor \log_2(n+1) \rfloor - 8(2^{\lfloor \log_2(n+1) \rfloor} - 1) \\ &= O(n \log n) \end{aligned}$$

So the amortized cost of adding an element is:

$$O\left(\frac{n \log n}{n}\right) = O(\log n)$$

□

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