



Algorithms and Data Structures Winter Term 2019/2020 Exercise Sheet 4

Remark: For this exercise, watch the relevant parts of the sixth and seventh video lecture.

Exercise 1: Hashing - Collision Resolution with Open Addressing

- (a) Let $h(s, j) := h_1(s) - 2j \pmod m$ and let $h_1(x) = x + 2 \pmod m$. Insert the keys 51, 13, 21, 30, 23, 72 into the hash table of size $m = 7$ using linear probing for collision resolution (the table should show the final state).

0	1	2	3	4	5	6

- (b) Let $h(s, j) := h_1(s) + j \cdot h_2(s) \pmod m$ and let $h_1(x) = x \pmod m$ and $h_2(x) = 1 + (x \pmod {m-1})$. Insert the keys 28, 59, 47, 13, 39, 69, 12 into the hash table of size $m = 11$ using the double hashing probing technique for collision resolution. The hash table below should show the final state.

0	1	2	3	4	5	6	7	8	9	10

- (c) Repeat part (a) using the “ordered hashing” optimization from the lecture.
 (d) Repeat part (b) using the “Robin-Hood hashing” optimization from the lecture.

Exercise 2: Amortized Analysis - Stack with Multipop

Consider the data structure “stack” in which elements can be stored in a *last in first out* manner. For a stack S we have the following operations:

- $S.\text{push}(x)$ puts element x onto S .
- $S.\text{pop}()$ deletes the topmost element of S . Assume $\text{pop}()$ is only called if S is nonempty.
- $S.\text{multipop}(k)$ removes the k top objects of S , popping the entire stack if S contains fewer than k objects.

Assume the costs of $S.\text{push}(x)$ and $S.\text{pop}()$ are 1 and the cost of $S.\text{multipop}(k)$ is $\min(k, |S|)$ where $|S|$ is the current number of elements in S .

Use the bank account paradigm to show that all three operations have constant amortized cost. Assume that S is initially empty.

Exercise 3: Amortized Analysis - a Hierarchy of Arrays

Consider the following data structure. We define arrays A_i (for $i = 0, 1, 2, \dots$), where A_i has size 2^i and stores integer keys in a sorted manner (ascending). During the runtime we ensure that each Array is either completely full, or completely empty.

We informally describe an operation $\text{insert}(k)$. It first tries to insert the key k into A_0 . If A_0 is empty we insert k into A_0 and are done. If A_0 happens to be already full (i.e. it contains one element), A_0 is *merged* with k to form a new sorted array B_1 of size 2. If A_1 is empty, B_1 becomes the new Array A_1 and we are done. Else B_1 is merged with A_1 into a sorted Array B_2 of size 4 and the same procedure is repeated with A_2, A_3, \dots until we find an Array A_i that is empty.

- (a) Describe a subprocedure $\text{merge}(A, B)$ (as pseudo code or as informal algorithm description) that merges the contents of two sorted Arrays A, B of size m into a new, sorted array of size $2m$ in $\mathcal{O}(m)$ runtime. Explain why your algorithm has the runtime $\mathcal{O}(m)$.
- (b) Show that any series of n insert -operations has an *amortized* runtime of at most $\mathcal{O}(\log n)$.