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# Algorithms and Data Structures Summer Term 2021 Sample Solution Exercise Sheet 2

#### Exercise 1: $\mathcal{O}$ -notation

Prove or disprove the following statements. Use the set definition of the  $\mathcal{O}$ -notation (lecture slides week 2, slides 11 and 12).

- (a)  $4n^3 + 8n^2 + n \in \mathcal{O}(n^3)$
- (b)  $2^n \in o(n!)$
- (c)  $2\log n \in \Omega((\log n)^2)$
- (d)  $\max\{f(n), g(n)\} \in \Theta(f(n) + g(n))$  for non-negative functions f and g.

#### Sample Solution

- (a) True. Choose  $n_0 = 1$  and c = 13. For  $n \ge n_0$  we have  $n^3 \ge n^2 \ge n$  and hence  $4n^3 + 8n^2 + n \le 13n^3 = cn^3$ .
- (b) True. Let c > 0. Choose  $n_0 = \max\{\frac{1}{c}, 8\}$ . For  $n \ge n_0$  we have

$$c \cdot n! \stackrel{n \ge 1/c}{\ge} \frac{1}{n} \cdot n! = (n-1)! \ge (n-1) \cdot (n-2) \cdot \ldots \cdot \lfloor \frac{n}{2} \rfloor \stackrel{n \ge 8}{\ge} 4^{\frac{n}{2}} = 2^n$$

(c) False. Let c > 0. We have

	$2\log n$	$\geq$	$c(\log n)^2$
$\Leftrightarrow$	2	$\geq$	$c\log n$
$\Leftrightarrow$	$\frac{2}{c}$	$\geq$	$\log n$
$\Leftrightarrow$	$4^{\frac{1}{c}}$	$\geq$	n

So for a given  $n_0 \ge 1$  choose  $n = \max\{n_0, 4^{\frac{1}{c}}\}+1$ . For this n we have  $n > n_0$  and  $2\log n < c(\log n)^2$ .

(d) True. Choose  $n_0 = 1$ ,  $c_1 = \frac{1}{2}$  and  $c_2 = 1$ . For  $n \ge n_0$  we have

$$c_1 \cdot (f(n) + g(n)) \le \max\{f(n), g(n)\} \stackrel{f,g \ge 0}{\le} c_2(f(n) + g(n))$$

#### Exercise 2: Sorting by asymptotic growth

Sort the following functions by their asymptotic growth. Write  $g <_{\mathcal{O}} f$  if  $g \in \mathcal{O}(f)$  and  $f \notin \mathcal{O}(g)$ . Write  $g =_{\mathcal{O}} f$  if  $f \in \mathcal{O}(g)$  and  $g \in \mathcal{O}(f)$  (no proof needed).

$\sqrt{n}$	$2^n$	n!	$\log(n^3)$
$3^n$	$n^{100}$	$\log(\sqrt{n})$	$(\log n)^2$
$\log n$	$10^{100}n$	(n+1)!	$n\log n$
$2^{(n^2)}$	$n^n$	$\sqrt{\log n}$	$(2^n)^2$

#### Sample Solution

$$\sqrt{\log n} <_{\mathcal{O}} \log(\sqrt{n}) =_{\mathcal{O}} \log n =_{\mathcal{O}} \log(n^3) <_{\mathcal{O}} (\log n)^2 <_{\mathcal{O}} \sqrt{n} <_{\mathcal{O}} 10^{100} n <_{\mathcal{O}} n \log n$$
$$<_{\mathcal{O}} n^{100} <_{\mathcal{O}} 2^n <_{\mathcal{O}} 3^n <_{\mathcal{O}} (2^n)^2 <_{\mathcal{O}} n! <_{\mathcal{O}} (n+1)! <_{\mathcal{O}} n^n <_{\mathcal{O}} 2^{(n^2)}$$

#### **Exercise 3: Stable Sorting**

A sorting algorithm is called stable if elements with the same key remain in the same order. E.g., assume you want to sort the following strings where the sorting key is *the first letter by alphabetic order*:

["tuv", "adr", "bbc", "tag", "taa", "abc", "sru", "bcb"]

A *stable* sorting algorithm must generate the following output:

["adr", "abc", "bbc", "bcb", "sru", "tuv", "taq", "taa"]

A sorting algorithm is not stable (with respect to the sorting key) if it outputs, e.g., the following:

["abc", "adr", "bbc", "bcb", "sru", "taa", "tag", "tuv"]

- (a) Which sorting algorithms from the lecture (except CountingSort) are *not* stable? Prove your statement by giving an appropriate example.
- (b) Describe a method to make any sorting algorithm stable, without changing the *asymptotic* runtime. Explain.

#### Sample Solution

- (a) Selection Sort is not stable. Consider as input the array [x, y, z] with x.key = y.key = 1 and z.key = 0. In the first step, x and z are swapped, because z has the smallest key in the array. So we get [z, y, x]. This array will not be altered in the second step (as y.key = x.key), i.e., it equals the output of Selection Sort. So x and y have been swapped.
  - Quicksort is not stable. Consider as input the array [x, y, z, w] with x.key = 1, y.key = z.key = 2 and w.key = 0 and assume x is taken as pivot. In the first divide step, y and w are swapped (i.e., we get [x, w, z, y]) and the array is divided into [x, w] and [z, y]. Recursive sorting yields [w, x] and [z, y] and thus [w, x, z, y] will be returned. So y and z have been swapped.
  - Mergesort: If you implement Mergesort according to the pseudocode on page 26 of lecture 01, Mergesort is not stable. The reason is the condition A[i] < A[j] in line 7 of the code which may cause elements with the same key to change order. If we instead use the condition  $A[i] \leq A[j]$ , we make the algorithm stable.
- (b) Add the number *i* to the key of the *i*-th element in the array (i.e., set A[i].key = (A[i].key, i)). Now run the given (non-stable) sorting algorithm according to the lexocographic ordering<sup>1</sup> on this new set of keys. That is, we sort according to the original keys and use the index in A as tie breaker.

Changing the keys takes time O(n). Additionally, each comparison between two elements is prolonged by an additional O(1) steps. As any sorting algorithm takes  $\Omega(n)$ , the asymptotic runtime does not change.

<sup>&</sup>lt;sup>1</sup>Let  $(A, <_A)$  and  $(B, <_B)$  be ordered sets. The lexicographic ordering  $<_{lex}$  on  $A \times B$  is defined by  $(a, b) <_{lex} (a', b') : \Leftrightarrow a <_A a' \lor (a = a' \land b <_B b')$ 

### Exercise 4: Running time

Give an asymptotically tight upper bound for the running time of the following algorithm as a function of n.

 $s \leftarrow 0$ for i = 1 to n do j = 1while j < i do  $s \leftarrow s + i \cdot j$  $j \leftarrow 2 \cdot j$ 

## Sample Solution

For each *i*, the running time of the internal while is proportional to  $\log_2 i$ . Hence, the total running time is proportional to  $\sum_{i=1}^{n} \log_2 i$ . For an upper bound, note that this sum is upper bounded by  $\sum_{i=1}^{n} \log_2 n = n \log_2 n = O(n \log n)$ . In order to show that it is tight, note that the sum is lower bounded by  $\sum_{i=n/2}^{n} \log_2(n/2) = (n/2) \log(n/2) = \Omega(n \log n)$ . Hence the running time is  $\Theta(n \log n)$ .