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Proseminar Algorithmen und Datenstrukturen

Exercise sheet 6

Exercise 1 (O Notation)

Compare the following pairs of functions in terms of order of magnitude. In each case, say whether $f(n) \in O(g(n)), f(n) \in \Omega(g(n))$ and/or $f(n) \in \Theta(g(n))$.

	f(n)	g(n)
a)	$n^3 + 3n + 1$	n^4
b)	$\log(n)$	n
c)	$n \cdot 2^n$	3^n
d)	n	$(\log(n))^5$
e)	$\log(n)$	$\log(n^2)$
f)	$100n + \log(n)$	$n + (\log(n))^2$

In order to show that $f(n) \in O(g(n))$ we have to find two constants c and N such that for all n > N we have that $f(n) \leq c \cdot g(n)$. Formally,

$$f(n) \in O(g(n)) \iff \exists c > 0 \ \exists N \ \forall n > N \quad f(n) \le c \cdot g(n) \tag{1}$$

Moreover, $f(n) \in \Omega(g(n))$ if and only if $g(n) \in O(f(n))$, and $f(n) \in \Theta(g(n))$ if and only if $f(n) \in O(g(n))$ and $g(n) \in O(f(n))$.

a) n^3+3n+1 is in $O(n^4)$. Setting c = 1 and N = 1, we have to show that $n^3+3n+1 \le n^4$ for all $n \ge 2$. A simple proof by induction on n will suffice.

 $\boxed{n=2} \ 2^3 + 3 * 2 + 1 = 15 \le 16 = 2^4$

$$n^{4} + 3n^{2} + 3n + 4 \leq (n+1)^{4}$$

$$n^{4} + 3n^{2} + 3n + 4 \leq n^{4} + 4n^{3} + 6n^{2} + 4n + 1$$

$$3 \leq 4n^{3} + 3n^{2} + n$$

And this last inequality is trivially satisfied for n > 2.

Alternatively, we can use the following reasoning to establish that $n^3 + 3n + 1$ is in $O(n^4)$.

Note that this proof also yields that $n^3 + 3n + 1$ is in $O(n^3)$. Finally, for the ones who are familiar with computing with limits, $\lim_{n\to\infty} \frac{f(n)}{g(n)} < \infty$ (i.e., the limit exists) implies condition (1). Hence,

$$\lim_{n \to \infty} \frac{n^3 + 3n + 1}{n^4} = \lim_{n \to \infty} \frac{1}{n} + 3\frac{1}{n^3} + \frac{1}{n^4} = 0$$

also tells us that $n^3 + 3n + 1$ is in $O(n^4)$.

Clearly, n^4 is not in $O(n^3+3n+1)$. The proof goes by contradiction. So assume that there exist constants c and N such that for all n > N we have $n^4 \le c(n^3+3n+1)$. This means that

$$c \ge \frac{n^4}{n^3 + 3n + 1} \ge \frac{n^4}{3n^3 + 3n + 3} \ge \frac{n^4}{9n^3} = \frac{n}{9}$$

Hence, $c \geq \frac{n}{9}$ for all n > N. Obviously such a constant c does not exist.

As a consequence of the results obtained above, $n^3 + 3n + 1$ is not in $\Theta(n^4)$.

b) For n > 0 we have $\log(n) \le n$ if and only if $e^{\log(n)} \le e^n$, which simplifies to $n \le e^n$ (remember that the exponential function is monotonically increasing). Establishing $n \le e^n$ for n > 0 is a trivial task if we consider the power series representation of the exponential function:

$$e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!} = 1 + x + \frac{x^2}{2!} + \dots$$

Hence, $\log(n)$ is in O(n). Reasoning with limits we obtain the same result somewhat faster

$$\lim_{n \to \infty} \frac{\log(n)}{n} = \lim_{n \to \infty} \frac{1/n}{1} = \lim_{n \to \infty} \frac{1}{n} = 0$$

by an application of de L'Hopital's rule for calculating limits.

However, n is not in $O(\log(n))$. Again the proof goes by contradiction. So assume that there exist constants c and N such that for all n > N we have $n \le c \cdot \log(n)$. This means that $c \ge n/\log(n)$, which implies that

$$e^{c} \ge e^{\frac{n}{\log(n)}} = e^{\frac{n\log(n)}{(\log(n))^{2}}} = n^{\frac{n}{(\log(n))^{2}}} \ge n^{\frac{n}{(\log(n))^{2}}}$$

(note that $n \ge (\log(n))^2$ for $n \ge 1$, a fact which we will prove later). Hence, $e^c \ge n$ for all n > N. Obviously such a constant c does not exist.

As a consequence of the results obtained above, $\log(n)$ is not in $\Theta(n)$.

c)

$$\lim_{n \to \infty} \frac{n \cdot 2^n}{3^n} = \lim_{n \to \infty} \frac{n}{(\frac{3}{2})^n} = \lim_{n \to \infty} \frac{1}{(\frac{3}{2})^n \cdot \log(\frac{3}{2})} = 0$$

Hence, $n \cdot 2^n$ is in $O(3^n)$.

However, 3^n is not in $O(n \cdot 2^n)$ as a simple proof by contradiction shows.

As a consequence of the results obtained above, $n \cdot 2^n$ is not in $\Theta(3^n)$.

d) $\log(n)^5$ is in O(n). It is even true that $(\log(n))^r$ is in O(n) for every natural number $r \ge 1$.

Proof. So we have to find constants c > 0 and N such that $(\log(n))^r \le c \cdot n$ for all n > N. The logarithm of n is non-negative for $n \ge 1$ and therefore $(\log(n))^r \le c \cdot n$ is equivalent to $\log(n) \le \sqrt[r]{c \cdot n}$. Now we conclude

$$\log(n) \le \sqrt[r]{c \cdot n} \Longleftrightarrow e^{\log(n)} \le e^{\sqrt[r]{c \cdot n}} \Longleftrightarrow n \le e^{\sqrt[r]{c \cdot n}} \Longleftrightarrow \frac{e^{\sqrt[r]{c \cdot n}}}{n} \ge 1$$

Next we use the power series representation of the exponential function to simplify $\frac{e^{\sqrt[n]{C \cdot n}}}{n}$.

$$\frac{e^{\sqrt[r]{c \cdot n}}}{n} = \frac{\sum_{k=0}^{\infty} \frac{(\sqrt[r]{c \cdot n})^k}{k!}}{n} = \frac{\sum_{k=0}^{\infty} \frac{(c \cdot n)^{\frac{k}{r}}}{k!}}{n} = \sum_{k=0}^{\infty} \frac{e^{\frac{k}{r}} n^{\frac{k}{r}-1}}{k!}$$

Remember that we want to find constants c > 0 and N such that this sum becomes 1 or greater for all n > N. Analyzing the structure of the sum, we see that all summands are non-negative. Hence, if only one of the summands is equal to 1 or greater then the whole sum is, too. Moreover, considering the summand for k = r

$$\frac{c^{\frac{r}{r}}n^{\frac{r}{r}-1}}{r!} = \frac{cn^0}{r!} = \frac{c}{r!}$$

we see that it is independent of n (which is very good!). So, if we set c to the faculty of r then the entire sum becomes 1 or greater, and $(\log(n))^r \leq r! \cdot n$ for all $n \geq 1$.

However, n is not in $O((\log(n))^r)$ as a proof by contradiction similar to the one used in b) shows.

As a consequence of the results obtained above, n is not in $\Theta((\log(n))^r)$.

e) Note that $\log(n^2) = 2 \cdot \log(n)$. Hence, $\log(n)$ is in $O(\log(n^2))$ and $\log(n^2)$ in $O(\log(n))$, which implies that $\log(n)$ is in $\Theta(\log(n^2))$.

$$\lim_{n \to \infty} \frac{100n + \log(n)}{n + (\log(n))^2} = \lim_{n \to \infty} \frac{100 + \frac{1}{n}}{1 + 2(\log(n))\frac{1}{n}} = \lim_{n \to \infty} \frac{100n + 1}{n + 2(\log(n))} = \lim_{n \to \infty} \frac{100}{1 + 2\frac{1}{n}} = 100$$

Hence, $100n + \log(n)$ is in $O(n + (\log(n))^2)$. The same calculation (with nominator and denominator exchanged) tells us that $n + (\log(n))^2$ is in $O(100n + \log(n))$. Together this implies that $100n + \log(n)$ is in $\Theta(n + (\log(n))^2)$.

Exercise 2 (O Notation)

The addition theorem for O states that $f(n) \in O(s(n))$ and $g(n) \in O(r(n))$ imply that $f(n) + g(n) \in O(s(n) + r(n))$. Now formulate the corresponding theorem for Ω and provide a proof for it.

Claim: $f(n) \in \Omega(s(n))$ and $g(n) \in \Omega(r(n))$ imply that $f(n) + g(n) \in \Omega(s(n) + r(n))$.

Proof. Let's assume that $f(n) \in \Omega(s(n))$ and $g(n) \in \Omega(r(n))$. By definition of Ω , we have

$$\begin{aligned} \exists c_1 > 0 \ \exists N_1 \ \forall n > N_1 \quad f(n) \geq c_1 \cdot s(n) \\ \exists c_2 > 0 \ \exists N_2 \ \forall n > N_2 \quad g(n) \geq c_2 \cdot r(n) \end{aligned}$$

Hence, $f(n) + g(n) \ge c_1 \cdot s(n) + c_2 \cdot r(n) \ge \min(c_1, c_2) \cdot (s(n) + r(n))$. Now we can prove that $\exists c > 0 \exists N \forall n > N$ $f(n) + g(n) \ge c \cdot (s(n) + r(n))$ by setting N to $\max(N_1, N_2)$ and c to $\min(c_1, c_2)$.

Exercise 3 (O Notation)

- a) Find a counterexample to the following claim: $f(n) \in O(s(n))$ and $g(n) \in O(r(n))$ imply that $f(n) g(n) \in O(s(n) r(n))$.
- b) Find a counterexample to the following claim: $f(n) \in O(s(n))$ and $g(n) \in O(r(n))$ imply that $f(n)/g(n) \in O(s(n)/r(n))$.

- a) Let $f(n) = 2n \in O(n)$ and $g(n) = n \in O(n)$.
- b) Let $f(n) = n \in O(n)$ and $g(n) = 1/n \in O(1)$.

Exercise 4 (O Notation)

Prove or disprove (by means of a counterexample) the following statements:

- a) $O(f(n) + g(n)) = O(\max(f(n), g(n)))$
- b) $O(f(n) + g(n)) = O(\min(f(n), g(n)))$
- c) $\Omega(f(n) + g(n)) = \Omega(\max(f(n), g(n)))$
- d) $\Omega(f(n) + g(n)) = \Omega(\min(f(n), g(n)))$
- e) $\Theta(f(n) + g(n)) = \Theta(\max(f(n), g(n)))$
- f) $\Theta(f(n) + g(n)) = \Theta(\min(f(n), g(n)))$

First, we prove a simple theorem concerning O notation. For every constant k > 0and $f : \mathbb{N}_0 \to \mathbb{R}_0^+$, we have

$$O(k \cdot f(n)) = O(f(n))$$

Proof. For all functions $g: \mathbb{N}_0 \to \mathbb{R}_0^+$ we have

$$g \in O(k \cdot f(n)) \iff \exists c > 0 \ \exists N \ \forall n > N \quad g(n) \le c \cdot (k \cdot f(n))$$
$$\iff \exists c > 0 \ \exists N \ \forall n > N \quad g(n) \le (c \cdot k) \cdot f(n)$$
$$\iff \exists c' > 0 \ \exists N \ \forall n > N \quad g(n) \le c' \cdot f(n)$$
$$g \in O(f(n))$$

a) $O(f(n) + g(n)) = O(\max(f(n), g(n)))$ is true. If f and g are functions from \mathbb{N}_0 to \mathbb{R}_0^+ then the following inequalities hold for all n:

$$\max(f(n), g(n)) \le f(n) + g(n) \le 2 \cdot \max(f(n), g(n))$$

From the first inequality we conclude that $O(\max(f(n), g(n))) \subseteq O(f(n) + g(n))$, since any function that is bounded by $\max(f, g)$ is also bounded by f + g.

Likewise, we conclude from the second inequality that $O(f(n) + g(n)) \subseteq O(2 \cdot \max(f(n), g(n)))$. But by the above theorem $O(2 \cdot \max(f(n), g(n))) = O(\max(f(n), g(n)))$, and hence $O(f(n) + g(n)) \subseteq O(\max(f(n), g(n)))$.

b) $O(f(n) + g(n)) = O(\min(f(n), g(n)))$ is false. For example, let f(n) = n and $g(n) = n^2$. Then $O(f(n) + g(n)) = O(n + n^2)$ which is equal to $O(n^2)$ by a), whereas $O(\min(f(n), g(n))) = O(n)$.

c) $\Omega(f(n) + g(n)) = \Omega(\max(f(n), g(n)))$ is true by similar reasoning as in a). If f and g are functions from \mathbb{N}_0 to \mathbb{R}_0^+ then the following inequalities hold for all n:

 $\max(f(n), g(n)) \le f(n) + g(n) \le 2 \cdot \max(f(n), g(n))$

From the first inequality we conclude that $\Omega(f(n) + g(n)) \subseteq \Omega(\max(f(n), g(n)))$, since any function that is bounded from below by f+g is also bounded by $\max(f, g)$.

Likewise, we conclude from the second inequality that $\Omega(2 \cdot \max(f(n), g(n))) \subseteq \Omega(f(n) + g(n))$. But $\Omega(2 \cdot \max(f(n), g(n))) = \Omega(\max(f(n), g(n)))$, and hence $\Omega(\max(f(n), g(n))) \subseteq \Omega(f(n) + g(n))$.

- d) $\Omega(f(n) + g(n)) = \Omega(\min(f(n), g(n)))$ is false by similar reasoning as in b).
- e) $\Theta(f(n) + g(n)) = \Theta(\max(f(n), g(n)))$ is true. By definition of Θ , we have $h \in \Theta(f(n) + g(n))$ if and only if $h \in O(f(n) + g(n))$ and $h \in \Omega(f(n) + g(n))$. According to a) and c) this is equivalent to $h \in O(\max(f(n), g(n)))$ and $h \in \Omega(\max(f(n), g(n)))$, which is equivalent to $h \in \Theta(\max(f(n), g(n)))$.
- f) $\Theta(f(n) + g(n)) = \Theta(\min(f(n), g(n)))$ is false by similar reasoning as in b).

Exercise 5 (O Notation)

Assuming that the bodies of the for-loops only contain simple statements (no loops, no function calls etc.), and hence have constant time complexity, what is the time complexity of each of the following code fragments?

c) for (i = 0; i < N; i++) { for (j = i; j < N; j++) { ... }

d) Consider this sample implementation for the longest upsequence algorithm (exercise 4 on exercise sheet 3). What is its complexity expressed in O notation? Consider both cases, binary and naive linear search.

```
#include <stdio.h>
```

```
int main() {
  int n = 8; /* array length */
  int X[] = \{1, 3, 4, 6, 2, 4, 4, 0\};
  int M[n]; for (int i=0; i<n; i++) M[i]=0;
  int i;
  int k = 0;
 M[0] = X[0];
  for (i = 1; i < n; i++){
    if (X[i] >= M[k]) \{
      k++;
      M[k] = X[i];
      }
    else if (X[i] < M[0]) {
      M[0] = X[i];
      }
    else {
      /* find j such that M[j-1] <= X[i] < M[j] */
      /* Possibility 1: naive search
      int \quad j = i;
      while (M/j - 1) > X/i)
        j - -; * /
      /* Possibility 2: binary search */
      int b, j = 1;
      for (b = (k+1)/2; M[j-1] > X[i] || X[i] >= M[j]; b /= 2)
        if (M[j-1] > X[i])
          j \rightarrow b;
        else
          j += b;
      M[j] = X[i];
    }
  }
```

printf("Longest_upsequence_has_length_%d\n", k + 1); return 0;

- a) The first loop is O(N) and the second loop is O(M). Since we don't know which one is bigger, we say this is O(N+M), or equivalently $O(\max(N, M))$. In the case where the second loop goes to N instead of M the complexity is O(N). You can see this from either expression above. O(N+M) becomes O(2N), and when you drop the constant it is O(N). $O(\max(N, M))$ becomes $O(\max(N, N))$ which is O(N).
- b) The first set of nested loops is $O(N^2)$ and the second loop is O(N). Hence, the overall complexity is $O(\max(N^2, N))$, which is $O(N^2)$.
- c) When *i* is 0, the inner loop executes *N* times. When *i* is 1, the inner loop executes N-1 times. In the last iteration of the outer loop, when *i* is N-1, the inner loop executes exactly once. Hence, the number of times the body of the inner loop executes is $N + (N-1) + ... + 2 + 1 = \sum_{i=1}^{N} i = N(N+1)/2$, which amounts to $O(N^2)$.
- d) The for-loop executes n-1 times, where n is the length of the input sequence. Most of the body of the for-loop is made up of basic statements that execute in constant time (i.e., the execution time does not depend on n), only the part where we search for a position to place the current element depends on n. Hence, (in the worst case) the complexity of the body of the for-loop is the complexity of the search algorithm we use.

In case of linear search, the iteration count of the while-loop is determined by the variable j which is initialized to the value of i, the loop index of the main loop. In the worst case j is n - 1. Consequently, the complexity of the search is O(n).

In case of binary search, the iteration count of the for-loop is determined by the variable b, which is initialized to (k + 1)/2. In the worst case b is n/2. Hence, the complexity of the search is $O(\log n)$.

Thus the total complexity is $O(n^2)$ using linear search and $O(n \cdot \log n)$ in case of binary search.

Exercise 6 (Linked List)

}

Reconsider your linked list implementation of last week, try to eliminate the bugs it has (if it has any) and provide a demo program that is equipped with a user interface. The user interface should consist of a menu that has entries for

- adding elements to a list,
- removing elements from a list,
- printing a list,

- inserting elements in order and
- searching for a specific element.

The main program main.c:

```
#include <stdio.h>
#include <string.h>
#include "list.h"
void printMenu()
{
  fprintf(stdout, "*\_[a]\_Add\_an\_element\_to\_the\_list\_\_\_=*\n");
  fprintf(stdout, "*__[r]_Remove_an_element_from_the_list_*\n");
  fprintf(stdout, "*__[p]_Print_the_list_____*\n");
  fprintf(stdout, "*\_[s]]Search_for_an_element____*\n");
  fprintf(stdout, "*\_\_[q]\_Quit\_the\_program\_\_\_\_\_\_====*\setminusn");
  fprintf(stdout, "\n");
}
char chooseCmd()
{
  char c = ' \setminus 0';
  char* s = NULL;
  do
  {
    fprintf(stdout, "Make_your_selection:_");
   c = getchar();
   //flush stdin: note fflush has undefined behaviour for input streams
   while (getchar() != ' n');
  } while (!(s = strchr("arpsq", c)));
  fprintf(stdout, "\n");
  return *s;
}
//read an integer from stdin, using the string s as prompt
int readInt(const char *s)
{
  int i = 0, ret = 0;
  do
  ł
   if (s)
```

}

{

```
printf("\%s", s);
    ret = scanf("%d", \&i);
    //flush stdin: note fflush has undefined behaviour for input streams
    while (getchar() != ' n');
  } while (ret != 1);
 return i;
int main()
  list_t * list = NULL; //empty list
  int stop = 0;;
  while (!stop)
  {
    printMenu();
    switch ((unsigned int) chooseCmd())
    {
      case 'a':
        list = addElement(list,
               readInt("Enter_an_element_to_add_to_the_list:_")
        );
        break;
      case 'r':
        list = removeElement(list,
               readInt("Enter_an_element_to_remove_from_the_list:_")
        );
        break;
      case 'p':
        printList(list);
        break;
      case 's':
        {
          int res = searchList(list,
               readInt("Enter_an_element_to_search_for_in_the_list:_")
          );
          printf("\nThis\_element\_is\%s\_present\_in\_the\_list\n", res ? "" : "\_not"
          break;
        }
      case 'q':
        stop = 1;
        break;
      default:
```

```
break;
    }
  }
  //free the memory occupied by the list nodes
  deleteList(list);
  return 0;
}
  The list interface file list.h:
#ifndef LIST_H
#define LIST_H
typedef struct list
{
  int data;
  struct list * next;
  list_t;
list_t * addElement(list_t * list, int x);
void printList(const list_t * list);
void deleteList(list_t * list);
int searchList(const list_t * list, int x);
list_t * removeHead(list_t * list);
list_t * removeElement(list_t * const list, int x);
#endif
  The list implementation file list.c:
#include <stdio.h>
#include <stdlib.h>
#include <assert.h>
#include "list.h"
list_t * addElement(list_t * list, int x)
{
  list_t * p = (list_t *) malloc(sizeof(list_t));
  if (p)
  {
    p \rightarrow data = x;
```

```
p \rightarrow next = list;
  }
  return p;
}
void printList(const list_t * list)
{
  fprintf(stdout, "[");
  while (list)
  {
    fprintf(stdout, "%d", list ->data);
    if (list ->next)
       fprintf(stdout, ";");
    list = list \rightarrow next;
  }
  fprintf(stdout, "]\n");
}
void deleteList(list_t * list)
{
  while (list)
  {
    list_t * p = list -> next;
    free(list);
    list = p;
  }
}
int searchList(const list_t * list, int x)
ł
  while (list)
  {
    if (list \rightarrow data = x)
       return 1;
    list = list \rightarrow next;
  }
  return 0;
}
list_t * removeHead(list_t * list)
ł
  list_t * p = NULL;
  if (list)
```

```
{
    p = list \rightarrow next;
     free(list);
  }
  return p;
}
list_t * removeElement(list_t * const list , int x)
{
  if (!list)
    return NULL;
  if (list \rightarrow data == x)
    return removeHead(list);
  list_t * prev = list;
  list_t * p = list ->next;
  while (p)
  {
     if (p \rightarrow data = x)
     {
       assert (prev != NULL);
       prev \rightarrow next = p \rightarrow next;
       free(p);
       return list;
     }
     else
     {
       prev = p;
       p = p - next;
     }
  }
  return list;
}
```