Algorithms and Data Structures Conditional Course

Lecture 3

Abstract Data Types, Simple Data Structures, Binary Search

FREIBUR

Fabian Kuhn Algorithms and Complexity

Data Structures

Algorithms

- How to solve a given problem efficiently?
- Goal: smallest possible complexity
 - small runtime / small memory usage
 - asymptotically, dependent on the problem size

Data Structures

- How to save data such that the access becomes as efficient as possible?
- Depends on the types of operations we have to support!
- Good data structures necessary to obtain fast algorithms
- One needs fast algorithms to carry out data structure operations optimally

Abstract Data Type:

- Specification which kind of data one can support
- Specification of the operations to access the data
 - including the semantics of the operations

Data Structure:

- A concrete way of implementing an abstract data type
- Depending on the implementation, the same operations might have different runtimes (complexities)

We will now first briefly discuss the most important abstract data types...

Array:

holds a collection of elements (of the same type)

Operations:

- create(n) : creates an array of length n
- *A.get(i)* : returns the element at position *i*
- A.set(x, i) : writes element x to position i
- *A.size()* : returns the length of the array (not always available)

For dynamic arrays (can change size):

- A.append(x) : appends element x at the end
- *A.deleteLast()* : deletes last element

Abstract Data Types: Examples

Dictionary: (also: maps, associative arrays)

 holds a collection of elements where each element is represented by a unique key

Operations:

- *create* : creates an empty dictionary
- *D.insert(key, value)* : inserts a new (key, value)-pair
 - If there already is an entry with the same key, the old entry is replaced
- *D.find(key)* : returns entry with key *key*
 - If there is such an entry (returns some default value otherwise)
- *D.delete(key)* : deletes entry with key *key*

Dictionary:

Additional possible operations:

- *D.minimum()* : returns smallest *key* in the data structure
- D.maximum() : returns largest key in the data structure
- *D.successor(key)* : returns next larger *key*
- *D.predecessor(key)* : returns next smaller *key*
- D.getRange(k1, k2) : returns all entries with keys in the interval [k1,k2]

Queue:

• Holds a collection (sequence) of elements

Operations:

- create : creates an empty queue
- *Q.enqueue(x)* : appends element *x* at the end
- Q.dequeue() : returns element at front element and removes it
- *Q.isEmpty()* : Is the queue empty?

Is also called FIFO queue (FIFO = first in first out)



Abstract Data Types: Examples

Stack:

• Holds a collection (sequence) of elements

Operations:

- create : creates an empty stack
- *S.push(x)* : puts an element *x* on the stack
- S.pop() : returns and deletes top element of stack
- *S.isEmpty()* : Is the stack empty?

Is also called LIFO queue (LIFO = last in first out)

Algorithms and Data Structures





Heap / Priority Queue :

• Holds a collection of *(key,value)* pairs

Operations:

- create() : creates an empty heap
- *H.insert(x, key)* : inserts element *x* with key *key*
- H.getMin() : returns element with smallest key
- *H.deleteMin()* : deletes element with smallest key
 - *H.getMin()* and *H.deleteMin()* have to be consistent
- *H.decreaseKey(x, newkey)* : If *newkey* is smaller than the current key of *x*, the key of x is set to *newkey*

Union-Find / Disjoint Sets:

• Manages a partition of elements

Operationen:

- *create()* : creates an empty union-find data structure
 - *U.makeSet(x)* : adds a set $\{x\}$ to the partition
- *U.find(x)* : returns the set containing element *x*
- U.union(S1, S2)
- : unites sets S1 and S2 to set $S1\cup S2$



Array Implementation of Stack

Let us try to implement the stack data type

- **Operations:** *create, push, pop, isEmpty*
- Assumption: Stack only needs to be able to hold NMAX elements

Variables to store the state of the stack:

- *stack* : array of length *NMAX*
- *size* : current number of elements in stack

```
create():
  stack = new array of length NMAX
  size = 0
```



Analysis: Array Implementation of Stack

Runtime (complexity) of the operations:

- create: O(1)
 - If we assume that memory can be allocated in O(1) time
- push: 0(1)
- pop: *O*(1)
- isEmpty: *O*(1)

Disadvantages of the implementation:

- Memory usage (space complexity) : $\Theta(NMAX)$
 - always needs the same amount of memory, no matter how many elements there are on the stack!
- The stack can only hold NMAX elements...
- We will see that both these things can be fixed...

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• Reversing a sequence: A, B, C

push(A), push(B), push(C), pop() \rightarrow C, pop() \rightarrow B, pop() \rightarrow A

- Undo operation for editors:
 - Put description of (reversible) edit operations on the stack
- Program stack for function / method calls
 - Remark: With a stack, it is possible to write down recursion explicitly

	Arguments and local variables of function f_2	def $f_1(x, y)$:	def $f_2(a)$:
	Arguments and local variables of function f_2	<i>f</i> ₂ (<i>z</i>) 	<i>f</i> ₂ (<i>b</i>)
	Arguments and local variables of function f_1		

Array Implementation of Queue

Let us try to implement the queue data type

- **Operations:** *create, enqueue, dequeue, isEmpty*
- Assumption: Queue only needs to be able to hold NMAX elements

Variables to store the state of the queue:

- *queue* : array of length *NMAX*
- head : position of the first element (cyclic)
 - if the queue is not empty.
- *size* : number of elements in the queue

create:

```
queue = new array of length NMAX
head = 0
size = 0
```

Z



- Q.dequeue() returns element at pos. head, if Q is not empty
- Q.enqueue(x) inserts element at pos. head + size



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Array Implementation of Queue

```
S.isEmpty():
  return (size == 0)
S.enqueue(x):
  if (size < NMAX)
    pos = (head + size) mod NMAX
    queue[pos] = x
    size += 1
S.dequeue():
  if (size == 0)
    report error (or return default value)
  else
    x = queue[head]
    head = (head + 1) \mod NMAX
    size = size - 1
    return x
```

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Analysis: Array Implementation of Queue

Runtime (complexity) of the operations:

- create: O(1)
 - If we assume that memory can be allocated in O(1) time
- enqueue : O(1)
- dequeue : *O*(1)
- is Empty : O(1)

Disadvantages of the implementation:

- Memory usage (space complexity) : $\Theta(NMAX)$
 - always needs the same amount of memory, no matter how many elements there are in the queue!
- The queue can only hold up to NMAX elements...
- We will see that both these things can be fixed ...

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Linked Lists

• Data structure to hold a list (sequence) of elements



Linked list:



Doubly linked list:



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List Elements





 Class to describe list elements



Python:

class ListElement:

```
def __init__(self, key=0, data=None, next=None, prev=None):
    self.key = key
    self.data = data
    self.next = next
    self.prev = prev
```

List Elements

Class to describe

list elements



Java:

•

```
public class ListElement {
```

```
int/String/... key;
Object/... data;
```

```
ListElement next;
ListElement prev;
```

C++:

class ListElement {
public/private:

int/... key; void*/... data;

ListElement* next; ListElement* prev;

}

}

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Singly Linked List:



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Stack and FIFO Queue with Lists

With singly linked lists, all operations in time $oldsymbol{O}(1)$



• Elements can be added (push) und deleted (pop) at front



- enqueue: add element at end (tail) of list
- dequeue: delete element at front (head) of list

N N N N N

Singly Linked List:



Goal: Find element with key *x*

```
current = first
while current != None and current.key != x:
    current = current.next
```

Runtime: List of length n : O(n)

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Insertion in Singly Linked Lists



y.next = x.next
x.next = y

Attention: Take care of special cases at beginning and end of list!



$$x.prev = y$$

Attention: Take care of special cases at beginning and end of list!

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Delete element x:



Assumption: Predecessor element *y* is given

y.next = x.next

 In C++ one would also need to free the memory used by element x, in Python / Java, this is done by the garbage collector

Attention: Take care of special cases at beginning and end of list!

Deletion in Doubly Linked Lists



x.prev.next = x.next
x.next.prev = x.prev

Attention: Take care of special cases at beginning and end of list!

Assumption: List of length n

Search for element with key x: O(n)

Insertion of an element: O(1)

• if reference to predecessor is given, otherwise O(n)

Deletion of an element: O(1)

• if ref. to predecessor (singly linked lists) or to element itself (doubly linked lists) is given, otherwise O(n)

Concatenation of two lists: O(1)

if last pointer to first list is given

Stack and Queue with linked lists:

- all operations in time O(1)
- Size not restricted, memory usage O(n)

Sentinel:



• A dummy element that form the start and end of the list

- list is accessed through *nil.next* instead of *first*
- replaces null pointer at the end of list
- empty list: sentinel points to itself (*nil.next = nil*)
- sentinel is just a part of the implementation and should not be visible from outside

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Sentinel for doubly linked lists:



- list is accessed through *nil.next, nil.prev* instead of *first, last*
- replaces null pointers at start and end of list
- results in a cyclic doubly linked list
- empty list: *nil.next = nil , nil.prev = nil*

Advantages:

- Avoids special cases at start / end of list when inserting / deleting
- Code becomes simpler and possibly also faster
- Null pointer exceptions are avoided ...
 - Not clear to what extent this improves robustness ...

Disadvantages:

- In case of many small lists, the additional memory useage for the sentinels might become relevant (never asymptotically)
- Sentinels make most sense if they really simplify the code

Abstract Data Types: Dictionary

Dictionary: (also: maps, associative arrays, symbol tables)

holds a collection of elements where each element is represented by a unique key

Operations:

- create : creates an empty dictionary
- *D.insert(key, value)* : inserts a new (key, value)-pair
 - If there already is an entry with the same key, the old entry is replaced
- *D.find(key)* : returns entry with key *key*
 - If there is such an entry (returns some default value otherwise)
- *D.delete(key)* : deletes entry with key *key*

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Dictionary

 In a first phase, we deal with implementing the basic operations *insert*, *find*, *delete* (und *create*)

Dictionary Examples:

- Dictionary (key: wort, value: definition / translation)
- Phone Book (key: name, value: phone number)
- DNS Server (key: URL, value: IP address)
- Python interpreter (key: variable name, value: value of variable)
 Java/C++ compiler (key: variable name, value: type information)

In all these cases, it is particularly important to have a fast *find* operation!

Operations:

- create:
 - creates an empty list
- D.insert(key, value):
 - inserts element at front
 - Assumption: There is no previous entry with key key
- D.find(key):
 - traverse list sequentially
- D.delete(key):
 - first search the list element with key key (as in find)
 - delete element from the list
 - For singly linked list, one has to stop as soon as current.next.key == key !

Runtimes:

create: O(1)

insert: O(1)

• If one does not need to check if key is already present

find: O(n)

• We potentially have to iterate over the whole list

delete: O(n)

• We potentially have to iterate over the whole list

Is this good?

• In particular *find* is very expensive!

Operations:

- create:
 - allocates a new array of size NMAX
- D.insert(key, value):
 - inserts new element at end (if there still is space)
 - Assumption: There is no previous entry with key key
- D.find(key):
 - Iterate over all the elements starting at front (or end)
- D.delete(key):
 - first, search for key
 - delete element from array, then:

Move all elements after the deleted element one position to the left!

Runtimes:

create: O(1)

insert: O(1)

find: O(n)

• We potentially need to iterate over the whole array

delete: O(n)

• We potentially have to iterate over the whole array and might need to copy $\Omega(n)$ elements

Better ideas?

• In particular *find* is still very expensive!

Use a Sorted Array?

- Expensive operation for list / array, in particular *find*
- If (as soon as) the entries do not change too much, *find* becomes the most important operation!
- Can we search for a given key faster if the entries in an array are sorted by their keys?
 - Example: Search phone number of a person in a phone book...

Ideas for searching x:

• We open the phone book approximately in the middle and check if the name we look for in before or after that position.



Binary Search

Use the divide-and-conquer idea!

Search for the number (the key) 19:

LN I



Algorithm (array A of length n, search for key x):

Manage left and right boundary l und r, s. t. (if x is contained in A)

 $A[l] \le x \le A[r]$

- At the beginning, we set l = 0 and r = n 1
- Go to the midele m = (l + r)/2
 - If $A[m] = x \implies x$ found!
 - If $A[m] < x \implies x$ is in right part $\implies l = m + 1$
 - $\text{ If } A[m] > x \implies x \text{ is in left part } \implies l = m 1$



Algorithm (array A of length n, search for key x):

Is the algorithm correct?

How can we verify this?

- Empirically: unit tests or more systematic tests...
- Formally?
 - Correctness is (usually) even more important than performance!

Hoare Logic

• We only look at the basic ideas

Precondition

Condition that holds at the beginning (of a method / loop / ...)

Postcondition

Condition that holds at the end (of a method / loop / ...)

Loop invariant

Condition that holds at beginning and/or end of each loop iteration

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```
l = 0; r = n - 1;
while r > l do
m = (l + r) / 2;
if A[m] < x then l = m + 1
else if A[m] > x then r = m - 1
else l = m; r = m
```

Precondition

• array is sorted, array is of length n

Postcondition

• If x is contained in array, then A[l] = x

Loop invariant

• If x is contained in array, then $A[l] \le x \le A[r]$

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Is the algorithm correct?

Precondition

• array is sorted, array is of length n

$$l = 0; r = n - 1;$$
 loop invariant

Loop invariant

- If x is contained in array, then $A[l] \le x \le A[r]$
- Precondition and assignment for l and $r \rightarrow loop$ invariant
 - Loop invariant holds at beginning of first loop iteration

Postcondition

- If x is contained in array, hen A[l] = x
- Termination condition of while loop $\rightarrow l \ge r$ and thus $A[l] \ge A[r]$
- If x is contained in array, then the loop invariant and the fact that A is sorted imply that A[l] = A[r] and thus A[l] = x

```
l = 0; r = n - 1;
while r > l do
m = (l + r) / 2;
if A[m] < x then l = m + 1
else if A[m] > x then r = m - 1
else l = m; r = m
```

Schleifeninvariante

- If x is contained in array, then $A[l] \le x \le A[r]$
 - $-\,$ The loop invariant holds at the beginning of the loop, it can only be invalidated if we change the variables l and r
 - If we set l = m + 1, we know that A[m] < x; therefore, we afterwards have $A[m + 1] \le x$ if x is contained in A.
 - Analogously, if we set r = m 1, we know that A[m] > x; therefore, we afterwards have $x \le A[m-1]$ if x is contained in A.

Does the algorithm terminate?

• Change of number of elements (r - l + 1) per iteration?

$$\begin{aligned} -l &= m+1: \\ r - (m+1) + 1 \leq r - \left(\frac{l+r}{2} + \frac{1}{2}\right) + 1 = \frac{r-l+1}{2} \\ -r &= m-1: \\ (m-1) - l + 1 \leq \frac{l+r}{2} - 1 - l + 1 = \frac{r-l}{2} < \frac{r-l+1}{2} \end{aligned}$$

- Otherwise x is found and r - l + 1 becomes 1

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Runtime

Does the algorithm terminate?

- INI
- The number of active elements is at least halved in each iteration
- The algorithm terminates!

Runtime?

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$$T(n) \leq T(\lfloor n/2 \rfloor) + c, \quad T(1) \leq c$$

$$T(n) \in T(\lceil n/2 \rceil) + c$$

$$\leq T(\lceil n/2 \rceil) + c + c$$

$$quest$$

$$\leq T(1) + c \cdot \log_2 n \leq c(\log_2 n + 1)$$

Runtime Binary Search

The algorithm terminates in time $O(\log n)$. $T(n) \leq T(\frac{1}{2}) + c$, $T(n) \leq c$ $\frac{1}{2}$ $\frac{1}$ base: N=1 $T(1) \leq C(O+1) = C$ $s_{ep}^{1} u_{7}^{1} T(u) \leq T(u_{2}^{2}) + C$ $\leq c\left(\log_2 \frac{N}{2} + 1\right) + c$ login = C(los, n + 1). Zä

Operations:

- create:
 - allocates new array of size NMAX
- D.find(key):
 - search for key by using binary search
- D.insert(key, value):
 - Search for key and insert element at the right position
 - Insertion: All elements after the insertion have to move one to the right!
- D.delete(key):
 - First search for *key* and remove the respective element
 - Deletion: All elements after the deletion have to move one to the left!

Runtimes:

- create: O(1)
- insert: O(n)
- find: $O(\log n)$
- delete: O(n)

Can we make all operations fast?

• and *find* even faster?

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