

# DATA STRUCTURES AND ALGORITHMS

## LECTURE 2

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# In Lecture 1...

- Course Organization
- Abstract Data Types and Data Structures
- Pseudocode
- Algorithm Analysis
  - $O$  - notation
  - $\Omega$  - notation
  - $\Theta$  - notation
  - Best Case, Worst Case, Average Case
  - Computing the complexity of recursive functions

- Dynamic Array
- Iterators
- ADT Bag

- A *record* (or *struct*) is a static data structure.
- It represents the reunion of a fixed number of components (which can have different types) that form a logical unit together.
- We call the components of a record *fields*.
- For example, we can have a record to denote a *Person* formed of fields for *name*, *date of birth*, *address*, etc.

## Person:

name: String

dob: String

address: String

etc.

# Arrays

- An array is one of the simplest and most basic data structures.
- An array can hold a fixed number of elements of the same type and these elements occupy a contiguous memory block.
- Arrays are often used as a basis for other (more complex) data structures.

- When a new array is created we have to specify two things:
  - The type of the elements in the array
  - The maximum number of elements that can be stored in the array (*capacity* of the array)
- The memory occupied by the array will be the capacity times the size of one element.
- The array itself is memorized by the address of the first element.

# Arrays - C++ Example 1

- Assume that we execute the following code in C++, which creates an array that can hold 20 boolean variables and prints the size of a boolean variable, the address of the array and the addresses of the first 5 elements.

```
void booleanExample() {  
    bool boolArray[20];  
    cout << "Size of boolean: " << sizeof(bool) << endl;  
    cout << "Address of array: " << boolArray << endl;  
    for (int i = 0; i < 8; i++) {  
        cout << "Address of element from position " << i << ":" << &boolArray[i]  
            << " " << (unsigned long long) & boolArray[i] << endl;  
    }  
}
```

# Arrays - C++ Example 1

- The output (if you run at home, you will probably get different values):

# Arrays - C++ Example 1

- The output (if you run at home, you will probably get different values):

```
Size of boolean: 1
Address of array: 00EFF760
Address of element from position 0: 00EFF760 15726432
Address of element from position 1: 00EFF761 15726433
Address of element from position 2: 00EFF762 15726434
Address of element from position 3: 00EFF763 15726435
Address of element from position 4: 00EFF764 15726436
Address of element from position 5: 00EFF765 15726437
Address of element from position 6: 00EFF766 15726438
Address of element from position 7: 00EFF767 15726439
```

- Can you guess the address of the element from position 8?

## Arrays - C++ Example 2

- Let's repeat it with an array of *integer* values (integer values occupy 4 bytes)

```
Size of int: 4
Address of array: 00D9FE6C
Address of element from position 0: 00D9FE6C 14286444
Address of element from position 1: 00D9FE70 14286448
Address of element from position 2: 00D9FE74 14286452
Address of element from position 3: 00D9FE78 14286456
Address of element from position 4: 00D9FE7C 14286460
Address of element from position 5: 00D9FE80 14286464
Address of element from position 6: 00D9FE84 14286468
Address of element from position 7: 00D9FE88 14286472
```

- Can you guess the address of the element from position 8?

## Arrays - C++ Example 3

- And repeat it with an array of *fraction* record values (the fraction record is composed of two integers)

```
Size of fraction: 8
Address of array: 007BF97C
Address of element from position 0: 007BF97C 8124796
Address of element from position 1: 007BF984 8124804
Address of element from position 2: 007BF98C 8124812
Address of element from position 3: 007BF994 8124820
Address of element from position 4: 007BF99C 8124828
Address of element from position 5: 007BF9A4 8124836
Address of element from position 6: 007BF9AC 8124844
Address of element from position 7: 007BF9B4 8124852
```

- Can you guess the address of the element from position 8?

- The main **advantage** of arrays is that any element of the array can be accessed in constant time ( $\Theta(1)$ ), because the address of the element can simply be computed with the following formula (considering that the first element is at position 0):

Address of  $i^{th}$  element = address of array +  $i * \text{size of an element}$

- The above formula works even if we consider that the first element is at position 1, but then we need to use  $i - 1$  instead of  $i$ .

- An array is a static structure: once the *capacity* of the array is specified, you cannot add or delete slots from it (you can modify the value of the elements from the slots, but the number of slots, the capacity, remains the same)
- This leads to an important **disadvantage**: we need to know/estimate from the beginning the number of elements:
  - if the capacity is too small: we cannot store every element we want to
  - if the capacity is too big: we waste memory

# Dynamic Array

- There are arrays whose size can grow or shrink, depending on the number of elements that need to be stored in the array: they are called *dynamic arrays* (or *dynamic vectors*).
- Dynamic arrays are still arrays, the elements are still kept at contiguous memory locations and we still have the advantage of being able to compute the address of every element in  $\Theta(1)$  time.

# Dynamic Array - Representation

- In general, for a Dynamic Array we need the following fields:
  - *cap* - denotes the number of slots allocated for the array (its capacity)
  - *nrElem* - denotes the actual number of elements stored in the array
  - *elems* - denotes the actual array with *capacity* slots for TElems allocated

DynamicArray:

cap: Integer

nrElem: Integer

elems: TElem[]

# Dynamic Array - Resize

- When the value of *nrElem* equals the value of *capacity*, we say that the array is full. If more elements need to be added, the *capacity* of the array is increased (usually doubled) and the array is *resized*.
- During the *resize* operation a new, bigger array is allocated and the existing elements are copied from the old array to the new one.
- Optionally, *resize* can be performed after delete operations as well: if the dynamic array becomes "too empty", a resize operation can be performed to shrink its size (to avoid occupying unused memory).

# Dynamic Array - DS vs. ADT

- Dynamic Array is a data structure:
  - It describes how data is actually stored in the computer (in a single contiguous memory block) and how it can be accessed and processed
  - It can be used as representation to implement different abstract data types
- However, Dynamic Array is so frequently used that in most programming languages it exists as a separate container as well.
  - The Dynamic Array is not really an ADT, since it has one single possible implementation, but we can still treat it as an ADT and discuss its interface.

- **Domain** of ADT DynamicArray

$$\mathcal{DA} = \{\mathbf{da} | da = (cap, nrElem, e_1 e_2 e_3 \dots e_{nrElem}), cap, nrElem \in N, nrElem \leq cap, e_i \text{ is of type TElem}\}$$

# Dynamic Array - Interface II

- What operations should we have for a *DynamicArray*?

# Dynamic Array - Interface III

- **init(da, cp)**

- **description:** creates a new, empty DynamicArray with initial capacity  $cp$  (constructor)
- **pre:**  $cp \in N^*$
- **post:**  $da \in DA$ ,  $da.cap = cp$ ,  $da.nrElem = 0$
- **throws:** an exception if  $cp$  is zero or negative

- `destroy(da)`
  - **description:** destroys a DynamicArray (destructor)
  - **pre:**  $da \in \mathcal{DA}$
  - **post:**  $da$  was destroyed (the memory occupied by the dynamic array was freed)

- `size(da)`
  - **description:** returns the size (number of elements) of the DynamicArray
  - **pre:**  $da \in \mathcal{DA}$
  - **post:**  $\text{size} \leftarrow da.\text{nrElem}$  (the number of elements)

- `getElement(da, i)`
  - **description:** returns the element from a position from the DynamicArray
  - **pre:**  $da \in DA, 1 \leq i \leq da.nrElem$
  - **post:**  $\text{getElement} \leftarrow e, e \in TElm, e = da.e_i$  (the element from position  $i$ )
  - **throws:** an exception if  $i$  is not a valid position

- `setElement(da, i, e)`
  - **description:** changes the element from a position to another value
  - **pre:**  $da \in DA, 1 \leq i \leq da.nrElem, e \in TElm$
  - **post:**  $da' \in DA, da'.e_i = e$  (the  $i^{th}$  element from  $da'$  becomes  $e$ ),  $da'.e_j = da.e_j \forall 1 \leq j \leq n, j \neq i.$   
 $\text{setElement} \leftarrow e_{old}, e_{old} \in TElm, e_{old} \leftarrow da.e_i$  (returns the old value from position  $i$ )
  - **throws:** an exception if  $i$  is not a valid position

# Dynamic Array - Interface VIII

- **addToEnd(da, e)**

- **description:** adds an element to the end of a DynamicArray.  
If the array is full, its capacity will be increased
- **pre:**  $da \in \mathcal{DA}$ ,  $e \in TElm$
- **post:**  $da' \in \mathcal{DA}$ ,  $da'.nrElem = da.nrElem + 1$ ;  $da'.e_{da'.nrElem} = e$  ( $da.cap = da.nrElem \Rightarrow da'.cap \leftarrow da.cap * 2$ )

- `addToPosition(da, i, e)`

- **description:** adds an element to a given position in the DynamicArray. If the array is full, its capacity will be increased
- **pre:**  $da \in DA$ ,  $1 \leq i \leq da.nrElem + 1$ ,  $e \in TElem$
- **post:**  $da' \in DA$ ,  $da'.nrElem = da.nrElem + 1$ ,  $da'.e_j = da.e_{j-1} \forall j = da'.nrElem, da'.nrElem - 1, \dots, i + 1$ ,  $da'.e_i = e$ ,  $da'.e_j = da.e_j \forall j = i - 1, \dots, 1$  ( $da.cap = da.nrElem \Rightarrow da'.cap \leftarrow da.cap * 2$ )
- **throws:** an exception if  $i$  is not a valid position ( $da.nrElem+1$  is a valid position when adding a new element)

- `deleteFromPosition(da, i)`
  - **description:** deletes an element from a given position from the DynamicArray. Returns the deleted element
  - **pre:**  $da \in DA, 1 \leq i \leq da.nrElem$
  - **post:**  $da' \in DA, da'.nrElem = da.nrElem - 1, da'.e_j = da.e_{j+1} \forall i \leq j \leq da'.nrElem, da'.e_j = da.e_j \forall 1 \leq j < i$   
 $\text{deleteFromPosition} \leftarrow e, e \in TElem, e = da.e_i;$
  - **throws:** an exception if  $i$  is not a valid position

- **iterator(da, it)**
  - **description:** returns an iterator for the DynamicArray
  - **pre:**  $da \in \mathcal{DA}$
  - **post:**  $it \in \mathcal{I}$ ,  $it$  is an iterator over  $da$ , the current element from  $it$  refers to the first element from  $da$ , or, if  $da$  is empty,  $it$  is invalid

- Other possible operations:
  - Delete all elements from the Dynamic Array (make it empty)
  - Verify if the Dynamic Array is empty
  - Delete an element (given as element, not as position)
  - Check if an element appears in the Dynamic Array
  - Remove the element from the end of the Dynamic Array
  - etc.

# Dynamic Array - Implementation

- Most operations from the interface of the Dynamic Array are very simple to implement.
- In the following we will discuss the implementation of two operations: *addToEnd*, *addToPosition*.
- For the implementation we are going to use the representation discussed earlier:

DynamicArray:

cap: Integer

nrElem: Integer

elems: TElem[]

# Dynamic Array - addToEnd - Case 1

51	32	19	31	47	95				
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 6

- Add the element 49 to the end of the dynamic array

# Dynamic Array - addToEnd - Case 1

51	32	19	31	47	95				
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 6

- Add the element 49 to the end of the dynamic array

51	32	19	31	47	95	49			
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 7

# Dynamic Array - addToEnd - Case 2

51	32	19	31	47	95
1	2	3	4	5	6

- capacity (cap): 6
- nrElem: 6

- Add the element 49 to the end of the dynamic array

# Dynamic Array - addToEnd - Case 2

51	32	19	31	47	95
1	2	3	4	5	6

- capacity (cap): 6
- nrElem: 6

- Add the element 49 to the end of the dynamic array

51	32	19	31	47	95
1	2	3	4	5	6

- capacity (cap): 12
- nrElem: 7

51	32	19	31	47	95	49					
1	2	3	4	5	6	7	8	9	10	11	12

# Dynamic Array - addToEnd

```
subalgorithm addToEnd (da, e) is:  
    if da.nrElem = da.cap then  
        //the dynamic array is full. We need to resize it  
        da.cap ← da.cap * 2  
        newElems ← @ an array with da.cap empty slots  
        //we need to copy existing elements into newElems  
        for index ← 1, da.nrElem execute  
            newElems[index] ← da.elems[index]  
        end-for  
        //we need to replace the old element array with the new one  
        //depending on the prog. lang., we may need to free the old elems array  
        da.elems ← newElems  
    end-if  
    //now we certainly have space for the element e  
    da.nrElem ← da.nrElem + 1  
    da.elems[da.nrElem] ← e  
end-subalgorithm
```

- What is the complexity of *addToEnd*?

# Dynamic Array - addToPosition

51	32	19	31	47	95				
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 6
- Add the element 49 to position 3

# Dynamic Array - addToPosition

51	32	19	31	47	95				
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 6

• Add the element 49 to position 3

The diagram shows a dynamic array with 10 slots. Slots 1 through 6 contain the values 51, 32, 19, 31, 47, and 95 respectively. Slots 7 through 10 are empty. Above each slot from 1 to 6, there is a curved arrow pointing downwards to the right, indicating a rightward shift of elements. The value 49 is placed in slot 3, which is highlighted in red. The original values 19, 31, 47, and 95 are also highlighted in red to show they have been moved.

51	32	49	19	31	47	95			
1	2	3	4	5	6	7	8	9	10

- capacity (cap): 10
- nrElem: 7

• Add the element 49 to position 3

**subalgorithm** addToPosition (da, i, e) **is:**

**if**  $i > 0$  **and**  $i \leq da.nrElem + 1$  **then**

**if**  $da.nrElem = da.cap$  **then** //the dynamic array is full. We need to  
resize it

da.cap  $\leftarrow$  da.cap \* 2

newElems  $\leftarrow$  @ an array with da.cap empty slots

**for** index  $\leftarrow 1$ , da.nrElem **execute**

newElems[index]  $\leftarrow$  da.elems[index]

**end-for**

da.elems  $\leftarrow$  newElems

**end-if** //now we certainly have space for the element e

da.nrElem  $\leftarrow$  da.nrElem + 1

**for** index  $\leftarrow$  da.nrElem, i+1, -1 **execute** //move the elements to the  
right

da.elems[index]  $\leftarrow$  da.elems[index-1]

**end-for**

da.elems[i]  $\leftarrow$  e

**else**

@throw exception

**end-if**

**end-subalgorithm**

- What is the complexity of *addToPosition*?

- **Obs.:** While it is not mandatory to double the capacity, it is important to define the new capacity as a product of the old one with a constant number greater than 1 (just adding one new slot or a constant number of new slots is not OK - you will see later why).
- The value used for increasing the capacity is often called the *growth factor*.

- How do dynamic arrays in other programming languages grow at resize? (I tried to look at capacities for 100 adds)
  - Microsoft Visual C++ <vector> - multiply by 1.5 (initially 0, then 1, 2, 3, 4, 6, 9, 13, 19, 28, 42, 63, 94, etc.)
    - But it also has a *resize* operation which allows you to resize it *manually* and the new size can be any value (less or greater than current size).
  - Java - has two containers based on Dynamic Array
    - *Vector* - multiply by 2 (initially 10, 20, 40, 80, 160)
    - *ArrayList* - not specified in the documentation: *The details of the growth policy are not specified beyond the fact that adding an element has constant amortized time cost.* Many implementations work with a factor of 1.5 (but it cannot be checked, unless you look at the actual implementation).

- Both Vector and ArrayList have an *ensureCapacity* method, where you can make sure that the capacity is at least the specified value. This is useful if a larger number of add operations are going to follow and we want to avoid intermediary resizes.
- Python *list* - this is also implementation dependent and hidden. However, by doing some tricks and accessing the total size of memory occupied by a list you can infer the capacity. In the Python implementation from Visual Studio, capacities are: 4, 8, 16, 24, 32, 40, 52, 64, 76, 92, 108 => the growth factor is about 1.25.
- C# - *List* in C# has a visible attribute *Capacity*, where you can see the current capacity. It is multiplied by 2 (initially 0, 4, 8, 16, 32, 64, 128).
  - This *Capacity* can be used to set the capacity as well, but you cannot make it less than the current number of elements (otherwise an Exception will be thrown).

# Dynamic Array

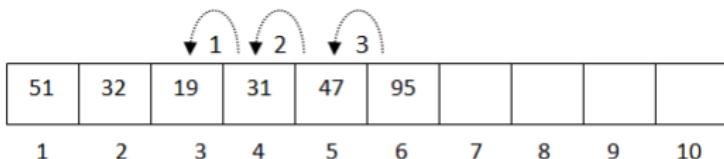
- After a resize operation the elements of the Dynamic Array will still occupy a contiguous memory zone, but it will be a different one than before.

# Dynamic Array

```
Address of the dynamic array variable: 0000002F364FF518
Capacity:3 Nr Elems: 3
Address of the actual array: 0000013712E56530
Address of the first 3 elements:
    Position 0 0000013712E56530
    Position 1 0000013712E56534
    Position 2 0000013712E56538
Address of the dynamic array variable: 0000002F364FF518
Capacity:6 Nr Elems: 6
Address of the actual array: 0000013712E54240
Address of the first 3 elements:
    Position 0 0000013712E54240
    Position 1 0000013712E54244
    Position 2 0000013712E54248
```

# Dynamic Array - delete operation

- To delete an element from a given position  $i$ , the elements after position  $i$  need to be moved one position to the left (element from position  $j$  is moved to position  $j-1$ ).



- capacity (cap): 10
  - nrElem: 5
- Delete the element from position 3

# Dynamic Array - Complexity of operations

- Usually, we can discuss the complexity of an operation for an ADT only after we have chosen the representation. Since the ADT Dynamic Array can be represented in a single way, we can discuss the complexity of its operations:
  - size -

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  - size -  $\Theta(1)$
  - getElement -

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  - getElement -  $\Theta(1)$
  - setElement -

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  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -

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  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -

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  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -  $O(n)$
  - deleteFromEnd -

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  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -  $O(n)$
  - deleteFromEnd -  $\Theta(1)$
  - deleteFromPosition -

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  - size -  $\Theta(1)$
  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -  $O(n)$
  - deleteFromEnd -  $\Theta(1)$
  - deleteFromPosition -  $O(n)$
  - deleteGivenElement -

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  - size -  $\Theta(1)$
  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -  $O(n)$
  - deleteFromEnd -  $\Theta(1)$
  - deleteFromPosition -  $O(n)$
  - deleteGivenElement -  $\Theta(n)$
  - addToEnd -

# Dynamic Array - Complexity of operations

- Usually, we can discuss the complexity of an operation for an ADT only after we have chosen the representation. Since the ADT Dynamic Array can be represented in a single way, we can discuss the complexity of its operations:
  - size -  $\Theta(1)$
  - getElement -  $\Theta(1)$
  - setElement -  $\Theta(1)$
  - iterator -  $\Theta(1)$
  - addToPosition -  $O(n)$
  - deleteFromEnd -  $\Theta(1)$
  - deleteFromPosition -  $O(n)$
  - deleteGivenElement -  $\Theta(n)$
  - addToEnd -  $\Theta(1)$  amortized

# Amortized analysis

- In *asymptotic* time complexity analysis we consider one single run of an algorithm.
  - *addToEnd* should have complexity  $O(n)$  - when we have to resize the array, we need to move every existing element, so the number of instructions is proportional to the length of the array.
  - Consequently, a sequence of  $n$  calls to the *addToEnd* operation would have complexity  $O(n^2)$ .

# Amortized analysis

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  - Consequently, a sequence of  $n$  calls to the *addToEnd* operation would have complexity  $O(n^2)$ .
- In *amortized* time complexity analysis we consider a sequence of operations and compute the average time for these operations.
  - In amortized time complexity analysis we will consider the total complexity of  $n$  calls to the *addToEnd* operation and divide this by  $n$ , to get the *amortized* complexity of the algorithm.

# Amortized analysis

- We can observe that if we consider a sequence of  $n$  operations, we rarely have to resize the array
- Consider  $c_i$  the cost ( $\approx$  number of instructions) for the  $i^{th}$  call to `addToEnd`
- Considering that we double the capacity at each resize operation, at the  $i^{th}$  operation we perform a resize if  $i-1$  is a power of 2. So, the cost of operation  $i$ ,  $c_i$ , is:

$$c_i = \begin{cases} i, & \text{if } i-1 \text{ is an exact power of 2} \\ 1 & \text{otherwise} \end{cases}$$

- Cost of  $n$  operations is:

$$\sum_{i=1}^n c_i \leq n + \sum_{j=0}^{[\log_2 n]} 2^j < n + 2n = 3n$$

- The sum contains at most  $n$  values of 1 (this is where the  $n$  term comes from) and at most (integer part of)  $\log_2 n$  terms of the form  $2^j$ .
- Since the total cost of  $n$  operations is  $3n$ , we can say that the cost of one operation is 3, which is constant.

# Amortized analysis

- While the worst case time complexity of *addToEnd* is still  $O(n)$ , the amortized complexity is  $\Theta(1)$ .
- The amortized complexity is no longer valid, if the resize operation just adds a constant number of new slots.
- In case of the *addToPosition* operation, both the worst case and the amortized complexity of the operation is  $O(n)$  - even if resize is performed rarely, we need to move elements to empty the position where we put the new element.

# When do we have amortized complexity?

- The reason why in case of *addToEnd* we can talk about amortized complexity is that the worst case situation (the resize) happens rarely.
- Whenever you have an algorithm and you want to determine whether amortized complexity computation is applicable, ask the following questions:
  - Can I have worst case complexity for two calls in a row (one after the another)?
- If the answer is YES, than you do not have a situation of amortized complexity computation. (If the answer is NO, it is still not sure that you do have amortized complexity, but if it is YES, you definitely do not have amortized complexity.)

# Amortized analysis

- In order to avoid having a Dynamic Array with too many empty slots, we can resize the array after deletion as well, if the array becomes "too empty".
- How empty should the array become before resize? Which of the following two strategies do you think is better? Why?
  - Wait until the table is only half full ( $da.nrElem \approx da.cap/2$ ) and resize it to the half of its capacity
  - Wait until the table is only a quarter full ( $da.nrElem \approx da.cap/4$ ) and resize it to the half of its capacity

- An *iterator* is an abstract data type that is used to iterate through the elements of a container.
- Containers can be represented in different ways, using different data structures. Iterators are used to offer a common and generic way of moving through all the elements of a container, independently of the representation of the container.
- Every container that can be iterated, has to contain in the interface an operation called *iterator* that will create and return an iterator over the container.

- An iterator usually contains:
  - a reference to the container it iterates over
  - a reference to a *current element* from the container
- Iterating through the elements of the container means actually moving this *current element* from one element to another until the iterator becomes *invalid*
- The exact way of representing the *current element* from the iterator depends on the representation of the container (data structure used for the implementation of the container and possibly other representation details). If the representation/implementation of the container changes, we need to change the representation/implementation of the iterator as well (think about the Bag from Seminar 1).

- **Domain of an Iterator**

$\mathcal{I} = \{\mathbf{it} | \text{it is an iterator over a container with elements of type TElm}\}$

- **Interface** of an Iterator:

- **init(it, c)**

- **description:** creates a new iterator for a container
- **pre:**  $c$  is a container
- **post:**  $it \in \mathcal{I}$  and  $it$  points to the first element in  $c$  if  $c$  is not empty or  $it$  is not valid

- **getCurrent(it)**

- **description:** returns the current element from the iterator
- **pre:**  $it \in \mathcal{I}$ ,  $it$  is valid
- **post:**  $\text{getCurrent} \leftarrow e$ ,  $e \in TElm$ ,  $e$  is the current element from  $it$
- **throws:** an exception if the iterator is not valid

- `next(it)`
- **description:** moves the current element from the container to the next element or makes the iterator invalid if no elements are left
- **pre:**  $it \in \mathcal{I}$ ,  $it$  is valid
- **post:**  $it' \in \mathcal{I}$ , the current element from  $it'$  points to the next element from the container or  $it'$  is invalid if no more elements are left
- **throws:** an exception if the iterator is not valid

- **valid(it)**

- **description:** verifies if the iterator is valid

- **pre:**  $it \in \mathcal{I}$

- **post:**

$$valid \leftarrow \begin{cases} True, & \text{if it points to a valid element from the container} \\ False & \text{otherwise} \end{cases}$$

- **first(it)**

- **description:** sets the current element from the iterator to the first element of the container
- **pre:**  $it \in \mathcal{I}$
- **post:**  $it' \in \mathcal{I}$ , the current element from  $it'$  points to the first element of the container if it is not empty, or  $it'$  is invalid

# Types of iterators I

- The interface presented above describes the simplest iterator: *unidirectional* and *read-only*
- A *unidirectional* iterator can be used to iterate through a container in one direction only (usually *forward*, but we can define a *reverse* iterator as well).
- A *bidirectional* iterator can be used to iterate in both directions. Besides the *next* operation it has an operation called *previous* and it could also have a *last* operation (the pair of *first*).

## Types of iterators II

- A *random access iterator* can be used to move multiple steps (not just one step forward or one step backward).
- A *read-only iterator* can be used to iterate through the container, but cannot be used to change it.
- A *read-write iterator* can be used to add/delete elements to/from the container.

# Using the iterator

- Since the interface of an iterator is the same, independently of the exact container or its representation, the following subalgorithm can be used to print the elements of any container.

**subalgorithm** printContainer(*c*) **is:**

//*pre*: *c* is a container

//*post*: the elements of *c* were printed

//we create an iterator using the iterator method of the container  
iterator(*c*, *it*)

**while** valid(*it*) **execute**

//get the current element from the iterator

getCurrent(*it*, *elem*)

**print** *elem*

//go to the next element

next(*it*)

**end-while**

**end-subalgorithm**

# Iterator for a Dynamic Array

- How can we define an iterator for a Dynamic Array?
- How can we represent that *current element* from the iterator?

# Iterator for a Dynamic Array

- How can we define an iterator for a Dynamic Array?
- How can we represent that *current element* from the iterator?
- In case of a Dynamic Array, the simplest way to represent a *current element* is to retain the position of the *current element*.

IteratorDA:

da: DynamicArray

current: Integer

- Let's see how the operations of the iterator can be implemented.

# Iterator for a Dynamic Array - init

- What do we need to do in the *init* operation?

- What do we need to do in the *init* operation?

**subalgorithm** init(it, da) *is:*

//*it is an IteratorDA, da is a Dynamic Array*

it.da  $\leftarrow$  da

it.current  $\leftarrow$  1

**end-subalgorithm**

- Complexity:

- What do we need to do in the *init* operation?

**subalgorithm** init(it, da) *is:*

//*it is an IteratorDA, da is a Dynamic Array*

it.da  $\leftarrow$  da

it.current  $\leftarrow$  1

**end-subalgorithm**

- Complexity:  $\Theta(1)$

# Iterator for a Dynamic Array - `getCurrent`

- What do we need to do in the `getCurrent` operation?

# Iterator for a Dynamic Array - getCurrent

- What do we need to do in the *getCurrent* operation?

```
function getCurrent(it) is:  
    if it.current > it.da.nrElem then  
        @throw exception  
    end-if  
    getCurrent ← it.da.elems[it.current]  
end-function
```

- Complexity:

# Iterator for a Dynamic Array - `getCurrent`

- What do we need to do in the `getCurrent` operation?

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- Complexity:  $\Theta(1)$

# Iterator for a Dynamic Array - next

- What do we need to do in the *next* operation?

# Iterator for a Dynamic Array - next

- What do we need to do in the *next* operation?

**subalgorithm** next(it) *is:*

**if** it.current > it.da.nrElem **then**

    @throw exception

**end-if**

    it.current  $\leftarrow$  it.current + 1

**end-subalgorithm**

- Complexity:

- What do we need to do in the *next* operation?

**subalgorithm** next(it) *is:*

**if** it.current > it.da.nrElem **then**

    @throw exception

**end-if**

    it.current  $\leftarrow$  it.current + 1

**end-subalgorithm**

- Complexity:  $\Theta(1)$

- What do we need to do in the *valid* operation?

# Iterator for a Dynamic Array - valid

- What do we need to do in the *valid* operation?

```
function valid(it) is:  
    if it.current <= it.da.nrElem then  
        valid ← True  
    else  
        valid ← False  
    end-if  
end-function
```

- Complexity:

- What do we need to do in the *valid* operation?

```
function valid(it) is:  
    if it.current <= it.da.nrElem then  
        valid  $\leftarrow$  True  
    else  
        valid  $\leftarrow$  False  
    end-if  
end-function
```

- Complexity:  $\Theta(1)$

# Iterator for a Dynamic Array - first

- What do we need to do in the *first* operation?

# Iterator for a Dynamic Array - first

- What do we need to do in the *first* operation?

**subalgorithm** first(it) *is:*

```
it.current ← 1
```

**end-subalgorithm**

- Complexity:  $\Theta(1)$

# Iterator for a Dynamic Array

- We can print the content of a Dynamic Array in two ways:
  - Using an iterator (as present above for a container)
  - Using the positions (indexes) of elements

# Print with Iterator

**subalgorithm** printDAWithIterator(da) **is:**

```
//pre: da is a DynamicArray  
//we create an iterator using the iterator method of DA  
iterator(da, it)  
while valid(it) execute  
    //get the current element from the iterator  
    elem ← getCurrent(it)  
    print elem  
    //go to the next element  
    next(it)  
end-while  
end-subalgorithm
```

- What is the complexity of *printDAWithIterator*?

# Print with indexes

```
subalgorithm printDAWithIndexes(da) is:  
    //pre: da is a Dynamic Array  
    for i ← 1, size(da) execute  
        elem ← getElement(da, i)  
        print elem  
    end-for  
end-subalgorithm
```

- What is the complexity of *printDAWithIndexes*?

# Iterator for a Dynamic Array

- In case of a Dynamic Array both printing algorithms have  $\Theta(n)$  complexity
- For other data structures/containers we need iterator because
  - there are no positions in the data structure/container
  - the time complexity of iterating through all the elements is smaller (in general we want all iterator operations to be  $\Theta(1)$ )

- The ADT Bag is a container in which the elements are not unique and they do not have positions.
- Interface of the Bag was discussed at Seminar 1.

- A Bag can be represented using several data structures, one of them being a dynamic array (others will be discussed later)
- Independently of the chosen data structure, there are two options for storing the elements:
  - Store separately every element that was added. (R1)
  - Store each element only once and keep a frequency count for it. (R2)

# ADT Bag - R1 example

- Assume a dynamic array as data structure for the representation (but the idea is applicable for other representations as well)
- Assume that we have a Bag with the following numbers: 4, 1, 6, 4, 7, 2, 1, 1, 1, 9
- In R1 the Bag looks in the following way:

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
4	1	6	4	7	2	1	1	1	9						

- Add element -5

# ADT Bag - R1 example

- Assume a dynamic array as data structure for the representation (but the idea is applicable for other representations as well)
- Assume that we have a Bag with the following numbers: 4, 1, 6, 4, 7, 2, 1, 1, 1, 9
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1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
4	1	6	4	7	2	1	1	1	9	-5					

# ADT Bag - R1 example

- Remove element 6

# ADT Bag - R1 example

- Remove element 6

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16
4	1	-5	4	7	2	1	1	1	9						

- Obs.** The order of the elements in a Bag is not important, you do not need to keep the same order in which they were added. So, when an element needs to be removed, you do not need to do the array-specific *move each element one position to the left* operation, you can simply take the last element and put it in the place of the removed one, for a small improvement in the number of operations.
- What is the complexity of the remove operation when:
  - we move elements to the left ?
  - we take the last element and put it to replace the removed one?

# ADT Bag - R2 example

- Assume the same elements as before: 4, 1, 6, 4, 7, 2, 1, 1, 1, 9
- In R2 the Bag looks in the following way:

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9			
freq	2	4	1	1	1	1			

- Add element -5

# ADT Bag - R2 example

- Assume the same elements as before: 4, 1, 6, 4, 7, 2, 1, 1, 1, 9
- In R2 the Bag looks in the following way:

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9			
freq	2	4	1	1	1	1			

- Add element -5

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9	-5		
freq	2	4	1	1	1	1	1		

- Add element 7

# ADT Bag - R2 example

- Assume the same elements as before: 4, 1, 6, 4, 7, 2, 1, 1, 1, 9
- In R2 the Bag looks in the following way:

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9			
freq	2	4	1	1	1	1			

- Add element -5

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9	-5		
freq	2	4	1	1	1	1	1		

- Add element 7

	1	2	3	4	5	6	7	8	9
elems	4	1	6	7	2	9	-5		
freq	2	4	1	2	1	1	1		

# ADT Bag - R2 example

- Remove element 6

# ADT Bag - R2 example

- Remove element 6

	1	2	3	4	5	6	7	8	9
elems	4	1	-5	7	2	9			
freq	2	4	1	2	1	1			

- Remove element 1

# ADT Bag - R2 example

- Remove element 6

	1	2	3	4	5	6	7	8	9
elems	4	1	-5	7	2	9			
freq	2	4	1	2	1	1			

- Remove element 1

	1	2	3	4	5	6	7	8	9
elems	4	1	-5	7	2	9			
freq	2	3	1	2	1	1			

- Today we have talked about:
  - Dynamic Array - the most basic data structure
  - Iterator
  - ADT Bag - different representations
- Extra reading:
  - How does the iterator look like in Python?