

Algorithms & Data Structures Notes - SoSe 24

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Preface

This is a Quarto book.

To learn more about Quarto books visit <https://quarto.org/docs/books>.

Part I

Introduction

1 Program Run-time Analysis

1.1 Recurrence Relations

Consider a very simple recurrence relation:

$$T(n) := \begin{cases} 1 & n = 1 \\ n + T(n-1), & n > 1 \end{cases}$$

With **mathematical induction** we can formally show that $T(n)$ is quadratic. But there is a simpler & more intuitive way:

$$\begin{aligned} T(n) &= n + T(n-1) && \text{(Def } T(\cdot)\text{)} \\ &= n + n - 1 + T(n-2) \\ &= \dots && \text{(Repeat } n-2 \text{ times)} \\ &= n + n - 1 + \dots + T(1) \\ &= n + n - 1 + \dots + 1 && \text{(Def } T(1)\text{)} \\ &= \frac{n(n+1)}{2} && \text{(Gauss)} \\ &\in \mathcal{O}(n^2) \end{aligned}$$

This method can be applied to the more complex divide-and-conquer recurrence relation from the lecture:

$$R(n) := \begin{cases} a, & n = 1 \\ cn + d \cdot R(\frac{n}{b}), & n > 1 \end{cases}$$

Applying the above method we expand $R(\cdot)$ repetitively according to its definition until we reach the base case, rearranging terms when necessary:

$$\begin{aligned}
R(n) &= c \cdot n + d \cdot R\left(\frac{n}{b}\right) && (\text{Def } R(\cdot)) \\
&= c \cdot n + d\left(c \frac{n}{b} + d \cdot R\left(\frac{n}{b^2}\right)\right) \\
&= c \cdot n + d\left(c \frac{n}{b} + d \cdot \left(c \cdot \frac{n}{b^2} + d \cdot R\left(\frac{n}{b^3}\right)\right)\right) \\
&= c \cdot n + d \cdot c \frac{n}{b} + d^2 c \frac{n}{b^2} + d^3 \cdot R\left(\frac{n}{b^3}\right) && (\text{Rearrange}) \\
&= c \cdot n \left(1 + \frac{d}{b} + \frac{d^2}{b^2}\right) + d^3 \cdot R\left(\frac{n}{b^3}\right) \\
&= \dots && (\text{Repeat } k\text{-times}) \\
&= c \cdot n \left(1 + \frac{d}{b} + \dots + \frac{d^{k-1}}{b^{k-1}}\right) + d^k \cdot R\left(\frac{n}{b^k}\right) \\
&= c \cdot n \sum_{i=0}^{k-1} \left(\frac{d}{b}\right)^i + d^k \cdot R\left(\frac{n}{b^k}\right) \\
&= c \cdot n \sum_{i=0}^{k-1} \left(\frac{d}{b}\right)^i + d^k \cdot R(1) && (\text{Ass } \frac{n}{b^k} = 1) \\
&= c \cdot n \sum_{i=0}^{k-1} \left(\frac{d}{b}\right)^i + a \cdot d^k && (\text{Def } R(1))
\end{aligned}$$

See lecture slides for the complexity analysis of final expression.

1.2 Master Theorem

For recurrence relations of the form:

$$T(n) := \begin{cases} a, & n = 1 \\ b \cdot n + c \cdot T\left(\frac{n}{d}\right), & n > 1 \end{cases}$$

Master theorem gives the solutions:

$$T(n) = \begin{cases} \Theta(n), & c < d \\ \Theta(n \log(n)), & c = d \\ \Theta(n^{\log_b(d)}), & c > d \end{cases}$$

Example: **Merge Sort**.

Complexity of merge sort satisfies the recurrence relation:

$$T(1) = 1$$

$$T(n) = \mathcal{O}(n) + 2 \cdot T\left(\frac{n}{2}\right)$$

Thus with $c = 2 = d$ the second case of MT applies: $T(n) = \Theta(n \log n)$

1.3 Amortized Analysis

Part II

Data Structures

2 Lists

2.1 Sequences as Arrays and Lists

Many terms for same thing: sequence, field, list, stack, string, **file**... Yes, files are simply sequences of bytes!

three views on lists:

- **abstract**: (2, 3, 5, 7)
- **functionality**: stack, queue, etc... What operations does it support?
- **representation**: How is the list represented in a given programming model/language/paradigm?

2.2 Applications of Lists

- Storing and processing any kinds of data
- Concrete representation of abstract data types such as: set, graph, etc...

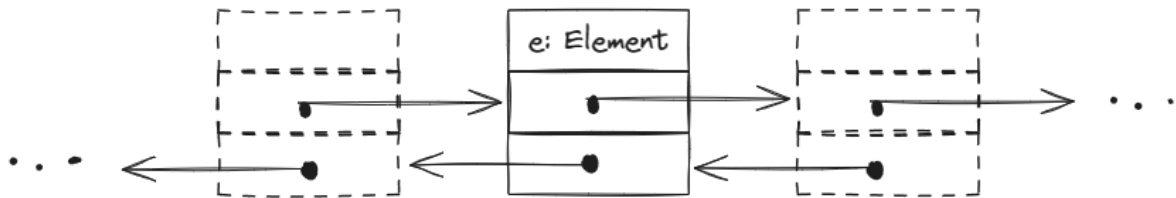
2.3 Linked and Doubly Linked Lists

	simply linked	doubly linked
lecture	<code>SList</code>	<code>List</code>
c++	<code>std::forward_list</code>	<code>std::list</code>

Doubly linked lists are usually **simpler** and require “only” double the space at most. Therefore their use is more widespread.

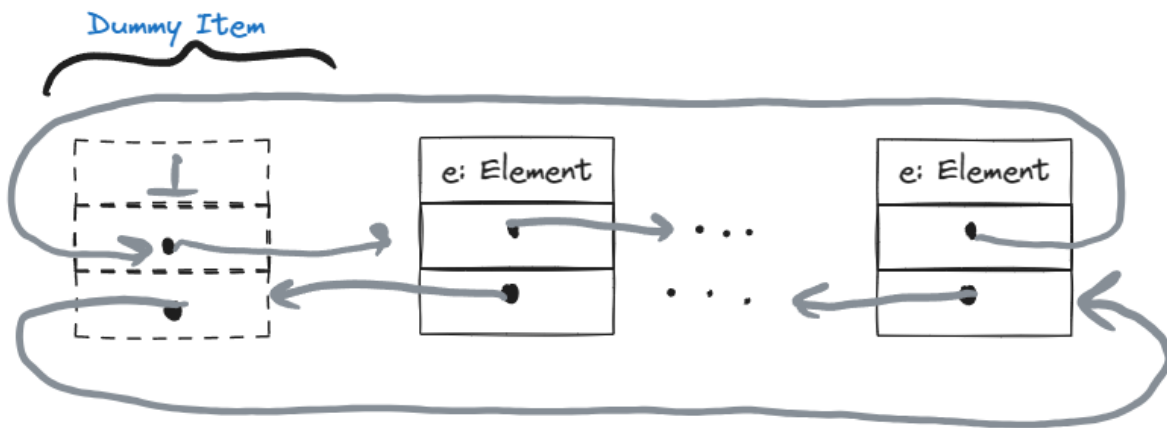
2.3.1 List Items

```
Class Item of T :=  
  e: T //Data item of type T  
  next: *Item //Pointer to Item  
  prev: *Item //Pointer to Item  
  invariant next->prev = this = prev->next
```



Problem: * predecessor of first list element? * successor of last list element?

Solution: Dummy Item with an empty data field as follows:



Advantages of this solution:

- **Invariant** is always satisfied
- Exceptions are avoided, thus making the coding more:
 - simple
 - readable
 - faster
 - elegant

Disadvantages: a little more storage space.

2.3.2 The List Class

```
Class List of T :=
  dummy := (
    Null : T
    &dummy : *T // initially list is empty, therefore next points to the
↪ dummy itself
    &dummy : *T // initially list is empty, therefore prev points to the
↪ dummy itself
  ) : Item

  // returns the address of the dummy, which represents the head of the
↪ list
  Function head() : *Item :=
    return address of dummy

  // simple access functions
  // returns true iff list empty
  Function is_empty() : Bool :=
    return dummy.next == dummy

  // returns pointer to first Item of the list, given list is not empty
  Function first() : *Item :=
    assert (not is_empty())
    return dummy.next

  // returns pointer to last Item of the list, given list is not empty
  Function last() : *Item :=
    assert (not is_empty())
    return dummy.prev

  /* Splice is an all-purpose tool to cut out parts from a list
     Cut out (a, ... b) from this list and insert after t */
  Procedure splice(a, b, t : *Item) :=
    assert (
      b is not before a
      and
      t not between a and b
    )
    // Cut out (a, ... , b)
    a->prev->next := b->next
    b->next->prev := a->prev
```

```

    // insert (a, ... b) after t
    t->next->prev := b
    b->next := t->next
    t->next := a
    a->prev := t

// Moving items by utilising splice
//Move item a after item b
Procedure move_after(a, b: *Item) :=
    splice(a, a, b)

// Move item a to the front of the list
Procedure move_to_front(a: *Item) :=
    move_after(a, dummy)

Procedure move_to_back(a: *Item) :=
    move_after(b, last())

// Deleting items by moving them to a global freeList
// remove item a
Procedure remove(a: *Item) :=
    move_after(b, freeList.dummy)

// remove first item
Procedure pop_front() :=
    remove(first())

//remove last item
Procedure pop_back() :=
    remove(last())

// Inserting Elements
// Insert an item with value x after item a
Function insert_after(x : T, a : *Item) : *Item :=
    checkFreeList() //make sure freeList is non empty
    b := freeList.first() // obtain an item b to hold x
    move_after(b, a) // insert b after a
    b->e := x // set the data item value of b to x
    return b

// Manipulating whole lists

```

```

Procedure concat(L : List) :=
    splice(L.first(), L.last(), last()) //move whole of L after last
↪ element of this list

Procedure clear()
    freeList.concat(this) //after this operation from from first to last
↪ element of this
                                // list are concatenated to the freeList,
↪ leaving only the
                                // dummy element in this list.

Fuction get(i )

```

Splicing

The code for splicing of the List class:

```

/* Splice is an all-purpose tool to cut out parts from a list
   Cut out (a, ... b) form this list and insert after t */
Procedure splice(a, b, t : *Item) :=
    assert (
        b is not before a
        and
        t not between a and b
    )
    // Cut out (a, ... , b)
    a->prev->next := b->next
    b->next->prev := a->prev

    // insert (a, ... b) after t
    t->next->prev := b
    b->next := t->next
    t->next := a
    a->prev := t

```

- Dlist cut-out (a, \dots, b) (see Figure 2.1):
- Dlist insert (a, \dots, b) after t (see Figure 2.2):

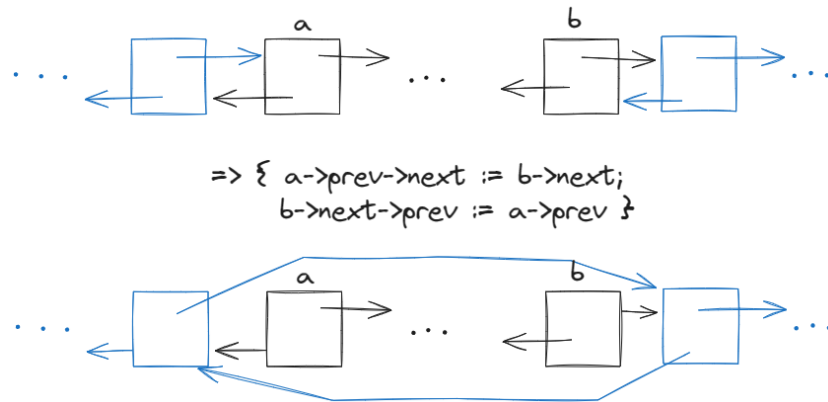


Figure 2.1: cutout

Speicherverwaltung ./FreeList

Methods (?):

- Naively: allocate memory for each new element, deallocate memory after deleting each element:
 - advantage: simplicity
 - disadvantage: requires a good implementation of memory management: potentially very slow
- “global” `freeList` (e.g. `static` member in C++)
 - doubly linked list of all not used elements
 - transfer ‘deleted’ elements in `freeList`.
 - `checkFreeList` allocates, in case the list is empty

Real implementations: * naiv but with well implemented, efficient memory management *
 refined Free List Approach (class-agnostic, release) * implementation-specific.

Deleting Elements

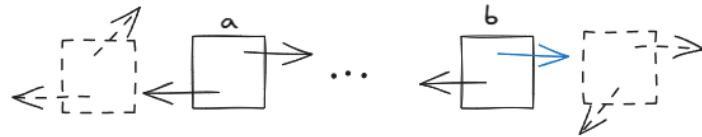
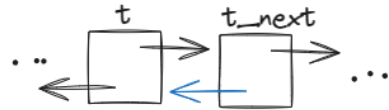
Deleting elements realised by moving them to the global `freeList`:

```

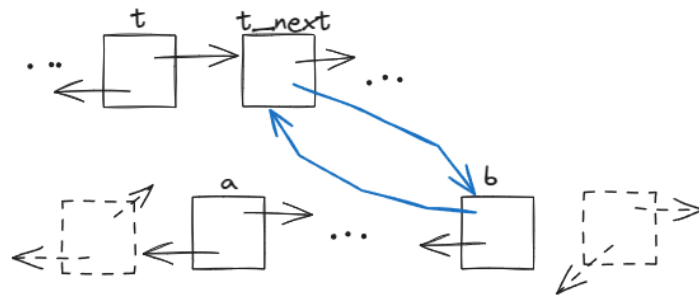
Procedure remove(a: *Item) :=
    move_after(a, freeList.dummy) // item a is now a 'free' item.
  
```

```

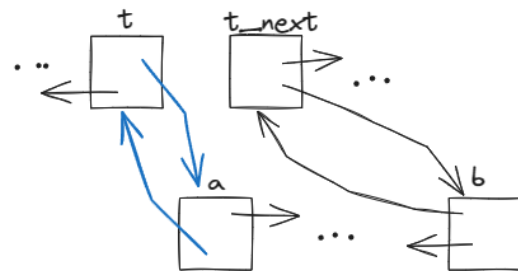
Procedure pop_front() :=
    remove(first())
  
```



$\Rightarrow \{ t \rightarrow \text{next} \rightarrow \text{prev} := b; \\ b \rightarrow \text{next} := t \rightarrow \text{next} \}$



$\Rightarrow \{ t \rightarrow \text{next} := a \\ a \rightarrow \text{prev} := t \}$



\Rightarrow

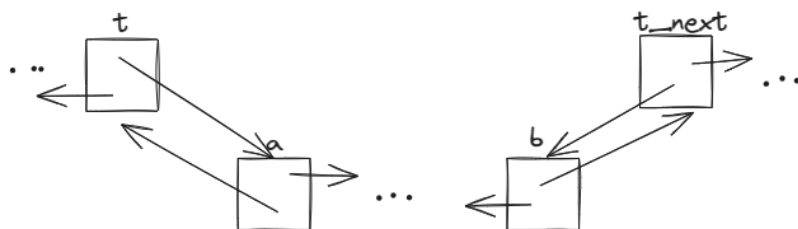


Figure 2.2: insert

```

Procedure pop_back() :=
    remove(last())

```

Inserting Elements

Inserting elements into a list l also utilizes `freeList`, by fetching its first element and moving it into l .

```

Function insert_after(x : T, a : *Item) : *Item :=
    checkFreeList() //make sure freeList is non empty
    b := freeList.first() // obtain an item b to hold x
    move_after(b, a) // insert b after a
    b->e := x // set the data item value of b to x
    return b

```

```

Function insert_before(x : T, b : *Item) : *Item :=
    return insert_after(x, b->prev)

```

```

Procedure push_front(x : T) :=
    insert_after(x, dummy)

```

```

Procedure push_back(x : T) :=
    insert_after(x, last())

```

Manipulating whole Lists

```

// Manipulating whole lists
Procedure concat(L : List) :=
    splice(L.first(), L.last(), last()) //move whole of L after last element
    ↪ of this list

Procedure clear()
    freeList.concat(this) //after this operation from first to last
    ↪ element of this
                                // list are concatenated to the freeList, leaving
    ↪ only the
                                // dummy element in this list.

```

This operations require **constant time** - indeendent of the list size!

3 Arrays

An array is a contiguous sequence of memory cells.

3.1 Bounded Arrays

Bounded arrays have fixed size and are an efficient data structure.

- Size must be known during compile time and is fixed.
- Its memory location in the stack allows many compiler optimizations.

3.2 Unbounded Arrays

The size of an **unbounded array** can dynamically change during run-time. From the user POV it provides the same behaviour as a linked list.

It allows the operations:

- `pushBack(e: T)`: insert an element at the end of the array
- `popBack(e: T)`: remove an element at the end of the array

3.2.1 Memory Management

- `allocate(n)`: request a n contiguous blocks of memory words and returns the address value of the first block. This we have the memory blocks:

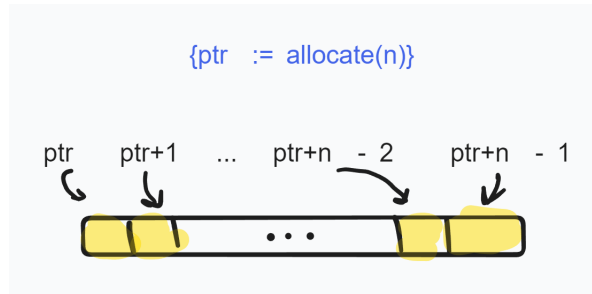


Figure 3.1: array memory allocation

where $\text{ptr} + i$ addresses are determined by pointer arithmetic.

- `dispose(ptr)` marks the memory address value held in `ptr` as free, effectively deleting the object held there.

In general, the allocated memory can't grow dynamically during life time, since the immediate memory block after the last one might get unpredictably occupied \Rightarrow If we need a new memory block of size $n' > n$, we must allocate a new block, copy the old block contents, and finally free it.

3.2.2 Implementation

First we consider a slow variant:

```

Class UArraySlow<T>:=
  c := 0 : Nat // capacity
  b : Array[0..c-1]<T> // the array itself

  pushBack(el : T) : void :=
    // c++
    // allocate new array on heap with new capacity
    // copy elements over from the old array
    // insert el at the last location

  popBack() : void :=
    // analagous

```

Problem: n `pushBack` operations require $1 + \dots + n \in \mathcal{O}(n^2)$ time \Rightarrow slow.

Solution:

Unbounded Arrays with Extra Memory

Idea: Request more memory than initial capacity. Reallocate memory only when array gets full or too empty:

Algorithm design principle: make common case fast.

```
Class UArray<T> :=
  c := 1 : Nat // capacity
  n := 0 : Nat // number of elements in the array

  //invariant  $n \leq c < 2n$  || ( $n == 0 \ \&\& \ c < 2$ )
  b : Array[0..c-1]<T>

  // Array access
  Operator [i : Nat] : T :=
    assert(0 <= i < n)
    return b[i]

  // accessor method for n
  Function size() : Nat := return n

  Procedure pushBack(e : T) :=
    if n == c :
      reallocate(2*n) // see definition below
    b[n] := e
    n++

  // reallocates a new memory with a given capacity c_new
  Procedure reallocate(c_new : Nat) :=
    c := c_new
    b_new := new Array[0..c_new - 1]<T>
    //copy elements over to new array
    for (i = 1 to n - 1) :
      b_new[i] := b[i]
    dispose(b)
    b := b_new

  Procedure popBack() :=
    // don't do anything for empty arrays
    assert n > 0
    n--
```

```
if 4*n <= c && n > 0 :  
    reallocate(2*n)
```

4 Sorting and Priority Queues

4.1 Sorting Algorithms

4.1.1 Insertion Sort

```
def insertion_sort(a) :
    n = len(a)
    # i = 1
    # sorted a[0..i-1]
    for i in range(1, n) :
        # insert i in the right position
        j = i - 1
        el = a[i]
        while el < a[j] and j > 0 :
            a[j + 1] = a[j]
            j = j - 1
        # el >= a[j] or j == 0
        if el < a[j] : # j == 0
            a[1] = a[0]
            a[0] = el
        else : # el >= a[j]
            a[j + 1] = el
    return a
```

testing insertion sort for some inputs:

```
import numpy as np
for i in range(2, 8) :
    randarr = np.random.randint(1, 20, i)
    print("in: ", randarr)
    print("out: ", insertion_sort(randarr))
```

```
in:  [19  8]
out:  [ 8 19]
in:  [12  6  1]
```

```

out: [ 1  6 12]
in:  [2 3 3 8]
out: [2 3 3 8]
in:  [11 4 18 16 14]
out: [ 4 11 14 16 18]
in:  [11 4 18 9 13 3]
out: [ 3 4 9 11 13 18]
in:  [ 8 6 4 11 3 10 7]
out: [ 3 4 6 7 8 10 11]

```

Following illustrates the state after each insertion (ith iteration):

```

def insertion_sort_print(a) :
    n = len(a)
    # i = 1
    # sorted a[0..i-1]
    for i in range(1, n) :
        # insert i in the right position
        j = i - 1
        el = a[i]
        while el < a[j] and j > 0 :
            a[j + 1] = a[j]
            j = j - 1
        # el >= a[j] or j == 0
        if el < a[j] : # j == 0
            a[1] = a[0]
            a[0] = el
        else : # el >= a[j]
            a[j + 1] = el
        print("after insertion ", i, ": ", a)
    # return a

a = np.random.randint(-20, 20, 8)
print("input:          ", a)
insertion_sort_print(a)

```

```

input:          [ 10  15  14   3  11  12 -15 -14]
after insertion 1 : [ 10  15  14   3  11  12 -15 -14]
after insertion 2 : [ 10  14  15   3  11  12 -15 -14]
after insertion 3 : [  3  10  14  15  11  12 -15 -14]
after insertion 4 : [  3  10  11  14  15  12 -15 -14]
after insertion 5 : [  3  10  11  12  14  15 -15 -14]
after insertion 6 : [-15   3  10  11  12  14  15 -14]

```

after insertion 7 : [-15 -14 3 10 11 12 14 15]

4.1.2 Selection Sort

4.1.3 Bubble Sort

4.1.4 Merge Sort

Given by the following python implementation:

```
def merge(a, b) :
    # assert: a and b are sorted
    c = []
    n1 = len(a)
    n2 = len(b)
    k1 = 0
    k2 = 0
    i = 0
    # invariant: merged a[0..k1 - 1] with b[0..k2 - 2]
    while k1 < n1 and k2 < n2 :
        if a[k1] <= b[k2] :
            c.append(a[k1])
            k1 = k1 + 1
        else :
            c.append(b[k2])
            k2 = k2 + 1
    # k1 >= n1 or k2 >= n2
    if k1 == n1 :
        while k2 < n2 :
            c.append(b[k2])
            k2 = k2 + 1
    if k2 == n2 :
        while k1 < n1 :
            c.append(a[k1])
            k1 = k1 + 1
    return c

def merge_sort(a) :
    if len(a) == 1 : return a[0:1]
    n = len(a)
    a1 = a[0 : n // 2]
```

```

a2 = a[n // 2 : ]
return merge(merge_sort(a1), merge_sort(a2))

```

We test on some arrays:

```

for i in range (2, 8) :
    randarr = np.random.randint(-20, 20, i)
    print("in: ", randarr)
    print("out: ", merge_sort(randarr))

```

```

in:  [-7 -7]
out: [-7, -7]
in:  [ 7 -8 -19]
out: [-19, -8, 7]
in:  [-11 -12 17 -2]
out: [-12, -11, -2, 17]
in:  [ 6 -6 0 -6 -16]
out: [-16, -6, -6, 0, 6]
in:  [ 11 -7 -19 -17 -12 13]
out: [-19, -17, -12, -7, 11, 13]
in:  [ 4 15 -14 -19 -6 -14 14]
out: [-19, -14, -14, -6, 4, 14, 15]

```

4.1.5 Quick Sort

Naively:

```

def quicksort(s) :
    if len(s) <= 1 : return s
    p = s[len(s) // 2]
    a = []
    b = []
    c = []
    for i in range(0, len(s)) :
        if s[i] < p : a.append(s[i])
    for i in range(0, len(s)) :
        if s[i] == p : b.append(s[i])
    for i in range(0, len(s)) :
        if s[i] > p : c.append(s[i])
    return quicksort(a) + b + quicksort(c)

```

testing this naive implementation for some arrays:


```

for i in range (2, 8) :
    randarr = np.random.randint(-10, 20, i)
    print("in: ", randarr)
    print("out: ", quicksort(randarr))

```

```

in:  [19  6]
out: [6, 19]
in:  [ 3 19 -4]
out: [-4, 3, 19]
in:  [14 -8 13 10]
out: [-8, 10, 13, 14]
in:  [-6 17 -5 14 -4]
out: [-6, -5, -4, 14, 17]
in:  [ 4  4  1  4 12 -1]
out: [-1, 1, 4, 4, 4, 12]
in:  [-7 -4 17 15 16 -8 14]
out: [-8, -7, -4, 14, 15, 16, 17]

```

4.2 Priority Queues and Heap Data Structure

A set M of Elements $e : T$ with Keys supporting two operations:

- `insert(e)`: Insert e into M .
- `delete_min()`: remove the min element from M and return it.

4.2.1 Applications

- Greedy algorithms (selecting the optimal local optimal solution)
- Simulation of discrete events
- branch-and-bound search
- time forward processing.

4.2.2 Binary Heaps

Heap Property:

- For any leaf $a \in M$ a is a heap.
- Let T_1, T_2 be heaps. If $a \leq x, \forall x \in T_1, T_2$, then $T_1 \circ a \circ T_2$ is also a heap.

Complete Binary Tree:

- A **complete** binary tree is a binary tree in which ever lebel, except possibly the last, is completely filled, and all nodes in the last level are as far left as possible.

Heap:

- A **heap** is a complete binary tree that satisfies the heap property:
- A heap can be succinctly represented as an array:

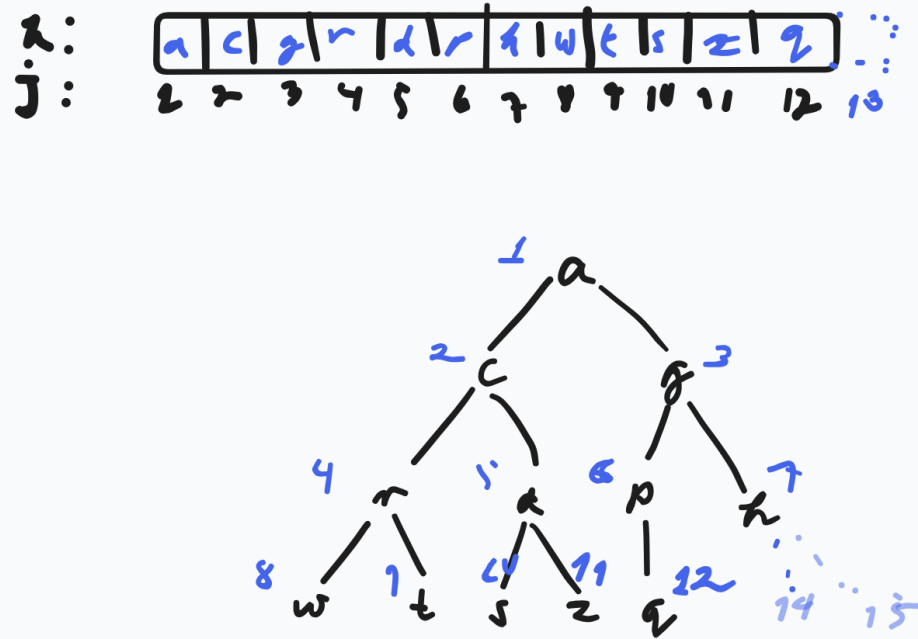


Figure 4.1: heap

- Array $h[1..n]$
- for any given node with the number j :
 - left child: $2*j$
 - right child $2*j + 1$
 - parent: $\text{bottom}(j/2)$

Pseudocode:

```

Class BinaryHeapPQ(capacity: Nat)<T> :=
  h : Array[1..capacity]<T>
  size := 0 : Nat // current amount of elements

```

```

// Heap-property
// invariant:  $h[\text{bottom}(j/2)] \leq h(j)$ , for all  $j == 2..n$ 

Function min() :=
  assert size > 0 // heap non-empty
  return h[1]

Procedure insert(e : T) :=
  assert size < capacity
  size++
  h[n] := e
  siftUp(n)

Procedure siftUp(i : Nat) :=
  // assert Heap-property violated at most at position i
  if i == 1 or  $h[\text{bottom}(i / 2)] \leq h[i]$  then return
  swap(h[i], h[bottom(i/2)])
  siftUp(bottom(i/2))

Procedure popMin : T :=
  result = h[1] : T
  h[1] := h[n]
  n--
  siftDown(1)
  return result

Procedure siftDown (i : Nat) :=
  // assert: Heap property is at most at position  $2*i$  or  $2*i + 1$ 
  ↪ violated
  if  $2i > n$  then return // i is a leaf

  // select the appropriate child
  if  $2*i + 1 > n$  or  $h[2*i] \leq h[2*i + 1]$  :
  //no right child exists or left child is smaller than right
    m := 2*i
  else : m := 2*i + 1
  siftDown(m)

Procedure buildHeap(a[1..n]<T>) :=
  h := a
  buildRecursive(1)

Procedure buildHeapRecursive(i : Nat) :=

```

```

        if 4*i <= n : // children are not leaves
            buildHeapRecursive(2*i) // assert: heap property holds for left
↪ subtree
            buildHeapRecursive(2*i + 1) // assert: heap property holds for
↪ right subtree
            siftDown(i) //assert Heap property holds for subtree starting at i

//alternatively
Procedure buildHeapBackwards :=
    for i := n/2 downto 1 :
        siftDown(i)

Procedure heapSort(a[1..n]<T>) :=
    buildHeap(a) // O(n)
    for i := n downto 2 do :
        h[i] := deleteMin(); // O(log(n))

```

Heap Insert

```

Procedure insert(e : T) :=
    assert size < capacity
    size++
    h[n] := e
    siftUp(n)

Procedure siftUp(i : Nat) :=
    // assert Heap-property violated at most at position i
    if i == 1 or h[bottom(i / 2)] <= h[i] then return
    swap(h[i], h[bottom(i/2)])
    siftUp(bottom(i/2))

```

Illustration of heap insert:

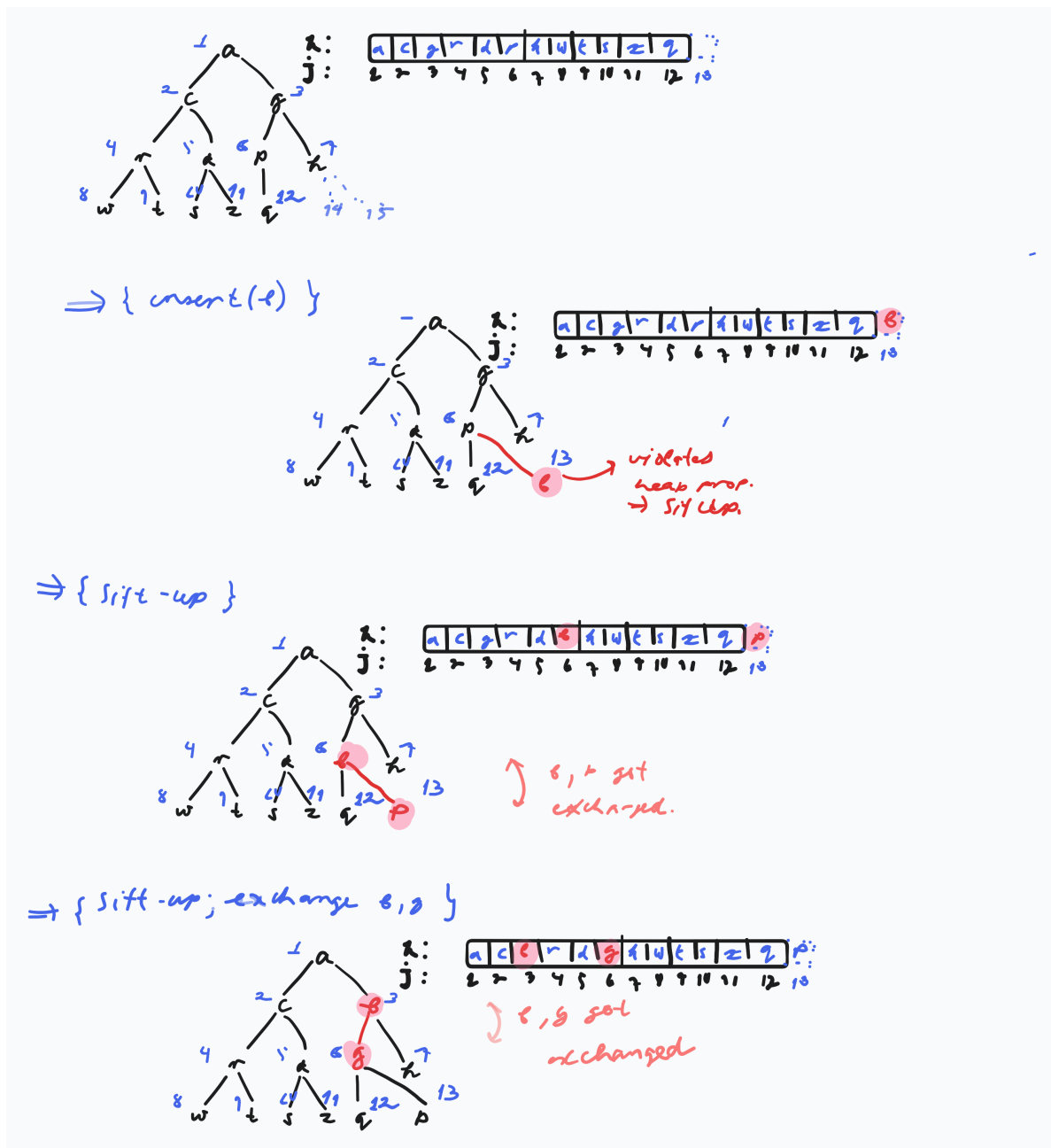


Figure 4.2: heap insert

Heap Pop Min (or Delete Min)

```
Procedure popMin : T :=  
  result = h[1] : T  
  h[1] := h[n]  
  n--  
  siftDown(1)  
  return result
```

```
Procedure siftDown (i : Nat)  
  // assert: Heap property is at most at position 2*i or 2*i + 1 violated  
  if 2i > n then return // i is a leaf  
  
  // select the appropriate child  
  if 2*i + 1 > n or h[2*i] <= h[2*i + 1] :  
  //no right child exists or left child is smaller than right  
    m := 2*i  
  else : m := 2*i + 1  
  siftDown(m)
```

Illustration of pop min:

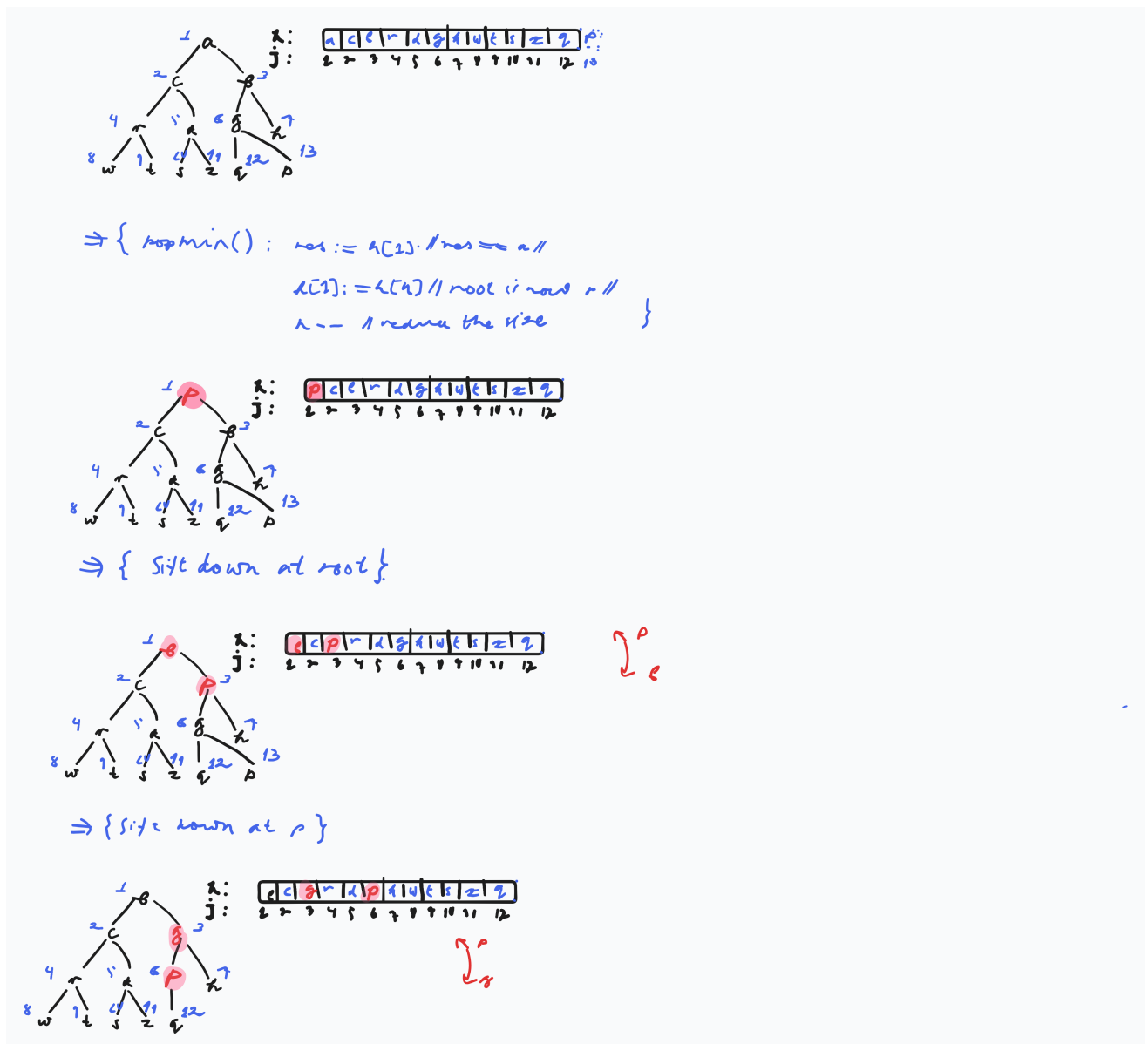


Figure 4.3: heap pop min

Construction of a Binary Heap

- Given are n numbers. Construct a heap from these numbers
- Naive Solution:** n calls to `insert()` $\Rightarrow \mathcal{O}(n \log(n))$
 - Problem: If numbers are given in an array, we can't perform the construction in

place.
 – It is slow

- we can do faster and in place in $\mathcal{O}(n)$ time.

Pseudocode for recursive implementation:

```

Procedure buildHeap(a[1..n] : T) :=
  h := a
  buildRecursive(1)

Procedure buildHeapRecursive(i : Nat) :=
  if 4*i <= n : // children are not leaves
    buildHeapRecursive(2*i) // assert: heap property holds for left
  ↪ subtree
    buildHeapRecursive(2*i + 1) // assert: heap property holds for right
  ↪ subtree
    siftDown(i) //assert Heap property holds for subtree starting at i

```

A simpler iterative one-liner:

```

Procedure buildHeapBackwards :=
  for i := n/2 downto 1 :
    siftDown(i)

```

$\lfloor i/2 \rfloor$ is the last non-leaf node.

Time complexity of these binary heap construction algorithms is $\mathcal{O}(n)$.

Heapsort

```

Procedure heapSort(a[1..n]<T>) :=
  buildHeap(a) //  $\mathcal{O}(n)$ 
  for i := n downto 2 do :
    h[i] := deleteMin(); //  $\mathcal{O}(\log(n))$ 

```

Sorts in decreasing order in $\mathcal{O}(n \log(n))$, by removing the minimal element and writing the return value to the end of the array in place.