

The logo of the Technical University of Cluj-Napoca is a large, light blue watermark in the background. It features a shield with a stylized 'T' and 'U' inside, and the text 'TECHNICAL UNIVERSITY' at the top, 'OF CLUJ-NAPOCA' in the middle, and 'Computer Science' at the bottom.

Fundamental Algorithms

Lecture #3

Computer Science

Agenda

- Master Theorem – to be remembered
- Features to evaluate – review
- Heap structure - review
- QuickSort
- i^{th} Selection
- QuickSort - updated

Master Theorem to remember/to keep close

a = number of recursive calls

b = division factor = ratio between original size over recursive size

c = degree of polynomial of the execution time of the sequence **outside recursive calls**: $f(n) = n^c$

$$t(n) = \begin{cases} t_0 & \text{if } n < n_0 \\ a t(n/b) + f(n) & \text{if } n \geq n_0 \end{cases}$$

1. $q < 1; a < b^c \Rightarrow O(n^c)$
2. $q = 1; a = b^c \Rightarrow O(n^c \log_b n)$
3. $q > 1; a > b^c \Rightarrow O(n^{\log_b a})$

Features to evaluate - review

- Correctness
 - Partial and total
- Efficiency vs. optimality
 - Cases – what do they depend on
 - The problem to be solved
 - The algorithm solving the problem
 - The implementation of the algorithm
- Stability:
 - Stable vs unstable algorithm
- Determinism:
 - Deterministic vs nondeterministic behavior

Heap – as a data structure

- Static data structure (an array)
- Heap utilization when its size changes
- Heap_size – a data field
- Operations:
 - pop_heap extract the top from the heap
 - push_heap add one item to the heap

Computer Science

Heap – as a data structure – cont.

- **pop_heap** Extracts the top element $O(1)$
 - To restore the heap property (after the pop_heap):
 - Move bottom (last) element on top
 - Decrements the heap size
 - Heapify the whole (from 1 to the new size), to update the heap structure $\Rightarrow O(\lg n)$ time to RESTORE the heap property
- **push_heap**
 - Increase the heap_size
 - Adds a new element at the bottom
 - Rebuild heap, a bottom-up approach (bubble the bottom element upper in the heap, until it finds a larger-value parent) $\Rightarrow O(h)=O(\lg n)$

Heap – as a data structure – cont.

- `build_heap`
 - Repeats `push_heap` procedure
 - It takes $1+2\cdot 1+4\cdot 2+\dots+n/2\cdot \lg n=O(n\lg n)$
- `heap_sort`
 - Build the heap (`build_heap` takes $O(n\lg n)$)
 - `pop_heap` (takes $O(\lg n)$)
 - add the popped element at `bottom+1` (i.e. out of the heap, in the array)
 - It takes $O(n\lg n)$ (to build the heap)+ $O(n\lg n)$ (n times a pop operation)

Heap – comparison in building the heap

Approach	Sol 1 (heapify)	Sol2 (pop/push)
1 el approach	sinks the root	bubbles a leaf
	$O(h)$	$O(h)$
all els(build heap)	bottom-up	top-down
approach	(starts with the last non-leaf el)	(adds a new leaf)
Time to build	$O(n)$	$O(n \lg n)$
advantage	faster	variable dim
drawback	fixed dim	slower
usage	sorting	priority queues

Sorting – optimal strategies

- Optimal sorting = algorithm to sort in place (constant additional space) in $O(n \lg n)$ time
- In practice, quicksort, even if not optimal (the original solution), behaves better than heapsort
- A good implementation of quicksort (by injecting various enhancements – see later) IS optimal

QuickSort

```
QuickSort (A, p, r)           //p, r -index of first, last el in
                                //the array A to order

if p < r                       //if proper array (=nonempty)

    then                       q <- partition (A, p, r) //q index returned
                                // at the boundary of the 2 partitions
    QuickSort (A, p, q)
    QuickSort (A, q+1, r)
```

$t(n)$: Master theorem: $f(n)=n \Rightarrow$

$c=1$ (partition, next slide)

$a=2$ (2 rec calls)

$b=?$

Partition (as Hoare originally proposed the algorithm; in the original textbook – first edition)

```
Partition(A, p, r)    //p, r -index of the first, last el in the array
x ← A[p]    i ← p-1    j ← r+1    //pivot is the first element in the array
while i ≤ j do        //as long as left index to the left of right index
    begin
        repeat    j ← j-1
        until    A[j] ≤ x    //stop at the first smaller or equal element to pivot
        repeat    i ← i+1
        until    A[i] ≥ x    //stop at the first greater or equal element to pivot
        if i < j
            then swap (A[i], A[j])
            else return j
    end
```

Partition (Hoare original)

Partition(A,p,r)

x←A[p] **i**←p-1 **j**←r+1

while **i**≤**j** do

begin

repeat **j**←**j**-1

until A[**j**]≤**x**

repeat **i**←**i**+1

until A[**i**]≥**x**

if **i**<**j**

then swap (A[**i**],A[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	9	3	12	5	7	2	9	5

i=0

j=9

Partition (Hoare original)

Partition(A,p,r)

x←A[p] **i**←p-1 **j**←r+1

while **i**≤**j** do

begin

repeat **j**←**j**-1

until **A**[**j**]≤**x**

repeat **i**←**i**+1

until **A**[**i**]≥**x**

if **i**<**j**

then swap (**A**[**i**],**A**[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	9	3	12	5	7	2	9	5
i =0								j =8

Partition (Hoare original)

Partition(A,p,r)

x←A[p] **i**←p-1 **j**←r+1

while **i**≤**j** do

begin

repeat **j**←**j**-1

until **A**[**j**]≤**x**

repeat **i**←**i**+1

until **A**[**i**]≥**x**

if **i**<**j**

then swap (**A**[**i**],**A**[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	9	3	12	5	7	2	9	5
	i =1							j =8

Partition (Hoare original)

```
Partition(A,p,r)
x←A[p]  i←p-1  j←r+1
while i≤j do
  begin
    repeat  j←j-1
    until   A[j]≤x
    repeat  i←i+1
    until   A[i]≥x
    if i<j
      then swap (A[i],A[j])
      else return j
  end
end
```

x=9

	1	2	3	4	5	6	7	8
A	5	3	12	5	7	2	9	9
	i=1							j=8

Partition (Hoare original)

Partition(A,p,r)

x←-A[p] **i**←-p-1 **j**←-r+1

while **i**≤**j** do

begin

repeat **j**←-**j**-1

until **A**[**j**]≤**x**

repeat **i**←-**i**+1

until **A**[**i**]≥**x**

if **i**<**j**

then swap (**A**[**i**],**A**[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	5	3	12	5	7	2	9	9
	i =1						j =7	

Partition (Hoare original)

Partition(A,p,r)

x←A[p] **i**←p-1 **j**←r+1

while **i**≤**j** do

begin

repeat **j**←**j**-1

until A[**j**]≤**x**

repeat **i**←**i**+1

until A[**i**]≥**x**

if **i**<**j**

then swap (A[**i**],A[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	5	3	12	5	7	2	9	9
			i=3				j=7	

Partition (Hoare original)

```
Partition(A,p,r)
x←A[p]  i←p-1  j←r+1
while i≤j do
  begin
    repeat  j←j-1
    until   A[j]≤x
    repeat  i←i+1
    until   A[i]≥x
    if i<j
      then swap (A[i],A[j])
      else return j
  end
```

x=9

	1	2	3	4	5	6	7	8
A	5	3	9	5	7	2	12	9
			i=3				j=7	

Partition (Hoare original)

Partition(A,p,r)

x ← A[p] i ← p-1 j ← r+1

while i ≤ j do

begin

repeat j ← j-1

until A[j] ≤ x

repeat i ← i+1

until A[i] ≥ x

if i < j

then swap (A[i],A[j])

else return j

end

x=9

	1	2	3	4	5	6	7	8
A	5	3	9	5	7	2	12	9
			i=3			j=6		

Partition (Hoare original)

Partition(A,p,r)

x←A[p] **i**←p-1 **j**←r+1

while **i**≤**j** do

begin

repeat **j**←**j**-1

until A[**j**]≤**x**

repeat **i**←**i**+1

until A[**i**]≥**x**

if **i**<**j**

then swap (A[**i**],A[**j**])

else return **j**

end

x=9

	1	2	3	4	5	6	7	8
A	5	3	9	5	7	2	12	9

j=6 **i**=7

Partition (Hoare original)

```
Partition(A,p,r)
x←A[p]  i←p-1  j←r+1
while i≤j do
  begin
    repeat  j←j-1
    until  A[j]≤x
    repeat  i←i+1
    until  A[i]≥x
    if i<j
      then swap (A[i],A[j])
      else return j
  end
```

x=9

	1	2	3	4	5	6	7	8
A	5	3	9	5	7	2	12	9

j=6 i=7

...returns 6

Partition (Hoare original)

```
Partition(A, p, r)    //p, r -index of the first, last el in the array
x<-A[p]  i<-p-1  j<-r+1 //pivot is the first element in the array
while i<=j do        //as long as left index to the left of right index
  begin
    repeat  j<-j-1
    until  A[j]<=x //stop at the first smaller or equal element to pivot
    repeat  i<-i+1
    until  A[i]>=x //stop at the first greater or equal element to pivot
    if i<j
      then swap (A[i],A[j])
      else return j
  end
```

Qs: (individual analysis! Hw!)

- the repeat-until loops stop on equal elements and swaps them. Why?
- the indexes i and j never go beyond the array boundaries. Why?
- First element pivot has an undesired worst case (leads $O(n^2)$ quicksort). Which is it? Why is it undesired?
- using A[p] as pivot is essential in this implementation. Why? Homework!
- using A[r] as pivot would cause error execution. Why? How can it be avoided? Homework!

Partition (Hoare's update)

```
Partition (A, p, r)           //p, r -index of first, last el in the array
x ← -A[(p+r) / 2]             //or p, or r; doesn't matter
i ← p
j ← r
repeat
    while A[i] < x do i = i + 1
    while A[j] > x do j = j - 1
    if i ≤ j then
        begin    swap(A[i], A[j])
                  i = i + 1  j = j - 1
        end
until (j < i)
```

Qs:

- symmetric method. Works the same, whatever (middle, first, last) pivot is chosen.
- the while loops stop on equal elements and swap them. Why? Why not allowing them in the partition they already belong and change the loops conditions to non-strict inequalities?
- In case $i=j$ elements are swapped. It is redundant! Why to swap them? So can we change if $i \leq j$ into if $i < j$? Any trap?

QuickSort – evaluation

b=? It depends on the case.

Cases DEPEND on the pivot choice, hence on the implementation!

Best: each partition divides the array into 2 equal parts => $b=2$ (in the Master theorem) => $O(n \lg n)$

Average: it can be shown it is close to the best case

Worst: each partition divides the array into arrays containing 1 element only and the rest of the elements => rec. calls each time on (1) and (n-1) elements respectively => $n + (n-1) + (n-2) + \dots = O(n^2)$
(for the first/last element chosen as pivot, **ordered array is the worst case!!!!**) TO BE AVOIDED!

QuickSort – evaluation – contd.

- Not an optimal algorithm: $O(n^2) > \Omega(n \cdot \lg n)$
- $O(n \lg n)$ for best and average case
- Worst case occurs seldom
 - How seldom?
- Property of data to enter worst case?
 - How does it depend on the implementation?
 - What factor(s) impact the case?
 - pivot (for Partition) first element worst case?
 - pivot middle element worst case?
- How can ensure we **NEVER** enter the worst case?
 - Always enter the best case
 - Do Partition based on the *median* (ensuring 2 equal halves)
 - Does this affect $f(n)$ (we should stay within $O(n)$)
 - Randomization (TBD)

Median selection

Put QS on hold for now...

- **Selection problem** = given an unordered array, find the element which in the ordered array would occur in the i^{th} position (obviously, without ordering the array)
- Median selection = selection when $i=n/2$

i^{th} Selection

- Selects the i^{th} smallest element from an unordered array
 - TBD on trees as well (dynamic structures)
- Hoare's algorithm – *QuickSelect*
- Resembles the QuickSort algorithm, but with just one recursive call

i^{th} Selection – code

```
QuickSelect(A, p, r, i) //p=first, r=last, i=desired rank
    if p=r    //got the  $i^{\text{th}}$  element in the right place
        then return A[p]
    q<-partition(A, p, r)    // q =index of the position
                             //where the partition stops
    k<-q-p+1    //k=length of the left partition, rank of the pivot
    if i=k
        then return A[q]
    else if i<k
        then return QuickSelect(A, p, q-1, i)
    else return QuickSelect(A, q+1, r, i-k)
```

QuickSelect – evaluation

- Problem lower bound: $\Omega(n)^1$
- Cases are similar to `QuickSort`, yet just a single recursive call
- Worst
 $t(n) = n + (n-1) + (n-2) + \dots = O(n^2) \Rightarrow$ **NOT** optimal
- Average
 $t(n) = n + n/2 + \dots = O(n)$
- Best
Element found after a single partition pass (no recursive call) $\Rightarrow O(n)$

Optimal Selection

- The same situation as for **QuickSort**: need to avoid worst case!
- *Akl's algorithm* = derived from parallel processing
- Splits the input data into a sub-arrays such that the selection is optimal

AklSelection

AklSelection ($A[1,n],i$)

1. Split the array into sub-arrays of dim **a** each A_i , $i=1,n/a$.
2. Direct sort each A_i , and find its median, **m_i** .
3. Generate the array of medians, and call the *AklSelection*($m[1,n/a],n/a$) on the new array, to select the median of medians (i.e. $M=m[n/a]$).
4. Partition the input array into elements \leq and $\geq M$ respectively. Assume there are k elements $\leq M$.
5. if $i \leq k$
 then *AklSelection*($A[1,k],i$)
 else *AklSelection*($A[k+1,n],i-k$)

AklSelection – algorithm evaluation

- **Determine a such that the alg. is optimal**
- **$\Omega(n)$ \Rightarrow it should be $O(n)$**
- Assume $t(n)$ the running time
- The steps:
 1. (split) $a=\text{constant} \Rightarrow \mathbf{c_1 \cdot n}$
 2. (sort) $O(1)$ for one seq, n/a seqs $\Rightarrow \mathbf{c_2 \cdot n}$
 3. (rec. call on n/a elements) $\Rightarrow \mathbf{t(n/a)}$
 4. (partition) $\Rightarrow \mathbf{c_4 \cdot n}$
 5. (rec. call on one partition) \Rightarrow at most **$t(3n/4)$**
(justification follows in 2 slides)

AklSelection – alg. eval. – contd.

We have: $t(n) = c \cdot n + t(n/a) + t(3n/4)$ (1)

We need: $t(n) \leq k \cdot n$ (2)

Therefore:

$$\begin{aligned} t(n) &= c \cdot n + t(n/a) + t(3n/4) \\ &\leq c \cdot n + k \cdot n/a + k \cdot 3n/4 \leq k \cdot n \end{aligned} \quad (3)$$

$$\Rightarrow c \cdot n \leq k \cdot (1/4 - 1/a) \cdot n$$

$$c > 0, a > 0 \Rightarrow 1/4 - 1/a > 0 \Rightarrow a > 4 \Rightarrow \mathbf{a_{min} = 5}$$

For $a=5$, we have that $\exists c$ s.t. $t(n) = c \cdot n = \mathbf{O(n)}$

OPTIMAL!

AklSelection – alg. eval. – contd.

- Why is step #5 $t(3n/4)$ at most?
 - $M \leq$ half of m_i 's $\Rightarrow \exists n/2a$ m_i 's such that

$$m_i \geq M \quad (1)$$
 - Each median m_i is \geq and \leq than exactly half of the nb. of elements in A_i , hence $\exists a/2$ A_i 's such that

$$m_i \leq A_i \quad (2)$$
- (1) $\Rightarrow M$ is \leq than $n/2a$ medians m_i
- (2) \Rightarrow Each such median $\leq a/2$ elements
- Overall: $M \leq$ than at least $n/2a \cdot a/2 = n/4$ elements

AklSelection – alg. eval. – contd.

- With a similar reasoning, $M \geq$ than at least $n/4$ elements
- How are the rest?
 - Unknown!
- So?
 - The longest recursive call is on $3n/4$
- Conclusion: AklSelection is optimal for $a \geq 5$
- In practice, for parallel execution, $a=8$ (or another power of 2; depends on the nb. of processing units available)

Median selection

- May use it in QS
 - its optimal version has $O(n)$
 - by median partition, QS enters best case always
- Resume QS

QuickSort revised (rv1)

QuickSort (A, p, r)

if $p < r$

then

$q \leftarrow \text{partition}(A, p, r)$

QuickSort (A, p, q)

QuickSort (A, q+1, r)

- Worst case running time: $O(n^2)$ due to uneven partitioning
- Avoid worst case: use the “right” partitioning sequence (i.e. split input data into 2 equal subsets)

QuickSort revised (rv1) – cont.

- Element to split the input data = median
(i.e. element which in the ordered array would occur in the middle)
- Use a Median Selection **before** partitioning (we'll see shortly that's actually **instead** of partitioning)
- Selection – revised
 - Hoare's alg.
 - kind of QS with only 1 recursive call
 - inefficient $O(n^2)$ worst case running time; no improvement
 - Akl's alg (the one described before)
 - Optimal for $a \geq 5 \Rightarrow O(n)$
 - Multiplicative ct. very large (i.e. in the average case, Hoare's alg. is much better!)

QuickSort revised (rv1) - transformation with selection

QuickSort (A, p, r)

if $p < r$

then

Ak1Select (A, p, r, $|A|/2$)

q ← partition (A, p, r) //use the element returned by Select

QuickSort (A, p, q)

QuickSort (A, p, $|A|/2$)

QuickSort (A, q+1, r)

QuickSort (A, $|A|/2+1$, r)

Q: what is the effect of partition?

Is it required any more?

Note: partitioning and the blue QS calls get out

QuickSort rv1

QuickSort (A, p, r)

if $p < r$

then

Ak1Select (A, p, r, $|A|/2$) //determines the median, and
//partitions based on the median

QuickSort (A, p, $|A|/2$)

QuickSort (A, $|A|/2+1$, r)

- How many rec. calls?
- Half done on leaves (i.e. empty data structures, thus call and return – takes time for doing nothing)
- What is the efficiency of rec. calls on small data structures?
- Avoid rec. calls on small data.

QuickSort rv1 enhanced

QuickSort (A, p, r)

if (r-p) < δ

then direct_sort (A, p, r) //which one?

else

 AklSelect (A, p, r, |A| / 2)

 QuickSort (A, p, |A| / 2)

 QuickSort (A, |A| / 2 + 1, r)

Enhancements

$p-r < \delta$ saves time (secs, overhead of calls/restores from calls),

Select ensures the optimality (always falling into the best case) of the alg

QuickSort revised (rv2)

- In rv1, **Ak1Select** guarantees best case always
- QuickSort is $O(n \lg n)$ in the average case
- **It's enough to avoid the worst case**
- A **random** partition ensures this!
- Before partitioning, at each step pick a **random** element to make the partitioning based on that element (so swap the random chosen element with the element placed in the position of the pivot – first/middle/last)

QuickSort rv2 – cont.

random_partition(A, p, r)

i ← random(p, r) //choose a random element

A[i] ← A[p] //put it in the first position

return partition(A, p, r) //here we have the
// regular one

QuickSort-Random(A, p, r)

if p < r

then q ← random_partition(A, p, r)

QuickSort(A, p, q)

QuickSort(A, q+1, r)

QuickSort rv2 enhanced

QuickSort-Random (A, p, r)

if $(r-p) < \delta$

then $\text{direct_sort}(A, p, r)$

else $q \leftarrow \text{random_partition}(A, p, r)$
 $\text{QuickSort}(A, p, q)$
 $\text{QuickSort}(A, q+1, r)$

Merge Sort

- Divide et impera
- Attempts (and succeeds) to always enter the best case
- Approach (see blackboard)
 - Divides the input into 2 equal partitions (chooses the middle)
 - Apply 2 recursions
 - Merge the resulting ordered halves
- Master Theorem: $a = 2; b = 2; c = 1 \Rightarrow O(n \lg n)$. Is it optimal? Why/why not?
- Where does it apply?

Sorting – conclusions

- No direct method is optimal; all are $O(n^2)$, even if some behave well in best/average cases
- **Heapsort – is optimal**
- Heaps often used in **Priority Queues**
- QuickSort
 - classic version not optimal
 - Improved versions optimal:
 - Choose a **random** element to make the split
 - Use an **optimal selection** alg. (Akl's) to find the "split" point
 - Augment the alg with a direct method for small arrays, s.t. improve time (in secs, not $t(n)$)

Required Bibliography

- From the Bible – Chapter 7 (QuickSort), Sections 9.2 and 9.3 (Selection problem algorithms)