

Lecture #3 Sorting. QuickSort

Fundamental Algorithms

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Agenda

- 1 Master Theorem - review
- 2 Algorithm features to evaluate - review
- 3 QuickSort
- 4 i^{th} selection



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Master Theorem - review

$$T(n) = \begin{cases} T_0, & \text{if } n < n_0 \\ a * T(\frac{n}{b}) + n^c, & \text{otherwise} \end{cases} \quad (1)$$

where ...

- a : # of recursive calls
- b : division factor = ratio between original size and recursive size
- c : degree of polynomial of the execution time of the sequence excepting the recursive calls: $f(n) = n^c$

- 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. $q > 1; a > b^c \Rightarrow O(n^{\log_b a})$



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Algorithm features to evaluate - review

- Correctness
 - *Partial vs total*
- Efficiency vs optimality
 - Cases depend on
 - the *problem* being solved
 - the *algorithm* solving the problem
 - the *implementation* of the algorithm
- Stability
 - *Stable vs unstable* algorithm
- Determinism
 - *Deterministic vs nondeterministic* behavior



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QuickSort

- In a nutshell...
 - base (vanilla) algorithm not optimal
 - better than *Heapsort* in practice
 - can be made optimal (sort in at most $O(n \lg n)$ time, with constant additional space)



QuickSort

QUICKSORT(A, p, r)

```
1  if  $p < r$  // if non empty array
2       $q = \text{PARTITION}(A, p, r)$ 
3      //  $q$  is an index, boundary between the two partitions
4      QUICKSORT( $A, p, q$ )
5      QUICKSORT( $A, q + 1, r$ )
```

- Why are the two partitions like that?
- Where is the pivot?



Partition (Hoare)

HOARE-PARTITION(A, p, r)

```
1   $x = A[p]$ 
2   $i = p - 1$ 
3   $j = r + 1$ 
4  while true
5      repeat
6           $j = j - 1$ 
7      until  $A[j] \leq x$ 
8      repeat
9           $i = i + 1$ 
10     until  $A[i] \geq x$ 
11     if  $i < j$ 
12         exchange  $A[i]$  with  $A[j]$ 
13     else return  $j$ 
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```

$x=9$

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

- What values will i and j have after the first **while** iteration ?
- How will A look after the first **while** iteration ?



Partition (Hoare)

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```

$x=9$

A

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

- How will the array look after the algorithm returns?
- What will the algorithm return?



Partition (Hoare)

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1   $x = A[p]$ 
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```

$x=9$

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

- How will the array look after the algorithm returns?
- What will the algorithm return?



Partition (Hoare)

!! Homework !!

HOARE-PARTITION(A, p, r)

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1   $x = A[p]$ 
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4  while true
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```

- i and j never go out of the array boundaries. Why?
- **repeat-until** loops stop on equal elements and swaps them. Why?
- $A[p]$ as pivot has an undesired worst case (leads QuickSort to $O(n^2)$). Which is it? Why is it undesired?
- $A[p]$ pivot is essential for correctness. Why? (e.g. $A[r]$ as pivot causes execution error. Why?)
- $T(n) = ?$



Partition (Hoare Update)

HOARE-PARTITION-UPDATE(A, p, r)

```
1   $x = A[(p + r)/2]$ 
2   $i = p$ 
3   $j = r$ 
4  repeat
5      while  $A[i] < x$ 
6           $i = i + 1$ 
7      while  $A[j] > x$ 
8           $j = j - 1$ 
9      if  $i \leq j$ 
10         exchange  $A[i]$  with  $A[j]$ 
11          $i = i + 1$ 
12          $j = j - 1$ 
13 until  $i < j$ 
14 return  $j$ 
```

- other pivot choices possible, but small adjustments needed (*Hw*)



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- other pivot choices possible, but small adjustments needed (*Hw*)
- Trace the execution on the array $A = \{9, 3, 12, 5, 7, 2, 9, 5\}$ (*Hw*)



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- other pivot choices possible, but small adjustments needed (*Hw*)
- Trace the execution on the array $A = \{9, 3, 12, 5, 7, 2, 9, 5\}$ (*Hw*)
- **while** loop stops on equal elements and swaps them. Can they be left in the original partition? (i.e. use non-strict inequalities)



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- other pivot choices possible, but small adjustments needed (*Hw*)
- Trace the execution on the array $A = \{9, 3, 12, 5, 7, 2, 9, 5\}$ (*Hw*)
- **while** loop stops on equal elements and swaps them. Can they be left in the original partition? (i.e. use non-strict inequalities)
- if $i = j$, elements are swapped. Redundant?



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- other pivot choices possible, but small adjustments needed (*Hw*)
- Trace the execution on the array $A = \{9, 3, 12, 5, 7, 2, 9, 5\}$ (*Hw*)
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- if $i = j$, elements are swapped. Redundant?
- $T(n) = ?$



QuickSort evaluation

QUICKSORT(A, p, r)

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- $a = ?$
- $b = ?$
- $c = ?$



QuickSort evaluation

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- $a = 2$
- $b = \dots$ depends on the case. On what, specifically?
- $c = 1$



QuickSort evaluation

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```

- $a = 2$
- $b = \dots$ depends on the pivot choice, so on the *implementation*!
- $c = 1$



QuickSort evaluation

b depends on the pivot choice, so on the *implementation*!

- **best** case: $b = 2$ (2 equal partitions); $T(n) = O(n \lg n)$
- **average** case: $b = 2$ – can be shown, on average, the partitions are balanced enough; $T(n) = O(n \lg n)$
- **worst** case: a partition with 1 element, the other with $n - 1$ elements; $T(n) = O(n) + T(n - 1)$ (why?)
So ... $T(n) = O(n^2)$
- Additional memory?



QuickSort evaluation

- NOT optimal: $T(n) = O(n^2) > \Omega(n \lg n)$
- BUT worst case occurs seldom
 - How seldom?
 - Property of the data to enter worst case?
 - What factor(s) impact the case?
 - How does it depend on the implementation?
- Can we ensure we NEVER enter the worst case?
 - Always enter best case (ensure balanced partitions, in $O(n)$) ... coming next
 - Randomization ... TBD



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i^{th} selection

- putting QS on hold for now, to discuss about:
 - the **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 = particular instance of the problem, when $i = n/2$
 - Algorithms:
 - QuickSelect (Hoare); based on QuickSort (only 1 recursive call)
 - AklSelect - strategy, parallel processing; optimal!



QuickSelect (Hoare)

QUICKSELECT(A, p, r, i)

```
1  //  $p$  - first,  $r$  - last,  $i$  desired rank
2  if  $p = r$  // we are on the correct array position
3      return  $A[p]$ 
4   $q = \text{HOARE-PARTITION}(A, p, r)$  //  $q$  - index where partition ends
5   $k = q - p + 1$  //  $k$  - length of the  $\leq$  partition
6  if  $i \leq k$ 
7      return QUICKSELECT( $A, p, q, i$ )
8  else return QUICKSELECT( $A, q + 1, r, i - k$ )
```

- Why only 1 recursive call?



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- Why only 1 recursive call?
- Why $i - k$ on recursive call in line 8?



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```

- Why only 1 recursive call?
- Why $i - k$ on recursive call in line 8?
- Trace execution for QUICKSELECT($A, 1, 8, 3$), for $A = \{4, 8, 1, 9, 3, 4, 2, 6\}$



QuickSelect (Hoare)

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```

- What changes need to be done to use LOMUTO-PARTITION(A, p, r) instead? (*Hw*)



QuickSelect (Hoare) evaluation

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- $a = ?$
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QuickSelect (Hoare) evaluation

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```

- $a = 1$
- $b = \dots$ depends on the pivot choice, so on the *implementation!*
- $c = 1$



QuickSelect (Hoare) evaluation

- Problem lower bound: $\Omega(n)$ ¹
- **best** case: element found after a single partition pass:
 $T(n) = O(n)$ - how?
- **average** case: $T(n) = n + n/2 + n/4 + \dots = O(n)$
- **worst** case: $T(n) = O(n) + T(n-1)$ (why?)
So ... $T(n) = O(n^2)$ - NOT optimal
- Additional memory?

¹<https://jeffe.cs.illinois.edu/teaching/497/02-selection.pdf>



Akl's Algorithm

- derived from parallel processing, strategy rather than algorithm
- idea - split data into sub-arrays, to make selection optimal



Akl's Algorithm

AKLSELECT($A[1..n]$, i)

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



AklSelect evaluation

- Problem lower bound: $\Omega(n)$
- Determine a such that the algorithm is optimal
- According to the algorithm steps: $T(n) =$



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 - 1. (split), a - cst.:



AklSelect evaluation

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- According to the algorithm steps: $T(n) =$
 - 1. (split), a - cst.: $c_1 * n$



AklSelect evaluation

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 - 1. (split), a - cst.: $c_1 * n$
 - 2. (sort), a - cst.:



AklSelect evaluation

- Problem lower bound: $\Omega(n)$
- Determine a such that the algorithm is optimal
- According to the algorithm steps: $T(n) =$
 - 1. (split), a - cst.: $c_1 * n$
 - 2. (sort), a - cst.: $O(1)$ for 1 seq., n/a seqs, so: $c_2 * n$



AklSelect evaluation

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- Determine a such that the algorithm is optimal
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 - 1. (split), a - cst.: $c_1 * n$
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 - 3. (rec. call on n/a elems):



AklSelect evaluation

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 - 1. (split), a - cst.: $c_1 * n$
 - 2. (sort), a - cst.: $O(1)$ for 1 seq., n/a seqs, so: $c_2 * n$
 - 3. (rec. call on n/a elems): $T(n/a)$



AklSelect evaluation

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 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition):



AklSelect evaluation

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 - 1. (split), a - cst.: $c_1 * n$
 - 2. (sort), a - cst.: $O(1)$ for 1 seq., n/a seqs, so: $c_2 * n$
 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition): $c_4 * n$



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 - 1. (split), a - cst.: $c_1 * n$
 - 2. (sort), a - cst.: $O(1)$ for 1 seq., n/a seqs, so: $c_2 * n$
 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition): $c_4 * n$
 - 7-9. (rec. call on one partition):



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 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition): $c_4 * n$
 - 7-9. (rec. call on one partition): at most $T(3n/4)$ (justification in 2 slides)



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 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition): $c_4 * n$
 - 7-9. (rec. call on one partition): at most $T(3n/4)$ (justification in 2 slides)
- $T(n) = c * n + T(n/a) + T(3n/4)$



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 - 3. (rec. call on n/a elems): $T(n/a)$
 - 4. (partition): $c_4 * n$
 - 7-9. (rec. call on one partition): at most $T(3n/4)$ (justification in 2 slides)
- $T(n) = c * n + T(n/a) + T(3n/4)$
- need $T(n) \leq k * n$



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 - 3. (rec. call on n/a elems): $T(n/a)$
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 - 7-9. (rec. call on one partition): at most $T(3n/4)$ (justification in 2 slides)
- $T(n) = c * n + T(n/a) + T(3n/4)$
- need $T(n) \leq k * n$
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- Determine a such that the algorithm is optimal
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- considering $c > 0, a > 0$; solve $- > a_{\min} = 5$
- So, for $a \geq 5, \exists c$ s.t. $T(n) = O(n)$, so it is OPTIMAL



AklSelect evaluation

- Why is the effort in steps 7-9. at most $T(3n/4)$?



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- Using a similar reasoning: $M \geq n/4$ elements
- The other at most $3n/4$ are unknown \Rightarrow longest recursive call is on $3n/4$ elements



QuickSort improvements (1)

QUICKSORTV2(A, p, r)

```
1  if  $p < r$  // if non empty array
2      AKLSELECT( $A, p, r, |A|/2$ )
3      // determines the median, and partitions based on the median
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- Avoid uneven partitioning by always splitting by the median
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- Avoid uneven partitioning by always splitting by the median
 - QuickSelect still has $O(n^2)$ worst case running time
 - AkSelect optimal for $a \geq 5$ but very large multiplicative constant!
 - QuickSelect is much better on average than AkSelect!



QuickSort improvements (1) + hybridization

QUICKSORTV21(A, p, r)

```
1  if  $r - p < \epsilon$  // if non empty array
2      DIRECTSORT( $A, p, r$ ) // which one?
3  else AKLSELECT( $A, p, r, |A|/2$ )
4      // determines the median, and partitions based on the median
5      QUICKSORTV21( $A, p, |A|/2$ )
6      QUICKSORTV21( $A, |A|/2 + 1, r$ )
```

- hybridization saves time from the overhead of calls/restores from calls (call stack operations)



QuickSort improvements (2) + hybridization

QUICKSORTV3(A, p, r)

```
1  if  $r - p < \epsilon$  // if non empty array
2      DIRECTSORT( $A, p, r$ )
3  else  $q = \text{RANDOMPARTITION}(A, p, r)$ 
4      QUICKSORTV3( $A, p, q$ )
5      QUICKSORTV3( $A, q + 1, r$ )
```

- In V2, AklSelect guarantees best partitioning always, but with large constant increase



QuickSort improvements (2) + hybridization

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- In V2, AkSelect guarantees best partitioning always, but with large constant increase
- QuickSort has a very low constant in the average case



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- QuickSort has a very low constant in the average case
- So, avoid the worst case
- A random partition ensures this!



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```

RANDOMPARTITION(A, p, r)

```
1   $i = \text{RANDOM}(p, r)$ 
2  exchange  $A[p]$  with  $A[i]$ 
3  return HOARE-PARTITION( $A, p, r$ )
```



MergeSort

MERGE_SORT(A, p, r)

```
1  if  $p \geq r$  // zero or one element?  
2      return  
3   $q = \lfloor (p + r) / 2 \rfloor$   
4  MERGE_SORT( $A, p, q$ )  
5  MERGE_SORT( $A, q + 1, r$ )  
6  MERGE( $A, p, q, r$ )
```

- also uses *divide et impera*
- partitions always balanced
- $T(n) = 2T(n/2) + O(n)$
- NOT optimal. Why?
- When do we use it?



Sorting - conclusions

- No direct method is optimal; all are $O(n^2)$, even if some behave well in best case
- HeapSort is **optimal**
- Heaps used to implement priority queues
- QuickSort
 - classic version not optimal
 - improved versions are optimal
 - Choose a **random** pivot to make the split
 - Use an **optimal selection** alg. (Akl) to find the “split” point
 - Augment the alg. with a direct method for small arrays, to improve time (in secs, not $T(n)$)



Required Bibliography

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