



INF421, Lecture 7

Sorting

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Course

- **Objective:** teach notions AND develop intelligence
- **Evaluation:** TP noté en salle info, Contrôle à la fin. Note:
 $\max(CC, \frac{3}{4}CC + \frac{1}{4}TP)$
- **Organization:** fri 31/8, 7/9, 14/9, 21/9, 28/9, 5/10, 12/10, 19/10, 26/10,
amphi 1030-12 (Arago), TD 1330-1530, 1545-1745 (SI:30-34)
- **Books:**
 1. K. Mehlhorn & P. Sanders, *Algorithms and Data Structures*, Springer, 2008
 2. D. Knuth, *The Art of Computer Programming*, Addison-Wesley, 1997
 3. G. Dowek, *Les principes des langages de programmation*, Editions de l'X, 2008
 4. Ph. Baptiste & L. Maranget, *Programmation et Algorithmique*, Ecole Polytechnique (Polycopié), 2006
- **Website:** www.enseignement.polytechnique.fr/informatique/INF421
- **Blog:** inf421.wordpress.com
- **Contact:** liberti@lix.polytechnique.fr (e-mail subject: INF421)



Lecture summary

- Sorting complexity in general
- Mergesort
- Quicksort
- Two-way partition



The minimal knowledge



```
mergeSort( $s_1, \dots, s_n$ )  
   $m = \lfloor \frac{n}{2} \rfloor$ ;  
   $s' = \text{mergeSort}(s_1, \dots, s_m)$ ;  
   $s'' = \text{mergeSort}(s_{m+1}, \dots, s_n)$ ;  
  merge  $s', s''$  such that result  $\bar{s}$  is sorted;  
  return  $\bar{s}$ ;
```

Split in half, recurse on shorter subsequences, then do some work to reassemble them



```
quickSort( $s_1, \dots, s_n$ )  
  
  choose a  $k \leq n$ ;  
   $s' = (s_i \mid i \neq k \wedge s_i < s_k)$ ;  
   $s'' = (s_i \mid i \neq k \wedge s_i \geq s_k)$ ;  
  return (quickSort( $s'$ ),  $s_k$ , quickSort( $s''$ ));
```

Choose a value s_k , split s.t. left subseq. has values $< s_k$, right subseq. has values $\geq s_k$, recurse on subseq.



```
twoWaySort( $s_1, \dots, s_n$ )  $\in \{0, 1\}^n$   
  
   $i = 1; j = n$   
  while  $i \leq j$  do  
    if  $s_i = 0$  then  $i \leftarrow i + 1$   
    else if  $s_j = 1$  then  $j \leftarrow j - 1$   
    else swap  $s_i, s_j; i++; j--$  endif  
  end while
```

Only applies to binary sequences. Move i to leftmost 1 and j to rightmost 0. These are out of place, so swap them; continue until i, j meet



The sorting problem

- Consider the following problem:

SORTING PROBLEM (SP). Given a sequence $s = (s_1, \dots, s_n)$, find a permutation $\pi \in S_n$ of n symbols with the following property:

$$\forall 1 \leq i < j \leq n \ (s_{\pi(i)} \leq s_{\pi(j)}),$$

where S_n is the *symmetric group of order n*



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Type of s influences efficiency: the more generic, the less efficient. E.g. mergeSort and quickSort OK for all types; twoWaySort only OK for boolean



Problem complexity

- **Algorithmic complexity**: worst-case run time over all inputs

- **Problem complexity**: worst case run time of most efficient algorithm for problem

- Usually: upper bound on problem complexity

Given problem P , find an $O(f)$ algorithm, say complexity of P is no worse than $O(f)$

- Lower bounds?

Given problem P , show that no algorithm for P can ever be better than $\Omega(f)$

- Seems to require listing all possible algorithms for P

An ill-defined question?

Comparisons



Sorting algorithms are **comparison-based**

given s_i, s_j , does $s_i \leq s_j$ hold?

Comparisons



- Sorting algorithms are **comparison-based**

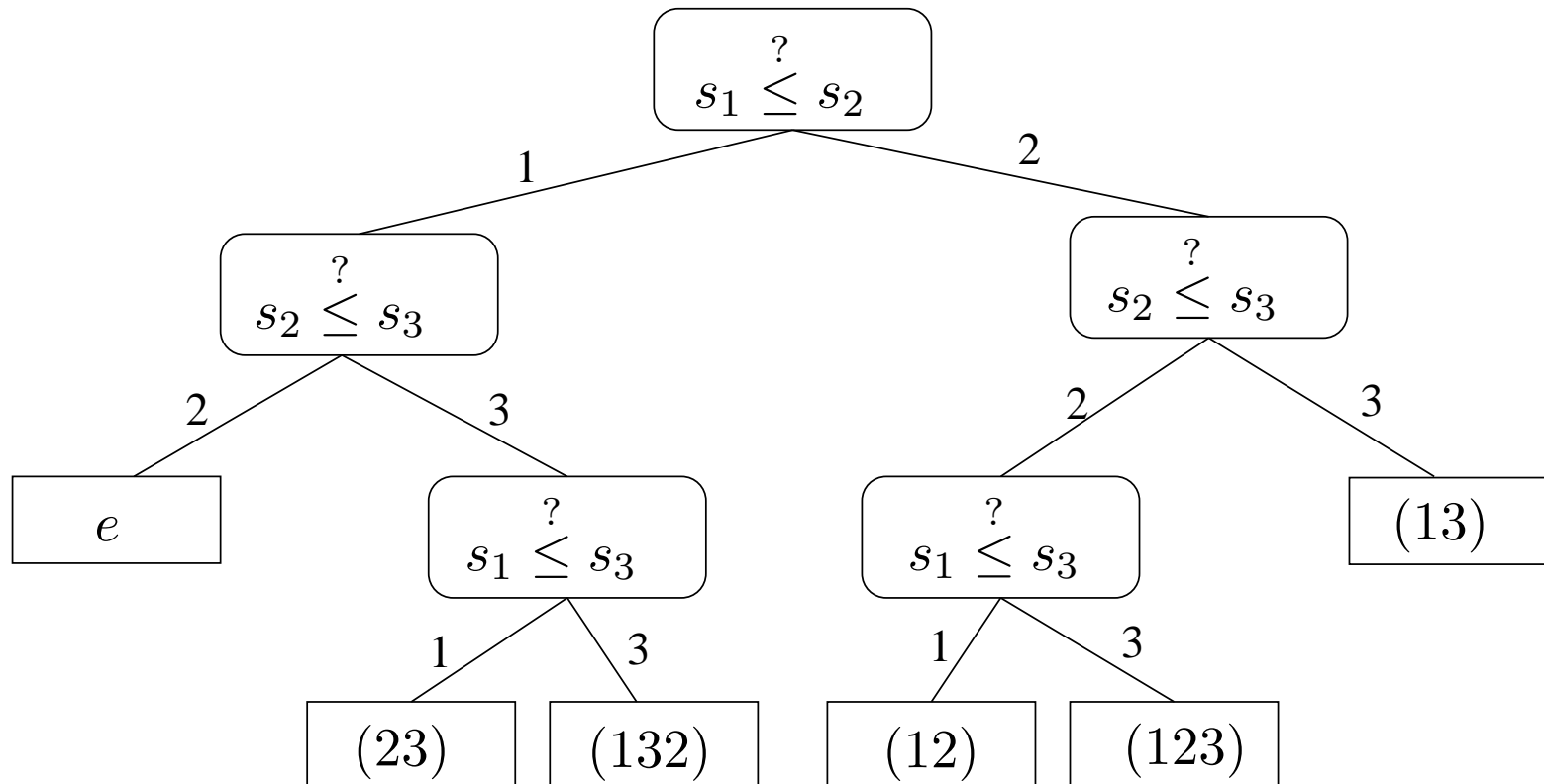
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- Describe *any* sorting algorithm by tracing calls to comparisons
sorting tree



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given s_i, s_j , does $s_i \leq s_j$ hold?
- Describe *any* sorting algorithm by tracing calls to comparisons
sorting tree
- E.g. sorting tree to order s_1, s_2, s_3 :



Sorting trees



● **Sorting tree:** different inputs \Rightarrow different paths *root* \rightarrow *leaf*

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- **Also:** best possible CB sorting algorithm = best possible sorting tree



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● \mathbb{T}_n = set of all sorting trees for n -sequences



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- For $T \in \mathbb{T}_n$ and $\pi \in S_n$,
 $\ell(T, \pi) = \text{length of path root} \rightarrow \text{leaf}(\pi) \text{ in } T$



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- For $T \in \mathbb{T}_n$ and $\pi \in S_n$,
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- $\ell(T, \pi)$: **trace length of a CB sorting algorithm on given input**
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- A binary tree T with depth $\leq k$ has $|V(T)| \leq 2^k$



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- By Stirling's approx., $\log n! = n \log n - \frac{1}{\ln 2}n + O(\log n)$ \square
- $\Rightarrow B_n$ is bounded below by a function proportional to $n \log n$
(we say B_n is $\Omega(n \log n)$)



Today's magic result: first part

Complexity of sorting:

$$\Omega(n \log n)$$



Simple sorting algorithms



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→ (4), (1, 2, 3)



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→ (1, 2, 3, 4)



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$(\boxed{3}, 1, 4, 2)$



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- Both are $O(n^2)$; insertion sort is fast for small $|s|$



Mergesort

Divide-and-conquer



• Let $s = (5, 3, 6, 2, 1, 9, 4, 3)$

Divide-and-conquer



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- Get $s' = (2, 3, 5, 6)$ and $s'' = (1, 3, 4, 9)$
- Merge s', s'' into a sorted sequence \bar{s} :

$$\begin{array}{l} (2,3,5,6) \\ (1,3,4, \boxed{9}) \end{array} \rightarrow (1, 2, 3, 3, 4, 5, 6)$$



Divide-and-conquer

- Let $s = (5, 3, 6, 2, 1, 9, 4, 3)$
- Split s midway: $s' = (5, 3, 6, 2)$, $s'' = (1, 9, 4, 3)$
- Sort s', s'' : $|s'| < |s|$ and $|s''| < |s| \Rightarrow$ use recursion
- **Base case:** *If $|s| \leq 1$ then s already sorted by definition*
- Get $s' = (2, 3, 5, 6)$ and $s'' = (1, 3, 4, 9)$
- Merge s', s'' into a sorted sequence \bar{s} :

$$\begin{pmatrix} 2,3,5,6 \\ 1,3,4,9 \end{pmatrix} \rightarrow (1, 2, 3, 3, 4, 5, 6, 9) = \bar{s}$$



Divide-and-conquer

- Let $s = (5, 3, 6, 2, 1, 9, 4, 3)$
- Split s midway: $s' = (5, 3, 6, 2)$, $s'' = (1, 9, 4, 3)$
- Sort s', s'' : $|s'| < |s|$ and $|s''| < |s| \Rightarrow$ use recursion
- **Base case:** *If $|s| \leq 1$ then s already sorted by definition*
- Get $s' = (2, 3, 5, 6)$ and $s'' = (1, 3, 4, 9)$
- Merge s', s'' into a sorted sequence \bar{s} :
$$\begin{pmatrix} 2,3,5,6 \\ 1,3,4,9 \end{pmatrix} \rightarrow (1, 2, 3, 3, 4, 5, 6, 9) = \bar{s}$$
- Return \bar{s}

Merge



• $\text{merge}(s', s'')$: merges two sorted sequences s', s'' in a sorted sequence containing all elements in s', s''

Merge



- $\text{merge}(s', s'')$: merges two sorted sequences s', s'' in a sorted sequence containing all elements in s', s''
- Since s', s'' are both already sorted, merging them so that the output is sorted is efficient
 - Read first (and smallest) elements of s', s'' : $O(1)$
 - Compare these two elements: $O(1)$
 - There are $|s|$ elements to process: $O(n)$

Merge



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- Since s', s'' are both already sorted, merging them so that the output is sorted is efficient
 - Read first (and smallest) elements of s', s'' : $O(1)$
 - Compare these two elements: $O(1)$
 - There are $|s|$ elements to process: $O(n)$
- You can implement this using lists: if s' is empty return s'' , if s'' is empty return s' , and otherwise compare the first elements of both and choose smallest

Recursive algorithm



```
● mergeSort(s) {  
  1: if  $|s| \leq 1$  then  
  2:   return s;  
  3: else  
  4:    $m = \lfloor \frac{|s|}{2} \rfloor$ ;  
  5:    $s' = \text{mergeSort}(e_1, \dots, e_m)$ ;  
  6:    $s'' = \text{mergeSort}(e_{m+1}, \dots, e_n)$ ;  
  7:   return merge( $s', s''$ );  
  8: end if  
}
```

By INF311, mergeSort has worst-case complexity
 $O(n \log n)$



Today's magic result: second part

Complexity of sorting:

$$\Theta(n \log n)$$

A function is $\Theta(g(n))$ if it is both $O(g(n))$ and $\Omega(g(n))$



Quicksort

Divide-and-conquer



• Let $s = (5, 3, 6, 2, 1, 9, 4, 3)$

Divide-and-conquer



- Let $s = (5, 3, 6, 2, 1, 9, 4, 3)$
- Choose a *pivot value* $p = s_1 = 5$ (no particular reason for choosing s_1)



Divide-and-conquer

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$$(5, \boxed{3}, 6, 2, 1, 9, 4, 3) \rightarrow \emptyset, \emptyset$$



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$$(5, 3, 6, 2, 1, 9, 4, 3) \rightarrow (3, 2), (6)$$

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- Sort $s' = (3, 2, 1, 4, 3)$ and $s'' = (6, 9)$: since $|s'| < |s|$ and $|s''| < |s|$ we can use recursion; base case $|s| \leq 1$



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- Update s to (s', p, s'')



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- Update s to (s', p, s'')

Notice: in `mergeSort`, we recurse *first*, then work on subsequences *afterwards*. In `quickSort`, we work on subsequences *first*, then recurse on them *afterwards*



Partition

- $\text{partition}(s)$: produces two subsequences s', s'' of (s_2, \dots, s_n) such that:
 - $s' = (s_i \mid i \neq 1 \wedge s_i < s_1)$
 - $s'' = (s_i \mid i \neq 1 \wedge s_i \geq s_1)$



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- Scan s : if $s_i < s_1$ put s_i in s' , otherwise put it in s''



Partition

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- Scan s : if $s_i < s_1$ put s_i in s' , otherwise put it in s''
- There are $|s| - 1$ elements to process: $O(n)$
- You can implement this using arrays; moreover, if you use a `swap` function such that, given i, j , swaps s_i with s_j in s , you don't even need to create any new temporary array: *you can update s "in place"*

Recursive algorithm



```
● quickSort(s) {  
  1: if  $|s| \leq 1$  then  
  2:   return ;  
  3: else  
  4:    $(s', s'') = \text{partition}(s)$ ;  
  5:   quickSort( $s'$ );  
  6:   quickSort( $s''$ );  
  7:    $s \leftarrow (s', s_1, s'')$ ;  
  8: end if  
}
```



Complexity

Worst-case complexity: $O(n^2)$

Average-case complexity: $O(n \log n)$

Very fast in practice

Worst-case complexity



- Consider the input $(n, n - 1, \dots, 1)$ with pivot s_1

Worst-case complexity



- Consider the input $(n, n - 1, \dots, 1)$ with pivot s_1
- Recursion level 1: $p = n$, $s' = (n - 1, \dots, 1)$, $s'' = \emptyset$

Worst-case complexity



- Consider the input $(n, n - 1, \dots, 1)$ with pivot s_1
- Recursion level 1: $p = n$, $s' = (n - 1, \dots, 1)$, $s'' = \emptyset$
- Recursion level 2: $p = n - 1$, $s' = (n - 2, \dots, 1)$, $s'' = \emptyset$

Worst-case complexity



- Consider the input $(n, n - 1, \dots, 1)$ with pivot s_1
- Recursion level 1: $p = n$, $s' = (n - 1, \dots, 1)$, $s'' = \emptyset$
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- And so on, down to $p = 1$ (base case)

Worst-case complexity



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- And so on, down to $p = 1$ (base case)
- Each partitioning call takes $O(n)$

Worst-case complexity



- Consider the input $(n, n - 1, \dots, 1)$ with pivot s_1
- Recursion level 1: $p = n$, $s' = (n - 1, \dots, 1)$, $s'' = \emptyset$
- Recursion level 2: $p = n - 1$, $s' = (n - 2, \dots, 1)$, $s'' = \emptyset$
- And so on, down to $p = 1$ (base case)
- Each partitioning call takes $O(n)$
- Get $O(n^2)$



2-Way partitioning

Definition by example



Input: $(1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$

Desired output: $(0, 0, 0, 0, 0, 1, 1, 1, 1, 1)$

Iterating swaps



• Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$

Iterating swaps



- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)

Iterating swaps



- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
- Swap them



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
- Swap them
- Increase leftmost counter, decrease rightmost counter



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
- Swap them
- Increase leftmost counter, decrease rightmost counter
- Repeat until counters become equal

$$(\boxed{1}, 0, 0, 1, 1, 0, 0, \boxed{0}, 1, 1)$$



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
- Swap them
- Increase leftmost counter, decrease rightmost counter
- Repeat until counters become equal

$(\mathbf{0}, 0, 0, 1, 1, 0, 0, \mathbf{1}, 1, 1)$



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
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- Increase leftmost counter, decrease rightmost counter
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$(0, 0, 0, \boxed{1}, 1, 0, \boxed{0}, 1, 1, 1)$



Iterating swaps

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- Find leftmost 1 and rightmost 0 (these are out of place)
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$(0, 0, 0, \mathbf{0}, 1, 0, \mathbf{1}, 1, 1, 1)$



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
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$(0, 0, 0, 0, \boxed{1}, \boxed{0}, 1, 1, 1, 1)$



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
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Iterating swaps

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$(0, 0, 0, 0, 0, 1, 1, 1, 1, 1)$



Iterating swaps

- Let $s = (1, 0, 0, 1, 1, 0, 0, 0, 1, 1)$
- Find leftmost 1 and rightmost 0 (these are out of place)
- Swap them
- Increase leftmost counter, decrease rightmost counter
- Repeat until counters become equal

$(1, 0, 0, 1, 1, 0, 0, 0, 1, 1) \rightarrow (0, 0, 0, 1, 1, 0, 0, 1, 1, 1) \rightarrow$
 $(0, 0, 0, 0, 1, 0, 1, 1, 1, 1) \rightarrow (0, 0, 0, 0, 0, 1, 1, 1, 1, 1)$



The algorithm

```
 $i = 0; j = n - 1;$   
while  $i < j$  do  
  if  $s_i = 0$  then  
     $i \leftarrow i + 1;$   
  else if  $s_j = 1$  then  
     $j \leftarrow j - 1;$   
  else  
     $\text{swap}(s, i, j);$   
     $i \leftarrow i + 1;$   
     $j \leftarrow j - 1;$   
  end if  
end while
```



Worst-case complexity

- Occurs with input $(1, \dots, 1, 0, \dots, 0)$ where number of 1's are around the same as the number of 0's
- Requires $\lfloor \frac{n}{2} \rfloor$ swaps
- Worst-case $O(n)$

A paradox?



- At the outset, we proved that sorting had complexity $\Theta(n \log n)$



A paradox?

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- Only apparent: the initial theorem was under the following assumptions:
 - no prior knowledge on the type of input (*“general input”*)
 - only **comparison-based algorithms** are concerned



A paradox?

- At the outset, we proved that sorting had complexity $\Theta(n \log n)$
- But 2-way partitioning requires only $O(n)$
- Contradiction? Paradox?
- Only apparent: the initial theorem was under the following assumptions:
 - no prior knowledge on the type of input (*“general input”*)
 - only **comparison-based algorithms** are concerned
- Neither assumption is true for 2-way partitioning
 - we know that the input sequence is of binary type
 - the algorithm never uses a comparison



Appendix

[P. Cameron, *Combinatorics*]



Quicksort: average complexity 1/10

- Let $n = |s|$
- Let q_n be the average number of comparisons made by quickSort to sort an n -sequence
- `partition(s)` involves $n - 1$ comparisons
- Assume the pivot $p = s_1$ is the k -th smallest element of s
- Then, recursion takes $q_{k-1} + q_{n-k}$ comparisons on average
- Average this over the n values that k can take
- This implies:

$$q_n = n - 1 + \frac{1}{n} \sum_{k=1}^n (q_{k-1} + q_{n-k}) \quad (1)$$



Quicksort: average complexity 2/10

- Notice that in the sum $\sum_{k=1}^n (q_{k-1} + q_{n-k})$, each q_k occurs twice

k	q_{k-1}	q_{n-k}
1	q_0	q_{n-1}
2	q_1	q_{n-2}
\vdots	\vdots	\vdots
$n - 1$	q_{n-2}	q_1
n	q_{n-1}	q_0

- Hence we can write:

$$q_n = n - 1 + \frac{2}{n} \sum_{k=0}^{n-1} q_k \quad (2)$$

Quicksort: average complexity 3/10

- Equation (2) is a *recurrence relation*
- A *solution* of a recurrence relation is a closed-form expression for q_n which does not include the symbol q_k for any integer $k \geq 0$
- One solution method consists in writing the solution as the infinite sequence $(q_0, q_1, q_2, \dots, q_n, \dots)$ as a *formal power series*:

$$Q(t) = \sum_{n \geq 0} q_n t^n \quad (3)$$

- If $Q(t)$ is known, then the value for each q_n can also be obtained:

Differentiate $Q(t)$ n times with respect to t , set $t = 0$, and divide the result by $n!$

Quicksort: average complexity 4/10

- Multiply each side of the recurrence relation (2) by nt^n and sum over all $n \geq 0$, get:

$$\sum_{n \geq 0} nq_n t^n = \sum_{n \geq 0} n(n-1)t^n + 2 \sum_{n \geq 0} \binom{n-1}{k=0} q_k t^n \quad (4)$$

- We now replace each of these three terms so as to be able to derive a more convenient expression for $Q(t)$



Quicksort: average complexity 5/10

- Differentiate $Q(t)$ with respect to t and multiply by t to get an expression for the first term:

$$t \frac{dQ(t)}{dt} = t \sum_{n \geq 0} n q_n t^{n-1} = \sum_{n \geq 0} n q_n t^n, \quad (5)$$

- For the second term: by lecture 1, $\sum_{n \geq 0} t^n = \frac{1}{1-t}$
- Differentiate this equation twice with respect to t , we get:

$$\sum_{n \geq 0} n(n-1)t^{n-2} = \frac{2}{(1-t)^3} \quad (6)$$

- Now multiply both members by t^2 to get an expression for the second term:

$$\sum_{n \geq 0} n(n-1)t^n = \frac{2t^2}{(1-t)^3} \quad (7)$$

Quicksort: average complexity 6/10

- Now for the third: the n -th term of the sum $\sum_{n \geq 0} (\sum_{k=0}^{n-1} q_k) t^n$ can be written as

$$\sum_{k=0}^{n-1} t^{n-k} (q_k t^k)$$

- Hence, the whole sum over n can be written as the following product (convince yourself that this is true):

$$(t + t^2 + t^3 + \dots)(q_0 + q_1 t + q_2 t^2 + q_3 t^3 + \dots)$$

- The first factor is $\sum_{n \geq 0} t^n = \frac{1}{1-t}$, and the second is simply the expression for $Q(t)$

- Hence, the third term is $\frac{2tQ(t)}{1-t}$



Quicksort: average complexity 7/10

- Putting it all together, we obtain a first-order differential equation for $Q(t)$:

$$tQ'(t) = \frac{2t^2}{(1-t)^3} + \frac{2t}{1-t}Q(t) \quad (8)$$

- Remark that if we differentiate the expression $(1-t)^2Q(t)$ (which I pulled out of a hat) w.r.t. t , we get:

$$\frac{d}{dt}((1-t)^2Q(t)) = (1-t)^2Q'(t) - 2(1-t)Q(t) \quad (9)$$

- We rearrange the terms of Eq. (8) to get:

$$tQ'(t) - \frac{2t}{1-t}Q(t) = \frac{2t^2}{(1-t)^3} \quad (10)$$

- We multiply Eq. (10) through by $\frac{(1-t)^2}{t}$ and get:

$$(1-t)^2Q'(t) - 2(1-t)Q(t) = \frac{2t}{1-t} \quad (11)$$



Quicksort: average complexity 8/10

- The RHS of Eq. (9) is the same as the LHS of Eq. (11), hence we can rewrite Eq. 9 as:

$$\frac{d}{dt}((1-t)^2 Q(t)) = \frac{2t}{1-t} \quad (12)$$

- Now, straightforward integration w.r.t. t yields:

$$Q(t) = \frac{-2(t + \log(1-t))}{(1-t)^2} \quad (13)$$

Quicksort: average complexity 9/10

- The next step consists in writing the power series for \log and $1/(1-t)^2$, rearrange them in a product, and read off the coefficient q_n of the term in t^n . Without going into details, this yields:

$$q_n = 2(n+1) \sum_{k=1}^n \frac{1}{k} - 4n \quad (14)$$

for all $n \geq 0$

- For all $n \geq 0$, the term $\sum_{k=1}^n \frac{1}{k}$ is an approximation of:

$$\int_1^n \frac{1}{x} dx = \log(n) + O(1) \quad (15)$$



Quicksort: average complexity 10/10

- Finally, we get an asymptotic expression for q_n :

$$\forall n \geq 0 \quad q_n = 2n \log(n) + O(n) \quad (16)$$

- This shows that the average number of comparisons taken by `quickSort` is $O(n \log n)$



End of Lecture 4