

Algorithms and Data Structures

Sorting: Merge Sort and Quick Sort

Ulf Leser

Content of this Lecture

- Merge Sort
- Quick Sort

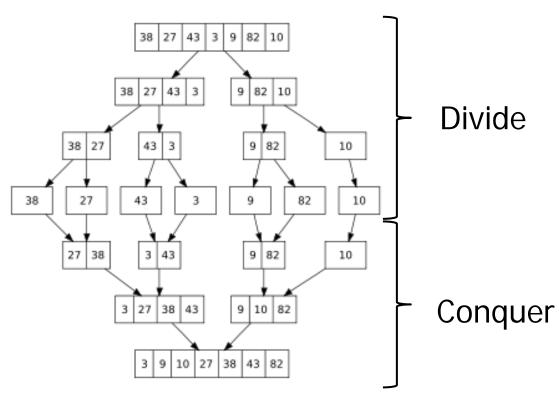
Central Idea for Improvement

- Methods we analyzed so-far did not optimally exploit transitivity of the "greater-or-equal" relationship
- If x≤y and y≤z, then x≤z
- If we compared x and y and y and z, there is no need any more to compare x and z
- The clue to lower complexities in sorting is finding ways to exploit such information

Merge Sort

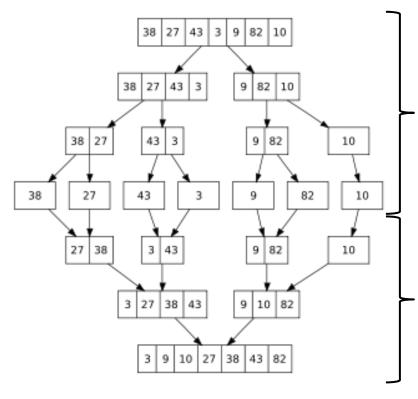
- Given the lower bound, we hope that we can do better
 - Not necessarily: The lower bound does not imply per-se that there is (and that we know) an algorithm which runs in this complexity
- Good news: There are various sort algorithms with O(n*log(n)) comparisons
- (Probably) Simplest one: Merge Sort
 - Divide-and-conquer algorithm
 - Break array in two partitions of equal size
 - Sort each partition recursively, if it has more than 1 elements
 - Merge sorted partitions
- Merge Sort is not in-place: Requires O(n) additional space

Illustration



Source: WikiPedia

Illustration



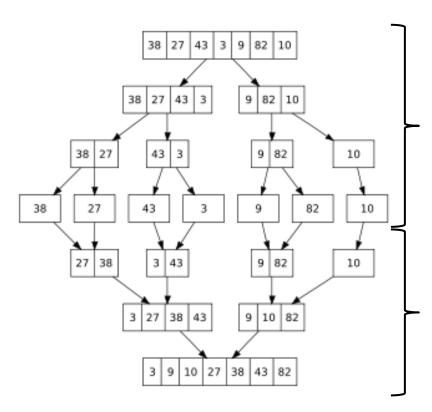
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Divide - Partition

Conquer - Merge

- Here we exploit transitivity
- We save comparisons during merge because both sub-lists are sorted

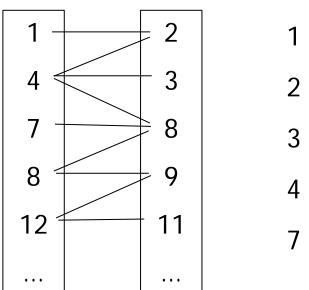
Algorithm



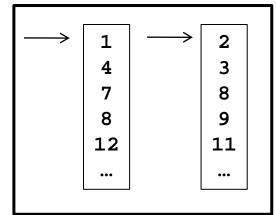
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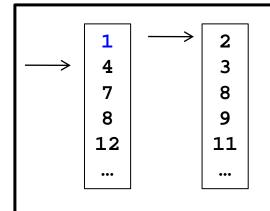
Merging Two Sorted Lists

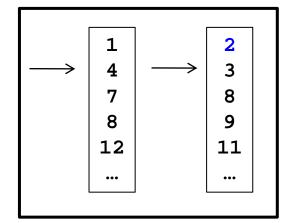
- We briefly looked at this problem before: Intersection of two sorted doc-lists in Information Retrieval
- Idea
 - Move one pointer through each list
 - Whatever element is smaller, copy to a new list and increment this pointer
 - "New list" requires additional space
 - Repeat until one list is exhausted
 - Copy rest of other list to new list

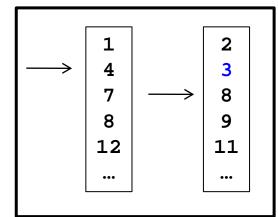


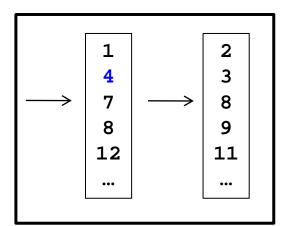
Example





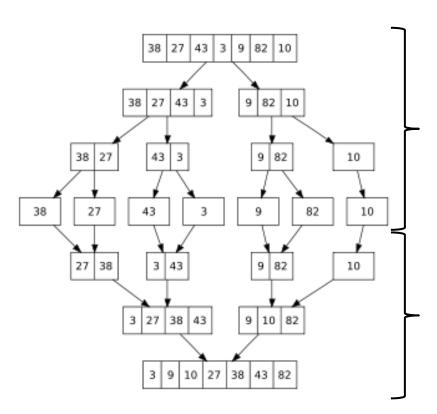






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Merge



Source: WikiPedia

```
function void merge(S array;
                     1,m,r integer) {
  B: array;
  i := 1;
                 # Start 1<sup>st</sup> list
                 # Start 2<sup>nd</sup> list
  i := m+1;
  k := 1;
                  # Target list
 while (i \le m) and (j \le r) do
    if S[i]<=S[j] then</pre>
      B[k] := S[i]; # From 1st list
      i := i+1;
    else
      B[k] := S[j]; \# From 2<sup>nd</sup> list
      j := j+1;
    end if;
    k := k+1;  # Next target
  end while;
  if i>m then # What remained?
    copy S[j..r] to B[k..k+r-j+1];
  else
    copy S[i..m] to B[k..k+m-i+1];
  end if:
    # Back to place
  copy B[1..k] to S[1..r]
```

Complexity Analysis

- Theorem
 Merge Sort requires Ω(n*log(n)) and O(n*log(n))
 comparisons
- Proof
 - Merging two sorted lists of size n requires O(n) comparisons
 - After every comp, 1 element is moved; there are only 2*n elements
 - Merge Sort calls Merge Sort twice with half of the array
 - Let T(n) be the number of comparisons
 - Thus: $T(n) = T(n/2) + T(n/2) + O(n) \sim 2*T(n/2) + n$
 - This is O(n*log(n))
 - See recursive solution of max subarray

Remarks

- Merge Sort is worst-case optimal: Even in the worst of all cases, it does not need more than (in the order of) the minimal number of comparisons
 - Given our lower bound for sorting
- But there are also disadvantages
 - O(n) additional space
 - Requires $\Omega(n^*\log(n))$ moves
 - Sorted sub-arrays get copied to new array in any case
 - See Ottmann/Widmayer for proof
- Note: Basis for sort algorithms on external memory

Summary

	Comparisons worst case	Comparisons best case	Additional space	Moves worst/best
Selection Sort	O(n ²)	O(n ²)	O(1)	O(n)
Insertion Sort	O(n ²)	O(n)	O(1)	O(n ²) / O(n)
Bubble Sort	O(n²)	O(n)	O(1)	O(n ²) / O(1)
Merge Sort	O(n*log(n))	O(n*log(n))	O(n)	O(n*log(n))

Content of this Lecture

- Merge Sort
- Quick Sort
 - Algorithm
 - Average Case Analysis
 - Improving Space Complexity

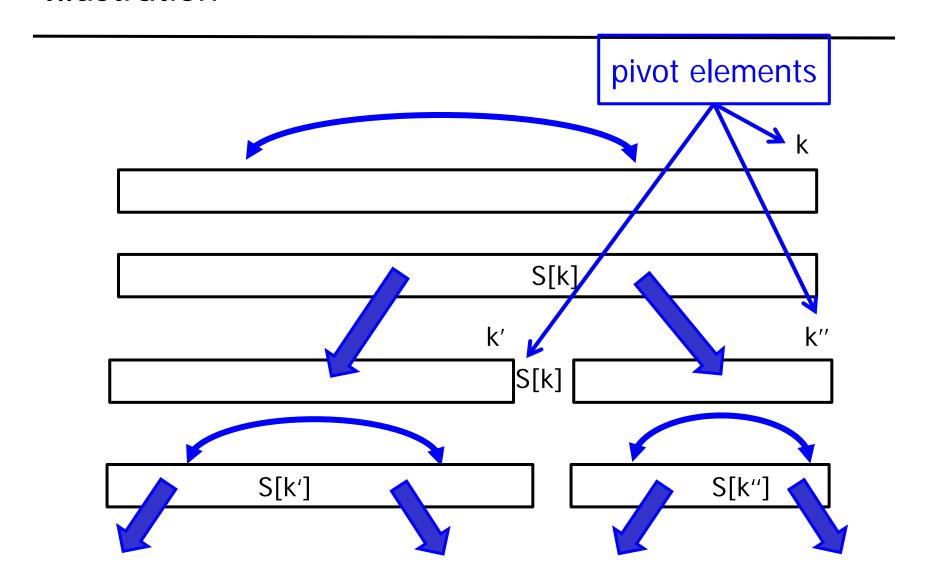
Comparison Merge Sort and Quick Sort

- What can we do better than Merge Sort?
 - The O(n) additional space is a problem
 - We need this space because the growing sorted runs have fixed sizes of up to 50% of |S| (2, 4, 8, ..., ceil(n/2))
 - We cannot easily merge two sorted lists in-place, because we have no clue how the numbers are distributed in the two lists
- Quick-sort uses a similar yet different way
 - We also recursively generate sort-of sorted runs
 - Whenever we create two such runs, we make sure that one contains only small and one contains only large values
 - Relative to a value that needs to be determined
 - This allows us to do a kind-of "merge" in-place

Main Idea

- Let k be an arbitrary index of S, 1≤k≤|S|
- Look at element S[k] (call it the pivot element)
- Modify S such that ∃i: ∀j≤i: S[j]≤S[k] and ∀l>i: S[k]≤S[l]
 - How? Wait a minute
 - S is broken in two subarrays S' and S"
 - S' with values smaller-or-equal than S[k]
 - S" with values larger-or-equal than S[k]
 - Note that afterwards S[k] is at its final position in the array
 - S' and S'' are smaller than S
 - But we don't know how much smaller depends on choice of k
- Treat S' and S" using the same method recursively
 - How often? Not clear depends on choice of k (again)

Illustration

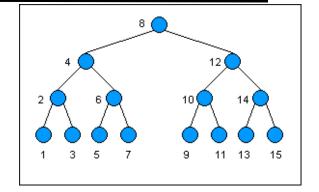


Quick Sort Framework

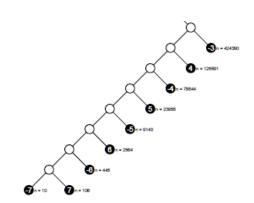
- Start with qsort(S, 1, |S|)
- 6: "Sort" S around the pivot element (divide)
 - Problem 1: Choose k
 - Problem 2: Do this in-place
- 7: Sort all values smaller-or equal than pivot element
- 8: Sort all values larger-orequal than pivot element
- Problem 3: How often do we need to do this?

Addressing P1 – approaching P3

- P1: We need to choose k (S[k])
- S[k] determines the sizes of S' and S"



- S[k] in the middle of the values of S
 - S' and S'' are of equal size $(\sim |S|/2)$
 - Creates a "nice" search tree
- S[k] at the border of the values of S
 - $|S'| \sim 0$ and $|S''| \sim |S| 1$ or vice versa
 - Creates a "bad" search tree
- Hint to P3: Somewhere in [log(n), n] times
 - Depending on choice of S[k]



Intermezzo: Mean and Median

- In statistics, one often tries to capture the essence of a (potentially large) set of values
- One essence: Mean
 - Average temperature per month, average income per year, average height of males at age of 18, average duration of study, ...
- Less sensible to outliers: Median
 - The middle value
 - Assume temps in June 25 24 24 23 25 25 24 4 -1 9 18 24
 - Which temperature do you expect for an average day in June?
 - Mean: 18.6
 - Median: 24 more realistic
 - How long will you need for your Bachelor? 6,35 semesters?

P1: Choosing k

- In the best case, S[k] is the median of S
- Approximations
 - If S is an array of people's income in Germany, we call the "Statistische Bundesamt" to ask for the mean of all incomes in Germany, and could scan the array until we find a value that is 10% or less different, and use this value as pivot
 - If S is large and randomly drawn from a set of incomes, this scan will be very short
 - If S is an array of family names in Berlin, we take the Berlin telephone book, open it roughly in the middle, and could scan the array until we find a value that is 10% or less different
- There is no exact and simple way to find the median of a large list of values (without sorting them)

P1: Choosing k - Again

- Option 1: Scan S to find min/max; search S[k]~(max-min)/2
 - Why should the values in S be equally distributed in this range?
 - For instance: Incomes are not equally distributed in their range
- Option 2: Choose a set of values X from S at random and determine S[k]~median(X)
 - X follows the same distribution (same median) as S, but |X| << |S|
 - Since this procedure would have to be performed for each qSort, only (too) small X do not influence runtime a lot
- More popular option 3: Choose k at random
 - For instance, simply use the last value in the array
 - Also relieves from searching an appropriate S[k]
 - We'll see that this already produces quite good result on average

Quick Sort Framework

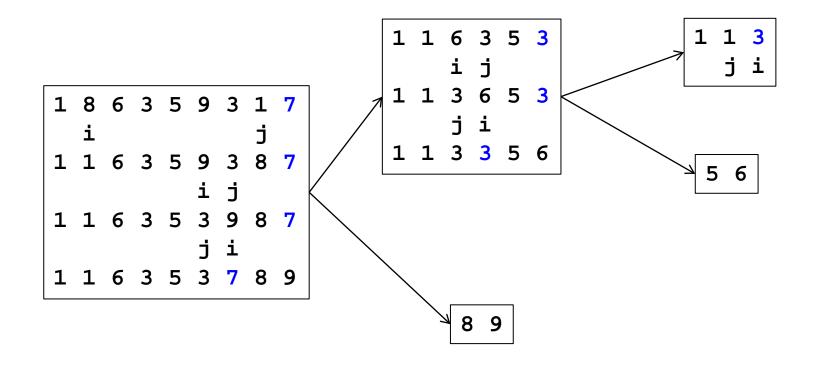
- Start with qsort(S, 1, |S|)
- 6: "Sort" S around the pivot element (divide)
 - Problem 1: Choose k
 - Problem 2: Do this in-place
- 7: Sort all values smaller-or equal than pivot element
- 8: Sort all values larger-orequal than pivot element
- Problem 3: How often do we need to do this?

P2: Do this in-place

- We use k=r
- Simple idea
 - Search from I towards r until a value greater-or-equal S[r]
 - Start from r towards I until a values smaller-or-equal S[r]
 - Swap values
 - Start again, if i has not yet reached j
 - When we have stopped, all values left from i are smaller than S[r], and all values right from j are larger than S[r] move S[r] right in the middle

```
1. func integer divide(S array;
2.
                        1,r integer) {
     val := S[r];
     i := 1-1;
     j := r;
     while true
7.
       repeat
8.
         i := i+1;
9.
       until S[i]>=val;
10.
       repeat
11.
         j := j-1;
12.
       until S[j]<=val or j<i;
13.
       if i>j then
14.
         break while;
15.
       end if;
16.
       swap( S[i], S[j]);
     end while;
17.
18.
     swap( S[i], S[r]);
19.
     return i;
20.}
```

Example



P2: Complexity

- # of comparisons: O(r-l)
 - Whenever we perform a comparison, either i or j are incremented / decremented
 - i starts from I, j starts from r, and the algorithm stops once they meet
 - This is worst, average and best case
- # of swaps: O(r-I) in worst case
 - Example: 8,7,8,6,1,3,2,3,5
 - Gives \sim (r-l)/2 swaps

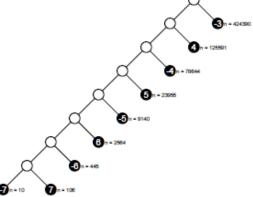
```
1. func integer divide(S array;
2.
                        1,r integer) {
     val := S[r];
4. i := 1-1;
5.
     j := r;
     while true
7.
       repeat
8.
         i := i+1;
9.
       until S[i]>=val;
10.
       repeat
11.
         j := j-1;
12.
       until S[j]<=val or j<i;
13.
       if i>j then
14.
         break while;
15.
       end if;
16.
       swap( S[i], S[j]);
     end while;
17.
18.
     swap( S[i], S[r]);
19.
     return i;
20.}
```

Worst-Case Complexity of Quick Sort

Worst case for number of comparisons:

A reverse-sorted list

- S[r] always is the smallest element
- Requires r-I comparisons in every call of divide()
- Every pair of qSort's has |S'|=0 and |S''|=n-1
- This gives $(n-1)+((n-1)-1)+...+1 = O(n^2)$



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Intermediate Summary

- Great disappointment
- We are in O(1) additional space, but as slow as our basic sorting algorithms
- But only in worst case
- Let's look at the average case

Average Case

- Without loss of generality, we assume that S contains all values 1...|S| in arbitrary order
 - If S had duplicates, we would at best save swaps (see code)
 - Sorting n different values is the same problem as sorting the values
 1...n replace each value by its rank
- For k, we choose any value in S with equal probability 1/n
- This choice divides S such that |S'|=k-1 and |S''|=n-k
- Let T(n) be the average # of comparisons. Then:

$$T(n) = \frac{1}{n} \sum_{k=1}^{n} \left(T(k-1) + T(n-k) \right) + bn = \frac{2}{n} \sum_{k=1}^{n-1} T(k) + bn$$

- Where bn is the time to divide the array and T(0)=0

Induction

We need to show that, for some c independent of n:

$$T(n) \le c * n * \log(n)$$

- We proof by induction (for n≥2)
 - Clearly, $T(1)=0 \le 1*log(1)$
 - We assume that the assumption holds for all i<n
 - Then

$$T(n) = \frac{2}{n} \sum_{k=1}^{n-1} T(k) + bn$$

$$\leq \frac{2c}{n} \sum_{k=1}^{n-1} k * \log(k) + bn$$

$$= \frac{2c}{n} \left[\sum_{k=1}^{n/2} k * \log(k) + \sum_{k=1}^{n/2-1} \left(\frac{n}{2} + k \right) * \log\left(\frac{n}{2} + k \right) \right] + bn$$

Continued

 $log(k) \le log(n)$

$$T(n) \le \frac{2c}{n} \left[\sum_{k=1}^{n/2} k * \log(k) + \sum_{k=1}^{n/2-1} \left(\frac{n}{2} + k \right) * \log\left(\frac{n}{2} + k \right) \right] + bn$$

$$\le \frac{2c}{n} \left[\sum_{k=1}^{n/2} k * \log(n) + \sum_{k=1}^{n/2-1} \left(\frac{n}{2} + k \right) * \log(n) \right] + bn$$

$$= \frac{2c}{n} \left[\left(\frac{n^2}{2} - \frac{n}{2} \right) * \log(n) - \frac{n^2}{8} - \frac{n}{4} \right] + bn$$

$$= c * n * \log(n) - c * \log(n) - \frac{cn}{4} - \frac{c}{2} + bn$$

$$\le c * n * \log(n) - cn/4 - c/2 + bn$$

$$\le c * n * \log(n)$$

Set c≥4b

Conclusion

- Although there are cases where we need O(n²)
 comparisons, these are so rare in the set of all possible
 permutations that we do not need more than O(n*log(n))
 comparisons on average
- In other words: If we sum the runtimes of Quick Sort over many (all) different orders of n values (for different n), then this sum will grow with n*log(n), not with n²
- One can show the same for the # of swaps
- Quick Sort is a fast general-purpose sorting algorithm

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 - Improving Space Complexity

Looking at Space Again

- We were quite sloppy
- Quick Sort does need extra space every recursive call puts some data on the stack
 - Array can be passed by reference or declared as a global variable
 - But we need to pass I and r
- Our current version has worst-case space complexity O(n)
 - Consider the worst-case of the time complexity
 - Reverse-sorted array
 - Creates 2*n recursive calls
 - This requires n times 2 integers on the stack

Improving Space Complexity

- In the recursive decent, always treat the smaller of the two sub-arrays first (S' or S", whatever is smaller)
- This branch of the search tree can generate at most O(log(n)) calls, as the smaller array always is smaller than |S|/2 (or it would not be the smaller one)
- Use iteration (no stack) to sort the bigger array afterwards
- Space complexity: O(log(n))

Implementation

```
    func integer qSort(S array;

2.
                       1,r int) {
     if r≤l then
       return:
   end if;
   val := S[r];
     i := 1-1;
     j := r;
     while true
10.
       repeat
11.
         i := i+1;
12.
      until S[i]>val;
13.
       repeat
         j := j-1;
14.
15.
      until S[j]<val or j<i;
16.
       if i>j then
         break while:
17.
18.
       end if;
19.
       swap( S[i], S[j]);
     end while;
20.
21.
     swap( S[i], S[r]);
22.
     qsort(S, 1, i-1);
23.
     qSort(S, i+1, r);
24.}
```

```
    func integer qSort++(S array;

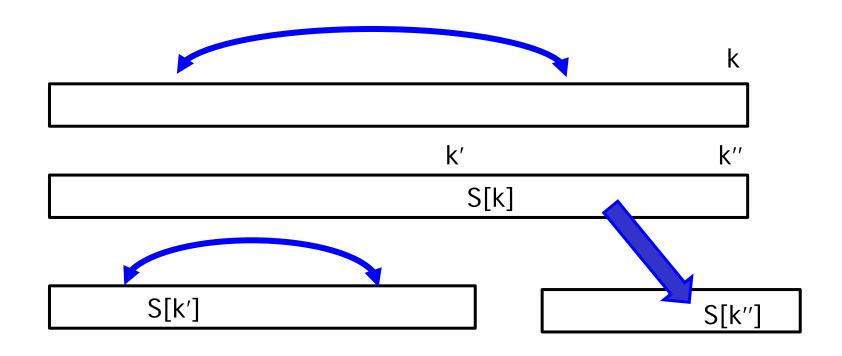
2.
                         1,r int) {
     if r≤l then
       return;
   end if;
     while r > 1 do
7.
       val := S[r];
       i := 1-1;
9.
       j := r;
       while true
10.
                 # as before
11.
12.
       end while;
13.
       swap( S[i], S[r]);
14.
       if(i-1-1) < (r-i-1) then
15.
         qsort(S, 1, i-1);
16.
         1 := i+1;
       else
17.
18.
       qSort(S, i+1, r);
19.
         r := i-1;
       end if;
20.
21.
     end while;
22.}
```

Implementation

- 14-20: Choose the smaller and sort it recursively
 - Note: Only one call is made for each division
- We adjust I/r and continue to sort the larger sub-array
 - New loop (6-21) applies the same procedure performing the next sort
- We turned a linear tail recursion into an iteration without stack

```
func integer qSort++(S array;
2.
                         1,r int) {
     if r≤l then
       return;
     end if:
     while r > 1 do
7.
       val := S[r];
       i := 1-1;
       j := r;
10.
       while true
11.
                  # as before
12.
       end while;
13.
       swap( S[i], S[r]);
14.
       if (i-1-1) < (r-i-1) then
15.
         qsort(S, 1, i-1);
16.
         1 := i+1;
       else
17.
18.
         qSort(S, i+1, r);
19.
         r := i-1;
20.
       end if:
21.
     end while;
22.}
```

Illustration



Improving Space Complexity Further

- Even O(1) space is possible
 - Do not store I/r, but search them at runtime within the array
 - Requires extra work in terms of runtime, but within the same complexity
 - See Ottmann/Widmayer for details
 - Is it worth it in practice?
 - Log(n) usually is not a lot of space

Summary

	Comps worst case	avg. case	best case	Additional space	Moves (wc / ac)
Selection Sort	O(n ²)		O(n ²)	O(1)	O(n)
Insertion Sort	O(n²)		O(n)	O(1)	O(n ²)
Bubble Sort	O(n²)		O(n)	O(1)	O(n²)
Merge Sort	O(n*log(n))		O(n*log(n))	O(n)	O(n*log(n))
QuickSort	O(n ²)	O(n*log(n)	O(n*log(n)	O(log(n))	O(n ²) / O(n*log(n))