Algorithms and Data Structures

Sorting:
Merge Sort and Quick Sort

Marius Kloft
Content of this Lecture

- Merge Sort
- Quick Sort
Central Idea for Improvements in Sorting

- Methods we analyzed so-far did not optimally exploit *transitivity of the „greater-or-equal“ relationship*
- If $x \leq y$ and $y \leq z$, then $x \leq z$
- 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 complexity algorithms for sorting** is finding ways to exploit such information
Merge Sort

- There are various algorithms with $O(n \cdot \log(n))$ comparisons
  - I.e., optimal
- (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 element
  - Merge sorted partitions
- Merge Sort is not in-place: $O(n)$ additional space
Illustration

Source: WikiPedia
Illustration

- **Divide - Partition**
- **Conquer - Merge**
  - Here we *exploit transitivity*
  - We save comparisons during merge because both sub-lists are sorted

Source: WikiPedia
function void mergesort(S array; l,r integer) {
    if (l<r) then
        #Sort each ~50% of array
        m := (r-l) div 2;
        mergesort(S, l, l+m);
        mergesort(S, l+m+1, r);
        #merges two sorted lists
        merge(S, l, l+m, r);
    else
        # Nothing to do, 1-element list
    end if;
}
Merging Two Sorted Lists

- Recall: Intersection of two sorted doc-lists in IR
- 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
function void merge(S array;
    l,m,r integer) {
    B: array[l..r-l+1];
    i := l;               # Start of 1st list
    j := m+1;             # Start of 2nd list
    k := 1;               # Target list

    while (i<=m) and (j<=r) do
        if S[i]<=S[j] then
            B[k] := S[i];       # From 1st list
            i := i+1;
        else
            B[k] := S[j];       # From 2nd 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];
    else
        copy S[i..m] to B[k..k+m-i];
    end if;

    # Back to original list
    copy B[1..r-l+1] to S[l..r];
}
Complexity

- **Theorem**
  
  Merge Sort requires \( \Omega(n \times \log(n)) \) and \( O(n \times \log(n)) \) comparisons

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

- \( \Omega(n \times \log(n)) \): # comparisons does not depend on data in S
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 $\Omega(n \cdot \log(n))$ moves
    - Sorted sub-arrays get copied to new array in any case
    - See Ottmann/Widmayer for proof
- Note: Basis for sorting algorithms on **external memory**
## Summary

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<td>$O(n^2)$</td>
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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, \(1 \leq k \leq |S|\)
- Look at element S[k] (call it the pivot element)
- Modify S such that \(\exists i: \forall j \leq i: S[j] \leq S[k]\) and \(\forall l > i: S[k] \leq S[l]\)
  - How? Wait a minute
  - S is broken in two subarrays S’ and S”
    - S’ with values smaller-or-equal than pivot element
    - S” with values larger-or-equal than pivot element
    - 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)
A Bad Case
Quick Sort Framework

- Start with qsort(S, 1, |S|)
- “Sort” S around the pivot element (divide)
  - Problem 1: Choose k
  - Problem 2: Do this in-place
- Recursively sort values smaller-or equal than pivot element
- Recursively sort values larger-or-equal than pivot element
- Problem 3: How often do we need to do this?

```c
1. func void qsort(S array;
2.     l,r integer) {
3.     if r≤l then
4.         return;
5.     end if;
6.     pos := divide( S, l, r);
7.     qsort( S, l, pos-1);
8.     qsort( S, pos+1, r);
9. }
```
Addressing Problem P1 – approaching P3

• P1: We need to choose k (S[k])
• S[k] determines the sizes of S’ and S”

• Best: S[k] in the middle of the values of S
  – S’ and S” are of equal size (\(\sim |S|/2\))
  – Creates a low search tree

• Worst: S[k] at the border of the values of
  – |S’|\(\sim 0\) and |S”|\(\sim |S|-1\) or vice versa
  – Creates a deep 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 sensitive 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 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, and open it roughly in the middle

- 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:** Find min/max in \( S \); search \( S[k] \approx (\text{min}+\text{max})/2 \)
  - Why should the values in \( S \) be **equally distributed in this range**?
  - For instance: Incomes are not equally distributed at all
- **Option 2:** Choose a (small) set of values \( X \) from \( S \) at random and determine \( S[k] \approx \text{median}(X) \)
  - \( X \) follows the same distribution (same median) as \( S \), but \( |X| < |S| \)
  - Since this procedure would have to be performed for each qSort, only small \( X \) do not influence runtime a lot
  - But: Small \( X \) will lead to bad median estimations
- **Option 3:** Choose \( k \) at random
  - For instance, simply use the last value in the array
  - We’ll see that this already produces **good results on average**
Quick Sort Framework

- Start with $qsort(S, 1, |S|)$
- “Sort” $S$ around the pivot element ($divide$)
  - Problem 1: Choose $k$
  - Problem 2: Do this in-place
- Recursively sort values smaller-or equal than pivot element
- Recursively sort values larger-or-equal than pivot element
- Problem 3: How often do we need to do this?

```c
1. func void qsort(S array;
2.    l,r integer) {
3.    if r≤l then
4.       return;
5.    end if;
6.    pos := divide( S, l, r);
7.    qsort( S, l, pos-1);
8.    qsort( S, pos+1, r);
9. }
```
Problem P2: Do this in-place

- We use \( k=r \)
- Simple idea
  - Search from \( l \) towards \( r \) until first value greater-or-equal \( S[r] \)
  - Search from \( r \) towards \( l \) until first value smaller-or-equal \( S[r] \)
  - Swap these two values
  - Repeat if \( i \) has not reached \( j \) yet
  - Result: Values left from \( i \) are smaller than \( S[r] \) and values right from \( j \) are larger than \( S[r] \)
  - Move \( S[r] \) into the middle

```c
1. func int divide(S array;  
2. l,r integer) {  
3. val := S[r];  
4. i := l;  
5. j := r-1;  
6. repeat  
7. while (S[i]<=val and i<r)  
8. i := i+1;  
9. end while;  
10. while (S[j]>=val and j>l)  
11. j := j-1;  
12. end while;  
13. if i<j then  
14. swap( S[i], S[j] );  
15. end if;  
16. until i>=j;  
17. swap( S[i], S[r] );  
18. return i;  
19. }
```
Example
P2: Complexity

- **# of comparisons: O(r-l)**
  - Whenever we perform a comparison, either i or j are incremented / decremented
  - i starts from l, j starts from r, and the algorithm stops once they meet
  - This is **worst, average and best case**

- **# of swaps: O(r-l)** in worst case
  - Example: 8,7,8,6,1,3,2,3,5
  - Requires \( \sim (r-l)/2 \) swaps

```plaintext
1. func int divide(S array; 
2. l,r integer) { 
3. val := S[r]; 
4. i := l; 
5. j := r-1; 
6. repeat 
7. while (S[i]<=val and i<r) 
8. i := i+1; 
9. end while; 
10. while (S[j]>=val and j>1) 
11. j := j-1; 
12. end while; 
13. if i<j then 
14. swap( S[i], S[j] ); 
15. end if; 
16. until i>=j; 
17. swap( S[i], S[r] ); 
18. return i; 
19. }
```
Worst-Case Complexity of Quick Sort

- **Worst case:** A **reverse-sorted** list
  - $S[r]$ in first iteration is the smallest element, later always the smallest or the largest
  - Requires $r-l$ comparisons in every call of $\text{divide()}$
  - Every pair of qSort’s has $|S'|=0$ and $|S''|=n-1$
  - This gives $(n-1)+((n-1)-1)+...+1 = O(n^2)$
Intermediate Summary

• Great disappointment
• We are in O(1) additional space, but as slow as our basic sorting algorithms in worst case
• Let’s look at the average case
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  - Average Case Analysis
  - Improving Space Complexity
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
  - 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:

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

- Where \( b*n \) is the time to divide the array and \( T(0)=0 \)
Induction

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

$$T(n) \leq c \cdot n \cdot \log(n)$$

• Proof by induction (for $n \geq 2$)
  – Clearly, $T(1) = b$, $T(2) = 3b \leq c \cdot 2 \cdot \log(2)$ if $c \geq 3b/2$
  – We assume the above assumption holds for all $2 \leq k < n$
  – We start with (for simplicity, assume $n = 2^x$ for some $x$):

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

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

\[ T(n) \leq c \cdot n \cdot \log(n) \]
\[ \leq \frac{2c}{n} \sum_{k=1}^{n-1} k \cdot \log(k) + bn + b \]
\[ = \frac{2c}{n} \left[ \sum_{k=1}^{\lfloor n/2 \rfloor} k \cdot \log(k) + \sum_{k=\lfloor n/2 \rfloor + 1}^{n-1} k \cdot \log(k) \right] + bn + b \]
Continued

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

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

\[ = \frac{2c}{n} \left[ \sum_{k=1}^{n/2} k \log(n) - n^2/8 - n/4 + \sum_{k=n/2+1}^{n-1} k \log(n) \right] + bn + b \]

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

\[ = c \cdot n \cdot \log(n) - c \cdot \log(n) - \frac{cn}{4} - \frac{c}{2} + bn + b \]

\[ \leq c \cdot n \cdot \log(n) - cn/4 - c/2 + bn + b \]

\[ \leq c \cdot n \cdot \log(n) \]

Set \( c \geq 4b \)
Conclusion

- Although there are cases where we need $O(n^2)$ 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 average over the runtimes of Quick Sort over many (all) different orders of $n$ values (for different $n$), then this average will grow with $n \log(n)$, not with $n^2$.
- One can show the same for the # of swaps.
- Quick Sort is a fast general-purpose sorting algorithm.
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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 l 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 recursion, 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

1. func integer qSort(S array;  
2. l,r int) {
3. if r≤l then
4. return;
5. end if;
6. val := S[r];
7. i := l;
8. j := r-1;
9. repeat
10. while (S[i]<val and i<r)  
11. i := i+1;
12. end while;
13. while (S[j]<val and j>l)  
14. j := j-1;
15. end while;
16. if i<j then
17. swap( S[i], S[j]);
18. end if;
19. until i>=j;
20. swap( S[i], S[r]);
21. qsort(S, l, i-1);
22. qSort(S, i+1, r);
23.}

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

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

```plaintext
1. func integer qSort++(S array; l,r int) {
2.   if r≤1 then
3.     return;
4.   end if;
5.   while r > l do
6.     val := S[r];
7.     i := l;
8.     j := r-1;
9.     repeat
10.    ... # as before
11.     until i>=j;
12.     swap( S[i], S[r]);
13.   if (i-1) < (r-i) then
14.     qsort(S, l, i-1);
15.     l := i+1;
16.   else
17.     qSort(S, i+1, r);
18.     r := i-1;
19.   end if;
20. end while;
21. }
```
Improving Space Complexity Further

- Even $O(1)$ space is possible
  - Do not store $l/r$, 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

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<tr>
<th></th>
<th>Comps worst case</th>
<th>avg. case</th>
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<td>$O(\log(n))$</td>
<td>$O(n^2) / O(n\times\log(n))$</td>
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Exemplary Questions

- Proof that any sort algorithm using only value comparisons needs $\Omega(n \times \log(n))$ comparisons in worst case
- Proof or refute: For every $n$, there exists a list with $n$ elements which is a best case for quick sort (choosing first element as pivot) and for bubble sort
- Give pseudo code for QuickSort with $O(\log(n))$ additional space
- Imagine your main memory can use only $n/16$ values. Recall that access disk is much more expensive than accessing memory. Which of the sorting algorithms can be used to keep disk IO low? Describe the algorithm in pseudo code and argue about the number of blocks read