

Algorithms and Data Structures

A7. Runtime Analysis: Bottom-Up Merge Sort

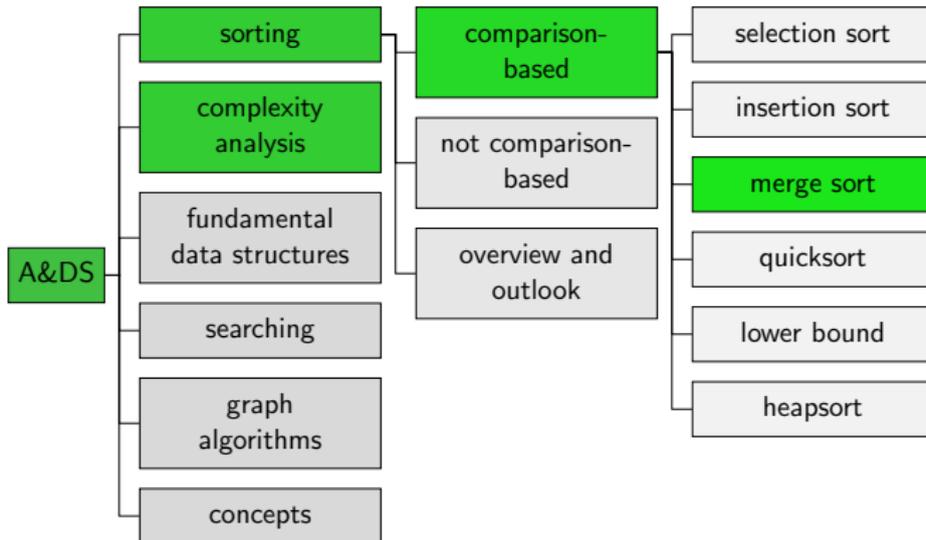
Gabriele Röger and Patrick Schneider

University of Basel

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Runtime Analysis: Bottom-Up Merge Sort

Content of the Course



Merge Step

```
1 def merge(array, tmp, lo, mid, hi):
2     i = lo
3     j = mid + 1
4     for k in range(lo, hi + 1): # k = lo, ..., hi
5         if j > hi or (i <= mid and array[i] <= array[j]):
6             tmp[k] = array[i]
7             i += 1
8         else:
9             tmp[k] = array[j]
10            j += 1
11    for k in range(lo, hi + 1): # k = lo, ..., hi
12        array[k] = tmp[k]
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We analyze the running time for $m := hi - lo + 1$
(number of elements that should be merged).

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Theorem

The merge step has *linear running time*, i.e., there are constants $c, c', n_0 > 0$ such that for all $n \geq n_0$: $cn \leq T(n) \leq c'n$.

Bottom-Up Merge Sort

```
1 def sort(array):
2     n = len(array)
3     tmp = list(array)
4     length = 1
5     while length < n:
6         lo = 0
7         while lo < n - length:
8             mid = lo + length - 1
9             hi = min(lo + 2 * length - 1, n - 1)
10            merge(array, tmp, lo, mid, hi)
11            lo += 2 * length
12            length *= 2
```

We use the following constants in the analysis:

c_1 lines 2–4

c_2 lines 6 and 12

c_3 lines 8,9,11

Assumption: merge requires

c_4 (hi-lo+1) operations.

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Total $T(n) \leq c_1 + \ell(c_2 + c_3n + c_4n) \leq \ell(c_1 + c_2 + c_3 + c_4)n$

Bottom-Up Merge Sort: Analysis II

What is the value of ℓ ?

- In iteration i we have $m = 2^i$ for the merge step.
- In iteration ℓ we have $m = 2^\ell = n$ for the merge step.
- Since $n = 2^k$ we have $\ell = k = \log_2 n$.

With $c := c_1 + c_2 + c_3 + c_4$ we get $T(n) \leq cn \log_2 n$.

Bottom-Up Merge Sort: Analysis III

What if n is not a power of two, so $2^{k-1} < n < 2^k$?

- Nevertheless k iterations of the outer loop.
- Inner loop does not perform more operations.
- $T(n) \leq cnk = cn(\lfloor \log_2 n \rfloor + 1) \leq 2cn \log_2 n$ (for $k > 2$)

Bottom-Up Merge Sort: Analysis IV

Analogous argument possible for lower bound.

→ Exercises

Theorem

*Bottom-up merge sort has **linearithmic running time**, i.e. there are constants $c, c', n_0 > 0$, such that for all $n \geq n_0$:*

$$cn \log_2 n \leq T(n) \leq c' n \log_2 n.$$

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Running time $n \log_2 n$ not much worse than linear running time

Merge Sort with Cost Model I

Key comparisons

- Only in merge.
 - Merging two ranges of length m and n requires in the best case $\min(n, m)$ and in the worst case $n + m - 1$ comparisons.
 - With two ranges of roughly equal length, this is a **linear** number of comparisons, i.e., there are $c, c' > 0$ such that the number of comparisons is between cn and $c'n$.
- Number of key comparisons that is performed for sorting the entire input sequence is **linearithmic** in the length of the sequence (analogously to the runtime analysis).

Merge Sort with Cost Model II

Movements of elements

- Only in merge.
- $2n$ movements for sequence of length n .
- Total for merge sort **linearithmic** (analogously to key comparisons).

Summary

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- Merge sort has linearithmic running time, key comparisons and movements of elements.