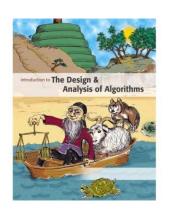




Introduction to

Algorithm Design and Analysis

[6] MergeSort



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In the Last Class...

Heap

- o Partial order property
 - FixHeap
 - ConstructHeap
- o Heap structure
 - Array-based implementation

HeapSort

- o Complexity
- o Accelerated HeapSort



MergeSort

- MergeSort
 - o Worst-case analysis of MergeSort
- Lower Bounds for comparison-based sorting
 - o Worst-case
 - o Average-case

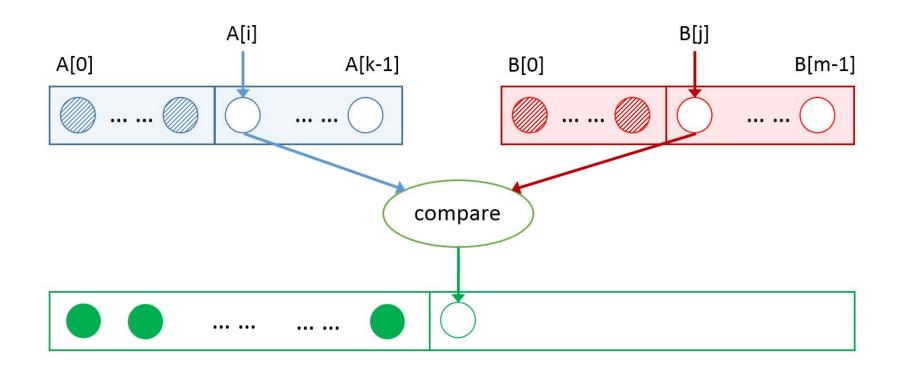
MergeSort: the Strategy

Easy division

- No comparison is conducted during the division
- Minimizing the size difference between the divided subproblems
- Merging two sorted subranges
 - o Using Merge



Merging Sorted Arrays





Merge: the Specification

Input

Array A with k elements and B with m elements,
 whose keys are in non-decreasing order

Output

- o Array C containing n = k + m elements from A and B in non-decreasing order
- o C is passed in and the algorithm fills it



Merge: Recursive Version

```
merge(A,B,C)
                                                 Base cases
  if (A is empty)
     rest of C = rest of B
  else if (B is empty)
     rest of C = rest of A
  else
     if (first of A \leq first of B)
       first of C = first of A
       merge(rest of A, B, rest of C)
     else
       first of C = first of B
       merge(A, rest of B, rest of C)
  return
```



Worst Case Complexity of Merge

Observations

- o Worst case is that the last comparison is conducted between A[*k*-1] and B[*m*-1]
 - After each comparison, one element is inserted into Array C, *at least*.
 - After entering Array C, an element will never be compared again
 - After the last comparison, at least two elements (the two just compared) have not yet been moved to Array
 C. So at most *n*-1 comparisons are done.
- In worst case, n-1 comparisons are done, where n=k+m



Optimality of Merge

- Any algorithm to merge two sorted arrays, each containing k=m=n/2 entries, by comparison of keys, does at least n-1 comparisons in the worst case.
 - o Choose keys so that:

$$b_0 < a_0 < b_1 < a_1 < \dots < b_i < a_i < b_{i+1}, \dots, < b_{m-1} < a_{k-1}$$

o Then the algorithm must compare a_i with b_i for every i in [0, m-1], and must compare a_i with b_{i+1} for every i in [0, m-2], so, there are n-1 comparisons.

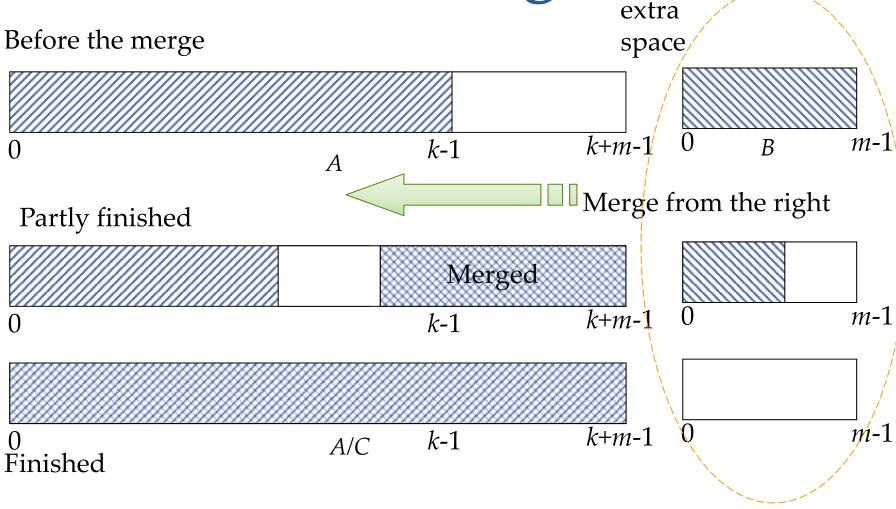
Valid for $|k-m| \le 1$, as well.

Space Complexity of Merge

- A algorithm is "in space", if the extra space it has to use is in $\Theta(1)$
- Merge *is not* a algorithm "in space", since it need enough extra space to store the merged sequence during the merging process.



Overlapping Arrays for Merge





MergeSort

- Input: Array E and indexes first, and last, such that the elements of E[i] are defined for $first \le i \le last$.
- Output: E[first],...,E[last] is a sorted rearrangement of the same elements.
- Procedure

```
void mergeSort(Element[] E, int first, int last)
if (first<last)
  int mid=(first+last)/2;
  mergeSort(E, first, mid);
  mergeSort(E, mid+1, last);
  merge(E, first, mid, last)
return</pre>
```



Analysis of MergeSort

• The recurrence equation for Mergesort

$$\circ$$
 W(n)=W($\lfloor n/2 \rfloor$)+W($\lceil n/2 \rceil$)+n-1

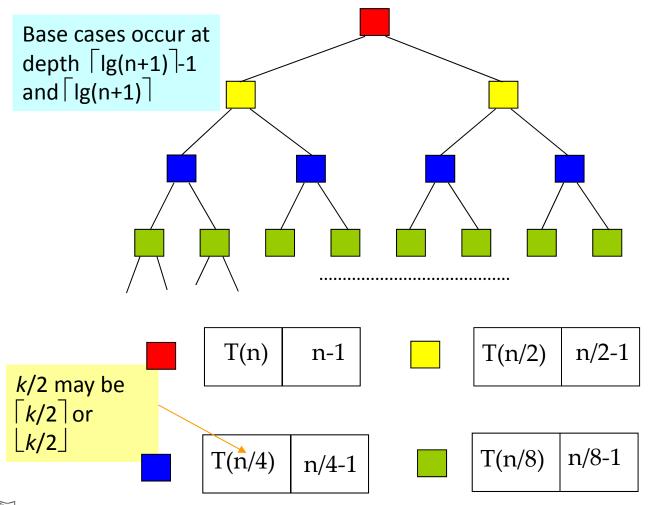
 $\circ W(1)=0$

Where *n*=last-first+1, the size of range to be sorted

• The *Master Theorem* applies for the equation, so:

 $W(n) \in \Theta(n \log n)$

Recursion Tree for Mergesort



n-1 Level 0

n-2 Level 1

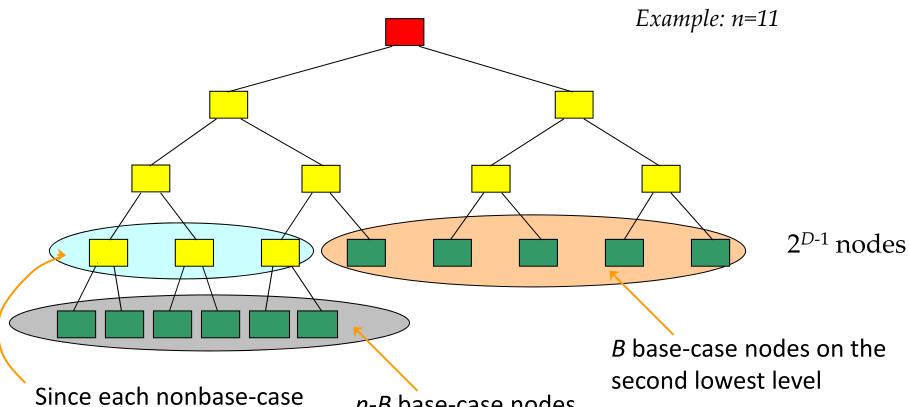
n-4 Level 2

n-8 Level 3

Note:

nonrecursive costs on level k is $n-2^k$ for all level without basecase node

Non-complete Recursion Tree



Since each nonbase-case node has 2 children, there are (*n-B*)/2 nonbase-case nodes at depth *D*-1

n-B base-case nodesNo nonbase-casenodesat this depth



Number of Comparison of MergeSort

- The maximum depth *D* of the recursive tree is $\lceil \log(n+1) \rceil$.
- Let *B* base case nodes on depth *D*-1, and *n*-*B* on depth *D*, (Note: base case node has nonrecursive cost 0).
- (n-B)/2 nonbase case nodes at depth D-1, each has nonrecursive cost 1.
- So:

$$W(n) = \sum_{d=0}^{D-2} (n-2^{d}) + \frac{n-B}{2} = n(D-1) \quad (2^{D-1}-1) + \frac{n-B}{2}$$
Since $(2^{D}-2B) + B = n$, that is $B = 2^{D} - n$
So, $W(n) = nD - 2^{D} + 1$

$$Let \frac{2^{D}}{n} = 1 + \frac{B}{n} = \alpha$$
, then $1 \le \alpha < 2$, $D = log n + log \alpha$
So, $W(n) = nlog n - (\alpha - log \alpha)n + 1$

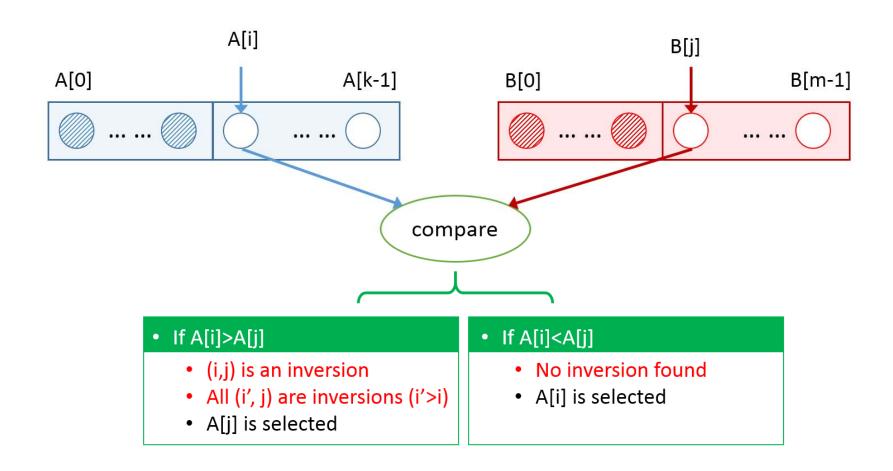
• $\lceil n \log(n) - n + 1 \rceil \le number of comparison \le \lceil n \log(n) - 0.914n \rceil$

The MergeSort D&C

- Counting the number of inversions
 - o Brute force: $O(n^2)$
 - o Can we use divide & conquer
 - In $O(nlogn) \Rightarrow$ combination in O(n)
- MergeSort as the carrier
 - o Sorted subarrays
 - A[0..k-1] and B[0..m-1]
 - o Compare the *left* and the *right* elements
 - A[i] vs. B[j]



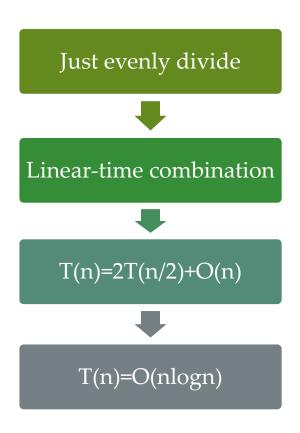
The MergeSort D&C





The MergeSort D&C

- Max-sum subsequence
- Counting inversions
- Finding the *frequent* element
- Find the nearest two points on the plane
- •





Lower Bounds for Comparison-based Sorting

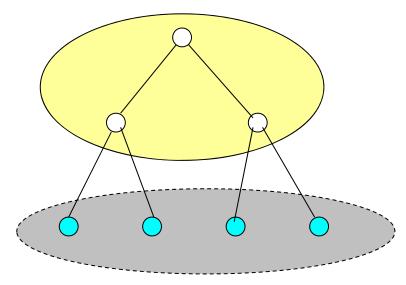
- Upper bound, e.g., worst-case cost
 - o For any possible input, the cost of the specific algorithm A is no more than the *upper bound*
 - $Max{Cost(i) | i is an input}$
- Lower bound, e.g., comparison-based sorting
 - For any possible (comparison-based) sorting algorithm A, the worst-case cost is no less than the lower bound
 - Min{*Worst-case*(a) | a is an algorithm}



2-Tree

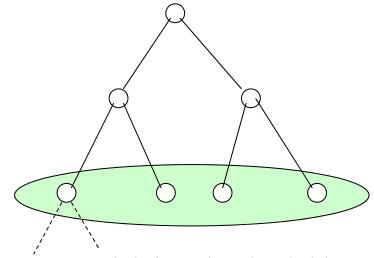
• 2-Tree

internal nodes



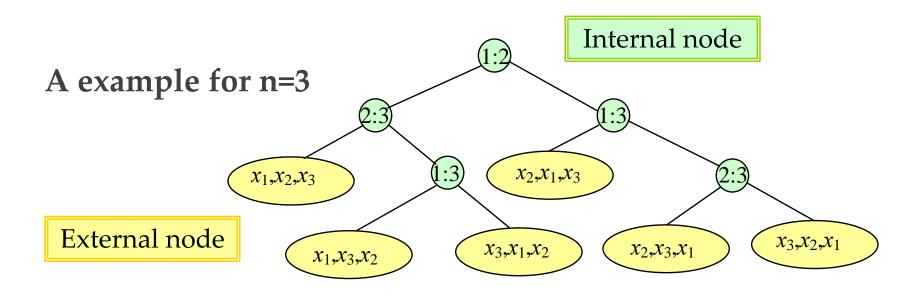
external nodes no child any type

Common Binary Tree



Both left and right children of these nodes are empty tree

Decision Tree for Sorting



- Decision tree is a 2-tree (assuming no same keys)
- The action of Sort on a particular input corresponds to following on path in its decision tree from the root to a leaf associated to the specific output



Characterizing the Decision Tree

- For a sequence of n distinct elements, there are n! different permutation
 - o So, the decision tree has at least n! leaves, and exactly n! leaves can be reached from the root.
 - o So, for the purpose of lower bounds evaluation, we use trees with exactly n! leaves.
- The number of comparison done in the *worst* case is the height of the tree.
- The average number of comparison done is the average of the lengths of all paths from the root to a leaf.



Lower Bound for Worst Case

- *Theorem*: Any algorithm to sort n items by comparisons of keys must do at least $\lceil \log n! \rceil$, or approximately $\lceil n \log n 1.443n \rceil$, key comparisons in the worst case.
 - o Note: Let L=n!, which is the number of leaves, then L≤ 2^h , where h is the height of the tree, that is h≥ $\lceil \log L \rceil = \lceil \log n! \rceil$
 - Lemma: let L be the number of leaves in a binary tree and h be its height. Then $L \le 2^h$
 - o For the asymptotic behavior:

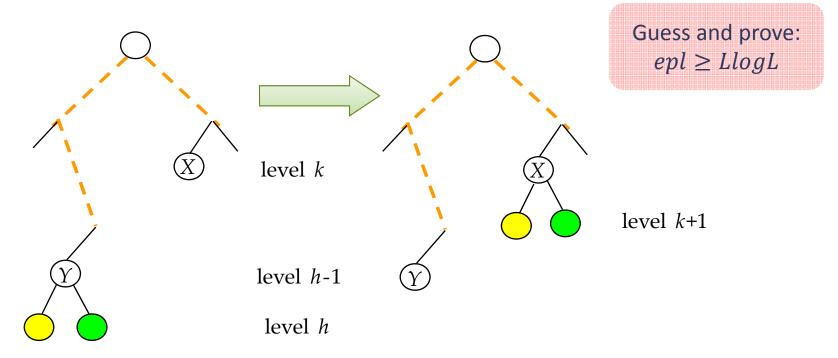
$$\log(n!) \ge \log(n(n-1)\cdots(\left\lceil \frac{n}{2}\right\rceil)) \ge \log(\frac{n}{2})^{\frac{n}{2}} = \frac{n}{2}\log(\frac{n}{2}) \in \Theta(n\log n)$$

External Path Length(EPL)

- EPL sum of path length to every leaf
 - o The EPL t is recursively defined as follows:
 - o [Base case] 0 for a single external node
 - o [Recursion] *t* is non-leaf with sub-trees *L* and *R*, then the sum of:
 - the external path length of *L*;
 - the number of external node of *L*;
 - the external path length of *R*;
 - the number of external node of *R*;



More Balanced 2-tree, Less EPL

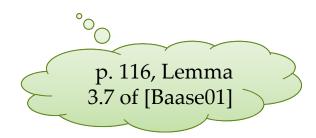


Assuming that h-k>1, when calculating epl, h+h+k is replaced by (h-1)+2(k+1). The net change in epl is k-h+1<0, that is, the epl decreases.



Properties of EPL

- Let *t* be a 2-tree, then the *epl* of *t* is the sum of the paths from the root to each external node.
- $epl \ge m \log(m)$, where m is the number of external nodes in t
 - $\circ epl=epl_L+epl_R+m \ge m_L\log(m_L)+m_R\log(m_R)+m,$
 - note $f(x)+f(y)\geq 2f((x+y)/2)$ for $f(x)=x\log x$
 - o so, $epl \ge 2((m_L + m_R)/2)\log((m_L + m_R)/2) + m$ $= m(\log(m)-1) + m = m\log m.$





Lower Bound for Average Behavior

- Since a decision tree with *L* leaves is a 2-tree, the average path length from the root to a leaf is *epl* .
 Recall that *epl* ≥ *L*log(*L*).
- Theorem: The average number of comparison done by an algorithm to sort n items by comparison of keys is at least log(n!), which is about nlogn-1.443n.

MergeSort Has Optimal Average Performance

- The average number of comparisons done by an algorithm to sort *n* items by comparison of keys is at least about *n*log*n*-1.443*n*
- The worst complexity of MergeSort is in Θ(nlogn)
- So, MergeSort is optimal as for its average performance

Thank you!

Q & A

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