Introduction to

Algorithm Design and Analysis

[06] MergeSort

Jingwei Xu https://ics.nju.edu.cn/~xjw/ Institute of Computer Software Nanjing University

In the last class ...

• Heap

- Partial order property
 - FixHeap
 - ConstructHeap
- Heap structure
 - Array-based implementation
- HeapSort
 - Complexity
 - Accelerated HeapSort

MergeSort

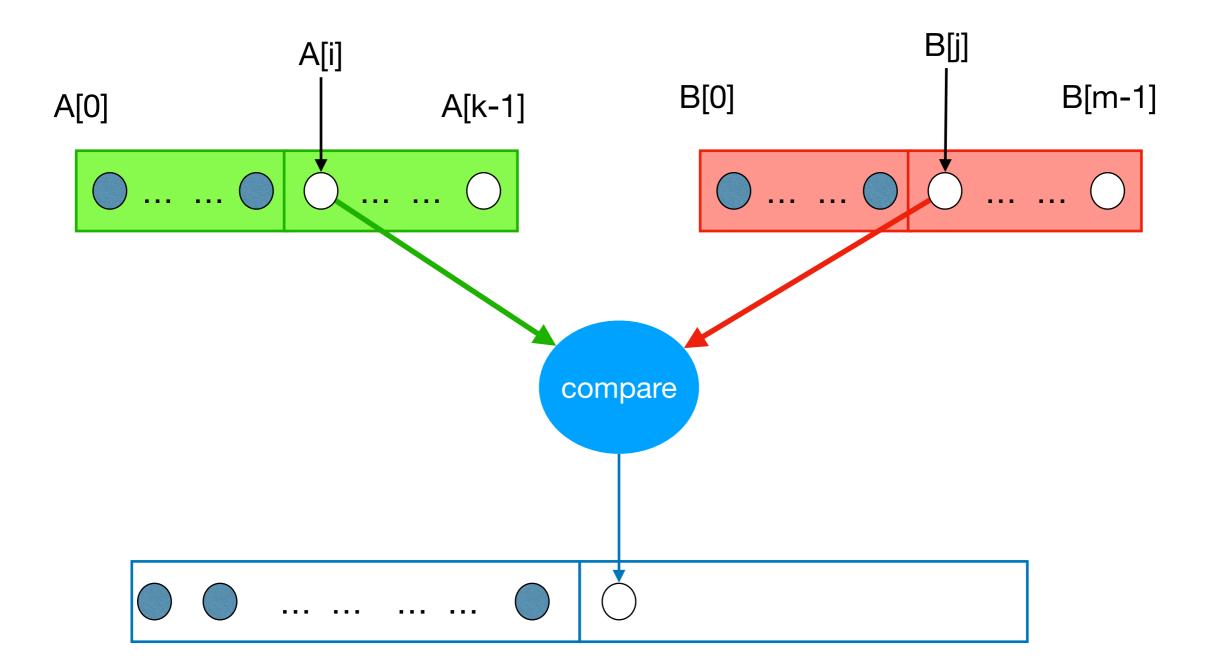
MergeSort

- Worst-case analysis of MergeSort
- Lower Bounds for comparison-based sorting
 - Worst-case
 - 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
 - Using Merge

Merging Sorted Arrays



Merge: the Specification

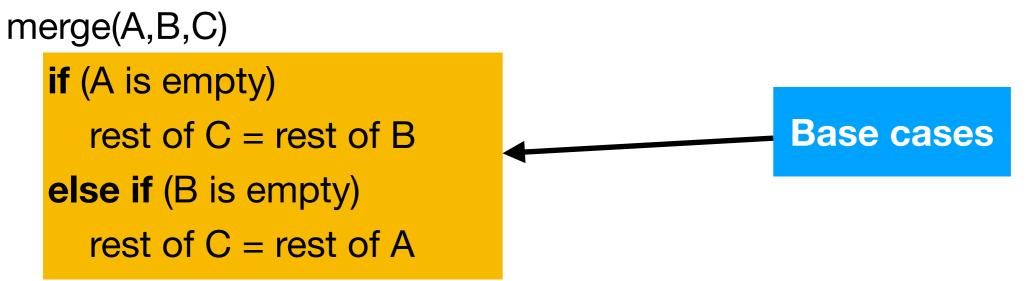
• Input

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

Output

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

Merge: Recursive Version



else

```
if (first of A ≤ 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

- Worst case is that the last comparison is conducted between A[k-1] and B[m-1]
 - After each comparison, at least 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 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

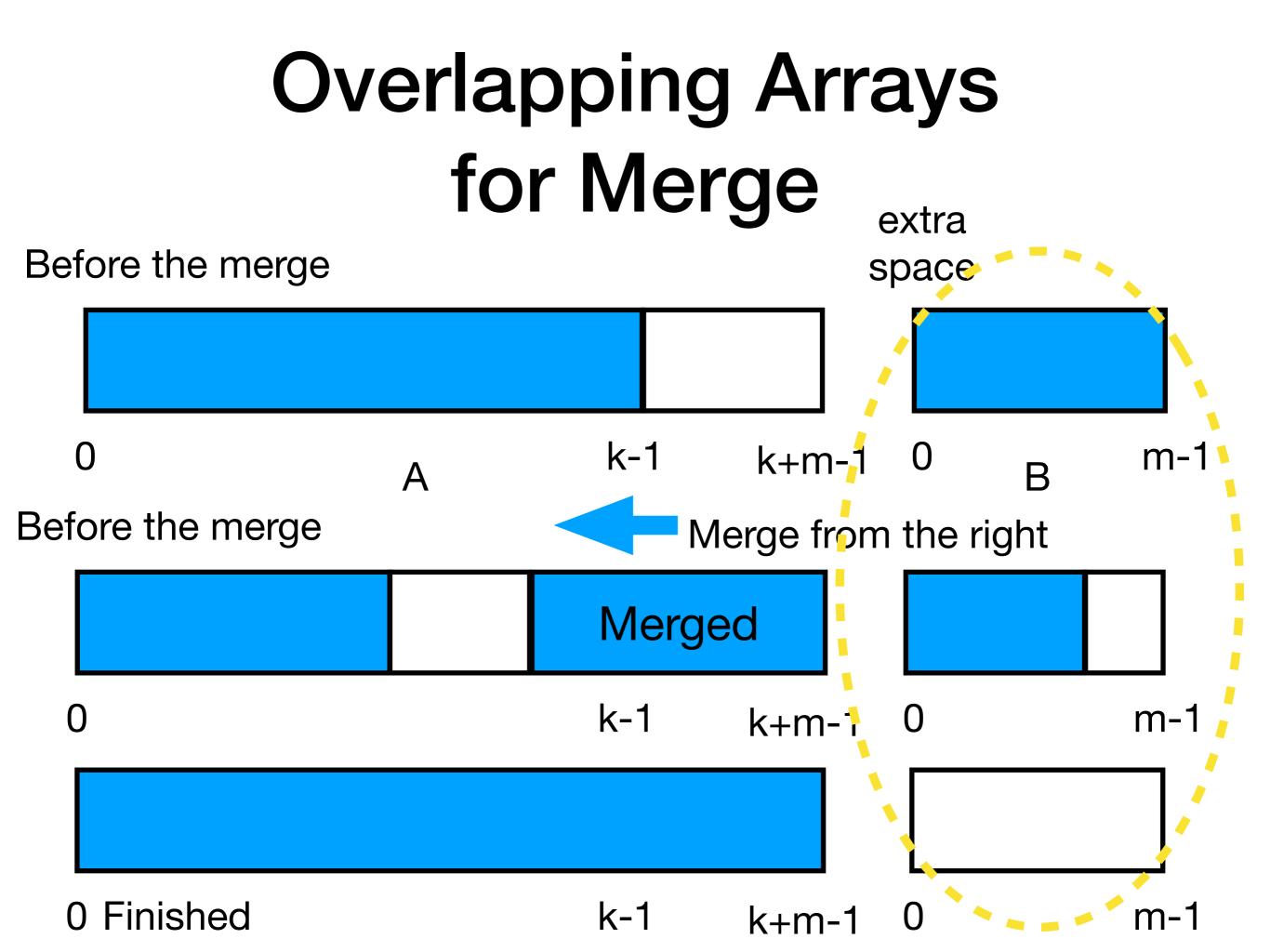
- Any algorithm to merge two sorted arrays, each containing k=m=n/2 entries, by comparison of keys, does at least n-1 comparisons in the worst case.
 - Choose keys so that:

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

 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.

Space Complexity of Merge

- An 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 needs O(n) extra space to store the merged sequence during the merging process.



MergeSort

- Input: Array E and indexes first, and last, such that the elements of E[i] are defined for first≤i≤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;

Analysis of MergeSort

• The recurrence equation for MergeSort

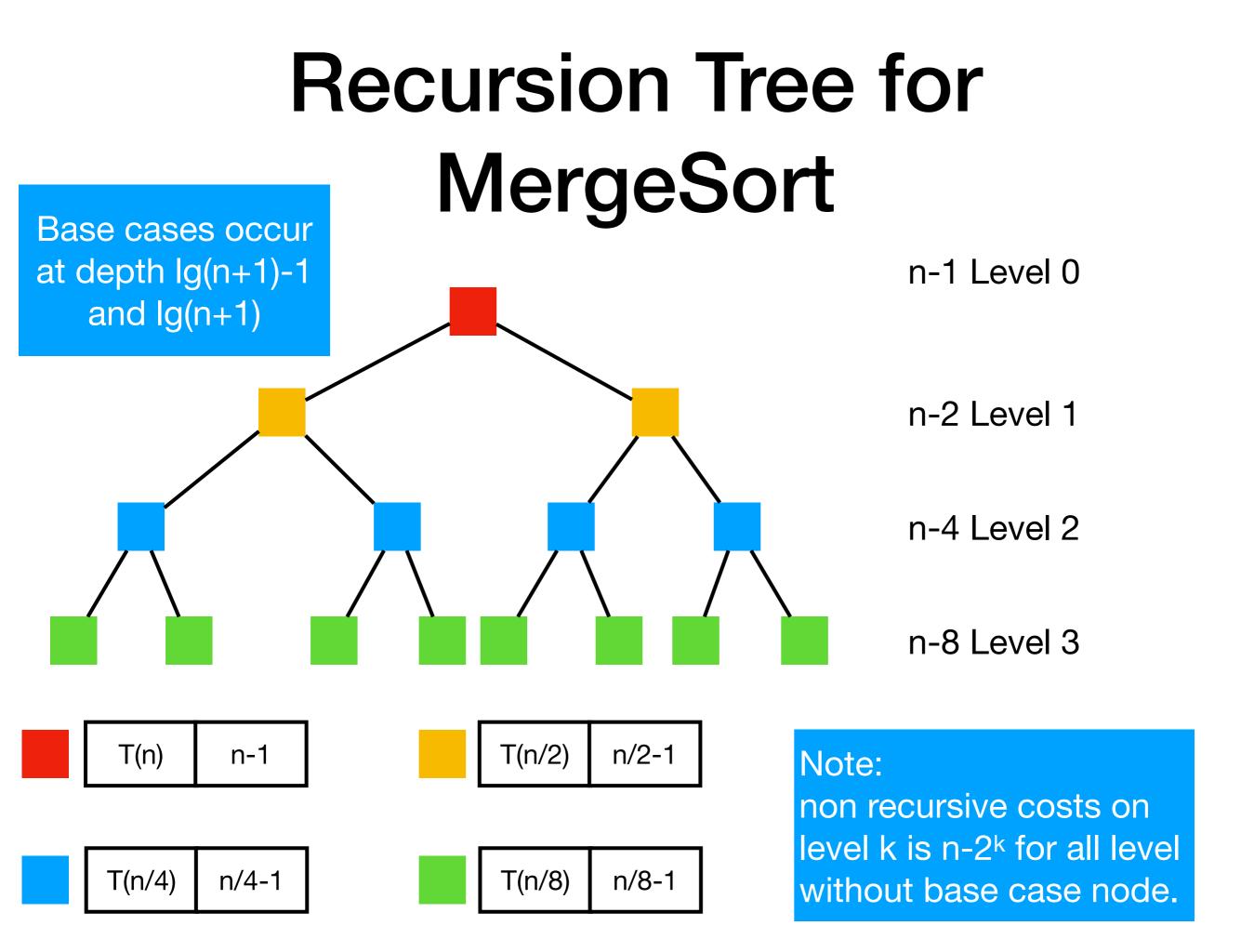
$W(n) = W(\lfloor n/2 \rfloor) + W(\lceil n/2 \rceil) + n - 1$ W(1) = 0

$$W(1) = 0$$

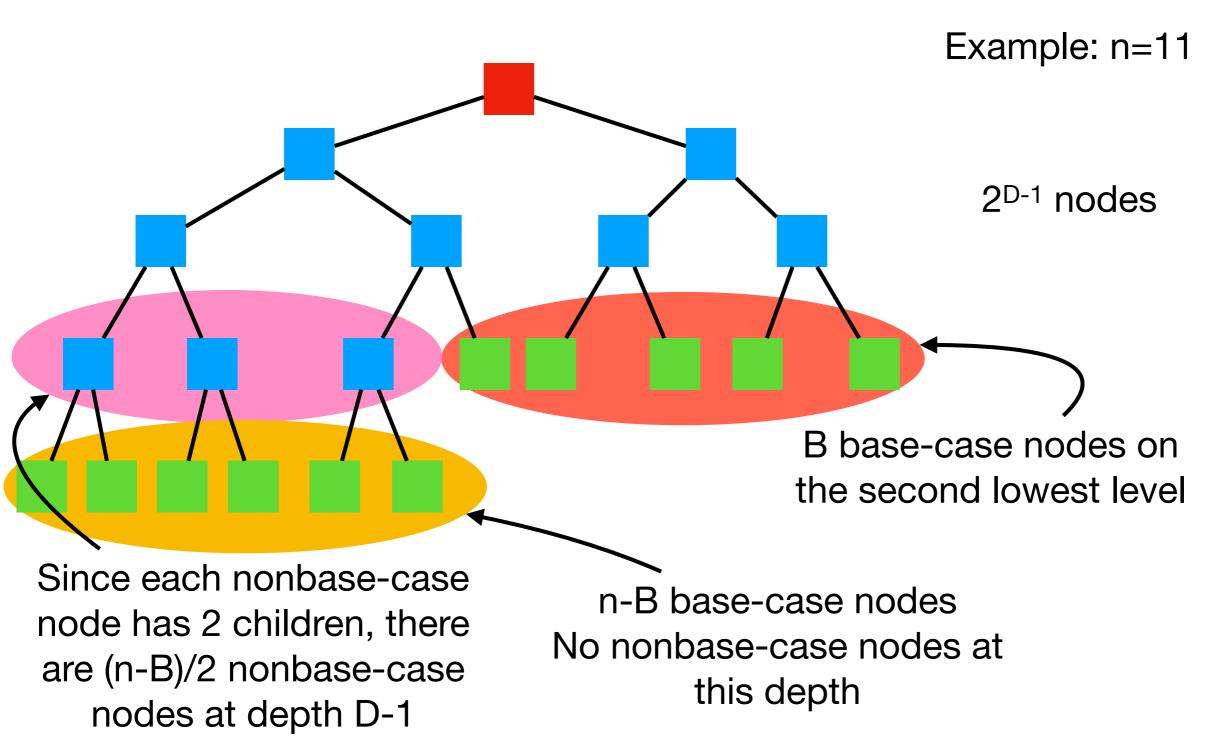
Where n = last - first + 1, the size of range to be sorted

The Master Theorem applies for the equation, so:

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



Non-complete Recursion Tree



Number of Comparison of MergeSort

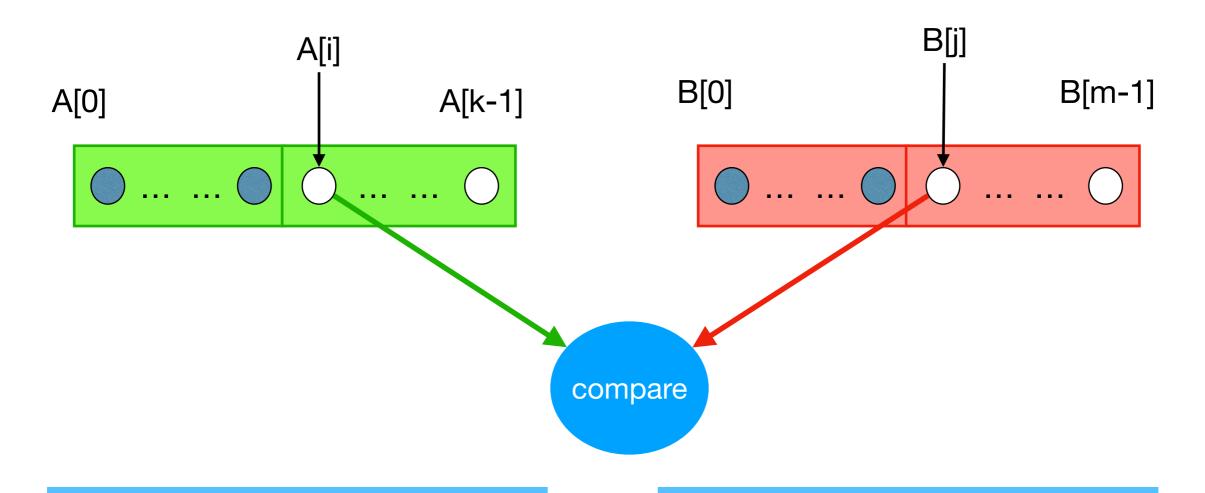
- Let B base case nodes on depth D-1, and n-B on depth D, (Note: base case node has non-recursive cost 0).
- (n-B)/2 non-base case nodes at depth D-1, each has non-recursive cost 1.
- So:

$$W(n) = \sum_{d=0}^{D-2} (n-2^{d}) + \frac{n-B}{2} = n(D-1) - (2^{D-1}-1) + \frac{n-B}{2}$$

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 $1 \le \alpha < 2$, $D = \lg n + \lg \alpha$
So, $W(n) = n \lg n - (\alpha - \lg \alpha)n + 1$

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

- Counting the number of inversions
 - Brute force: O(n²)
 - Can we use divide & conquer
 - In O(nlogn)=>combination in O(n)
- MergeSort as the carrier
 - Sorted subarrays
 - A[0..k-1] and B[0..m-1]
 - Compare the left and right elements
 - A[i] v.s. B[j]

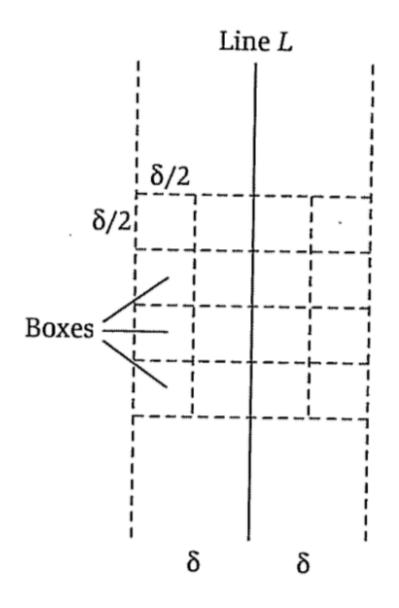


if A[i]>B[j] (i,j) is an inversion All (i',j) are inversions (i'>i) B[j] is selected

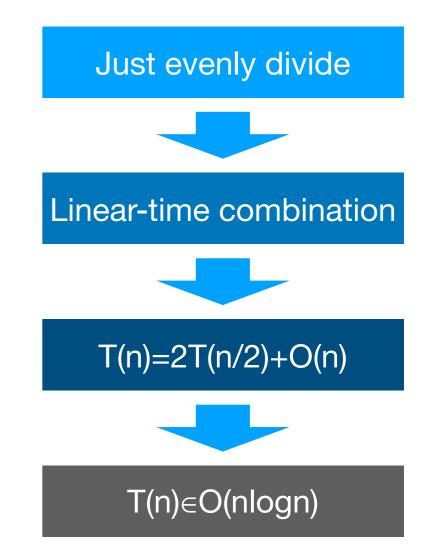
if A[i]<B[j] No inversion found A[i] is selected

The nearest pair

- n nodes on a plane
- The pair with the minimum distance
- The MergeSort D&C
 - T(n)=2T(n/2)+f(n)
 - f(n) must be O(n)
 - How?



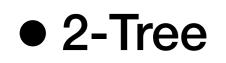
- Max-sum subsequence
- Maxima on a plane
- Finding the frequent element
- Integer/matrix multiplication



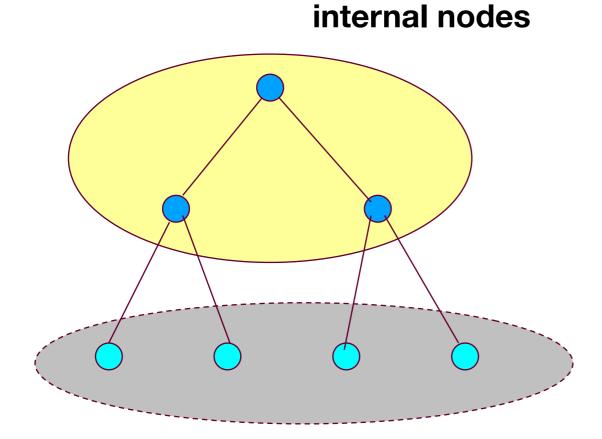
Lower Bounds for Comparison-based Sorting

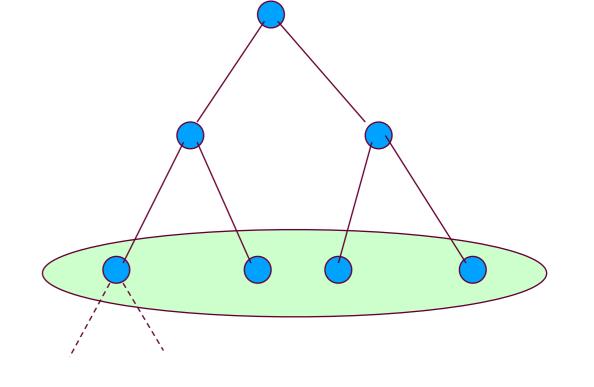
- Upper bound, e.g., worst-case cost
 - For any possible input, the cost of the specific algorithm A is no more than the upper bound
 - Max{cost(i) | i is an input}
- Lower bound, e.g., comparison-based sorting
 - For any possible (comparison-based) sorting algorithm A, the worst-case cost is no less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

2-Tree



Common Binary Tree

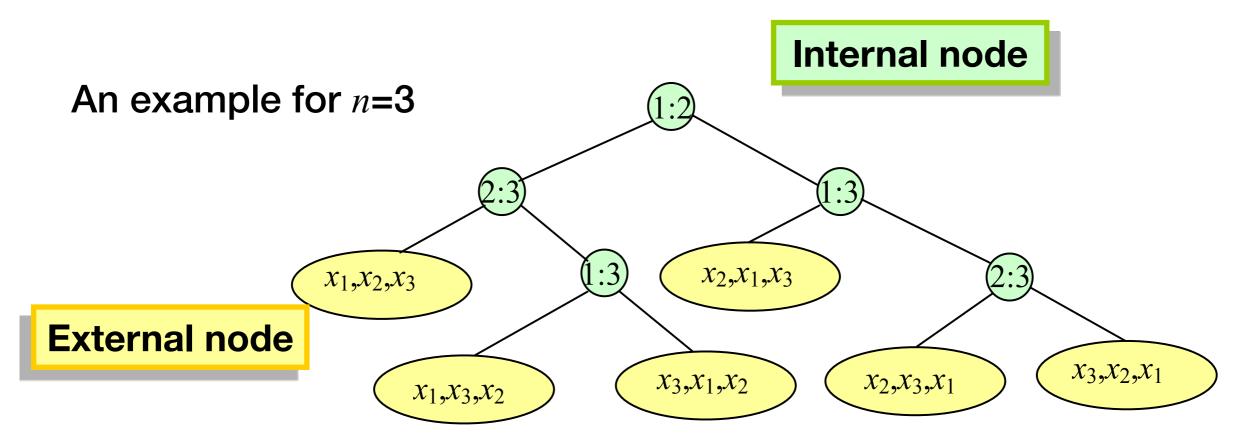




external nodes no child any type

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
 - So, the decision tree has at least n! leaves, and exactly n! leaves can be reached from the root.
 - 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 [Ign!], or approximately [nlgn-1.443n], key comparisons in the worst case.
 - Note: Let L=n!, which is the number of leaves, then $L \le 2^h$, where *h* is the height of the tree, that is $h \ge \lceil \lg L \rceil = \lceil \lg n! \rceil$
 - For the asymptotic behavior:

$$\lg(n!) \ge \lg[n(n-1)...\left(\left\lceil \frac{n}{2} \right\rceil\right)] \ge \lg\left(\frac{n}{2}\right)^{\frac{n}{2}} = \frac{n}{2}\lg\left(\frac{n}{2}\right) \in \Theta(n\lg n)$$

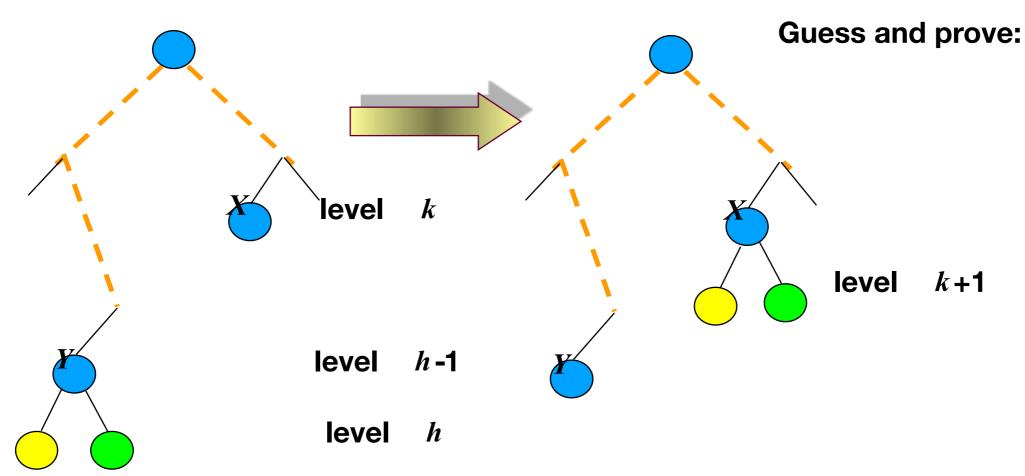
derived using:
$$\lg n! = \sum_{j=1}^{n} \lg(j)$$

External Path Length (EPL)

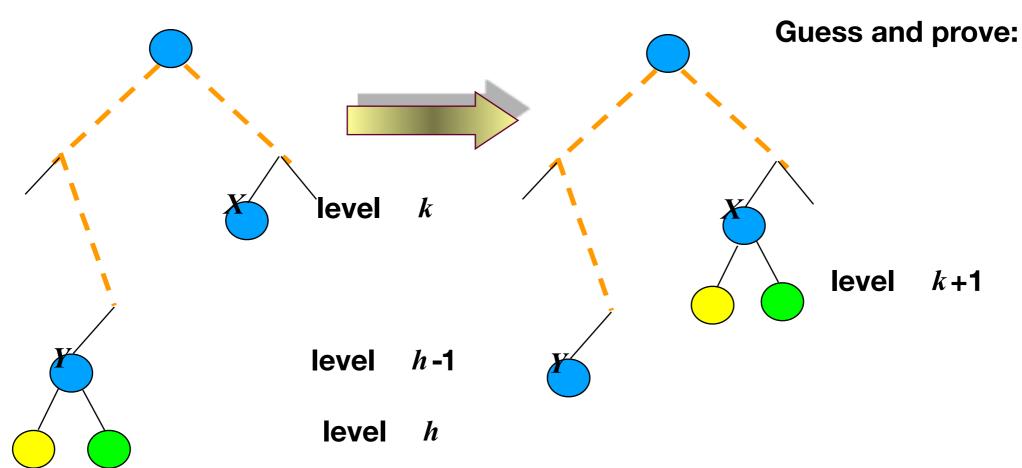
• The EPL of a 2-tree t is defined as follows:

- [Base case] 0 for a single external node
- [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



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.

So, more balanced 2-tree has smaller epl.

Properties of EPL

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

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 $\frac{epl}{L}$
 - Recall that $epl \ge Llg(L)$.
- **Theorem**: The average number of comparison done by an algorithm to sort *n* items by comparison of keys is at least lg(*n*!), which is about *n*lg*n*-1.443*n*.

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*lg*n*-1.443*n*
- The worst complexity of MergeSort is in $\Theta(n \lg n)$
- But, the average performance can not be worse than the worst case performance.
- So, MergeSort is optimal as for its average performance.

Thank you! Q&A