

Merge-sort

COMS10017 - (Object-Oriented Programming and) Algorithms

Dr Christian Konrad

Sorting Problem

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- **Output:** A reordering of A s.t. $A[0] \leq A[1] \leq \dots \leq A[n - 1]$

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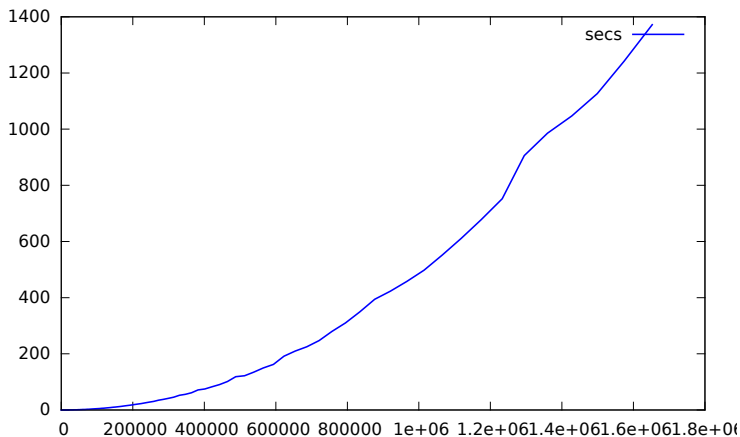
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Insertion Sort

- Worst-case runtime $O(n^2)$
- Surely we can do better?!

Insertion sort in Practice on Worst-case Instances



n	46929	102428	364178	1014570
secs	1.03084	4.81622	61.2737	497.879

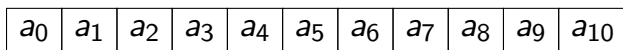
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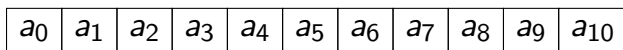
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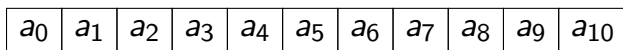


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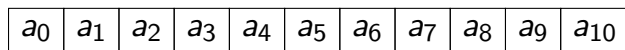
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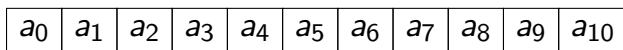
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A sorting algorithm is *stable* if any pair of equal numbers in the input array appear in the same order in the sorted array

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family name	first name	data of birth	role
Smith	Peter	02.10.1982	lecturer
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Observe: Stability makes more sense when sorting complex data as opposed to numbers

Merge Sort

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Merge Operation

- Copy left half of A to new array B
- Copy right half of A to new array C
- Traverse B and C simultaneously from left to right and write the smallest element at the current positions to A

Example: Merge Operation

A

1	4	9	10	3	5	7	11
---	---	---	----	---	---	---	----

Example: Merge Operation

A

1	4	9	10	3	5	7	11
---	---	---	----	---	---	---	----

B

1	4	9	10
---	---	---	----

C

3	5	7	11
---	---	---	----

Example: Merge Operation

A

--	--	--	--	--	--	--	--

B

1	4	9	10
---	---	---	----

C

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---	---	---	----

Example: Merge Operation

A

--	--	--	--	--	--	--	--

B

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C

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---	---	---	----

Example: Merge Operation

A

1							
---	--	--	--	--	--	--	--

B

1	4	9	10
---	---	---	----

C

3	5	7	11
---	---	---	----

Example: Merge Operation

A

1	3						
---	---	--	--	--	--	--	--

B

1	4	9	10
---	---	---	----

C

3	5	7	11
---	---	---	----

Example: Merge Operation

A

1	3	4					
---	---	---	--	--	--	--	--

B

1	4	9	10
---	---	---	----

C

3	5	7	11
---	---	---	----

Example: Merge Operation

A

1	3	4	5				
---	---	---	---	--	--	--	--

B

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---	---	---	----

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Example: Merge Operation

A

1	3	4	5	7			
---	---	---	---	---	--	--	--

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Example: Merge Operation

A

1	3	4	5	7	9	10	11
---	---	---	---	---	---	----	----

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Divide and Conquer!

Merge Sort: A Divide and Conquer Algorithm

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Require: Array  $A$  of  $n$  numbers  
if  $n = 1$  then  
    return  $A$   
 $A[0, \lfloor \frac{n}{2} \rfloor] \leftarrow \text{MERGESORT}(A[0, \lfloor \frac{n}{2} \rfloor])$   
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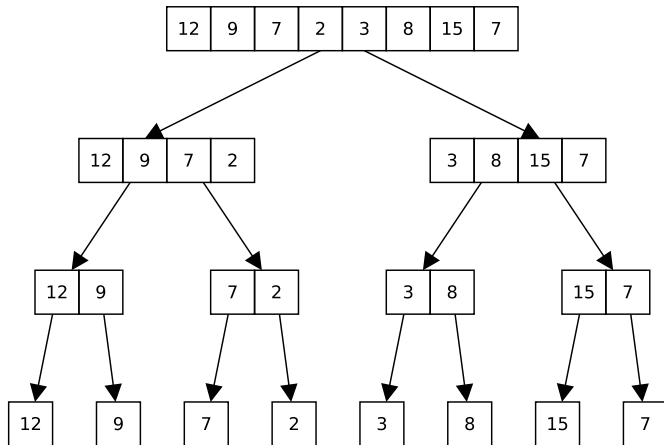
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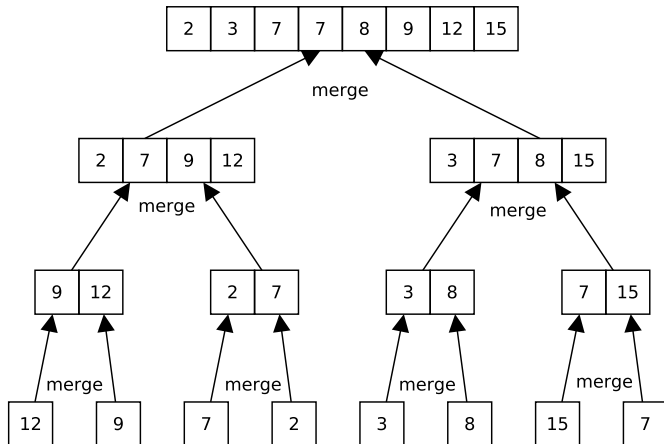
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- **Combine** the solutions to the subproblems into the solution for the original problem.

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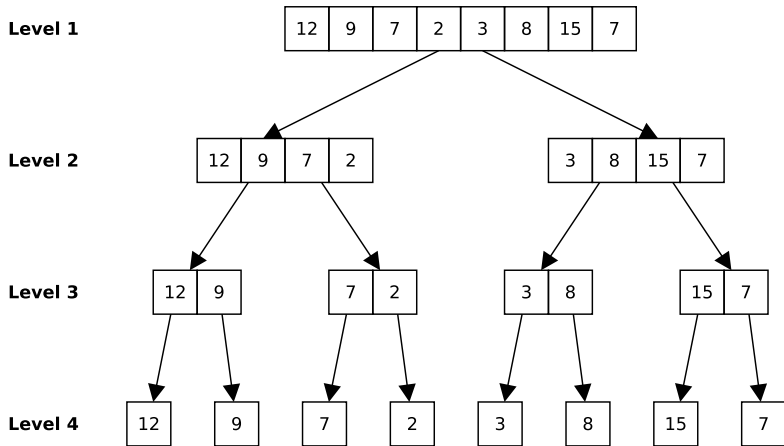
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- How many nodes per level?
- Time spent per node?

Number of Levels



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$$\frac{n}{2^{l-2}} > 1$$

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$$\frac{n}{2^{l-2}} > 1 \Rightarrow n > 2^{l-2} \Rightarrow \log(n) + 2 > l$$

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- Array length in level i is $\lceil \frac{n}{2^{i-1}} \rceil$ (at most)
- Runtime of merge operation for each node in level i : $O(\frac{n}{2^{i-1}})$

Number of Levels:

- Array length in last level l is 1: $\lceil \frac{n}{2^{l-1}} \rceil = 1$

$$\frac{n}{2^{l-1}} \leq 1 \Rightarrow n \leq 2^{l-1} \Rightarrow \log(n) + 1 \leq l$$

- Array length in last but one level $l - 1$ is 2: $\lceil \frac{n}{2^{l-2}} \rceil = 2$

$$\frac{n}{2^{l-2}} > 1 \Rightarrow n > 2^{l-2} \Rightarrow \log(n) + 2 > l$$

$$\log(n) + 1 \leq l < \log(n) + 2$$

Number of Levels (2)

Level i :

- 2^{i-1} nodes (at most)
- Array length in level i is $\lceil \frac{n}{2^{i-1}} \rceil$ (at most)
- Runtime of merge operation for each node in level i : $O(\frac{n}{2^{i-1}})$

Number of Levels:

- Array length in last level l is 1: $\lceil \frac{n}{2^{l-1}} \rceil = 1$

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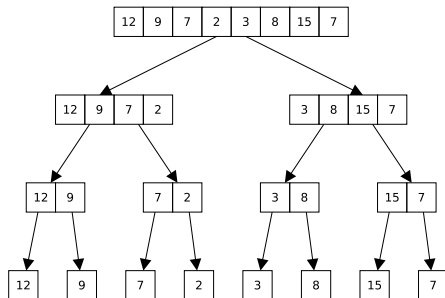
$$\log(n) + 1 \leq l < \log(n) + 2$$

Hence, $l = \lceil \log n \rceil + 1$.

Runtime of Merge Sort

Sum up Work:

- Levels:
 $l = \lceil \log n \rceil + 1$
- Nodes on level i :
at most 2^{i-1}
- Array length in level i :
at most $\lceil \frac{n}{2^{i-1}} \rceil$

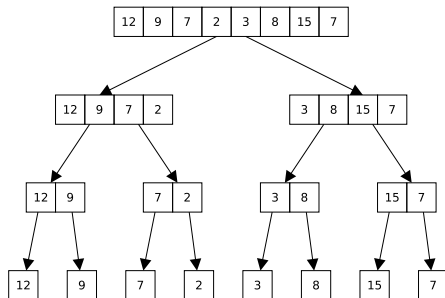


Runtime of Merge Sort

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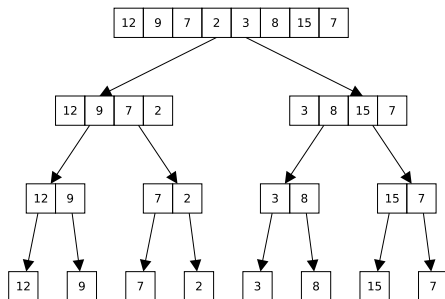
Worst-case Runtime:



Runtime of Merge Sort

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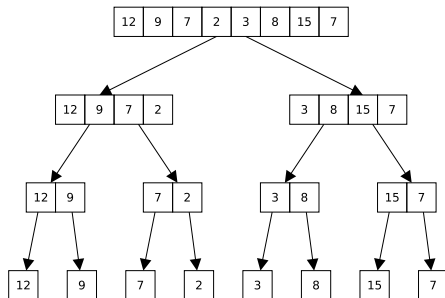
Worst-case Runtime:

$$\sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\lceil \frac{n}{2^{i-1}} \rceil\right)$$

Runtime of Merge Sort

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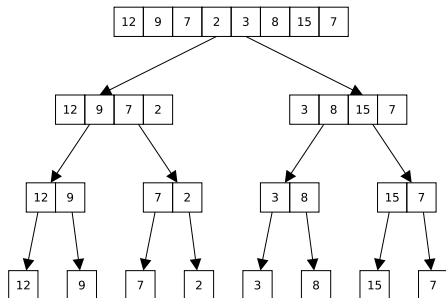
Worst-case Runtime:

$$\sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\left\lceil \frac{n}{2^{i-1}} \right\rceil\right) = \sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\frac{n}{2^{i-1}}\right)$$

Runtime of Merge Sort

Sum up Work:

- Levels:
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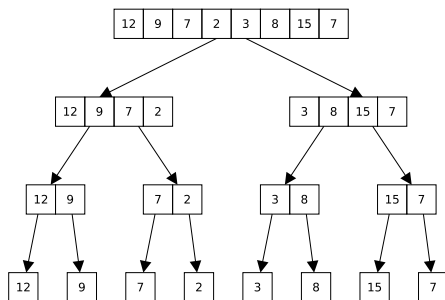
Worst-case Runtime:

$$\begin{aligned} \sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\left\lceil \frac{n}{2^{i-1}} \right\rceil\right) &= \sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\frac{n}{2^{i-1}}\right) \\ &= \sum_{i=1}^{\lceil \log n \rceil + 1} O(n) \end{aligned}$$

Runtime of Merge Sort

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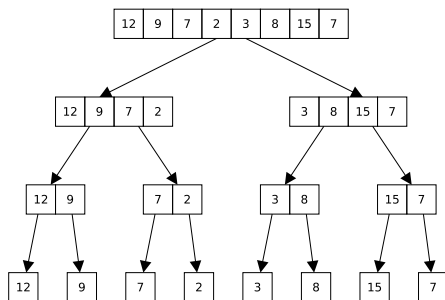
Worst-case Runtime:

$$\begin{aligned} \sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\lceil \frac{n}{2^{i-1}} \rceil\right) &= \sum_{i=1}^{\lceil \log n \rceil + 1} 2^{i-1} O\left(\frac{n}{2^{i-1}}\right) \\ &= \sum_{i=1}^{\lceil \log n \rceil + 1} O(n) = (\lceil \log n \rceil + 1) O(n) \end{aligned}$$

Runtime of Merge Sort

Sum up Work:

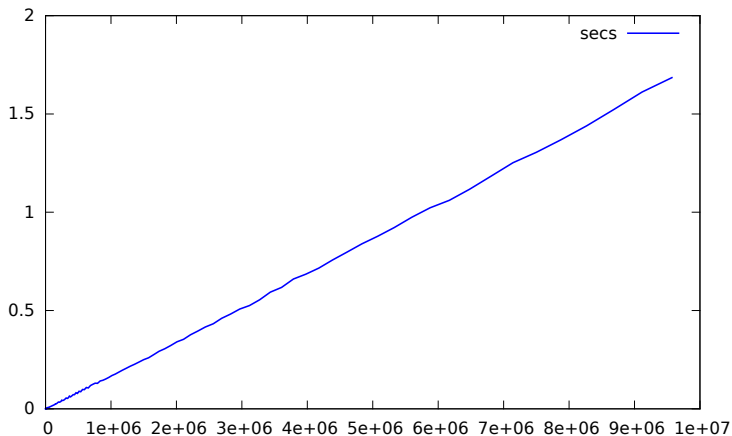
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Merge sort in Practice on Worst-case Instances



n	46929	102428	364178	1014570
secs	1.03084	4.81622	61.2737	497.879 (Insertion-sort)
secs	0.007157	0.015802	0.0645791	0.169165 (Merge-sort)

Stability and In Place Property?

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- Merge sort is stable

Stability and In Place Property?

- Merge sort is stable
- Merge sort does not sort in place

Divide and Conquer Algorithm:

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Then:

A has a runtime of $O(n \log n)$.