Merge Sort

https://cs.pomona.edu/classes/cs140/

Outline

Topics and Learning Objectives

- Learn how the merge sort algorithm operates
- Become aware of the "Divide and Conquer" algorithmic paradigm by analyzing merge sort

Exercise

• Recursion tree

Extra Resources

• Chapter 4: divide-and-conquer

Divide and Conquer

- This is an algorithm design paradigm
- Most divide and conquer algorithms are recursive in nature
- The basic idea is to break the problem into easier-to-solve subproblems
- What's easier to do:
 - Sort 0, 1, or 2 numbers, or
 - Sort 10 numbers

Merge Sort

- This is a "Divide and Conquer"-style algorithm
- Improves over insertion sort in the worst case
- Unlike insertion sort, the best/average/worst case running times of merge sort are all the same

```
FUNCTION MergeSort(array)
n = array.length
IF n == 1
RETURN array
```



left_sorted = MergeSort(array[0 ..< n//2])
right_sorted = MergeSort(array[n//2 ..< n])</pre>

array_sorted = Merge(left_sorted, right_sorted)

RETURN array_sorted

FUNCTION MergeSort(array) O(1) n = array.length O(1) IF n == 1 O(1) RETURN array

array_sorted = Merge(left_sorted, right_sorted)

RETURN array_sorted

```
T(n) FUNCTION MergeSort(array)
        0(1) n = array.length
        0(1) IF n == 1
        0(1) RETURN array
```

T(n/2) left_sorted = MergeSort(array[0 ..< n//2])
T(n/2) right_sorted = MergeSort(array[n//2 ..< n])</pre>

array_sorted = Merge(left_sorted, right_sorted)

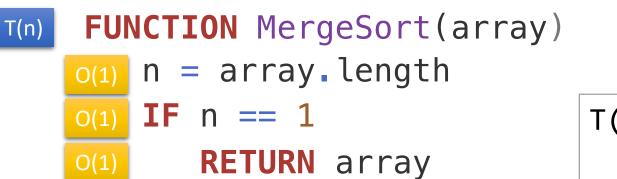
RETURN array_sorted

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T(n) FUNCTION MergeSort(array)
O(1) n = array.length
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T(n/2) left_sorted = MergeSort(array[0 ..< n//2])
T(n/2) right_sorted = MergeSort(array[n//2 ..< n])</pre>

o(?) array_sorted = Merge(left_sorted, right_sorted)

O(1) **RETURN** array_sorted



$$T(n) = 2 T(n/2) + 0(?) + 4 0(1)$$

= 2 T(n/2) + 0(?)

T(n/2) left_sorted = MergeSort(array[0 ..< n//2])
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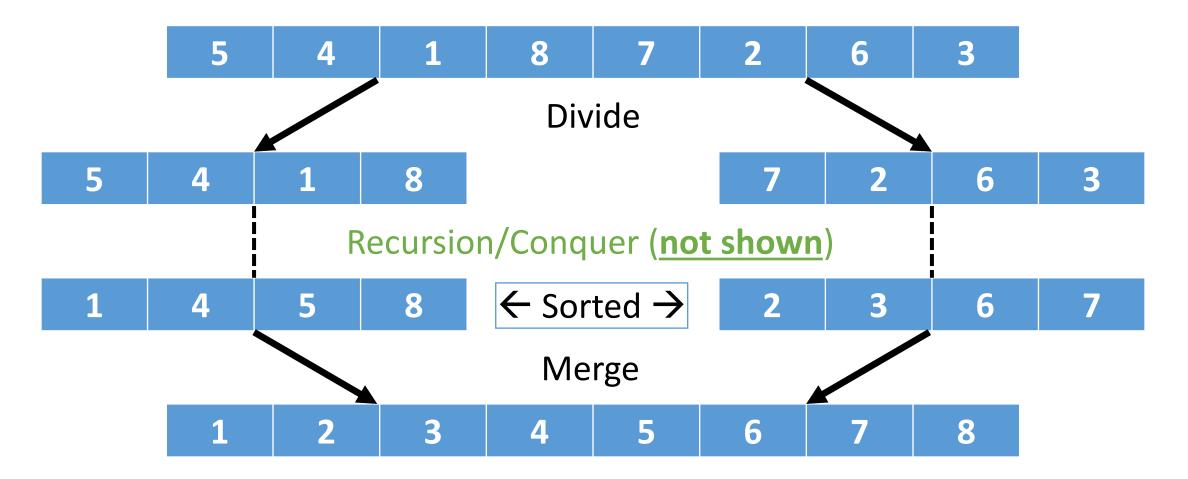
o(?) array_sorted = Merge(left_sorted, right_sorted)

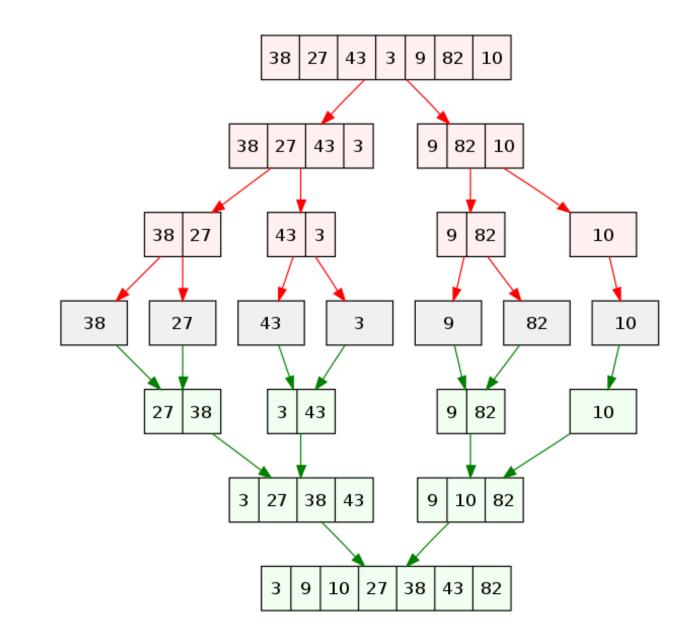


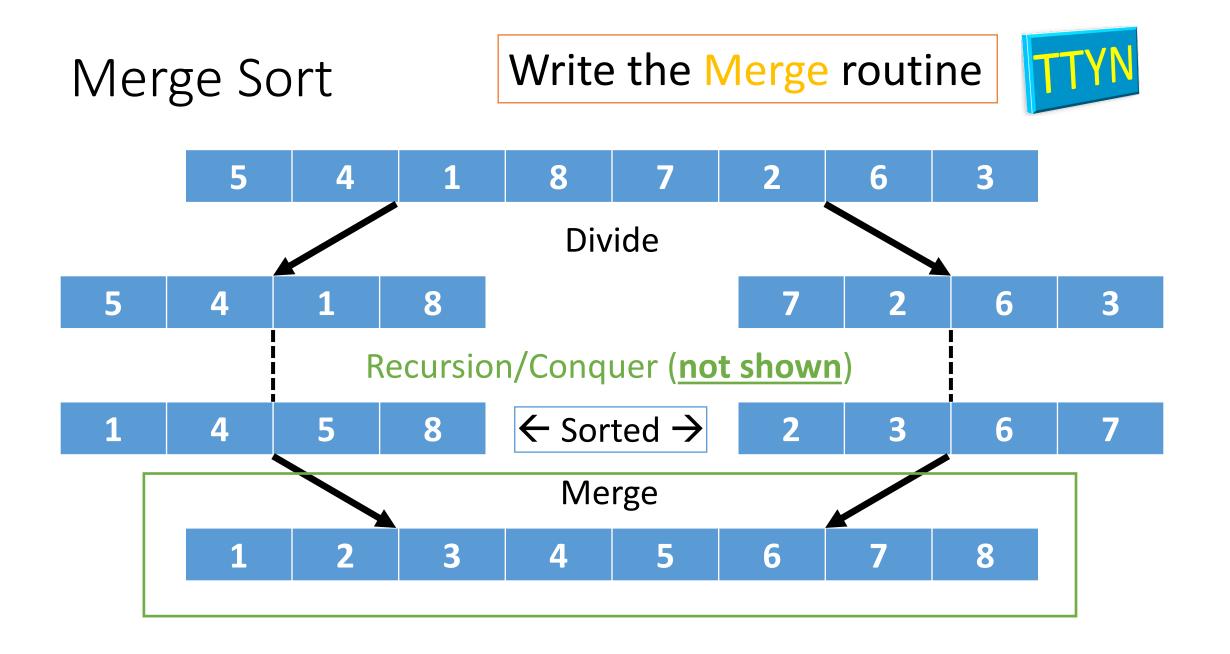
Recurrence Equation

T(n) = 2 T(n/2) + O(?) + 4 O(1)= 2 T(n/2) + O(?)

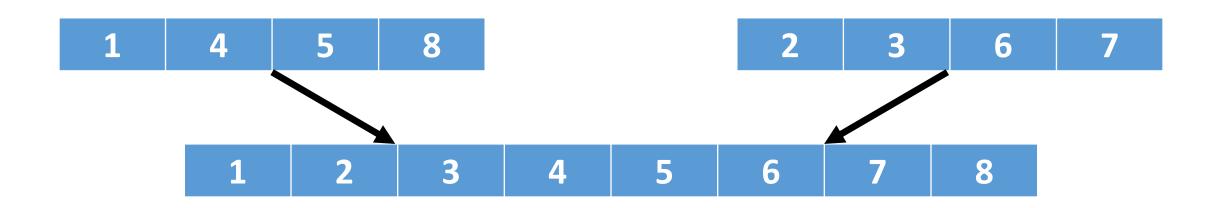
Merge Sort







FUNCTION Merge(one, two) out[one.length + two.length] # Declare array

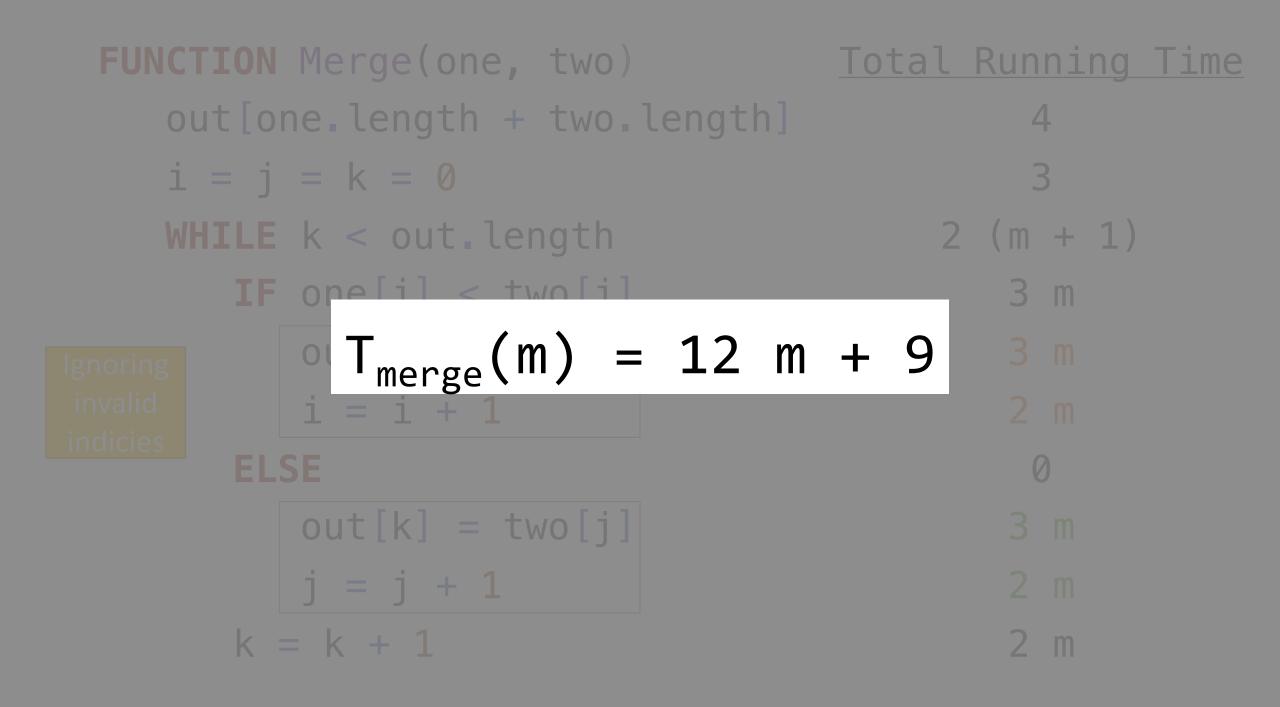


FUNCTION Merge(one, two) out[one.length + two.length] i = j = k = 0WHILE k < out.length IF one[i] < two[j]</pre> out[k] = one[i] Ignoring invalid i = i + 1indices ELSE out[k] = two[j]j = j + 1k = k + 1

What is the total running time?



FUNCTION Merge(one, two)Total Running Timeout[one.length + two.length]4
$$i = j = k = 0$$
3WHILE k < out.length2 (m + 1)IF one[i] < two[j]3 mout[k] = one[i]3 m $i = i + 1$ 2 minvalid $j = j + 1$ 3 m $j = j + 1$ 2 m $k = k + 1$ 2 m



Simplifying the running time

- We don't need to be *exactly* correct with the running time of Merge
- We will eventually remove lower order terms anyway
- Let's simplify the expression a bit:

$$T_{merge}(m) = 12m + 9$$

$$T_{merge}(m) \le 12m + 9m$$

 $T_{merge}(m) \leq 21m$

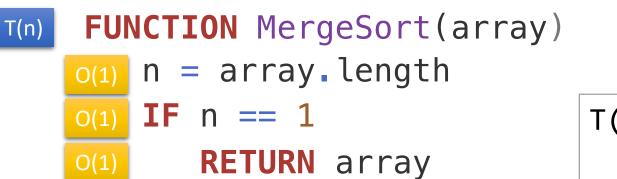
Merging

We have an idea of the cost of an <u>individual call to merge</u>:

 $T(m) \leq 21m$

What else do we need to know to calculate the total time of **MergeSort**?

- 1. How many times do we merge in total?
- 2. What is the size of each merge? (In other words: What is m?)



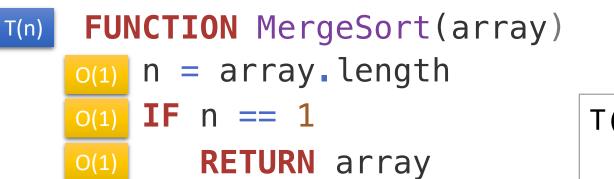
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o(?) array_sorted = Merge(left_sorted, right_sorted)





$$T(n) = 2 T(n/2) + O(n) + 4 O(1)$$

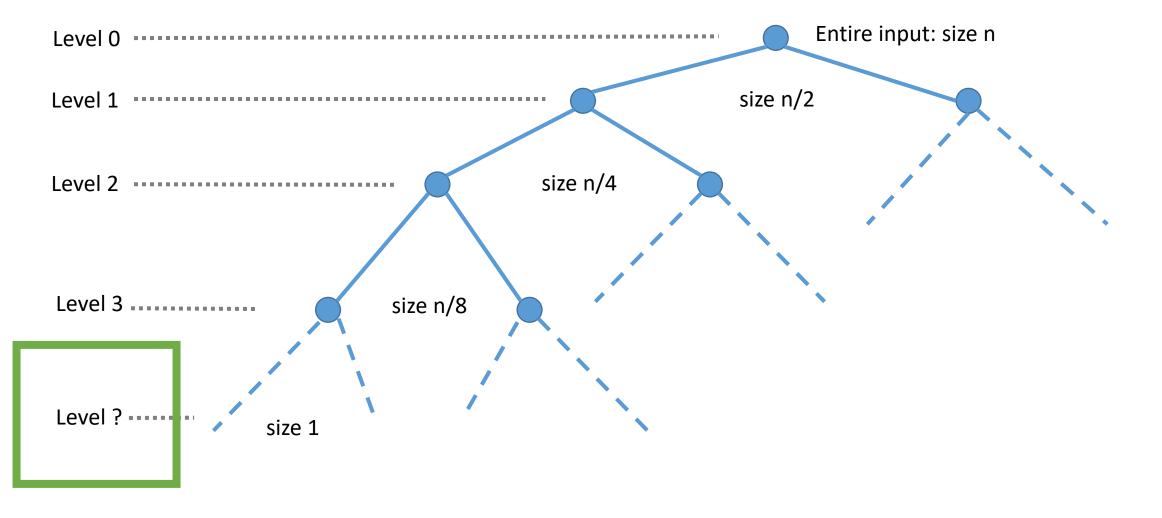
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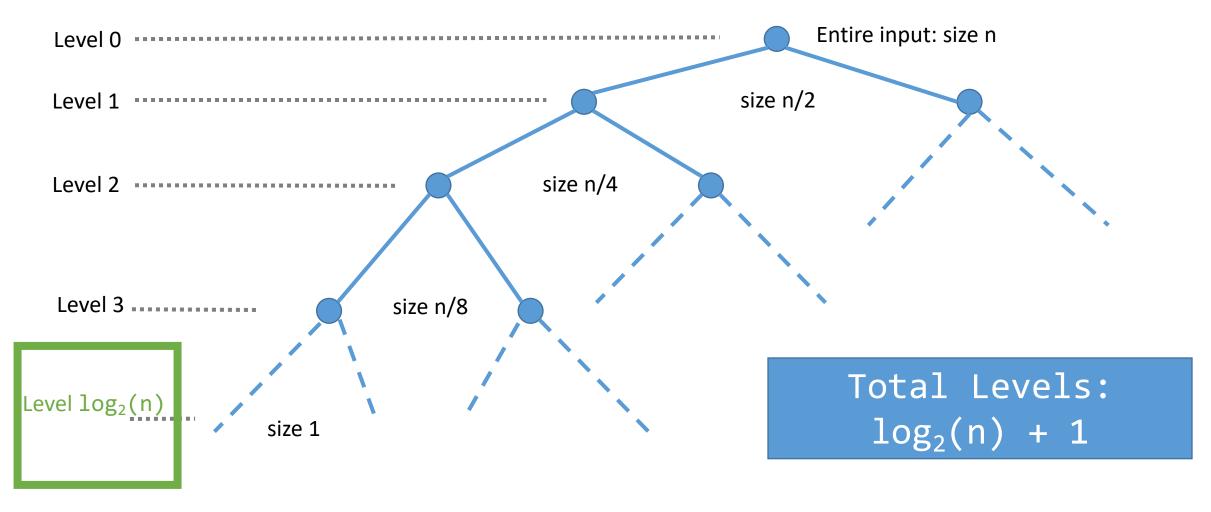
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How many times do we call Merge?



How many times do we call Merge?



Exercise



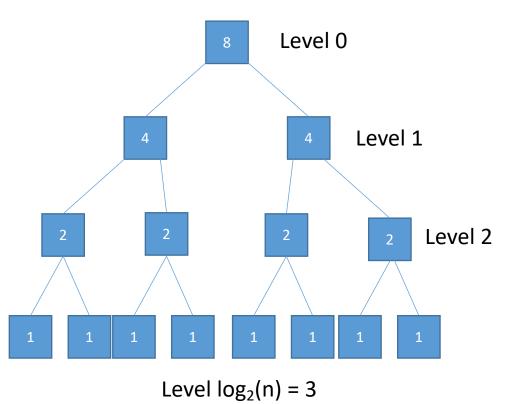
How many sub-problems are there at level L? The top level is Level 0, the second level is Level 1, and the bottom level is Level log₂(n)

Answer: 2^L

How many elements are there for a given sub-problem found in level L?

Answer: n/2^L

How many computations are performed at a given level? The cost of a Merge was 21m.



Exercise



How many sub-problems are there at level L? The top level is Level 0, the second level is Level 1, and the bottom level is Level $\log_2(n)$

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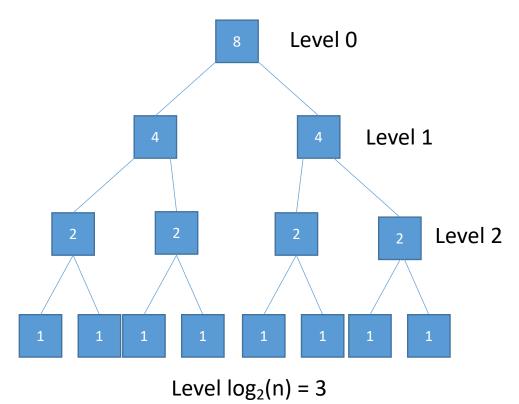
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Answer: $2^{L} 21(n/2^{L}) \rightarrow 21n$

What is the total computational cost of merge sort?

Answer: $21n (log_2(n) + 1)$



Exercise

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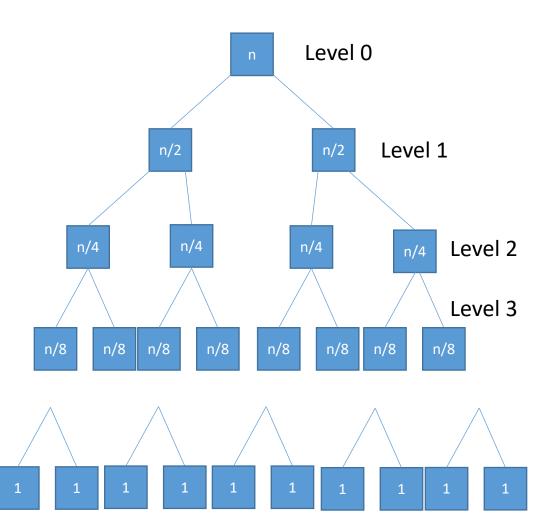
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Answer: $21n (log_2(n) + 1)$



Merge Sort

Divide and Conquer

constantly halving the problem size and then merging

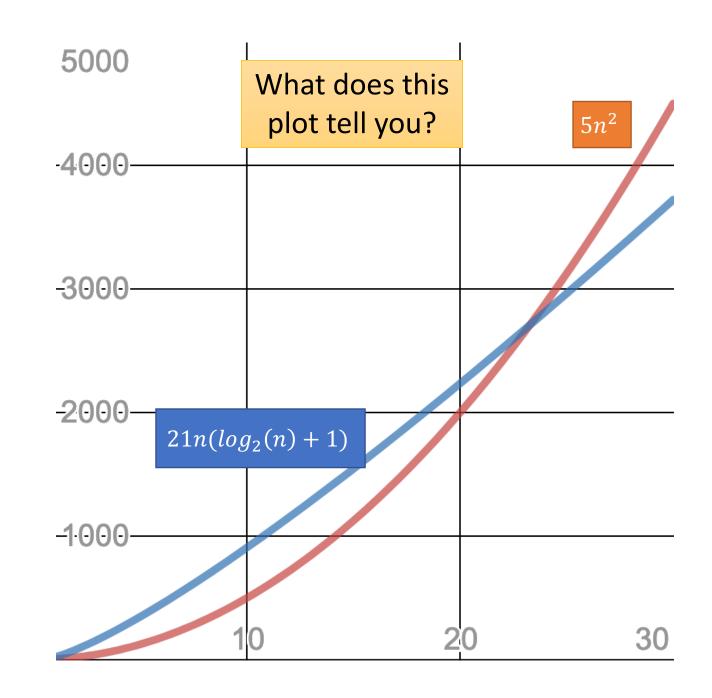
Total running time of roughly $21n \log_2(n) + 21n$

Compared to insertion sort with an average total running time of ½ n²

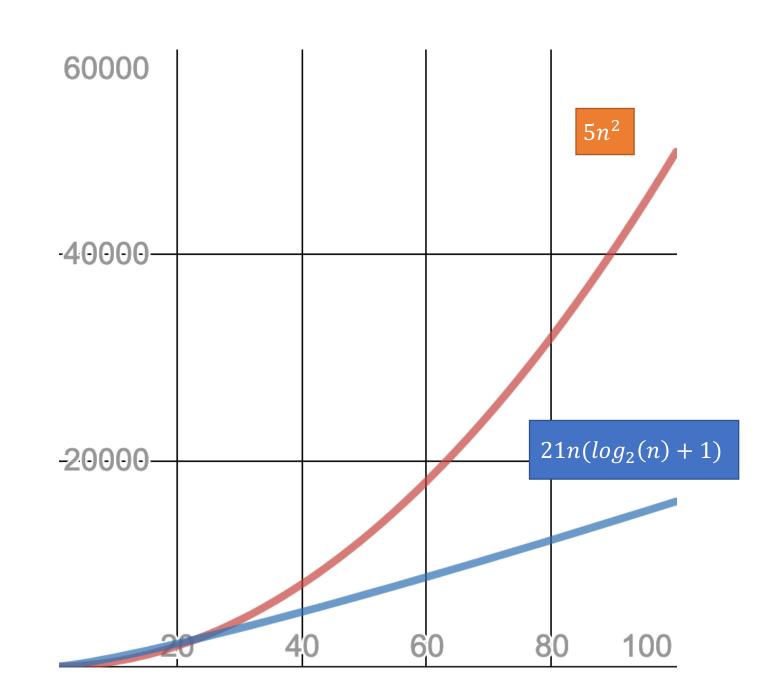
• For small values of n, insertion sort is better

Which algorithm is **better**?

Merge Sort Verse Insertion Sort Worst-Case



Merge Sort Verse Insertion Sort Worst-Case



Constants

