

# Merge Sort



# More sorting?

Heapsort had complexity  $O(n \log n)$ , and sorted in-place.  
Is there anything else worth looking at?

- Are there algorithms with better worst cases complexity?
  - even if they have the same complexity, maybe the lower order terms are better?
- Are there algorithms with better average complexity?
- Do we need to worry about in-place sorting?
- Are there other problem-solving strategies we could look at?

# Divide and Conquer

If a problem is very simple, solve it in a single step.  
If a problem is too complex to solve in a single step,  
    divide it into multiple pieces  
    solve the individual pieces  
    combine the pieces together to get a solution

Typically implemented using multiple recursion

A general problem solving strategy  
used throughout computing

# Divide and Conquer: sorting

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If a problem is too complex to solve in a single step,  
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## Sorting a list:

If an input list is of size 1 (or 0), do nothing.

If an input list is of size 2 or more

    split it into two almost equal sublists

*#divide*

    sort the first sublist

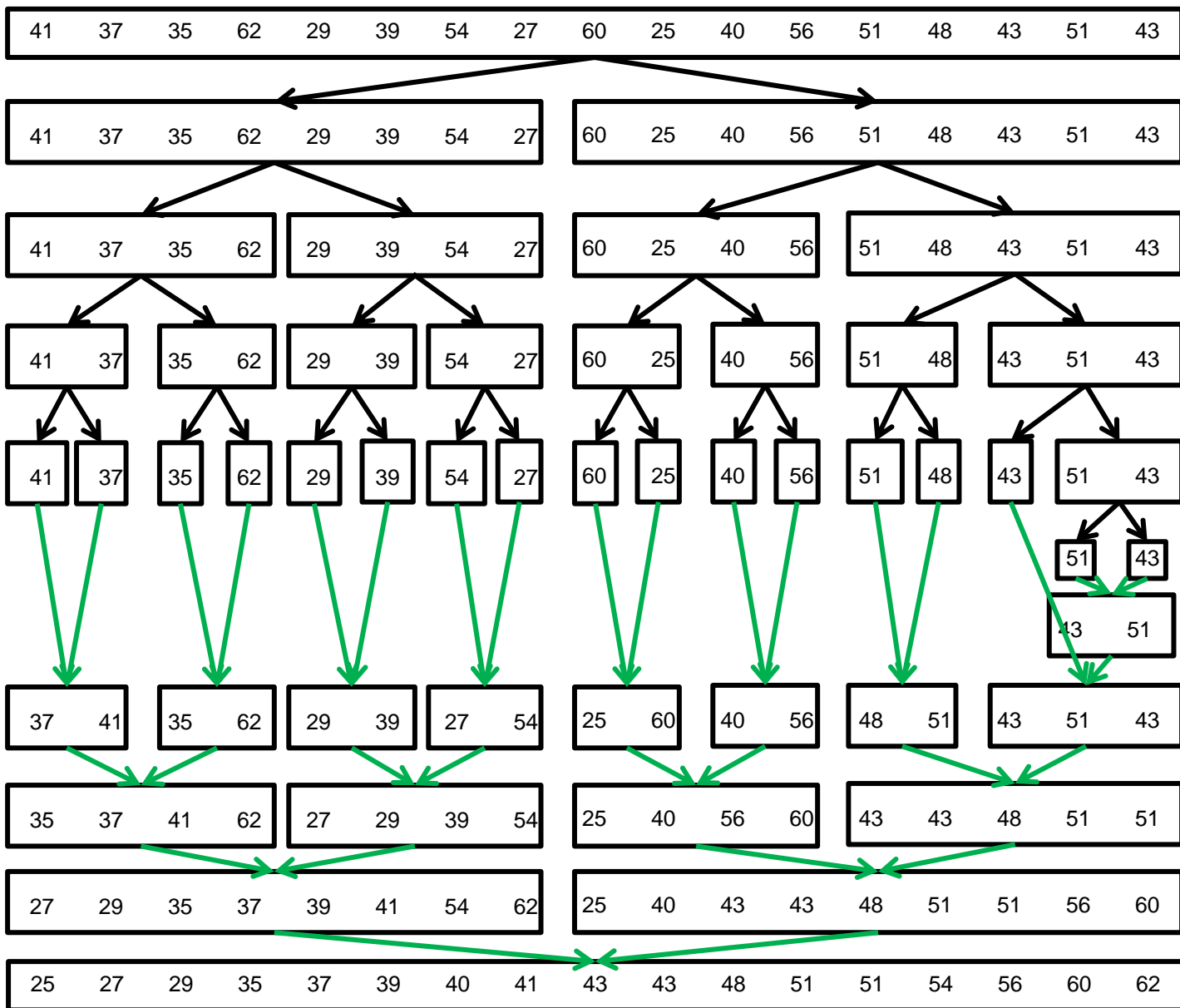
*#conquer*

    sort the second sublist

    merge the two sublists into a combined list

*#combine*

41	37	35	62	29	39	54	27	60	25	40	56	51	48	43	51	43
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```
def mergesort(mylist):  
    n = len(mylist)  
    if n > 1:  
        list1 = mylist[:n//2]  
        list2 = mylist[n//2:]  
        mergesort(list1)  
        mergesort(list2)  
        merge(list1, list2, mylist)
```

Slicing creates a new list each time, so not in-place.  
But it is difficult to write an in-place mergesort  
without increasing the time complexity

Merge:

35

37

41

62

27

29

39

54



```
def merge(list1, list2, mylist):  
    f1 = 0  
    f2 = 0  
    while f1 + f2 < len(mylist):  
        if f1 == len(list1):  
            mylist[f1+f2] = list2[f2]  
            f2 += 1  
        elif f2 == len(list2):  
            mylist[f1+f2] = list1[f1]  
            f1 += 1  
        elif list2[f2] < list1[f1]:  
            mylist[f1+f2] = list2[f2]  
            f2 += 1  
        else:  
            mylist[f1+f2] = list1[f1]  
            f1 += 1
```

**Note: written for clarity. Repeated code is not a good idea, so should rewrite to require only 1 test in loop body**

```
def merge(list1, list2, mylist):  
    f1 = 0  
    f2 = 0  
    while f1 + f2 < len(mylist):  
        if f1 == len(list1):  
            mylist[f1+f2] = list2[f2]  
            f2 += 1  
        elif f2 == len(list2):  
            mylist[f1+f2] = list1[f1]  
            f1 += 1  
        elif list2[f2] < list1[f1]:  
            mylist[f1+f2] = list2[f2]  
            f2 += 1  
        else:  
            mylist[f1+f2] = list1[f1]  
            f1 += 1
```

Analysis: ( $|mylist| = n$ )

round the loop  $n$  times

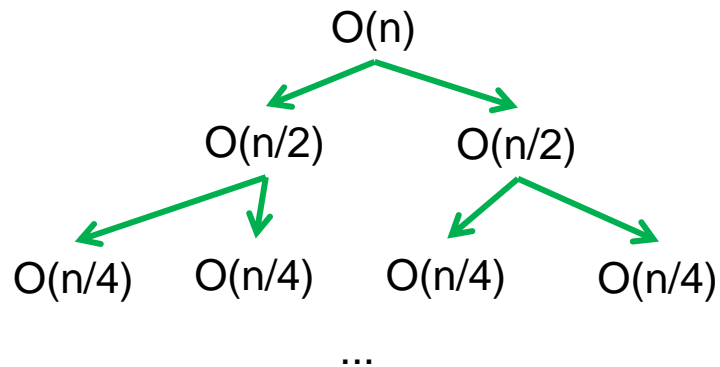
inside the loop, at most  
3 tests, 2 calls to `len(.)`,  
and 2 assignments

So  $O(1)$  inside the loop.

So function has worst  
case time  $O(n)$

Note: writing the result into the 3<sup>rd</sup> input list, so  
we do not occupy any extra space.

```
def mergesort(mylist):
    n = len(mylist)
    if n > 1:
        list1 = mylist[:n//2]
        list2 = mylist[n//2:]
        mergesort(list1)
        mergesort(list2)
        merge(list1, list2, mylist)
```



Each call creates new smaller lists,  
so space complexity (of this  
implementation) is same as time –  $O(n \log n)$

## Analysis:

Each call (without recursion) takes:

- $n$  assignments to create the slices
- $O(n)$  for the merge function

So  $O(n)$

Each recursive call is for a list of  
size  $n/2$ , and so takes  $O(n/2)$ , etc.

So we have  $O(n)$  at each level in  
the call tree.

The depth of the tree is either  
 $\log_2(n)$  or  $\log_2(n) + 1$

So  $O(n \log n)$  in total

# Alternative Analysis: recurrence equations

The base case is  $O(1)$ . Merge is  $O(n)$ , so we will write as  $c*n$ .

Time to sort a list of length  $n$ ,  $t(n)$ , for  $n > 1$ , is then:

$$t(n) = 2*t(n/2) + c*n.$$

But  $t(n/2)$  must then be  $2*t(n/4) + c*n/2$ . So

$$t(n) = 2*(2*t(n/4) + c*n/2) + c*n = 4*t(n/4) + 2c*n$$

$$t(n) = \dots = 8*t(n/8) + 3c*n$$

$$\text{So } t(n) = 2^k*t(n/2^k) + kc*n$$

This eventually stops when the list is of size 1, which happens when  $k = \log_2 n$ . But  $t(n/2^k) = t(n/2^{\log_2 n}) = O(1)$  since list is length 1. Also,  $2^{\log_2 n} = n$ .

$$\text{So } t(n) = n + \log_2 n * c * n \quad \text{which is } O(n \log n)$$

# Alternative Mergesort implementations

1. implementing mergesort on linked lists is (probably) easier
2. Mergesort on arrays can be implemented bottom-up rather than top down, using just  $O(n)$  extra space:

Create a new empty list of size  $n$ , called list 2

For each pair of cells in original list

merge into sorted pair in corresponding cells in list 2

For each successive group of 4 cells in list 2

merge into sorted group of 4 in corresponding cells in original list

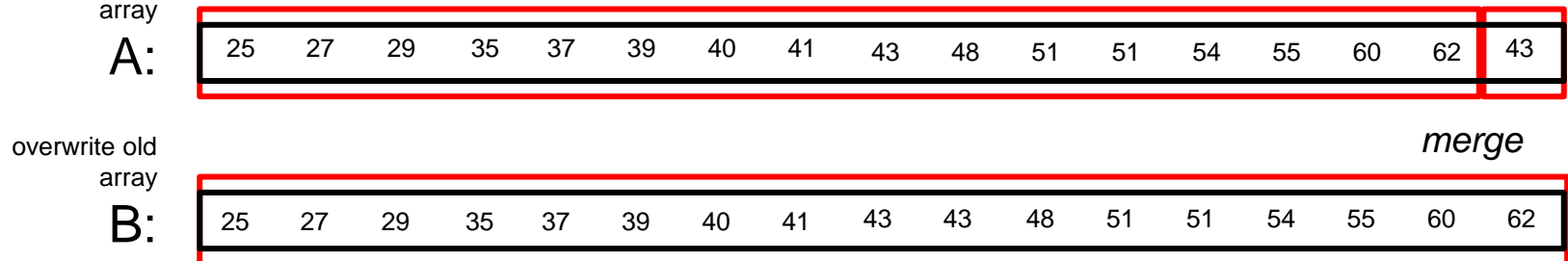
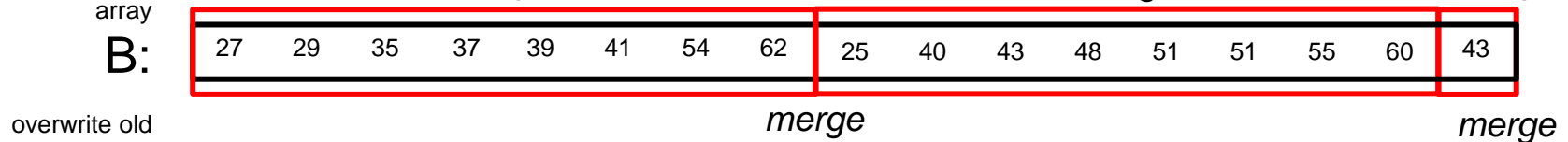
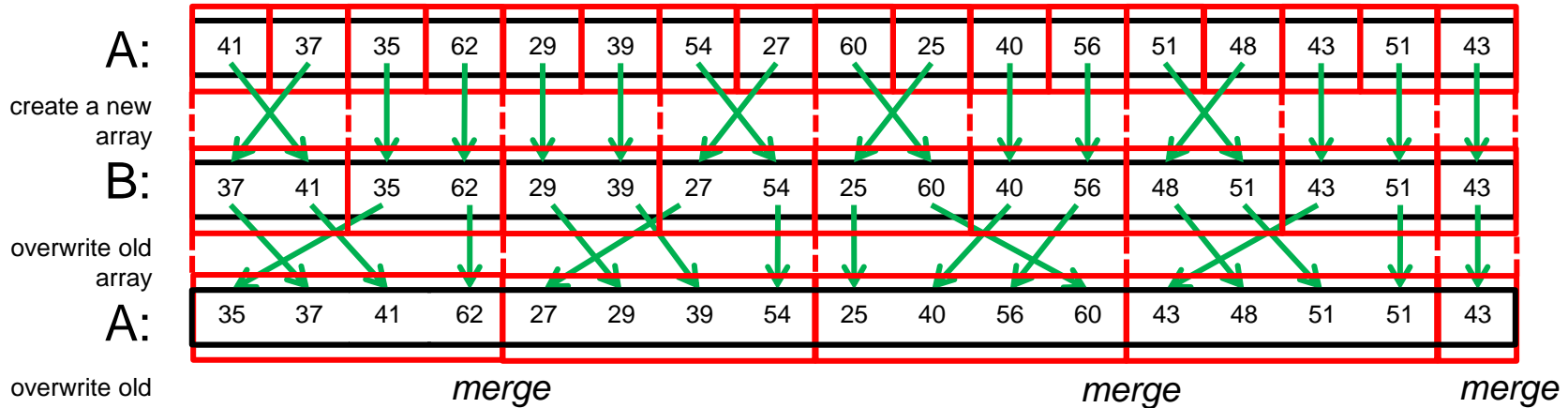
For each successive group of 8 cells ...

continuing until entire list is sorted.

Treat each cell in the array as a list of size 1

A: 

41	37	35	62	29	39	54	27	60	25	40	56	51	48	43	51	43
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# Next Lecture

Quicksort