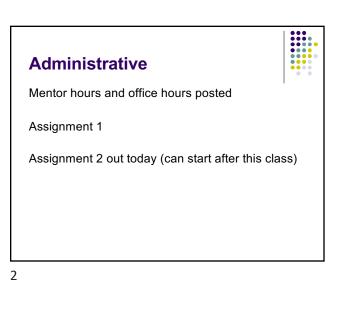




Divide and Conquer Divide: Break the problem into smaller sub-problems Conquer: Solve the sub-problems. Generally, this involves waiting for the problem to be small enough that it is trivial to solve (i.e. 1 or 2 items)

Combine: Given the results of the solved sub-problems, combine them to generate a solution for the complete problem





Divide and Conquer: some thoughts

Often, the sub-problem is the same as the original problem

Dividing the problem in half frequently does the job

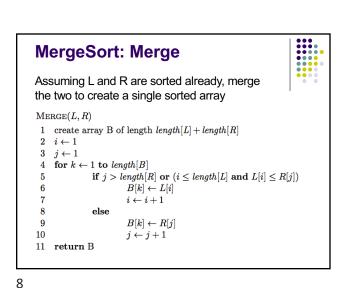
May have to get creative about how the data is split

Splitting tends to generate run times with log *n* in them

Divide and conquer One approach: - Pretend like you have a working version of your function, but it only works on smaller subproblems - If you split up the current problem in some way (e.g. in half) and solved those sub-problems, how could you then get the solution to the larger problem?

5

MergeSort Merge-Sort(A)1 if length[A] == 1 $\mathbf{2}$ return A 3 else 4 $q \leftarrow \lfloor length[A]/2 \rfloor$ $\mathbf{5}$ create arrays L[1..q] and R[q + 1.. length[A]]6 copy A[1..q] to L copy A[q+1.. length[A]] to R7 8 $LS \leftarrow \text{Merge-Sort}(L)$ $RS \leftarrow MERGE-SORT(R)$ 9 return MERGE(LS, RS) 10



Divide and Conquer: Sorting

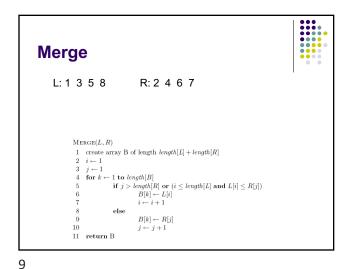
What are the sub-problems we need to solve?

How do we combine the results from these sub-

How should we split the data?

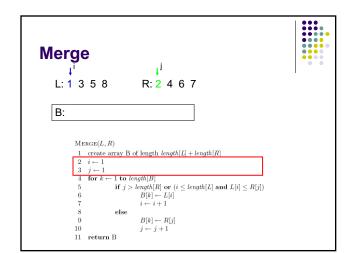
problems?

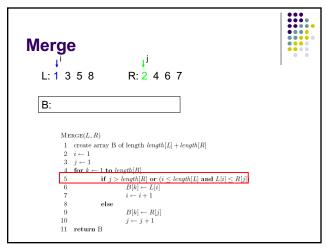
6

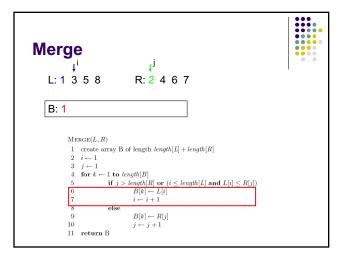


Merge L:1358 R:2467 B: MERGE(L, R)1 create array B of length length[L] + length[R]2 $i \leftarrow 1$ $j \leftarrow 1$ for $k \leftarrow 1$ to length[B] $\frac{3}{4}$ $\begin{array}{l} \text{if } j > length[D] \\ \text{if } j > length[R] \text{ or } (i \leq length[L] \text{ and } L[i] \leq R[j]) \\ B[k] \leftarrow L[i] \\ i \leftarrow i+1 \end{array}$ 5 6 7 8 \mathbf{else} $\begin{array}{l} B[k] \leftarrow R[j] \\ j \leftarrow j+1 \end{array}$ 9 10 11 return B

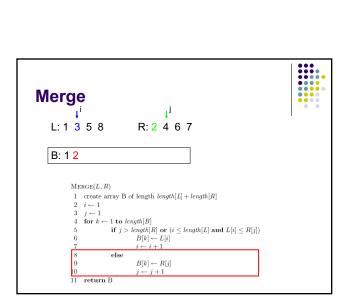
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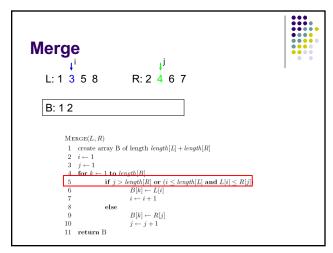






13





16

Merge

B: 1

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9

10

14

11 return B

L:1 3 5 8

MERGE(L, R)

↓^j R: 2 4 6 7

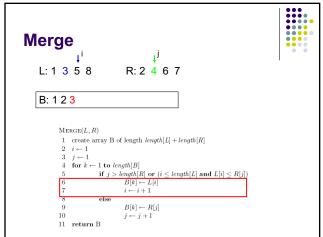
1 create array B of length length[L] + length[R]2 $i \leftarrow 1$ 3 $j \leftarrow 1$

 $\begin{array}{l} B[k] \leftarrow R[j] \\ j \leftarrow j+1 \end{array}$

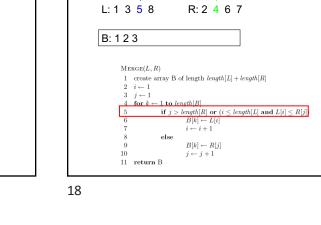
 $\begin{array}{l} \text{if } j > length[R] \text{ or } (i \leq length[L] \text{ and } L[i] \leq R[j] \\ \hline B[k] \leftarrow L[i] \\ i \leftarrow i+1 \end{array}$

for $k \leftarrow 1$ to length[B]

else



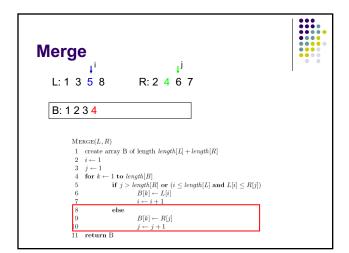
17

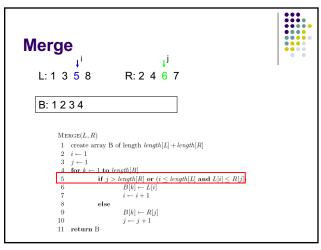


, tj

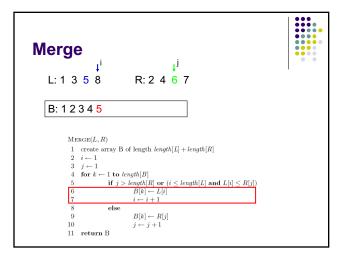
Merge

____↓ⁱ

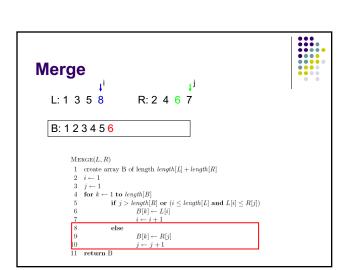


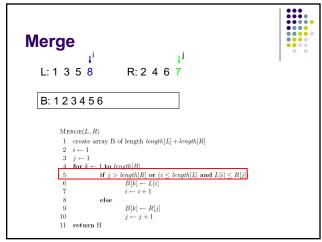


20









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 $\begin{array}{l} \text{if } j > length[R] \text{ or } (i \leq length[L] \text{ and } L[i] \leq R[j] \\ \hline B[k] \leftarrow L[i] \\ i \leftarrow i+1 \end{array}$

R:2467

1 create array B of length length[L] + length[R]2 $i \leftarrow 1$ 3 $j \leftarrow 1$

 $\begin{array}{l} B[k] \leftarrow R[j] \\ j \leftarrow j+1 \end{array}$

for $k \leftarrow 1$ to length[B]

else

24

Merge

_ ↓ⁱ

L:1358

B: 1 2 3 4 5

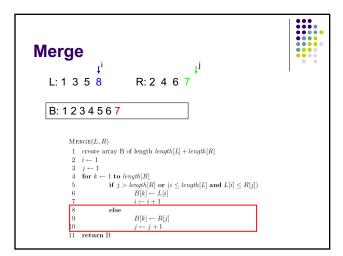
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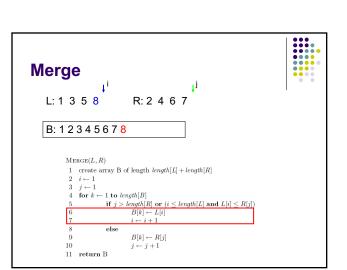
22

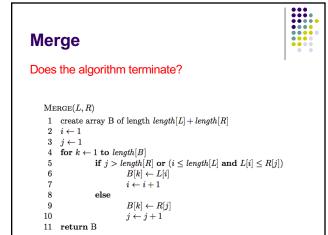
11 return B

MERGE(L, R)



25







Merge

L:1358

 $\begin{array}{ll} 2 & i \leftarrow 1 \\ 3 & j \leftarrow 1 \end{array}$

9

10 11 return B

26

B: 1234567

MERGE(L, R)

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 $\begin{array}{l} \text{if } j > length[R] \text{ or } (i \leq length[L] \text{ and } L[i] \leq R[j] \\ \hline B[k] \leftarrow L[i] \\ i \leftarrow i+1 \end{array}$

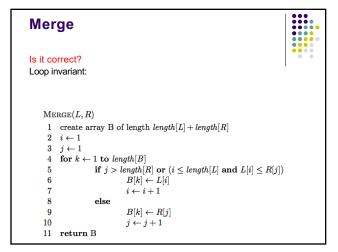
R:2467

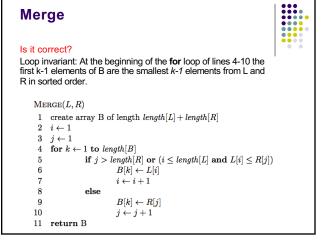
1 create array B of length length[L] + length[R]

 $\begin{array}{l} B[k] \leftarrow R[j] \\ j \leftarrow j+1 \end{array}$

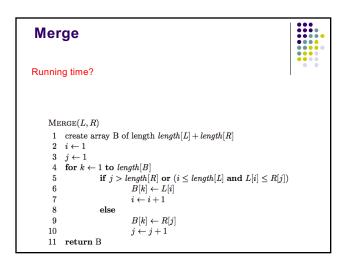
for $k \leftarrow 1$ to length[B]

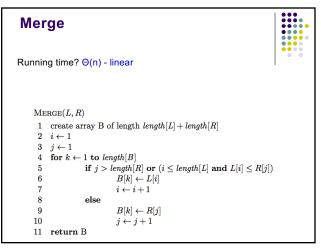
else

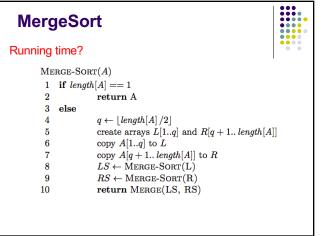


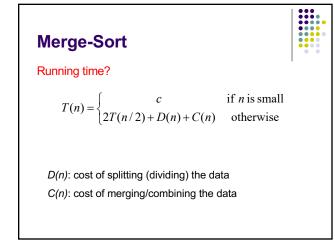


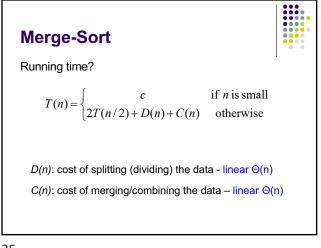


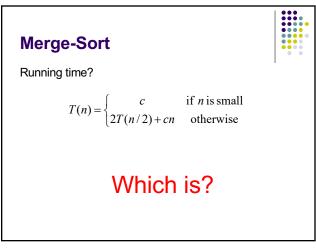


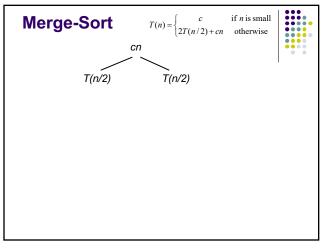


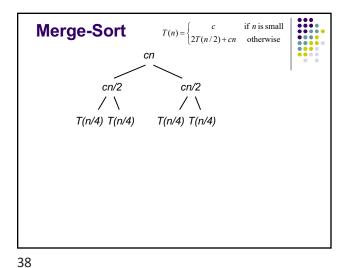


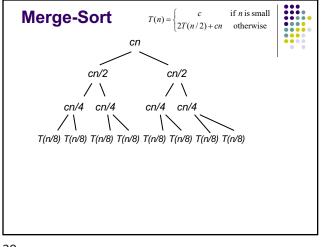


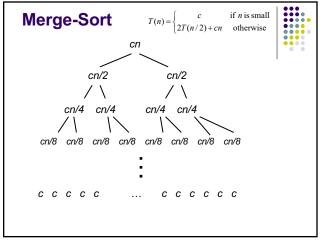


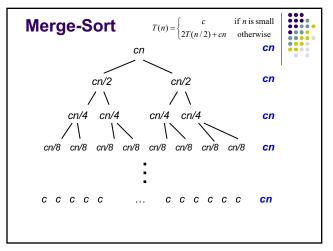


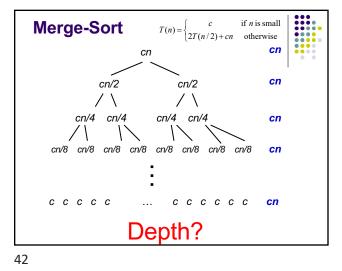


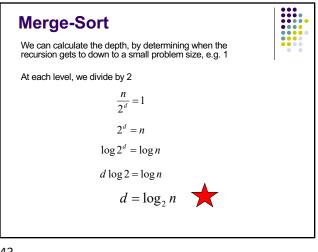


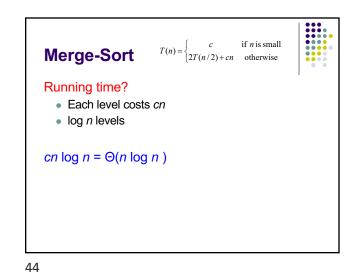












Recurrence

A function that is defined with respect to itself on smaller inputs

$$T(n) = 2T(n/2) + n$$
$$T(n) = 16T(n/4) + n$$
$$T(n) = 2T(n-1) + n^{2}$$

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Computational cost of divide and conquer algorithms

$$T(n) = aT(n/b) + D(n) + C(n)$$

• a subproblems of size n/b

- D(n) the cost of dividing the data
- *C*(*n*) the cost of recombining the subproblem solutions

In general, the runtimes of most recursive algorithms can be expressed as recurrences

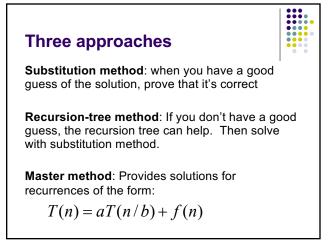
46

The challenge

Recurrences are often easy to define because they mimic the structure of the program

But... they do not directly express the computational cost, i.e. $n, n^2, ...$

We want to remove self-recurrence and find a more understandable form for the function



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Substitution method

Guess the form of the solution Then prove it's correct by induction

T(n) = T(n/2) + d

Halves the input then constant amount of work Guesses?

49

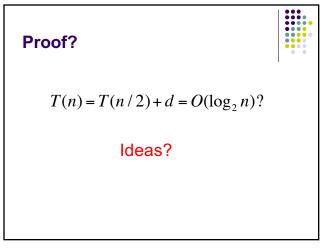
Substitution method

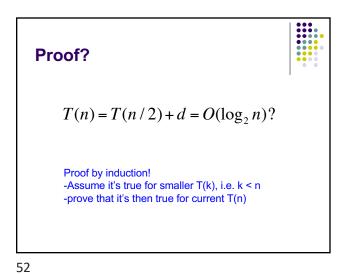
Guess the form of the solution Then prove it's correct by induction

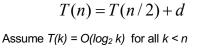
T(n) = T(n/2) + d

Halves the input then constant amount of work Similar to binary search: Guess: O(log₂ n)

50







Assume $T(k) = O(\log_2 k)$ for all k < rShow that $T(n) = O(\log_2 n)$

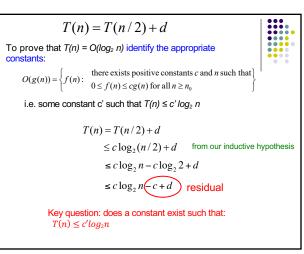
From our assumption, $T(n/2) = O(\log_2 n)$:

 $O(g(n)) = \begin{cases} f(n): & \text{there exists positive constants } c \text{ and } n \text{ such that} \\ 0 \le f(n) \le cg(n) \text{ for all } n \ge n_0 \end{cases}$

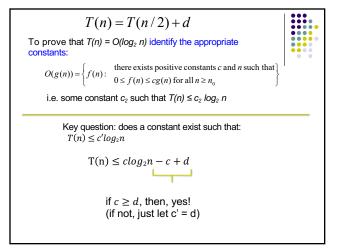
From the definition of big-O: $T(n/2) \le c \log_2(n/2)$

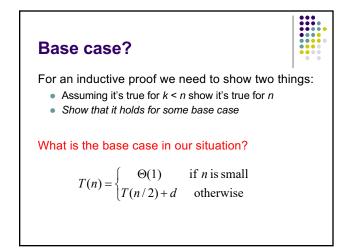
How do we now prove $T(n) = O(\log n)$?

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$$T(n) = T(n-1) + n$$

Guess the solution?

At each iteration, does a linear amount of work (i.e. iterate over the data) and reduces the size by one at each step $O(n^2)$

Assume $T(k) = O(k^2)$ for all k < n• again, this implies that $T(n-1) \le c(n-1)^2$ Show that $T(n) = O(n^2)$, i.e. $T(n) \le c'n^2$

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