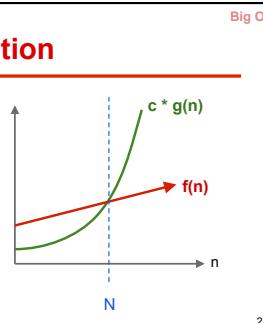


Recitation 11

Analysis of Algorithms and Inductive Proofs

Review: Big O definition

$f(n)$ is $O(g(n))$
iff
There exists $c > 0$ and $N > 0$ such that:
 $f(n) \leq c * g(n)$ for $n \geq N$



Example: $n+6$ is $O(n)$

$$\begin{aligned} & n + 6 \quad \text{---this is } f(n) \\ & \leq \quad \text{---if } 6 \leq n, \text{ write as} \\ & \quad n + n \\ & = \quad \text{---<arith>} \\ & \quad 2n \\ & \quad \quad \text{---choose } c = 2 \\ & = \quad c \cdot n \quad \text{---this is } c \cdot g(n) \end{aligned}$$

$f(n)$ is $O(g(n))$: There exist $c > 0, N > 0$ such that:
 $f(n) \leq c * g(n)$ for $n \geq N$

So choose $c = 2$ and $N = 6$

Review: Big O

Is used to classify algorithms by how they respond to changes in input size n .

Important vocabulary:

- Constant time: $O(1)$
- Logarithmic time: $O(\log n)$
- Linear time: $O(n)$
- Quadratic time: $O(n^2)$
- Exponential time: $O(2^n)$

Review: Big O

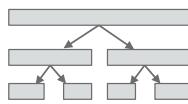
1. $\log(n) + 20$	is	$O(\log(n))$	(logarithmic)
2. $n + \log(n)$	is	$O(n)$	(linear)
3. $n/2$ and $3 \cdot n$	are	$O(n)$	
4. $n \cdot \log(n) + n$	is	$n \cdot \log(n)$	
5. $n^2 + 2 \cdot n + 6$	is	$O(n^2)$	(quadratic)
6. $n^3 + n^2$	is	$O(n^3)$	(cubic)
7. $2^n + n^5$	is	$O(2^n)$	(exponential)

Merge Sort

Merge Sort

```
/** Sort b[h..k]. */
public static void mS(Comparable[] b, int h, int k) {
    if (h >= k) return;
    int e = (h+k)/2;
    mS(b, h, e);
    mS(b, e+1, k);
    merge(b, h, e, k);
}

mS is mergeSort for readability
```



Merge Sort

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Merge Sort

```
/** Sort b[h..k]. */
public static void mS(Comparable[] b, int h, int k) {
    if (h >= k) return;
    int e = (h+k)/2;
    mS(b, h, e);
    mS(b, e+1, k);
    merge(b, h, e, k);
}

mS is mergeSort for readability
```

- We will *count* the number of comparisons mS makes
- Use $T(n)$ for the number of array element comparisons that mS makes on an array segment of size n

Merge Sort

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Merge Sort

```
/** Sort b[h..k]. */
public static void mS(Comparable[] b, int h, int k) {
    if (h >= k) return;
    int e = (h+k)/2;
    mS(b, h, e);
    mS(b, e+1, k);
    merge(b, h, e, k);
}

Use T(n) for the number of array element comparisons that mergeSort makes on an array of size n
```

$T(0) = 0$
 $T(1) = 0$

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Merge Sort

Merge Sort

```
/** Sort b[h..k]. */
public static void mS(Comparable[] b, int h, int k) {
    if (h >= k) return;
    int e = (h+k)/2;
    mS(b, h, e);           T(e+1-h) comparisons = T(n/2)
    mS(b, e+1, k);         T(k-e)   comparisons = T(n/2)
    merge(b, h, e, k);     How long does merge
    take?
}
```

Merge Sort

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Merge

```
pseudocode for merge
/** Pre: b[h..e] and b[e+1..k] are already sorted */
merge(Comparable[] b, int h, int e, int k)
Copy both segments
While both copies are non-empty
    Compare the first element of each segment
    Set the next element of b to the smaller value
    Remove the smaller element from its segment
    One comparison, one add, one remove

k-h loops must empty one segment      Runtime is O(k-h)
```

Merge Sort

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Merge

```
/** Sort b[h..k]. */
public static void mS(Comparable[] b, int h, int k) {
    if (h >= k) return;
    int e = (h+k)/2;
    mS(b, h, e);           T(e+1-h) comparisons = T(n/2)
    mS(b, e+1, k);         T(k-e)   comparisons = T(n/2)
    merge(b, h, e, k);     O(k-h)   comparisons =
    O(n)
}

Recursive Case:
T(n) = 2T(n/2) + O(n)
```

Merge Sort

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Merge Sort

Runtime

We determined that

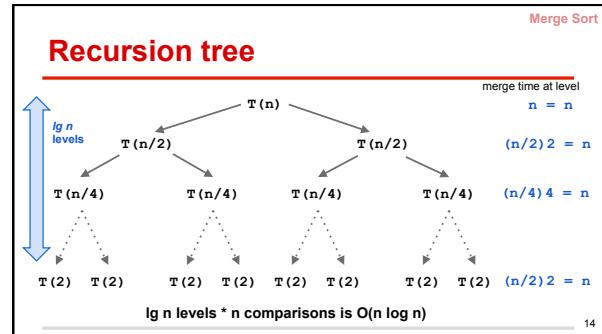
$$T(1) = 0$$

$$T(n) = 2T(n/2) + n \text{ for } n > 1$$

We will prove that

$$T(n) = n \log_2 n \text{ (or } n \lg n \text{ for short)}$$

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Merge Sort

Proof by induction

To prove $T(n) = n \lg n$, we can assume true for smaller values of n (like recursion)

$$\begin{aligned} T(n) &= 2T(n/2) + n \\ &= 2(n/2)\lg(n/2) + n \quad \text{Property of logarithms} \\ &= n(\lg n - \lg 2) + n \\ &= n(\lg n - 1) + n \\ &= n \lg n - n + n \quad \log_2 2 = 1 \\ &= n \lg n \end{aligned}$$

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Heap Sort

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Heap Sort

Very simple idea:

1. Turn the array into a max-heap
2. Pull each element out

```
/** Sort b */
public static void heapSort(Comparable[] b) {
    heapify(b);
    for (int i = b.length-1; i >= 0; i--) {
        b[i] = poll(b, i);
    }
}
```

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Heap Sort

```
/** Sort b */
public static void heapSort(Comparable[] b) {
    heapify(b);
    for (int i = b.length-1; i >= 0; i--) {
        b[i] = poll(b, i);
    }
}
```

Why does it have to be a max-heap?

The diagram shows an array represented as a tree. The root node is labeled "max-heap". Below it, another node is labeled "max-heap" with a "sorted" section to its right. Ellipses indicate more nodes below. The final node is labeled "sorted". A red arrow points from the question "Why does it have to be a max-heap?" to the "max-heap" label of the second node.

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Heap Sort runtime

```
/** Sort b */
public static void heapSort(Comparable[] b) {
    heapify(b);           ← O(n lg n)
    for (int i = b.length-1; i >= 0; i--) {
        b[i] = poll(b, i); ← loops n times
    }
}                                ← O(lg n)

Total runtime:
O(n lg n) + n * O(lg n) = O(n lg n)
```

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