

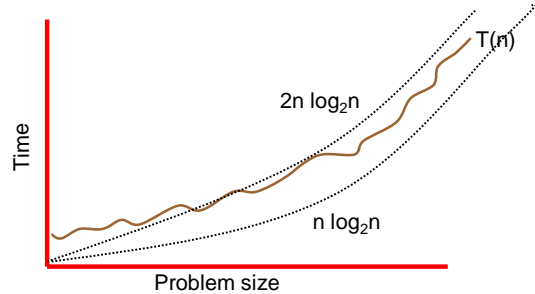
CSE 417: Algorithms and Computational Complexity

3: Complexity (cont.)

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Complexity



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O-notation etc

- Given two functions f and $g: \mathbb{N} \rightarrow \mathbb{R}$
 - $f(n)$ is $O(g(n))$ iff there is a constant $c > 0$ so that $c g(n)$ is eventually always $\geq f(n)$
 - $f(n)$ is $\Omega(g(n))$ iff there is a constant $c > 0$ so that $c g(n)$ is eventually always $\leq f(n)$
 - $f(n)$ is $\Theta(g(n))$ iff there are constants c_1 and $c_2 > 0$ so that eventually always $c_1 g(n) \leq f(n) \leq c_2 g(n)$

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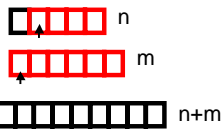
Example

- Mergesort
 - on a problem of size at least 2
 - Sort the first half of the numbers
 - Sort the second half of the numbers
 - Merge the two sorted lists
 - on a problem of size 1 do nothing

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Cost of Merge

- Given two lists to merge size n and m
 - Maintain pointer to head of each list
 - Move smaller element to output and advance pointer



Worst case $n+m-1$ comparisons
Best case $\min(n,m)$ comparisons

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Recurrence relation for Mergesort

- In total including other operations let's say each merge costs 3 per element output

$$T(n) = T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + 3n \quad \text{for } n \geq 2$$

$$T(1) = 1$$

- Can use this to figure out T for any value of n

$$\begin{aligned} T(5) &= T(3) + T(2) + 3 \times 5 \\ &= (T(2) + T(1) + 3 \times 3) + (T(1) + T(1) + 3 \times 2) + 15 \\ &= ((T(1) + T(1) + 6) + 1 + 9) + (1 + 1 + 6) + 15 \\ &= 8 + 10 + 8 + 15 = 41 \end{aligned}$$

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Insertion Sort

- For $i=2$ to n do
 - $j \leftarrow i$
 - while ($j > 1$ & $X[j] > X[j-1]$) do
 - swap $X[j]$ and $X[j-1]$
- i.e., For $i=2$ to n do
 - Insert $X[i]$ in the sorted list $X[1], \dots, X[i-1]$

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Recurrence relation for Insertion Sort

- Let $T(n,i)$ be the **worst case cost** of creating list that has first i elements sorted out of n .
 - We want $T(n,n)$
- The insertion of $X[i]$ makes up to $i-1$ comparisons in the worst case
- $T(n,i) = T(n,i-1) + i - 1$ for $i > 1$
- $T(n,1) = 0$ since a list of length 1 is always sorted
- Therefore $T(n,n) = n(n-1)/2$

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Solving recurrence relations

- e.g. $T(n) = T(n-1) + f(n)$ for $n \geq 1$
 $T(0) = 0$
 - solution is $T(n) = \sum_{i=1}^n f(i)$
- Insertion sort: $T_n(i) = T_n(i-1) + i - 1$
 - so $T_n(n) = \sum_{i=1}^n (i-1) = n(n-1)/2$

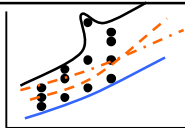
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Arithmetic Series

- $S = 1 + 2 + 3 + \dots + (n-1)$
- $S = (n-1) + (n-2) + (n-3) + \dots + 1$
- $2S = n + n + n + \dots + n$ { $n-1$ terms}
- $2S = n(n-1)$ so $S = n(n-1)/2$
- Works generally when $f(i) = ai + b$ for all i
- Sum = average term size \times # of terms

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Complexity analysis



- Problem size n
 - Worst-case complexity:** **max** # steps algorithm takes on any input of size n
 - Best-case complexity:** **min** # steps algorithm takes on any input of size n
 - Average-case complexity:** **avg** # steps algorithm takes on inputs of size n

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Why Worst-Case Analysis?

- Appropriate for time-critical applications, e.g. avionics
- Unlike Average-Case, no debate about what the right definition is
- Analysis often easier
- Result is often representative of "typical" problem instances
- Of course there are exceptions...

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