

Analysis of Algorithms

The theoretical study of computer programs performance and resource usage

Whats more important than perf?

Correctness, User friendliness, Scalability, efficiency, etc.

Why study algs and perf?

- Enables efficient, scalable software ..
- Helps in choosing optimal problem solving methods.
- Ensures programs run fast & use resources wisely
- Prepares for real world demands and constraints.

Problem: Sorting

Input: sequence $\langle a_1, a_2, \dots, a_n \rangle$
of numbers.

Output: permutation $\langle a'_1, a'_2, \dots, a'_n \rangle$
such that $a'_1 \leq a'_2 \leq \dots \leq a'_n$

Insertion sort

Insertion-Sort (A, n) // Sorts $A[1 \dots n]$

for $j \leftarrow 2$ to n

do $Key \leftarrow A[j]$

$i \leftarrow j - 1$

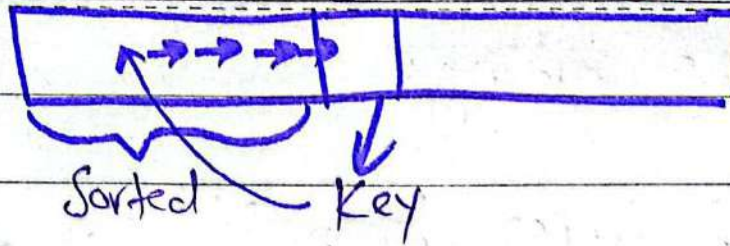
while $i > 0$ and $A[i] > Key$

do $A[i+1] \leftarrow A[i]$

$i \leftarrow i - 1$

$A[i+1] \leftarrow Key$

A:



Ex: $\rightarrow 8$ (2) 4 9 36

2 $\rightarrow 8$ (4) 9 36

2 4 8 (9) 36

2 4 8 9 (3) 6

2 3 4 8 9 (6)

2 3 4 6 8 9 done.

Running time

- Depends on input (e.g. already sorted, in reverse order)
- Depends on inputs size (6 elem vs 6×10^9)
 - parametrize in input size

talk about time as the function of the size of the input that

we are sorting.

- Want upper bounds

We want to know that the time is ^{no} more ^{than} certain amount and the reason is because that represents a guarantee to the user.

Kinds of Analysis. Focus.

- Worst-case (usually)

$T(n)$ = max time on any input of size n .

- Average case (sometimes)

$T(n)$ = expected time over all inputs of size n

- Expected type of input
- To know the type of input
(Need assumption of statistical distribution of inputs)

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Day: ___

No good

- Best-case (bogus)
 - You can cheat
 - A slow algorithm that works fast on some input

What is insertion sort's worst-case time?

Depends on computer.

- relative speed (on same machine)
- absolute speed (on diff. machines)

BIG IDEA!

Asymptotic Analysis

1. Ignore machine dependent constants
2. Look at growth of $T(n)$ as $n \rightarrow \infty$

In short, to look at the efficiency of an algorithm and correctness irrelevant of the power of a machine.

Asymptotic Notation

↳ Theta

- Θ - notation:

Drop lower-order terms

Ignore leading constants

Ex:

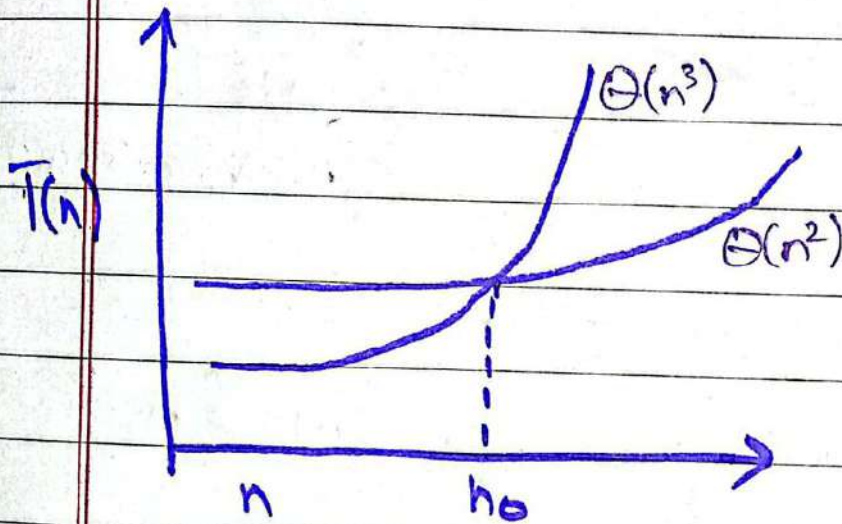
$$\begin{array}{c} \text{Leading} \nearrow \\ \text{constant} \end{array} \quad \underline{3n^3 + 90n^2 - 5n + 6046} \quad \begin{array}{c} \text{lower-order terms,} \\ \Rightarrow \Theta(n^3) \end{array}$$

— Represents the upper and the lower bounds of the running time of an algorithm.

— Used to represent Average-case complexity

— You add the running times for each possible input combination and take the average in the average case

- As $n \rightarrow \infty$, $\Theta(n^2)$ alg. always beats a $\Theta(n^3)$ alg.



- Even though alg. with $\Theta(n^3)$ time may be slower than algo. with $\Theta(n^2)$ time but they can be faster on reasonable size of inputs.

Insertion Sort Analysis:

Worst Case: input reverse sorted.

$$T(n) = \sum_{j=2}^n \Theta(j) = \Theta(n^2)$$

↑
(Arithmetic series)

Is insertion sort fast?

- Moderately so, for small n
- Not at all for large n .

Now, we look at a faster algorithm than insertion sort which is merge sort.

Merge Sort

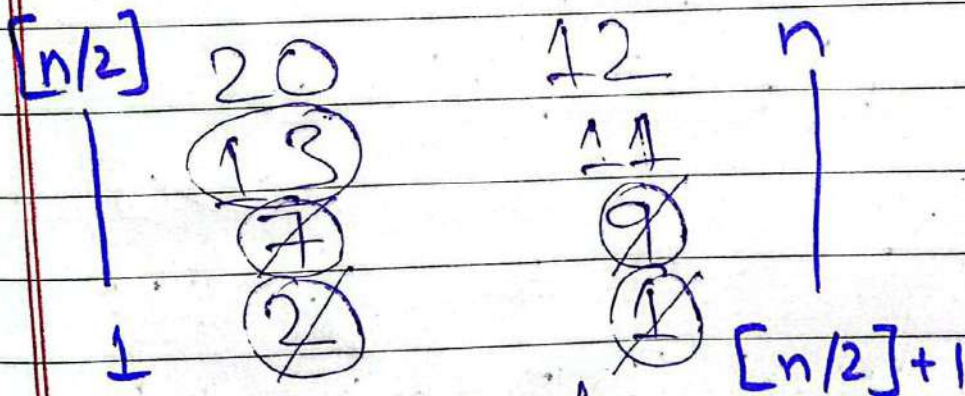
$T(n)$ Merge-Sort $A[1 \dots n]$

$\Theta(1)$ 1. If $n=1$, done

$2T(n/2)$ 2. Recursively sort $A[1 \dots \lfloor n/2 \rfloor]$ and $A[\lfloor n/2 \rfloor + 1 \dots n]$

$\Theta(n)$ 3. Merge 2 sorted lists.

Key Subroutine: Merge



Compare each element of both list like as $1 < 2, 2 < 9, 7 < 9, 9 < 13$
1 2 7 9 and so on

Time = $\Theta(n)$ on n total elements.

Recurrence

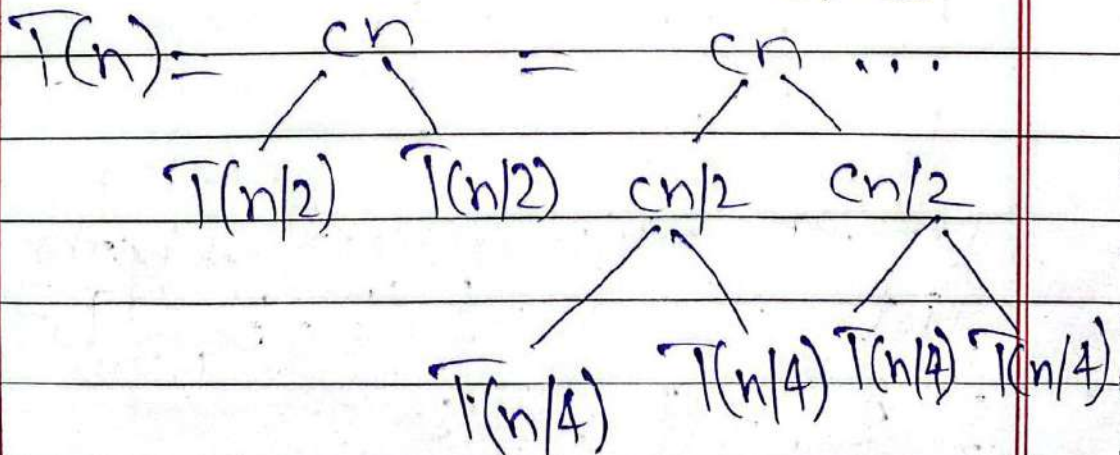
$$T(n) = \begin{cases} \Theta(1) & \text{if } n=1 \\ 2T(n/2) + \Theta(n) & \text{if } n>1 \end{cases}$$

[1] We usually omit ~~base~~ ^{base} cases in recurrence because if you're something on constant size input it takes constant time.

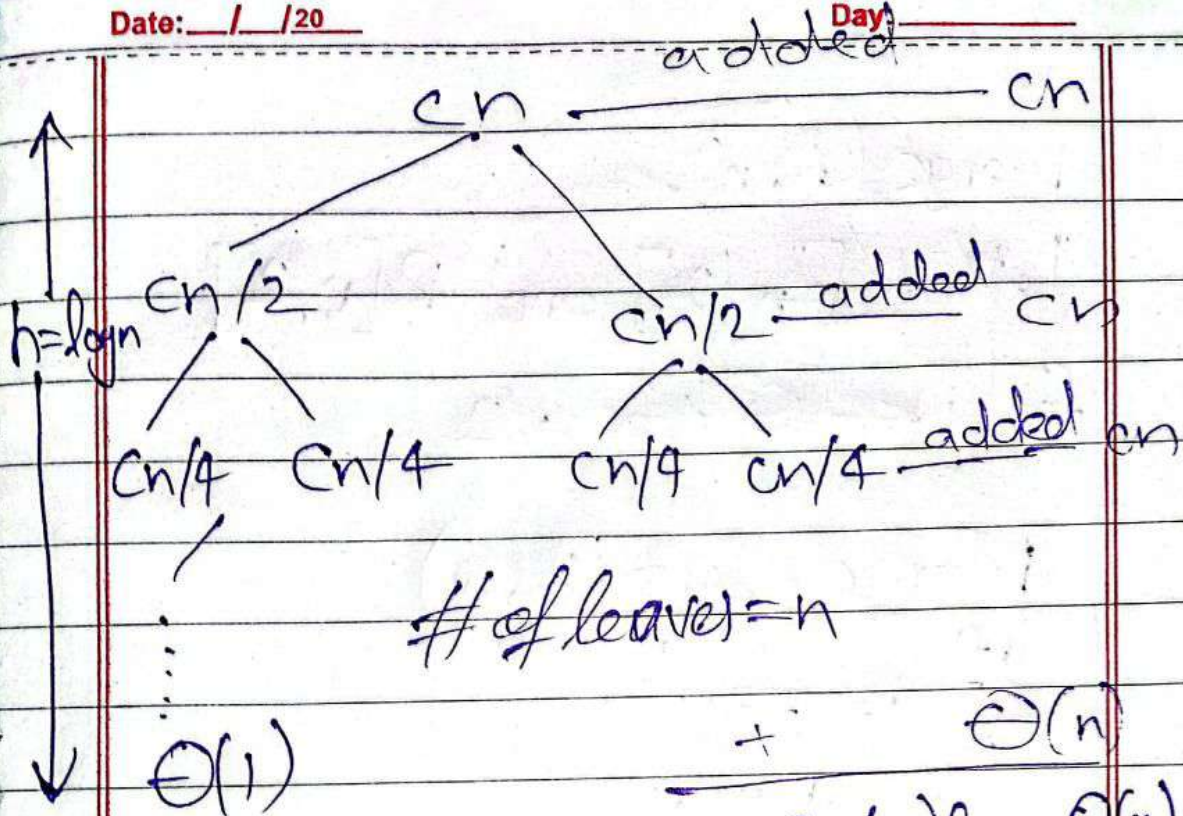
Recursion Tree:

$$T(n) = 2T(n/2) + cn$$

$c > 0$



Keep doing until you end up with.



$$\begin{aligned}
 & \quad \quad \quad + \quad \quad \quad \Theta(n) \\
 & \quad \quad \quad \hline
 \text{Total} = (cn) \lg n + \Theta(n) \\
 & \quad \quad \quad = \Theta(n \lg n)
 \end{aligned}$$

$\Theta(n \lg n)$ is faster than $\Theta(n^4)$.

That's why Merge sort is faster than insertion sort. This goes for all algs.

So algs with time $\Theta(n \lg n)$ like Merge sort will beat algs. with time $\Theta(n^2)$ like insertion sort on a large enough database.

$A = \text{array}$, $p = 1^{\text{st}} \text{ elem}$, $q = \text{midpoint}$, $r = \text{last elem}$

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Merge (A, p, q, r)

$$n_L \leftarrow q - p + 1$$

$$n_R \leftarrow r - q$$

let $L[0:n_L]$ and $R[0:n_R]$

for $i \leftarrow 0$ to n_L

$$L[i] \leftarrow A[p+i]$$

for $j \leftarrow 0$ to n_R

$$R[j] \leftarrow A[q+j+1]$$

$$i \leftarrow 0$$

$$j \leftarrow 0$$

$$k \leftarrow p$$

while $i < n_L$ and $j < n_R$

$$\text{if } L[i] \leq R[j]$$

$$A[k] \leftarrow L[i]$$

$$i \leftarrow i + 1$$

$$\text{else } A[k] \leftarrow R[j]$$

$$j \leftarrow j+1$$
$$k \leftarrow k+1$$

while $i < n_L$
 $A[k] \leftarrow L[i]$
 $i \leftarrow i+1$
 $k \leftarrow k+1$

while $j < n_R$
 $A[k] \leftarrow R[j]$
 $j \leftarrow j+1$
 $k \leftarrow k+1$

Delete Left and Right.

Merge-Sort (A, p, r)

if $p \geq r$ return

$$q = (p + \overset{r}{r}) / 2$$

Merge-Sort (A, p, q)

Merge-Sort ($A, q+1, r$)

Merge (A, p, q, r)