

- Exam 1 : Mon, Feb 20.
- No lecture on Mon, Feb 20.
- Recitation on Saturday.

Sorting.

Input: Array A of n distinct integers.

Objective: Sort A in ascending order.

IS ($A[1..n]$)

$\xrightarrow{\text{for } j \leftarrow 2 \text{ to } n \text{ do}} \rightarrow n \text{ times.}$
 $\quad \text{key} \leftarrow A[j] \quad \left. \begin{array}{l} \text{key} \leftarrow A[j] \\ i \leftarrow j-1 \end{array} \right\} \text{const } O(1)$
 $\quad i \leftarrow j-1$

$\rightarrow \text{while } i > 0 \text{ and } A[i] > \text{key do}$
 $\quad A[i+1] \leftarrow A[i] \quad \left. \begin{array}{l} A[i+1] \leftarrow A[i] \\ i \leftarrow i-1 \end{array} \right\} \text{const } O(1)$
 $\quad i \leftarrow i-1$
 $\quad A[i+1] \leftarrow \text{key}$

$O(n^2)$ in the worst case

Example .

A :

18	<u>3</u>	24	16
1	2	3	4

$j = \cancel{2} \ 3 \ 4$

key : $\cancel{3} \ \cancel{24}^{16}$

$i = \cancel{1} \ \cancel{0} \ 2$

$3 \neq 1$

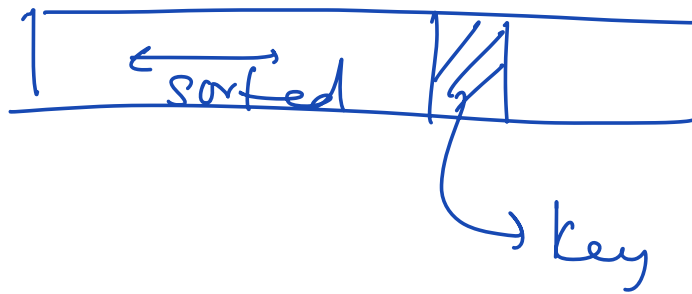
A :

3 18	18	24	16
1	2	3	4

3	<u>18</u>	24	24
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3	18 16	18	24
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→



Running time :

Best case : $\Theta(n)$.
 ↳ tight bound.

Worst case : $\Theta(n^2)$.

Asymptotic notation.

Ω ↳ asymptotic lower bound
 Θ ↳ tight bound.
 O , Σ , \ominus , o , ω .

↳ asymptotic upper bound.
asymptotic
 $n \rightarrow \infty$

Rest of the course: ^{worst case input}
↑
worst-case
running time of algorithm.

$$T(n) = O(n^2),$$

- in the worst case

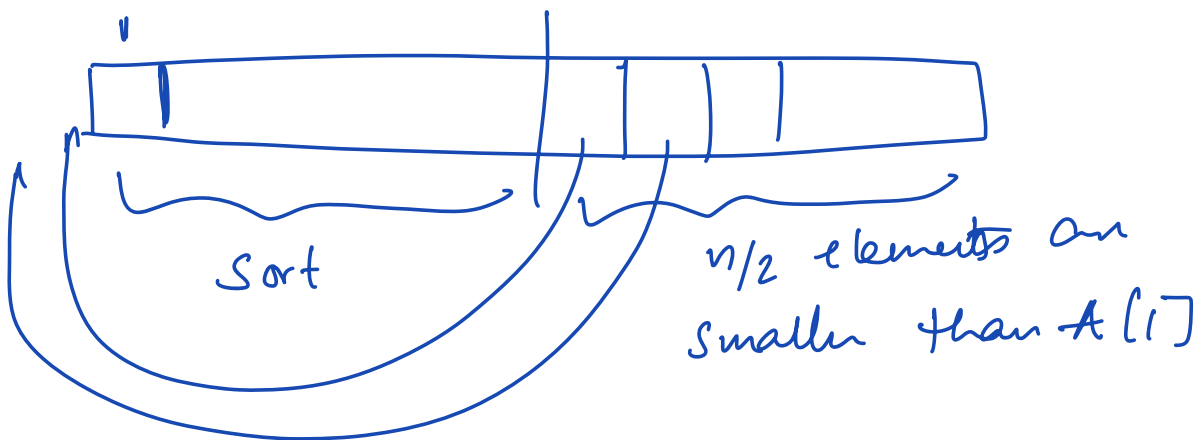
IS will not take more
than n^2 time within
constants -

$$\underline{\underline{7n^2}} + \underline{\underline{18n}} - 500$$

Why is $T(n) = \Omega(n^2)$?

Why is it true that

It takes $\approx n^2$ time in
the worst case?



Each item will need to make $\geq n/2$ swaps. There are $n/2$ such elements.

$$\therefore T(n) \geq \frac{n}{2} \cdot \frac{n}{2} = \frac{n^2}{4}$$

$$= \Omega(n^2).$$

$$\therefore T(n) = \Theta(n^2).$$

Suppose there is an alg. A
for which $T(n) = O(n^3)$, ✓
 $T(n) = \Omega(n^2)$.

$$T(n) = \Theta(\underline{n^2}) \quad \checkmark \checkmark$$

$$\Theta(\underline{n^3}) \quad \checkmark \checkmark$$

$$\Theta(n^{2.5}) \quad \checkmark \checkmark$$

$$\begin{array}{l} \checkmark \checkmark \\ \neq \end{array} \quad \begin{array}{l} O \\ \Theta \end{array}$$

$$n^{1.5} = O(n^2)$$

$$\begin{array}{l} \checkmark \\ \neq \end{array} \quad \begin{array}{l} \Omega \\ \Theta \end{array}$$

$$\underline{n^{1.5}} = \omega(n)$$

$$\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \begin{matrix} 0 & \Rightarrow & f(n) = o(g(n)) \\ \infty & \Rightarrow & f(n) = \omega(g(n)) \\ c & \Rightarrow & f(n) = \Theta(g(n)) \end{matrix}$$

RAM model of computation.

- all high-level operations take Const time.

- Comparisons, Arithmetic, return, fn. call, array indexing.

- all int & nos can be stored in a word of memory.
- all memory access takes const time.

Divide & Conquer.

Input: int $n \geq 0$

Obj: To compute 2^n .

① $\text{pow}(\text{int } n) \quad T(n)$

if $n = 0$ then $\left. \begin{array}{l} \text{return } 1 \end{array} \right\} O(1)$

else

return $2 * \text{pow}(n-1)$ $T(n-1)$

$n = 100$

1

2^{n-1}

$\text{pow}(n-1)$

③

poz (int n)

if n=0 then

return 1

else

return poz(n-1) + poz(n-1)

~~>> 100 million years~~
~~≈ 1~~
 why is that bad!?

④

poz (int n)

$T(n)$

if n=0 then

return 1

$O(1)$

≈ 1.2

else

Dynamic Prog.

x ← poz(n-1) $T(n-1)$
 return x + x c

⑤

poz (int n) → $T(n)$

if n=0 then

return 1

else

$tmp \leftarrow pow(\lfloor n/2 \rfloor) \quad T(n/2)$

if n is even then

return $tmp * tmp$

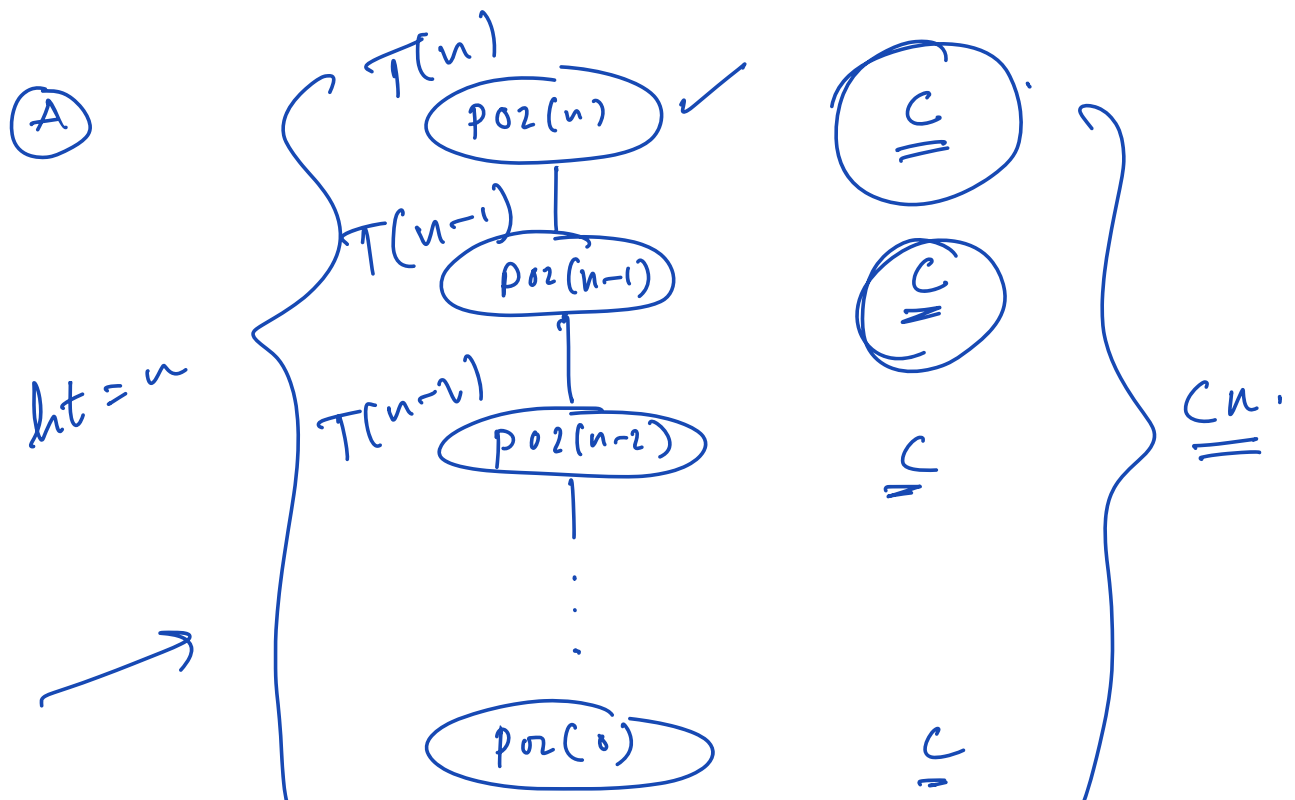
} $O(1)$

else

return $tmp * tmp * 2$.

Runtime analysis.

$T(n)$: worst case running time of pow on an input of size n .



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

$$\quad \quad \quad \downarrow$$

$$= (T(n-2) + c) + c$$

$$\rightarrow = T(n-3) + 3c$$

⋮

$$= T(n-k) + kc$$

Recursion bottoms out when

$n-k = 0$, i.e., when $k=n$.

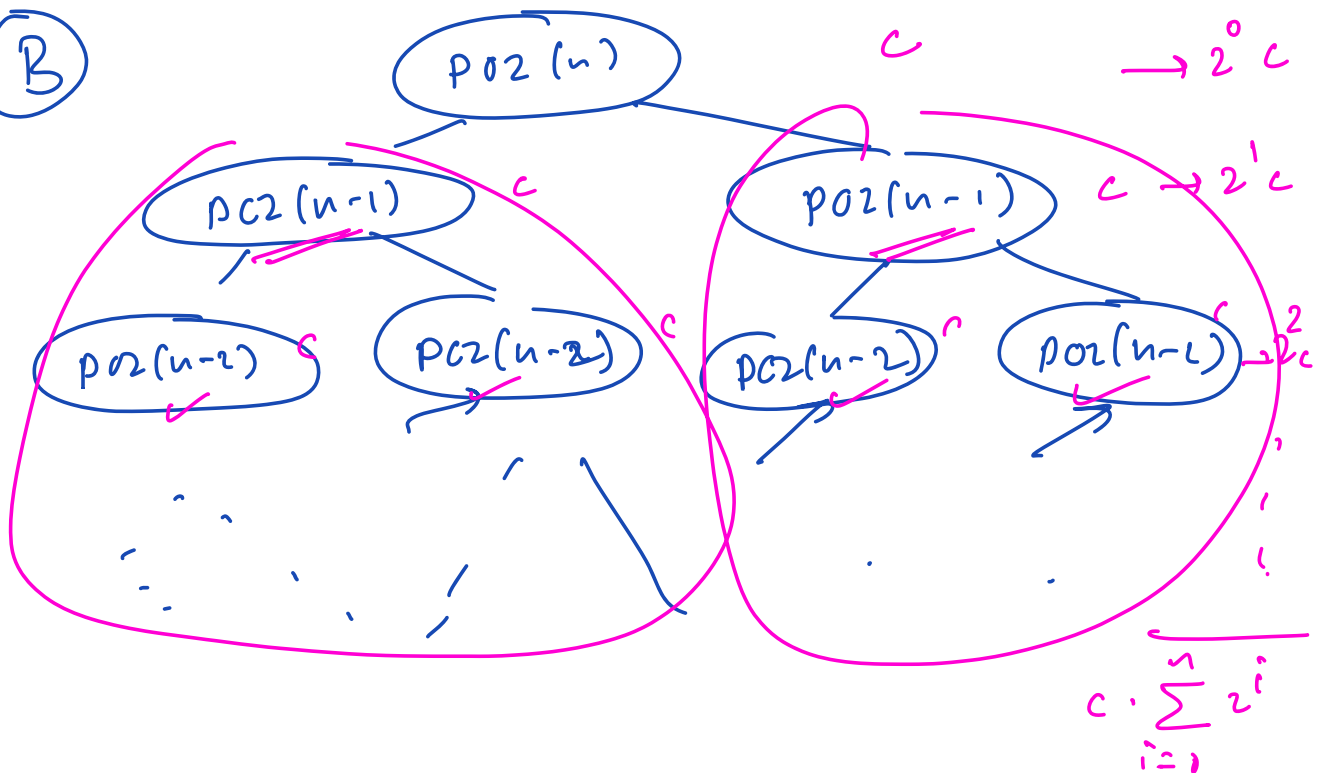
When that happens, we have

$$T(n) = T(0) + nc$$

$$= O(1) + n c$$

$$= \boxed{\Theta(n)}$$

(B)



Complete bin. tree with ht. n .

This has $\Omega(2^n)$ nodes.

$$T(n) = \begin{cases} O(1) , & n=0 \text{ (Base case)} \\ 2T(n-1) + c , & \text{otherwise.} \end{cases}$$

$$T(n) = 2T(n-1) + c$$

$$= 2(2T(n-2) + c) + c$$

$$= 2^2 T(n-2) + 2^1 c + 2^0 c$$

$$= 2^2 (2T(n-3) + c) + 2^1 c + 2^0 c$$

$$= 2^3 T(n-3) + 2^2 c + 2^1 c + 2^0 c$$

⋮

$$= 2^k T(n-k) + c \cdot \sum_{i=0}^{k-1} 2^i$$

Recursion bottoms out when $n-k \geq 0$,
i.e., $k = n$.

When this happens, we have

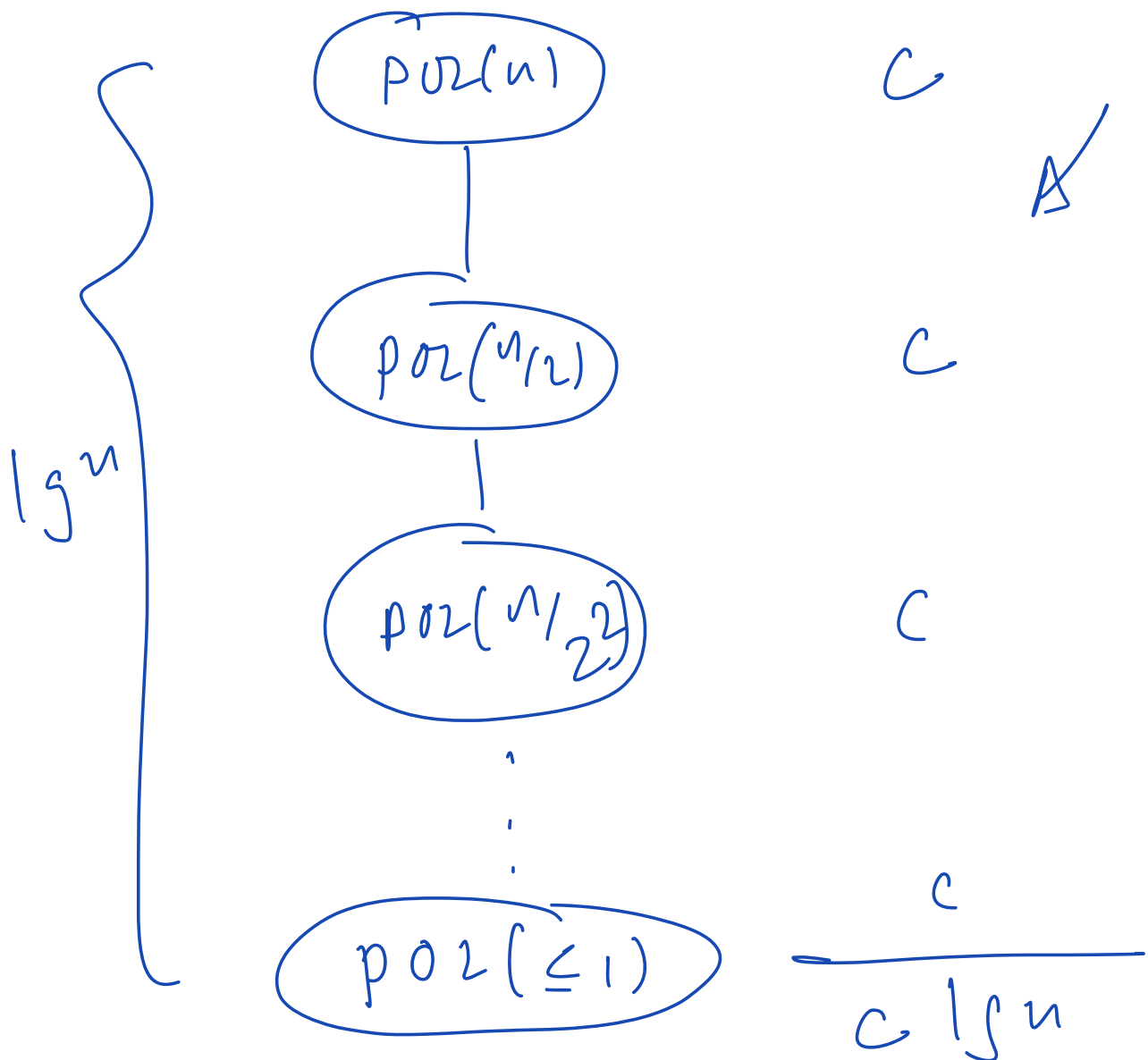
$$\begin{aligned} T(n) &= 2^n T(0) + c \cdot \sum_{i=0}^{n-1} 2^i \\ &= 2^n O(1) + c \cdot (2^n - 1) \\ &= \Theta(2^n). \end{aligned}$$

$$\textcircled{c} \cdot T(n) = \begin{cases} O(1), & n=0 \\ T(n-1) + c, & \text{o.w.} \end{cases}$$

Same as \textcircled{A} .

④ Assuming n is an exact power of 2.

$$T(n) = \begin{cases} O(1), & \text{if } n=0 \\ T\left(\frac{n}{2}\right) + c, & \text{o.w.} \end{cases}$$



$$T(n) = T\left(\frac{n}{2}\right) + \underline{\underline{C}}$$

$$= T\left(\frac{n}{2^2}\right) + 2C$$

$$= T\left(\frac{n}{2^3}\right) + 3C$$

⋮

$$= T\left(\frac{n}{2^k}\right) + kC$$

Recursion bottoms out when

$$\frac{n}{2^k} < 1 \Rightarrow k > \lg n.$$

$$\therefore T(n) \leq T(0) + (\lg n + 1) c.$$

$$= \underline{\underline{O(\lg n)}}.$$

Input: - Array A of n distinct int.

- int k

Obj: To output Yes, if $k \in A$.
No, o.w.

LS (A[1..n], k)

if $n == 1$ then
return A[1] == k.

else

return $(A[n] == k)$ or $(LS(A[1..n-1], k))$
 $\underbrace{\hspace{1.5cm}}_{C} \quad \underbrace{\hspace{1.5cm}}_{T(n-1)}$

Runtime recurrence.

$T(n)$: worst case running time of LS
on an i/p of size n .

$$T(n) = \begin{cases} O(1), & n = 1 \\ T(n-1) + C. \end{cases}$$

$$T(n) = \Theta(n).$$