

Analysis of Algorithms:

theoretical study of computer program performance and resource usage.

What's more important than performance?

- correctness, simplicity, maintainability, modularity, security, ...
- user friendliness

Why study algo performance

- feasible vs. infeasible
- currency to pay for other important properties
- Bottom of the heap

Sorting Problem

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

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

s.t. (\$) $a'_1 \leq a'_2 \leq \dots \leq a'_n$

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

for $j \leftarrow 2$ to n

do $key \leftarrow A[j]$

$i \leftarrow j-1$

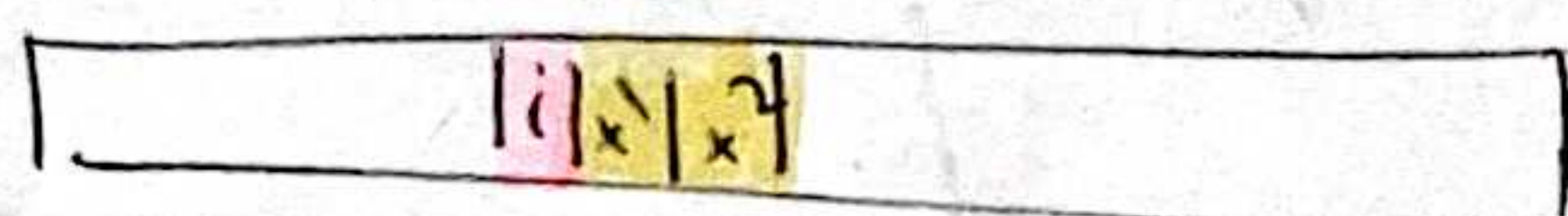
while $i > 0$ and $A[i] > key$

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

$i \leftarrow i-1$

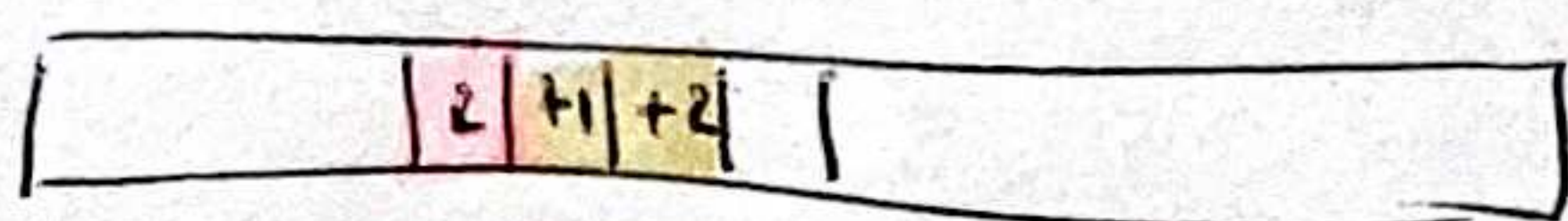
$A[i+1] \leftarrow key$

after completing a while loop iteration

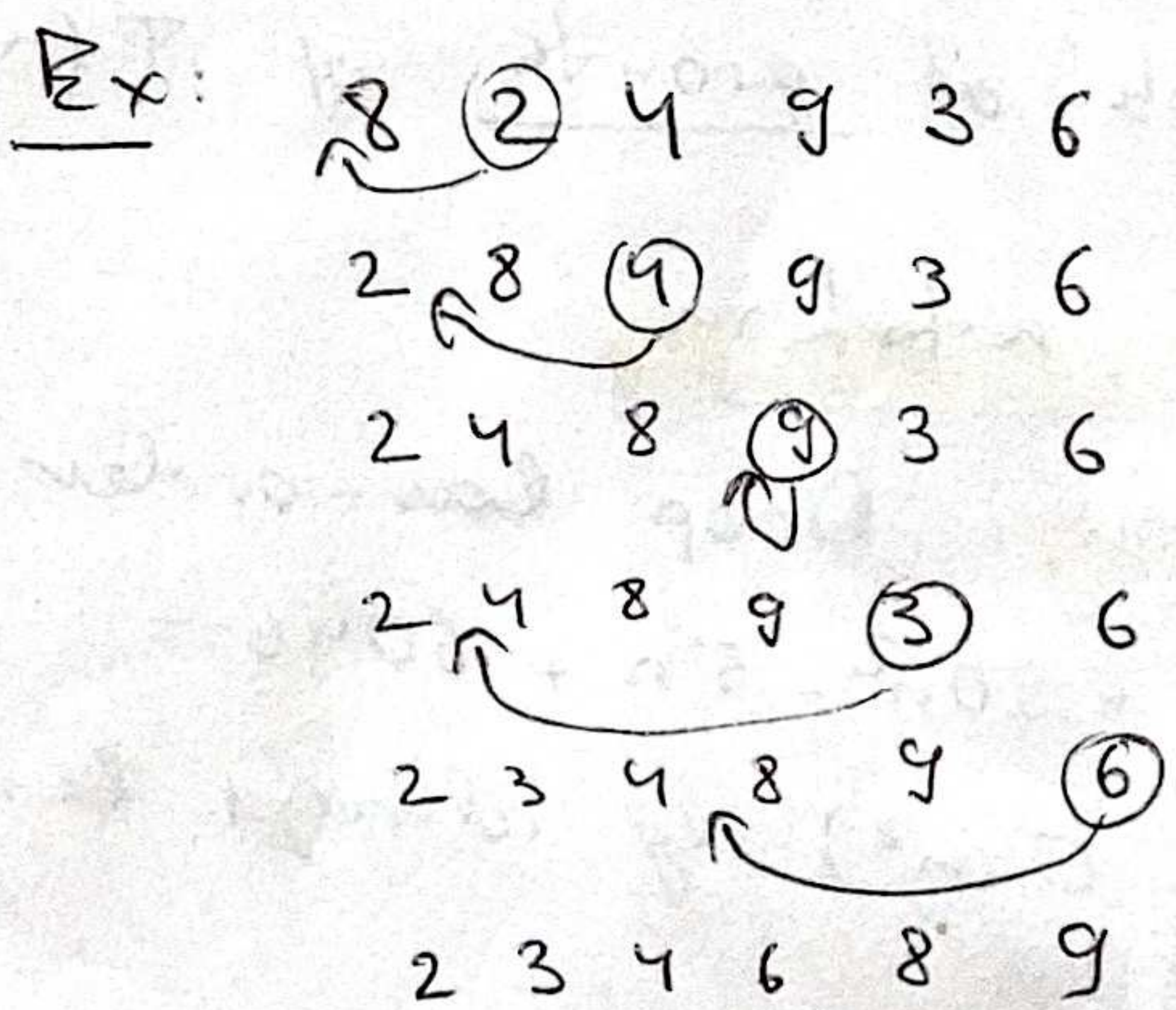
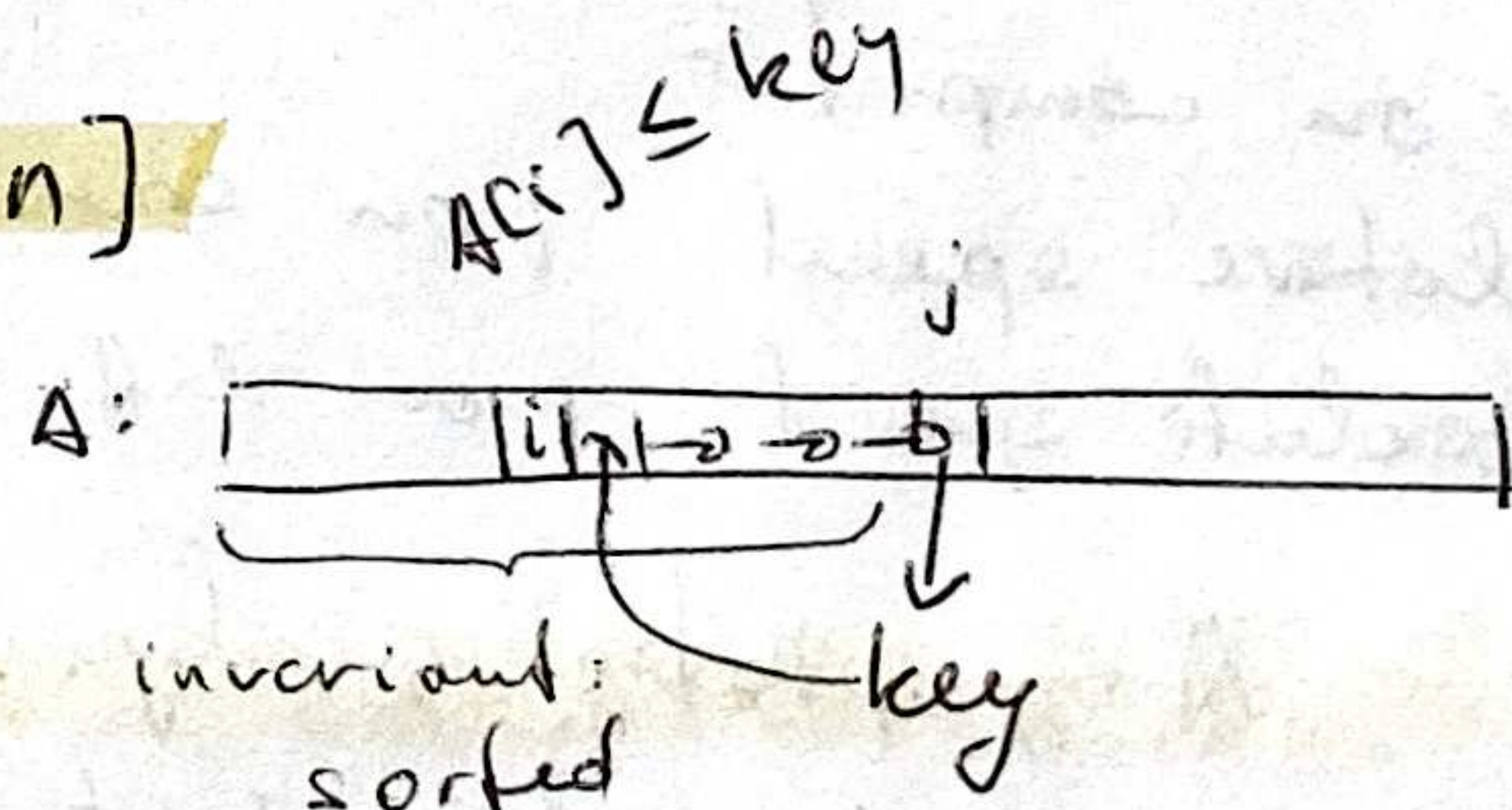
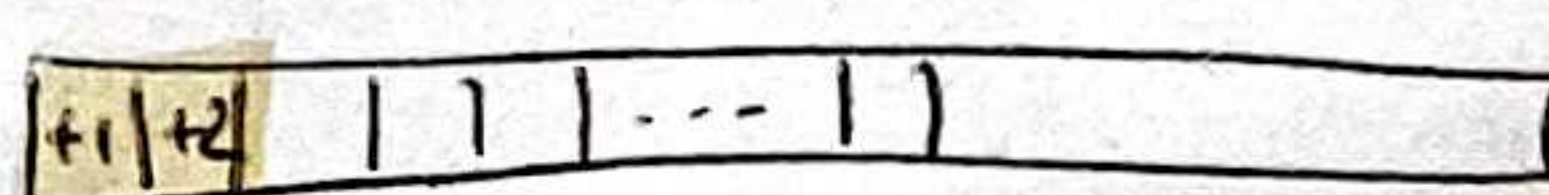


$A[i+1] = A[i+2]$

- so key can be inserted into $A[i+1]$ if $A[i] \leq k$
- or $A[i]$ can be copied into $A[i+1]$ to continue



border case



Running time

- depends on input (already sorted, or reverse-sorted)
- depends on input size (6 elem. vs. 6×10^9)
 - parameterize in input size
- want upper bounds
 - represents a guarantee to user

Kinds of analysis

Worst-case (usually):

$T(n)$ = max time on any input of size n . max turns relation to function

Average case (sometimes):

$T(n)$ = expected time over all inputs of size n
(need an assumption of statistical distribution of inputs)

Best case: (bogus)

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

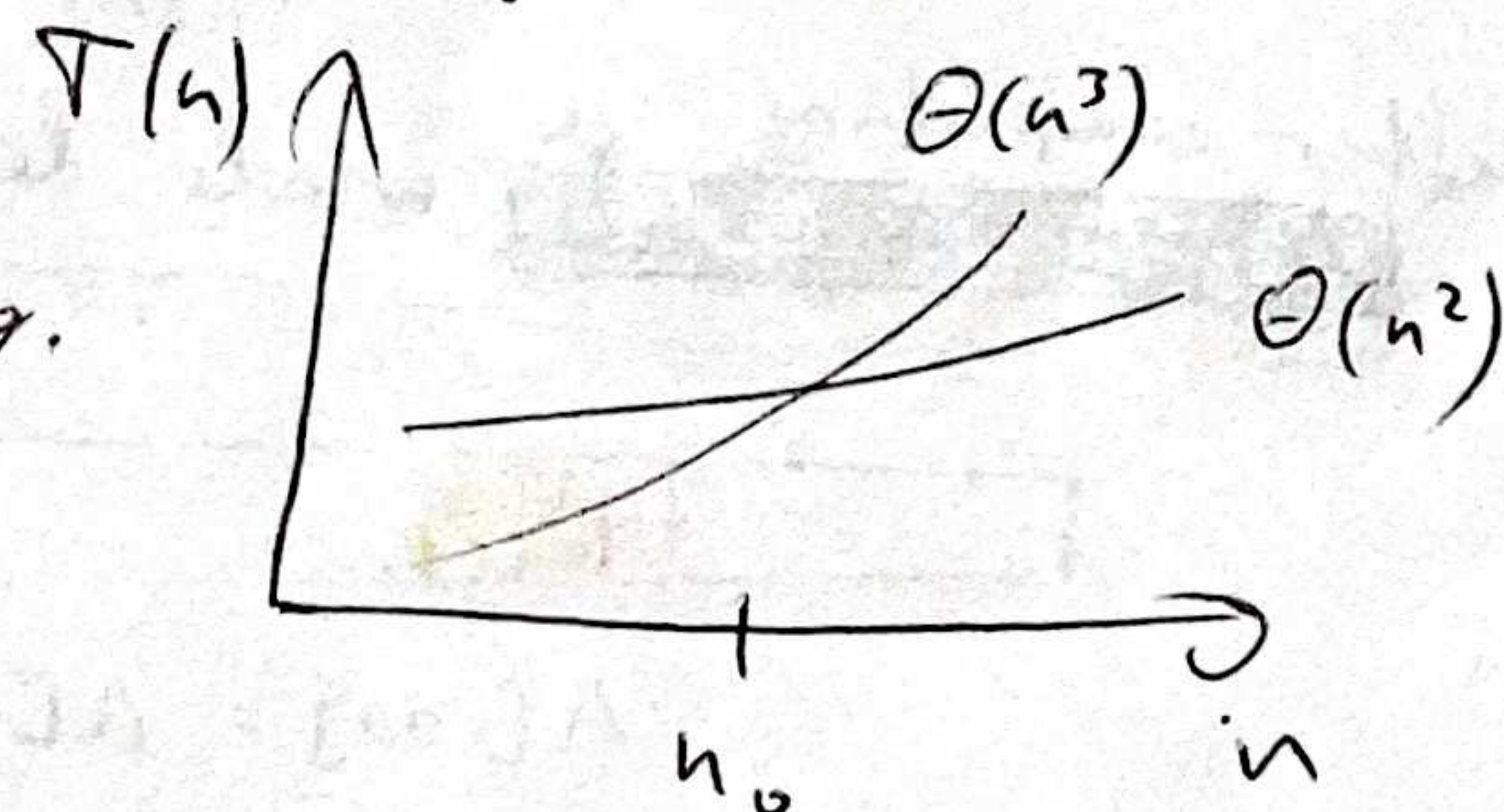
- ignore machine-dependent constants
- look at growth of $T(n)$ as $n \rightarrow \infty$

Asymptotic notation

Θ -notation: Drop low-order terms, ignore leading constants

Ex: $3n^3 + 90n^2 - 5n + 6046 = \Theta(n^3)$

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



Insertion sort analysis

worst case: input reverse-sorted

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

arithmetic series

$$\sum_{k=1}^n k = \frac{n(n+1)}{2}$$

0, 1, ..., 99, 100

50 times (2)

Is insertion sort fast?

- moderately so
- not for large n

Merge sort

$$T(n)$$

abuse of notation $\rightarrow \Theta(1)$

$$\text{sloppy} \rightarrow 2T(n/2)$$

$$\Theta(n)$$

Merge sort $A[1 \dots n]$

1. If $n=1$ done
2. recursively sort $A[1 \dots \lceil n/2 \rceil]$ and $A[\lceil n/2 \rceil + 1 \dots n]$
3. merge 2 sorted lists

Recurrence:

usually omit const. base case

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

key subroutine: Merge

20 12
13 11
7 9
2 1

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

Recursion tree:

$$T(n) = 2T(n/2) + cn, \quad c > 0$$

$$T(n) = \begin{array}{c} cn \\ / \quad \backslash \\ T(n/2) \quad T(n/2) \end{array}$$

$$= \begin{array}{c} cn \\ / \quad \backslash \\ cn/2 \quad cn/2 \\ / \quad \backslash \quad / \quad \backslash \\ T(n/4) \quad T(n/4) \quad T(n/4) \quad T(n/4) \end{array}$$

$$= \dots = \begin{array}{c} cn \\ / \quad \backslash \\ cn/2 \quad cn/2 \\ / \quad \backslash \quad / \quad \backslash \\ cn/4 \quad cn/4 \quad cn/4 \quad cn/4 \\ \vdots \quad \vdots \quad \vdots \quad \vdots \\ \Theta(1) \end{array}$$

cn
 cn
 \vdots
 cn
 $\log n$
 $\Theta(n)$

optimally combine

insertion sort at lower levels

on a large n
merge sort is faster than
insertion sort $\sim n > 30$

leaves = n

$$\text{Total: } cn \log n + \Theta(n) = \Theta(n \log n)$$