

Lecture 2:

From last time, running time of insertion sort is

$$2n - 1 + 3 \sum_{j=2}^n t_j$$

where $t_j = \#$ times we execute the loop

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

starting with $i = j-1$.

when $i = j-1$

- Best Case: exit loop because $A[i] \leq A[j] \Rightarrow t_j = 1$
- Worst case: exit loop because $i = 0 \Rightarrow t_j = j$
- Avg Case: $t_j \approx j/2$

Best Case: $2n - 1 + 3 \sum_{j=2}^n 1 = 2n - 1 + 3(n - 1) = 5n - 4 = \Theta(n)$

Worst Case:

$$\begin{aligned} 2n - 1 + 3 \sum_{j=2}^n j &= 2n - 1 + 3 \left(\sum_{j=1}^n j - 1 \right) \\ &= 2n - 1 + 3 \sum_{j=1}^n j - 3 \\ &= 2n - 4 + 3 \frac{n(n+1)}{2} \end{aligned}$$

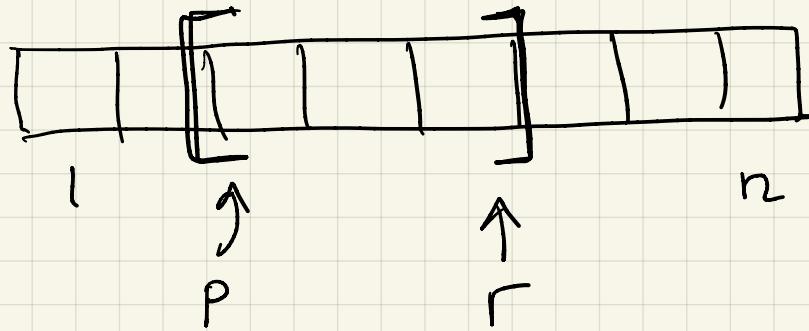
Note: $\sum_{j=1}^n j = 1 + 2 + 3 + \dots + n = n(n+1)/2$

$$2n - 4 + 3 \frac{n(n+1)}{2} = \Theta(n^2)$$

Merge Sort based on divide and conquer.

Sort subproblems, and combine them in some way.

$A[p \dots r]$ defines a subproblem



p & r change over time

of course we start with $p=1$, $r=n$

Divide: Split array $A[p \dots r]$ into two subarrays

$A[p \dots q]$ $A[q+1 \dots r]$, choose q

to be half-way between p and r

Merge Sort (A, p, r)

if $p < r$

then $q \leftarrow \lfloor \frac{p+r}{2} \rfloor$

Conquer step: Recursively sort $A[p \dots q]$ & $A[q+1 \dots r]$

Combination: Merge the two sorted arrays

$A[p \dots q]$ and $A[q+1 \dots r]$ into
one sorted array.

Merge Sort (A, p, q)

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

Merge (A, p, q, r)

Example:

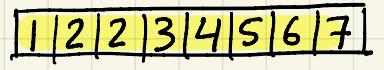
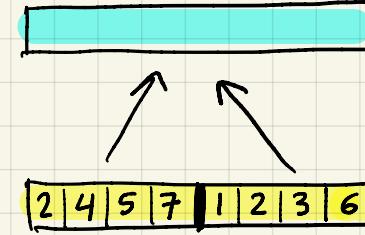
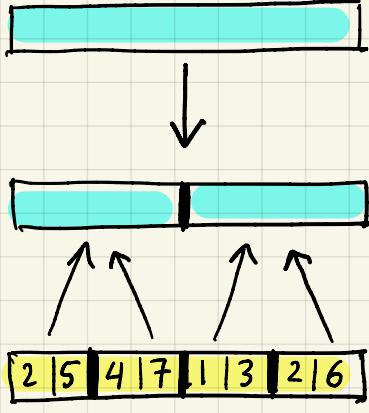
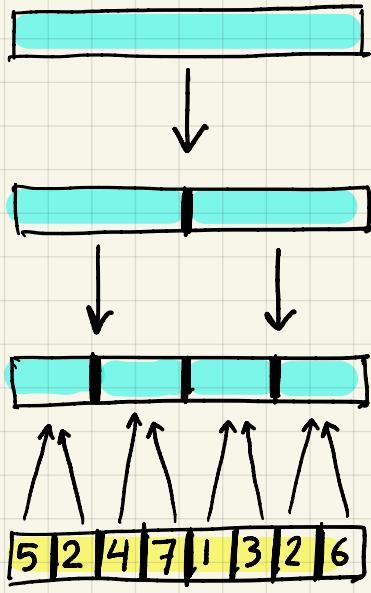
1	2	2	3	4	5	6	7
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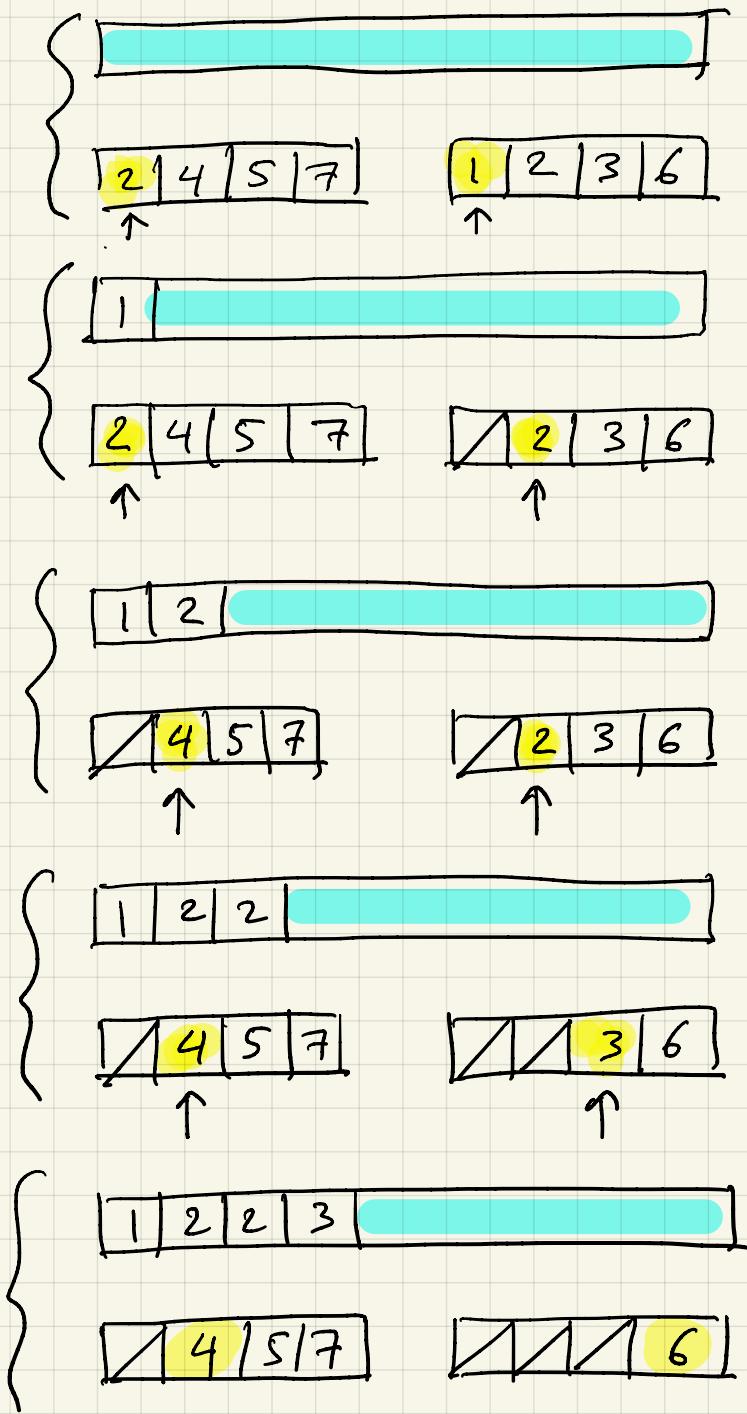
"merge magic"

2	4	5	7		1	2	3	6
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2	5		4	7		1	3		2	6
---	---	--	---	---	--	---	---	--	---	---

5	2		4	7		1	3		2	6
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Merge is NOT in place

Merging takes linear

time, that's the
benefit we are
getting.

etc...

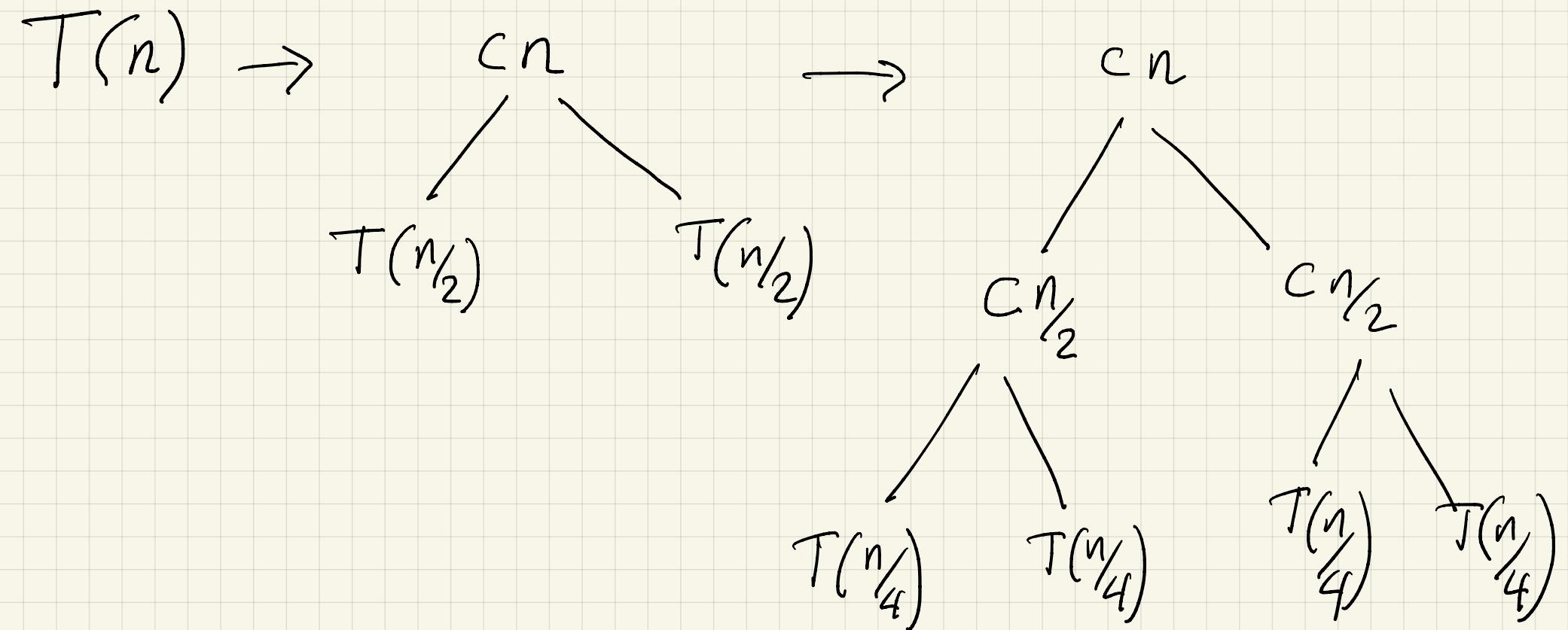
Recurrence: Let $T(n)$ be time needed
to sort an array of size n .

$$T(n) = \begin{cases} 2 T\left(\frac{n}{2}\right) + Cn & n > 1 \\ C & n = 1 \end{cases}$$

Annotations:

- $\overbrace{2}^{\text{\# of subproblem}}$ $\xrightarrow{\text{size of subproblem}}$
- C $\xrightarrow{\text{\# of subproblem in the divide step}}$

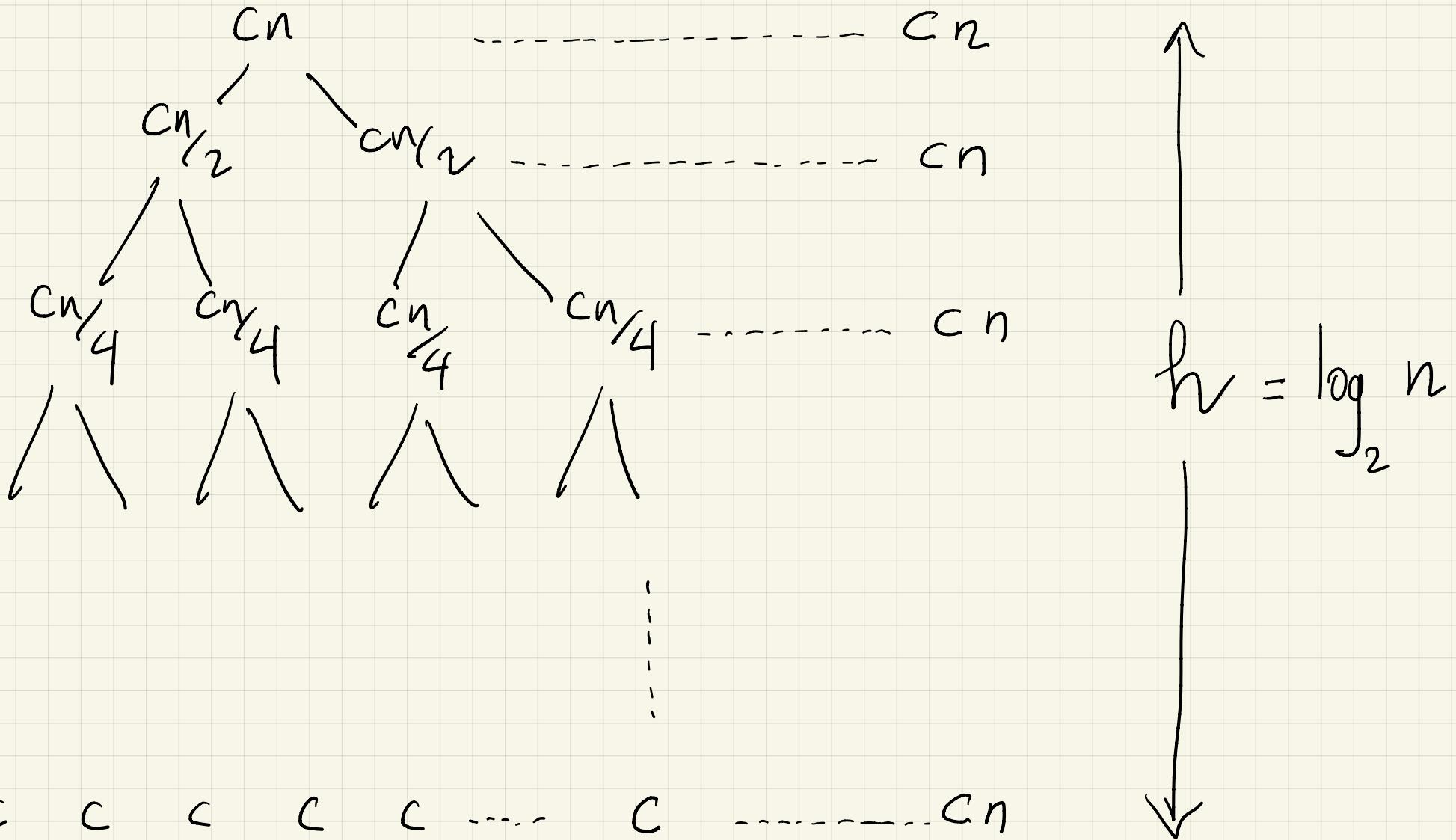
I am using the same constant C
for simplicity.



• • •

Finally $T(1) \rightarrow c$

Assume n
is power of 2
for simplicity



$$\# \text{ total time} = cn \times (\# \text{ levels})$$

$$\begin{aligned}
 & cn \times (\text{height tree} + 1) \\
 &= cn(\log_2 n + 1) = cn\log_2 n + cn
 \end{aligned}$$

$$cn \log_2 n + cn = \Theta(n \log_2 n) = \Theta(n \lg n)$$

But base does not
matter as we
will see later

Strassen's Alg.

$$\begin{bmatrix} r & s \\ t & u \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \times \begin{bmatrix} e & f \\ g & h \end{bmatrix}$$

$$r = ae + bg$$

$$u = cf + dh$$

Straight forward algorithm: For each entry multiply a row by a col.

$n \times n$ matrix \Rightarrow n^2 entries to compute.

each requires a row of size n to be mult. by a col of size n

row / col combination \Rightarrow n multiplication, $n-1$ addition
 $\Rightarrow \Theta(n)$ operations / entry

This alg. require $n^2 \Theta(n) \approx \Theta(n^3)$ time

Divide & Conquer for $n \times n$ matrix

$$\begin{matrix} & \leftarrow \overset{n}{\overbrace{\quad}} \rightarrow \\ \begin{matrix} R \\ T \end{matrix} & \left[\begin{matrix} R & S \\ T & U \end{matrix} \right] = \left[\begin{matrix} A & B \\ C & D \end{matrix} \right] \times \left[\begin{matrix} E & F \\ G & H \end{matrix} \right] \\ \uparrow \downarrow & \end{matrix}$$

$$R = AE + BG$$

$$T(n) = \begin{cases} 8T(n/2) + cn^2 & n > 1 \\ c & n = 1 \end{cases}$$