

LECTURE 15

COMPARISON SORTS

MCS 275 Spring 2023

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LECTURE 15: COMPARISON SORTS

Reminders and announcements:

- Homework 6 posted.
- Project 2 due 6pm Fri Feb 24. Autograder opens by Monday.
- Starting new topic (trees) next week.

EVALUATING SORTS

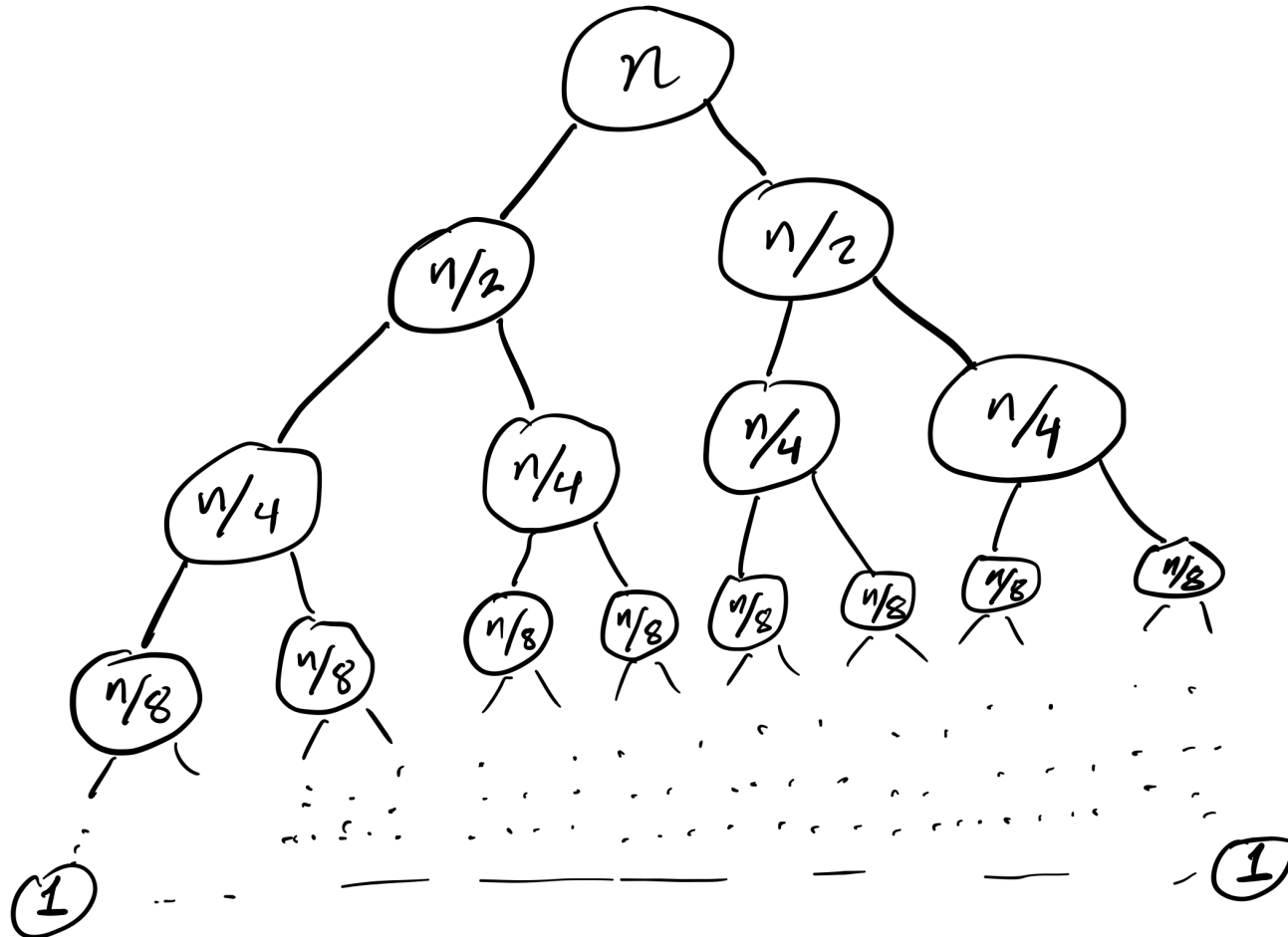
On Monday we discussed and implemented mergesort, developed by von Neumann (1945) and Goldstine (1947).

On Wednesday we discussed quicksort, first described by Hoare (1959), with the simpler partitioning scheme introduced by Lomuto.

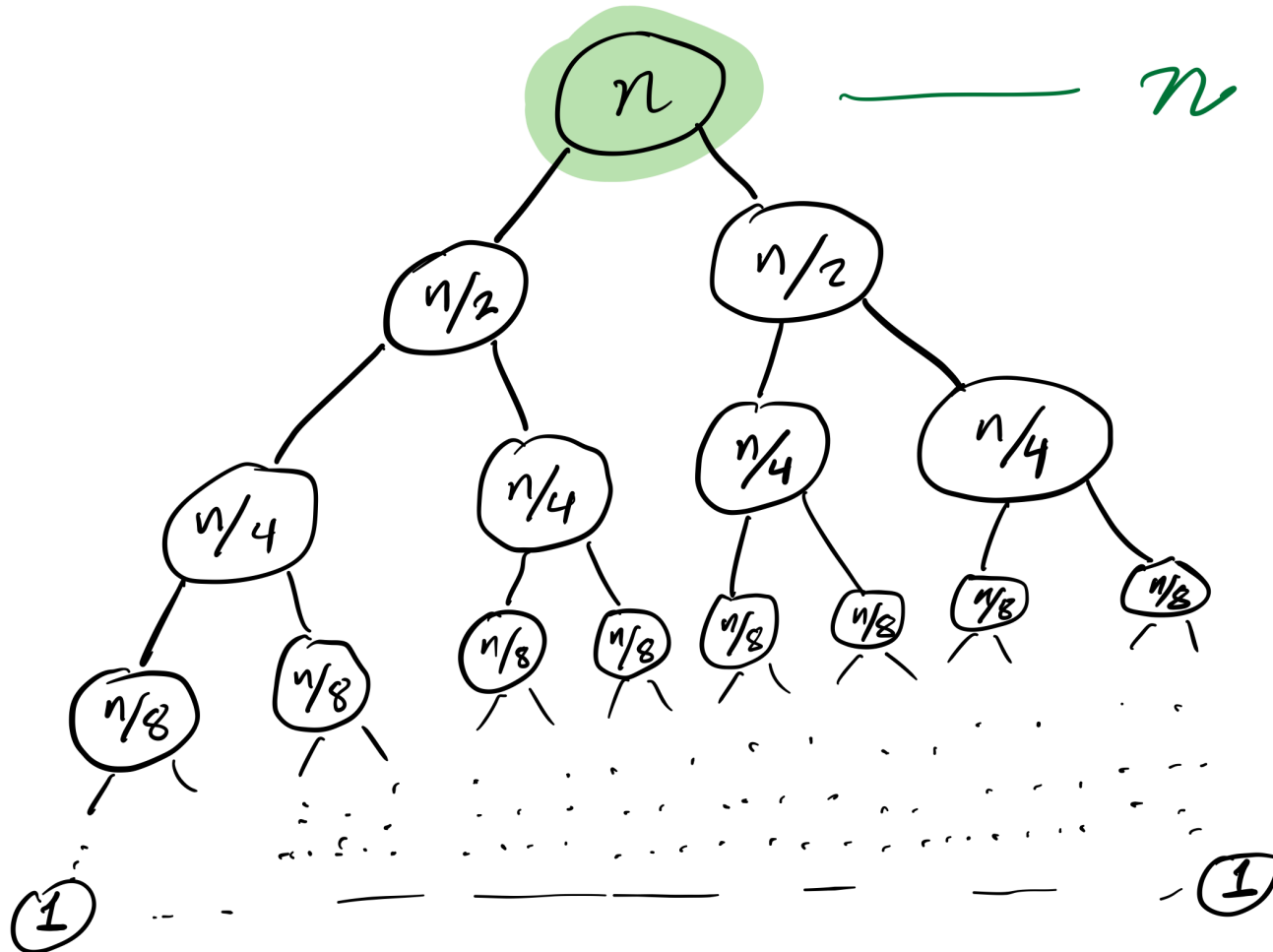
But are these actually good ways to sort a list?

MERGESORT RECURSION TREE

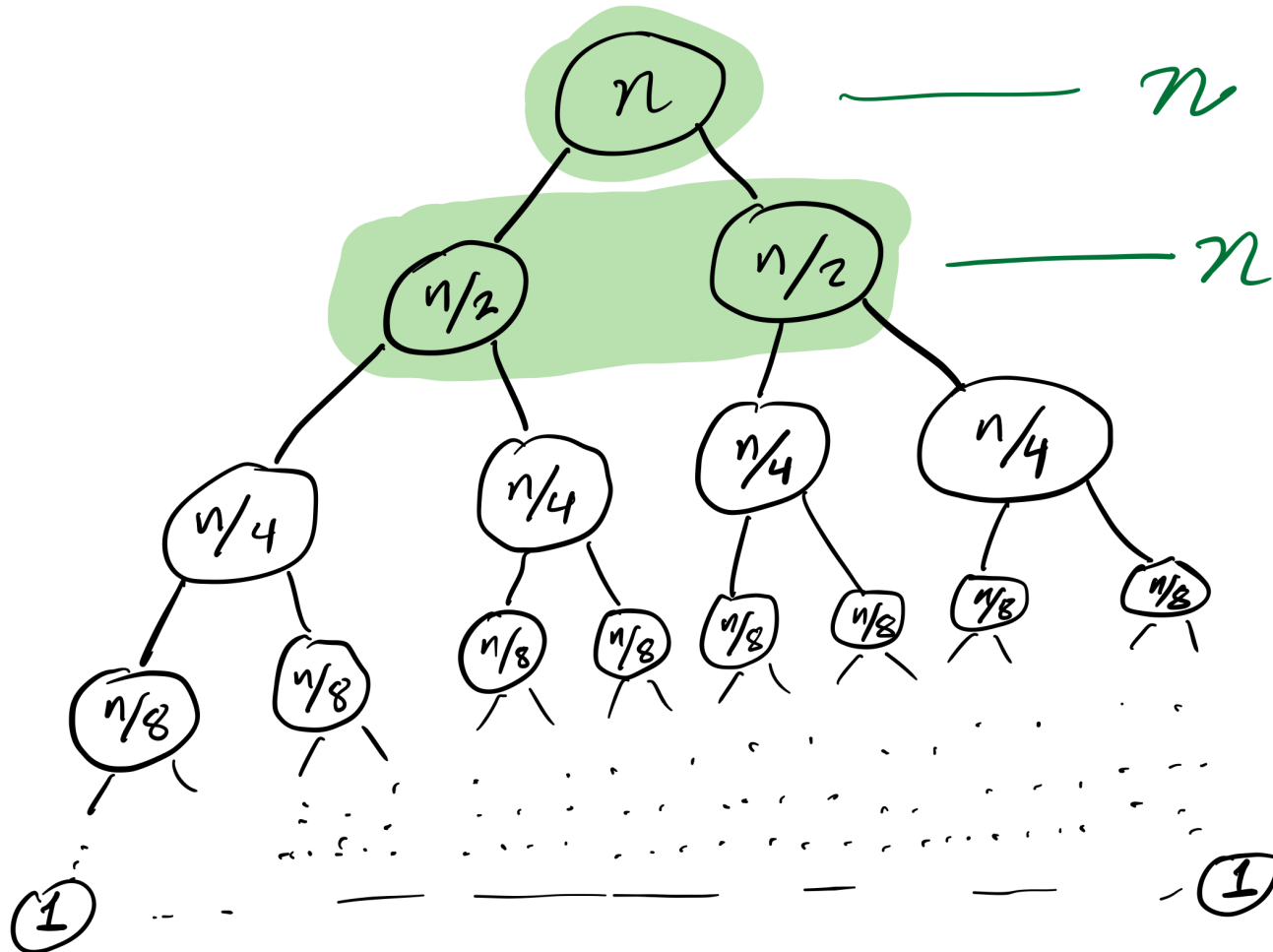
MERGESORT RECURSION TREE



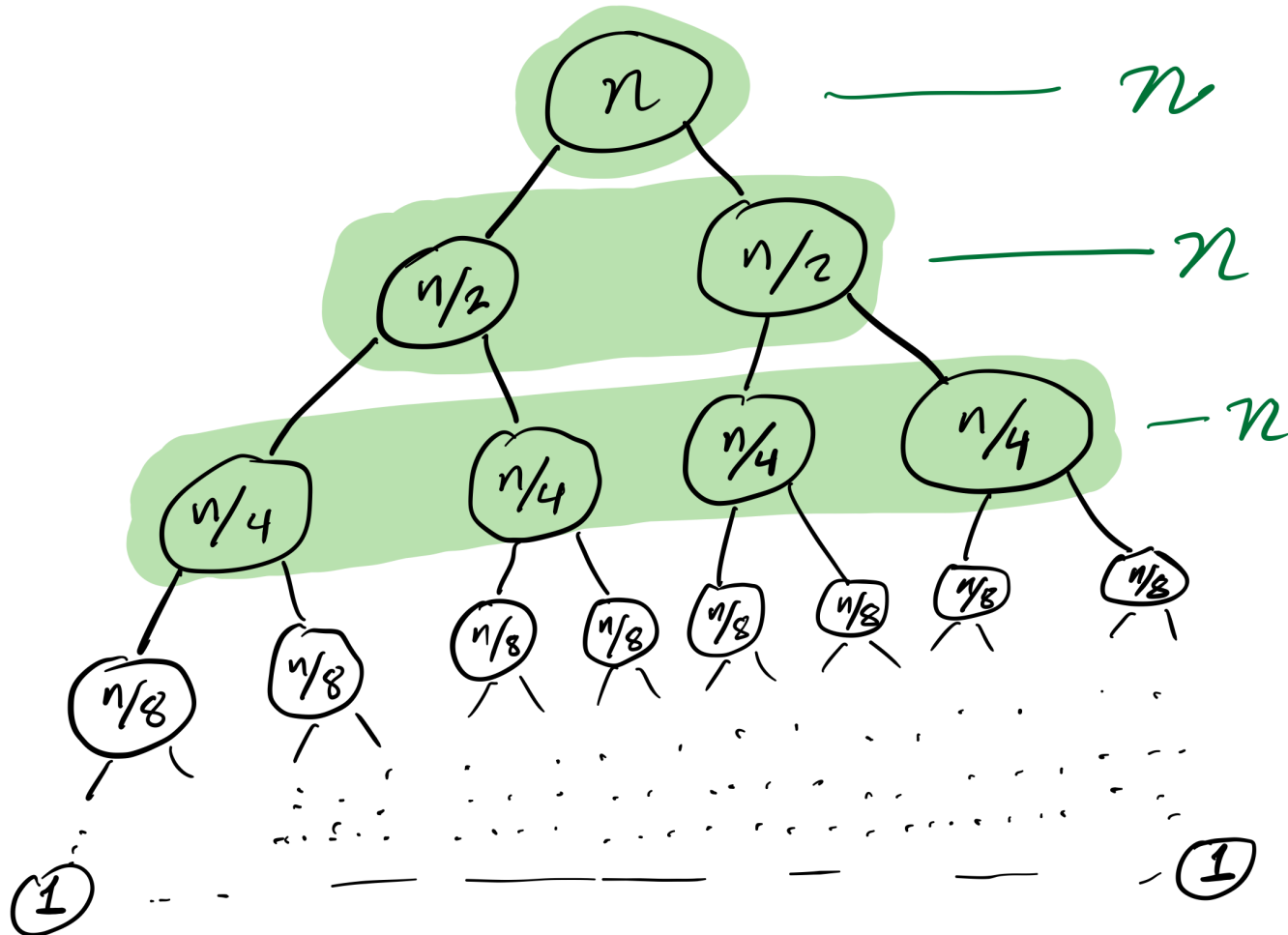
MERGESORT RECURSION TREE



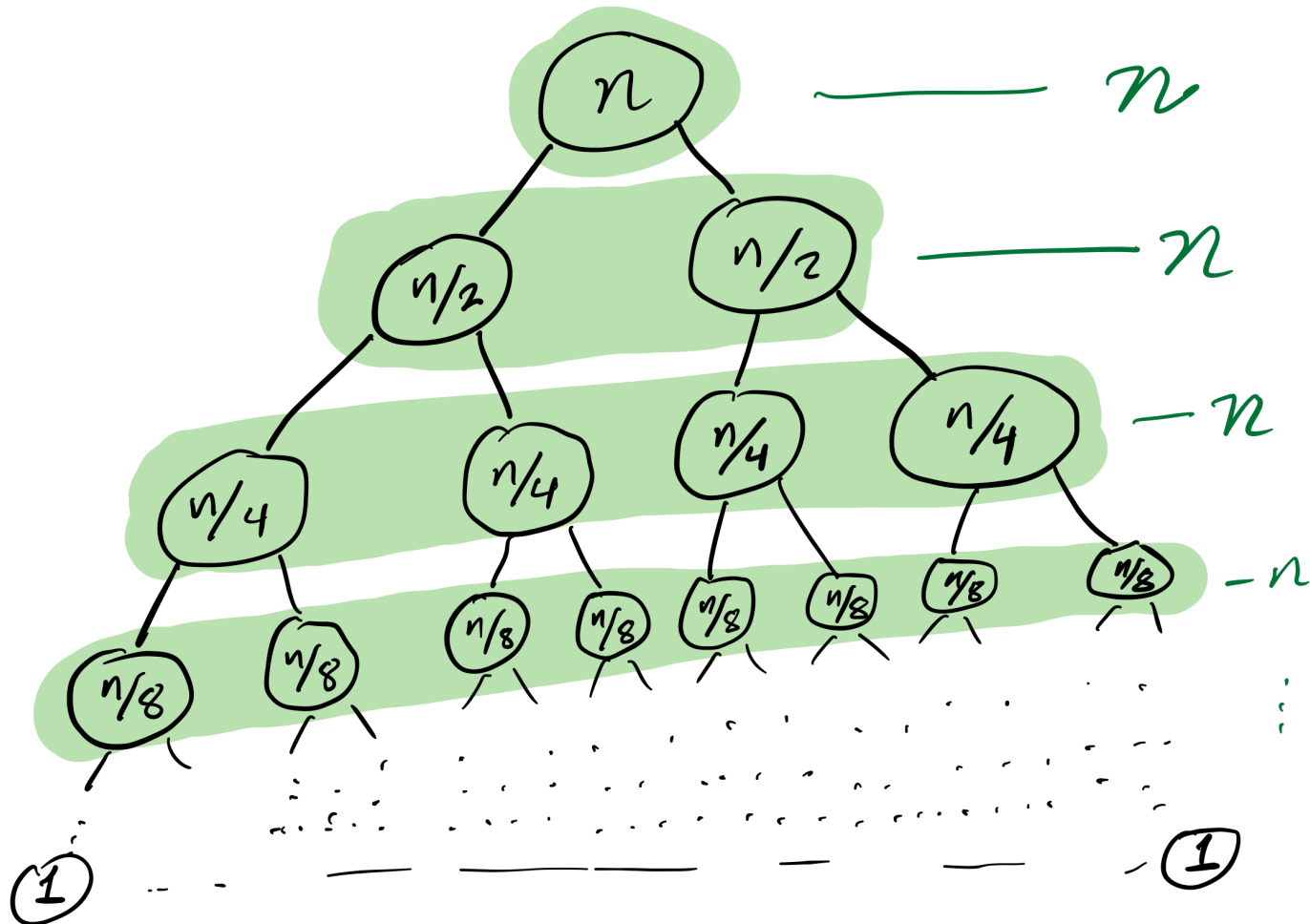
MERGESORT RECURSION TREE



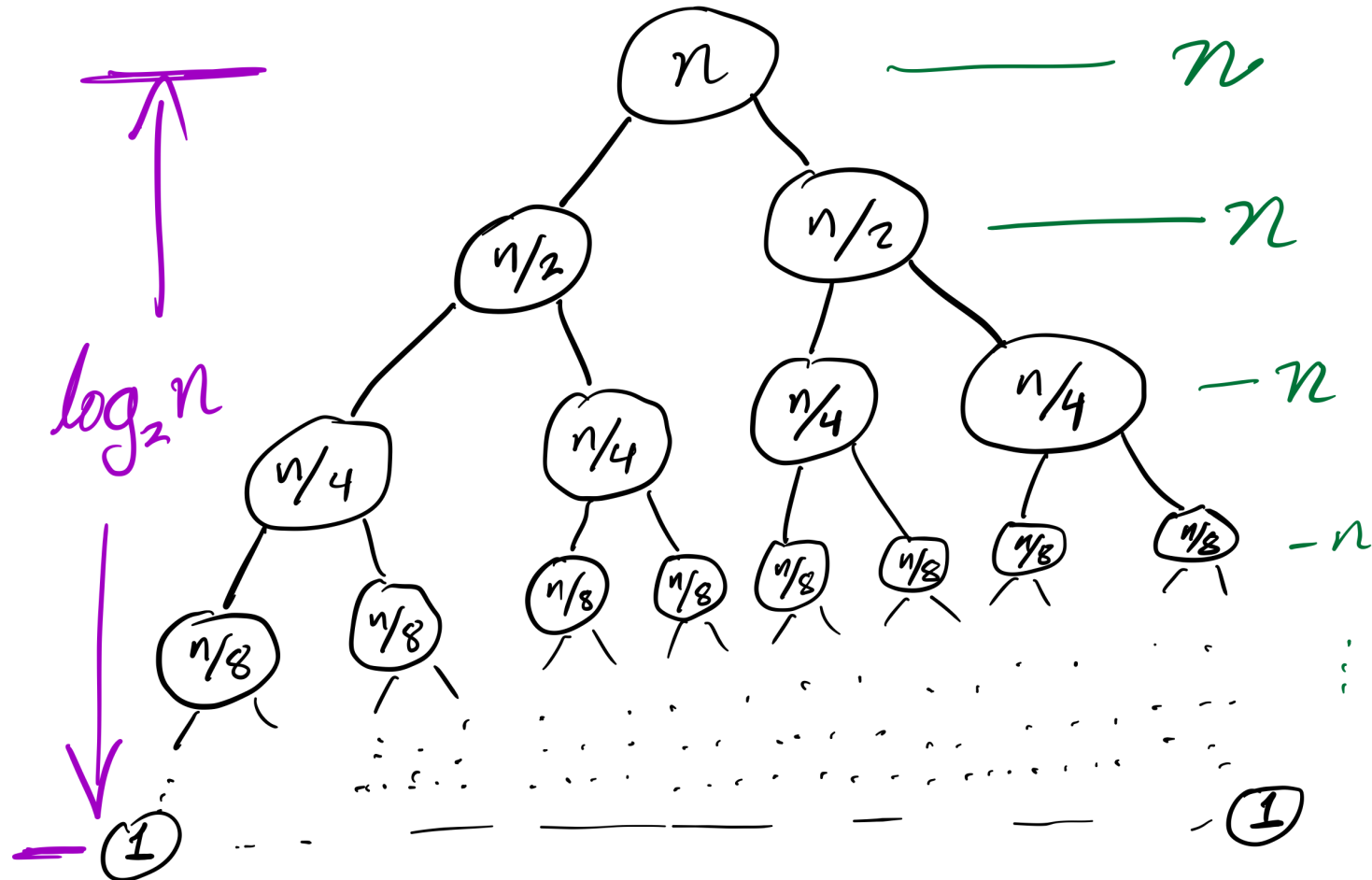
MERGESORT RECURSION TREE



MERGESORT RECURSION TREE



MERGESORT RECURSION TREE



MERGESORT RECURSION TREE

$$\left(\log_2 n \right) \left(n \right)$$

levels per level.

MERGESORT RECURSION TREE

$$n \log_2 n$$

MERGESORT RECURSION TREE

$$C n \log n$$

time cost of mergesort

for a list of length n .

EFFICIENCY

Theorem: If you measure the time cost of mergesort in any of these terms

- Number of comparisons made
- Number of assignments (e.g. $L[i] = x$ counts as 1)
- Number of Python statements executed

then the cost to sort a list of length n is less than $Cn \log(n)$, for some constant C that only depends on which expense measure you chose.

ASYMPTOTICALLY OPTIMAL

$Cn \log(n)$ is pretty efficient for an operation that needs to look at all n elements. It's not linear in n , but it only grows a little faster than linear functions.

Furthermore, $Cn \log(n)$ is the best possible time for comparison sort of n elements (though different methods might have better C).

LOOKING BACK ON QUICKSORT

It ought to be called **partitionsort** because the algorithm is simply:

- Partition the list
- Quicksort the part before the pivot
- Quicksort the part after the pivot

OTHER PARTITION STRATEGIES

We used the last element of the list as a pivot. Other popular choices:

- The first element, $L[start]$
- A random element of $L[start:end]$
- The element $L[(start+end) // 2]$
- An element near the median of $L[start:end]$
(more complicated to find!)

HOW TO CHOOSE?

Knowing something about your starting data may guide choice of partition strategy (or even the choice to use something other than quicksort).

Almost-sorted data is a common special case where first or last pivots are bad.

EFFICIENCY

Theorem: If you measure the time cost of quicksort in any of these terms

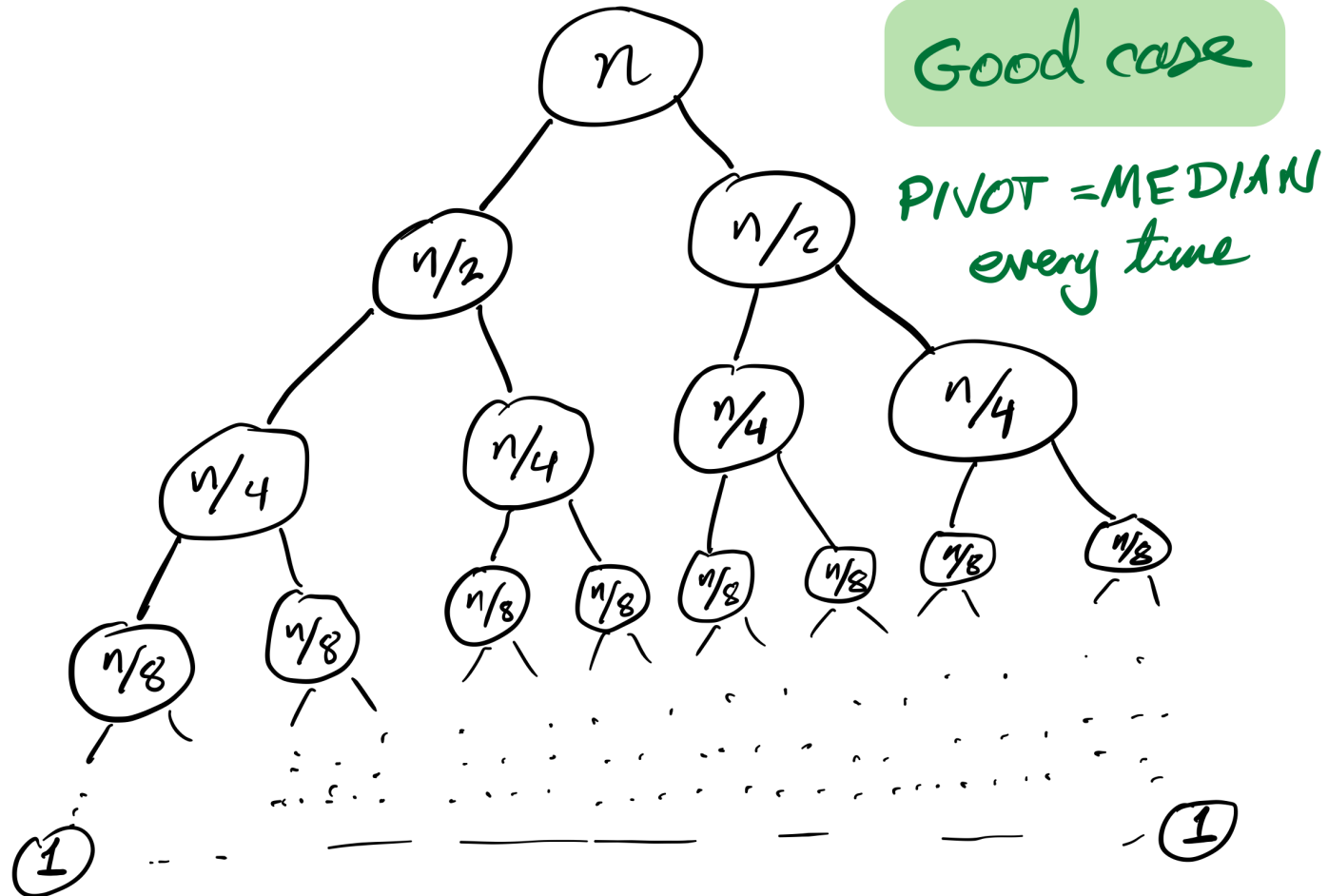
- Number of comparisons made
- Number of swaps or assignments
- Number of Python statements executed

then the cost to sort a list of length n is less than Cn^2 , for some constant C .

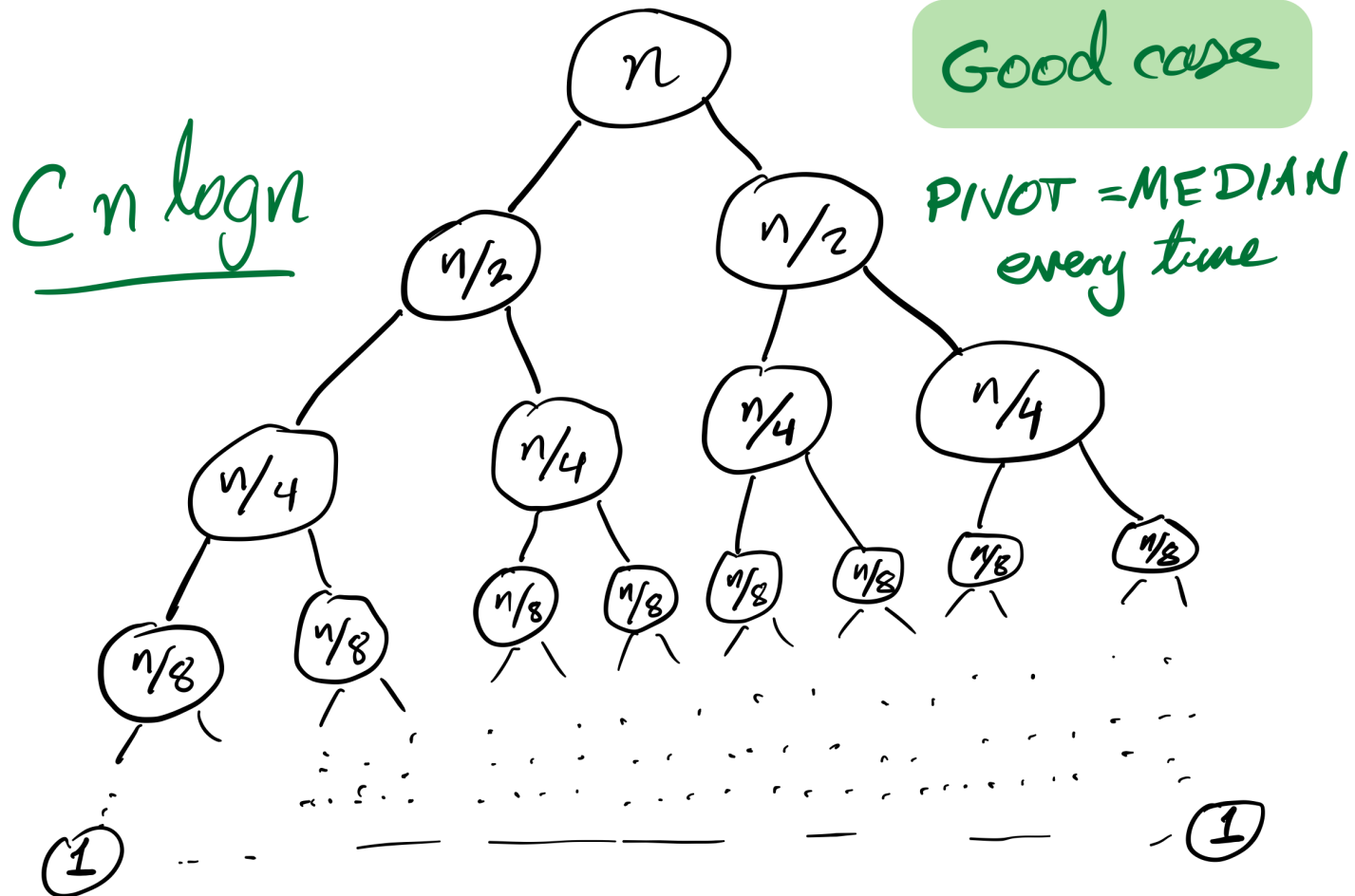
But if you average over all possible orders of the input data, the result is less than $Cn \log(n)$.

QUICKSORT RECURSION TREE

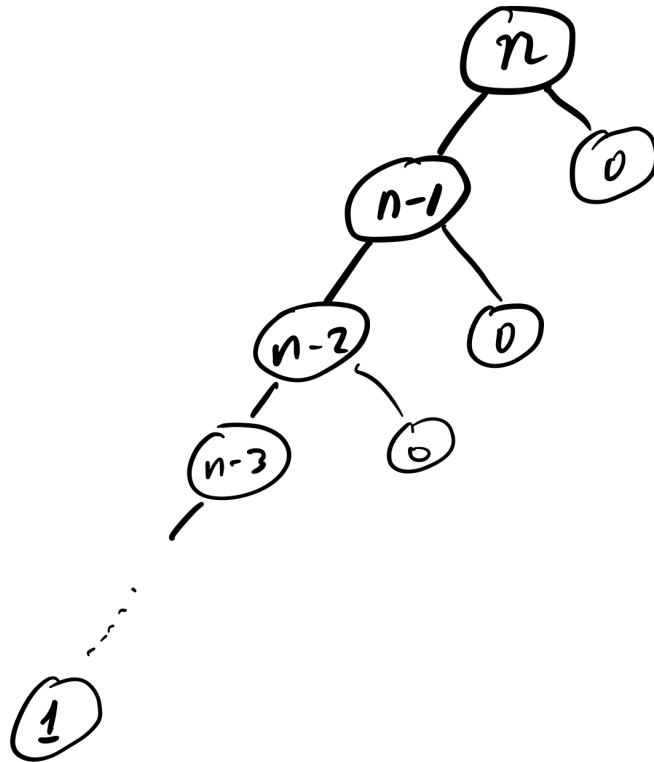
QUICKSORT RECURSION TREE



QUICKSORT RECURSION TREE



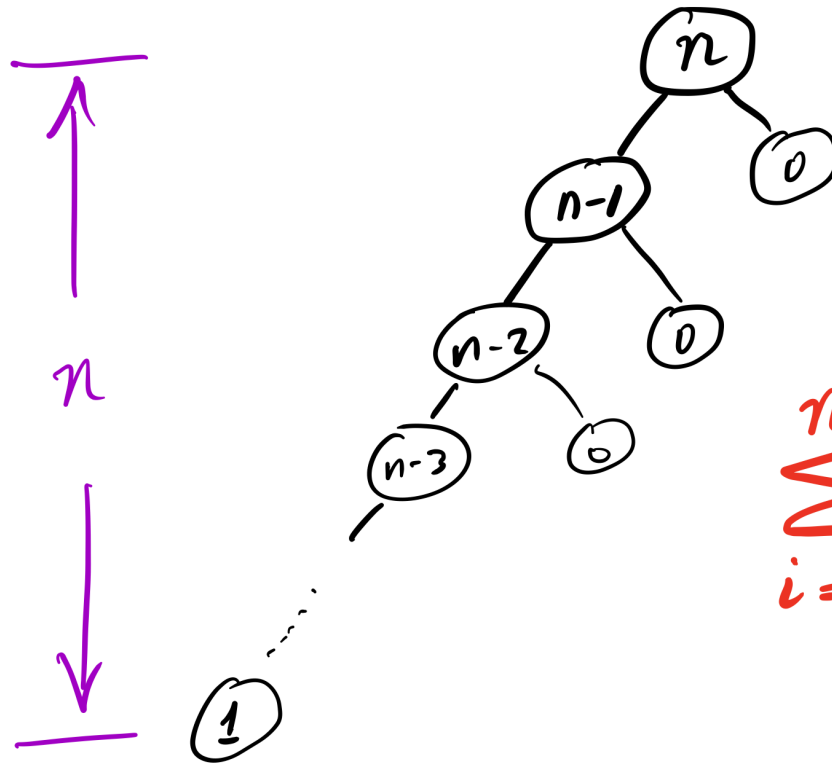
QUICKSORT RECURSION TREE



WORST CASE

Pivot = max or min
every time.

QUICKSORT RECURSION TREE



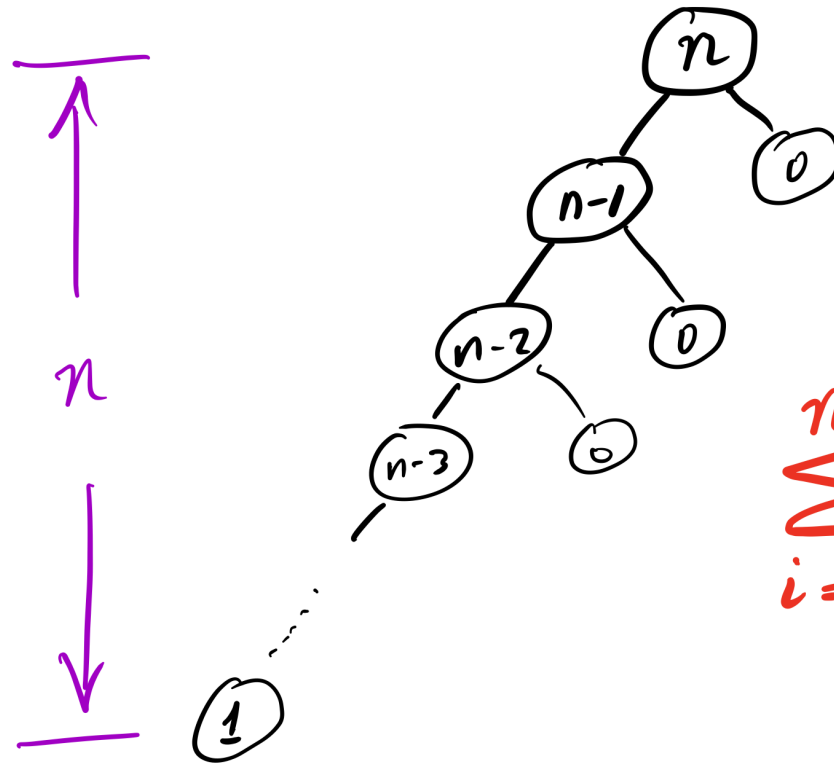
WORST CASE

Pivot = max or min
every time.

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

$$= Cn^2 + \text{smaller terms}$$

QUICKSORT RECURSION TREE



WORST CASE

Pivot = max or min
every time.

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

$$= Cn^2 + \text{smaller terms}$$

SORTED INPUT + LAST ELT PIVOT DOES THIS !!

BAD CASE

What if we ask our version of `quicksort` to sort a list that is already sorted?

Recursion depth is n (whereas if the pivot is always the median it would be $\approx \log_2 n$).

Number of comparisons $\approx Cn^2$. Very slow!

STABILITY

A sort is called **stable** if items that compare as equal stay in the same relative order after sorting.

This could be important if the items are more complex objects we want to sort by one attribute (e.g. sort alphabetized employee records by hiring year).

As we implemented them:

- Mergesort is stable
- Quicksort is not stable

EFFICIENCY SUMMARY

Algorithm	Time (worst)	Time (average)	Stable?	Space
Mergesort	$Cn \log(n)$	$Cn \log(n)$	Yes	Cn
Quicksort	Cn^2	$Cn \log(n)$	No	C

(Every time C is used, it represents a different constant.)

OTHER COMPARISON SORTS

- **Insertion sort** — Convert the beginning of the list to a sorted list, starting with one element and growing by one element at a time.
- **Bubble sort** — Process the list from left to right. Any time two adjacent elements are in the wrong order, switch them. Repeat n times.

EFFICIENCY SUMMARY

Algorithm	Time (worst)	Time (average)	Stable?	Space
Mergesort	$Cn \log(n)$	$Cn \log(n)$	Yes	Cn
Quicksort	Cn^2	$Cn \log(n)$	No	C
Insertion	Cn^2	Cn^2	Yes	C
Bubble	Cn^2	Cn^2	Yes	C

(Every time C is used, it represents a different constant.)

CLOSING THOUGHTS ON SORTING

Mergesort is rarely a bad choice. It is stable and sorts in $Cn \log(n)$ time. Nearly sorted input is not a pathological case. Its main weakness is its use of memory proportional to the input size.

Heapsort, which we may discuss later, has $Cn \log(n)$ running time and uses constant space, but it is not stable.

There *are* stable comparison sorts with $Cn \log(n)$ running time and constant space (best in every category!) though they tend to be more complex.

If swaps and comparisons have very different cost, it may be important to select an algorithm that minimizes one of them. Python's `list.sort` assumes that comparisons are expensive, and uses **Timsort**.

QUADRATIC DANGER

Algorithms that take time proportional to n^2 are a big source of real-world trouble. They are often fast enough in small-scale tests to not be noticed as a problem, yet are slow enough for large inputs to disable the fastest computers.

REFERENCES

- An algorithms textbook like [Algorithms by Jeff Erickson](#) will discuss analysis of running time for sorting algorithms in more depth.

REVISION HISTORY

- 2022-02-21 Last year's lecture on this topic finalized
- 2023-02-17 Updated for 2023

