

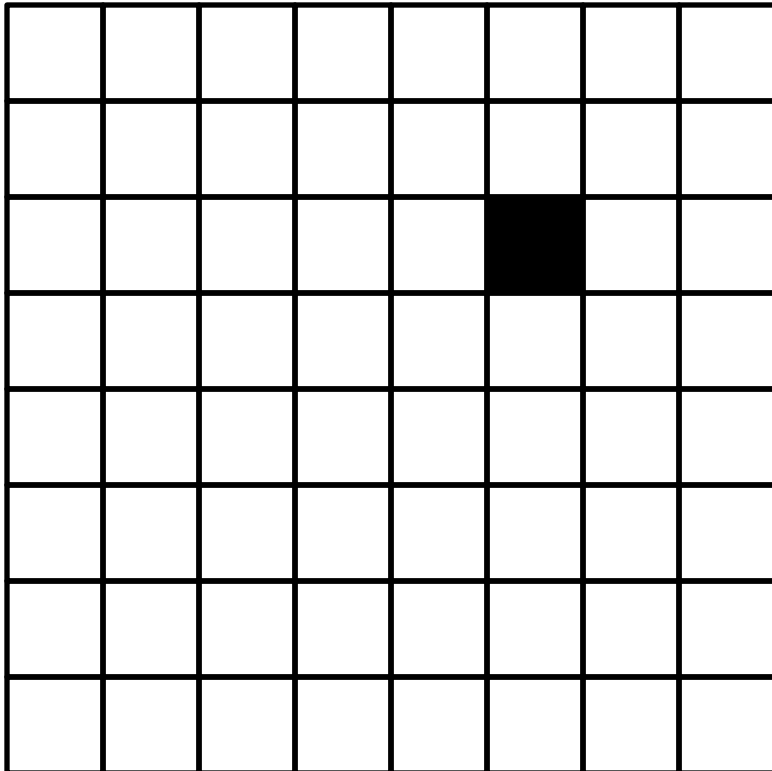
CS4102 Algorithms

Fall 2018

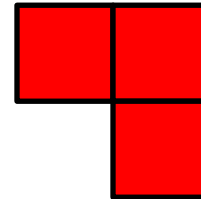
Warm up

Can you cover an 8×8 grid with 1 square missing using “trominoes”?

Can you cover this?



With these?



Today's Keywords

- Recursion
- Recurrences
- Asymptotic notation
- Divide and Conquer
- Trominos
- Merge Sort

CLRS Readings

- Chapters 3 & 4

Homeworks

- Hw0 due 11pm Wednesday, Sept 5
 - Submit 2 attachments (zip and pdf)
- Hw1 released Monday, Sept 3
 - Due 11pm Wednesday, Sept 12
 - Written (use Latex!)
 - Asymptotic notation
 - Recurrences
 - Divide and conquer

Attendance

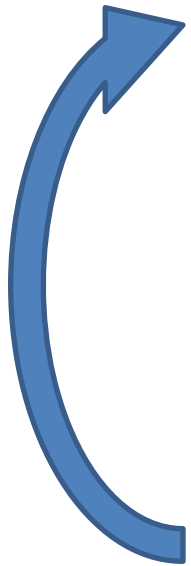
- How many people are here today?
- Naïve algorithm
 1. Everyone stand
 2. Professor walks around counting people
 3. When counted, sit down
- Run time?
 - Class of n students
 - $O(n)$
- Other suggestions?

Better Attendance

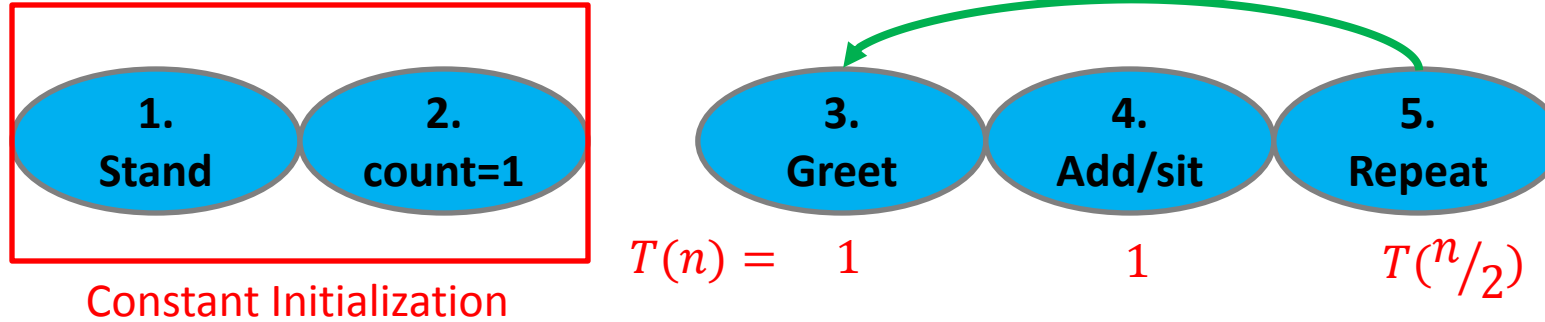
1. Everyone Stand
2. Initialize your “count” to 1
3. Greet a neighbor who is standing: share your name, full date of birth(pause if odd one out)
4. If you are older: give “count” to younger and sit.
Else if you are younger: add your “count” with older’s
5. If you are standing and have a standing neighbor, go to 3

What was the
run time of this
algorithm?

What are we
going to count?



Attendance Algorithm Analysis



$$T(n) = 1 + 1 + T(n/2) \quad \text{How can we "solve" this?}$$

$$T(1) = 3 \quad \text{Base case?}$$

Do not need to be exact, asymptotic bound is fine.

Why?

Let's solve the recurrence!

Special case: $n = 2^k$

$T(1) = 3$

$T(n) = 2 + \cancel{T(n/2)}$

$2 + \cancel{T(n/4)}$

$2 + \cancel{T(n/8)}$

\dots

3

k

$$T(n) = 3 + \sum_{i=0}^{\log_2 n} 2 = 2 \log_2 n + 3$$

What if $n \neq 2^k$?

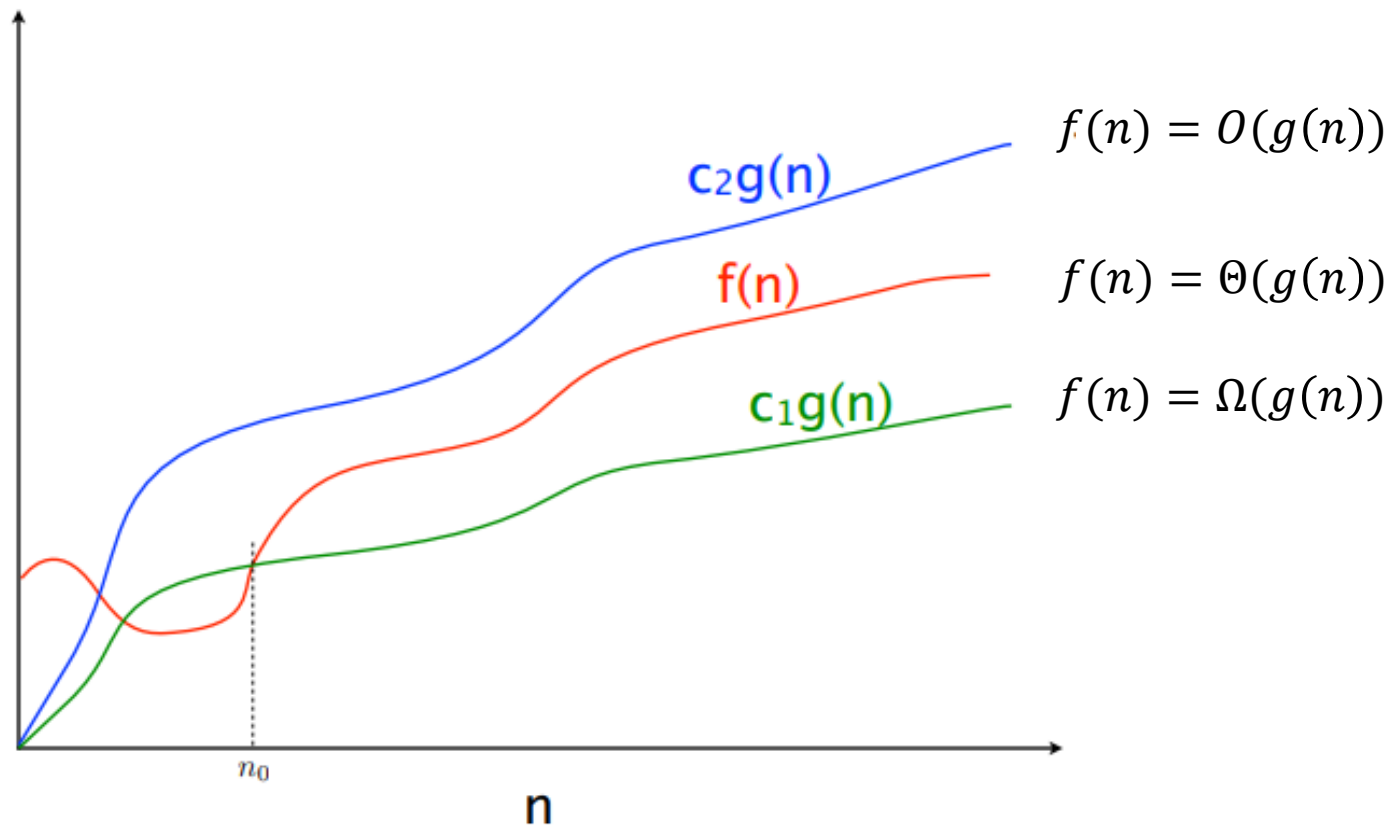
- More people in the room \rightarrow more time
- $\forall 0 < n < m, T(n) < T(m)$
- $T(n) \leq T(2^{\lceil \log_2 n \rceil}) = 2 \lceil \log_2 n \rceil + 3 = O(\log n)$



These are unimportant.
Why?

Asymptotic Notation*

- $O(g(n))$
 - At most within constant of g for large n
 - $\{\text{functions } f \mid \exists \text{ constants } c > 0, n_0 > 0 \text{ s.t. } \forall n > n_0, f(n) \leq c \cdot g(n)\}$
- $\Omega(g(n))$
 - At least within constant of g for large n
 - $\{\text{functions } f \mid \exists \text{ constants } c > 0, n_0 > 0 \text{ s.t. } \forall n > n_0, f(n) \geq c \cdot g(n)\}$
- $\Theta(g(n))$
 - “Tightly” within constant of g for large n
 - $\Omega(g(n)) \cap O(g(n))$



Asymptotic Notation Example

- To Show: $n \log n \in O(n^2)$
 - Find $c, n_0 > 0$ s.t. $\forall n > n_0, n \log n \leq c \cdot n^2$
 - Let $c = 1, n_0 = 1$
 - $(1) \log(1) = 0, 1 \cdot 1^2 = 1$
 - $\forall n \geq 1, \log(n) < n \Rightarrow n \log n \leq n^2$

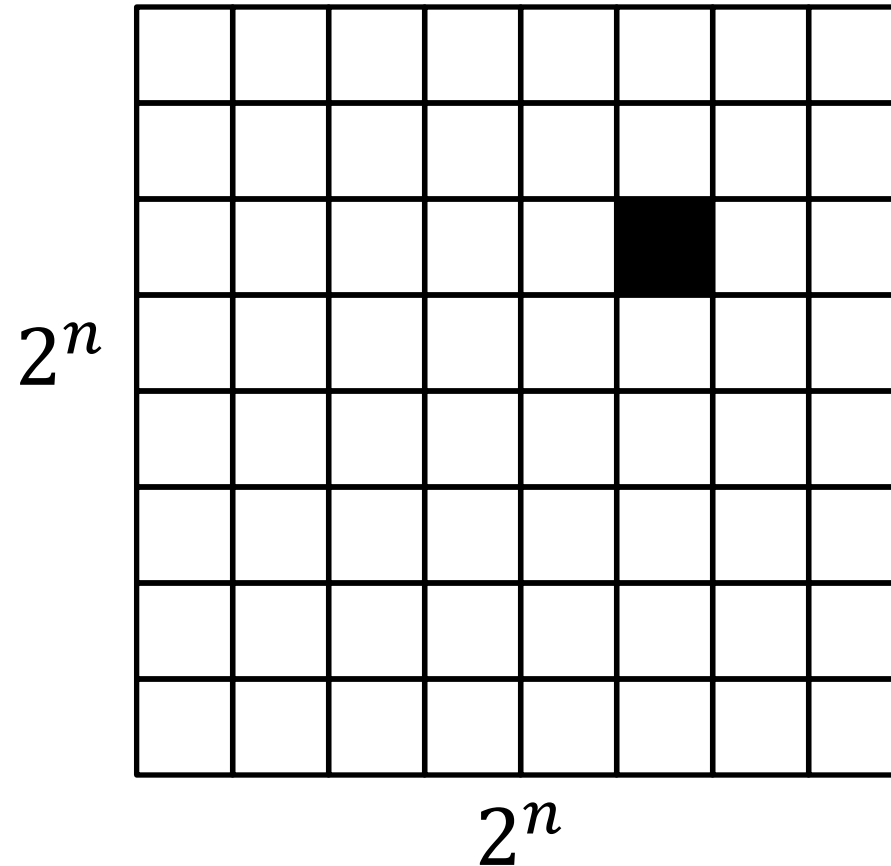
Asymptotic Notation

- $o(g(n))$
 - Below *any* constant of g for large n
 - $\{\text{functions } f \mid \forall \text{ constants } c > 0, \exists n_0 \text{ s.t. } \forall n > n_0, f(n) < c \cdot g(n)\}$
- $\omega(g(n))$
 - Above *any* constant of g for large n
 - $\{\text{functions } f \mid \forall \text{ constants } c > 0, \exists n_0 \text{ s.t. } \forall n > n_0, f(n) > c \cdot g(n)\}$
- $\theta(g(n))$?
 - $o(g(n)) \cap \omega(g(n)) = \emptyset$

Asymptotic Notation Example

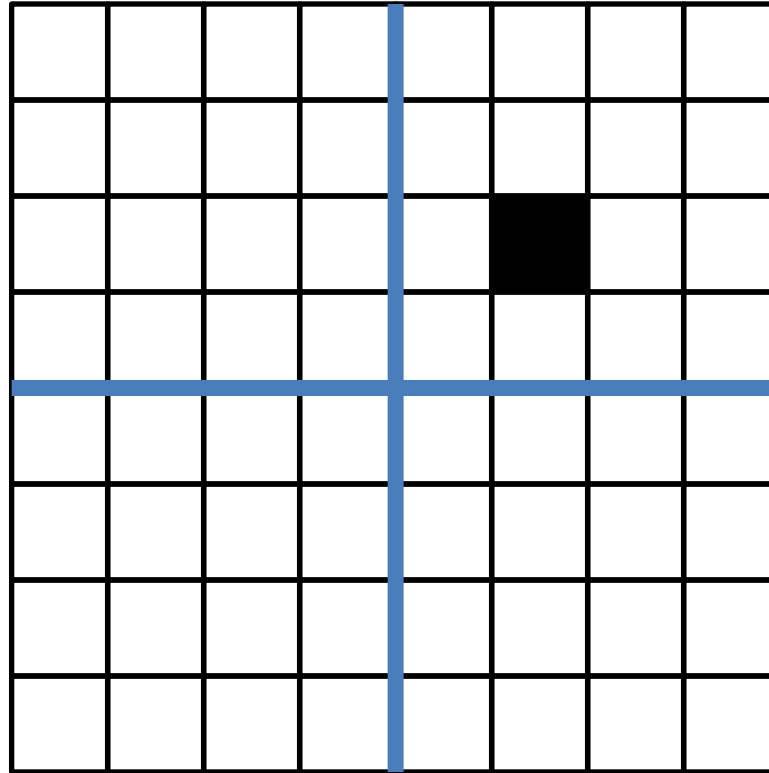
- $o(g(n)) = \{\text{functions } f \mid \forall \text{ constants } c, \exists n_0 \text{ s.t. } \forall n > n_0, f(n) < c \cdot g(n)\}$
- To Show: $n \log n \in o(n^2)$
 - given any c find a $n_0 > 0$ s.t. $\forall n > n_0, n \log n < c \cdot n^2$
 - Find a value of n in terms of c : $n \log n < c \cdot n^2$
 - $n \log n < c \cdot n^2$
 - $\log n < c \cdot n$
 - For a given c , select any value of n such that $\frac{\log n}{n} < c$

Trominoes Puzzle Solution



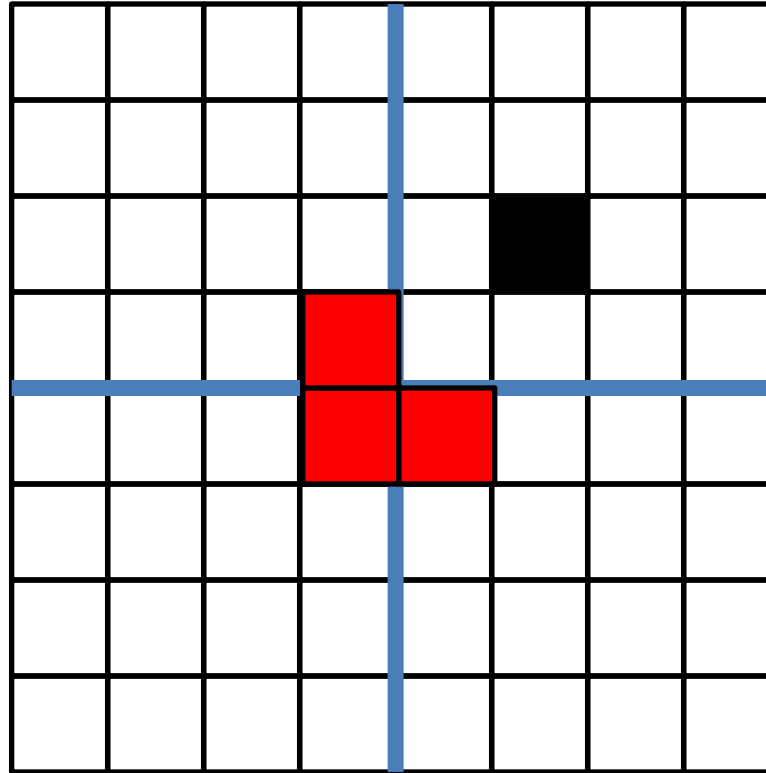
What about larger boards?

Trominoes Puzzle Solution



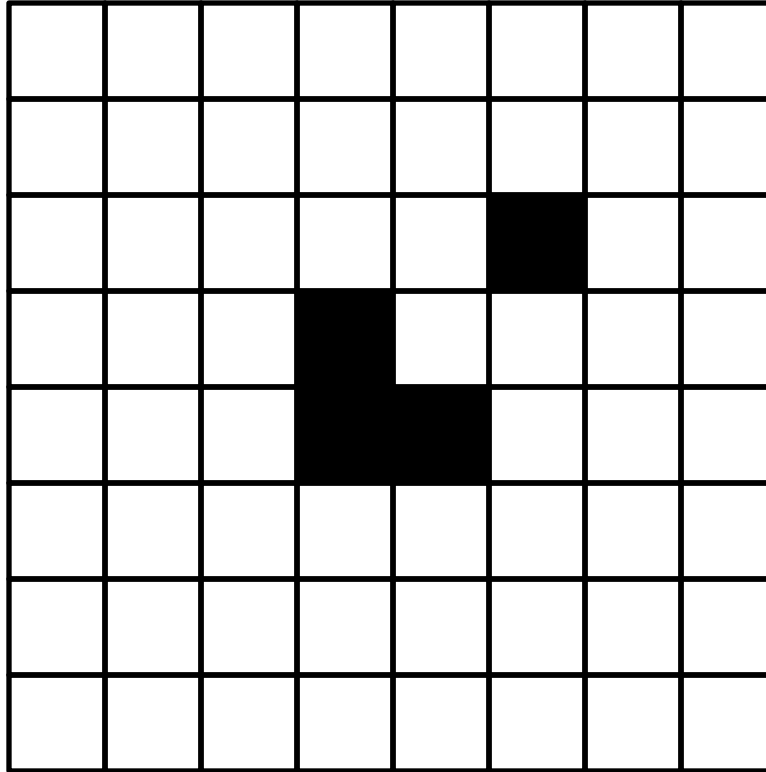
Divide the board into quadrants

Trominoes Puzzle Solution



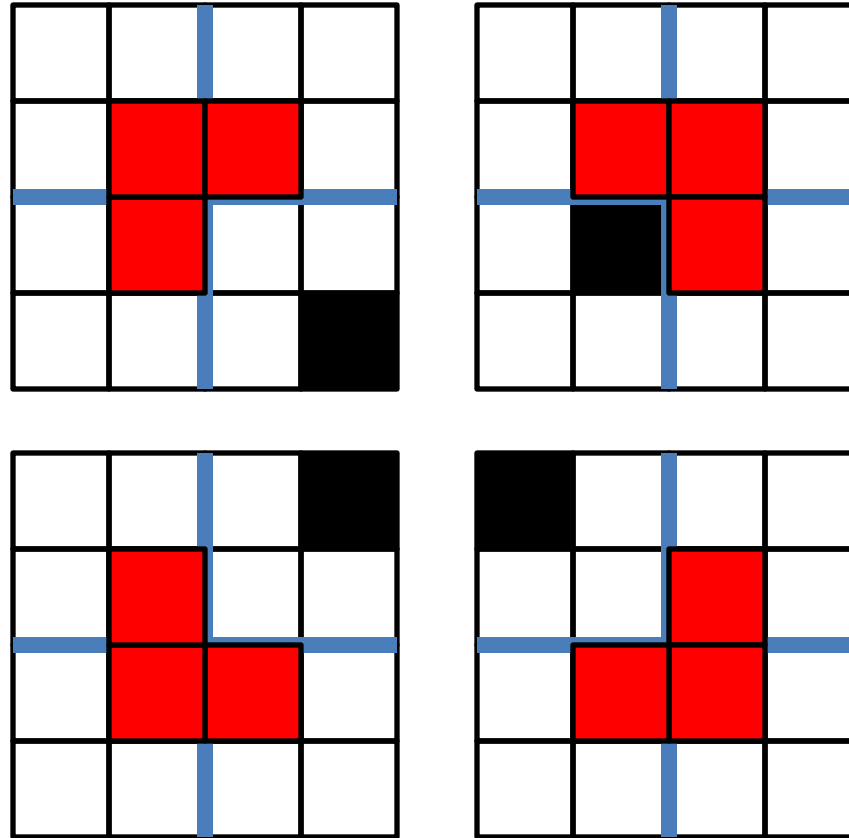
Place a tromino to occupy the three quadrants without the missing piece

Trominoes Puzzle Solution



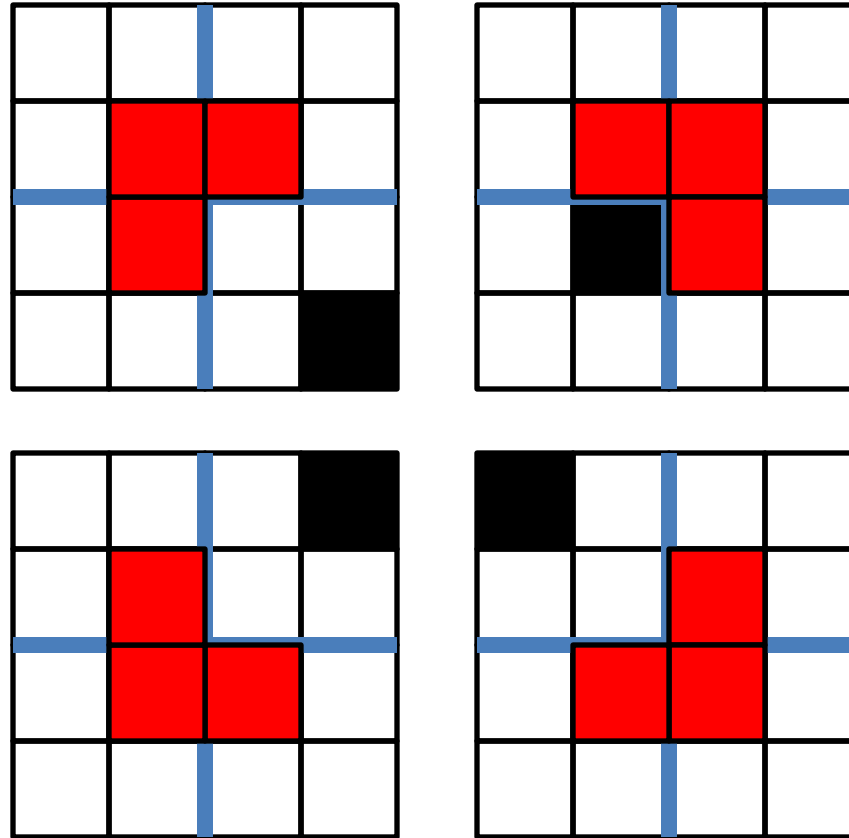
Each quadrant is now a smaller subproblem

Trominoes Puzzle Solution



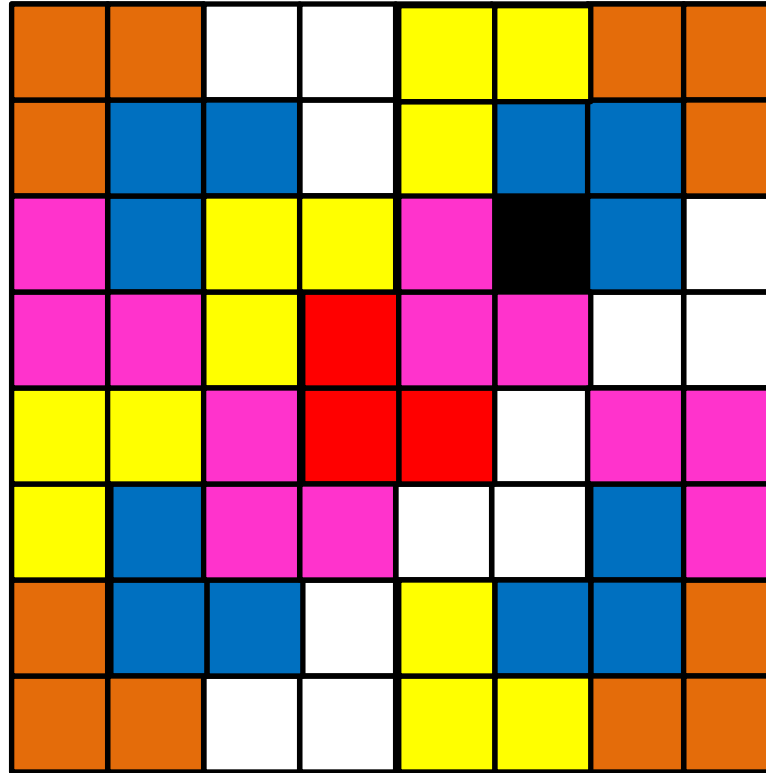
Solve **Recursively**

Divide and Conquer



Our first algorithmic technique!

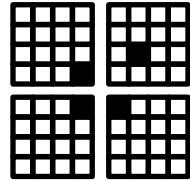
Trominoes Puzzle Solution



Divide and Conquer*

When is this a
good strategy?

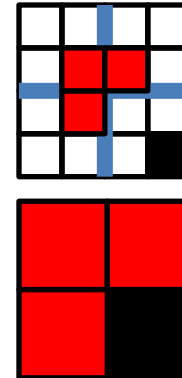
- **Divide:**



- Break the problem into multiple **subproblems**, each smaller instances of the original

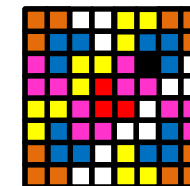
- **Conquer:**

- If the subproblems are “large”:
 - Solve each subproblem **recursively**
- If the subproblems are “small”:
 - Solve them directly (**base case**)



- **Combine:**

- Merge together solutions to subproblems



Analyzing Divide and Conquer

1. Break into smaller **subproblems**
 2. Use **recurrence** relation to express recursive running time
 3. Use **asymptotic** notation to simplify
- **Divide:** $D(n)$ time,
 - **Conquer:** recurse on small problems, size s
 - **Combine:** $C(n)$ time
 - **Recurrence:**
 - $T(n) = D(n) + \sum T(s) + C(n)$

Recurrence Solving Techniques



Tree



Guess/Check



“Cookbook”



Substitution

Merge Sort

- **Divide:**
 - Break n -element list into two lists of $n/2$ elements
- **Conquer:**
 - If $n > 1$:
 - Sort each sublist **recursively**
 - If $n = 1$:
 - List is already sorted (**base case**)
- **Combine:**
 - Merge together sorted sublists into one sorted list

Merge

- **Combine:** Merge sorted sublists into one sorted list
- We have:
 - 2 sorted lists (L_1, L_2)
 - 1 output list (L_{out})

While (L_1 and L_2 not empty):

 If $L_1[0] \leq L_2[0]$:

$L_{out}.append(L_1.pop())$

 Else:

$L_{out}.append(L_2.pop())$

$L_{out}.append(L_1)$

$L_{out}.append(L_2)$

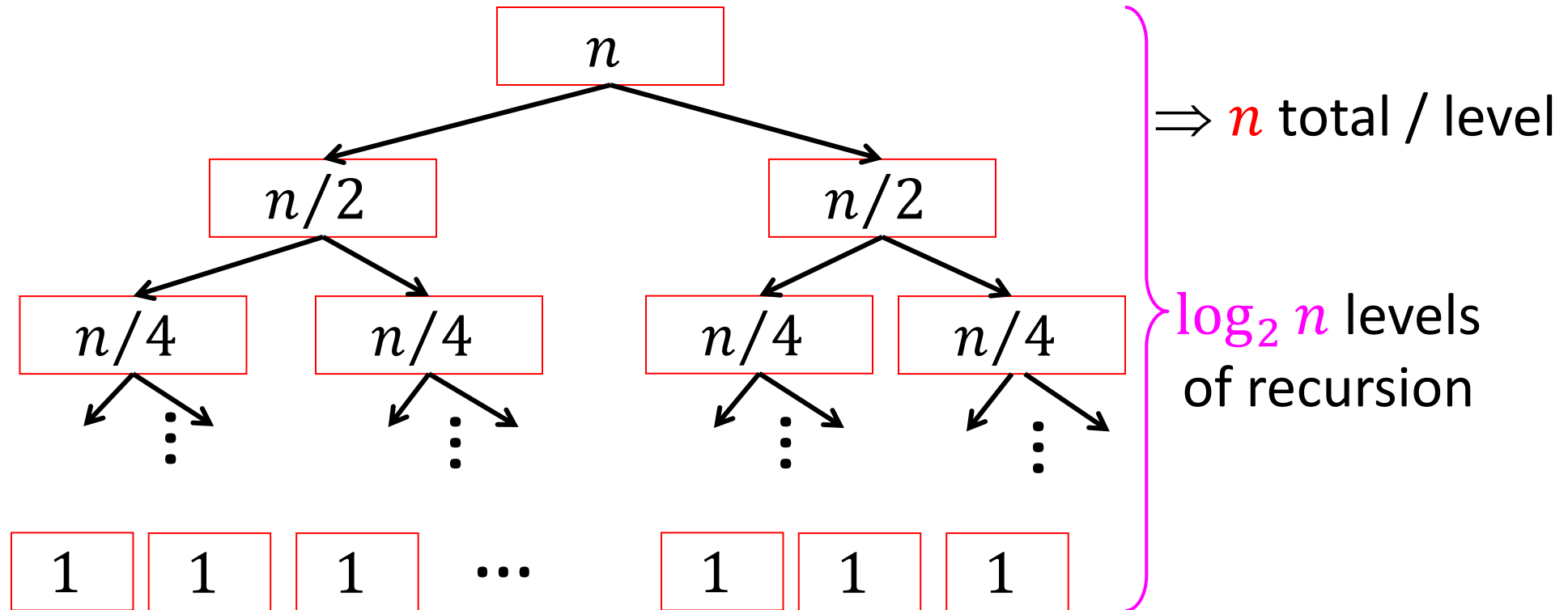
Analyzing Merge Sort

1. Break into smaller **subproblems**
2. Use **recurrence** relation to express recursive running time
3. Use **asymptotic** notation to simplify

- **Divide**: 0 comparisons
- **Conquer**: recurse on 2 small problems, size $\frac{n}{2}$
- **Combine**: n comparisons
- **Recurrence**:
 - $T(n) = 2T(\frac{n}{2}) + n$

Tree method

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



$$T(n) = \sum_{i=1}^{\log n} n = n \log_2 n$$