

Algorithm Design Strategy

"Divide and Conquer"

More examples of Divide and Conquer

☐ Review

- ☐ Divide & Conquer Concept
- ☐ Binary search
- ☐ Merge sort

☐ More examples

- ☐ Quicksort
- ☐ Tromino tiling
- ☐ Finding closest pair of points
- ☐ Matrix Multiplication Algorithm

Divide and Conquer Approach

- Concept: D&C is a general strategy for algorithm design

It involves three steps:

- (1) Divide an instance of a problem into one or more smaller instances
- (2) Conquer (solve) each of the smaller instances. Unless a smaller instance is sufficiently small, use recursion to do this
- (3) If necessary, combine the solutions to the smaller instances to obtain the solution to the original instances (e.g., Merge sort)

- Approach:
 - Recursion (Top-down approach)

Quicksort

```
quicksort(L)
{
  if (length(L) < 2) return L
  else
  {
    pick some x in L      // x is the pivot element
    L1 = { y in L : y < x }
    L2 = { y in L : y > x }
    L3 = { y in L : y = x }
    quicksort(L1)
    quicksort(L2)
    return concatenation of L1, L3, and L2
  }
}
```

Time Complexity of Quicksort

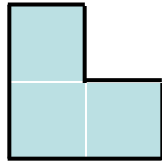
- $T(n) = T(n_1) + T(n_2) + n-1$
 - $n_1 = \text{length of } L_1, n_2 = \text{length of } L_2$
 - $n_1 + n_2 = n - 1$ if all the elements in L are unique
- Best case
 - L is divided into two halves at every recursion
 - $T(n) = 2T(n/2) + n-1 = \Theta(n \lg n)$
- Average case
 - $\Theta(n \lg n)$
 - See the textbook for a proof
- Worst case
 - $T(n) = T(0) + T(n-1) + n-1$ when L is already sorted (As a result, the pivot is always the smallest)
 - So, $T(n) = \Theta(n^2)$
 - Why is it called quicksort then?

How to make Quicksort efficient

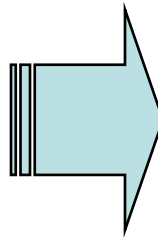
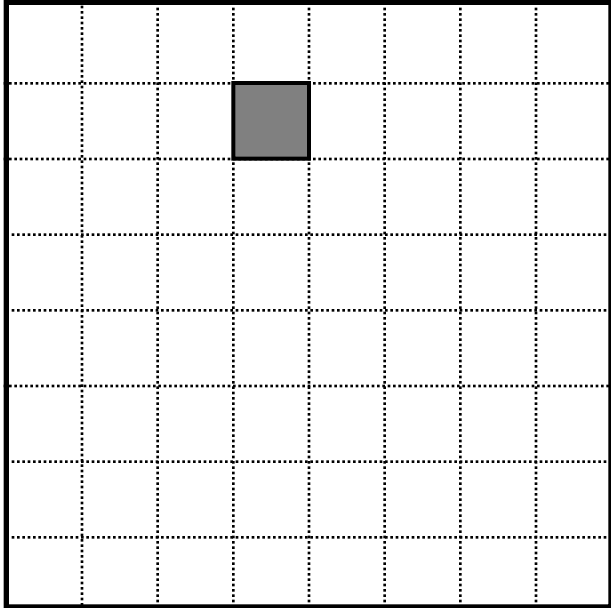
- Randomization
 - Randomly pick an element in L as the pivot
 - So, each item has the same probability of $1/n$ to be selected as the pivot
 - Avoid the worst case (what is the worst case in quicksort?)
 - $O(n \lg n)$ in the worst case
 - c.f., See blackboard for a proof (optional - not required)

Tromino Tiling

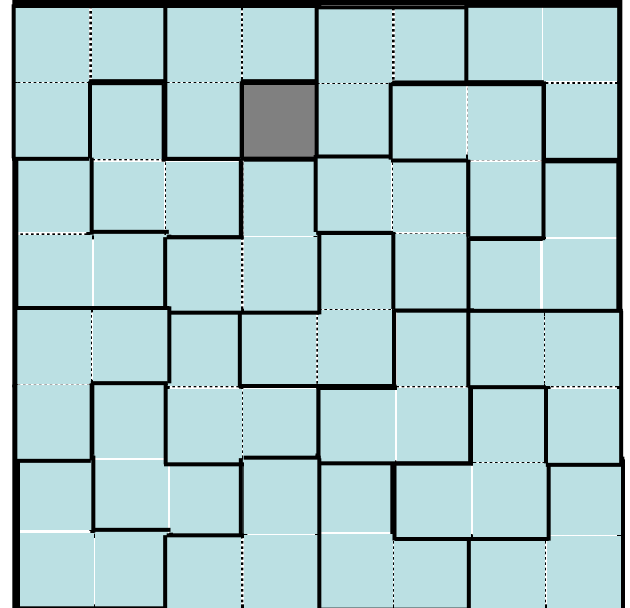
A tromino tile:



And a $n \times n$ ($2^k \times 2^k$) board with a hole:



A tiling of the board with trominos:

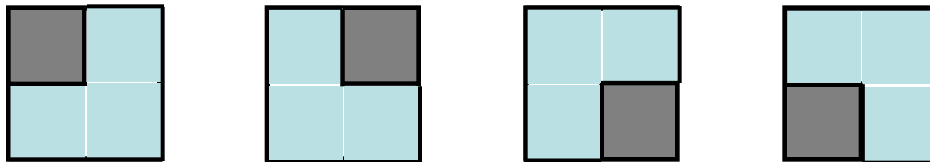


Solvable?

- Total number of squares:
 - $n * n - 1 = 2^k * 2^k - 1 = 2^{2k} - 1 = 4^k - 1$
- Size of one tromino: 3
- So, $4^k - 1$ should be divisible by 3
- Proof by induction
 - Basis: If $n = 2$, $4^1 - 1$ is divisible by 3
 - Induction hypothesis: Assume $4^{k-1} - 1$ is divisible by 3
 - Induction step: $4^k - 1 = 4(4^{k-1} - 1) + 3$

Tiling: Trivial Case ($n = 2$)

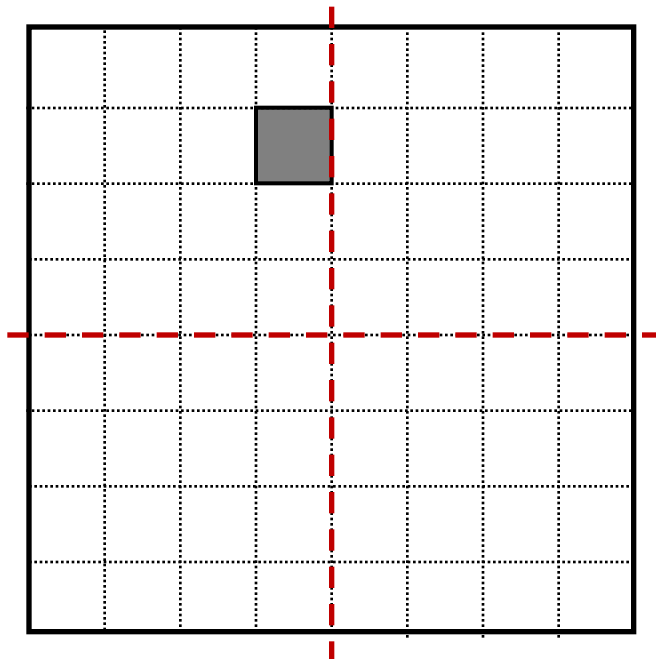
- Trivial case ($n = 2$): tiling a 2×2 board with a hole:



- Idea – try somehow to reduce the size of the original problem, so that we eventually get to the 2×2 boards which we know how to solve...

Tiling: Dividing the Problem

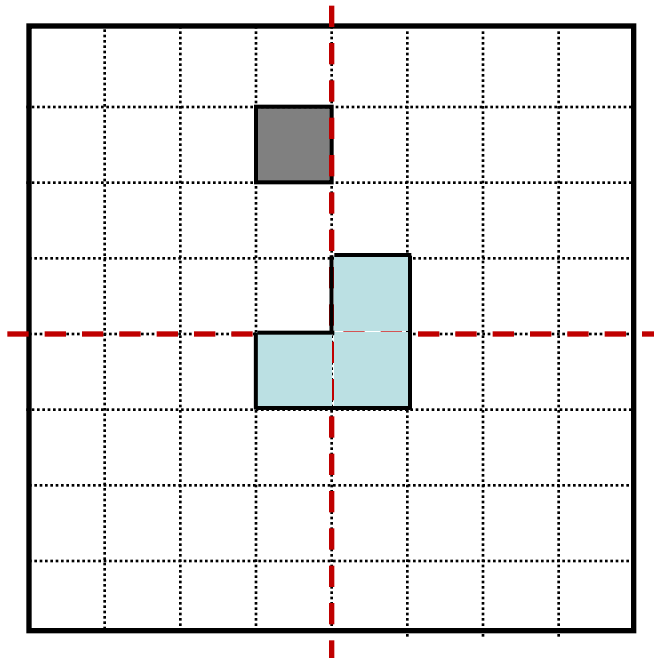
- To get smaller square boards, let's divide the original board into four boards



- Great! We have one problem of the size $(n/2) \times (n/2)$
- But: The other three problems are not similar to the original problems – they do not have holes!

Tiling: Dividing the Problem

- Idea: insert one tromino at the center to get three *imaginary* holes in each of the three smaller boards



- Now we have four boards with holes of the size $(n/2) \times (n/2)$
- Keep doing this division, until we get the 2×2 boards with holes – we know how to tile those

Tiling: Algorithm

INPUT: n – the board size ($2^k \times 2^k$ board where $n = 2^k$), L – location of the hole.

OUTPUT: tiling of the board

Tile(k , L)

if $k = 1$ **then**

Trivial case

 Tile with one tromino

return

Divide the board into four equal-sized boards

Place one tromino at the center to cut out 3 additional holes (orientation based on where existing hole, L , is)

Let L_1 , L_2 , L_3 , L_4 denote the positions of the 4 holes

Tile($k-1$, L_1)

Tile($k-1$, L_2)

Tile($k-1$, L_3)

Tile($k-1$, L_4)

Tiling: Divide-and-Conquer

- Tiling is a divide-and-conquer algorithm:
 - Just do it trivially if the board is 2×2 , else:
 - **Divide** the board into four smaller boards (introduce holes at the corners of the three smaller boards to make them look like original problems)
 - **Conquer** using the same algorithm recursively
 - **Combine** by placing a single tromino in the center to cover the three introduced holes

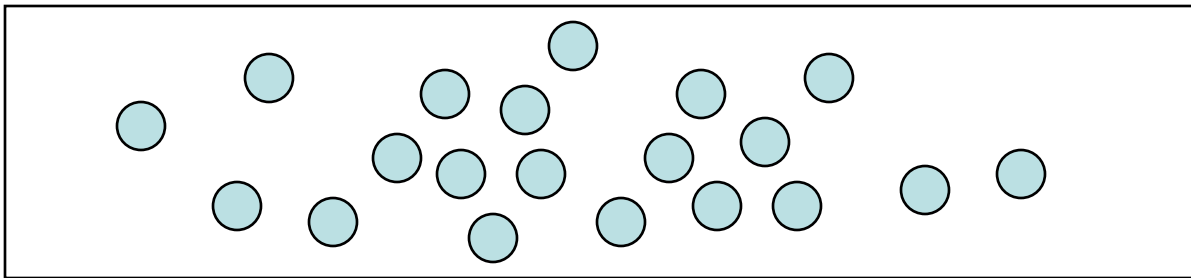
Time Complexity of Tromino Tiling

- $T(n) = 4T(n/2) + 1$
- $T(2) = 1$
- $O(n^2)$ by master theorem (or solve it iteratively)

Closest Pair Problem

□ Finding the closest pair of points

- Find the closest pair of points in a set of points. The set consists of points in two dimension plane
- Given a set P of N points, find $p, q \in P$, such that the distance $d(p, q)$ is minimum



- Application:
 - Traffic control systems: A system for controlling air or sea traffic might need to know which two vehicles are too close in order to detect potential collisions.
 - Computational geometry

Brute force algorithm

Input: set S of points

Output: closest pair of points

```
min_distance = infinity
```

```
for each point  $x$  in  $S$ 
```

```
    for each point  $y$  in  $S$ 
```

```
        if  $x \neq y$  and  $\text{distance}(x,y) < \text{min\_distance}$ 
```

```
        {
```

```
            min_distance =  $\text{dist}(x,y)$ 
```

```
            closest_pair =  $(x,y)$ 
```

```
        }
```

Time Complexity: $O(n^2)$

1 Dimension Closest Pair Problem

- Brute-force algorithm: Find all the distances $D(p, q)$ and find the minimum distance
 - Time complexity: $O(n^2)$
- 1D problem can be solved in $O(n \lg n)$ via sorting
- But, sorting does not generalize to higher dimensions
- Let's develop a divide & conquer algorithm for 2D problem.

2D Closest Pair Problem

□ Finding the closest pair of points

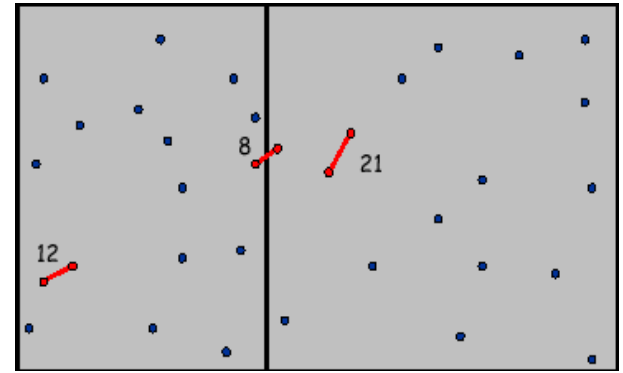
- The "closest pair" refers to the pair of points in the set that has the smallest Euclidean distance,

Distance between points $p_1=(x_1,y_1)$ and $p_2=(x_2,y_2)$

$$D(p_1, p_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

- If there are two identical points in the set, then the closest pair distance in the set will obviously be zero.

2D Closest Pair Problem



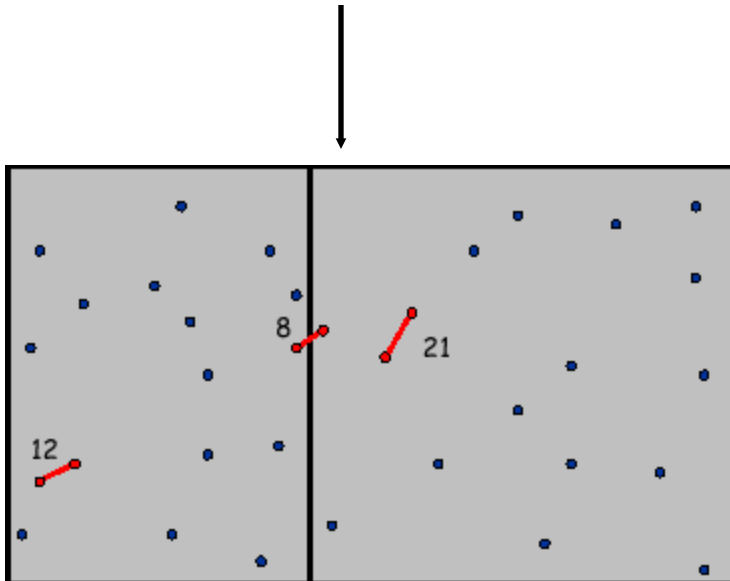
□ Finding the closest pair of points

- Divide: Sort the points by x-coordinate; draw vertical line to have roughly $n/2$ points on each side
- Conquer: Find closest pair in each side recursively
- Combine: Find closest pair with one point in each side
- Return: best of three solutions

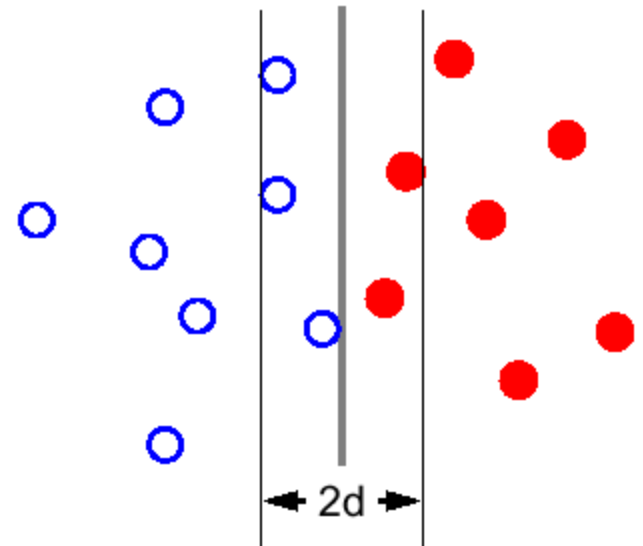
Key observation

Find the closest pair in a strip of width $2d$

Example: $d = \min(12, 21)$

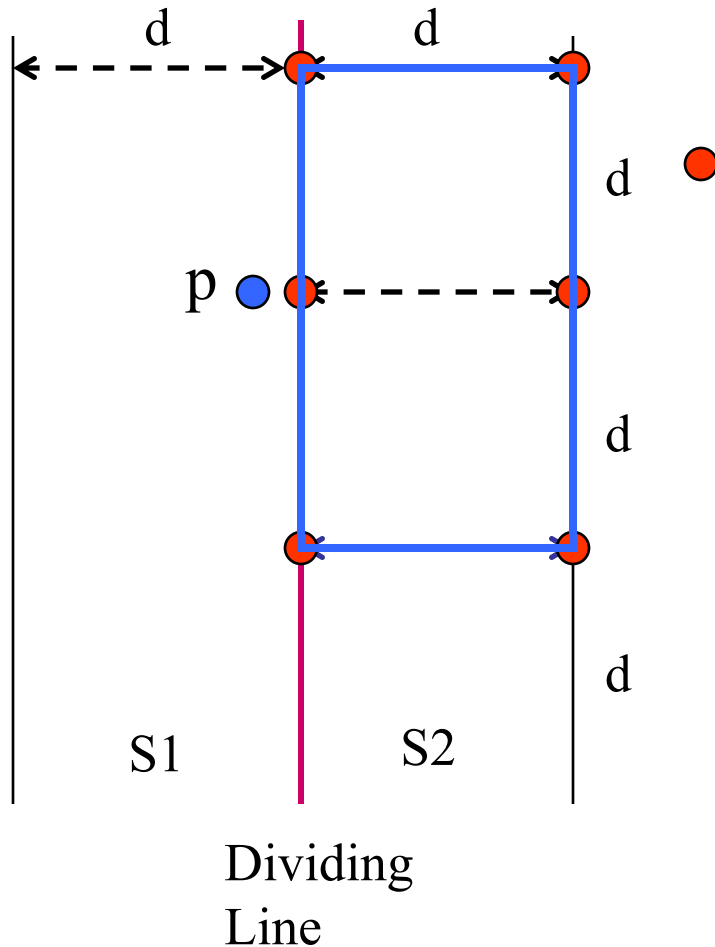


- Find the closest (\circ, \bullet) pair in a strip of width $2d$, knowing that no (\circ, \circ) or (\bullet, \bullet) pair is closer than d .



2D Closest Pair

$$d = \text{minimum}(d_{Lmin}, d_{Rmin})$$



For each point p on the left of the dividing line, we have to compare d to the distances between p and the points in the blue rectangle.

Note there must be no point inside the blue rectangle, because $d = \text{minimum}(d_{Lmin}, d_{Rmin})$. Thus, for each point p , we only have to consider 6 points – 6 red circles on the blue rectangles.

Also, note that there can be no red point between two red points on the blue box.

So, we need at most $6 \cdot n/2$ distance comparisons

2D Closest Pair

- Time Complexity
 - $T(n) = 2T(n/2) + \theta(n) = \theta(n \lg n)$
 - Solve this recurrence equation yourself by applying the iterative method

Strassen's matrix multiplication algorithm

□ Example: 2 by 2 matrix multiplication

$$m1 = (a11 + a22)(b11+b22)$$

$$m2 = (a21 + a22) b11$$

$$m3 = a11 (b12 - b22)$$

$$m4 = a22 (b21 - b11)$$

$$m5 = (a11 + a12) b22$$

$$m6 = (a21 - a11) (b11+b12)$$

$$m7 = (a12 - a22)(b21 + b22)$$

$$C = \begin{bmatrix} c11 & c12 \\ c21 & c22 \end{bmatrix} = \begin{bmatrix} a11 & a12 \\ a21 & a22 \end{bmatrix} \times \begin{bmatrix} b11 & b12 \\ b21 & b22 \end{bmatrix}$$

$$C = \begin{bmatrix} m1 + m4 - m5 + m7 & m3 + m5 \\ m2 + m4 & m1 + m3 - m2 + m6 \end{bmatrix}$$

Example 4 (cont'd)

□ Strassen's matrix multiplication algorithm

- partition $n \times n$ matrix into sub-matrices $n/2 \times n/2$
(assume n is power of 2)
- apply the seven basic calculations:

$$\begin{aligned} M1 &= (A11 + A22)(B11+B22) \\ M2 &= (A21 + A22) B11 \\ M3 &= A11 (B12 - B22) \\ M4 &= A22 (B21 - B11) \\ M5 &= (A11 + A12) B22 \\ M6 &= (A21 - A11) (B11+B12) \\ M7 &= (A12 - A22)(B21 + B22) \end{aligned}$$
$$C = \begin{bmatrix} C11 & C12 \\ C21 & C22 \end{bmatrix} = \begin{bmatrix} A11 & A12 \\ A21 & A22 \end{bmatrix} \times \begin{bmatrix} B11 & B12 \\ B21 & B22 \end{bmatrix}$$
$$C = \begin{bmatrix} M1+M4-M5+M7 & M3+M5 \\ M2+M4 & M1+M3-M2+M6 \end{bmatrix}$$

Strassen's matrix multiplication example

$$C = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 2 & 2 \\ 1 & 1 & 2 & 2 \\ 3 & 3 & 4 & 4 \\ 3 & 3 & 4 & 4 \end{bmatrix} \times \begin{bmatrix} 5 & 5 & 6 & 6 \\ 5 & 5 & 6 & 6 \\ 7 & 7 & 8 & 8 \\ 7 & 7 & 8 & 8 \end{bmatrix}$$

$$M1 = \left(\begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} + \begin{bmatrix} 4 & 4 \\ 4 & 4 \end{bmatrix} \right) \times \left(\begin{bmatrix} 5 & 5 \\ 5 & 5 \end{bmatrix} + \begin{bmatrix} 8 & 8 \\ 8 & 8 \end{bmatrix} \right)$$

$$C = \begin{bmatrix} M1 + M4 - M5 + M7 & M3 + M5 \\ M2 + M4 & M1 + M3 - M2 + M6 \end{bmatrix}$$

Strassen's matrix multiplication algorithm

□ Strassen's matrix multiplication:

input: n, A, B ;
output: C

```
strassen_matrix_multiply(int n, A, B, C)
{
    divide A into  $A_{11}, A_{12}, A_{21}, A_{22}$ ;
    divide B into  $B_{11}, B_{12}, B_{21}, B_{22}$ ;

    strassen_matrix_multiply( $n/2, A_{11}+A_{22}, B_{11}+B_{22}, M_1$ );
    strassen_matrix_multiply( $n/2, A_{21}+A_{22}, B_{11}, M_2$ );
    .....
    strassen_matrix_multiply( $n/2, A_{12}-A_{22}, B_{21}+B_{22}, M_7$ );

    Compose  $C_{11}, \dots, C_{22}$  by  $M_1, \dots, M_7$ ;
}
```

Time Complexity

□ Strassen's matrix multiplication algorithm

- Every case Time complexity

$$\text{Multiplications: } \begin{aligned} T(n) &= 7 T(n/2); \\ T(1) &= 1 \end{aligned}$$

$$T(n) = n^{\lg 7} \in \Theta(n^{2.81})$$

(while brute-force approach: $T(n) = \Theta(n^3)$)

$$\text{Additions/Subtractions: } \begin{aligned} T(n) &= 7 T(n/2) + 18 (n/2)^2 \\ T(1) &= 0 \end{aligned}$$

$$T(n) = O(n^{2.81}) \text{ by Master theorem (a=7, b=2, k=2)}$$

Combine the D&C algorithm with other simple algorithm

□ Switching point

- Recursive method may have no advantage for small n compared to an alternative, non-recursive, algorithm
- Recursive algorithm requires a fair amount of overhead
- Stop the dividing process at a certain switching point (or threshold) for recursive algorithm, and then use the alternative algorithm

□ Example

- Use the standard matrix multiplication to multiply two 2×2 matrices

When not to use D&C algorithm

□ Two cases:

- An instance of size n is divided into two or more instances each almost of size n
- An instance of size n is divided into almost n instances of size n/c (c is constant)

Fibonacci number: Divide & Conquer based on recursion is a poor choice!

