

DIVIDE-AND-CONQUER

Divide-and-Conquer paradigm

In divide-and-conquer ,the problem is solved by applying the 3 steps at each level of recursion

- **Divide** the problem into a number of subproblems that are smaller instances of the same problem.
- **Conquer** the subproblems by solving them recursively. If the subproblem sizes are small enough, however, just solve the subproblems in a straightforward manner.
- **Combine** the solutions to the subproblems into the solution for the original problem.

Recurrences are used to characterize the running times of divide-and –conquer algorithms.

A *recurrence* is an equation or inequality that describes a function in terms of its value on smaller inputs.

- If the problem size is small enough, say $n \leq c$ for some constant c , the straightforward solution takes constant time, which we write as $\Theta(1)$.
- Suppose that our division of the problem yields a subproblems, each of which is n/b the size of the original and so it takes time $aT(n/b)$ to solve a of them.
- If we take $D(n)$ time to divide the problem into subproblems and $C(n)$ time to combine the solutions to the subproblems into the solution to the original problem, we get the recurrence

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c , \\ aT(n/b) + D(n) + C(n) & \text{otherwise .} \end{cases}$$

Merge sort

The *merge sort* algorithm follows the divide-and-conquer paradigm. Intuitively, it operates as follows.

Divide: Divide the n -element sequence to be sorted into two subsequences of $n/2$ elements each.

Conquer: Sort the two subsequences recursively using merge sort.

Combine: Merge the two sorted subsequences to produce the sorted answer.

MERGE-SORT(A, p, r)

```
1  if  $p < r$ 
2       $q = \lfloor (p + r)/2 \rfloor$ 
3      MERGE-SORT( $A, p, q$ )
4      MERGE-SORT( $A, q + 1, r$ )
5      MERGE( $A, p, q, r$ )
```

MERGE(A, p, q, r)

```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else  $A[k] = R[j]$ 
17          $j = j + 1$ 
```

MERGE(A, p, q, r)

1	$n_1 = q - p + 1$	}		Constant- $\Theta(1)$	
2	$n_2 = r - q$				
3	let $L[1..n_1 + 1]$ and $R[1..n_2 + 1]$ be new arrays				
4	for $i = 1$ to n_1	}	$\Theta(n_1)$	}	$\Theta(n)$
5	$L[i] = A[p + i - 1]$				
6	for $j = 1$ to n_2				
7	$R[j] = A[q + j]$	}	$\Theta(n_2)$		
8	$L[n_1 + 1] = \infty$				
9	$R[n_2 + 1] = \infty$				
10	$i = 1$	}		Constant- $\Theta(1)$	
11	$j = 1$				
12	for $k = p$ to r	}		$\Theta(n)$	
13	if $L[i] \leq R[j]$				
14	$A[k] = L[i]$				
15	$i = i + 1$				
16	else $A[k] = R[j]$				
17	$j = j + 1$				

Merge procedure runs in $\Theta(n)$

Recurrence Relation

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c, \\ aT(n/b) + D(n) + C(n) & \text{otherwise.} \end{cases}$$

Divide: compute the middle of the subarray, which takes constant time. Thus $D(n) = \Theta(1)$.

Conquer: Recursively solve the two subproblems of size $n/2$, which contains $2T(n/2)$ to the running time.

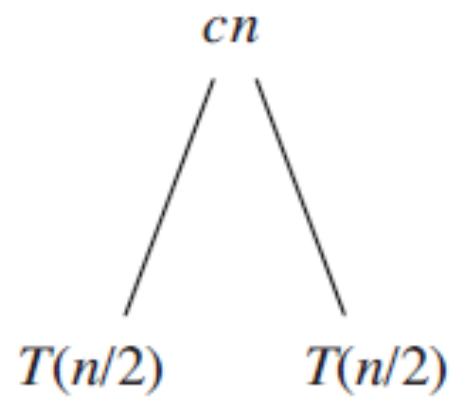
Combine: Merge procedure on an n -element subarray takes time $\Theta(n)$ and so $C(n) = \Theta(n)$

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

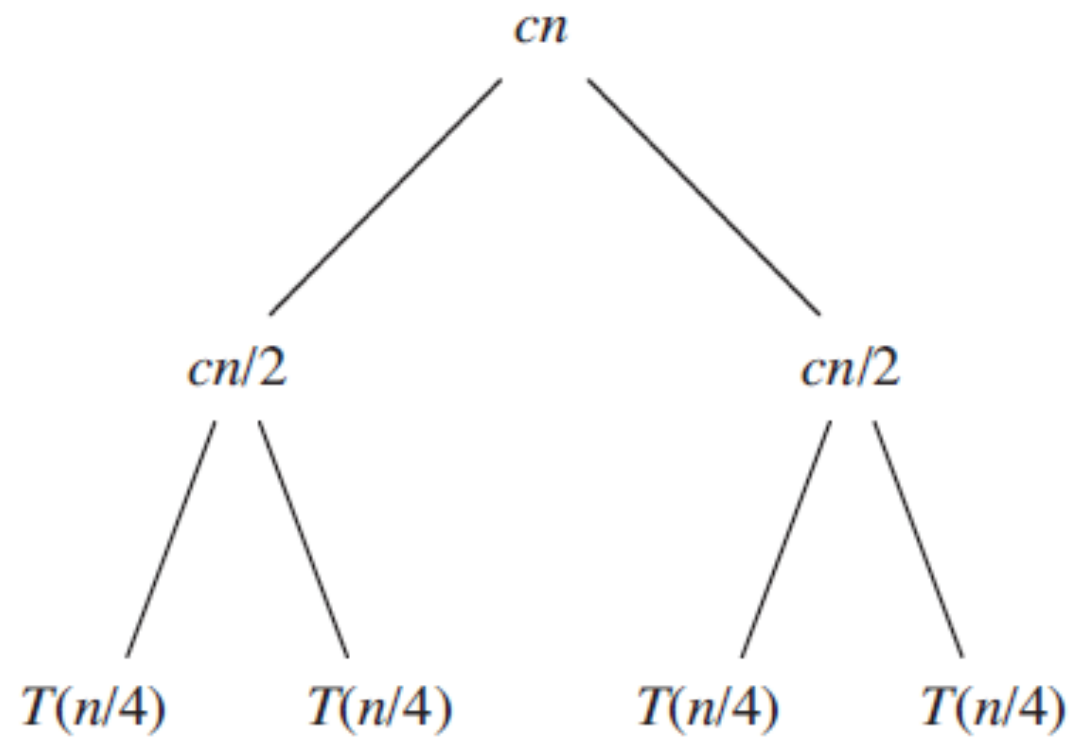
Solve the recursion relation

Recursion tree

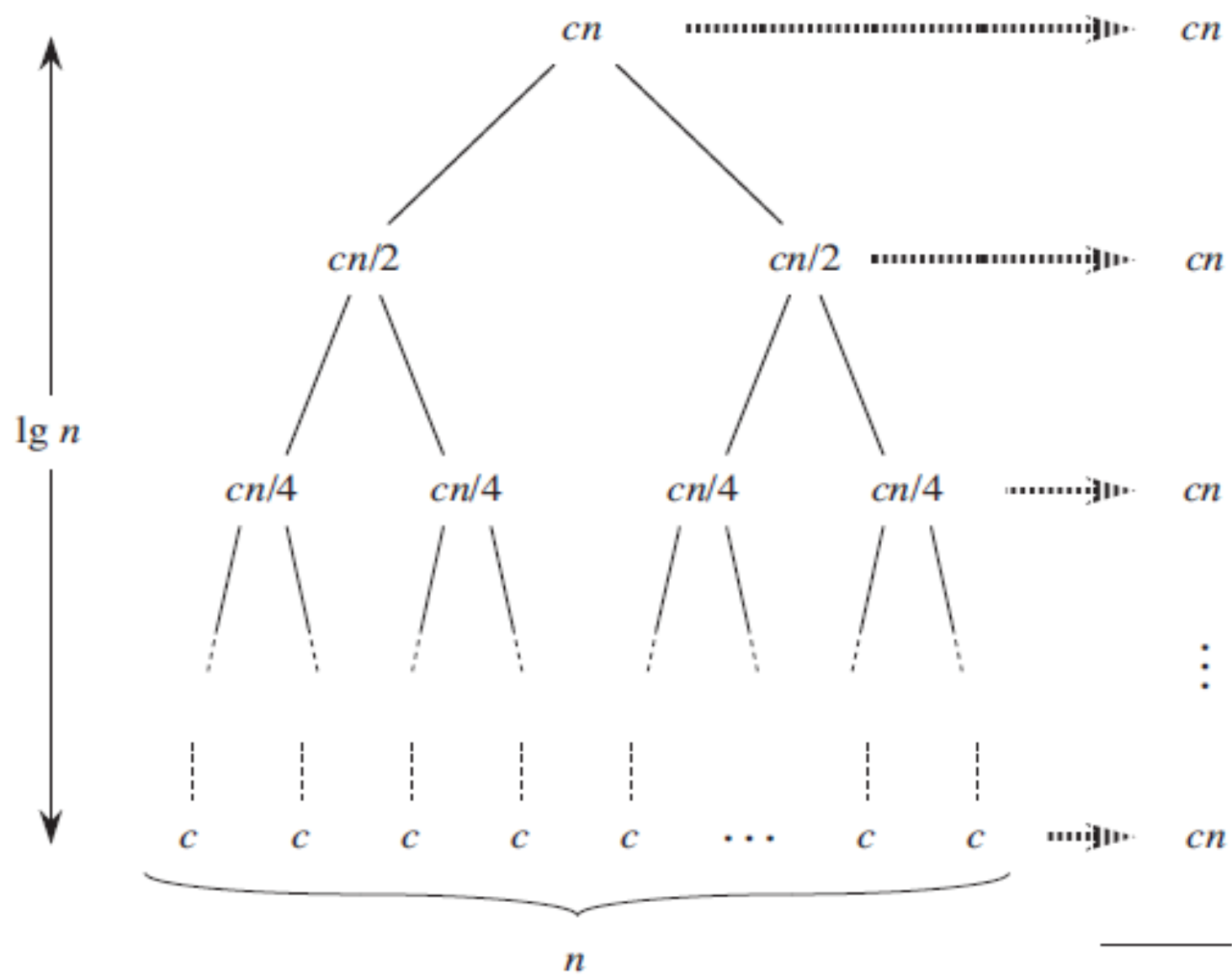
$T(n)$



(a)



(c)



(d)

Total: $cn \lg n + cn$

Quick sort

Divide: Partition (rearrange) the array $A[p..r]$ into two (possibly empty) subarrays $A[p..q-1]$ and $A[q+1..r]$ such that each element of $A[p..q-1]$ is less than or equal to $A[q]$, which is, in turn, less than or equal to each element of $A[q+1..r]$. Compute the index q as part of this partitioning procedure.

Conquer: Sort the two subarrays $A[p..q-1]$ and $A[q+1..r]$ by recursive calls to quicksort.

Combine: Because the subarrays are already sorted, no work is needed to combine them: the entire array $A[p..r]$ is now sorted.

QUICKSORT(A, p, r)

```
1  if  $p < r$ 
2       $q = \text{PARTITION}(A, p, r)$ 
3      QUICKSORT( $A, p, q - 1$ )
4      QUICKSORT( $A, q + 1, r$ )
```

PARTITION(A, p, r)

```
1   $x = A[r]$ 
2   $i = p - 1$ 
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
7  exchange  $A[i + 1]$  with  $A[r]$ 
8  return  $i + 1$ 
```

Quick sort Algorithm

QUICKSORT(A, p, r)

```
1  if  $p < r$ 
2       $q = \text{PARTITION}(A, p, r)$ 
3      QUICKSORT( $A, p, q - 1$ )
4      QUICKSORT( $A, q + 1, r$ )
```

HOARE-PARTITION(A, p, r)

```
1   $x = A[p]$ 
2   $i = p - 1$ 
3   $j = r + 1$ 
4  while TRUE
5      repeat
6           $j = j - 1$ 
7      until  $A[j] \leq x$ 
8      repeat
9           $i = i + 1$ 
10     until  $A[i] \geq x$ 
11     if  $i < j$ 
12         exchange  $A[i]$  with  $A[j]$ 
13     else return  $j$ 
```

Performance Analysis

- The worst-case behavior for quicksort occurs when the partitioning routine produces one subproblem with $n - 1$ elements and one with 0 elements.

Recursive call of array size $n-1$ is $T(n-1)$

Recursive call of array size 0 is $T(0)$ is $\Theta(1)$

Partition takes $\Theta(n)$

$$\begin{aligned} T(n) &= T(n-1) + T(0) + \Theta(n) \\ &= T(n-1) + \Theta(n) . \end{aligned}$$

The running time is $\Theta(n^2)$

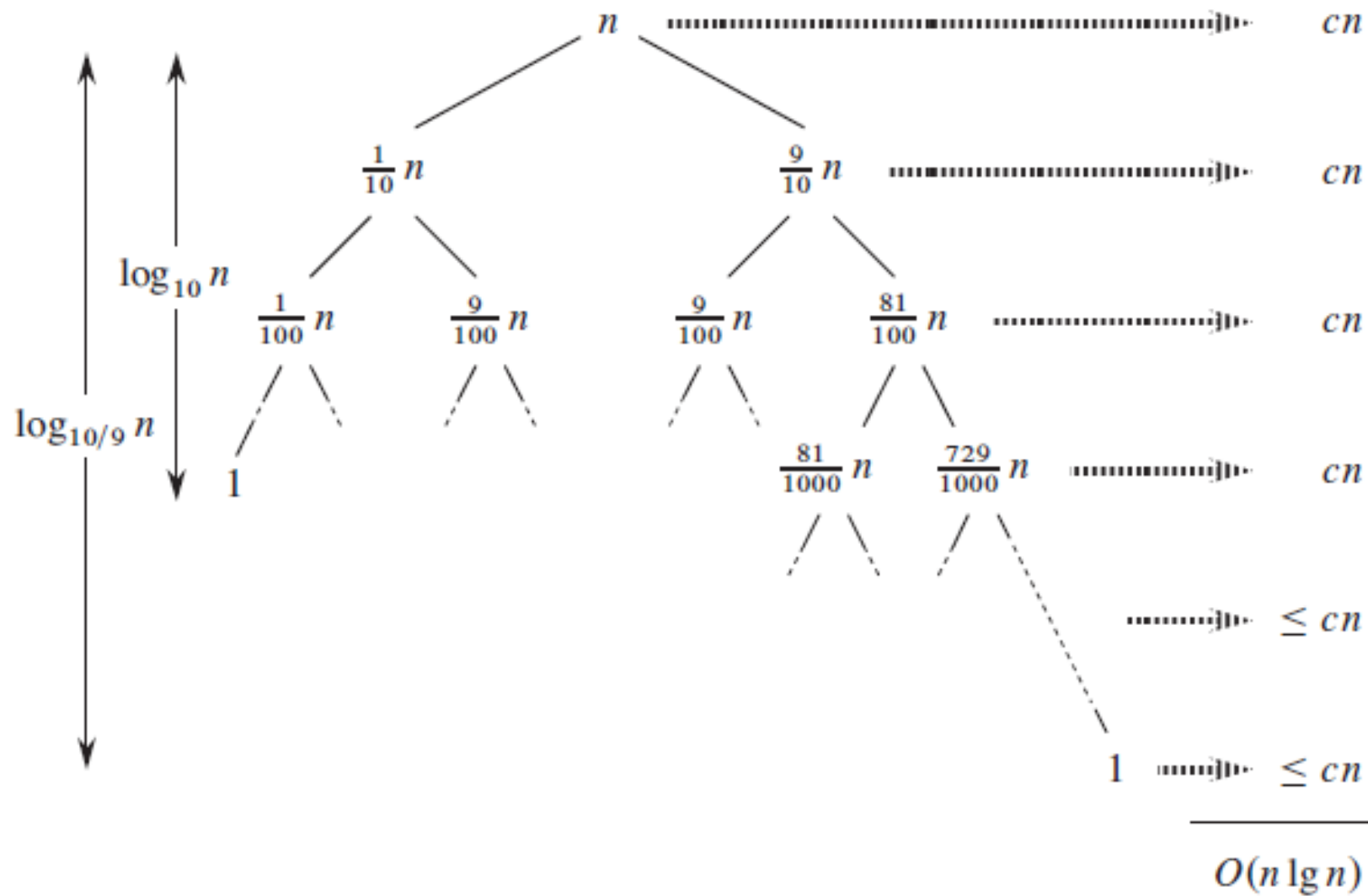
Best-case partitioning

- PARTITION produces two subproblems, each of size no more than $n/2$, since one is of size $\lfloor n/2 \rfloor$ and one of size $\lceil n/2 \rceil - 1$.

$$T(n) = 2T(n/2) + \Theta(n) ,$$

The running time is $\Theta(n \log n)$

Balanced partitioning



Finding the Maximum and Minimum

1. Algorithm StraightMaxMin (a, n, max, min)

// set max to the maximum and min to the minimum of A[1:n]

{

2. *max := min := a[1];*

3. *for i: 2 to n do*

{

4. *if (a[i] > max then max := a[i];*

5. *if a[i] < min then min := a[i];*

}

}

StraightMaxMin requires $2 \cdot (n - 1)$ element comparisons in the best, average , and worst case.

1. Algorithm MaxMin (a, i, j, Max, Min)

// i and j are the lower and upper bounds of an array 'a'. Max and min contains the maximum and minimum elements of an array 'a'

2. { if (i == j) then max:=min:= a[i]; *// Small(P)*

3. else if (i == j -1) then *// another case of Small(P)*

{

4. if (a[i] < a[j]) then

5. max: = a[j] ; min:=a[i];

else

6. max: = a[i] ; min:=a[j];

}

7. else {

// If P is not small divide P into subproblems .Find where to split the set

8. Mid:= floor((i+j)/2)

// solve the subproblems

9. MaxMin(i ,mid,max,min);

10. MaxMin(mid+1, j ,max1,min1);

// Combine the solutions

11. if (max<max1) then max:=max1;

12. if (min>min1) then min:=min1;

}

}

Maximum
and
Minimum
using
Divide-and
Conquer

22	13	-5	-8	15	60	17	31	47
1	2	3	4	5	6	7	8	9

i j max min

1	9	60	-8
---	---	----	----

1	5	22	-8
---	---	----	----

6	9	60	17
---	---	----	----

1	3	22	-5
---	---	----	----

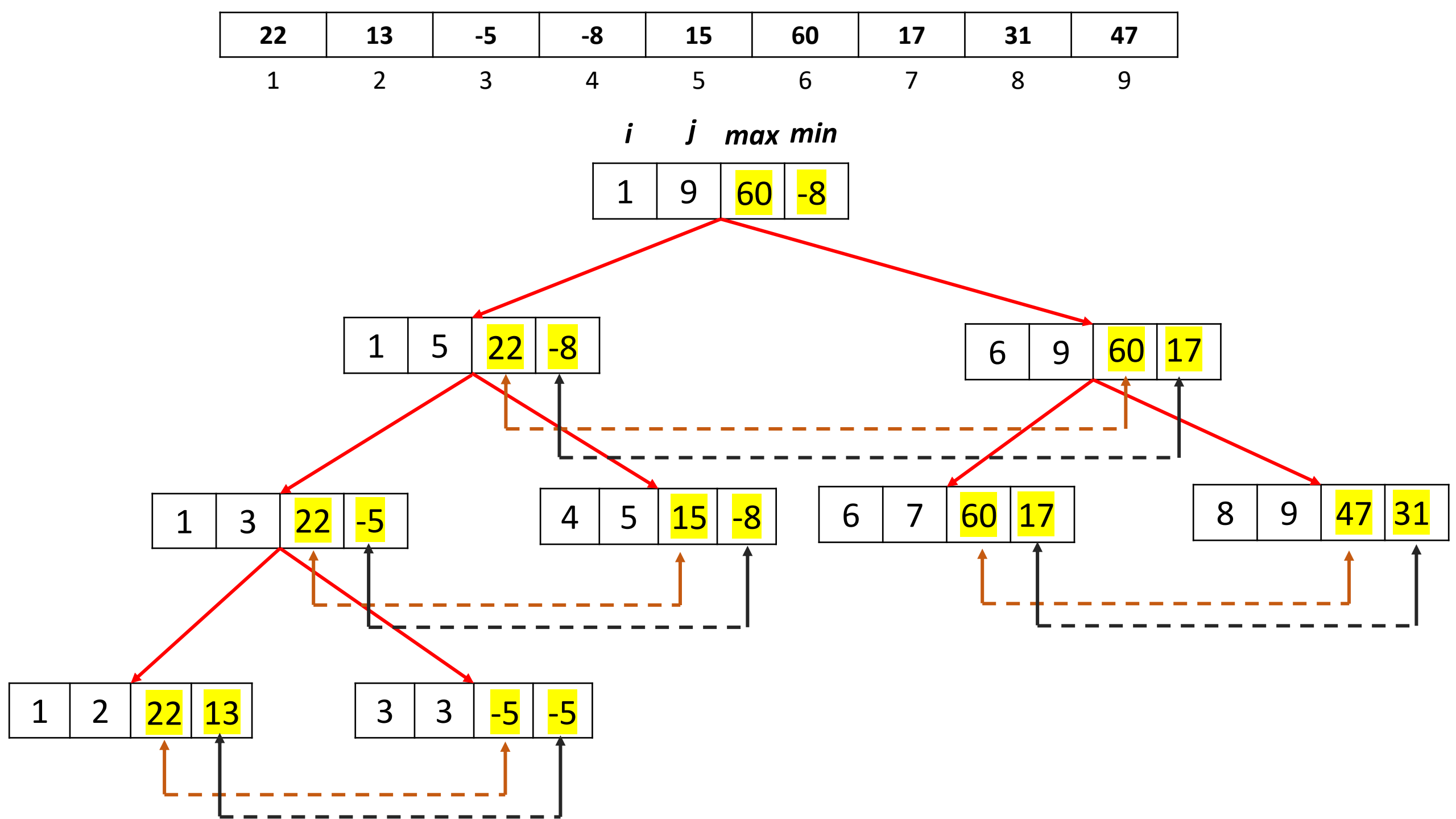
4	5	15	-8
---	---	----	----

6	7	60	17
---	---	----	----

8	9	47	31
---	---	----	----

1	2	22	13
---	---	----	----

3	3	-5	-5
---	---	----	----



Matrix Multiplication

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix}$$

$$c_{11} = a_{11} * b_{11} + a_{12} * b_{21}$$

$$c_{12} = a_{11} * b_{12} + a_{12} * b_{22}$$

$$c_{21} = a_{21} * b_{11} + a_{22} * b_{21}$$

$$c_{22} = a_{21} * b_{12} + a_{22} * b_{22}$$

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} C_{11} & C_{12} & C_{13} & C_{14} \\ C_{21} & C_{22} & C_{23} & C_{24} \\ C_{31} & C_{32} & C_{33} & C_{34} \\ C_{41} & C_{42} & C_{43} & C_{44} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & A_{34} \\ A_{41} & A_{42} & A_{43} & A_{44} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} & B_{13} & B_{14} \\ B_{21} & B_{22} & B_{23} & B_{24} \\ B_{31} & B_{32} & B_{33} & B_{34} \\ B_{41} & B_{42} & B_{43} & B_{44} \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{aligned} C_{11} &= A_{11} * B_{11} + A_{12} * B_{21} \\ C_{12} &= A_{11} * B_{12} + A_{12} * B_{22} \\ C_{21} &= A_{21} * B_{11} + A_{22} * B_{21} \\ C_{22} &= A_{21} * B_{12} + A_{22} * B_{22} \end{aligned}$$

Each recursive call multiplies two $n/2 \times n/2$ matrices.

Matrix Multiplications : 8
Matrix Additions : 4

Matrix Multiplication Algorithm

SQUARE-MATRIX-MULTIPLY-RECURSIVE(A, B)

```
1   $n = A.rows$ 
2  let  $C$  be a new  $n \times n$  matrix
3  if  $n == 1$ 
4       $c_{11} = a_{11} \cdot b_{11}$ 
5  else partition  $A, B$ , and  $C$  :
6       $C_{11} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{11}, B_{11})$ 
           +  $\text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{12}, B_{21})$ 
7       $C_{12} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{11}, B_{12})$ 
           +  $\text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{12}, B_{22})$ 
8       $C_{21} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{21}, B_{11})$ 
           +  $\text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{22}, B_{21})$ 
9       $C_{22} = \text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{21}, B_{12})$ 
           +  $\text{SQUARE-MATRIX-MULTIPLY-RECURSIVE}(A_{22}, B_{22})$ 
10 return  $C$ 
```

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1, \\ 8T(n/2) + \Theta(n^2) & \text{if } n > 1. \end{cases}$$

Recurrence relation

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1, \\ 8T(n/2) + \Theta(n^2) & \text{if } n > 1. \end{cases}$$

If $n=1$ only one scalar multiplication

If $n>1$

8 matrix multiplications of size $n/2 \times n/2$. Each recursive call takes $T(n/2)$ running time. So, 8 calls need $8 T(n/2)$

4 matrix additions of size $n/2 \times n/2$. Addition of two matrices of size $n/2 \times n/2$ takes $n^2/4$. So, 4 additions take n^2

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & A_{11} & 7 & A_{12} \\ 9 & A_{21} & 11 & A_{22} \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & B_{11} & 1 & B_{12} \\ 0 & 0 & 1 & 0 \\ 0 & B_{21} & 0 & B_{22} \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} + \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 5 & 6 \end{bmatrix} \times \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix} \times \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 5 & 6 \end{bmatrix}$$

$$\begin{aligned}
 C_{11} &= 1 \times 1 + 2 \times 0 \\
 C_{12} &= 1 \times 0 + 2 \times 1 \\
 C_{21} &= 5 \times 1 + 6 \times 0 \\
 C_{22} &= 5 \times 0 + 6 \times 1
 \end{aligned}$$

$$\begin{aligned}
 C_{11} &= 3 \times 0 + 4 \times 0 \\
 C_{12} &= 3 \times 0 + 4 \times 0 \\
 C_{21} &= 7 \times 0 + 8 \times 0 \\
 C_{22} &= 7 \times 0 + 8 \times 0
 \end{aligned}$$

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & A_{11} & 7 & A_{12} \\ 9 & A_{21} & 11 & A_{22} \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & B_{11} & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & B_{21} & 0 & 1 \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} + \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix}$$

$$\begin{aligned}
 C_{11} &= 1 \times 0 + 2 \times 0 \\
 C_{12} &= 1 \times 0 + 2 \times 0 \\
 C_{21} &= 5 \times 0 + 6 \times 0 \\
 C_{22} &= 5 \times 0 + 6 \times 0
 \end{aligned}$$

$$\begin{aligned}
 C_{11} &= 3 \times 1 + 4 \times 0 \\
 C_{12} &= 3 \times 0 + 4 \times 1 \\
 C_{21} &= 7 \times 1 + 8 \times 0 \\
 C_{22} &= 7 \times 0 + 8 \times 1
 \end{aligned}$$

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & A_{11} & 7 & A_{12} \\ 9 & A_{21} & 11 & A_{22} \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & B_{11} & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & B_{21} & 0 & 1 \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ 9 & 10 \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ 1 & 0 \\ B_{21} & B_{22} \end{bmatrix} + \begin{bmatrix} A_{22} & A_{12} \\ 11 & 12 \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ 0 & 0 \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} 9 & 10 \\ 13 & 14 \end{bmatrix} + \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 9 & 10 \\ 13 & 14 \end{bmatrix}$$

$$\begin{aligned}
 C_{11} &= 9 \times 1 + 10 \times 0 \\
 C_{12} &= 9 \times 0 + 10 \times 1 \\
 C_{21} &= 13 \times 1 + 14 \times 0 \\
 C_{22} &= 13 \times 0 + 14 \times 1
 \end{aligned}$$

$$\begin{aligned}
 C_{11} &= 11 \times 0 + 12 \times 0 \\
 C_{12} &= 11 \times 0 + 12 \times 0 \\
 C_{21} &= 15 \times 0 + 16 \times 0 \\
 C_{22} &= 15 \times 0 + 16 \times 1
 \end{aligned}$$

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & A_{11} & 7 & A_{12} \\ 9 & A_{21} & 11 & A_{22} \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & B_{11} & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & B_{21} & 0 & 1 \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ 9 & 10 \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ 0 & 0 \\ B_{21} & B_{22} \end{bmatrix} + \begin{bmatrix} A_{22} & A_{12} \\ 11 & 12 \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ 1 & 0 \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 11 & 12 \\ 15 & 16 \end{bmatrix} = \begin{bmatrix} 11 & 12 \\ 15 & 16 \end{bmatrix}$$

$$\begin{aligned}
 C_{11} &= 9 \times 0 + 10 \times 0 \\
 C_{12} &= 9 \times 0 + 10 \times 0 \\
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 C_{22} &= 13 \times 0 + 14 \times 0
 \end{aligned}$$

$$\begin{aligned}
 C_{11} &= 11 \times 1 + 12 \times 0 \\
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 C_{21} &= 15 \times 1 + 16 \times 0 \\
 C_{22} &= 15 \times 0 + 16 \times 1
 \end{aligned}$$

Matrix Multiplication- using divide and conquer

n/2

$$\begin{bmatrix} C_{11} & C_{12} & C_{13} & C_{14} \\ C_{21} & C_{22} & C_{23} & C_{24} \\ C_{31} & C_{32} & C_{33} & C_{34} \\ C_{41} & C_{42} & C_{43} & C_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

n/2

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 5 & 6 \end{bmatrix} \times \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 5 & 6 \end{bmatrix}$$

$$\begin{aligned}
 C_{11} &= 1 \times 1 + 2 \times 0 \\
 C_{12} &= 1 \times 0 + 2 \times 1 \\
 C_{21} &= 5 \times 1 + 6 \times 0 \\
 C_{22} &= 5 \times 0 + 6 \times 1
 \end{aligned}$$

Performance Analysis

$$T(n) = 8 * T(n/2) + cn^2$$

$$= 8 * [8 * T(n/4) + cn^2/4] + cn^2$$

$$= 2^6 * [8 * T(n/8) + cn^2/8] + 2cn^2 + \cancel{cn^2} + 2^6 * T(n/4) + 2cn^2 + cn^2$$

$$= 2^9 * T(n/8) + 4cn^2 + 2cn^2 + cn^2$$

...

...

$$= (2^3)^k * T(n / 2^K) + (2^K - 1) cn^2$$

$$= (8)^k * T(n / 2^K) + (2^K - 1) cn^2$$

Let $2^K = n$ then $k = \log_2 n$

$$= (8)^{\log_2 n} * T(n / n) + (n - 1) cn^2$$

$$= (2^3)^{\log_2 n} * T(n / n) + (n - 1) cn^2$$

$$= (2)^{\log_2 n^3} + (n - 1) cn^2$$

$$= n^3 + (n - 1) cn^2$$

$$= \Theta(n^3)$$

Strassen's Matrix Multiplication

- The key to Strassen's method is to perform seven recursive multiplications instead of performing eight recursive multiplications of $n/2 \times n/2$.
- The cost of eliminating one matrix multiplication will create several new additions of $n/2 \times n/2$ matrices, but still only a constant number .

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1 , \\ 7T(n/2) + \Theta(n^2) & \text{if } n > 1 . \end{cases}$$

Strassen's matrix multiplication method

1. Divide the input matrices A and B and output matrix C into $n/2 \times n/2$ submatrices. This step takes $\Theta(1)$ time by index calculation, just as in SQUARE-MATRIX-MULTIPLY-RECURSIVE.
2. Create 10 matrices S_1, S_2, \dots, S_{10} , each of which is $n/2 \times n/2$ and is the sum or difference of two matrices created in step 1. We can create all 10 matrices in $\Theta(n^2)$ time.
3. Using the submatrices created in step 1 and the 10 matrices created in step 2, recursively compute seven matrix products P_1, P_2, \dots, P_7 . Each matrix P_i is $n/2 \times n/2$.
4. Compute the desired submatrices $C_{11}, C_{12}, C_{21}, C_{22}$ of the result matrix C by adding and subtracting various combinations of the P_i matrices. We can compute all four submatrices in $\Theta(n^2)$ time.

Step 2

$$S_1 = B_{12} - B_{22} ,$$

$$S_2 = A_{11} + A_{12} ,$$

$$S_3 = A_{21} + A_{22} ,$$

$$S_4 = B_{21} - B_{11} ,$$

$$S_5 = A_{11} + A_{22} ,$$

$$S_6 = B_{11} + B_{22} ,$$

$$S_7 = A_{12} - A_{22} ,$$

$$S_8 = B_{21} + B_{22} ,$$

$$S_9 = A_{11} - A_{21} ,$$

$$S_{10} = B_{11} + B_{12} .$$

Step 3

$$P_1 = A_{11} \cdot S_1 = A_{11} \cdot B_{12} - A_{11} \cdot B_{22} ,$$

$$P_2 = S_2 \cdot B_{22} = A_{11} \cdot B_{22} + A_{12} \cdot B_{22} ,$$

$$P_3 = S_3 \cdot B_{11} = A_{21} \cdot B_{11} + A_{22} \cdot B_{11} ,$$

$$P_4 = A_{22} \cdot S_4 = A_{22} \cdot B_{21} - A_{22} \cdot B_{11} ,$$

$$P_5 = S_5 \cdot S_6 = A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22} ,$$

$$P_6 = S_7 \cdot S_8 = A_{12} \cdot B_{21} + A_{12} \cdot B_{22} - A_{22} \cdot B_{21} - A_{22} \cdot B_{22} ,$$

$$P_7 = S_9 \cdot S_{10} = A_{11} \cdot B_{11} + A_{11} \cdot B_{12} - A_{21} \cdot B_{11} - A_{21} \cdot B_{12} .$$

Step 4

$$C_{11} = P_5 + P_4 - P_2 + P_6 .$$

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{22} = P_5 + P_1 - P_3 - P_7 ,$$

Matrix Multiplication- using divide and conquer

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} & c_{13} & c_{14} \\ c_{21} & c_{22} & c_{23} & c_{24} \\ c_{31} & c_{32} & c_{33} & c_{34} \\ c_{41} & c_{42} & c_{43} & c_{44} \end{bmatrix} = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 5 & A_{11} & 7 & A_{12} \\ 9 & A_{21} & 11 & A_{22} \\ 13 & 14 & 15 & 16 \end{bmatrix} \times \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & B_{11} & 1 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & B_{21} & 0 & 1 \end{bmatrix}$$

$n/2$

$$\begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} + \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 \\ 5 & 6 \end{bmatrix} \times \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} + \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix} \times \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 7 & 8 \end{bmatrix}$$

In place of recursive calls straseen's use his formulas
 S_1 to S_{10} ,
 P_1 to P_{10} ,
 C_{11} to C_{22}

$$\begin{aligned}
 C_{11} &= 1 \times 0 + 2 \times 0 \\
 C_{12} &= 1 \times 0 + 2 \times 0 \\
 C_{21} &= 5 \times 0 + 6 \times 0 \\
 C_{22} &= 5 \times 0 + 6 \times 0
 \end{aligned}$$

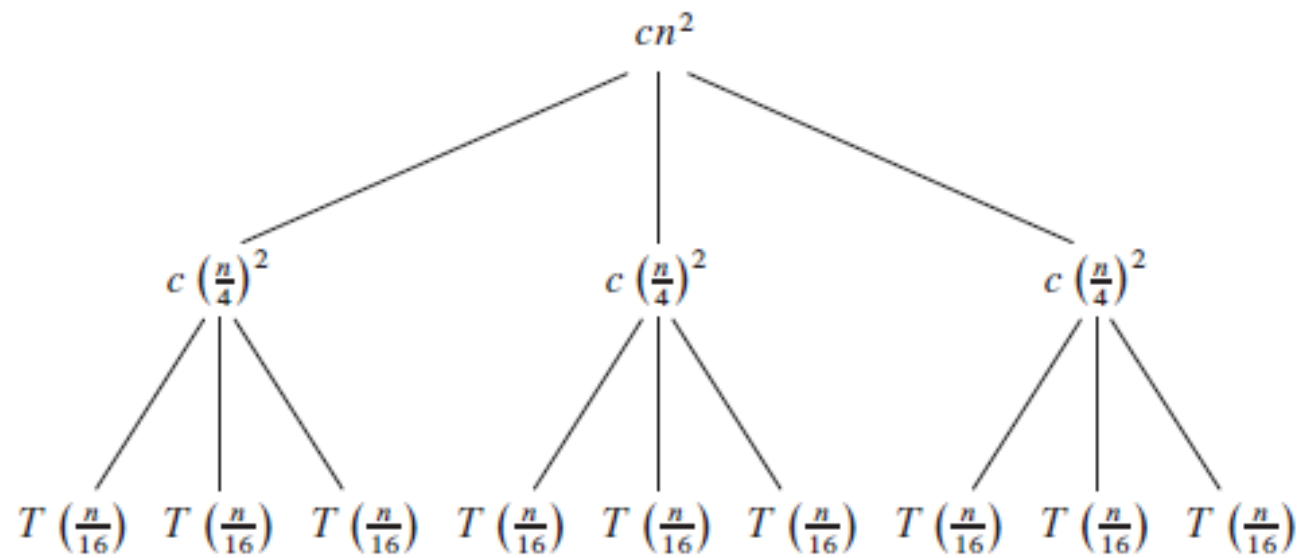
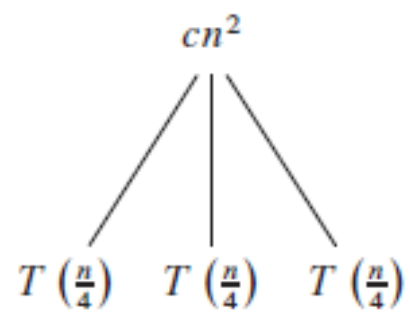
$$\begin{aligned}
 C_{11} &= 3 \times 1 + 4 \times 0 \\
 C_{12} &= 3 \times 0 + 4 \times 1 \\
 C_{21} &= 7 \times 1 + 8 \times 0 \\
 C_{22} &= 7 \times 0 + 8 \times 1
 \end{aligned}$$

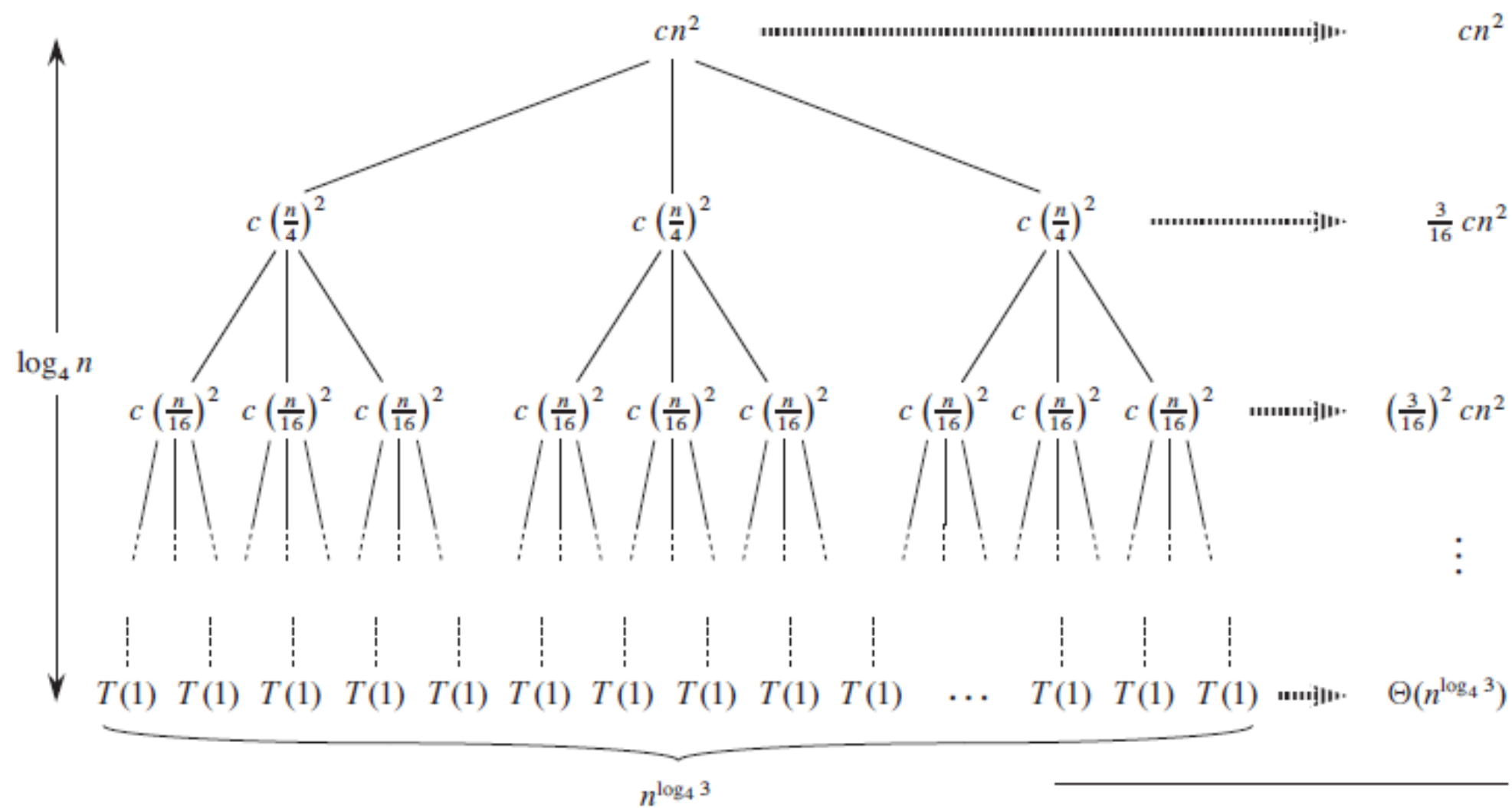
Performance Analysis of Strassen's Matrix multiplication

$$\begin{aligned}T(n) &= 7 * T(n/2) + cn^2 \\&= 7 * [7 * T(n/4) + cn^2/4] + cn^2 \\&= 7^2 * [7 * T(n/8) + cn^2/16] + 7cn^2 /4 + cn^2\end{aligned}$$

Solving the Recurrence Relation using recursion tree

$$T(n) = 3T(n/4) + cn^2,$$





(d)

Total: $O(n^2)$

Master's Theorem

Let $a \geq 1$ and $b > 1$ be constants, let $f(n)$ be a function, and let $T(n)$ be defined on the nonnegative integers by the recurrence

$$T(n) = aT(n/b) + f(n) ,$$

where we interpret n/b to mean either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then $T(n)$ has the following asymptotic bounds:

1. If $f(n) = O(n^{\log_b a - \epsilon})$ for some constant $\epsilon > 0$, then $T(n) = \Theta(n^{\log_b a})$.
2. If $f(n) = \Theta(n^{\log_b a})$, then $T(n) = \Theta(n^{\log_b a} \lg n)$.
3. If $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and if $af(n/b) \leq cf(n)$ for some constant $c < 1$ and all sufficiently large n , then $T(n) = \Theta(f(n))$. ■