



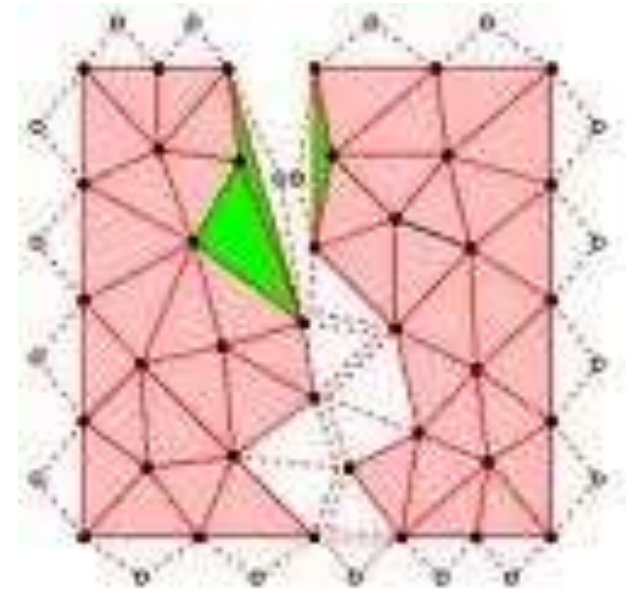
Divide & Conquer Strategy

General Method, Finding the maximum and minimum,
Merge Sort, Quick Sort

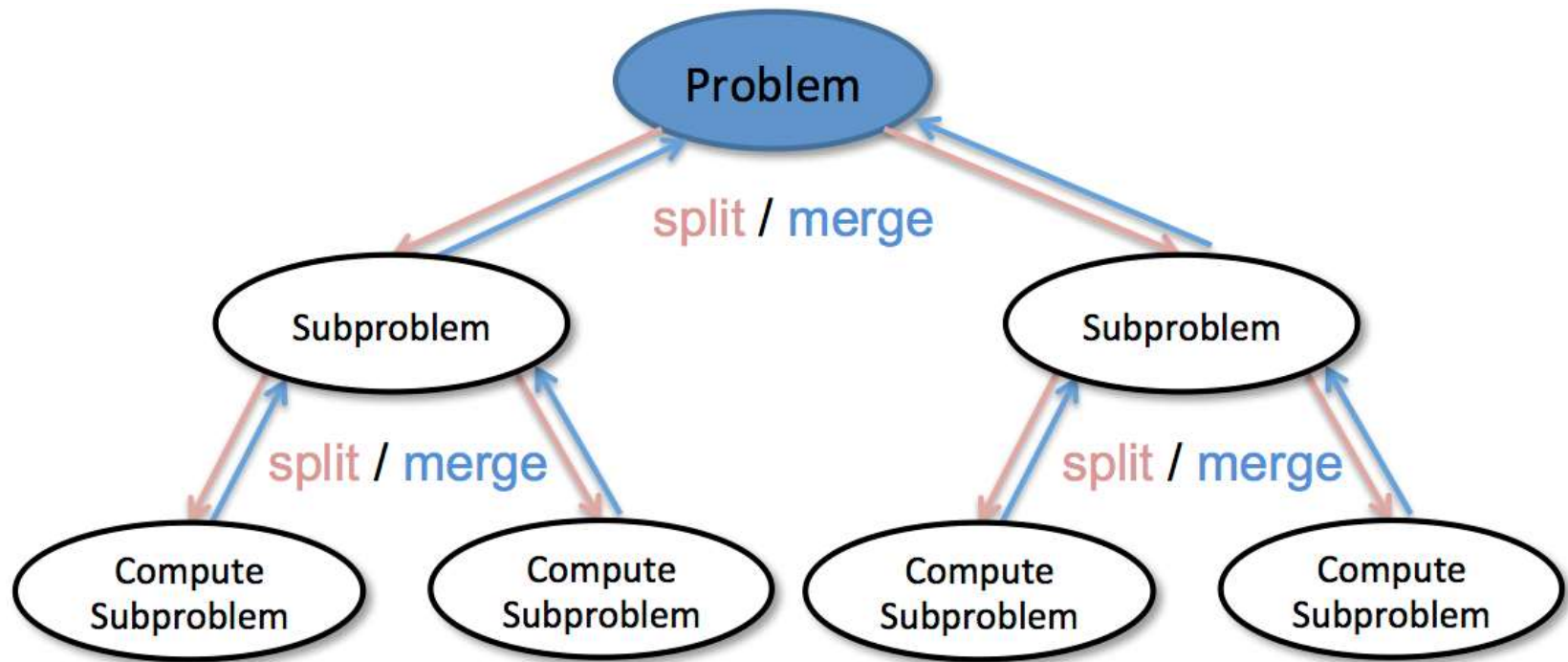
Divide And Conquer Method



- ▶ General Method
- ▶ Finding the maximum and minimum
- ▶ Binary Search
- ▶ Merge Sort
- ▶ Quick Sort



General Method



General Concept of Divide & Conquer



- ▶ Given a function to compute on n inputs, the divide-and-conquer strategy consists of:
 - ▶ splitting the inputs into k distinct subsets, $1 < k \leq n$, yielding k subproblems.
 - ▶ solving these subproblems
 - ▶ combining the subsolutions into solution of the whole.
 - ▶ if the subproblems are relatively large, then divide_Conquer is applied again.
 - ▶ if the subproblems are small, they are solved without splitting.

The Divide and Conquer Algorithm



```
Divide_Conquer(problem P)
{
    if Small(P) return S(P);
    else {

        divide P into smaller instances  $P_1, P_2, \dots, P_k, k \geq 1$ ;

        Apply Divide_Conquer to each of these subproblems;

        return Combine(Divide_Conquer( $P_1$ ), Divide_Conquer( $P_2$ ), ...,
            Divide_Conquer( $P_k$ ));
    }
}
```

Divide_Conquer recurrence relation



- ▶ The computing time of Divide_Conquer is

$$T(n) = \begin{cases} g(n) & n \text{ small} \\ T(n_1) + T(n_2) + \dots + T(n_k) + f(n) & \text{otherwise} \end{cases}$$

- ▶ $T(n)$ is the time for Divide_Conquer on any input size n .
- ▶ $g(n)$ is the time to compute the answer directly (for small inputs)
- ▶ $f(n)$ is the time for dividing P and combining the solutions.

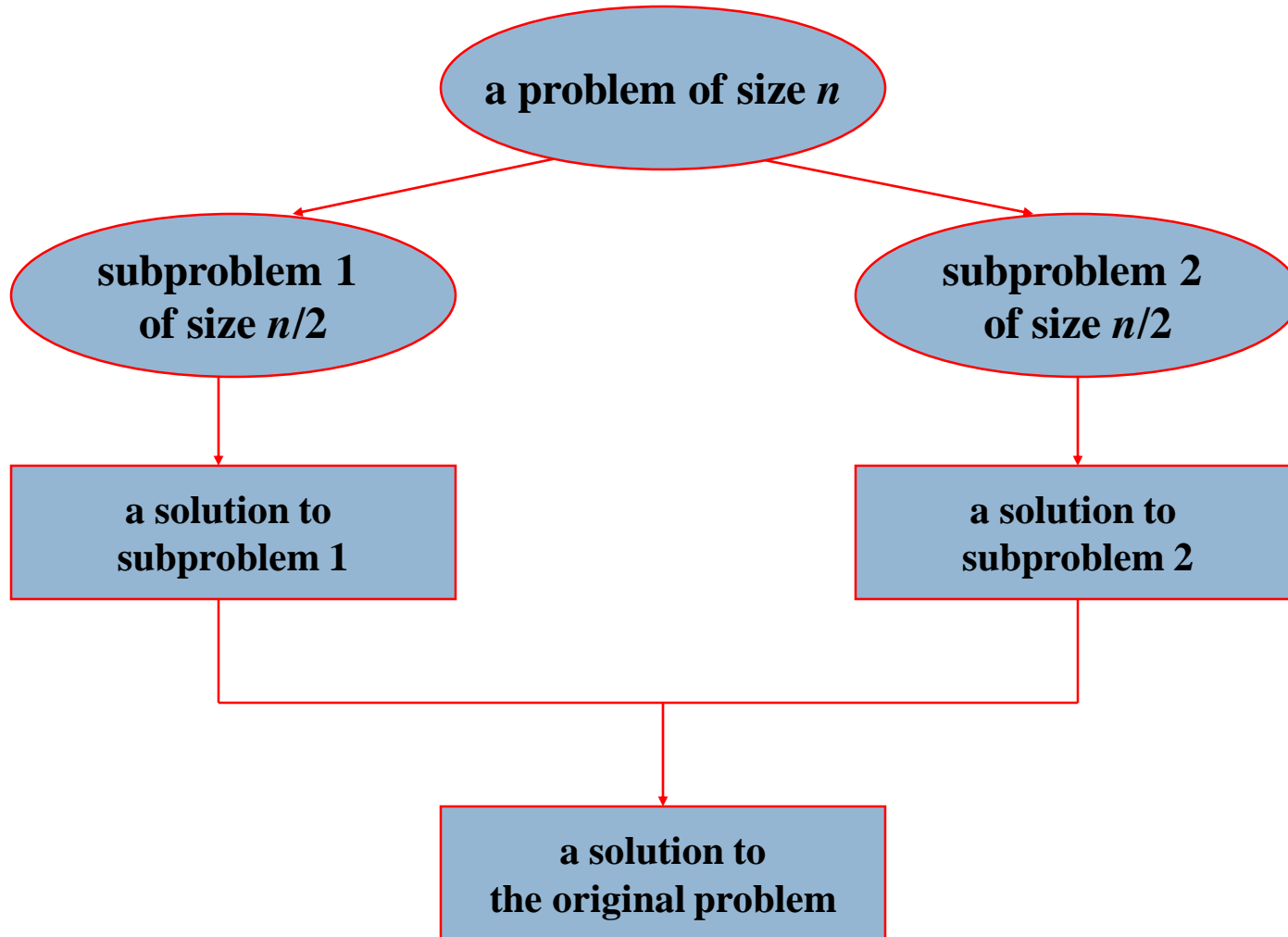


Three Steps of The Divide and Conquer Approach

The most well known algorithm design strategy:

1. **Divide** the problem into two or more smaller subproblems.
2. **Conquer** the subproblems by solving them recursively.
3. **Combine** the solutions to the subproblems into the solutions for the original problem.

A Typical Divide and Conquer Case



An Example: Calculating $a_0 + a_1 + \dots + a_{n-1}$



ALGORITHM RecursiveSum($A[0..n-1]$)

//Input: An array $A[0..n-1]$ of orderable elements

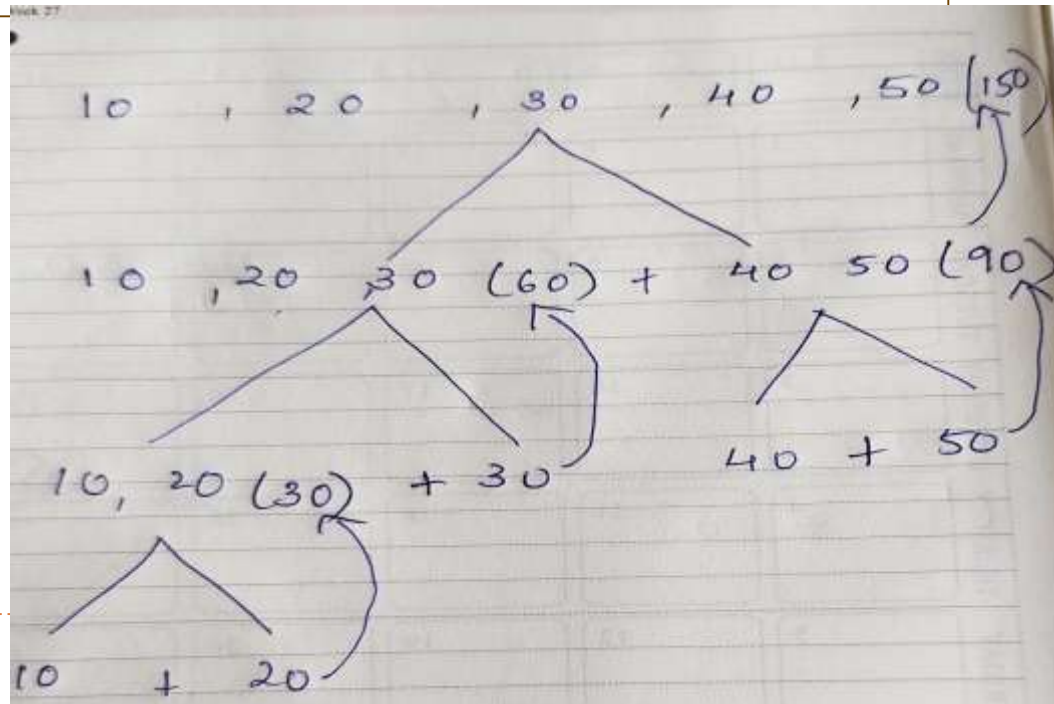
//Output: the summation of the array elements

if $n > 1$

return (RecursiveSum($A[0.. \lfloor n/2 \rfloor - 1]$) + RecursiveSum($A[\lfloor n/2 \rfloor .. n-1]$))

Efficiency: (for $n = 2^k$)

- ▶ $A(n) = 2A(n/2) + 1, n > 1$
- ▶ $A(1) = 1;$





$$A(n) = 2A(n/2) + 1, n > 1$$

$$A(1) = 1;$$

$$A(n) = 2A(n/2) + 1$$

$$= 2[2A(n/4) + 1] + 1$$

$$= 4A(n/4) + 2 + 1$$

$$= 4[2A(n/8) + 1] + 2 + 1$$

$$= 8A(n/8) + 4 + 2 + 1$$

$$= 2^3 A(n/2^3) + 2^2 + 2 + 1$$

$$= 2^3 A(n/2^3) + 2^2 + 2 + 1 \dots$$

$$= 2^k A(n/2^k) + 2^{k-1} + \dots + 2 + 1$$

$$\text{for } A(1), n/2^k = 1 \Rightarrow n = 2^k$$

$$= 2^k + 2^{k-1} + \dots + 2 + 1 \quad 2^k + 2^{k-1} + \dots + 2 + 1 = 2^{k+1} - 1$$

$$= 2^{k+1} - 1 = 2^k \cdot 2 - 1 = 2n - 1 = O(n)$$



Solving Recurrence Equations



▶ Substitution method

▶ Forward

- ▶ Uses initial condition in the initial term and generates next term.
- ▶ Repeated until some formula is guessed

▶ Backward

- ▶ Values are substituted recursively to derive some formula

▶ Master's Method



Recurrence Relation



- ▶ Equation that defines a sequence recursively
- ▶ Eg1: $T(n)=T(n-1)+n, n>0$ \Rightarrow Recurrence relation
 $T(0)=0$ \Rightarrow Initial Condition
- Eg2: $f(n)= 2f(n-1)+1$ for $n>1$
 $f(0)=0$, Solve the recurrence relation





Substitution Method

► $T(n) = T(n-1) + n, n > 0$ *Recurrence relation*

$T(0) = 0, n \leq 0$ *Initial Condition*

Forward Substitution

$$T(n) = T(n-1) + n$$

$$T(0) = 0$$

$$T(1) = T(0) + 1 = 0 + 1$$

$$T(2) = T(1) + 2 = 0 + 1 + 2$$

$$T(3) = T(2) + 3 = 0 + 1 + 2 + 3 \dots$$

$$T(n) = T(n-1) + n = 0 + 1 + 2 + 3 + \dots + n$$

$$T(n) = 0 + 1 + 2 + 3 + \dots + n = n * (n+1) / 2 = \mathbf{O(n^2)}$$

Backward Substitution

$$T(n) = T(n-1) + n$$

$$= [T(n-2) + n-1] + n$$

$$= [T(n-3) + n-2] + (n-1) + n \dots$$

$$= [T(n-k) + n-(k-1)] + \dots + (n-2) + (n-1) + n$$

If it terminates at the kth step i.e $n-k=0$

$$= T(n-k) + n-(k-1) + \dots + (n-2) + (n-1) + n$$

$$= T(0) + 1 + 2 + \dots + (n-2) + (n-1) + n$$

$$= 0 + 1 + 2 + \dots + (n-2) + (n-1) + n$$

$$= n * (n+1) / 2 = \mathbf{O(n^2)}$$





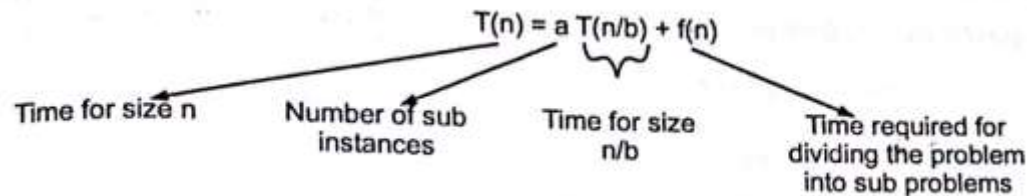
The Master Method

- ▶ The master method is used to solve recurrences of the type $T(n) = aT(n/b) + f(n)$ where $a \geq 1$ and $b > 1$
- ▶ The complexity of the divide and conquer algorithms is given by recurrences of the form

$$T(n) = \begin{cases} T(1) & n=1 \\ aT(n/b) + f(n) & n>1 \end{cases}$$

Where a and b are constants

We assume that $T(1)$ is known and n is a power of b that is $n = b^k$.





General Divide and Conquer recurrence

The Master Theorem

the time spent on solving a subproblem of size n/b .

$$T(n) = aT(n/b) + f(n), \text{ where } f(n) \in \Theta(n^k)$$

Master Theorem can be stated for efficiency analysis as:

1. $a < b^k$

$$T(n) \in \Theta(n^k)$$

2. $a = b^k$

$$T(n) \in \Theta(n^k \lg n)$$

3. $a > b^k$

$$T(n) \in \Theta(n^{\log_b a})$$

the time spent on dividing the problem
into smaller ones and combining their solutions.

Solve the recurrence relation: $T(n) = 4T(n/2) + n$