

# Recursion, Divide & Conquer

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# Function

- Function in Math:
- e.g.  $f(x) = x^2 + 2x + 1$ ,  $g(x) = \sin x + x$ , ...
- Process the input  $x$  and return the computed value
  
- Function in Programming:
- Similar to that in Math
- A subroutine that processes the input (or parameters) and return some outputs

# Function in OI

```
int f(int x){  
    int y = x * x + 2 * x + 1;  
    return y;  
}
```

```
int main(){  
    int ans = f(10);  
    cout << ans;  
    return 0;  
}
```

Define the function **name** and **input**  
Process the given input  
Output the **required value**

Call the function with  $x = 10$

# Function in OI

Functions can have several parameters:

```
int dist(double x1, double y1, double x2, double y2){  
    double distance = sqrt((x1 - x2) * (x1 - x2) + (y1 - y2) * (y1 - y2));  
    return distance;  
}
```

# Procedure

- Similar to functions
- A subroutine that processes the input (or parameters) but does not return any outputs

# Procedure

```
void hello(int n){  
    for (int i = 1; i <= n; i++)  
        cout << "Hello world!\n";  
}
```

```
int main(){  
    hello(10);  
    return 0;  
}
```

Define the procedure **name** and **input**  
Process the given input

Call the procedure with  $n = 10$

# Recursion

- A function / procedure can be called inside itself

```
int factorial(int x){  
    int y = factorial(x - 1) * x;  
    return y;  
}
```

```
void helloWorlds(int n){  
    cout << "Hello world " << n << endl;  
    helloWorlds(n - 1);  
}
```



# Recursion

- Consider the following code:

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

```
int main(){  
    cout << fact(3);  
}
```

# Recursion

```
int main(){  
    cout << fact(3);  
}
```

# Recursion

```
int main(){  
    cout << fact(3);  
}  
    ↓  
    (x = 3)  
    int fact(int x){  
        →    if (x == 1) return 1;  
        int y = fact(x - 1) * x;  
        return y;  
    }
```

# Recursion

```
int main(){  
    cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



# Recursion

```
int main(){  
    cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



(x = 2)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
    cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



(x = 2)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
  cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
  if (x == 1) return 1;  
  int y = fact(x - 1) * x;  
  return y;  
}
```



(x = 2)

```
int fact(int x){  
  if (x == 1) return 1;  
  int y = fact(x - 1) * x;  
  return y;  
}
```



(x = 1)

```
int fact(int x){  
  if (x == 1) return 1;  
  int y = fact(x - 1) * x;  
  return y;  
}
```



Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
    cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



(x = 2, y = 2)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = 1 * x;  
    return y;  
}
```



(x = 1)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

Each function call has their own local variables (x and y)  
They are only used in their own subroutine



# Recursion

```
int main(){  
    cout << fact(3);  
}
```



(x = 3)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```



(x = 2, y = 2)

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = 1 * x;  
    return y;  
}
```



```
(x = 1)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
    cout << fact(3);  
}
```



```
(x = 3, y = 6)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 2 * x;  
    return y;  
}
```



```
(x = 2, y = 2)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 1 * x;  
    return y;  
}
```

```
(x = 1)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
    cout << fact(3);  
}
```



```
(x = 3, y = 6)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 2 * x;  
    return y;  
}
```



```
(x = 2, y = 2)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 1 * x;  
    return y;  
}
```

```
(x = 1)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

Each function call has their own local variables (x and y)  
They are only used in their own subroutine

# Recursion

```
int main(){  
    cout << 6;  
}
```

```
(x = 3, y = 6)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 2 * x;  
    return y;  
}
```

```
(x = 2, y = 2)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = 1 * x;  
    return y;  
}
```

```
(x = 1)  
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

# Recursion

- Using recursion, some computations can be done with simpler code
- More importantly, recursion help us to solve problems with the following properties:
  - The problem can be divided into same problem with smaller parameter
  - We need information of the sub-problem(s) to solve the current one

# Recursion

```
int fact(int x){  
    if (x == 1) return 1;  
    int y = fact(x - 1) * x;  
    return y;  
}
```

Base case: to stop the recursion

Recurrent relation: to further  
divide the problem into smaller one

# Example – Fibonacci Numbers

- Find the n-th Fibonacci Number
- Base case: When  $n = 1$  or  $2$ ,  $f(n)$  is  $1$
- Recurrence relation: Otherwise  $f(n) = f(n - 1) + f(n - 2)$

# Example – Fibonacci Numbers

- Find the n-th Fibonacci Number
- Base case: When  $n = 1$  or  $2$ ,  $f(n)$  is  $1$
- Recurrence relation: Otherwise  $f(n) = f(n - 1) + f(n - 2)$

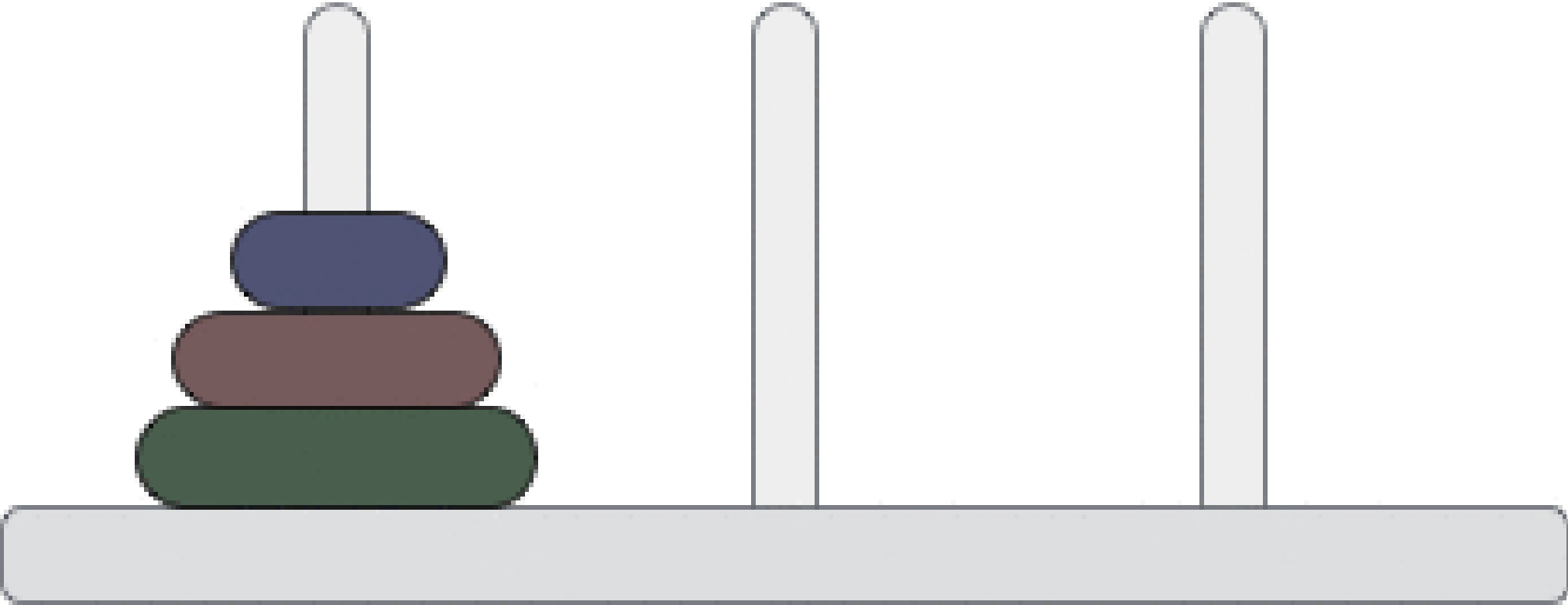
```
int f(int x){  
    if (x == 1 || x == 2) return 1;  
    else return f(x - 1) + f(x - 2);  
}
```



# Example – Tower of Hanoi

- There are three pegs numbered 0, 1 and 2.
- There are  $N$  disks. The number  $N$  is constant while working the puzzle.
- All disks are different in size.
- The disks are initially stacked on peg 0 so that they increase in size from the top to the bottom.
- The goal of the puzzle is to transfer the entire tower from the 0 peg to one of the others pegs.
- One disk at a time can be moved from the top of a stack either to an empty peg or to a peg with a larger disk than itself on the top of its stack.

Step: 0



# Example – Tower of Hanoi

- Define the problem as  $P(n, \text{start}, \text{end}, \text{intermediate})$ , meaning that the disk 1 to  $n$  are moved from the **start** peg to the **end** peg, using the **intermediate** peg for transition
  - No need to care about the positions of disk  $n + 1, n + 2, \dots$  for now
- Base case: when  $n = 1$ , simply move the disk 1 from **start** to **end**
- Recurrence Relation: when  $n > 1$ ,
  - Move disk 1 to  $n - 1$  from **start** to **intermediate**, using **end** for transition
  - Move disk  $n$  from **start** to **end**
  - Move disk 1 to  $n - 1$  from **intermediate** to **end**, using **start** for transition

# Example – Tower of Hanoi

```
int P(int n, int start, int end, int inter){  
    if (n == 1){  
        [Move peg n from start to end]  
    }else{  
        P(n - 1, start, inter, end);  
        [Move peg n from start to end]  
        P(n - 1, inter, end, start);  
    }  
}
```

# Exhaustion

- Sometimes we don't know a fast algorithm to solve the problem
- We can enumerate all possible solutions
- Check whether each of them satisfies the problem's statement
- Find the best one among them

# Example 0

- Given a list of integers and an integer  $M$
- Choose a pair of integers such that the sum of them is equal to  $M$
- E.g.  $A = \{1,2,4,8,16\}$ ,  $M = 10 \rightarrow$  Choose 2 and 8

# Example 0

- The number of possible solutions is small
  - Number of pairs =  $N(N-1)/2$ , where  $N$  is number of integers
- We can try to form all pairs and check whether they satisfy the requirement

For  $i = 1$  to  $N-1$

    For  $j = i+1$  to  $N$

        if  $A[i] + A[j] = M$  then

            return  $\{A[i], A[j]\}$

# Example 1 - Subset

- Given a list of positive integers and an integer  $M$
- Find a **subset** such that the sum of the integers is equal to  $M$
- E.g.  $A = \{2, 9, 15, 16\}$ ,  $M = 27 \rightarrow$  Output  $\{2, 9, 16\}$



# Example 1 - Subset

- Number of possible solutions =  $2^N$
- {}
- {2}, {9}, {15}, {16}
- {2, 9}, {2, 15}, {2, 16}, {9, 15}, {9, 16}, {15, 16}
- {2, 9, 15}, {2, 9, 16}, {2, 15, 16}, {9, 15, 16}
- {2, 9, 15, 16}
- When N is small, we can just check all of them

# Example 1 - Subset

- Method in example 1 can not solve the problem

```
if (n == 1){
    for (int choose_1 = 0; choose_1 <= 1; choose_1++)
        if (a[1] * choose_1 == m) return 1;
} else {
    for (int choose_1 = 0; choose_1 <= 1; choose_1++)
        for (int choose_2 = 0; choose_2 <= 1; choose_2++)
            if (a[1] * choose_1 + a[2] * choose_2 == m) return 1;
} else {
    for (int choose_1 = 0; choose_1 <= 1; choose_1++)
        for (int choose_2 = 0; choose_2 <= 1; choose_2++)
            for (int choose_3 = 0; choose_3 <= 1; choose_3++)
                if (a[1] * choose_1 + a[2] * choose_2 + a[3] * choose_3 == m) return 1;
} else ...
```

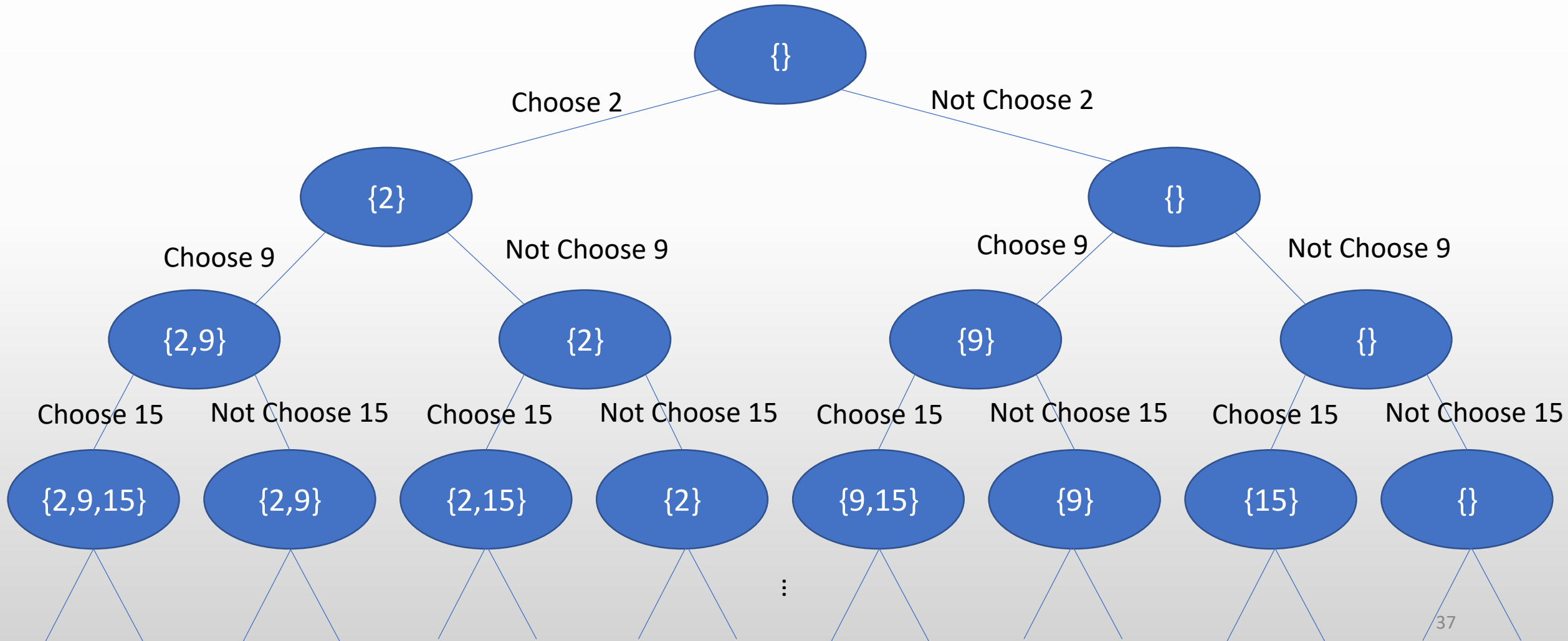
# Example 1 - Subset

- 1 for loop to choose 1 integer, 2 for loops to choose 2 integers, 3 for loops to choose 3 integers, ...
- Each integer requires its own for loop for the 2 branches
- Recursion provides a better way to generate all possible subsets

# Example 1 - Subset

- Procedure  $\text{Exhaustion}(x, \textit{current})$  generates all subsets from  $a[x \dots n]$  together with the set *current*
- Base case: When  $x > n$ , the only result is the set *current*
- Recurrence relation:  $\text{Exhaustion}(x, \textit{current})$  can lead to two cases:
  - Pick  $a[x]$  into the subset  $\rightarrow \text{Exhaustion}(x + 1, \textit{current} + \{a[x]\})$
  - Not pick  $a[x]$  into the subset  $\rightarrow \text{Exhaustion}(x + 1, \textit{current})$

# Example 1 - Subset



# Example 1 – Pseudocode

Procedure exhaustion(integer  $x$ , set current)

  If  $x \leq N$  then

    exhaustion( $x + 1$ , current + { $a[x]$ })

    exhaustion( $x + 1$ , current)

  Else

    if sum of the integers in the set current =  $M$  then

      output the set current

# Example 1 – Code

In reality, we can store the chosen subset outside the procedure (as global variables) instead of passing as parameters

```
void exhaustion(int x, int sum){           // we can store the sum instead
    if (x <= N){
        chosen[x] = true;
        exhaustion(x + 1, sum + a[x]);
        chosen[x] = false;
        exhaustion(x + 1, sum);
    } else {
        if (sum == M)
            [Output the chosen numbers]
    }
}
```

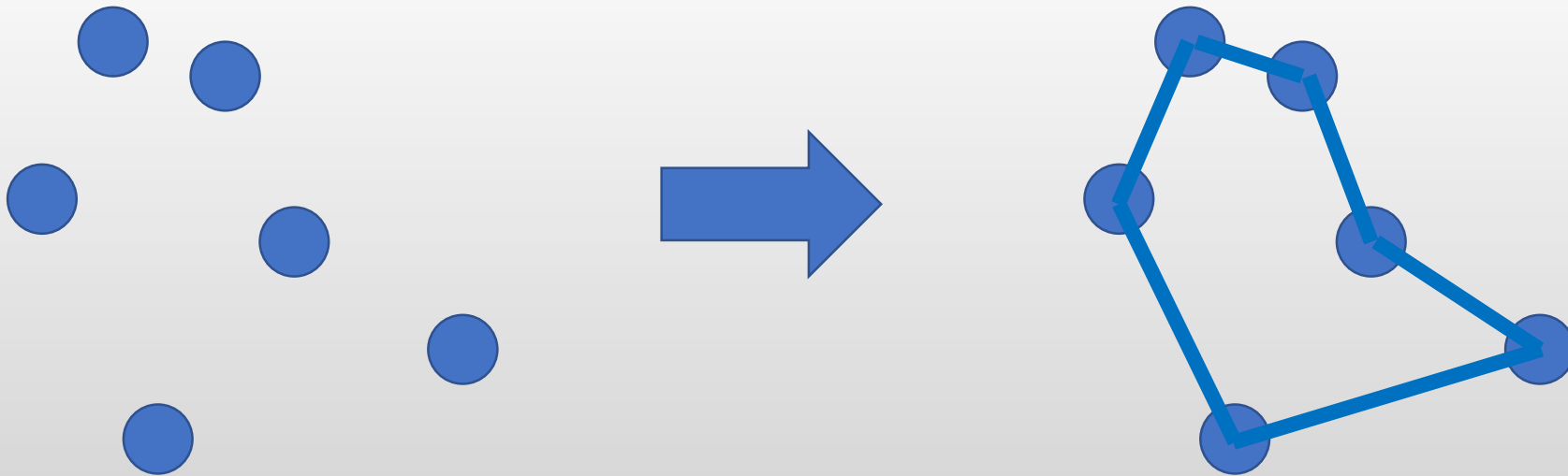
# Example 1

- Time complexity:  $O(2^N)$
- There exists other more efficient solutions
- Exhaustion is enough to solve the problem if  $N$  is small (e.g.  $N \leq 20$ )



# Example 2 - Permutation

- Given the coordinates of N points
- Find out the shortest possible route that visits each point exactly once and returns to the origin point
- Example:



# Example 2 - Permutation

- The famous “Travelling salesman problem”
- Proven to be an NP-hard problem
- Cannot be solved in polynomial time (for now)
  
- The number of possible routes is  $N!$
- Every permutation of the points gives a possible route
- Enumerate all of them to find the shortest one

# Example 2 - Permutation

- Procedure  $\text{Exhaustion}(x, \textit{current})$  generates all permutation starting with the permutation *current* (which has the size of  $x - 1$ )
- Base case: When  $x > n$ , the permutation *current* contains all integers already
- Recurrence relation: In  $\text{Exhaustion}(x, \textit{current})$ , we can pick any point that is not in *current* as the  $x$ -th point of the permutation
  - Calls  $\text{Exhaustion}(x + 1, \textit{current} + [i])$  for all  $i$  that  $i$  is not chosen yet

# Example 2 – Pseudocode 1

Procedure exhaustion(integer x, list current)

  If  $x \leq N$  then

    for i from 1 to N

      if the i-th point is not in current

        exhaustion( $x + 1$ , current + [i])

  Else

    Calculate the length of the route

    Check if it is the shortest one

Time complexity:  $O(N \cdot N!)$

## Example 2 – Code

Similarly, we can store the permutation as global variables

```
void exhaustion(int x){
    if (x <= N){
        for (int i = 1; i <= N; i++){
            if (!chosen[i]){
                chosen[i] = true; per[x] = i; // add i to the list
                exhaustion(x + 1);
                chosen[i] = false; per[x] = 0; // revert the changes
            }
        }
    }else{
        [Calculate and compare the lengths using the array per]
    }
}
```

## Example 2 – Pseudocode 2

The procedure has time complexity of  $O(N)$ , which can be optimized

Let  $P[i]$  denotes the  $i$ -th point

Procedure exhaustion(int  $x$ , double total)

  If  $x \leq N$  then

    for  $i$  from  $x$  to  $N$

      swap  $P[x]$  and  $P[i]$

      exhaustion( $x + 1$ , total + distance between  $P[x]$  and  $P[x-1]$ )

  Else

    total = total + distance between  $P[N]$  and  $P[1]$

    Check if it is the shortest one

Time complexity:  $O(N!)$

# Example 3 – Distinct Permutations

- HKOJ 01031
- Similar to the previous problem
- Given a string of alphabets
- Generate DISTINCT permutations

# Example 3 – Distinct Permutations

- Methods from example 3 may fail
- If the input is ACBA, 2 AABC may be generated
- Because 2 'A's are treated as different elements



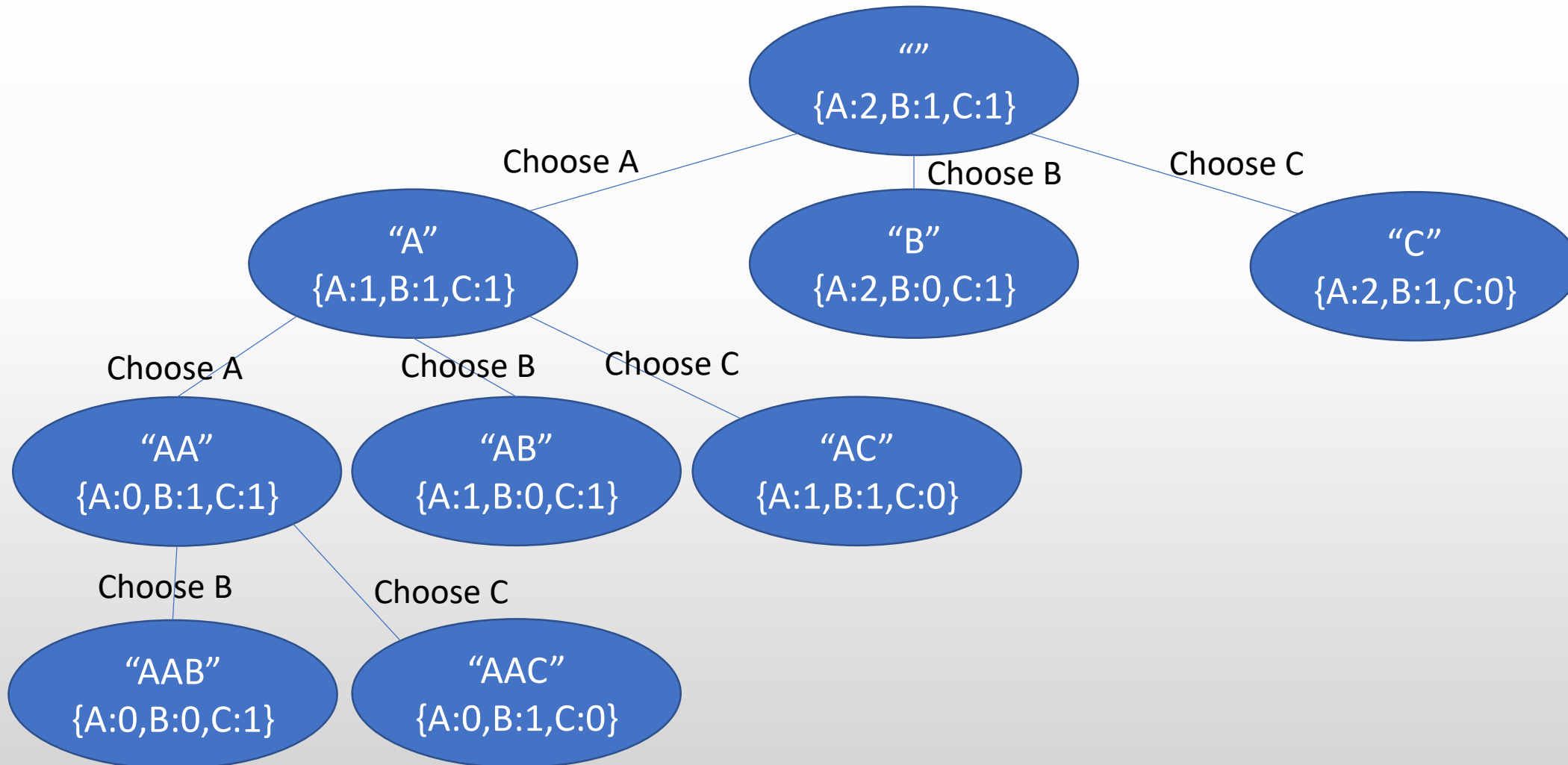
# Example 3 – Distinct Permutations

- Solution 1:
- After generate the permutations
- Delete the repeated ones by sorting (or using set/map in c++)
- Okay for this problem, but more time and memory is wasted

# Example 3 – Distinct Permutations

- Solution 2:
- Count the frequency of each alphabet
- During the recursion, choose an alphabet that still has quotas

# Example 3 – Distinct Permutations



# Example 3 – Pseudocode

Let  $s[1..K]$  be the output string

Procedure exhaustion(int x)

  If  $x \leq K$  then

    for i from 'A' to 'Z'

      if frequency[i] > 0 then

        frequency[i] = frequency[i] - 1

        s[x] = i

        exhaustion(x + 1)

        frequency[i] = frequency[i] + 1

  Else

    print s

# Exhaustion

- Sometimes the way we do exhaustion may affect the efficiency
- A part of the answer may determine the rest
- Smaller part is unknown -> less states to be searched

# Example 4 – Toggle

- A board with  $N \times N$  cells
- Each cell is colored either black or white initially
- In each move, you can either 'toggle' a row or a column so that the color of the cells on the entire row/column changes. If the color of a cell is black, it becomes white after it is toggled, and vice versa.
- Finds a sequence of moves that maximizes the number of white cells.

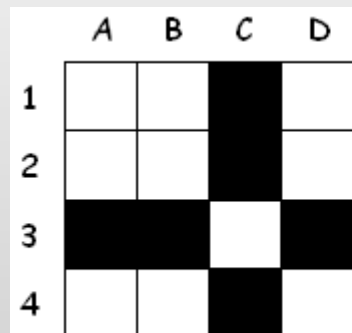


Figure 1

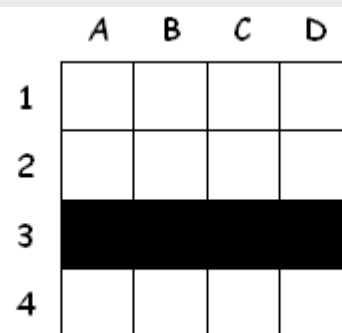


Figure 2

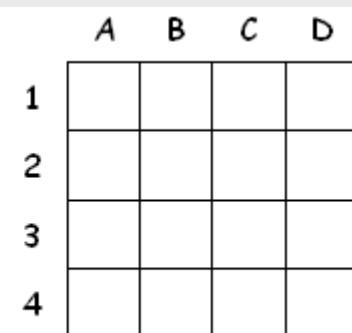


Figure 3

# Example 4 – Toggle

- Observations:
  - The order of row/column to be toggled does not matter
  - Each row/column will be toggled at most once
- Try toggling all combinations of rows and columns
- See which one gives the most number of white cells

# Example 4 – Toggle

- Number of combinations =  $2^{2N} = 4^N$
- Time complexity :  $O(N^2 * 4^N)$
  
- Works when  $N \leq 4$
- Too slow when  $N = 16$



# Example 4 – Toggle

- If the combination of rows to be toggled is fixed, we don't need to try toggling all combinations of columns
- If a column has more white cells than black cells, we should not toggle it
- If a column has more black cells than white cells, we must toggle it

# Example 4 – Toggle

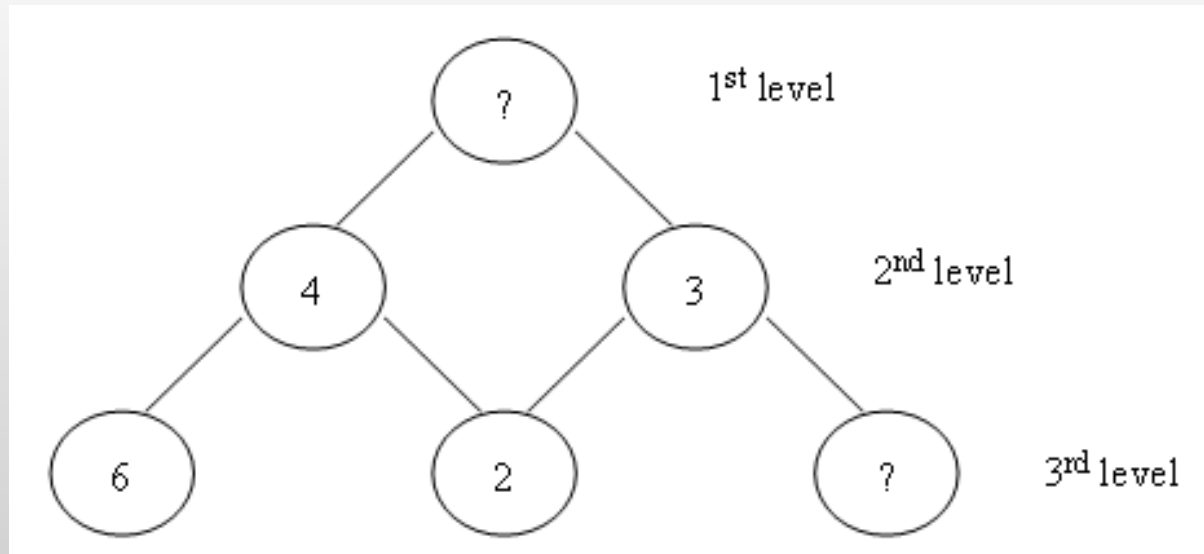
- Try toggling all combinations of rows
- For each column, toggle it only if it has more black cells than white cells
- Check whether it has the most number of white cells
  
- Number of combinations =  $2^N$
- Time complexity :  $O(N^2 * 2^N)$

# Example 4 - Code

```
void exhaustion(int x){
    if (x <= N){
        flip_row[x] = true;
        exhaustion(x + 1);
        flip_row[x] = false;
        exhaustion(x + 1);
    }else{
        [Flip the rows according to flip_row]
        for (int i = 1; i <= n; i++)
            [Flip the column i if there are more black there]
        [Compare with the current answer]
    }
}
```

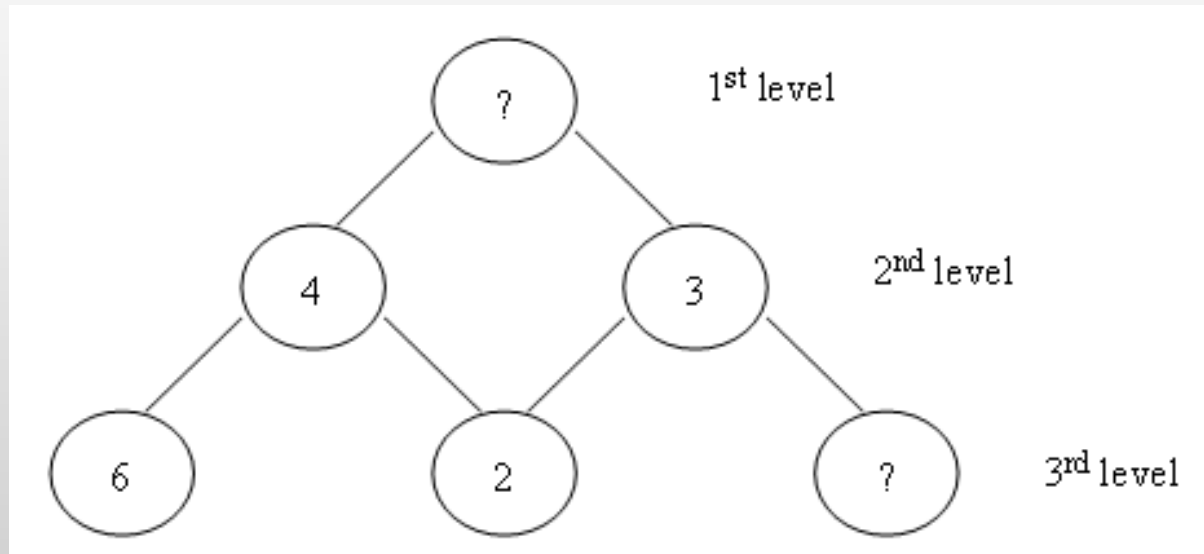
# Example 5 – Magic Triangle

- There is a triangle with N levels ( $N \leq 5$ )
- The i-th level has i nodes
- Each node except those in the Nth level has two children



# Example 5 – Magic Triangle

- Assign 1 to  $N*(N+1)/2$  into the nodes
- The value of some nodes are predetermined
- Output an arrangement such that the value of each node = The absolute difference of the value of its 2 children



# Example 5 – Magic Triangle

- Solution 1 :
- Try all permutations of numbers
- Check whether they fulfill the condition
- Number of permutations =  $(N*(N+1)/2)!$
- Time complexity :  $O((N*(N+1)/2)!)$

# Example 5 – Magic Triangle

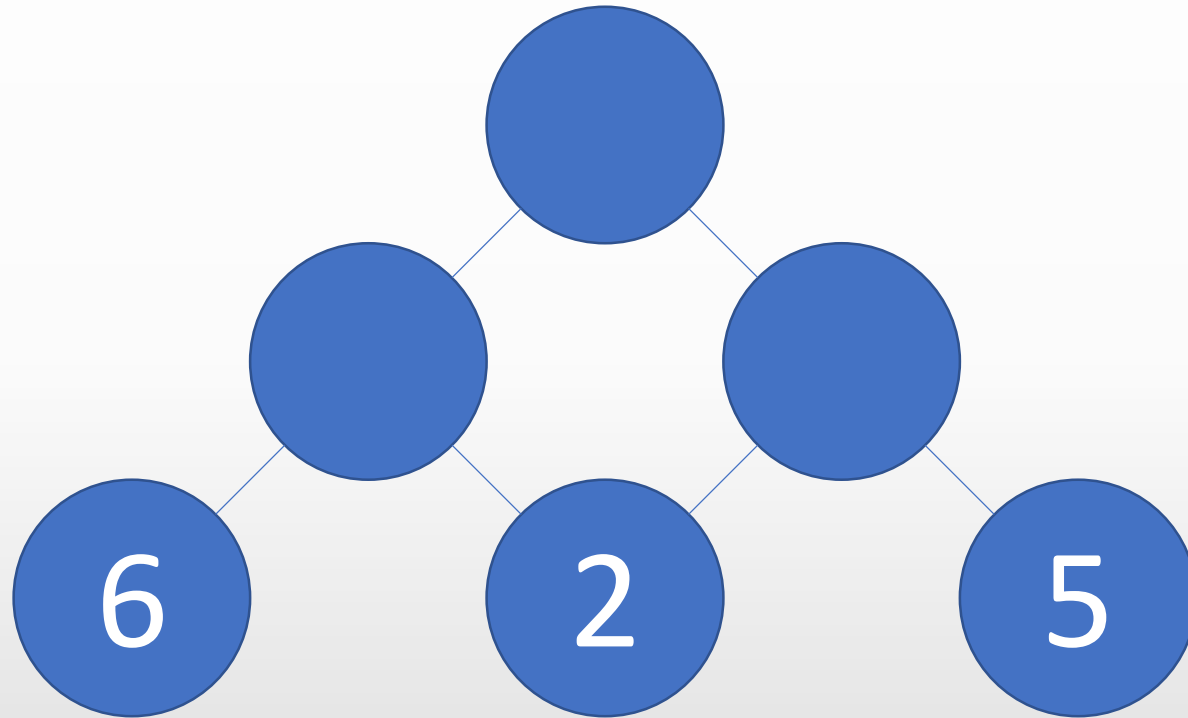
- When  $N = 5$ , number of permutations = 1307674368000
- Constant optimization is not enough
- Even by immediately discarding the permutations that does not match with the predetermined numbers, there are still too many
- A faster solution is needed

# Example 5 – Magic Triangle

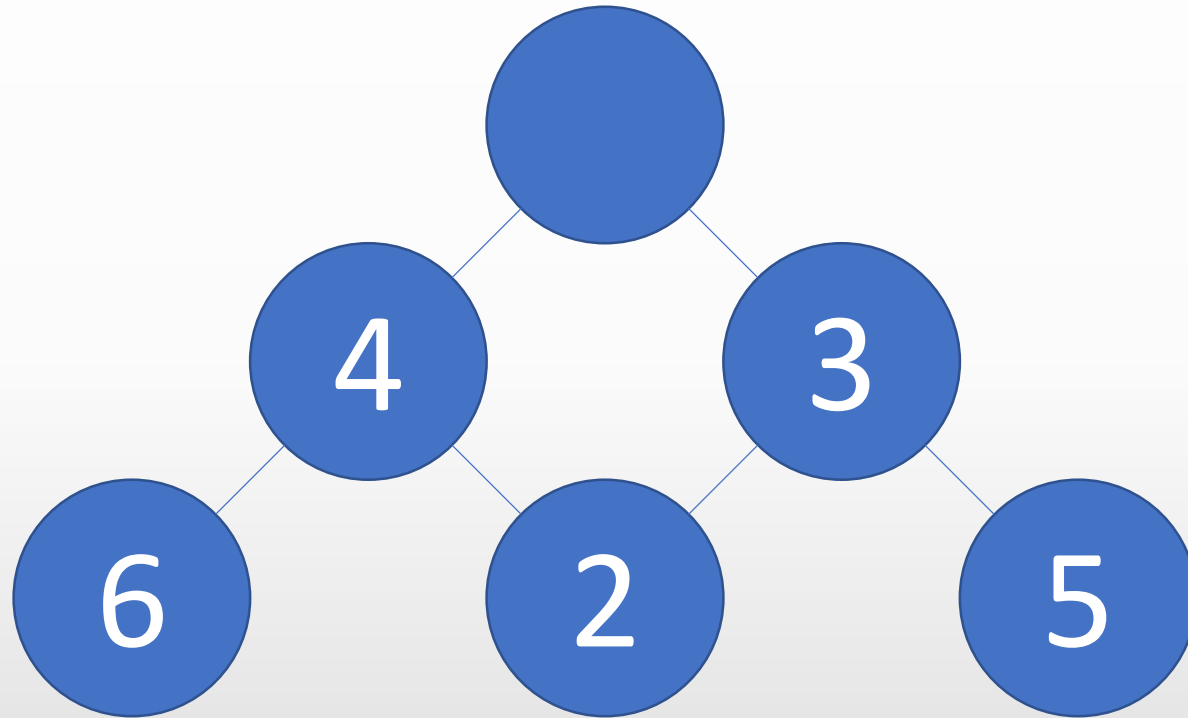
- Consider the values in the Nth level
- If these values are fixed, the values in the (N-1)th level can be calculated
- Use values in the (N-1)th level to calculate those in the (N-2)th level
- ...
- ...
- Use values in the 2<sup>nd</sup> level to calculate the value in the 1<sup>st</sup> level
- The whole arrangement is fixed!



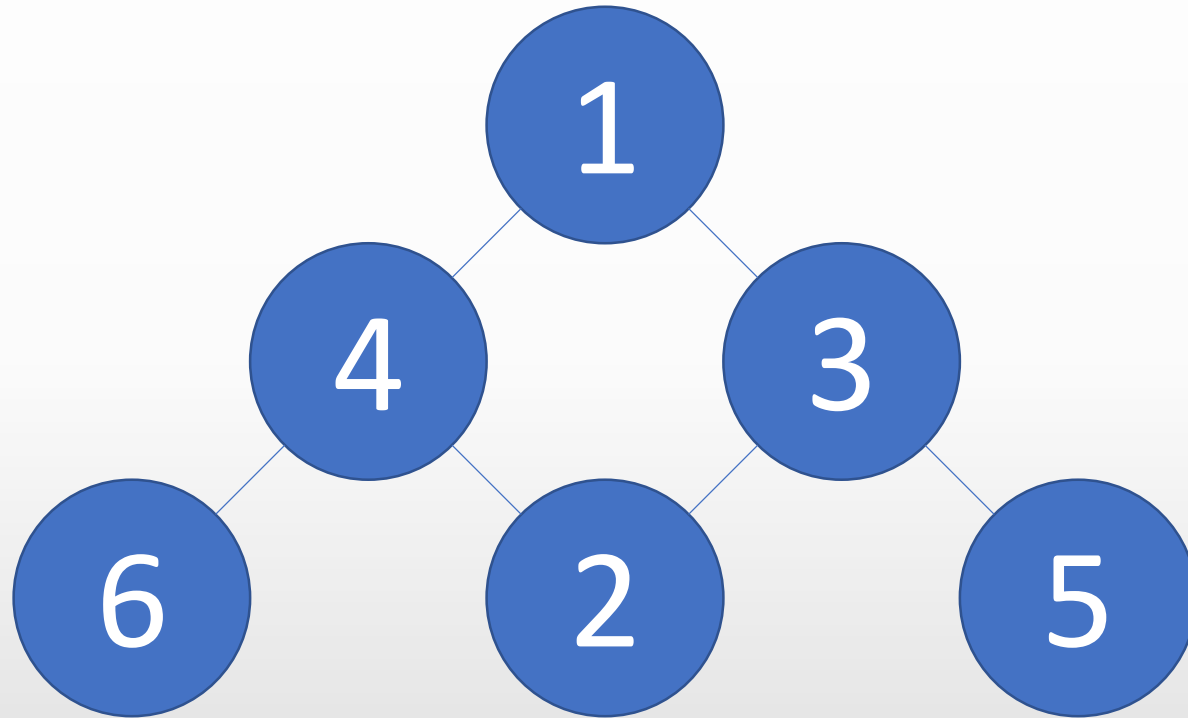
# Example 5 – Magic Triangle



# Example 5 – Magic Triangle



# Example 5 – Magic Triangle



# Example 5 – Magic Triangle

- Solution 2
  - Exhaust all permutations of the Nth level
  - Calculate the values of the whole triangle from the bottom to top
  - Verify if the numbers are distinct
- 
- Number of permutations =  $(N*(N+1)/2)PN$
  - When N is 5, the number of permutations = 360360
  - Fast enough to solve the problem

# Practice problem

- 01046 One-Step Tower of Hanoi
- 01014 Stamps
- 01031 Permutations
- 01035 Combinations
- 20296 Safecracker

# Branch & Bound

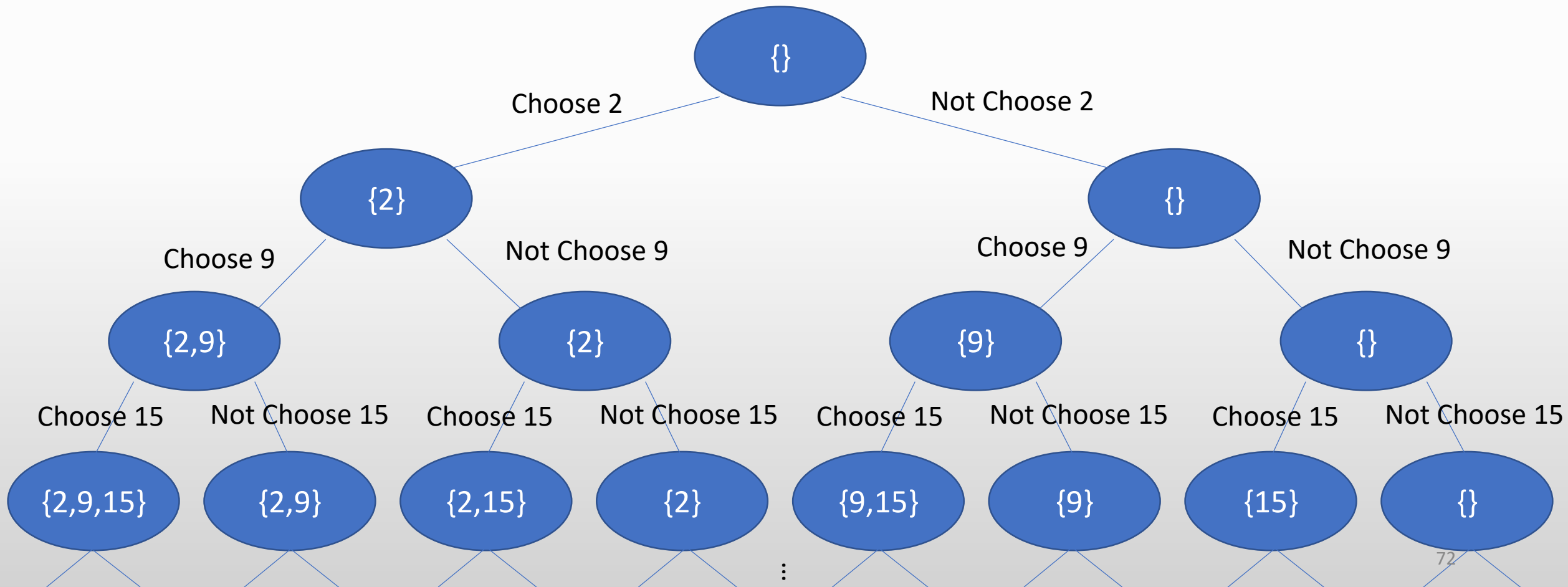
- During the searching, some invalid states may be visited
- When the intermediate state is known to be invalid, stop branching from it
- When the intermediate state gives an unsatisfied answer, stop branching from it

# Example 1 – again

- Let's go back to example 1
- Now  $A = \{2, 9, 10, 3, 7, 2, 5, \dots\}$  and  $M = 16$
- Use the same technique again

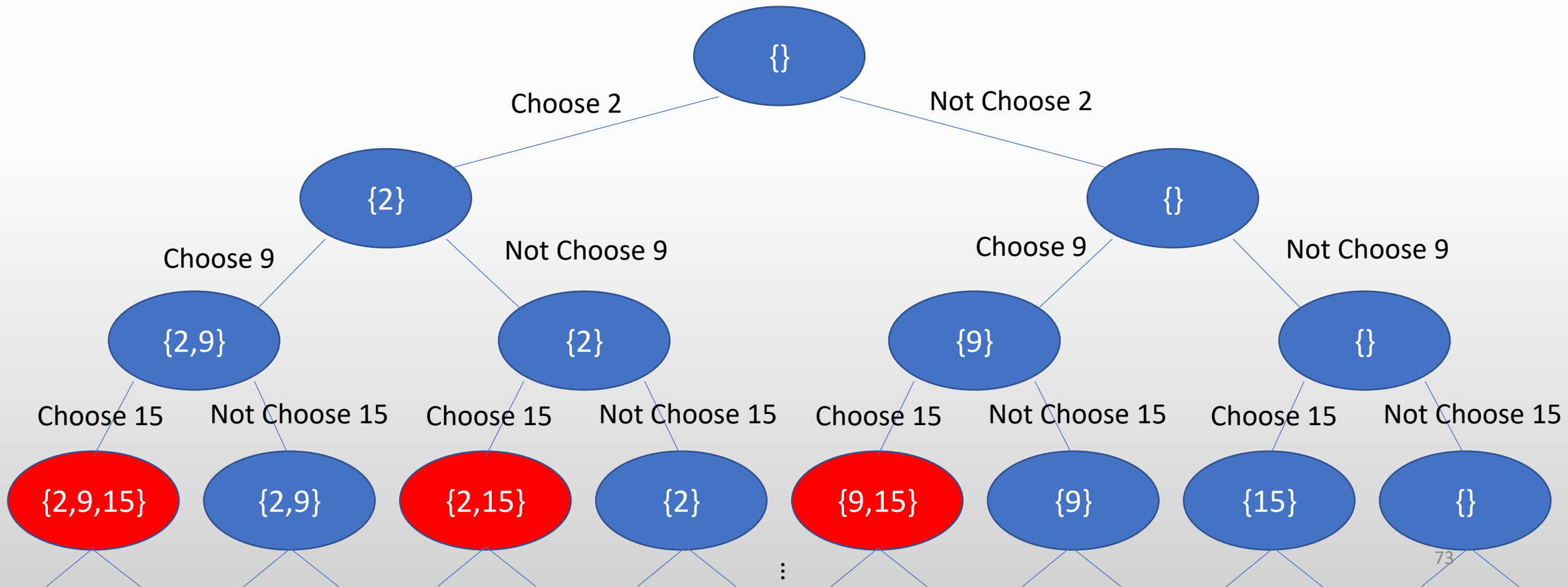
# Example 1 – again

- Some of the states are useless!



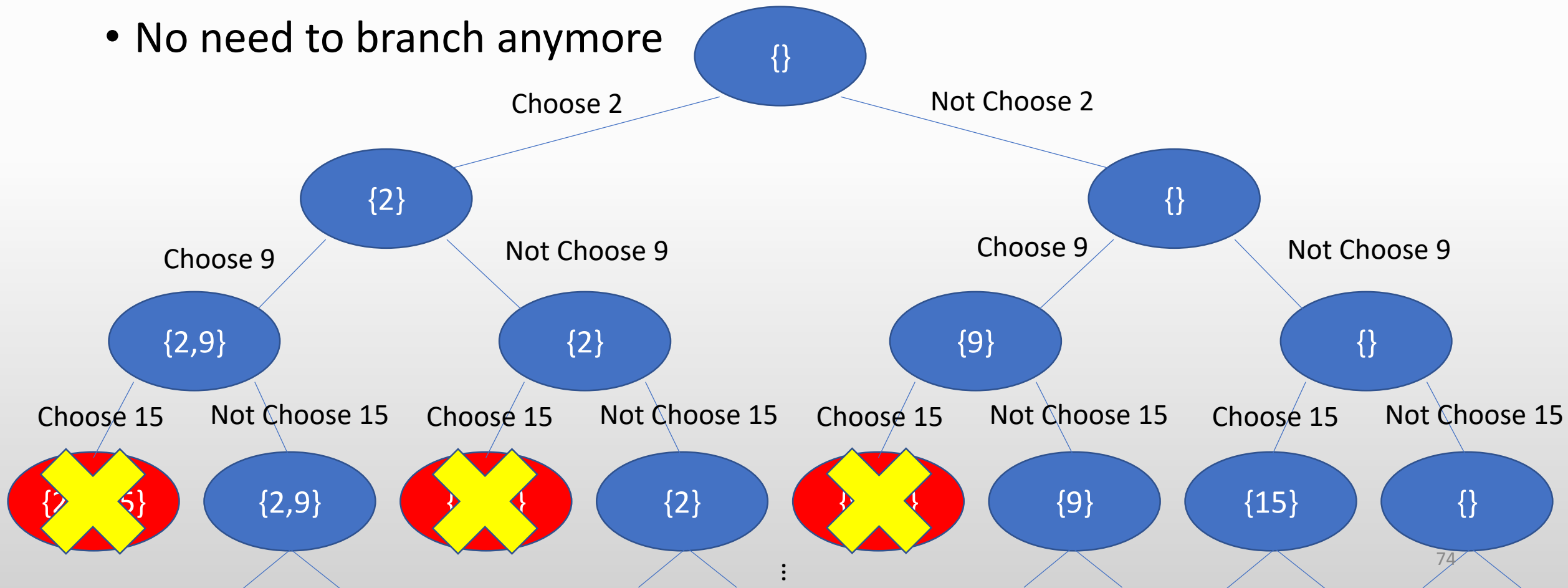


# Example 1 – again



# Example 1 – again

- Sum of the chosen numbers  $> 16$
- No need to branch anymore



# Example 1 – Code 2

```
void exhaustion(int x, int sum){
    if (sum > M) return;
    if (x <= N){
        choose[x] = true;
        exhaustion(x + 1, sum + A[x]);
        choose[x] = false;
        exhaustion(x + 1, sum);
    }else{
        if (sum == M)
            [Output the chosen numbers]
    }
}
```

# Example 1 – Code 3

```
void exhaustion(int x, int sum){
    if (sum > M) return;
    if (solution_found) return;
    if (x <= N){
        choose[x] = true;
        exhaustion(x + 1, sum + A[x]);
        choose[x] = false;
        exhaustion(x + 1, sum);
    }else{
        if (sum == M){
            [Output the chosen numbers]
            solution_found = true;
        }
    }
}
```

# Example 1 – again

- Constant optimization
- The time complexity is not affected much – still  $O(2^N)$
- But much less operations are done
- Search meaningful states only

# Example 6 - Chocolate

- HKOJ 01049
- Given the length of the divided pieces, find the minimum possible length of the original bars

# Example 6 - Chocolate

- For all possible lengths, try to arrange the pieces to see if they can form complete bars
- Possible solutions:
- Each time we pick an unused piece that can fit into the current bar until a complete bar is formed
- Repeat until all piece is used

# Example 6 – Code 1

```
void exhaustion(int x, int current){
    if (x > n){
        solution_found = true;
        return;
    }
    for (int i = 1; i <= n; i++)
        if (!used[i] && current + a[i] <= length){
            used[i] = 1;
            if (current + a[i] == length) exhaustion(x + 1, 0);
            else exhaustion(x + 1, current + a[i]);
            used[i] = 0;
        }
}
```



# Example 6 - Chocolate

- The previous solution is not very efficient
- Different orders of pieces are generated
  - $(\{4, 2\}, \{3, 3\}, \dots)$ ,  $(\{2, 4\}, \{3, 3\}, \dots)$  and  $(\{3, 3\}, \{2, 4\}, \dots)$  are all generated
- But the order doesn't matter actually
  - They are all the same
- Solution 2:
- Each time we put the first unused piece into one of the incomplete bar, or start forming a bar by itself

## Example 6 – Code 2

```
void exhaustion(int x, int m){           // m denotes the current number of
    if (x > n){                         incomplete/complete bars
        [check if all m pieces are complete]
        return;
    }
    for (int i = 1; i <= m; i++){
        if (current[i] + a[x] <= length){
            current[i] += a[x];
            exhaustion(x + 1, m);
            current[i] -= a[x];
        }
        current[m + 1] = a[x];
        exhaustion(x + 1, m + 1);
        current[m + 1] = 0;
    }
}
```

# Example 6 - Chocolate

- Still not fast enough?
- If the piece can form a complete bar with an incomplete bar, no capacity will be wasted
- Must be the best arrangement for that piece
- No need to consider putting it into other incomplete bars

# Example 6 – Code 3

```
void exhaustion(int x, int m){
    if (x > n){
        [check if all m pieces are complete]
        return;
    }
    for (int i = 1; i <= m; i++){
        if (current[i] + a[x] == length){
            current[i] += a[x]; exhaustion(x + 1, m); current[i] -= a[x];
            return;
        }
    }
    for (int i = 1; i <= m; i++){
        if (current[i] + a[x] <= length){
            current[i] += a[x]; exhaustion(x + 1, m); current[i] -= a[x];
        }
    }
    current[m + 1] = a[x]; exhaustion(x + 1, m + 1); current[m + 1] = 0;
}
```

# Example 6 - Chocolate

- Still not fast enough???????
- Sorting the pieces in descending order may be useful, as the smaller pieces will have less choices at last
- Stop branching when too many bars are formed
- ...
- Be creative :O)

# Practice problems

- HKOJ 01049 Chocolate
- HKOJ 01050 Bin Packing
- HKOJ 20750 8 Queens Chess Problem
- UVA 307 Sticks
- UVA 524 Prime Ring Problem

# Divide and Conquer

1. Divide the problem into smaller and independent sub-problems that are same as the original one
2. If the sub-problems are easy to solve, solve them directly; otherwise solve the sub-problems recursively
3. Combine the solutions of the sub-problems to form the solution for the original problem

# Example 1 – Big Mod

- Given B, P and M, find  $B^P \% M$

- Naïve solution: multiply B for P times

```
ans = 1;
```

```
for (int i = 1; i <= P; i++) ans = ans * B % M;
```

or a recursive one using  $B^P \% M = B * B^{P-1} \% M$

```
int f(int B, int P, int M){  
    if (P == 0) return 1 % M;  
    else return f(B, P - 1, M) * B % M;  
}
```

- Time complexity:  $O(P)$



# Example 1 – Big Mod

- Given  $B$ ,  $P$  and  $M$ , find  $B^P \% M$
- Notice that when  $P$  is even, a better way would be

$$B^P \% M = B^{P/2} \% M * B^{P/2} \% M$$

- When  $P$  is odd, we can still have  $B^P \% M = B * B^{P-1} \% M$ , or

$$B^P \% M = B^{P/2} \% M * B^{P/2} \% M * B \% M$$

# Example 1 – Big Mod

- Base case: when  $P = 0$ ,  $f(B, P, M) = 1 \% M$
- Recurrence relation:
  - If  $P$  is even,  $f(B, P, M) = f(B, P / 2, M) * f(B, P / 2, M) \% M$
  - If  $P$  is odd,  $f(B, P, M) = f(B, P / 2, M) * f(B, P / 2, M) \% M * B \% M$
- Two calls  $f(B, P / 2, M)$  are the same so we can just call once

```
int f(int B, int P, int M){
    if (P == 0) return 1 % M;
    int tmp = f(B, P / 2, M);
    if (P % 2 == 0) return tmp * tmp % M;
    else return tmp * tmp % M * B % M;
}
```

# Example 1 – Big Mod

- Each time we reduce  $P$  to  $P / 2$  until  $P$  become 0
- e.g. When  $P = 30$ , we only call  $f(30)$ ,  $f(15)$ ,  $f(7)$ ,  $f(3)$ ,  $f(1)$ , which is around  $\log(P)$  times
- So the time complexity is  $O(\log(P))$
- By dividing the problem properly, the time complexity may be reduced

# Master Theorem

- In fact there are better ways to analyze the time complexity of the divide and conquer algorithms
- Master Theorem: For an algorithm run in time complexity  $T(n)$
- If  $T(n)$  can be written as a recurrence formula in form of  $T(n) \leq aT\left(\frac{n}{b}\right) + O(n^d)$ , where  $a \geq 1, b > 1$ ,
- We have  $T(n) = \begin{cases} O(n^d) & \text{if } d > \log_b a \\ O(n^d \log n) & \text{if } d = \log_b a \\ O(n^{\log_b a}) & \text{if } d < \log_b a \end{cases}$
- $aT\left(\frac{n}{b}\right)$  denotes the time complexity for the divide part: we divide the original problem to  $a$  sub-problems, the parameter of each sub-problem reduces from  $n$  to  $\frac{n}{b}$
- $O(n^d)$  denotes the time complexity for the combine part: we spend  $O(n^d)$  time to combine the solutions to the one for the original problem

# Example 1 – Big Mod

```
int f(int B, int P, int M){
    if (P == 0) return 1 % M;
    int tmp = f(B, P / 2, M);
    if (P % 2 == 0) return tmp * tmp % M;
    else return tmp * tmp % M * B % M;
}
```

- We divide the problem  $f(P)$  into **ONE** sub-problem  $f(P/2)$ , and we takes  $O(1)$  (or  $O(n^0)$ ) time in combining the answer
- Now  $a = 1, b = 2, d = 0, T(n) \leq T\left(\frac{n}{2}\right) + O(1)$
- Since  $\log_b a = 0 = d$ , the time complexity is  $O(n^d \log n) = O(\log n)$

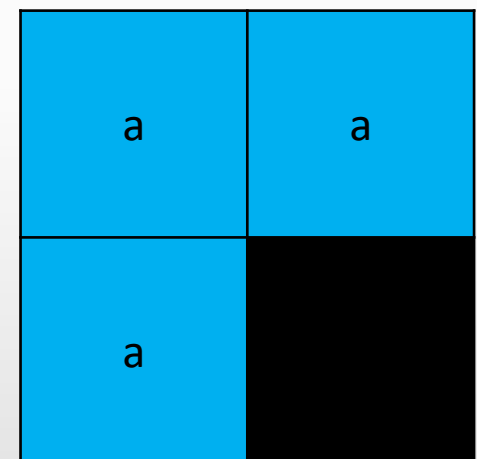
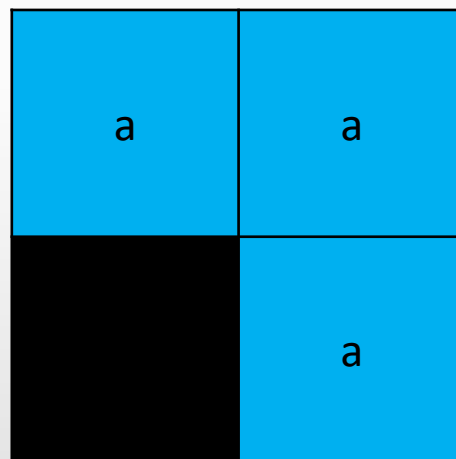
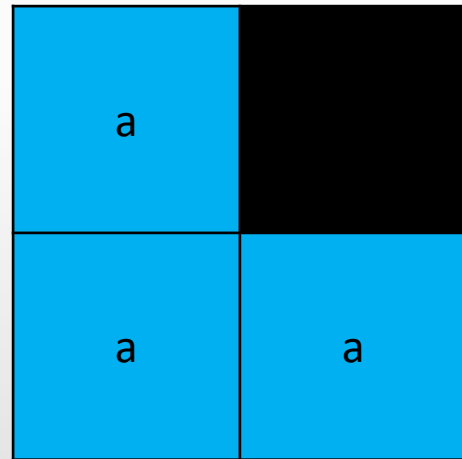
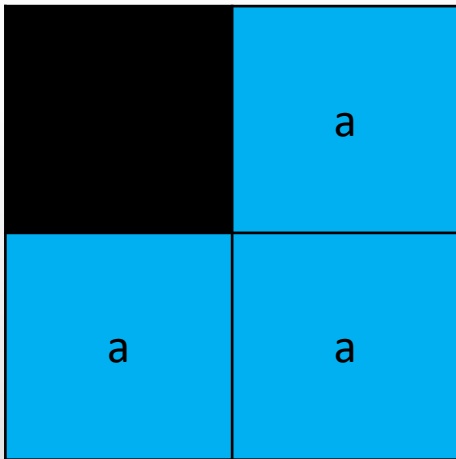
## Example 2 – L pieces

- Given a  $N * N$  grid with a empty cell. Output a way to use L-pieces to cover the whole grid where  $N$  is a power of 2
- e.g.  $N = 4$ , empty cell is (3, 4):

c	c	d	d
c	b	b	d
e	b	a	
e	e	a	a

# Example 2 – L pieces

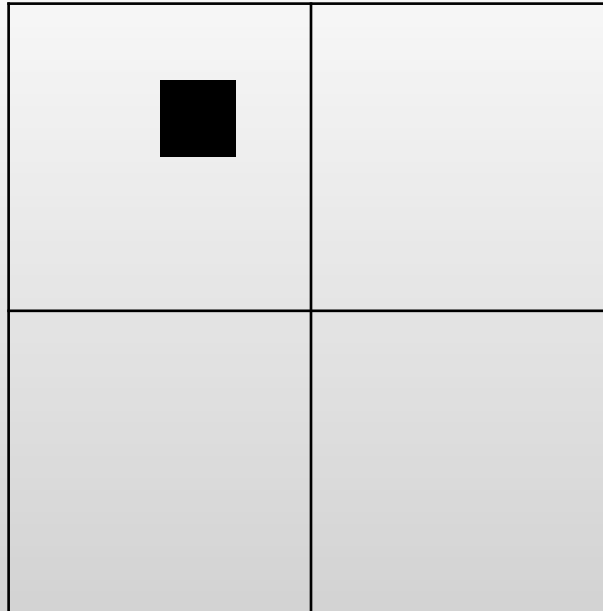
- Base Case: When  $N = 2$ , the solution is trivial
- Just put a L-piece on the remaining three cell



- How about the larger ones?

## Example 2 – L pieces

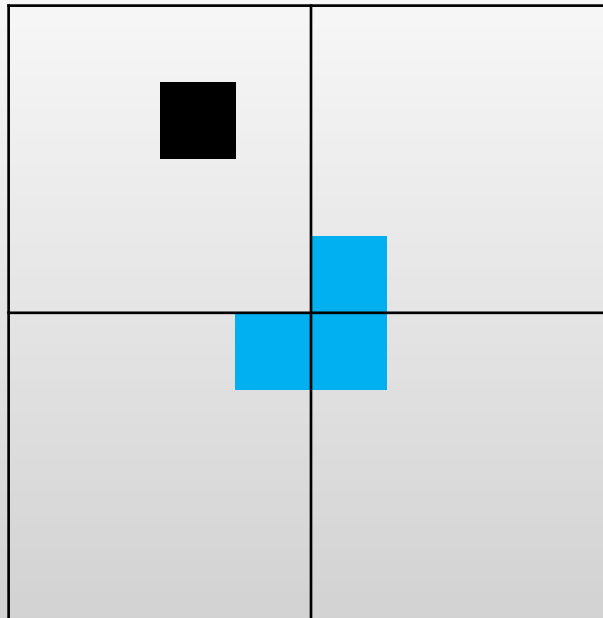
- We can divide the  $N \times N$  grid into 4  $(N/2) \times (N/2)$  smaller grids
- The empty cell must be in one of the smaller grids
- The problem is divided, but they are not the same problem
- The other three smaller grids does not have empty cell





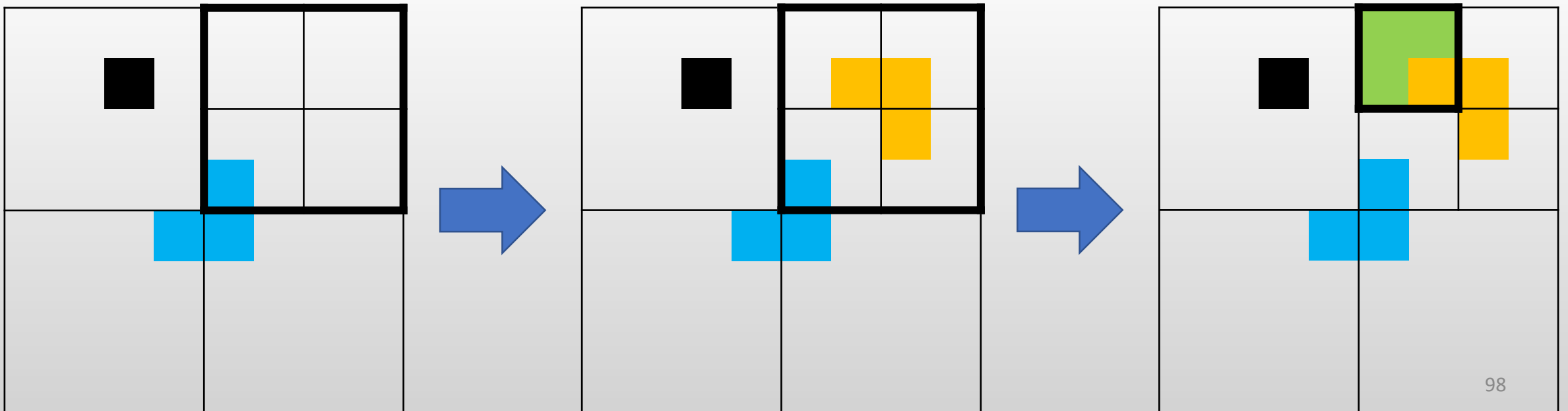
## Example 2 – L pieces

- Add one L-piece at the center of the grid that covers the remaining three grids
- Now every smaller grid has one empty cell and becomes a sub-problem that is same as the original one



## Example 2 – L pieces

- The same method can be applied on the smaller grid as well
- Solve them recursively until it is divided into a 2 x 2 grid



# Example 2 – L pieces

- Solution:

- Divide the grid to 4 smaller grids: upper-left, upper-right, lower-right, lower-left
- Put a L-piece in a suitable orientation in the center of the grid such that each smaller grid has exactly 1 empty cell
- Repeat the above steps recursively on the smaller grids

- Time complexity:

- $T(n) = 4T\left(\frac{n}{2}\right) + O(1), a = 4, b = 2, d = 0, \log_b a = 2 > d$
- $T(n) = O(n^{\log_2 4}) = O(n^2)$

# Example 3 – Number of Inversions

- Given a list of distinct integers  $a[1..n]$
- Find the number of pairs  $(a[i], a[j])$  such that  $i < j$  and  $a[i] > a[j]$
- e.g.  $n = 6$ ,  $a = [1, 2, 6, 3, 5, 4]$
- Answer = 4, the pairs are  $(6, 3)$ ,  $(6, 5)$ ,  $(6, 4)$  and  $(5, 4)$

## Example 3 – Number of Inversions

- We can split the array  $a$  into two halves  $a_l$  and  $a_r$
- $a = [1, 2, 6, 3, 5, 4] \rightarrow a_l = [1, 2, 6]$  and  $a_r = [3, 5, 4]$
- # of inversions of  $a = (\text{\# of inversions of } a_l) + (\text{\# of inversions of } a_r) + (\text{\# of inversions between } a_l \text{ and } a_r)$
- To calculate the # of inversions between  $a_l$  and  $a_r$ , for each integer  $a_r[i]$ , we need to count the number of integers in  $a_l$  that is greater than  $a_r[i]$

## Example 3 – Number of Inversions

- Once the # of inversions of  $a_l$  is obtained, we do not care about the order in  $a_l$  anymore, and same for  $a_r$
- Sorting  $a_l$  and  $a_r$  will not affect the answer now
- To count the number of inversions between  $a_l$  and  $a_r$ , we merge the sorted arrays  $a_l$  and  $a_r$  into one sorted array (like Merge Sort)
- Whenever an element from  $a_r$  is pushed into the array, count the number of elements from  $a_l$  that are not yet pushed into the array
  - These are the inversion pairs between  $a_l$  and  $a_r$

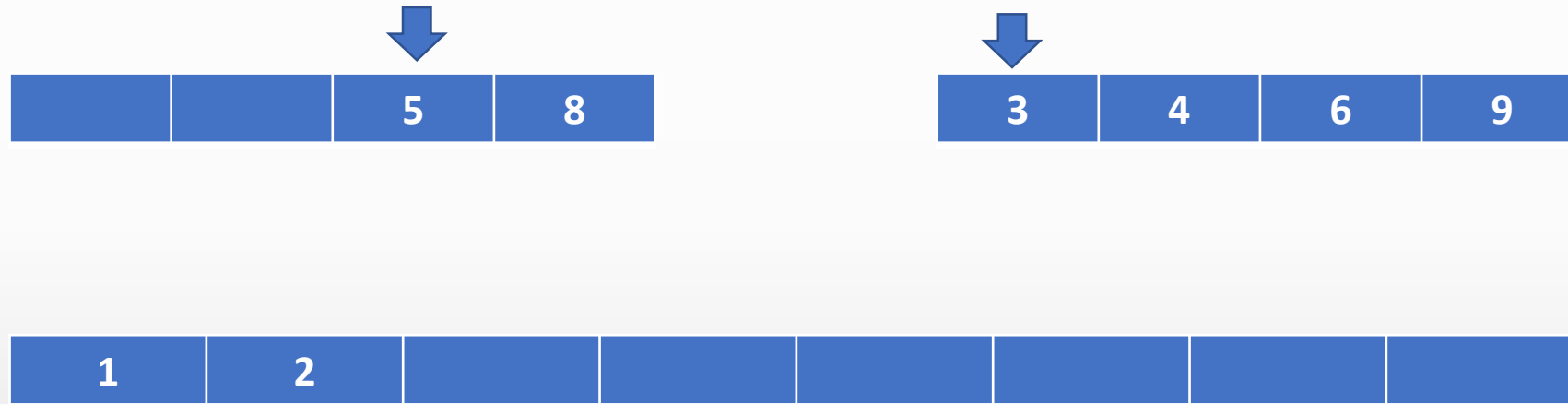
# Example 3 – Number of Inversions

- Answer = 0



# Example 3 – Number of Inversions

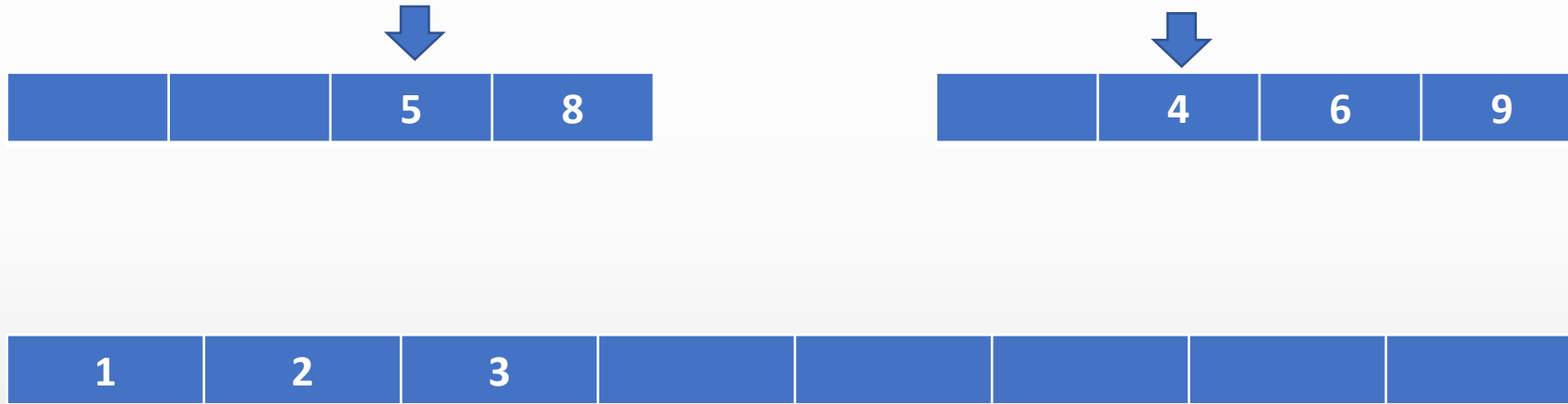
- Answer = 0





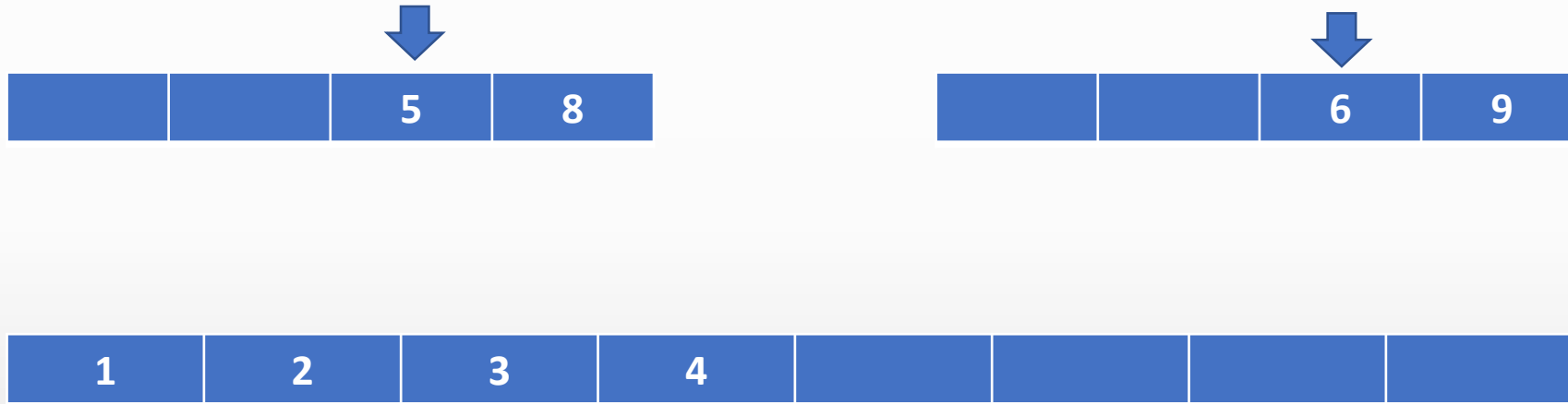
# Example 3 – Number of Inversions

- Answer = 2, as there are 2 remaining integers in  $a_l$



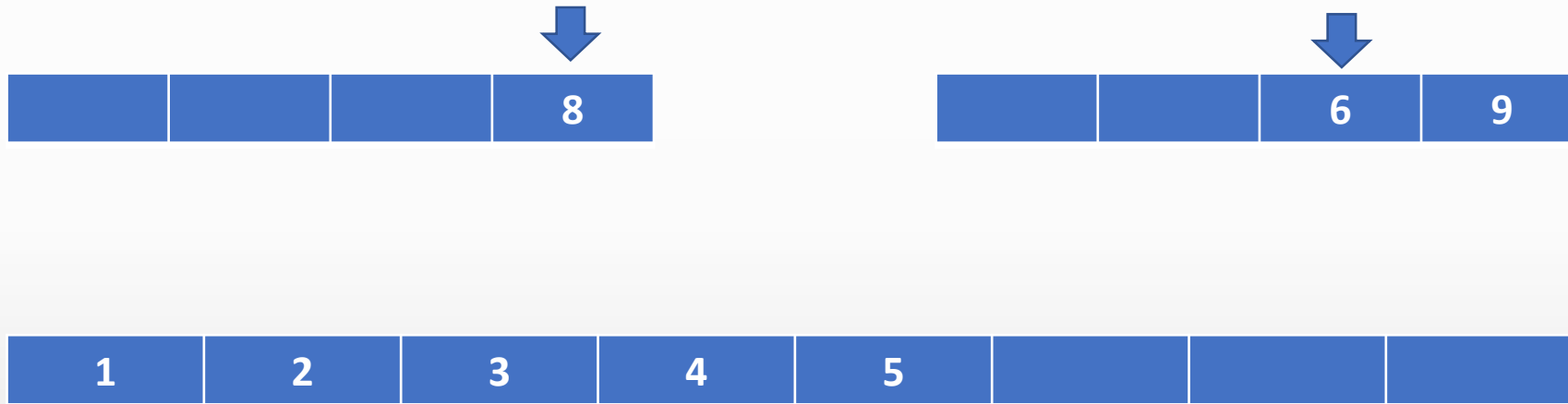
# Example 3 – Number of Inversions

- Answer = 4, as there are 2 remaining integers in  $a_l$



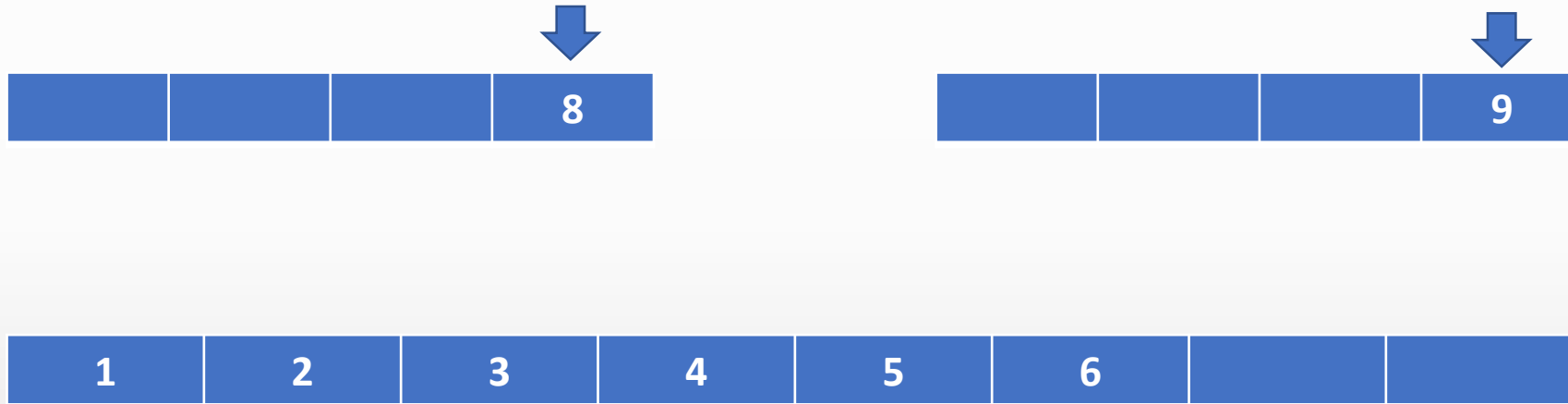
# Example 3 – Number of Inversions

- Answer = 4



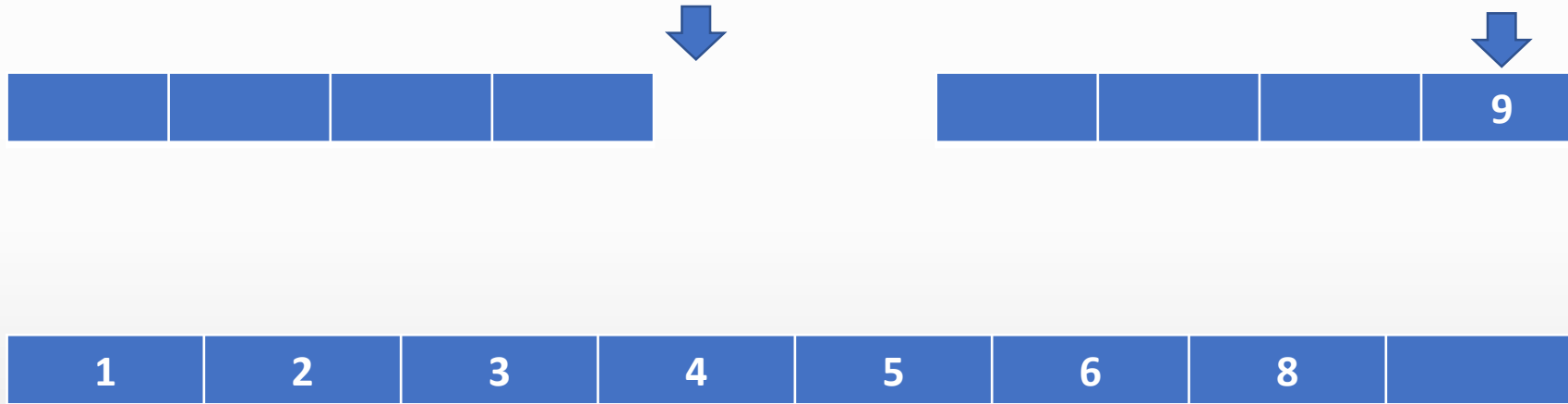
# Example 3 – Number of Inversions

- Answer = 5, as there are 1 remaining integer in  $a_l$



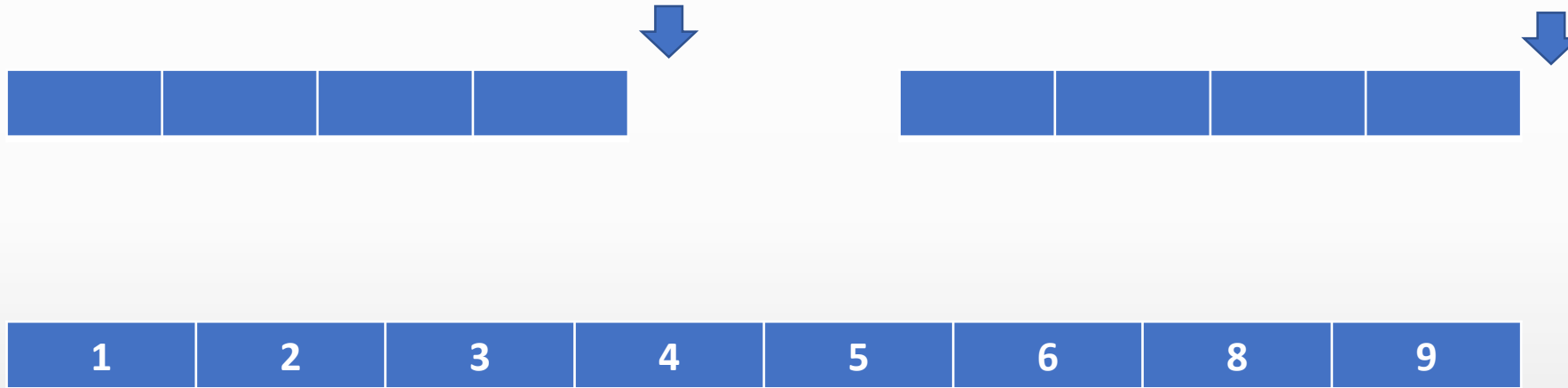
# Example 3 – Number of Inversions

- Answer = 5, as there are 1 remaining integer in  $a_l$



# Example 3 – Number of Inversions

- Answer = 5, as there are 0 remaining integers in  $a_l$



# Example 3 – Number of Inversions

- Solution:

- Split the array into two halves  $a_l$  and  $a_r$
- Compute the number of inversions of the two halves and sort them recursively
- Merge the two halves and compute the number of inversions between two halves
- Return the sum of the three parts as the answer

- Time complexity:

- $T(n) = 2T\left(\frac{n}{2}\right) + O(n), a = 2, b = 2, d = 1, \log_b a = 1 = d$
- $T(n) = O(n^d \log n) = O(n \log n)$

# Example 3 – Number of Inversions

```
function compute_and_sort(integer x, y){           // processing the subarray a[x .. y]
    if x = y then return 0
    ans1 = compute_and_sort(x, (x + y) / 2)
    ans2 = compute_and_sort((x + y) / 2 + 1, y)
    ptr_l = x, ptr_r = (x + y) / 2 + 1
    ans3 = 0
    for i from x to y
        if ptr_l > (x + y) / 2 then                b[i] = a[ptr_r++]
        else if ptr_r > y then                      b[i] = a[ptr_l++]
        else if a[ptr_l] < a[ptr_r] then           b[i] = a[ptr_l++]
        else
            b[i] = a[ptr_r++]
            ans3 += (x + y) / 2 + 1 - ptr_l
    a[x .. y] = b[x .. y]
    return ans1 + ans2 + ans3
}
```



# Summary

- Recursion is useful and common
- Don't hesitate to use exhaustion to get some scores or find patterns
- Don't ignore it and spend all the time on other subtasks
- Divide and Conquer is essential for many data structures, and is often related to other algorithms
- Practice more :O)

# Practice Problems

- 01003 L-pieces
- 01046 One-Step Tower of Hanoi
- 01047 Partially-Solved Tower of Hanoi
- M0713 Cream Soda
- J173 Fibonacci Word
- S134 Unfair Santa Claus
- S163 Arithmetic Sequence