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SCHOOL OF INFORMATION AND COMMUNICATION TECHNOLOGY



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# DATA STRUCTURES AND ALGORITHMS

## Algorithmic paradigms

# CONTENT

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- Recursion
- Recursion with memoization
- Backtracking
- Branch and Bound
- Greedy
- Divide and conquer
- Dynamic Programming

# Recursion

- A code in which the name of a function (procedure) appears in the function itself
- Base case
  - Input parameter of the function is small enough so that the result of the function can be computed directly without calling the function recursively

Example:  $S = 1 + 2 + \dots + n$

```
int sum(int n) {  
    if (n <= 1) return 1;  
    int s = sum(n-1);  
    return n + s;  
}
```

# Recursion

- Example

$$C(k, n) = \frac{n!}{k!(n-k)!}$$

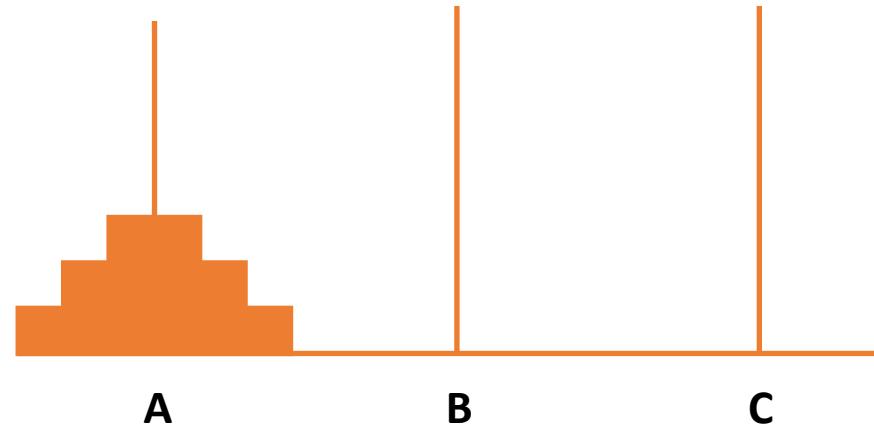
- Recurrent relation
  - $C(k, n) = C(k-1, n-1) + C(k, n-1)$
- Base case:

$$C(0, n) = C(n, n) = 1$$

```
int C(int k, int n) {  
    if (k == 0 || k == n)  
        return 1;  
    int C1 = C(k-1,n-1);  
    int C2 = C(k,n-1);  
    return C1 + C2;  
}
```

# Hanoi tower

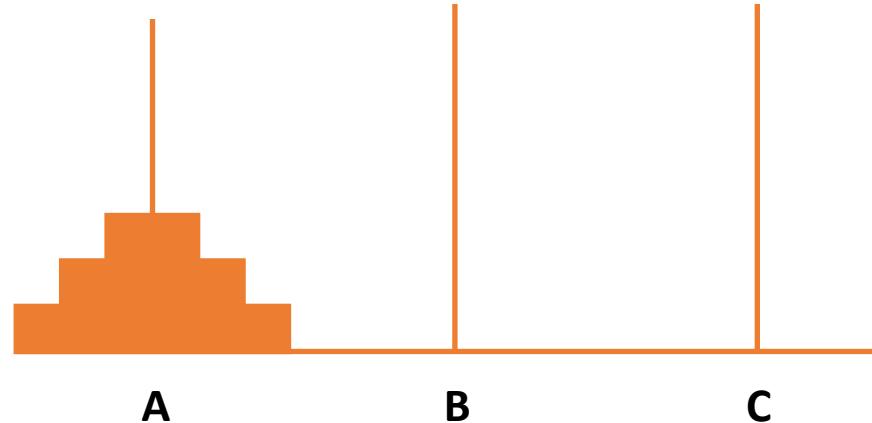
- There are  $n$  disks with different sizes and 3 rods A, B, C
- Initialization:  $n$  disks are in the rod A such that larger disks are below smaller ones



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# Hanoi tower

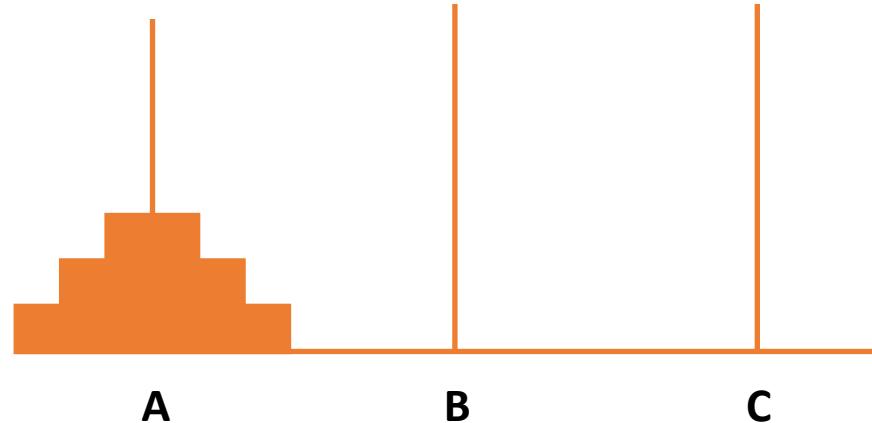
- There are  $n$  disks with different sizes and 3 rods A, B, C
- Initialization:  $n$  disks are in the rod A such that larger disks are below smaller ones
- Goal: Move  $n$  disks from A to B such that
  - Only one disk can be moved at a time
  - Each move consists of taking the upper disk from one of the stacks and placing it on top of another stack or on an empty rod
  - No larger disk may be placed on top of a smaller disk



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# Hanoi tower

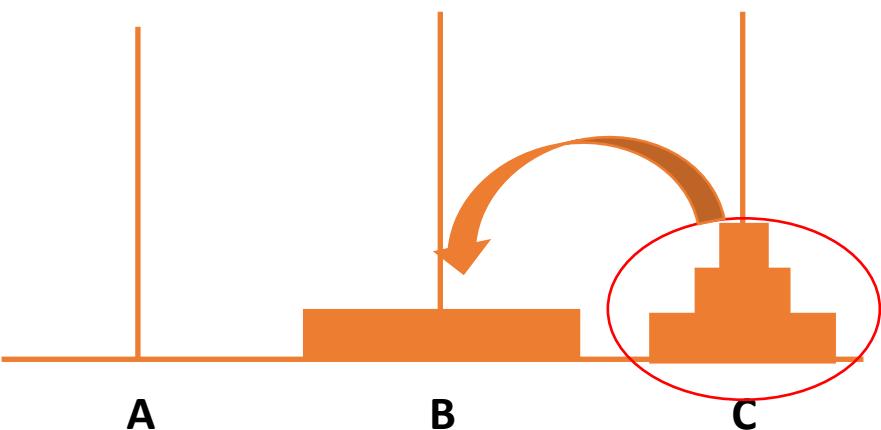
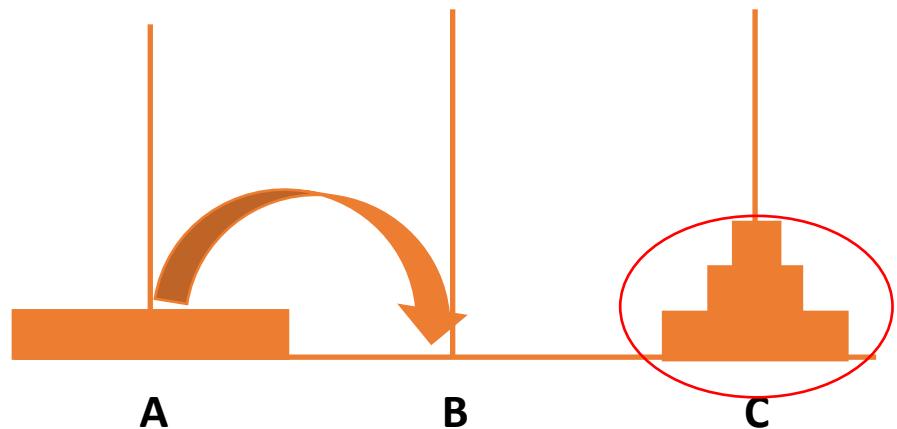
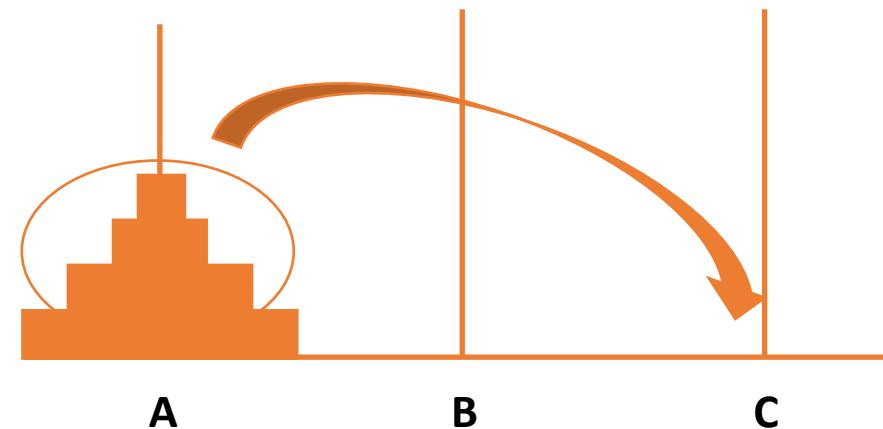
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- Initialization:  $n$  disks are in the rod A such that larger disks are below smaller ones
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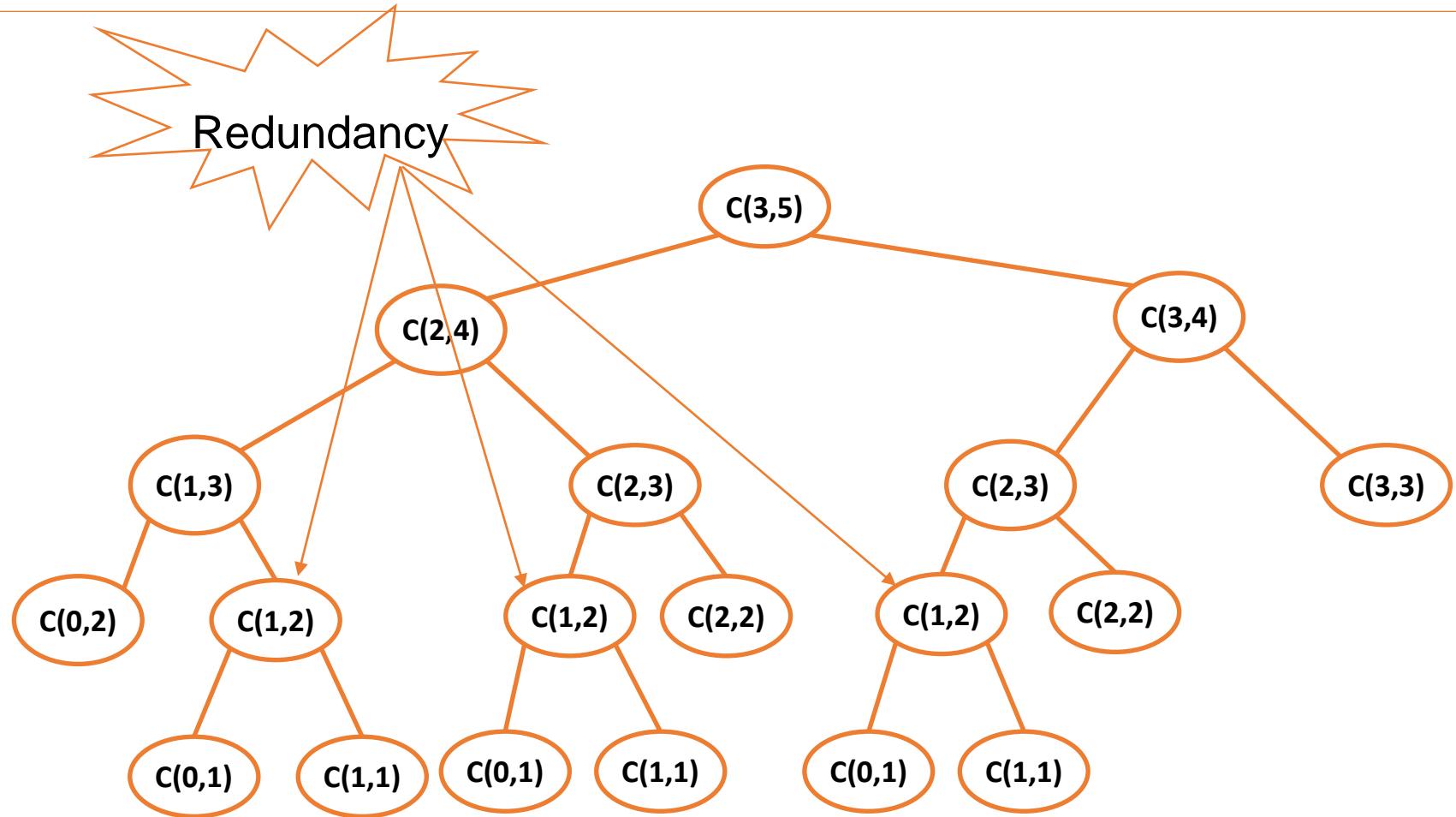
Lời giải  
B1: A → B  
B2: A → C  
B3: B → C  
B4: A → B  
B5: C → A  
B6: C → B  
B7: A → B

# Hanoi tower



```
move(n, A, B, C) {  
    if(n == 1) {  
        print("Move 1 disk from A to B")  
    }else{  
        move(n-1, A, C, B);  
        move(1, A, B, C);  
        move(n-1, C, B, A);  
    }  
}
```

# Recursion with memoization



# Recursion with memoization

- Use memory (table) to store results of subproblems
- Map each subproblem with a specified parameter to an element of the table
- Initialize the table with special values indicating that no results of any subproblem are available

```
int m[MAX][MAX];  
  
int C(int k, int n) {  
    if (k == 0 || k == n)  
        m[k][n] = 1;  
    else {  
        if(m[k][n] < 0){  
            m[k][n] = C(k-1,n-1) +  
                      C(k,n-1);  
        }  
    }  
    return m[k][n];  
}  
  
int main() {  
    for(int i = 0; i < MAX; i++)  
        for(int j = 0; j < MAX; j++)  
            m[i][j] = -1;  
    printf("%d\n",C(16,32));  
}
```

# Backtracking

---

- Application: solve combinatorial enumeration/optimization problems
- $A = \{(x_1, x_2, \dots, x_n) \mid x_i \in A_i, \forall i = 1, \dots, n\}$
- Generate all configurations  $x \in A$  satisfying some property  $P$
- Recursive procedure Try( $k$ ):
  - Try all values  $v$  that can be assigned to  $x_k$  without violating  $P$
  - For each feasible value  $v$ :
    - Assign  $v$  to  $x_k$
    - If  $k < n$ : call recursively Try( $k+1$ ) to try values for  $x_{k+1}$
    - If  $k = n$ : get a solution

# Backtracking

---

```
Try( $k$ ){  
    foreach  $v$  of  $A_k$   
        if check( $v, k$ ) /* check feasibility of  $v$  */  
        {  
             $x_k = v$ ;  
            [Update a data structure D]  
            if( $k = n$ ) solution();  
            else Try( $k+1$ );  
            [Recover D when backtracking]  
        }  
    }  
Main(){  
    Try(1);  
}
```

# Backtracking

- Example: generate all binary sequences of length n
- Modelling:
  - Array  $x[n]$  where  $x[i] \in \{0,1\}$  is the  $i^{\text{th}}$  bit of the sequence  
( $i = 0, \dots, n-1$ )

```
void solution(){
    for(int k = 0; k < n; k++)
        printf("%d",x[k]);
    printf("\n");
}

int TRY(int k) {
    for(int v = 0; v <= 1; v++){
        x[k] = v;
        if(k == n-1) solution();
        else TRY(k+1);
    }
}

int main() {
    TRY(0);
}
```

# Backtracking

- Example: generate all binary sequences of length n containing no 2 consecutive bit 1
- Modelling: Array  $x[n]$  where  $x[i] \in \{0,1\}$  is the  $i^{\text{th}}$  bit of the sequence ( $i = 0, \dots, n-1$ )

```
void solution(){
    for(int k = 0; k < n; k++)
        printf("%d", x[k]);
    printf("\n");
}
int check(int v, int k){
    if(k == 0) return 1;
    return v + x[k-1] <= 1;
}
```

```
int TRY(int k) {
    for(int v = 0; v <= 1; v++){
        if(check(v,k)){
            x[k] = v;
            if(k == n-1) solution();
            else TRY(k+1);
        }
    }
}
int main() {
    TRY(0);
}
```

# Backtracking

- Generate all  $k$ -subsets (sets having  $k$  elements) of  $1, 2, \dots, n$
- Modelling: Array  $x[k]$  where  $x[i] \in \{1, \dots, n\}$  is the  $i^{th}$  of the configuration ( $i = 1, \dots, k$ )
- Constraint  $P$ :  $x[i] < x[i+1]$ , for all  $i = 1, 2, \dots, k-1$

```
int TRY(int i) {  
    for(int v = x[i-1]+1; v <= n-k+i;  
        v++){  
        x[i] = v;  
        if(i == k)  
            printSolution();  
        else TRY(i+1);  
    }  
}  
  
int main() {  
    x[0] = 0;  
    TRY(1);  
}
```

# Backtracking

- Generate all permutations of  $1, 2, \dots, n$
- Modelling: Array  $x[1, \dots, n]$  where  $x[i] \in \{1, \dots, n\}$  is the  $i^{th}$  element of the permutation ( $i = 1, \dots, n$ )
- Property  $P$ :  $x[i] \neq x[j]$ , với mọi  $1 \leq i < j \leq n$
- Makring:  $m[v] = \text{true}$  (false) if  $v$  has been appeared (not appeared) in the partial solution, for all  $v = 1, \dots, n$

```
void TRY(int i) {  
    for(int v = 1; v <= n; v++){  
        if(!m[v]) {  
            x[i] = v;  
            m[v] = true; // mark v as appeared  
            if(i == n)  
                solution();  
            else TRY(i+1);  
            m[v] = false;// recover status of v  
        }  
    }  
}  
  
void main() {  
    for(int v = 1; v <= n; v++)  
        m[v] = false;  
    TRY(1);  
}
```

# Backtracking: queen

- Place  $n$  queens on a chess board such that no two queens attack each other
- Modelling:  $x[1, \dots, n]$  where  $x[i]$  is the row index of the queen on column  $i$ ,  $i = 1, \dots, n$
- Property  $P$ 
  - $x[i] \neq x[j]$ ,  $1 \leq i < j \leq n$
  - $x[i] + i \neq x[j] + j$ ,  $1 \leq i < j \leq n$
  - $x[i] - i \neq x[j] - j$ ,  $1 \leq i < j \leq n$

	1	2	3	4
1		X		
2				X
3	X			
4			X	

solution  $x = (3, 1, 4, 2)$

# Backtracking: queen

```
int check(int v, int k) {  
    for(int i = 1; i <= k-1; i++) {  
        if(x[i] == v) return 0;  
        if(x[i] + i == v + k) return 0;  
        if(x[i] - i == v - k) return 0;  
    }  
    return 1;  
}
```

```
void TRY(int k) {  
    for(int v = 1; v <= n; v++) {  
        if(check(v,k)) {  
            x[k] = v;  
            if(k == n) solution();  
            else TRY(k+1);  
        }  
    }  
}  
void main() {  
    TRY(1);  
}
```

# Backtracking: sudoku

- Generate all the ways to fill numbers 1, ..., 9 to cells of a grid 9x9 such that
  - Numbers of each row are different
  - Numbers of each column are different
  - Numbers of each sub-grid (3x3) are different

1	2	3	4	5	6	7	8	9
4	5	6	7	8	9	1	2	3
7	8	9	1	2	3	4	5	6
2	1	4	3	6	5	8	9	7
3	6	5	8	9	7	2	1	4
8	9	7	2	1	4	3	6	5
5	3	1	6	4	2	9	7	8
6	4	2	9	7	8	5	3	1
9	7	8	5	3	1	6	4	2

# Backtracking: sudoku

- Modelling: Array  $x[0..8, 0..8]$
- Property P
  - $x[i, j_2] \neq x[i, j_1]$ , with  $i = 0, \dots, 8$ , và  $0 \leq j_1 < j_2 \leq 8$
  - $x[i_1, j] \neq x[i_2, j]$ , with  $j = 0, \dots, 8$ , và  $0 \leq i_1 < i_2 \leq 8$
  - $x[3I+i_1, 3J+j_1] \neq x[3I+i_2, 3J+j_2]$ , with  $I, J = 0, \dots, 2$ , and  $i_1, j_1, i_2, j_2 \in \{0, 1, 2\}$  such that  $i_1 \neq i_2$  or  $j_1 \neq j_2$

1	2	3	4	5	6	7	8	9
4	5	6	7	8	9	1	2	3
7	8	9	1	2	3	4	5	6
2	1	4	3	6	5	8	9	7
3	6	5	8	9	7	2	1	4
8	9	7	2	1	4	3	6	5
5	3	1	6	4	2	9	7	8
6	4	2	9	7	8	5	3	1
9	7	8	5	3	1	6	4	2

# Backtracking: sudoku

- Order of variables to be considered:  
from up to down, and from left to right
- First tried variable is  $x[0,0]$

1	2	3	4	5	6	7	8	9
4	5	6	7	8	9	1	2	3
7	8	9	1	2	3	4	5	6
2	1	4	3	6	5	8	9	7
3	6	5	8	9	★			

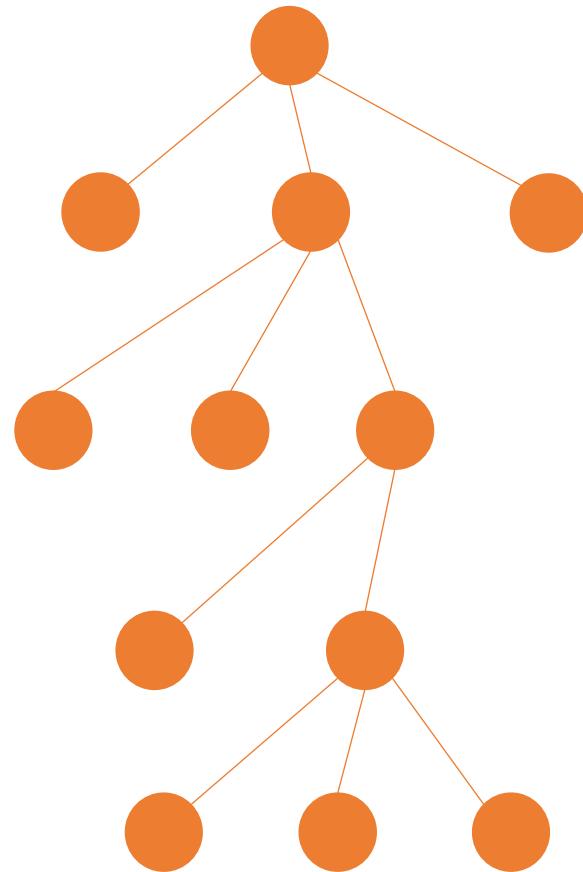
# Backtracking: sudoku

```
bool check(int v, int r, int c){  
    for(int i = 0; i <= r-1; i++)  
        if(x[i][c] == v) return false;  
    for(int j = 0; j <= c-1; j++)  
        if(x[r][j] == v) return false;  
    int I = r/3; int J = c/3;  
    int i = r - 3*I; int j = c - 3*J;  
    for(int i1 = 0; i1 <= i-1; i1++)  
        for(int j1 = 0; j1 <= 2; j1++)  
            if(x[3*I+i1][3*J+j1] == v)  
                return false;  
    for(int j1 = 0; j1 <= j-1; j1++)  
        if(x[3*I+i][3*J+j1] == v)  
            return false;  
    return true;  
}
```

```
void TRY(int r, int c){  
    for(int v = 1; v <= 9; v++){  
        if(check(v,r,c)){  
            x[r][c] = v;  
            if(r == 8 && c == 8){  
                solution();  
            }else{  
                if(c == 8) TRY(r+1,0);  
                else TRY(r,c+1);  
            }  
        }  
    }  
}  
void main(){  
    TRY(0,0);  
}
```

# Branch and Bound

- Solve combinatorial optimization problems: find a solution that minimizes or maximizes an objective function
- Base on exhaustive search
- Each node of the search tree corresponds to a ***partial solution***
- Evaluate the lower/upper bound of the objective function of solutions developed from a current partial solution
  - For minimization problem: if the lower bound is greater than or equal to the value of the objective function of the best solution found so far, then ***cut-off*** (do not expand from the current partial solution to compete solutions)
  - For maximization problem: if the upper bound is less than or equal to the value of the objective function of the best solution found so far, then cut-off



# Branch and Bound

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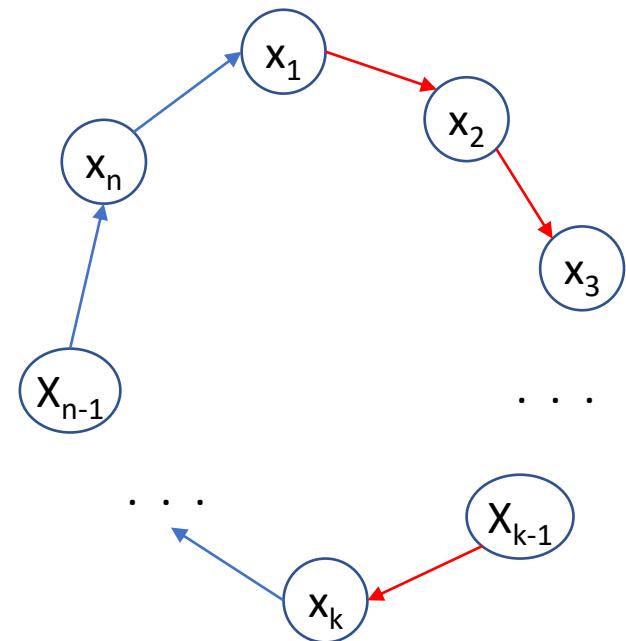
- Travelling Salesman Problem (TSP)
  - There are  $n$  cities  $1, 2, \dots, n$ . The cost for travelling from city  $i$  to  $j$  is  $c(i, j)$ . Find a closed tour starting from city 1, visiting other cities exactly once and coming back to 1 having minimal total cost
- Modelling: Solution  $x = (x_1, x_2, \dots, x_n)$  where  $x_i \in \{1, 2, \dots, n\}$ 
  - Constraint:  $x_i \neq x_j, \forall 1 \leq i < j \leq n$
  - Objective function

$$f(x) = c(x_1, x_2) + c(x_2, x_3) + \dots + c(x_n, x_1) \rightarrow \min$$

# Branch and Bound: TSP

- $c_m$  : minimum cost among the cost between two cities
- Partial solution  $(x_1, \dots, x_k)$ 
  - Cost of the partial solution  $f = c(x_1, x_2) + c(x_2, x_3) + \dots + c(x_{k-1}, x_k)$
  - Lower bound

$$g(x_1, \dots, x_k) = f + c_m \times (n-k+1)$$



# Branch and Bound: TSP

```
void TRY(int k){  
    for(int v = 1; v <= n; v++){  
        if(marked[v] == false){  
            a[k] = v;  
            f = f + c[a[k-1]][a[k]];  
            marked[v] = true;  
            if(k == n){  
                solution();  
            }else{  
                int g = f + cmin*(n-k+1);  
                if(g < f_min)  
                    TRY(k+1);  
            }  
            marked[v] = false;  
            f = f - c[a[k-1]][a[k]];  
        }  
    }  
}
```

```
void solution() {  
    if(f + c[x[n]][x[1]] < f_min){  
        f_min = f + c[x[n]][x[1]];  
    }  
}  
void main() {  
    f_min = 9999999999;  
    for(int v = 1; v <= n; v++)  
        marked[v] = false;  
    x[1] = 1; marked[1] = true;  
    TRY(2);  
}
```

# Greedy algorithms

---

- Problem-based heuristics
- Each step, choose the best decision based on local information, do not take into account negative impacts of that decision in the future
- Not sure to find global optimal solutions

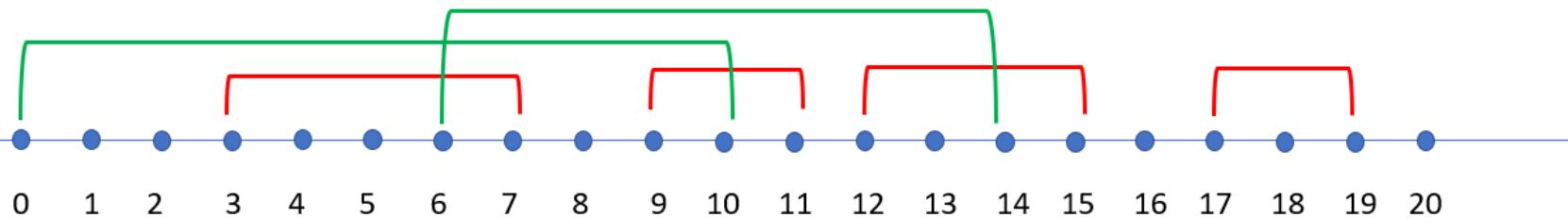
# Greedy algorithms

- S: solution under construction (set of components)
- C: set of candidates
- select(C): select the most promising component to add to S
- solution(S): return true if S is a solution to the problem
- feasible(S): return true if S does not violate any constraints

```
Greedy() {  
    S = {};  
    while C ≠ Ø and  
        not solution(S){  
        x = select(C);  
        C = C \ {x};  
        if feasible(S ∪ {x}) {  
            S = S ∪ {x};  
        }  
    }  
    return S;  
}
```

# Greedy algorithms: disjoint segments

- Given a set of segments  $X = \{(a_1, b_1), \dots, (a_n, b_n)\}$  in which  $a_i < b_i$  are coordinates of the segment  $i$  on a line,  $i = 1, \dots, n$ .
- Goal: Find a subset of  $X$  having largest cardinality in which no two segments of the subset intersect



Optimal solution:  $S = \{(3,7), (9,11), (12,15), (17,19)\}$

# Greedy algorithms: disjoint segments

```
Greedy1() {  
    S = {};  
    L = sort segments of X in a non-decreasing order of  $a_i$ ;  
    while ( $X \neq \emptyset$ ) {  
         $(a_c, b_c)$  = select first segment of L;  
        remove  $(a_c, b_c)$  from L;  
        if feasible( $S \cup \{(a_c, b_c)\}$ ) {  
            S =  $S \cup \{(a_c, b_c)\}$ ;  
        }  
    }  
    return S;  
}
```

# Greedy algorithms: disjoint segments

---

- Greedy 1 cannot ensure to find optimal solutions, example

$$X = \{(1,11), (2,5), (6,10)\}$$

→ Greedy 1 returns solutions  $\{(1,11)\}$ , however, optimal solution is  $\{(2,5), (6,10)\}$

# Greedy algorithms: disjoint segments

```
Greedy2() {  
    S = {};  
    L = sort segments of X in a non-decreasing order of  $b_i - a_i$ ;  
    while ( $X \neq \emptyset$ ) {  
         $(a_c, b_c)$  = select first segment of L;  
        remove  $(a_c, b_c)$  from L;  
        if feasible( $S \cup \{(a_c, b_c)\}$ ) {  
            S =  $S \cup \{(a_c, b_c)\}$ ;  
        }  
    }  
    return S;  
}
```

# Greedy algorithms: disjoint segments

---

- Greedy 2 cannot ensure to find optimal solutions, example

$$X = \{(1,5), (4,7), (6,11)\}$$

→ Greedy 2 returns solution  $\{(4,7)\}$ , however, optimal solution is  $\{(1,5), (6,11)\}$

# Greedy algorithms: disjoint segments

```
Greedy3() {
    S = {};
    L = sort segments of X in a non-decreasing order of  $b_i$ ;
    while ( $X \neq \emptyset$ ) {
         $(a_c, b_c)$  = select first segment of L;
        remove  $(a_c, b_c)$  from L;
        if feasible( $S \cup \{(a_c, b_c)\}$ ) {
            S =  $S \cup \{(a_c, b_c)\}$ ;
        }
    }
    return S;
}
```

# Greedy algorithms: exercise

---

- Given a set of jobs  $1, 2, \dots, n$  where each job  $i$  has a deadline  $d(i)$  and associated profit  $p(i)$  if the job is finished before the deadline. It is assumed that every job takes the single unit of time and each job is executed at a time. Compute a subset of jobs to be scheduled such that each job is not finished after the deadline and total profits of jobs is maximal.

# Greedy algorithms: exercise

---

- Example: there are 4 jobs with deadline  $d[i]$  and profit  $p[i]$  are given as follows:
  - Job 1: 4 20
  - Job 2: 1 10
  - Job 3: 1 40
  - Job 4: 1 30
- Optimal solution consists of jobs 1 and 3 with the following schedule:
  - Job 3 starts at time-point 0 and finishes at time-point 1
  - Job 1 starts at time-point 1 and finishes at time-point 2
- Total profit is  $20 + 40 = 60$

# Divide and conquer

---

- Divide the original problem into independent subproblems
- Solve recursively subproblems
- Combine the results of subproblems to establish the solution to the original problem

# Divide and conquer: binary search

- Given a sequence  $x[1..n]$  which is sorted in an increasing order and a value  $y$ . Find the index  $i$  such that  $x[i] = y$

```
bSearch(x, start, finish, y) {  
    if(start == finish) {  
        if(x[start] == y)  
            return start;  
        else return -1;  
    }else{  
        m = (start + finish)/2;  
        if(x[m] == y) return m;  
        if(x[m] < y)  
            return bSearch(x, m+1,finish,y);  
        else  
            return bSearch(x,start,m-1,y);  
    }  
}
```

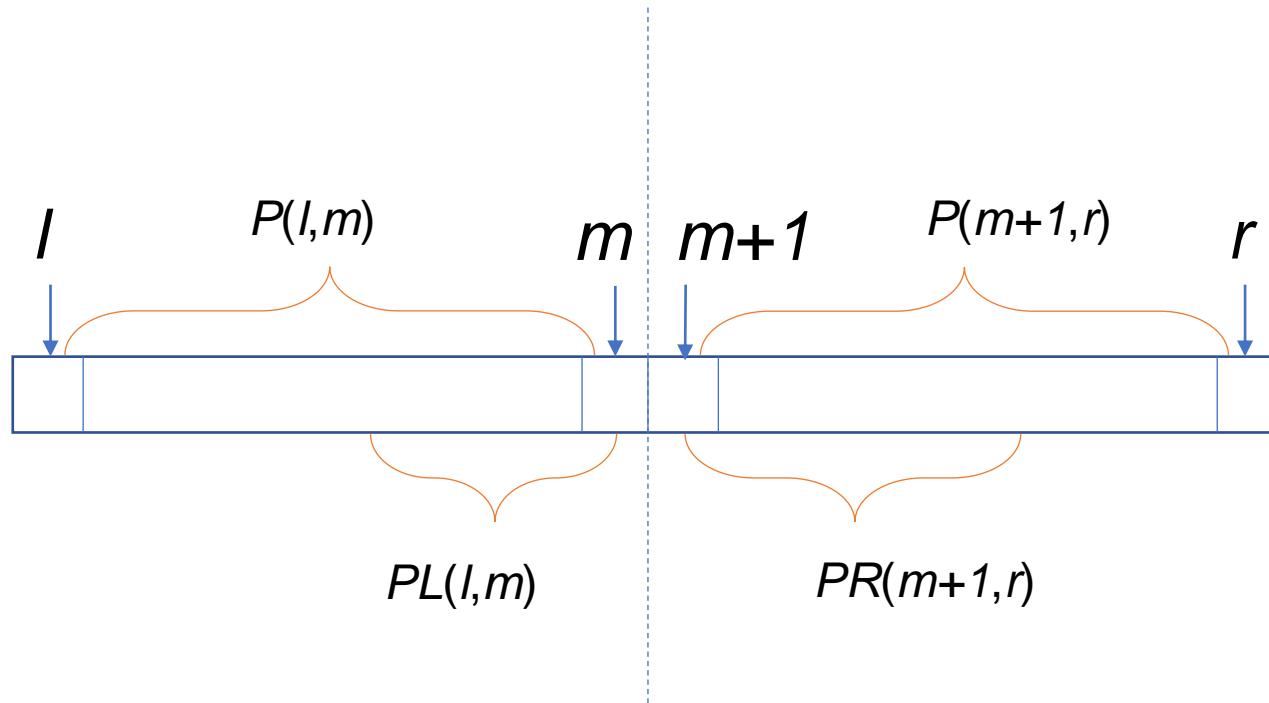
# Divide and conquer: largest subsequence

---

- Given a sequence of numbers  $a = a_1, a_2, \dots, a_n$ . Find a subsequence (consecutive elements) such that the sum of its elements is largest
- Division:  $P(i, j)$  is the sum of elements of largest subsequence of  $a_i, a_{i+1}, \dots, a_j$
- Combination
  - Denote  $PL(i, j)$  the sum of elements of the largest subsequence of  $a_i, a_{i+1}, \dots, a_j$  such that the last element of that largest subsequence is  $a_j$
  - Denote  $PR(i, j)$  the sum of elements of the largest subsequence of  $a_i, a_{i+1}, \dots, a_j$  such that the first element of that largest subsequence is  $a_i$

# Divide and conquer: largest subsequence

- Consider the interval  $[l, l+1, \dots, r]$ . Denote  $m = (l+r)/2$
- $P(l, r) = \text{MAX}\{P(l, m), P(m+1, r), PL(l, m) + PR(m+1, r)\}$



# Divide and conquer: largest subsequence

```
maxLeft(a, l, r){  
    max = -∞; s = 0;  
    for i = r downto l do{  
        s += a[i];  
        if(s > max) max = s;  
    }  
    return max;  
}  
  
maxRight(a, l, r){  
    max = -∞; s = 0;  
    for i = l to r do{  
        s += a[i];  
        if(s > max) max = s;  
    }  
    return max;  
}
```

```
maxSeq(a[...], l, r){  
    if l = r then return a[r];  
    m = (l+r)/2;  
    ml = maxSeq(a,l,m);  
    mr = maxSeq(a,m+1,r);  
    m lr = maxLeft(a,l,m) +  
           maxRight(a,m+1,r);  
    return max(ml, mr, m lr);  
}  
  
main() {  
    input a[1..n]  
    result = maxSeq(a,1,n);  
}
```

# Divide and conquer: running time analysis

- Divide the original problem into a subproblems of size  $n/b$
- $T(n)$ : running time of the problem of size  $n$
- Division time (line 4):  $D(n)$
- Combination time (line 6):  $C(n)$
- Recurrence relation:

$$T(n) = \begin{cases} \Theta(1), & n \leq n_0 \\ aT\left(\frac{n}{b}\right) + C(n) + D(n), & n > n_0 \end{cases}$$

```
procedure D-and-C(n) {  
1. if( $n \leq n_0$ )  
2.   Process directly  
3. else{  
4.   Divide original problem  
       into  $a$  subproblems of size  $n/b$   
5.   Solve  $a$  subproblems recursively  
6.   Combination of solutions  
7. }  
}
```

# Master theorem

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- Recurrence relation:

$$T(n) = aT(n/b) + cn^k, \text{ with constants } a \geq 1, b > 1, c > 0$$

- If  $a > b^k$  then  $T(n) = \Theta(n^{\log_b a})$
- If  $a = b^k$  then  $T(n) = \Theta(n^k \log n)$  with  $\log n = \log_2 n$
- If  $a < b^k$  then  $T(n) = \Theta(n^k)$

# Dynamic programming

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- Break the given problem down into a set of simpler subproblems
- Solve these subproblems once and store solutions in memory (e.g., tables) in a bottom-up fashion
- Smallest subproblems can be solved trivially
- Recurrence relation: solutions to a subproblem can be computed based on solutions (solved and stored in memory) to smaller subproblems

# Dynamic programming

---

- Given an integers sequence  $a = (a_1, a_2, \dots, a_n)$ . A subsequence of  $a$  is defined to be  $a_i, a_{i+1}, \dots, a_j$ . The weight of a subsequence is the sum of its elements. Find the subsequence having the highest weight

# Dynamic programming

- Subproblem definition
  - $S_i$ : weight of the highest subsequence of  $a_1, \dots, a_i$  in which the last element of the subsequence is  $a_i$ ,  $i = 1, \dots, n$
  - Smallest problem can be solved directly:  $S_1 = a_1$
  - Recurrence relation

$$S_i = \begin{cases} S_{i-1} + a_i, & \text{if } S_{i-1} > 0 \\ a_i, & \text{otherwise} \end{cases}$$

# Dynamic programming

---

- Longest increasing subsequence: given a sequence  $a = a_1, a_2, \dots, a_n$ . A subsequence of  $a$  is created by removing some elements of  $a$ . Find a subsequence of  $a$  which is an increasing sequence and the length is longest

# Dynamic programming

---

- Division
  - Denote  $P_i$  the problem of finding the longest increasing subsequence of  $a_1, \dots, a_i$  such that the last element is  $a_i$ , for all  $i = 1, \dots, n$
  - Denote  $S_i$  the number of element of the solution to  $P_i$ ,  $\forall i = 1, \dots, n$

# Dynamic programming

---

- Division
  - Denote  $P_i$  the problem of finding the longest increasing subsequence of  $a_1, \dots, a_i$  such that the last element is  $a_i$ , for all  $i = 1, \dots, n$
  - Denote  $S_i$  the number of elements of the solution to  $P_i$ ,  $\forall i = 1, \dots, n$
  - $S_1 = 1$
  - $S_i = \max\{1, \max\{S_j + 1 | j < i \wedge a_j < a_i\}\}$
  - Solution to the original problem is  $\max\{S_1, S_2, \dots, S_n\}$

# Dynamic programming

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  - Solution to the original problem is  $\max\{S_1, S_2, \dots, S_n\}$

```
void solve(a[1...n]){
    S[1] = 1;
    rs = S[1];
    for i = 2 to n do{
        S[i] = 1;
        for j = i-1 downto 1 do{
            if(a[i] > a[j]){
                if(S[j] + 1 > S[i])
                    S[i] = S[j] + 1;
            }
        }
        rs = max(S[i], rs);
    }
    return rs
}
```

# Dynamic programming

---

- Longest common subsequence
  - Denote  $X = \langle X_1, X_2, \dots, X_n \rangle$ , a subsequence of  $X$  is created by removing some element from  $X$
  - Input: Given two sequences  $X = \langle X_1, X_2, \dots, X_n \rangle$  and  $Y = \langle Y_1, Y_2, \dots, Y_m \rangle$
  - Output: Find a common subsequence of  $X$  and  $Y$  such that the length is longest

# Dynamic programming

- Longest common subsequence
  - Division
    - Denote  $S(i, j)$  the length of the longest common subsequence of  $\langle X_1, \dots, X_i \rangle$  and  $\langle Y_1, \dots, Y_j \rangle$ ,  $\forall i = 1, \dots, n$  and  $j = 1, \dots, m$
    - Base case
      - $\forall j = 1, \dots, m$ :  $S(1, j) = \begin{cases} 1, & \text{if } X_1 \text{ appears in } Y_1, \dots, Y_j \\ 0, & \text{otherwise} \end{cases}$
      - $\forall i = 1, \dots, n$ :  $S(i, 1) = \begin{cases} 1, & \text{if } Y_1 \text{ appears in } X_1, \dots, X_i \\ 0, & \text{otherwise} \end{cases}$
    - Aggregation

$$S(i, j) = \begin{cases} S(i-1, j-1) + 1, & \text{if } X_i = Y_j \\ \max\{S(i-1, j), S(i, j-1)\}, & \text{otherwise} \end{cases}$$

# Dynamic programming

X

3	7	2	5	1	4	9
---	---	---	---	---	---	---

Y

4	3	2	3	6	1	5	4	9	7
---	---	---	---	---	---	---	---	---	---

$S(i,j)$

1 2 3 4 5 6 7 8 9 10

1	0	1	1	1	1	1	1	1	1	1
2	0	1	1	1	1	1	1	1	1	2
3	0	1	2	2	2	2	2	2	2	2
4	0	1	2	2	2	2	3	3	3	3
5	0	1	2	2	2	3	3	3	3	3
6	1	1	2	2	2	3	3	4	4	4
7	1	1	2	2	2	3	3	4	5	5

# Dynamic programming

```
solve(X[1..n], Y[1..m]){
    rs = 0;
    for i = 0 to n do S[i][0] = 0;
    for j = 0 to m do S[0][j] = 0;

    for i = 1 to n do{
        for j = 1 to m do{
            if(X[i] == Y[j]) S[i][j] = S[i-1][j-1] + 1;
            else{
                S[i][j] = max(S[i-1][j], S[i][j-1]);
            }
            rs = max(S[i][j], rs);
        }
    }
    return rs;
}
```