



# Summary

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1. Definition and divide-and-conquer strategies
2. Recursion: design tips
3. Simple recursive algorithms
  1. Fibonacci numbers
  2. Dicothomic search
  3. X-Expansion
  4. Anagrams
  5. Knapsack
4. Recursive vs Iterative strategies
5. More complex examples of recursive algorithms
  1. Knight's Tour
  2. Proposed exercises



# Definition and divide-and-conquer strategies

Recursion

# Why recursion?

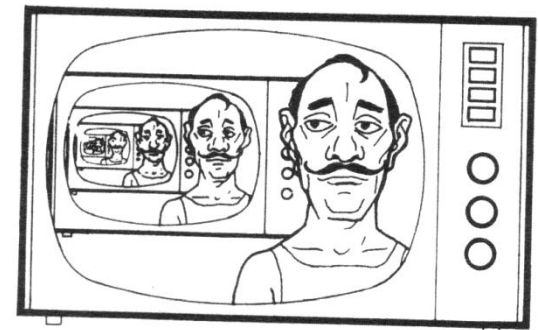
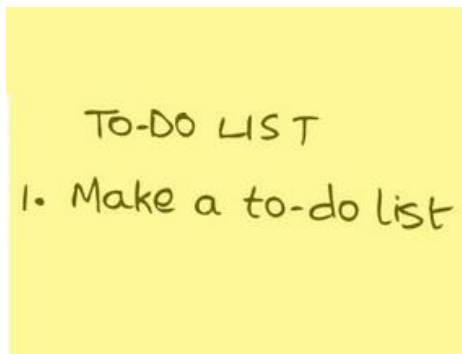
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- ▶ Divide et impera
- ▶ Systematic exploration/enumeration
- ▶ Handling recursive data structures

# Definition

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- ▶ A method (or a procedure or a function) is defined as recursive when:
  - ▶ Inside its definition, we have a call to the same method (procedure, function)
  - ▶ Or, inside its definition, there is a call to another method that, directly or indirectly, calls the method itself
- ▶ An algorithm is said to be recursive when it is based on recursive methods (procedures, functions)





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devi capire la  
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# Example: Factorial

$$\left\{ \begin{array}{l} 0! \stackrel{\text{def}}{=} 1 \\ \forall N \geq 1: \\ N! \stackrel{\text{def}}{=} N \times (N-1)! \end{array} \right.$$

```
public long recursiveFactorial(long N)
{
    long result = 1 ;

    if ( N == 0 )
        return 1 ;
    else {
        result = recursiveFactorial(N-1) ;
        result = N * result ;
        return result ;
    }
}
```

# Motivation

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- ▶ Many problems lend themselves, naturally, to a recursive description:
  - ▶ We define a method to solve sub-problems similar to the initial one, but smaller
  - ▶ We define a method to combine the partial solutions into the overall solution of the original problem



Gaius Julius Caesar



# Recursion

---

## ▶ **Divide et Impera**

- ▶ Split a problem  $\mathcal{P}$  into  $\{Q_i\}$  where  $Q_i$  are still complex, yet *simpler* instances of the same problem.
- ▶ Solve  $\{Q_i\}$ , then merge the solutions
- ▶ Merge & split must be “simple”
- ▶ A.k.a., *Divide 'n Conquer*

## ▶ **Exploration**

- ▶ Systematic procedure to enumerate all possible solutions
- ▶ Solutions (built stepwise)
  - ▶ Paths
  - ▶ Permutations
  - ▶ Combinations
- ▶ Divide et Impera, by “dividing” the possible solutions

# Divide et Impera – Divide and Conquer

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- ▶ Solution = **Solve** ( Problem ) ;
- ▶ **Solve** ( Problem ) {
  - ▶ List<SubProblem> subProblems = **Divide** ( Problem ) ;
  - ▶ For ( each subP[i] in subProblems ) {
    - ▶ SubSolution[i] = **Solve** ( subP[i] ) ;
  - ▶ }
  - ▶ Solution = **Combine** ( SubSolution[ 1..N ] ) ;
  - ▶ return Solution ;
- ▶ }

# Divide et Impera – Divide and Conquer

---

▶ Solution = **Solve** ( Problem ) ;

▶ **Solve** ( Problem ) {

▶ List<SubProblem> subProblems = **Divide** ( Problem ) ;

▶ For ( each subP[i] in subProblems ) {

▶ SubSolution[i] = **Solve** ( subP[i] ) ;

▶ }

▶ Solution = **Combine** ( SubSolution[ 1..N ] ) ;

▶ return Solution ;

▶ }

recursive call

“a” sub-problems, each  
“b” times smaller than  
the initial problem

# How to stop recursion?

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- ▶ Recursion **must not** be infinite
  - ▶ Any algorithm must always terminate!
- ▶ After a sufficient nesting level, sub-problems become so small (and so easy) to be solved:
  - ▶ Trivially (ex: sets of just one element)
  - ▶ Or, with methods different from recursion

# Warnings

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- ▶ Always remember the “termination condition”
- ▶ Ensure that all sub-problems are strictly “smaller” than the initial problem

# Divide et Impera – Divide and Conquer

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- ▶ **Solve** ( Problem ) {
  - ▶ if( problem is trivial )
    - ▶ Solution = **Solve\_trivial** ( Problem ) ;
  - ▶ else {
    - ▶ List<SubProblem> subProblems = **Divide** ( Problem ) ;
    - ▶ For ( each subP[i] in subProblems ) {
      - SubSolution[i] = **Solve** ( subP[i] ) ;
    - ▶ }
    - ▶ Solution = **Combine** ( SubSolution[ 1..N ] ) ;
  - ▶ }
  - ▶ return Solution ;
  - ▶ }

do recursion

# What about complexity?

---

- ▶  $a$  = number of sub-problems for a problem
- ▶  $b$  = how smaller sub-problems are than the original one
- ▶  $n$  = size of the original problem
- ▶  $T(n)$  = complexity of **Solve**
  - ▶ ...our unknown complexity function
- ▶  $\Theta(1)$  = complexity of **Solve\_trivial**
  - ▶ ...otherwise it wouldn't be trivial
- ▶  $D(n)$  = complexity of **Divide**
- ▶  $C(n)$  = complexity of **Combine**

# Divide et Impera – Divide and Conquer

- ▶ **Solve** ( Problem ) {
  - ▶ if( problem is trivial )
    - ▶ Solution = **Solve\_trivial** ( Problem ) ;  $\Theta(1)$
  - ▶ else {
    - ▶ List<SubProblem> subProblems = **Divide** ( Problem ) ;  $D(n)$
    - ▶ For ( each subP[i] in subProblems ) {  $a$  times
      - SubSolution[i] = **Solve** ( subP[i] ) ;  $T(n/b)$
    - ▶ }
    - ▶ Solution = **Combine** ( SubSolution[ 1.. $a$  ] ) ;  $C(n)$
  - ▶ }
  - ▶ return Solution ;
- ▶ }



# Complexity computation

---

- ▶  $T(n) =$ 
  - ▶  $\Theta(1)$  for  $n \leq c$
  - ▶  $D(n) + aT(n/b) + C(n)$  for  $n > c$
- ▶ Recurrence Equation not easy to solve in the general case
- ▶ Special case:
  - ▶ If  $D(n)+C(n)=\Theta(n)$
  - ▶ We obtain  **$T(n) = \Theta(n \log n)$** .

# Exploration

---

- ▶ **Explore** ( **S** ) {
  - ▶ List<Step> steps = **PossibleSteps** ( Problem, **S** ) ;
  - ▶ for ( each **p** in steps ) {
    - ▶ **S.Do** ( **p** )
    - ▶ **Explore** ( **S** ) ;
    - ▶ **S.Undo** ( **p** ) ;
  - ▶ }
- ▶ }

# Exploration

The “status” of the problem

- ▶ **Explore** ( **S** ) {
  - ▶ List<Step> steps = **PossibleSteps** ( Problem, **S** );
  - ▶ for ( each **p** in steps ) {
    - ▶ **S.Do** ( **p** )
    - ▶ **Explore** ( **S** );
    - ▶ **S.Undo** ( **p** );
  - ▶ }
- ▶ }

Local variable

“Try” the step

Recursion

Backtrack!



# Goal

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1. Analysis of a problem to be solved with recursive techniques
2. Identification of the main design choices
3. Identification of the main implementation strategies

# Analizzare il problema

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- ▶ Come imposto in generale la ricorsione?
- ▶ Che cosa mi rappresenta il "livello"?
- ▶ Com'è fatta una soluzione parziale?
- ▶ Com'è fatta una soluzione totale?

# Generale le possibili soluzioni

---

- ▶ Qual è la regola per generare tutte le soluzioni del livello+1 a partire da una soluzione parziale del livello corrente?
- ▶ Come faccio a riconoscere se una soluzione parziale è anche completa? (terminazione con successo)
- ▶ Come viene avviata la ricorsione (livello 0)?

# Identificare le soluzioni valide

---

- ▶ Data una soluzione **parziale**, come faccio a
  - ▶ sapere se è valida (e quindi continuare)?
  - ▶ sapere se non è valida (e quindi terminare la ricorsione)?
  - ▶ nb. magari non posso...
- ▶ Data una soluzione **completa**, come faccio a
  - ▶ sapere se è valida?
  - ▶ sapere se non è valida?
- ▶ Cosa devo fare con le soluzioni complete valide?
  - ▶ Fermarmi alla prima?
  - ▶ Generarle e memorizzarle tutte?
  - ▶ Contarle?



# Progettare le strutture dati

---

- ▶ Qual è la struttura dati per memorizzare una soluzione (parziale o completa)?
- ▶ Qual è la struttura dati per memorizzare lo stato della ricerca (della ricorsione)?

# Scheletro del codice

---

```
// Struttura di un algoritmo ricorsivo generico

void recursive (... , level) {

    // E -- sequenza di istruzioni che vengono eseguite sempre
    // Da usare solo in casi rari (es. Ruzzle)
    doAlways();

    // A
    if (condizione di terminazione) {
        doSomething;
        return;
    }

    // Potrebbe essere anche un while ()
    for () {

        // B
        generaNuovaSoluzioneParziale;

        if (filtro) { // C
            recursive (... , level + 1);
        }

        // D
        backtracking;
    }
}
```

# Riempire lo scheletro (del codice)

Blocco	Frammento di codice
A	
B	
C	
D	
E	

```
// Struttura di un algoritmo ricorsivo
void recursive (... , level) {
    // E -- sequenza di istruzioni che ve
    // Da usare solo in casi rari (es. R
    doAlways();

    // A
    if (condizione di terminazione) {
        doSomething;
        return;
    }

    // Potrebbe essere anche un while ()
    for () {

        // B
        generaNuovaSoluzioneParziale;

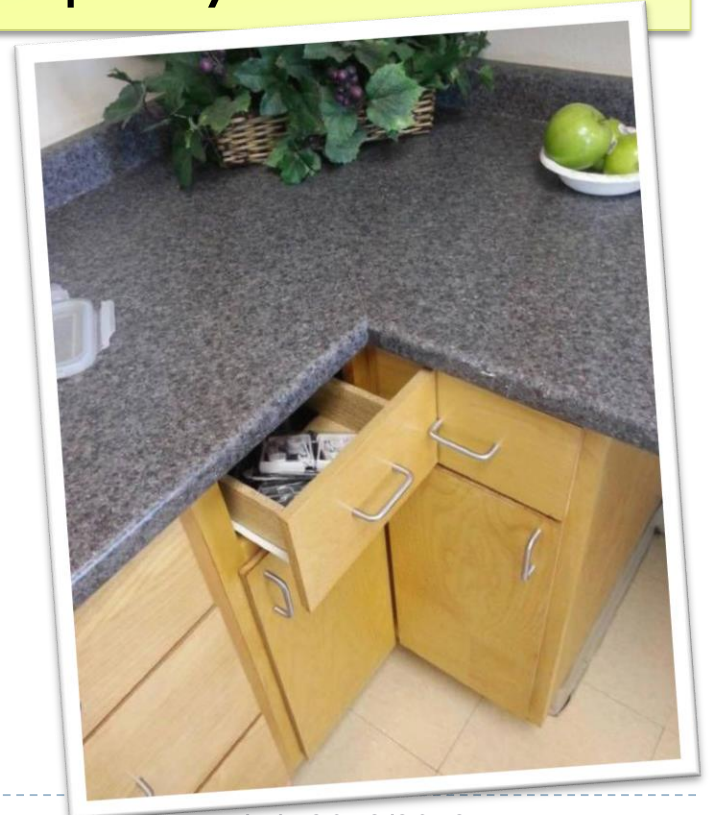
        if (filtro) { // C
            recursive (... , level + 1);
        }

        // D
        backtracking;
    }
}
```

# Recursion myths

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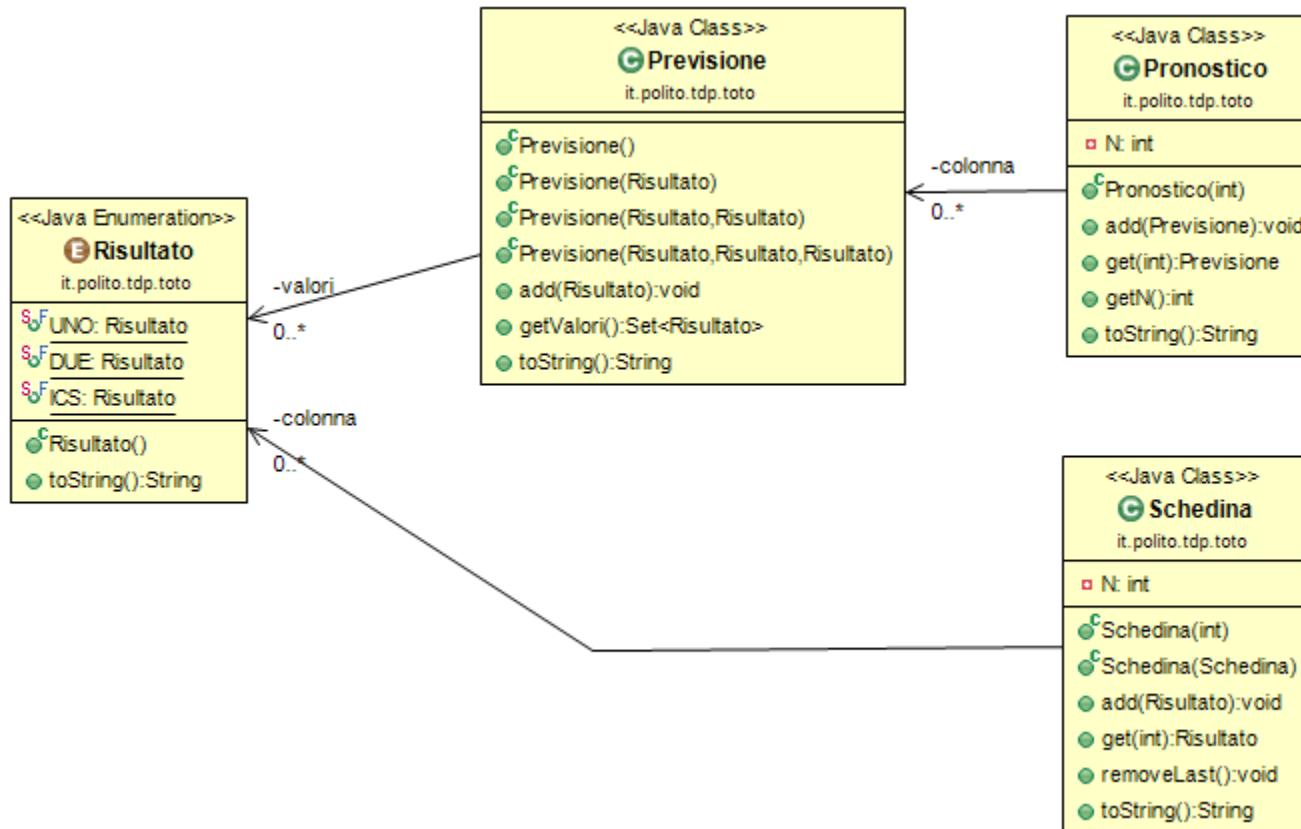
- ▶ Recursive algorithms are  $O(n \log n)$
- ▶ Recursive algorithms are better than non-recursive ones
- ▶ Recursive algorithms can be coded quickly







# Classi



# Fibonacci Numbers

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▶ **Problem:**

- ▶ Compute the N-th Fibonacci Number

▶ **Definition:**

- ▶  $FIB_{N+1} = FIB_N + FIB_{N-1}$  for  $N > 0$
- ▶  $FIB_1 = 1$
- ▶  $FIB_0 = 0$



# Recursive solution

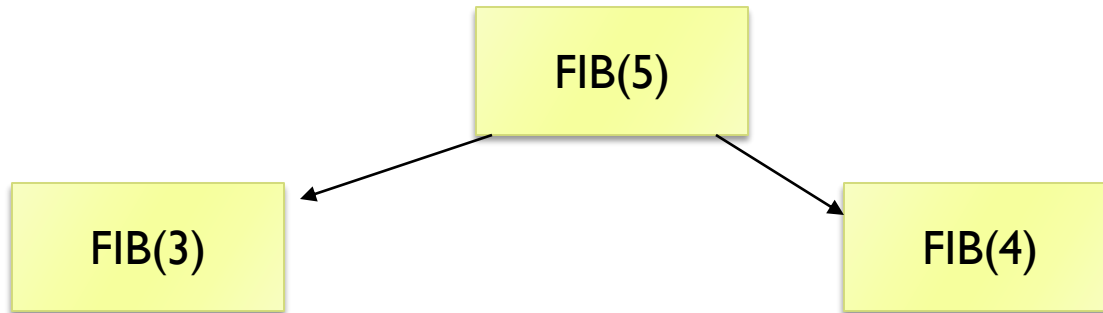
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```
public long recursiveFibonacci(long N) {  
    if(N==0)  
        return 0 ;  
    if(N==1)  
        return 1 ;  
  
    long left = recursiveFibonacci(N-1) ;  
    long right = recursiveFibonacci(N-2) ;  
  
    return left + right ;  
}
```

```
Fib(0) = 0  
Fib(1) = 1  
Fib(2) = 1  
Fib(3) = 2  
Fib(4) = 3  
Fib(5) = 5
```

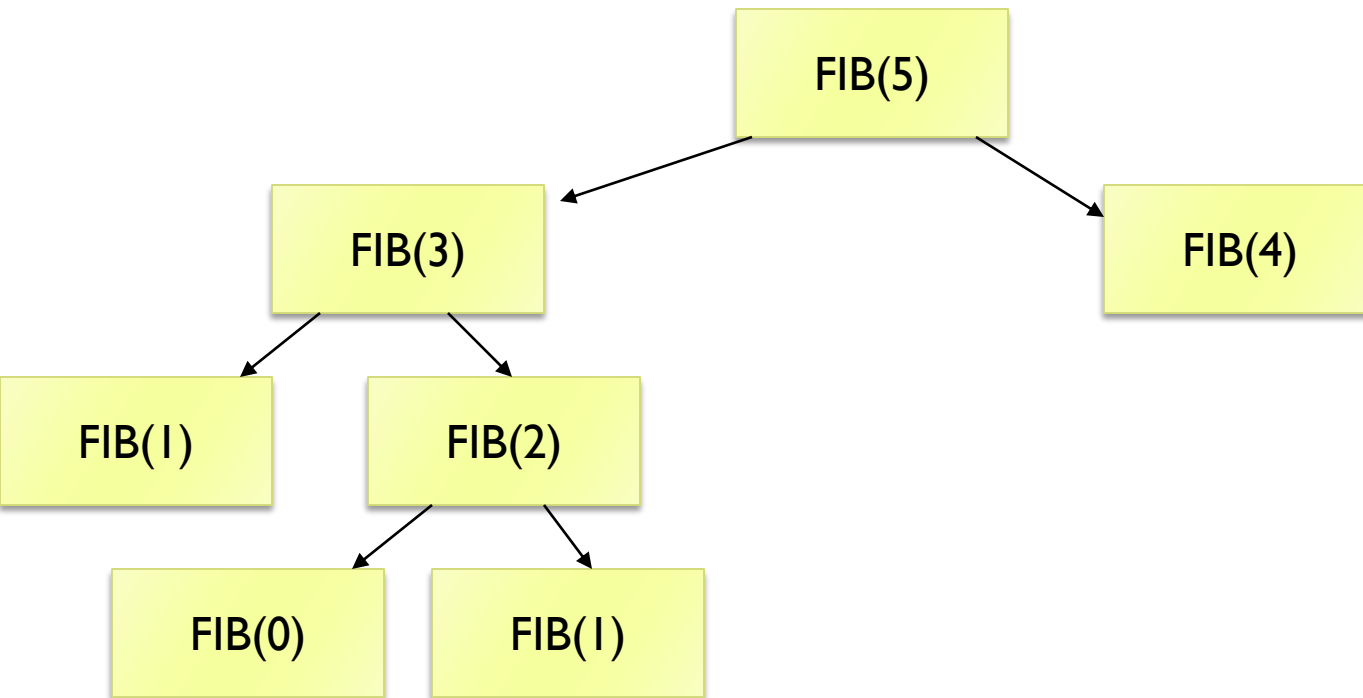
# Analysis

---



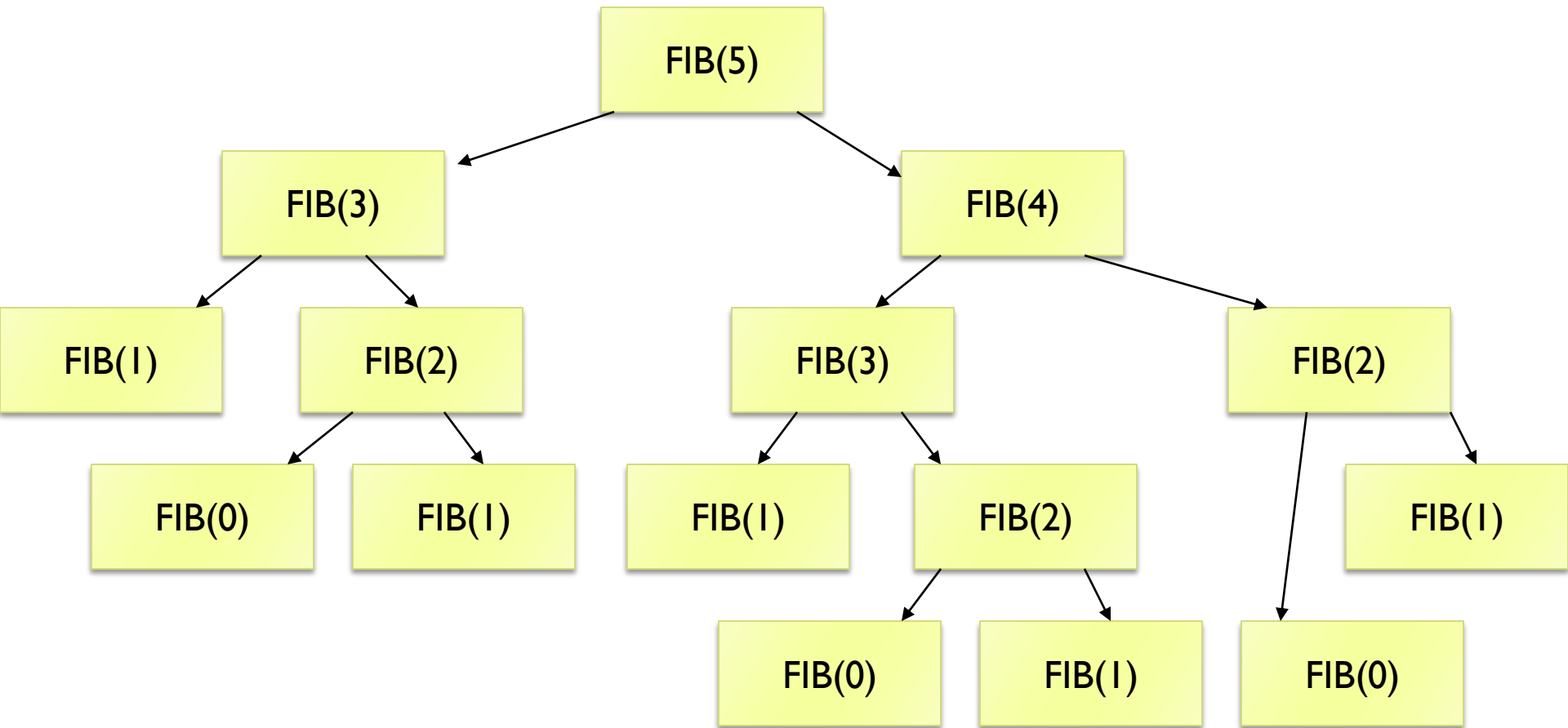
# Analysis

---



# Analysis

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# Example: dichotomic search

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## ▶ Problem

- ▶ Determine whether an element  $x$  is **present** inside an ordered **vector**  $v[N]$

## ▶ Approach

- ▶ Divide the vector in two halves
- ▶ Compare the middle element with  $x$
- ▶ Reapply the problem over one of the two halves (left or right, depending on the comparison result)
- ▶ The other half may be ignored, since the vector is ordered

# Example

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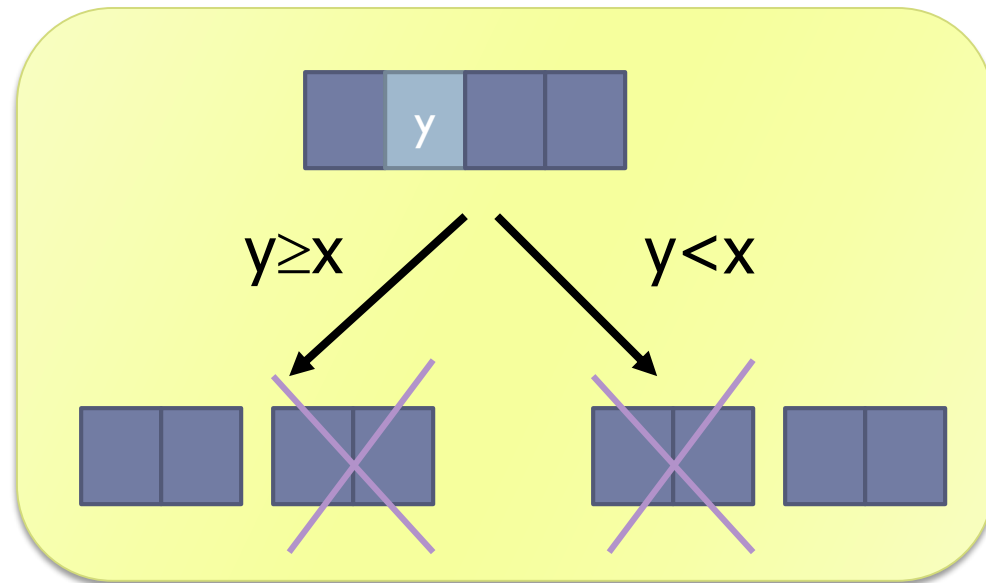
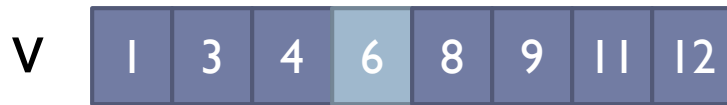
v

1	3	4	6	8	9	11	12
---	---	---	---	---	---	----	----

x

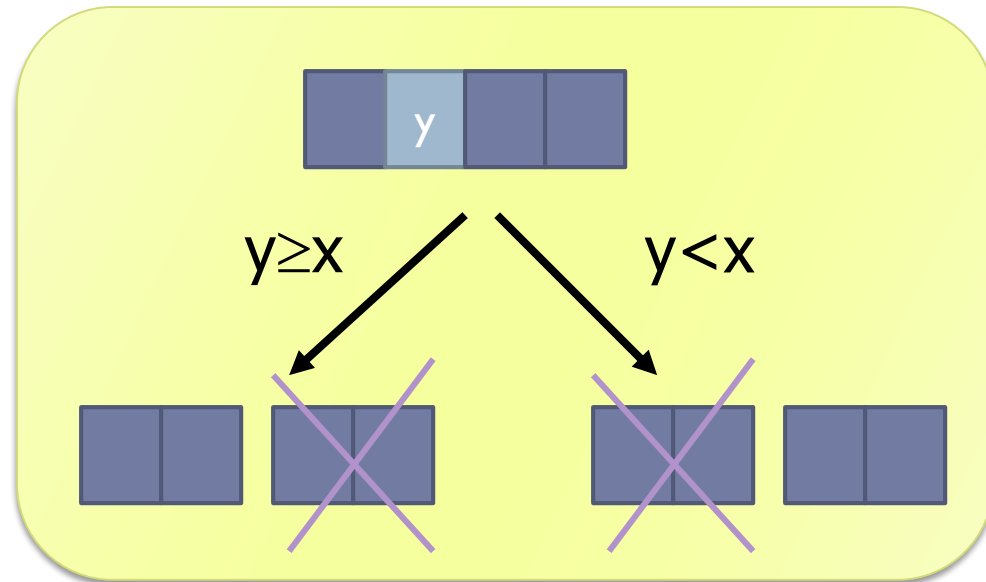
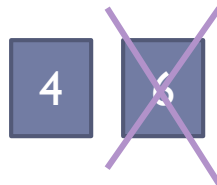
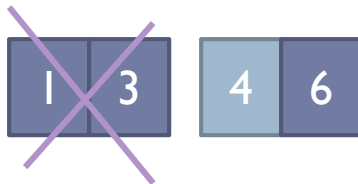
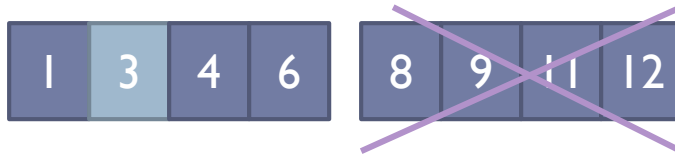
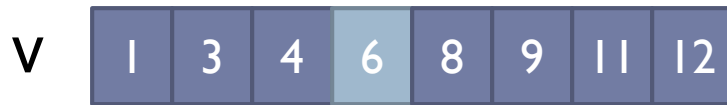
4
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# Example





# Example



# Solution

```
public int find(int[] v, int a, int b, int x)
{
    if(b-a == 0) { // trivial case
        if(v[a]==x) return a ; // found
        else return -1 ;      // not found
    }

    int c = (a+b) / 2 ; // splitting point
    if(v[c] >= x)
        return find(v, a, c, x) ;
    else return find(v, c+1, b, x) ;
}
```

## Solution


```
public int find(v, a, b, x)
{
    if(b-a < 1)
        return v[a];

    int c = (a+b) / 2; // fitting point
    if(v[c] >= x)
        return find(v, a, c, x) ;
    else return find(v, c+1, b, x) ;
}
```

**Beware of integer-arithmetic approximations!**

# Quick reference

BINARY SEARCH		
Best	Average	Worst
$O(1)$	$O(\log n)$	$O(\log n)$



Array

Divide and Conquer

```

search (A, t)
1.  low = 0
2.  high = n - 1
3.  while (low ≤ high) do
4.    ix = (low + high) / 2
5.    if (t = A[ix]) then
6.      return true
7.    else if (t < A[ix]) then
8.      high = ix - 1
9.    else low = ix + 1
10. return false
end
    
```

*search (A, 11)*

*low*                      *ix*                      *high*

*first pass*

1	4	8	9	11	15	17
---	---	---	---	----	----	----

*low*   *ix*   *high*

*second pass*

1	4	8	9	11	15	17
---	---	---	---	----	----	----

*low*

*ix*

*high*

*third pass*

1	4	8	9	11	15	17
---	---	---	---	----	----	----

} explored elements

## Exercise: Value X

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- ▶ When working with Boolean functions, we often use the symbol  $X$ , meaning that a given variable may have indifferently the value  $0$  or  $1$ .
- ▶ Example: in the OR function, the result is  $1$  when the inputs are  $01$ ,  $10$  or  $11$ . More compactly, if the inputs are  $X1$  or  $1X$ .

# X-Expansion

---

- ▶ We want to devise an algorithm that, given a binary string that includes characters 0, 1 and X, will compute all the possible combinations implied by the given string.
- ▶ Example: given the string 01X0X, algorithm must compute the following combinations
  - ▶ 01000
  - ▶ 01001
  - ▶ 01100
  - ▶ 01101

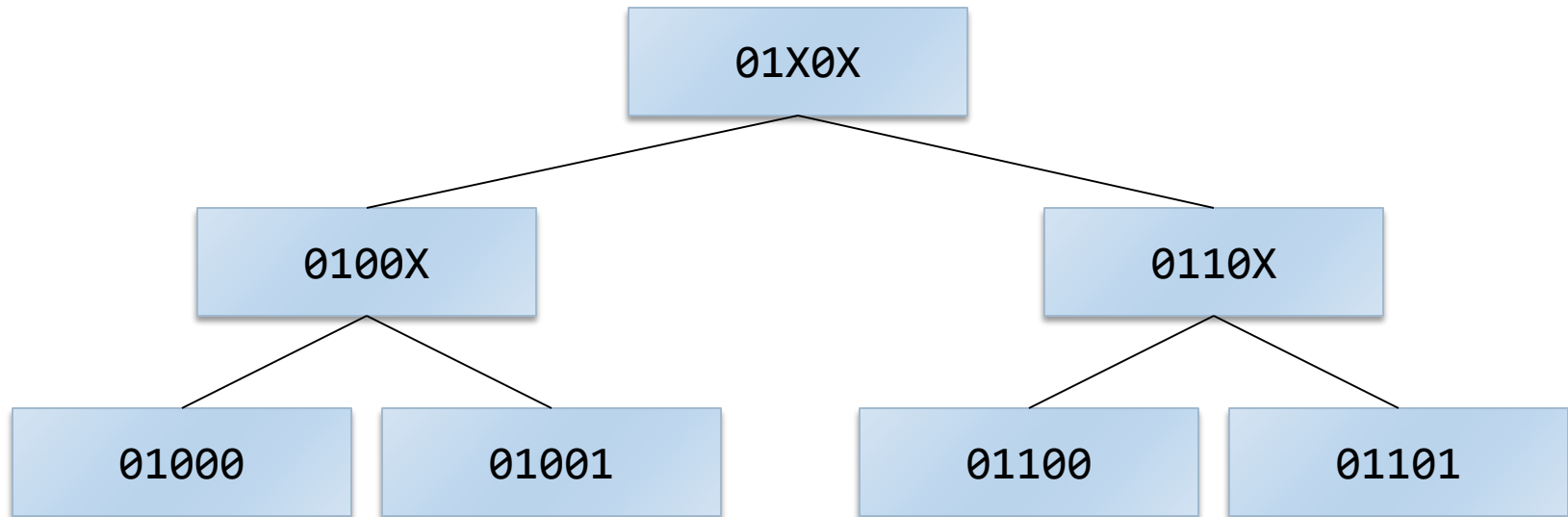
# Solution

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- ▶ We may devise a recursive algorithm that explores the complete 'tree' of possible compatible combinations:
  - ▶ Transforming each  $X$  into a  $\emptyset$ , and then into a  $1$
  - ▶ For each transformation, we recursively seek other  $X$  in the string
- ▶ The number of final combinations (leaves of the tree) is equal to  $2^N$ , if  $N$  is the number of  $X$ .
- ▶ The tree height is  $N+1$ .

# Combinations tree

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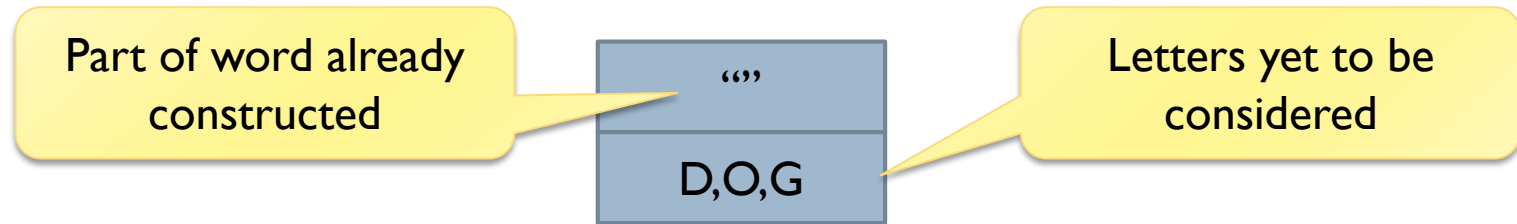
# Exercise: Anagram

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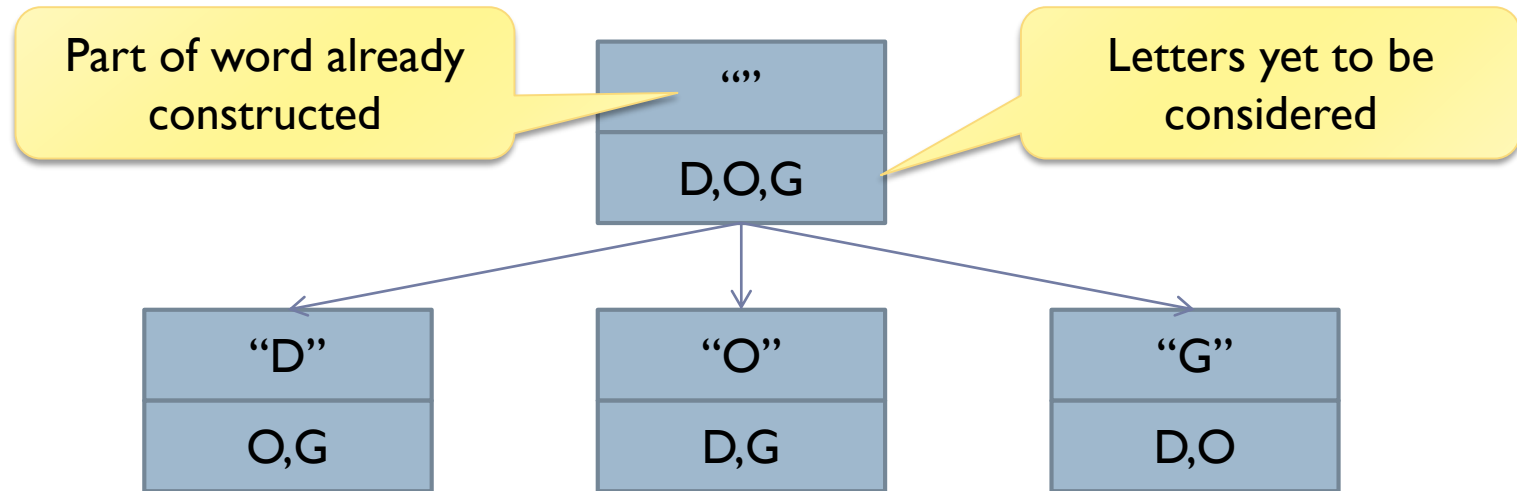
- ▶ Given a word, find all possible anagrams of that word
  - ▶ Find all permutations of the elements in a set
  - ▶ Permutations are  $N!$
- ▶ E.g.: «Dog» → dog, dgo, god, gdo, odg, ogd

# Anagrams: recursion tree

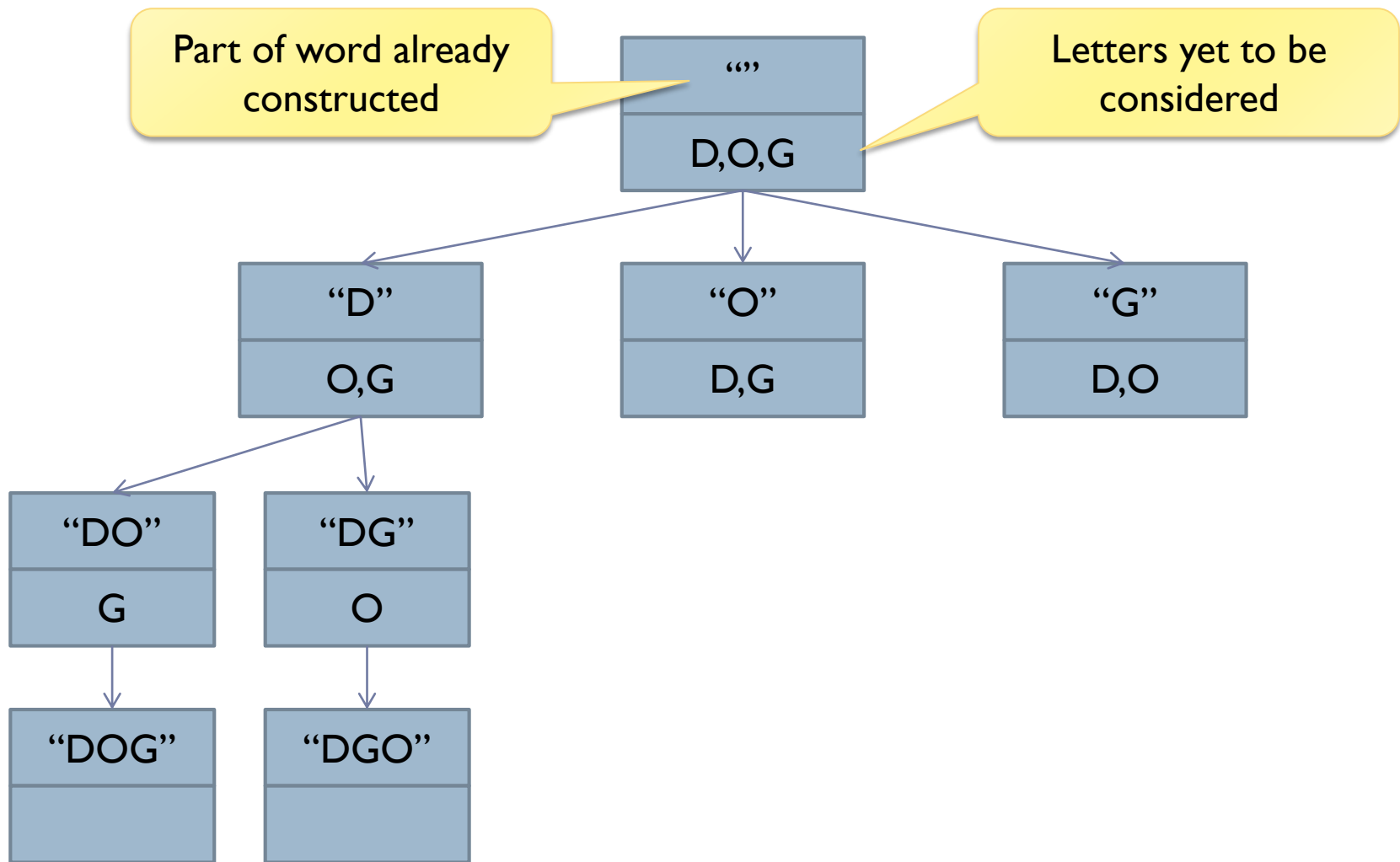
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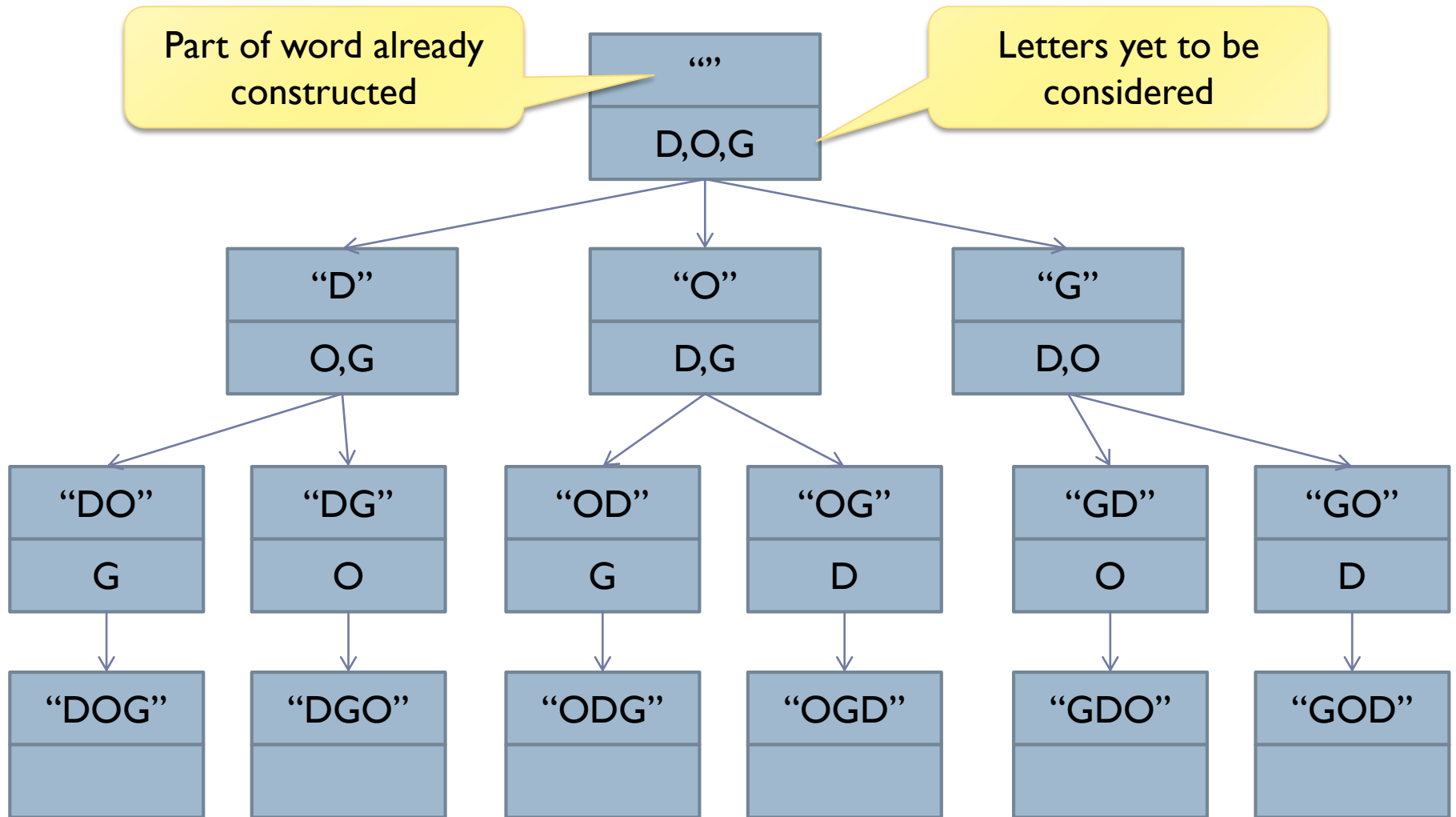
# Anagrams: recursion tree



# Anagrams: recursion tree



# Anagrams: recursion tree



# Anagrams: problem variants

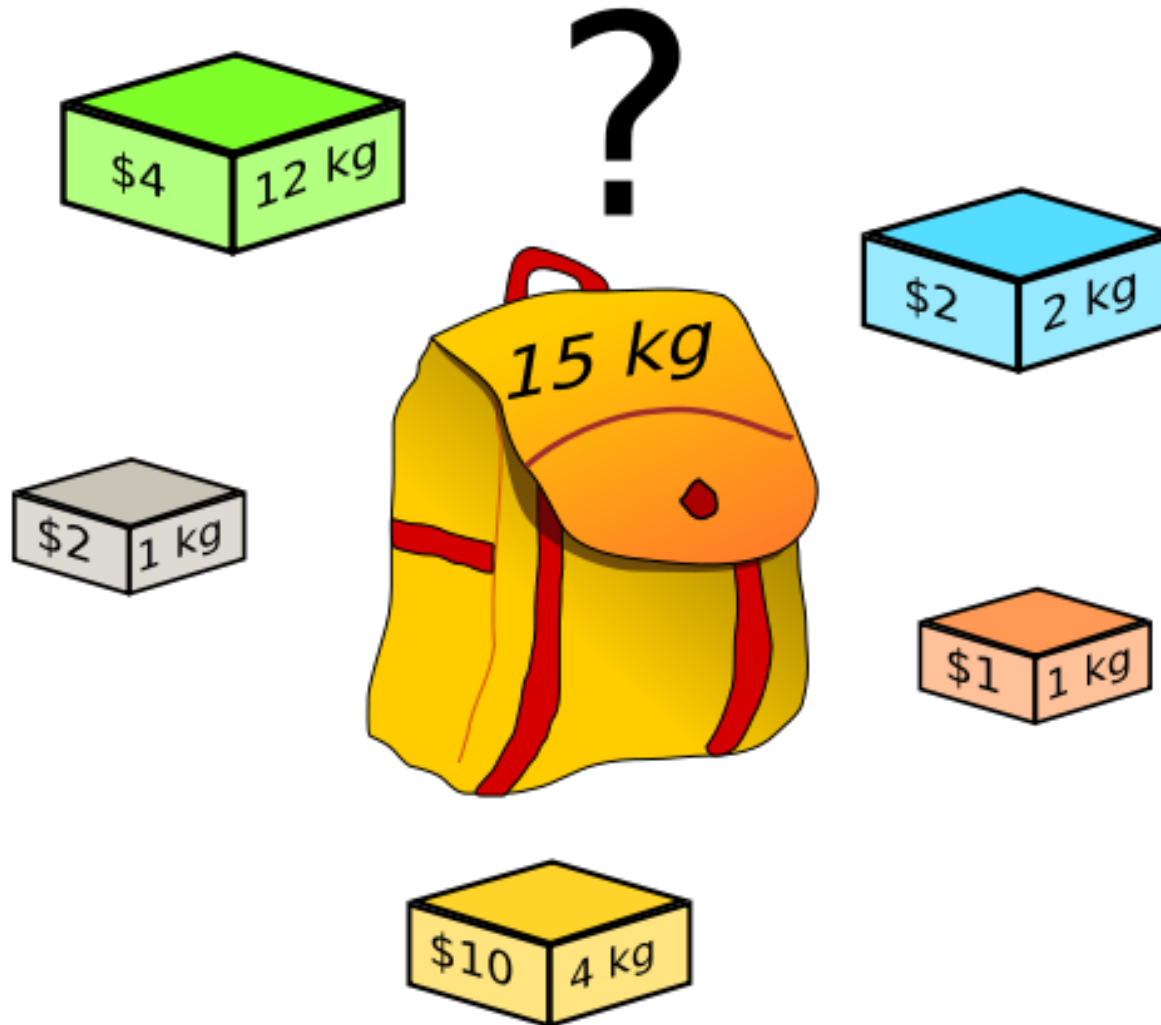
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- ▶ **Generate only anagrams that are “valid” words**
  - ▶ At the end of recursion, check the dictionary
  - ▶ During recursion, check whether the current prefix exists in the dictionary
- ▶ **Handle words with multiple letters: avoid duplicate anagrams**
  - ▶ E.g., “seas” → **s** seas and seas **s** are the same word
  - ▶ Generate all and, at the end of recursion, check if repeated
  - ▶ Constrain, during recursion, duplicate letters to always appear in the same order (e.g, **s** always before **s**)

<http://wordsmith.org/anagram/index.html>

# The Knapsack Problem

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# The Knapsack Problem

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**Input:** Weight of N items  $\{w_1, w_2, \dots, w_n\}$   
Cost of N items  $\{c_1, c_2, \dots, c_n\}$   
Knapsack limit S

**Output:** Selection for knapsack:  $\{x_1, x_2, \dots, x_n\}$   
where  $x_i \in \{0, 1\}$ .

## Sample input:

$$w_i = \{1, 1, 2, 4, 12\}$$

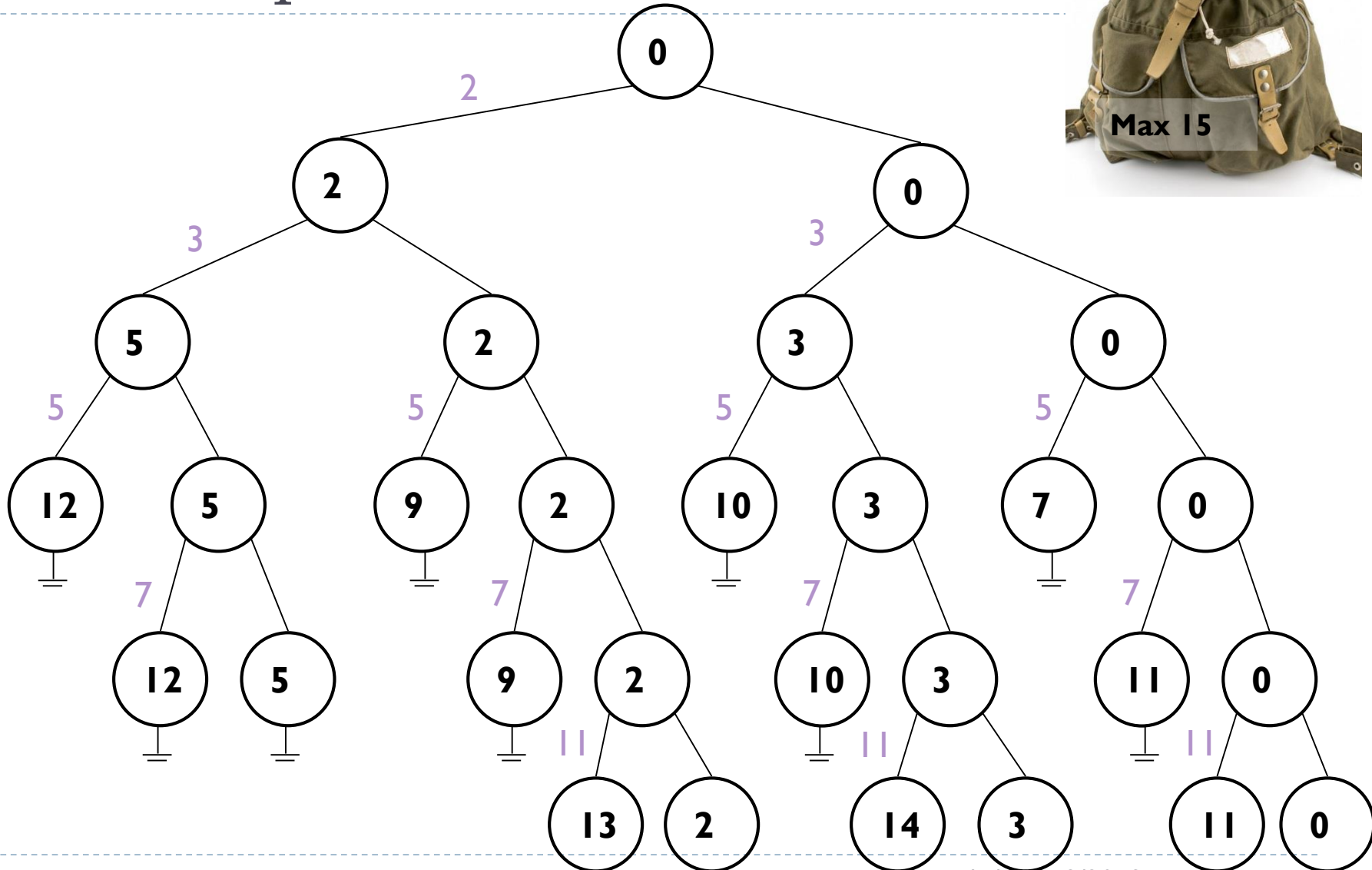
$$c_i = \{1, 2, 2, 10, 4\}$$

$$S = 15$$





# The Knapsack Problem





8	2	5	5	6	7	3	9
1	2	4	1	9	2	3	1
2	2	5	2	42	7	9	7
8	2	5	6	6	6	3	9
1	2	4	1	2	3	1	9
2	7	1	1	4	7	8	9
2	3	5	3	1	8	9	9
8	2	3	1	6	7	3	9





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



# Exercise: Binomial Coefficient

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- ▶ Compute the Binomial Coefficient  $\binom{n}{m}$  exploiting the recurrence relations (derived from Tartaglia's triangle):

$$\left\{ \begin{array}{l} \binom{n}{m} = \binom{n-1}{m-1} + \binom{n-1}{m} \\ \binom{n}{n} = \binom{n}{0} = 1 \\ 0 \leq n, \quad 0 \leq m \leq n \end{array} \right.$$

# Exercise: Determinant

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- ▶ Compute the determinant of a square matrix
- ▶ Remind that:
  - ▶  $\det(M_{1 \times 1}) = m_{1,1}$
  - ▶  $\det(M_{N \times N}) =$  sum of the products of all elements of a row (or column), times the determinants of the  $(N-1) \times (N-1)$  submatrices obtained by deleting the row and column containing the multiplying element, with alternating signs  $(-1)^{i+j}$ .

$$\det(A) = \sum_{j=1}^n (-1)^{i+j} a_{i,j} M_{i,j} = \sum_{i=1}^n (-1)^{i+j} a_{i,j} M_{i,j}.$$

Laplace's Formula, at

<http://en.wikipedia.org/wiki/Determinant>



# Recursion and iteration

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- ▶ Every **recursive** program can **always** be implemented in an **iterative** manner
- ▶ The best solution, in terms of efficiency and code clarity, depends on the problem

# Example: Factorial (iterative)

$$\left\{ \begin{array}{l} 0! \stackrel{\text{def}}{=} 1 \\ \forall N \geq 1: \\ N! \stackrel{\text{def}}{=} N \times (N-1)! \end{array} \right.$$

```
public long iterativeFactorial(long N)
{
    long result = 1 ;

    for (long i=2; i<=N; i++)
        result = result * i ;

    return result ;
}
```



# Fibonacci (iterative)

---

```
public long iterativeFibonacci(long N) {
    if(N==0) return 0 ;
    if(N==1) return 1 ;

    // now we know that N >= 2
    long i = 2 ;
    long fib1 = 1 ; // fib(N-1)
    long fib2 = 0 ; // fib(N-1)

    while( i<=N ) {
        long fib = fib1 + fib2 ;
        fib2 = fib1 ;
        fib1 = fib ;
        i++ ;
    }

    return fib1 ;
}
```

# Dichotomic search (iterative)

```
public int findIterative(int[] v, int x) {
    int a = 0 ;
    int b = v.length-1 ;

    while( a != b ) {
        int c = (a + b) / 2; // middle point
        if (v[c] >= x) {
            // v[c] is too large -> search left
            b = c ;
        } else {
            // v[c] is too small -> search right
            a = c+1 ;
        }
    }
    if (v[a] == x)
        return a;
    else
        return -1;
}
```

# Exercises

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- ▶ Create an iterative version for the computation of the binomial coefficient  $\binom{n}{m}$ .
- ▶ Analyze a possible iterative version for computing the determinant of a matrix. What are the difficulties?
- ▶ Can you find a simple iterative solution for the X-Expansion problem? And for the Anagram problem?

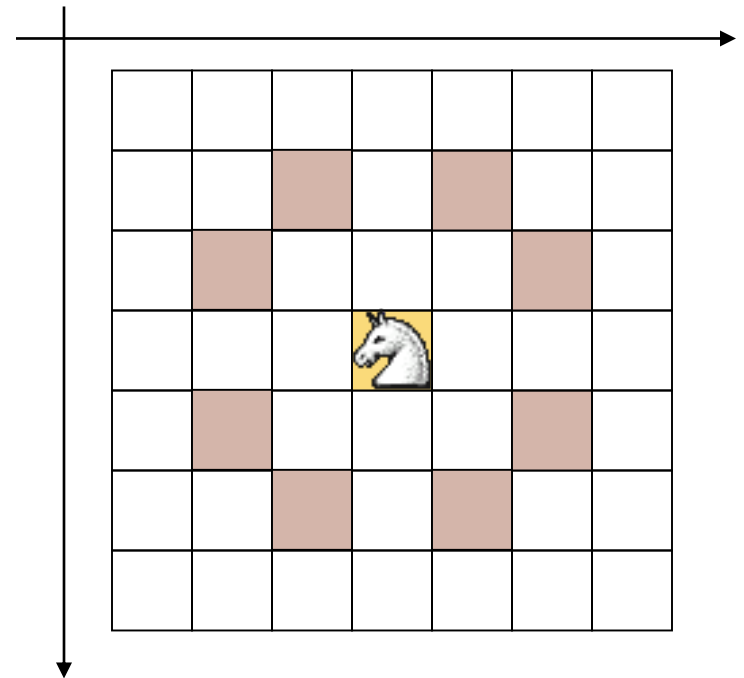


# More complex examples of recursive algorithms

Recursion

# Knight's tour

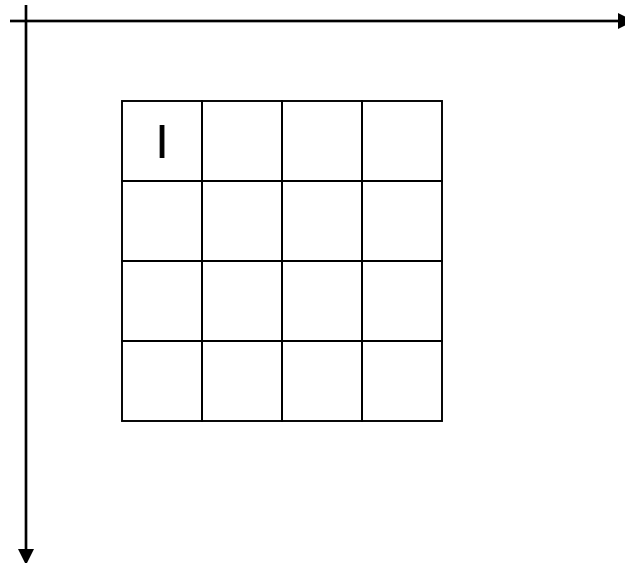
- ▶ Consider a  $N \times N$  chessboard, with the Knight moving according to Chess rules
  - ▶ The Knight may move in 8 different cells
- ▶ We want to find a **sequence** of moves for the Knight where
  - ▶ **All** cells in the chessboard are visited
  - ▶ Each cell is touched exactly **once**
- ▶ The starting point is arbitrary



# Analysis

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► Assume  $N=4$



# Move 1

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I			

Level of the next move  
to try

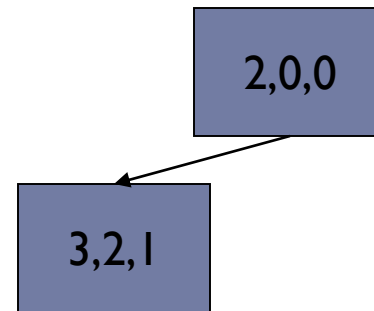
2,0,0

Coordinates of the last  
move

# Move 2

---

1			
	2		

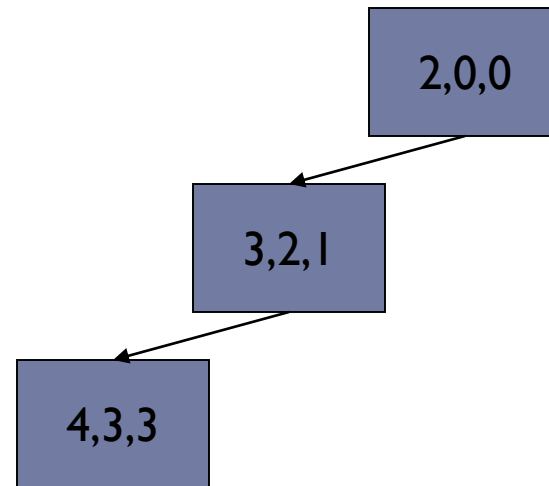




# Move 3

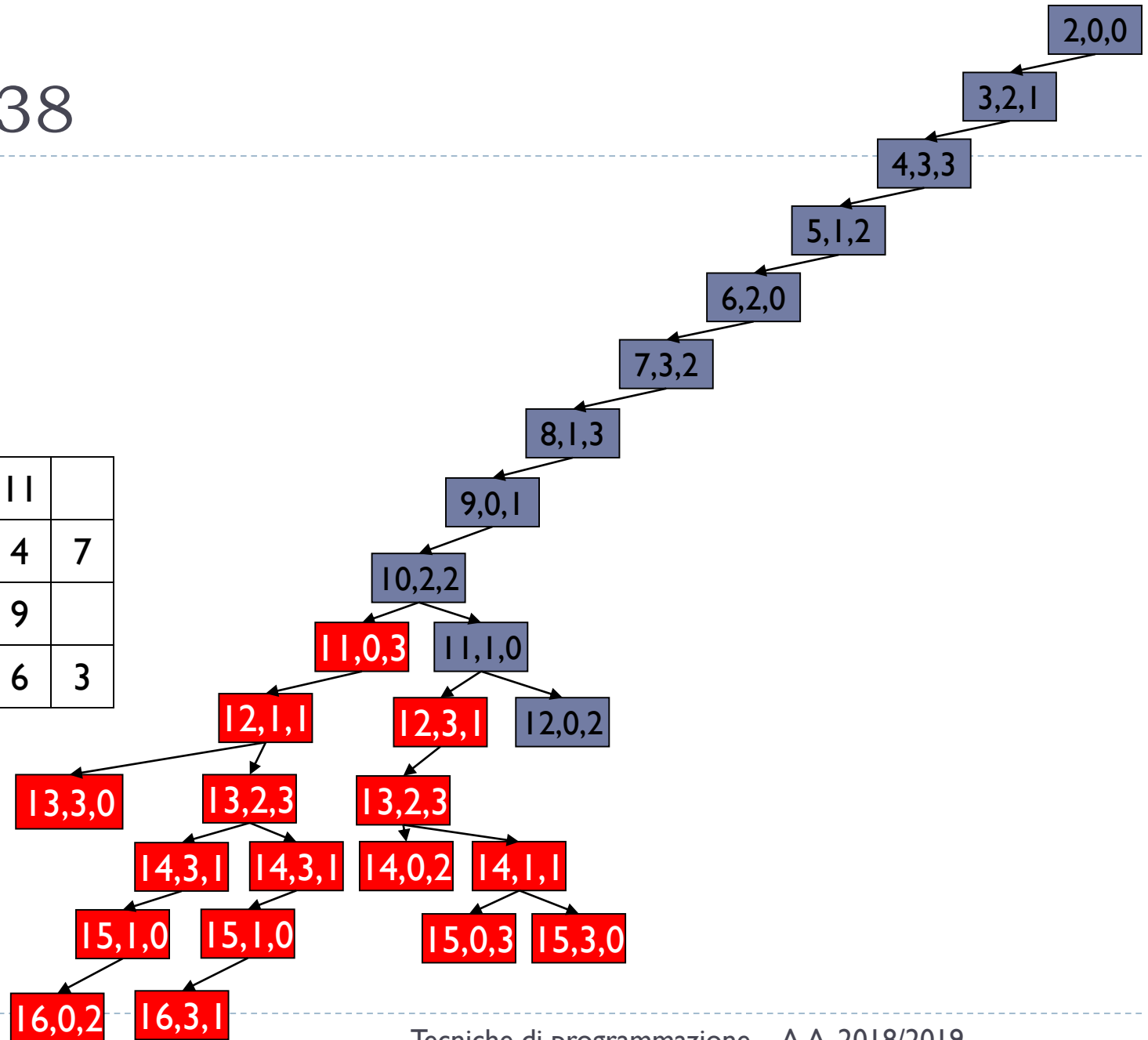
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1			
	2		
			3



# Move 38

1	8	11	
10		4	7
5	2	9	
		6	3

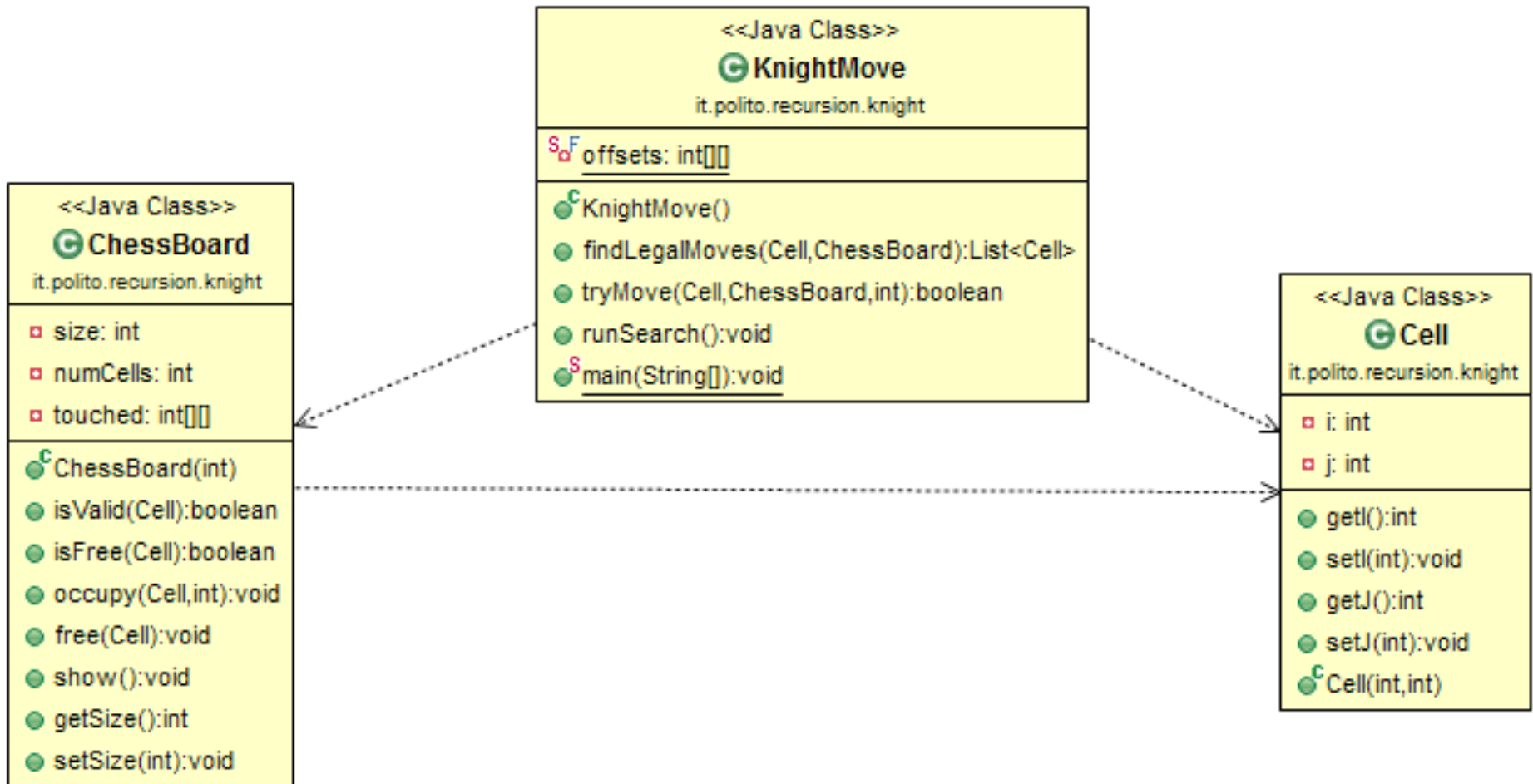


# Complexity

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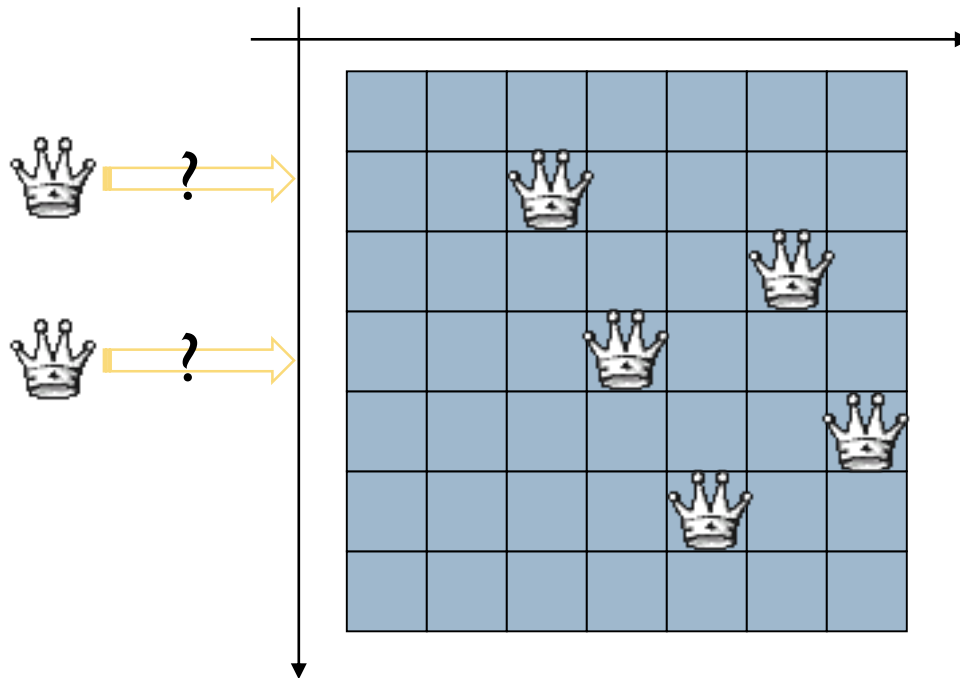
- ▶ The number of possible moves, at each step, is at most 8.
- ▶ The number of steps is  $N^2$ .
- ▶ The solution tree has a number of nodes  $\leq 8^{N^2}$ .
- ▶ In the worst case
  - ▶ The solution is in the right-most leaf of the solution tree
  - ▶ The tree is complete
- ▶ The number of recursive calls, in the worst case, is therefore  $\Theta(8^{N^2})$ .

# Implementation



# The N Queens

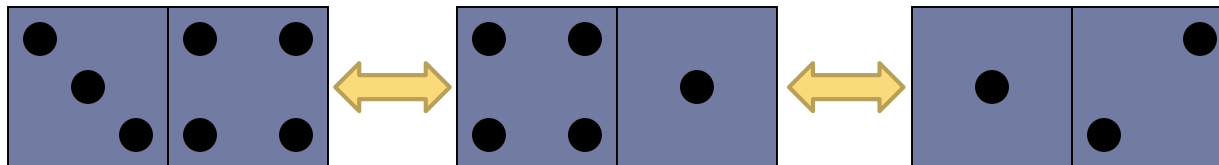
- ▶ Consider a  $N \times N$  chessboard, and  $N$  Queens that may act according to the chess rules
- ▶ Find a position for the  $N$  queens, such that no Queen is able to attack any other Queen



# Domino game

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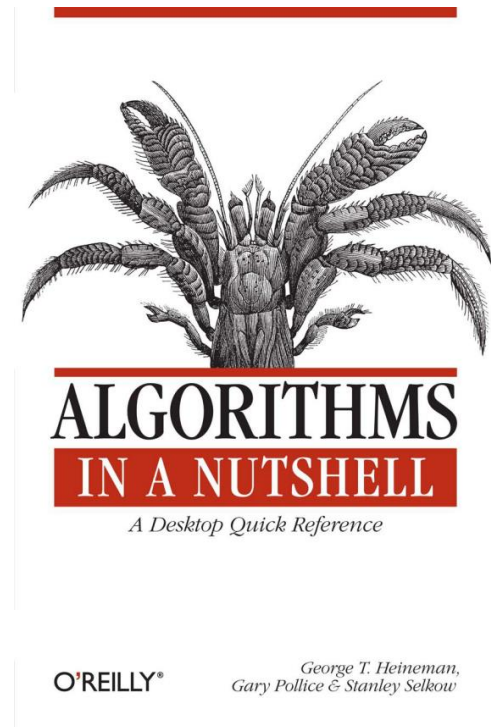
- ▶ Consider the game of Domino, composed of two-sided pieces: each side is labeled with a number from 0 to 6. All combinations of number pairs are represented exactly once.
- ▶ Find the longest possible sequence of pieces, such that consecutive pieces have the same value on the adjacent sides.



# Resources






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- ▶ Algorithms in a Nutshell, By George T. Heineman, Gary Pollice, Stanley Selkow, O'Reilly Media



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