

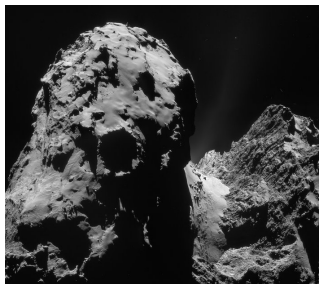
SAT: Introduction

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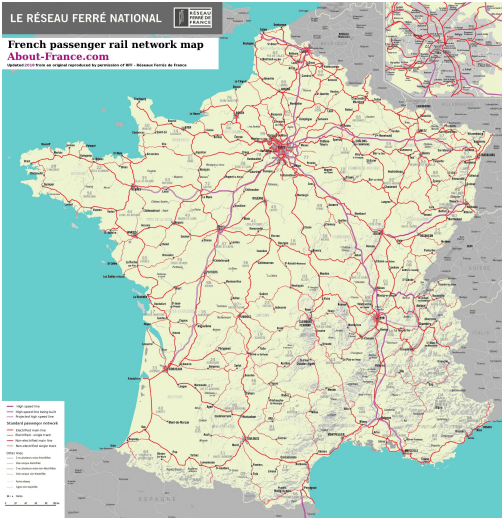
January 18, 2022

Context: Solving (Very) Hard Combinatorial Problems



<https://homepages.laas.fr/ehebrard/rosetta.html>

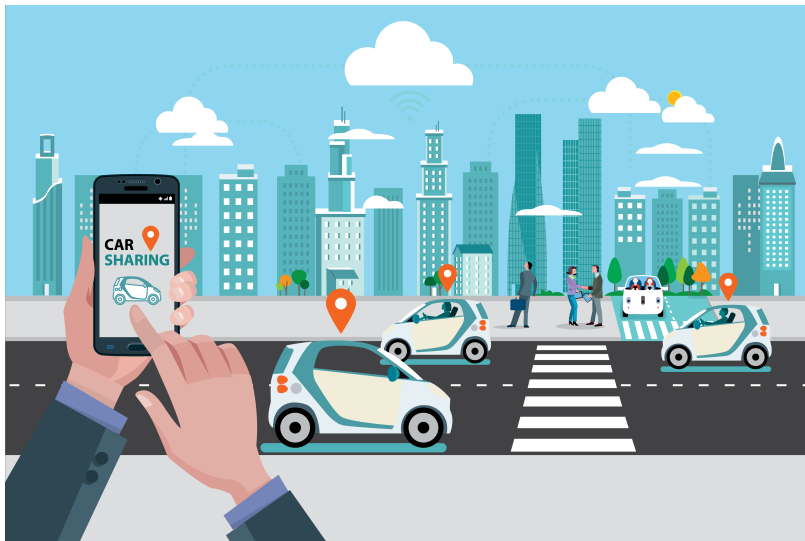
Context: Solving (Very) Hard Combinatorial Problems



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Why this Lecture?

- I noticed that most graduate students are doing software development.
- We are missing job opportunities in optimisation!
- Resources: many.. a good start would be the online course on discrete optimisation
<https://www.coursera.org/learn/discrete-optimization>

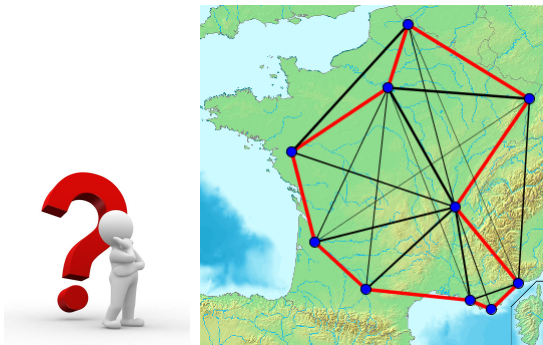
Solving Methodologies

- ① Adhoc methods
 - ① Specific exact algorithm
 - ② Heuristic method
 - ③ Meta-heuristic (genetic algorithms, ant colony, ..)
- ② Declarative Approached
 - ① (Mixed) Integer Programming,
 - ② Constraint Programming
 - ③ Boolean Satisfiability (SAT)
 - ④ ...

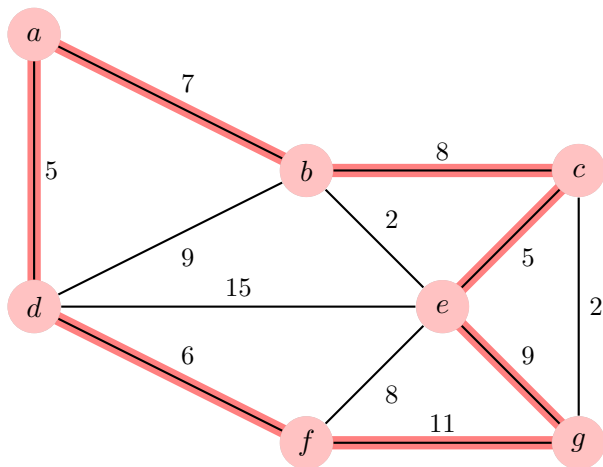
Why Declarative Approaches?

- They are problem independent! The user models the problem in a specific language and the solver do the job!
- Very active community

Travelling Salesman Problem

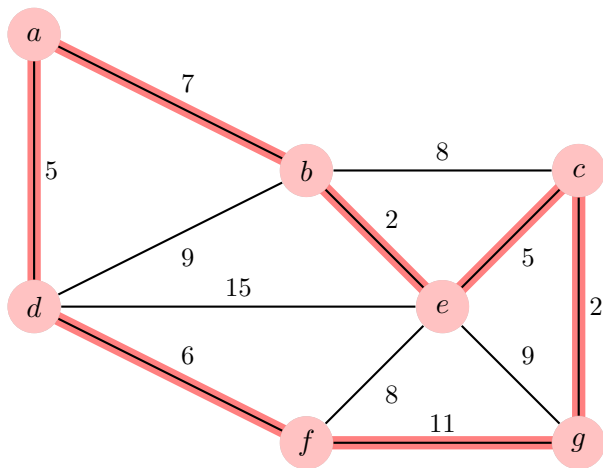


Exemple



-- \rightarrow Cost : $5 + 7 + 8 + 5 + 9 + 11 + 6 = 53Km$

Example



-- > Cost : $5 + 7 + 2 + 5 + 2 + 11 + 6 = 38Km$

What if we check all possibilities?

- 2 Cities $\rightarrow 1$
- 5 Cities $\rightarrow 24$
- 8 Cities $\rightarrow 4032$
- 40 Cities $\rightarrow 2.10^{46}$ (with a modern machine: 3.10^{27} years!)
- 95 Cities, if we use a Plack (the shortest possible time interval that can be measured) processor and fill the universe with a processor per mm^3 , we need $3 \times$ the age of the universe

The problem is inherently hard. However, the Concorde algorithm can solve instances up to 86 000 cities!

A step back: Problems, Instances, and Algorithms

- A **problem** is a question that associates an input of an output
- Many **instances** (instantiation of the input) for the same problem
- Many **algorithms** (methodologies) to solve the same problem

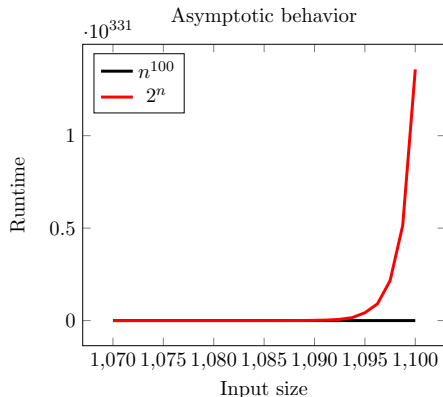
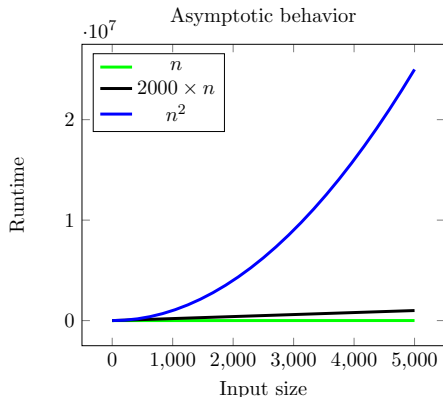
Example: The Sorting Integers problem

- Problem: sort a given sequence of n integers.
- Instance: a sequence of n integers
- A simple algorithm:
 - Scan the list to look for the smallest element
 - Swap it with the first position
 - Repeat for the list of remaining elements
- Example with the instance : 9, 3, 8, 7, 2
 - 2, 9, 3, 8, 7
 - 2, 3, 9, 8, 7
 - 2, 3, 7, 9, 8
 - 2, 3, 7, 8, 9
 - 2, 3, 7, 8, 9

Complexity

- Complexity: a measure to analyze/classify algorithms based on the amount of resource required (Time and Memory)
- Time Complexity: number of operations as a function of the size of the input
- Space Complexity: memory occupied by the algorithm as a function of the size of the input
- The evaluation is made usually by reasoning about the worst case.
- The analysis is given with regard with the asymptotic behaviour

Asymptotic behaviour



- If f is a polynomial and g is exponential then $f \in O(g)$.
For instance $n^{10000} \in O(2^n)$
- Convention:
 - Easy/Tractable Problem: We know a polynomial time algorithm to solve the problem
 - Hard/Intractable: No known polynomial algorithm
- Example: Th sorting problem is easy because we have an algorithm that runs in the worst case in $O(n^2)$ (and actually the same for memory consumption)
- What if we don't know if a problem has a polynomial time algorithm?

Classes of problems

- **P** is the class of problems that are **solvable** in polynomial time (easy problems)
- **NP** is the class of problems that are **verifiable** in polynomial time algorithm
- We know that $P \in NP$ (if you can solve then you can verify)
- For many Problems in NP , we don't know if a polynomial time algorithm exists.
- **1 Million \$** question: Is $P=NP$?

The Boolean Satisfiability Problem (SAT)

Definitions

- Atoms (Boolean variables): x_1, x_2, \dots
- Literal: $x_1, \neg x_1$
- Clauses: a clause is a disjunction of literals
- Example of clause: $(\neg x_1 \vee \neg x_4 \vee x_7)$
- Propositional formula Φ given in a **Conjunctive Normal Form** (CNF) $\Phi : c_1 \wedge \dots \wedge c_n$

Given a set of Boolean variables x_1, \dots, x_n and a CNF formulae Φ over x_1, \dots, x_n , the Boolean Satisfiability problem (SAT) is to find an assignment of the variables that satisfies all the clauses.

Why SAT?

- SAT is the first problem that is shown to be in the class NP-Complete (the hardest problems in NP)
- Many theoretical properties
- Huge practical improvements in the past 2 decades
- Is considered today as a powerful technology to solve computational problems

In this lecture, we focus on the practical side

- How to use it to solve problems (Modelling)
- Discover some efficient implementations

Example

$$x \vee \neg y \vee z$$

$$\neg x \vee \neg z$$

$$y \vee w$$

$$\neg w \vee \neg x$$

A possible solution:

$$x \leftarrow 1; y \leftarrow 1; z \leftarrow 0; w \leftarrow 0$$