

INRAE



université
PARIS-SACLAY

> Discrete optimization

Alberto TONDA

*UMR 518 MIA-PS, INRAE, AgroParisTech, Université Paris-Saclay
UAR 3611, Institut des Systèmes Complexes de Paris Île-de-France*

➤ Outline

- Definition of discrete optimization
- Combinatorial optimization
- Mixed-integer linear programming
- SAT solvers (Boolean satisfiability problem)
- Evolutionary algorithms (again)

➤ Definition of discrete optimization

- Some (or all) variables values are in \mathbb{Z} or \mathbb{N}
 - It is commonly assumed that variable values can be sorted
 - If we can enumerate all candidates, **combinatorial optimization**

- Example: mixed integer sphere

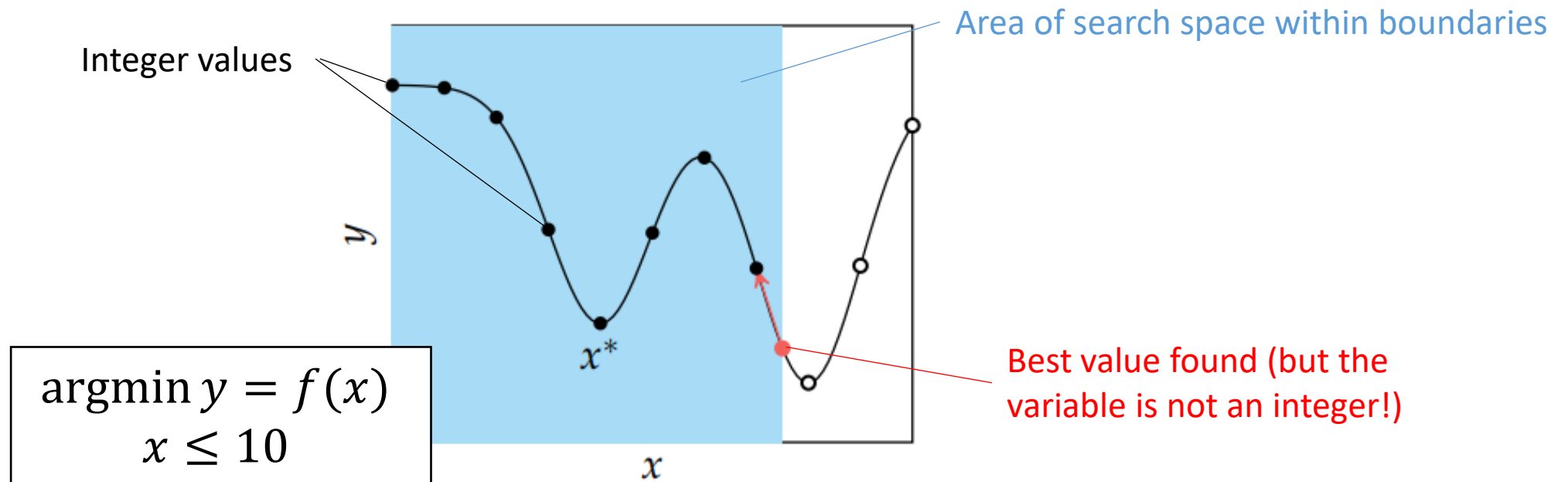
$$f(x_1, x_2) = x_1^2 + x_2^2 \text{ where } x_1 \in \mathbb{R}, x_2 \in \mathbb{Z}$$

➤ Definition of discrete optimization

- Variable values could also be **categorical**
 - E.g. $x = \{\text{red, green, blue}\}$; there is no default way of sorting
 - Some algorithms can manage them natively
 - Others encode them as **bitstrings** with constraints
 - In ML sometimes called “one-hot encoding”
- Example: $x \rightarrow x_r, x_g, x_b \in \{0,1\}; x_r + x_g + x_b = 1$

➤ Discrete optimization

- Trivial (and *ineffective*) solution
 - “Relax” the problem into a continuous one
 - Round up variable values of best solution to the closest integer



➤ Branch and bound

- If you can enumerate all possible solutions of a problem
- Exhaustive approach: evaluate them all, find global optimum
- Branch & Bound guarantees finding the global optimum
 - Without exploring everything, creates a tree
 - Branch: divide search space into partitions with extra constraints
 - Bound: compute lower bound for a partition, linear programming
 - Pruning: decide what branches are not worthy of being explored

➤ Branch and bound

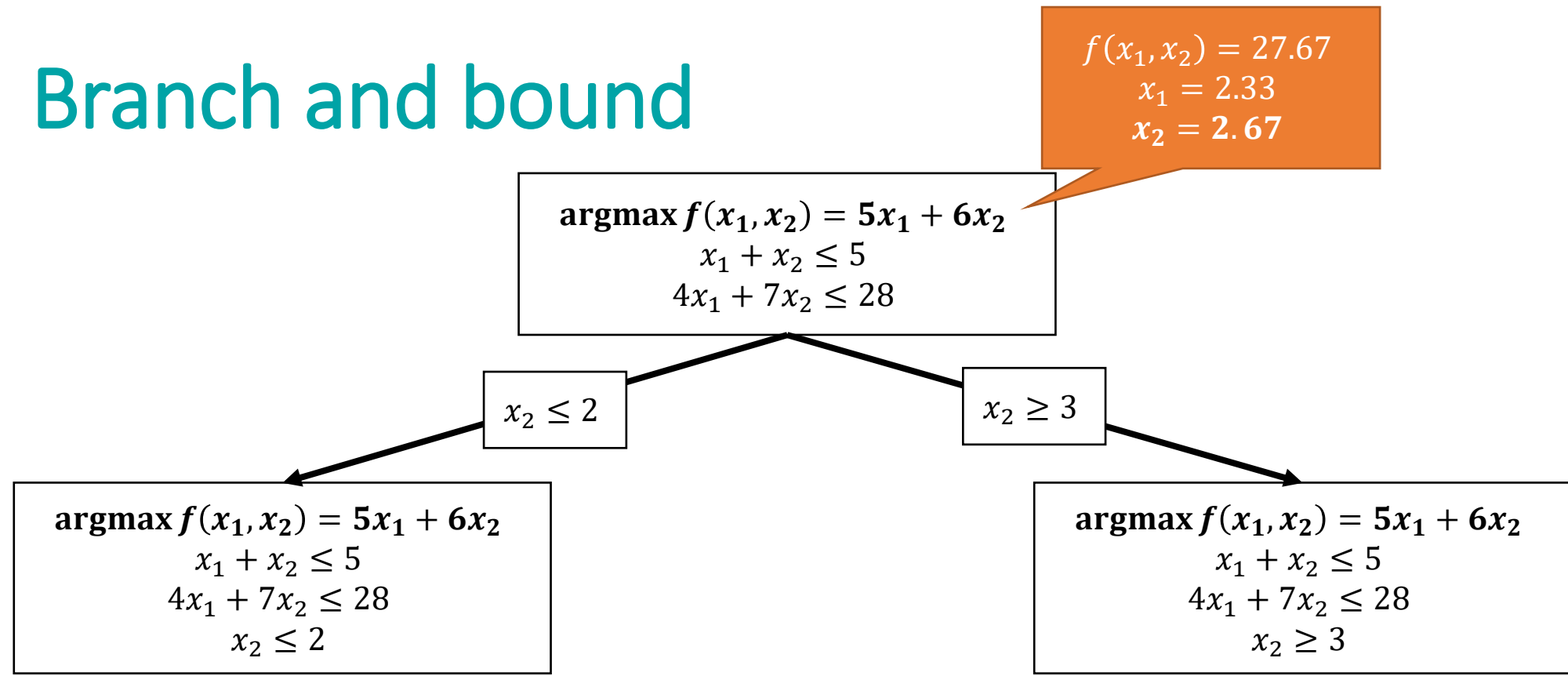
$$\begin{aligned}\text{argmax } f(x_1, x_2) &= 5x_1 + 6x_2 \\ x_1 + x_2 &\leq 5 \\ 4x_1 + 7x_2 &\leq 28 \\ x_1, x_2 &\geq 0 \\ x_1, x_2 &\in \mathbb{N}\end{aligned}$$

➤ Branch and bound

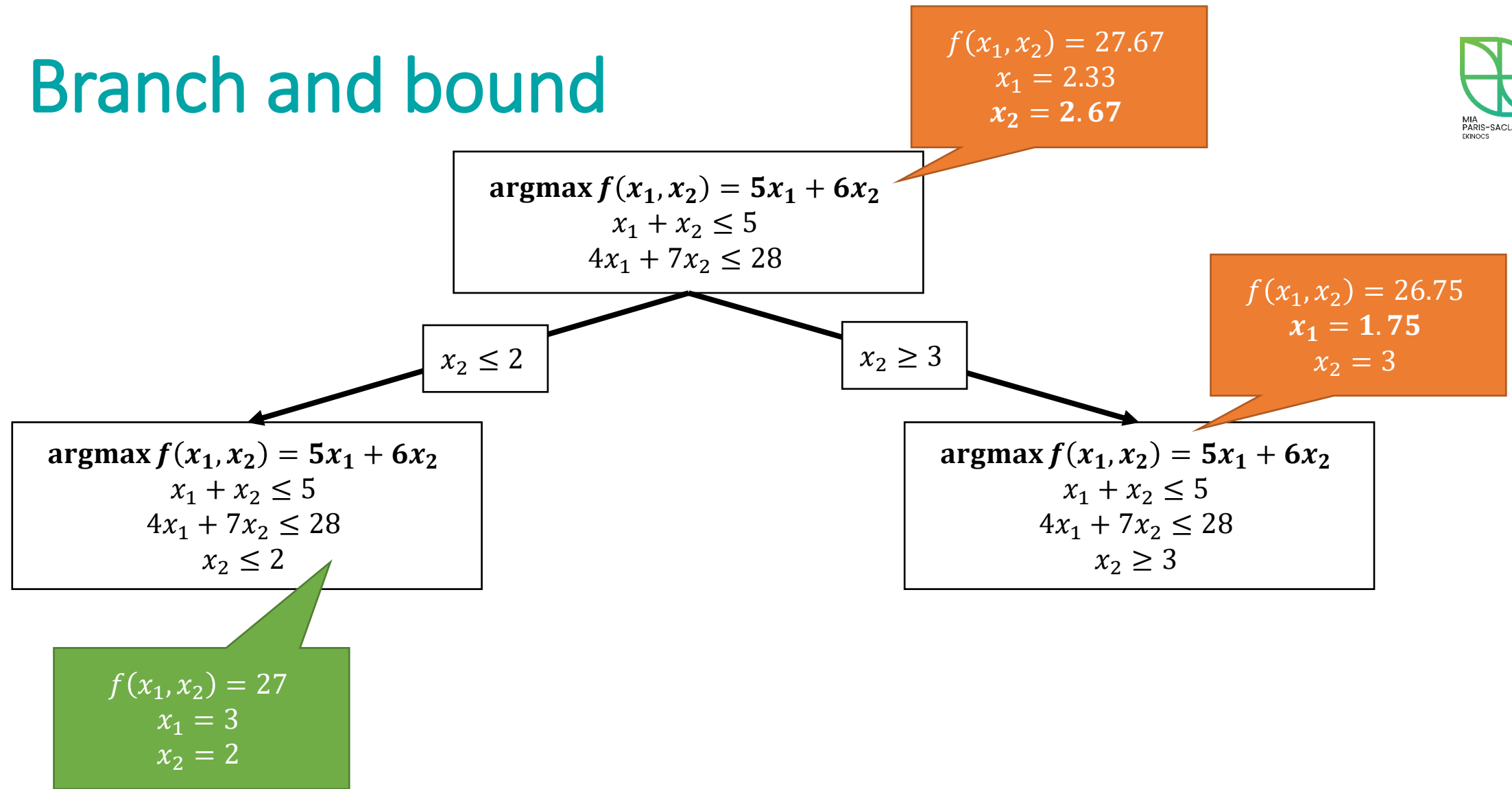
$$\begin{aligned}f(x_1, x_2) &= 27.67 \\x_1 &= 2.33 \\x_2 &= 2.67\end{aligned}$$

$$\begin{aligned}\text{argmax } f(x_1, x_2) &= 5x_1 + 6x_2 \\x_1 + x_2 &\leq 5 \\4x_1 + 7x_2 &\leq 28 \\x_1, x_2 &\geq 0 \\x_1, x_2 &\in \mathbb{N}\end{aligned}$$

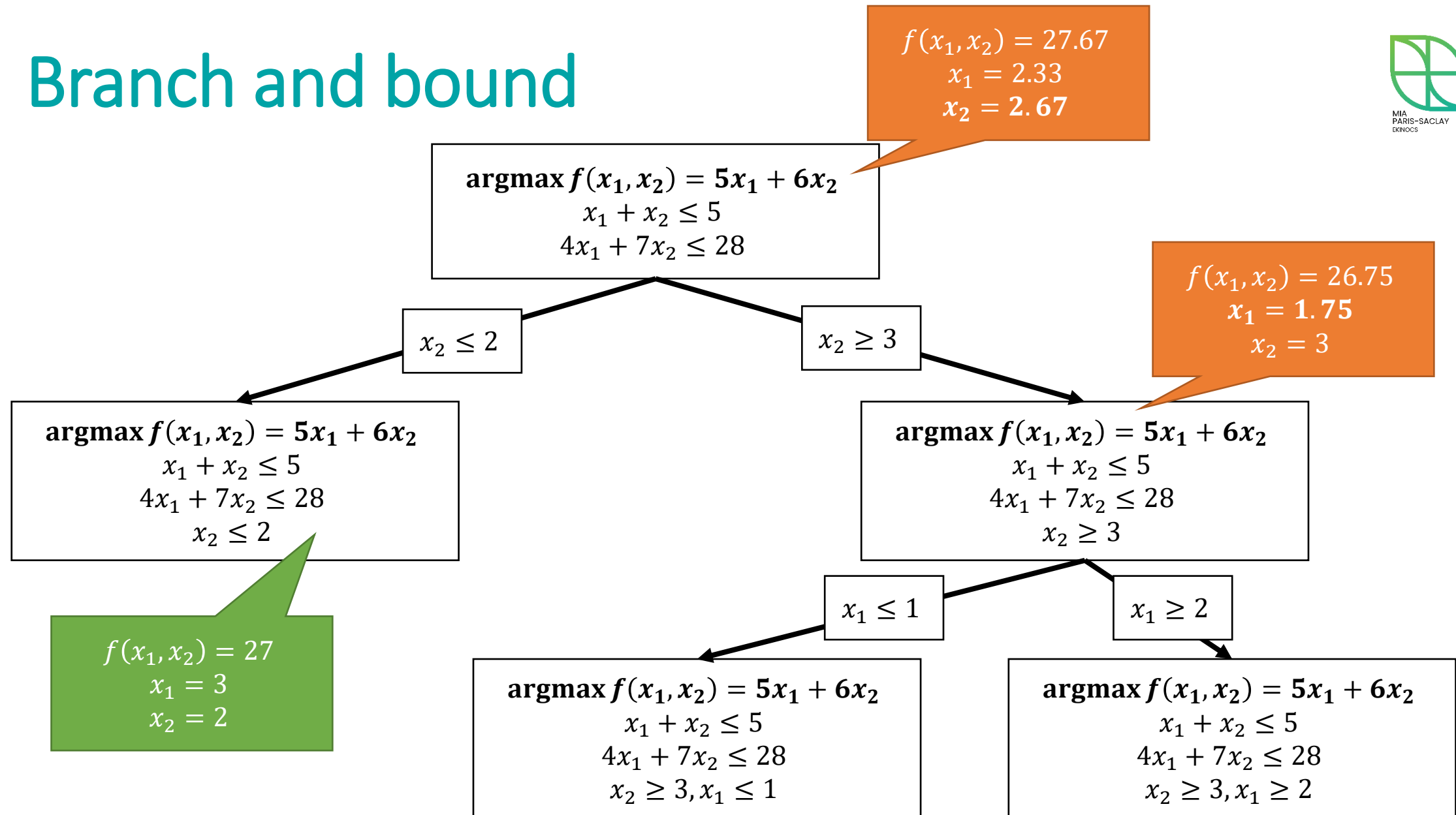
➤ Branch and bound



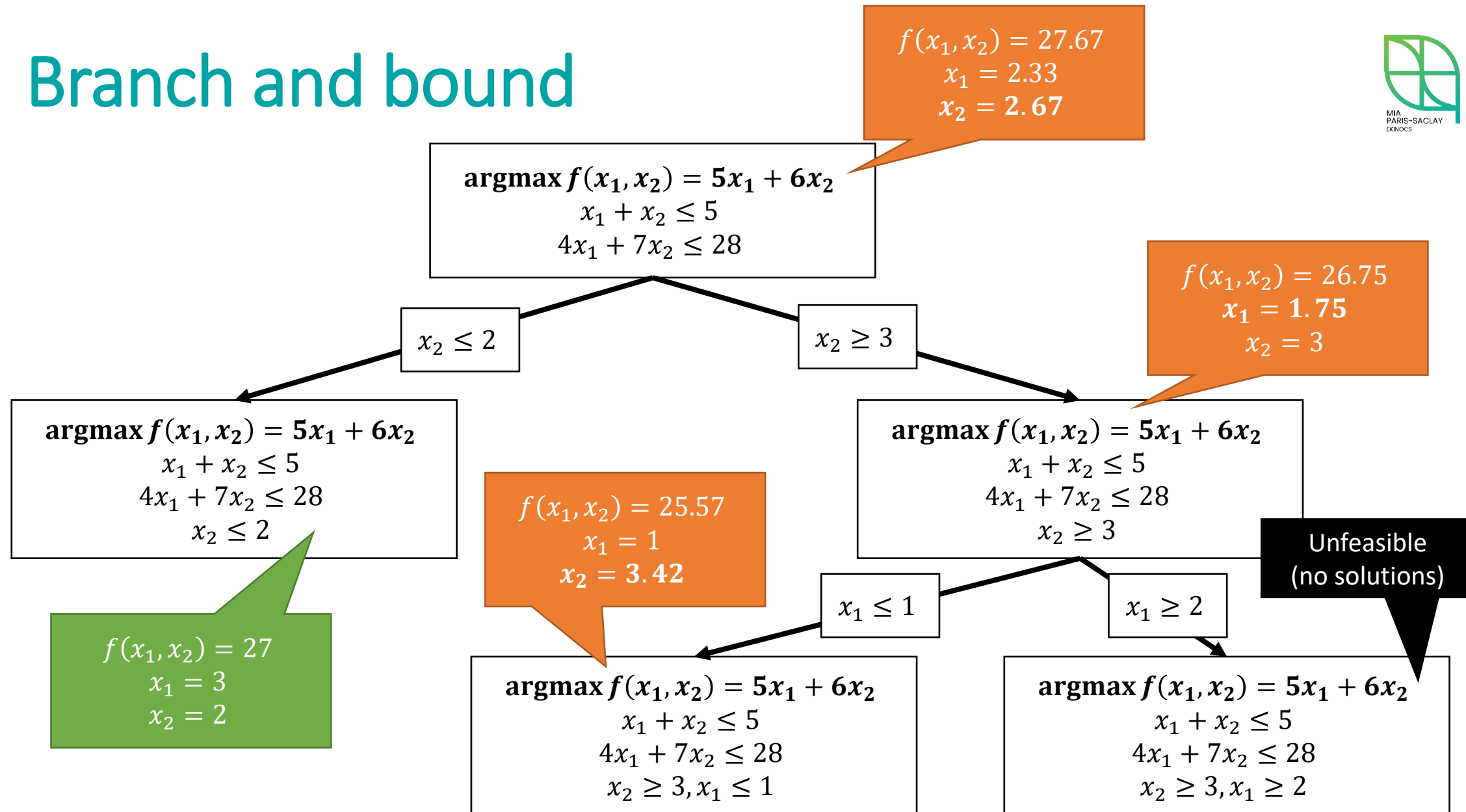
➤ Branch and bound



➤ Branch and bound

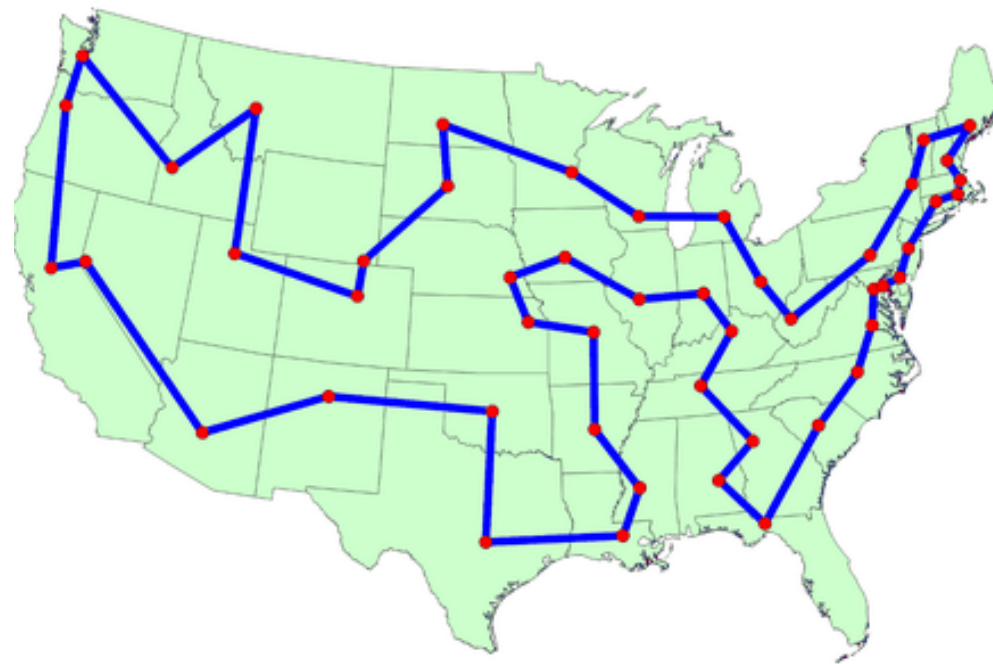


➤ Branch and bound



➤ Traveling salesman problem

- Salesman has to visit multiple cities once, in any order
- What is the optimal order, to minimize total time?



➤ Traveling salesman problem

- Combinatorial: compute total number of possible paths

$$n_p = \frac{(n_c - 1)!}{2}$$

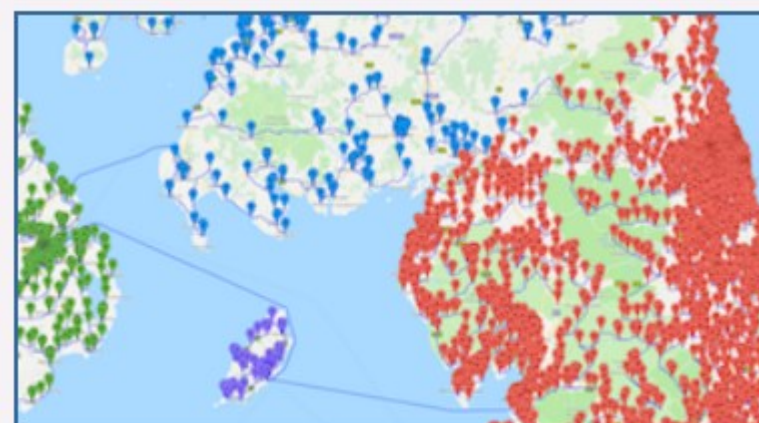
- With n_c being the number of cities
 - For 30 cities, 4.42×10^{30} possible paths
 - For 100 cities, 4.67×10^{155}
 - For 1,000 cities, 2.01×10^{249}
- Estimated number of atoms in the universe: $10^{78} - 10^{82}$

➤ Traveling salesman problem

- The best algorithm is (arguably) *heuristic*
 - Mix of linear programming, evolutionary algorithms, ...
 - CONCORDE: <https://www.math.uwaterloo.ca/tsp/index.html>



Winning \$100,000 in the Amazon Last Mile Routing Research Challenge.



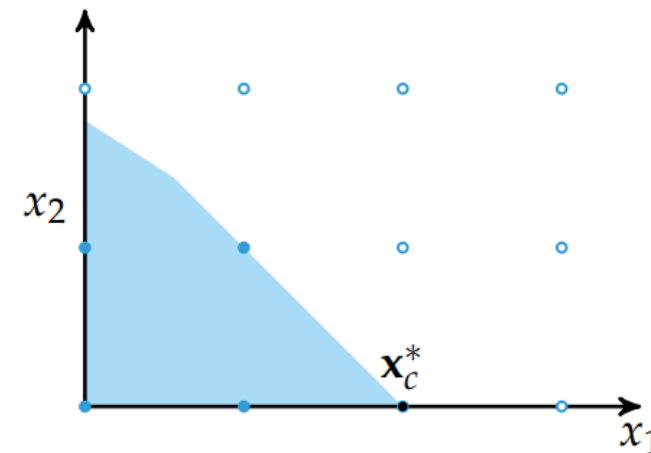
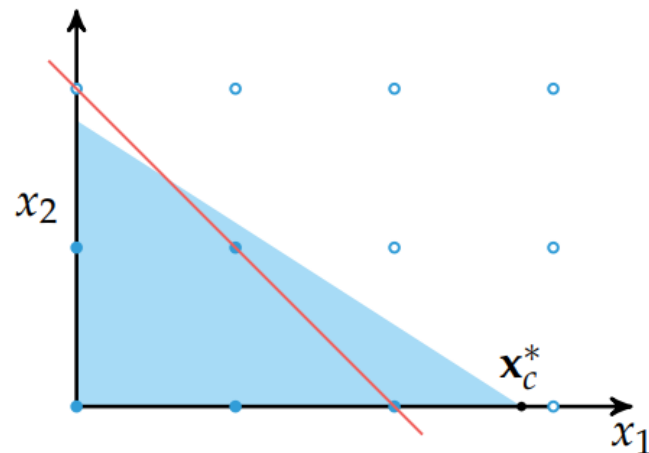
Optimal crawl to 49,687 pubs in the UK.



Visit 49,603 historic sites in the US.

➤ Mixed-integer linear programming

- MILP: Class of problems with different solvers
- Example: Branch-and-cut
 - Similar to Branch and Bound + cutting planes (multiple variables)
 - Different strategies to find best possible cutting planes



➤ Boolean satisfiability (SAT solvers)

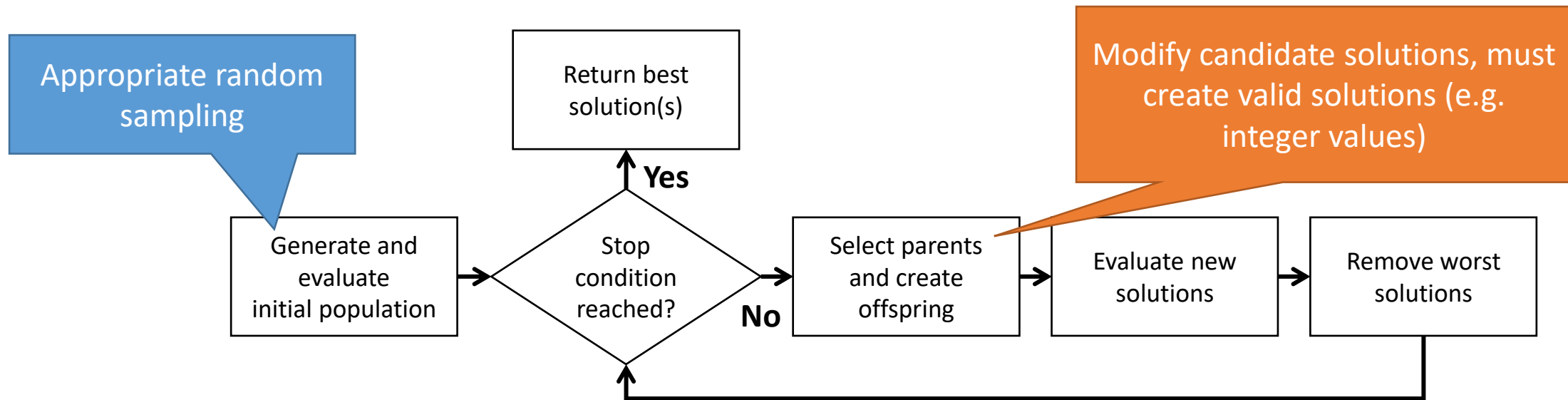
- Boolean expression (binary variables)
- Existence of a candidate solution that outputs *true*?

$$f(\mathbf{x}) = x_1 \wedge (x_2 \vee \neg x_3) \wedge (\neg x_1 \vee \neg x_2)$$

- Complex problem
- Specialized techniques called SAT solvers

➤ Evolutionary algorithms (again)

- Internal representation of a candidate solution mixed integer
 - Requires problem-specific variators (int for int parameters, ...)
 - But in general, it's relatively straightforward
 - No guarantee to find the global optimum



INRAE



université
PARIS-SACLAY

➤ Questions?

Bibliography

- Kochenderfer & Wheeler, *Algorithms for Optimization*, MIT Press, 2019
- Vanderbei, *Linear Programming: Foundations and Extensions*, 2014
- Hamano et al., *CMA-ES with Margin: Lower-Bounding Marginal Probability for Mixed-Integer Black-Box Optimization*, 2022
- Applegate et al., *Solution of a Min-Max Vehicle Routing Problem*, 2002

Images: unless otherwise stated, I stole them from the Internet. I hope they are not copyrighted, or that their use falls under the Fair Use clause, and if not, I am sorry. Please don't sue me.