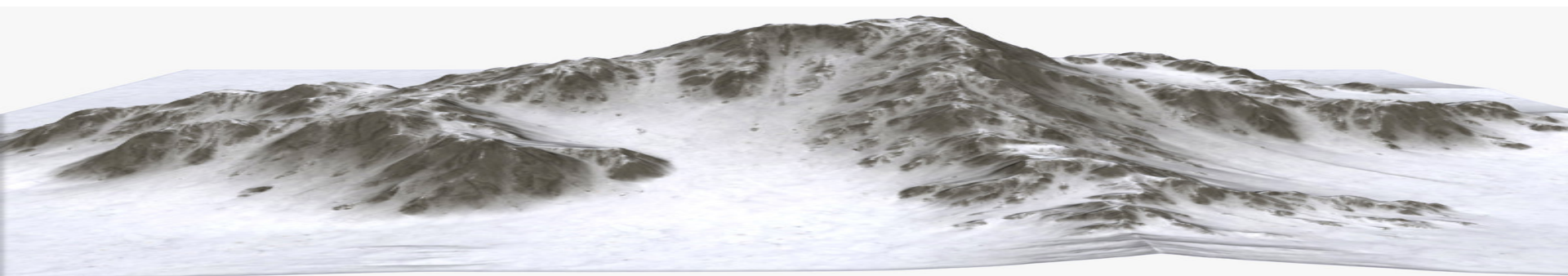




Objective Landscapes for Constraint Programming

Philippe Laborie, IBM
laborie@fr.ibm.com



Goal

- Improve performance of a **Black-Box CP Solver** (in our case IBM ILOG CP Optimizer) by exploiting some information on the relation between **decision variables** and **objective value**
- This idea is not new !
 - Impact-Based Search [Refalo-2014]
 - Activity-Based Search [Michel&Van Hentenryck-2012]
 - Linear Relaxations [Beck&Refalo-2003, Laborie&Rogerie-2016]
 - ..., ..., ...
- But we had specific requirements:
 - It should work for variables with **huge domains**, typically with continuous semantics (start/end times of activities)
 - It should scale to **large problems** with several 1000s of variables

Goal

- Typical examples of objective functions in the real-world:
 - Scheduling in semiconductor manufacturing:

$$\sum_i C_i \cdot x_i + \sum_j W_j \cdot F_j(e_j) + \sum_k U_k \cdot \max(0, d_k - D_k)^2$$

- x_i : Boolean variables (allocations of tasks to machines)
 - e_j : End time of task j , F_j : Earliness/Tardiness cost function
 - d_k : Delay variable between some particular tasks
- Resource Constrained Project Scheduling with Net Present Value:

$$\sum_p C_p \cdot \exp(-R \cdot e_p) - \sum_c C_c \cdot \exp(-R \cdot e_c) + W \cdot \max_i e_i$$

- e_p : End times of activities producing cash
- e_c : End times of activities consuming cash

The Philosophy of Objective Landscapes

“Lazy Programmers for Lazy Machines.”

- The new technique should (almost) not add any computational cost to the existing search algorithms
 - The new technique should reuse existing algorithms as much as possible
- Use existing algorithms in a different way !

Objective Landscape Definition

Context:

- A constrained-optimization problem:

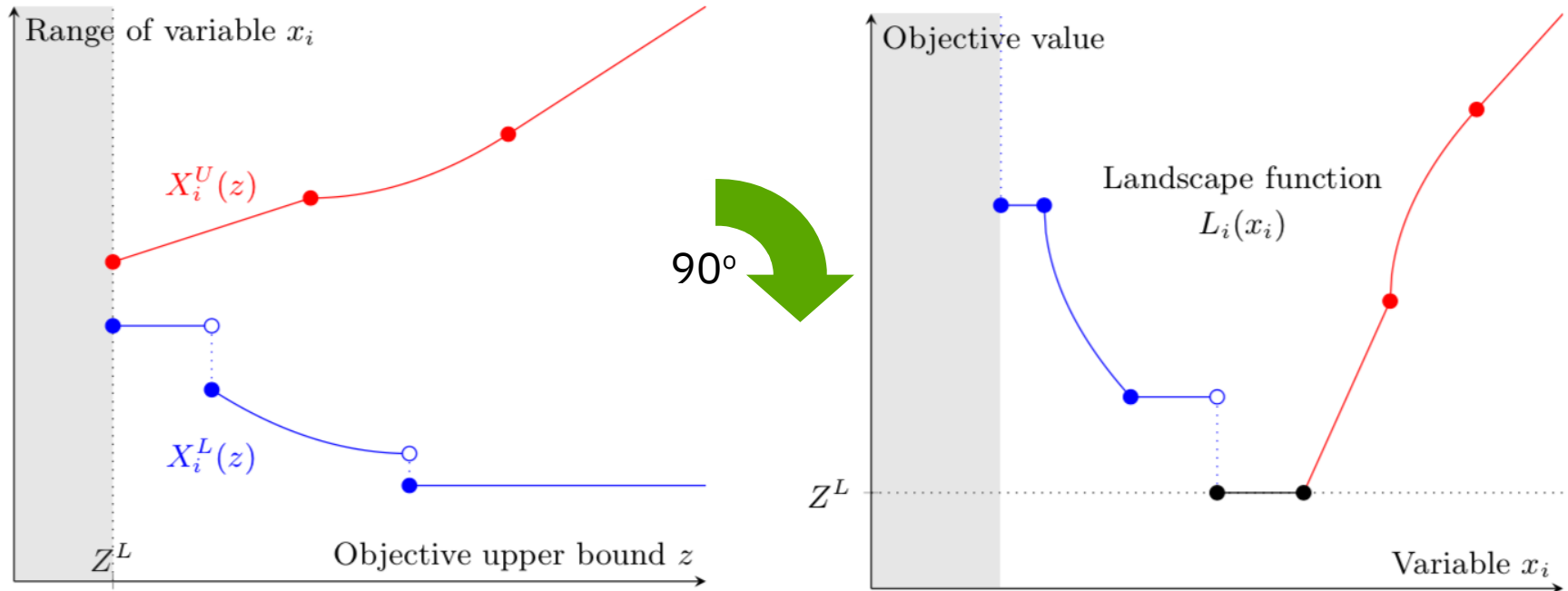
$$\begin{array}{ll}\text{minimize} & f(x_1, \dots, x_n) \\ \text{subject to} & c(x_1, \dots, x_n, y_1, \dots, y_m)\end{array}$$

$x_1, \dots, x_n, y_1, \dots, y_m$ are decision variables

- A **constraint propagation algorithm** \mathcal{P} that propagates the constraint $f(x_1, \dots, x_n) \leq z$ for a given value z
- Let $X_i^L(z)$ (resp. $X_i^U(z)$) denote the min (resp. max) value in the domain of x_i after propagation of $f(x_1, \dots, x_n) \leq z$ by \mathcal{P}

Objective Landscape Definition

- Landscape function L_i of a variable x_i



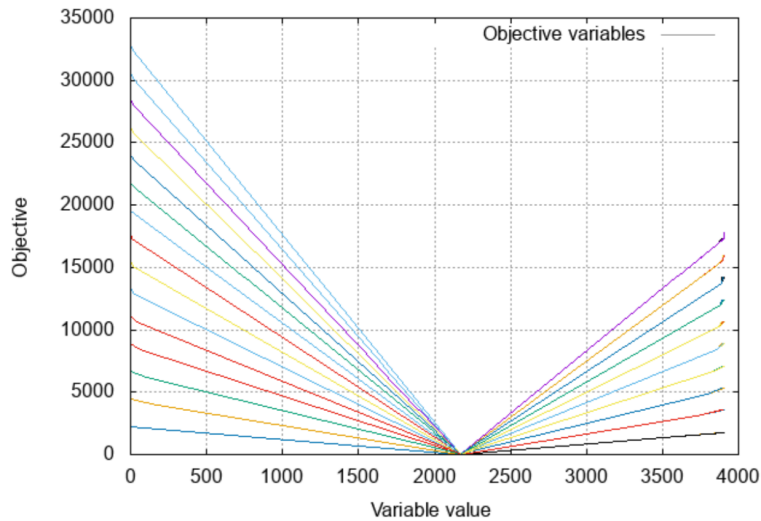
- $L_i(v)$ can be interpreted as an optimistic estimation of the objective value when $x_i=v$ (more formalization in the paper)

Objective Landscape Computation

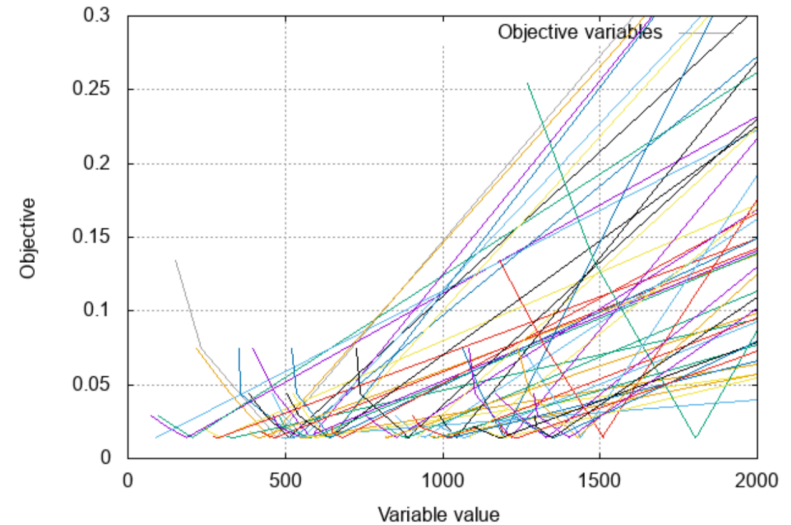
- Objective Landscape functions can be **quickly** computed **once and for all** before entering the search by sampling a discrete set of objective values z and running propagation algorithm \mathcal{P} and recording the bounds for all variables x_i
- They are represented using a **light-weight** structure (array of pairs variable value/objective value)
- Computing the landscape value $L_i(v)$ for a value v in the domain of x_i is **fast**: it can be done in $O(\log(N))$ if N is the number of samples (generally only a few tens) by simple binary search

Examples of Landscape Functions

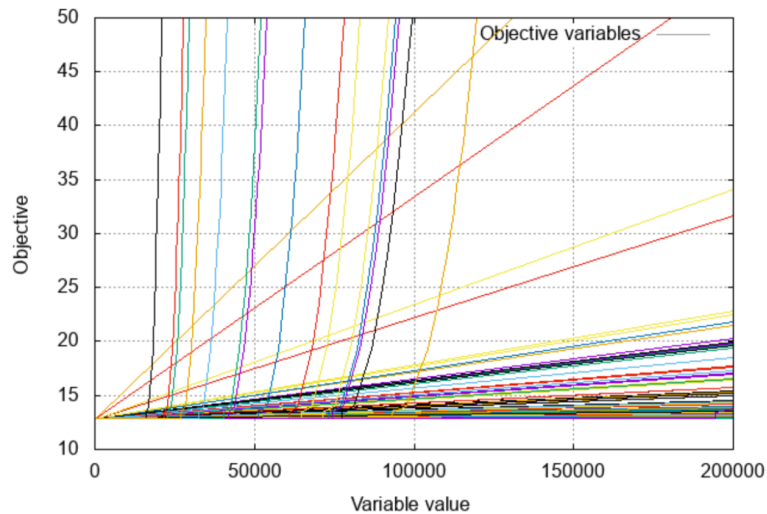
Objective landscapes for a one-machine problem with common due dates



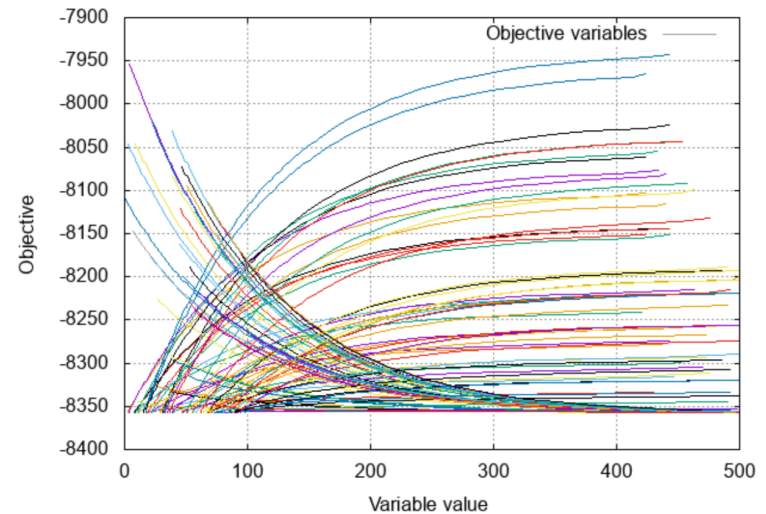
Objective landscapes for a flow-shop with earliness-tardiness costs



Objective landscapes for a photolithography scheduling problem



Objective landscapes for RCPSP with net present value

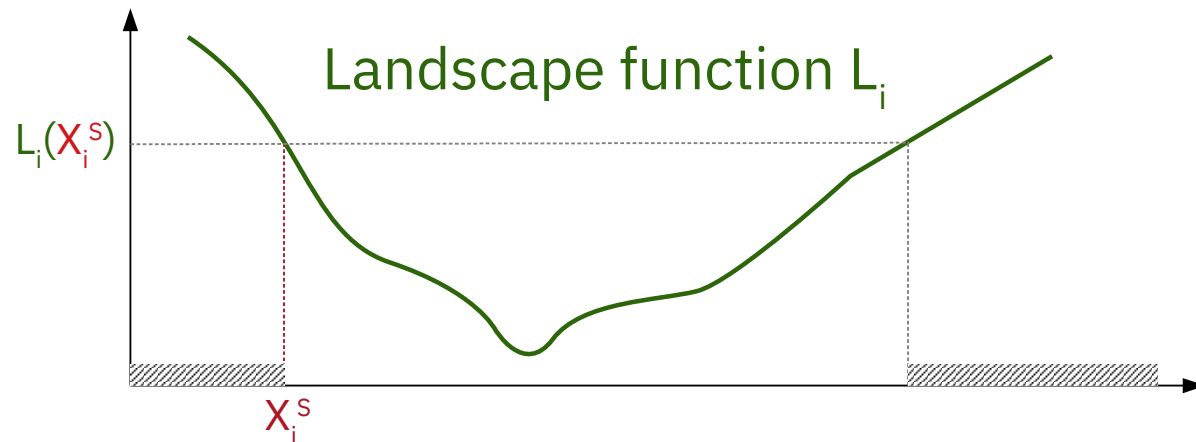


Objective Landscape Exploitation

- In CP, landscape functions can be used in many ways during the search:
 - Selection of variables
 - Selection of values
 - In a Large Neighborhood Search (LNS) context:
 - Selection of LNS fragment
 - Prevent cost degradation for some variables in LNS moves
 - Selection of variables and values in LNS completion
- We still did not fully explore all these ideas

Objective Landscape Exploitation

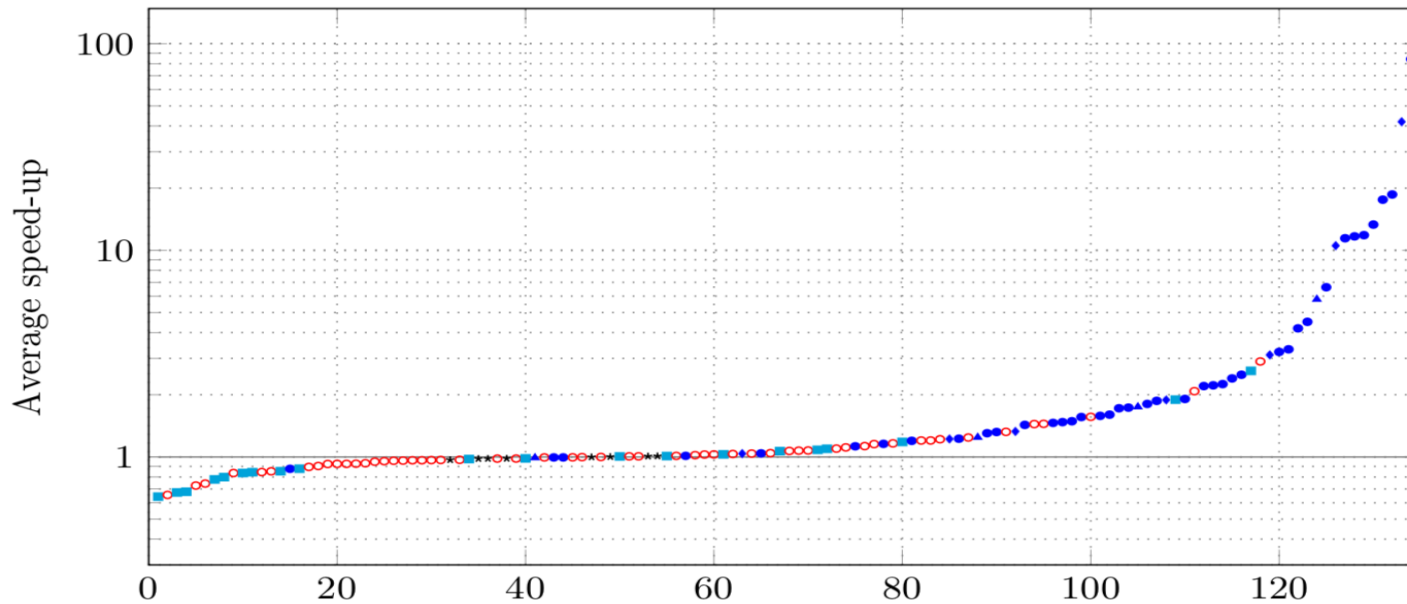
- Preventing cost degradation for some variables in LNS moves
 - At the top level of a move, we have the **current solution S** and a **relaxed model** around S that we want to re-optimize
 - For a given variable x_i , let X_i^S denote the value of x_i in solution **S**



- For a certain number of variables x_i (randomly selected) we just consider the additional constraint $L_i(x_i) \leq L_i(X_i^S)$

Experimental Results

- Average speed-up when using objective landscapes on 135 different families of scheduling problems



Aggregation type	Main variables type	Number of families	Average speed-up	Mark on Fig. 10
Max	Start/End	55	1.05	○
Sum	Start/End	40	2.62	●
Sum	Presence	29	0.98	■
Sum	Length	7	3.31	◆
Sum	Height	4	1.88	▲

Conclusion and Future Work

- Implemented in automated search of IBM ILOG CP Optimizer V12.7.1/12.8
- Average speed-up x1.5 between 12.7 and 12.8
- Up to two orders of magnitude speed-up for some problems with objective expressed as a weighted sum of variables with continuous semantics (very common in real problems)
- Future work:
 - Investigate other forms of landscape exploitation in the search
 - Objective landscapes can be generalized to store holes in the domains
 - Constraints of the problem could be exploited too