

Temporal Linear Relaxation in IBM ILOG CP Optimizer

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Rationale

- We present a loose coupling between a linear relaxation and a CP engine Large Neighborhood Iteration.
- The relaxation exploits the dominant structures of the problem to provide well informed guidelines for the completion strategy.
- We apply this idea for having a robust automatic search for scheduling problems involving irregular costs
 - tardiness cost, non convex temporal costs, non execution cost, execution mode cost, or length and workload costs

IBM ILOG CP Optimizer

- A constraint programming library
 - integer variable and numerical expressions
 - Algebraic language and global logical constraints
 - Extension interface by encoding your own constraints and search traversal solution generation
- + Automatic search and optimization
- + Specific algebraic modeling language for temporal problems

Temporal Model in CPO

- Based on the notion of **interval** as a temporal segment

$$[s, e) \text{ } s \leq e \text{ integers, } l = e - s$$

- **Interval variables**

- Decisions
 - Is the interval present in the solution
 - Start and end points of a present interval

- Domain

$$\text{dom}(a) \subseteq \{\perp\} \cup \{ [s,e) \mid s,e \in \mathbb{Z}, s \leq e \}$$

- Ranges

- start, end, and length integer ranges (notation s, e, l), presence Boolean range (notation x)

- Presence Constraint

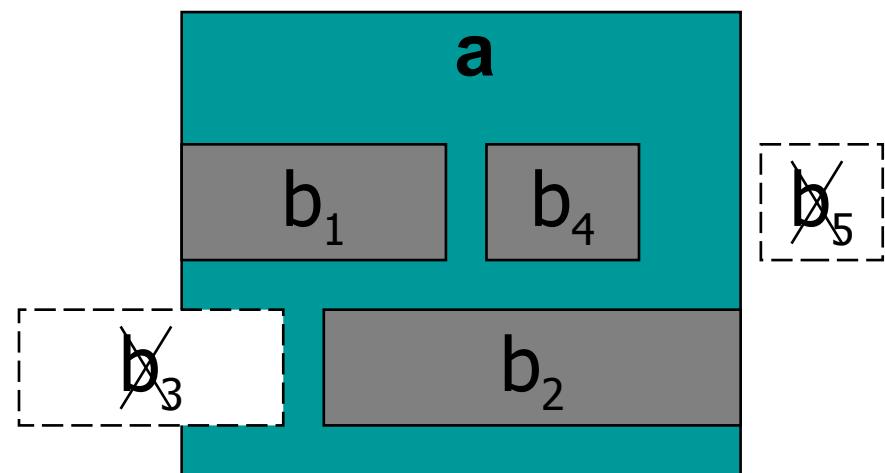
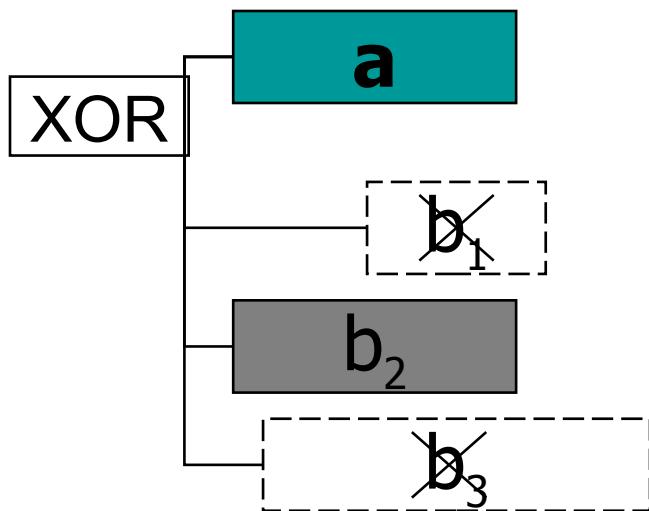
$$\text{presenceOf}(a) \text{ states } \perp \text{ not in } \text{dom}(a)$$

Precedence and Implication graphs

- All precedences between interval variables are collected in a precedence graph
 - a ends before start of b means:
if a and b are present then $s(a) \leq e(b)$
- All implications between interval variables are collected in an implication graph
 - The constraint $\text{presenceOf}(a) \leq \text{presenceOf}(b)$ adds the arc **(a implies b)**

Breakdown structure

- The breakdown structure of a task is stated thanks to **alternative** and **span** constraints on interval variables.

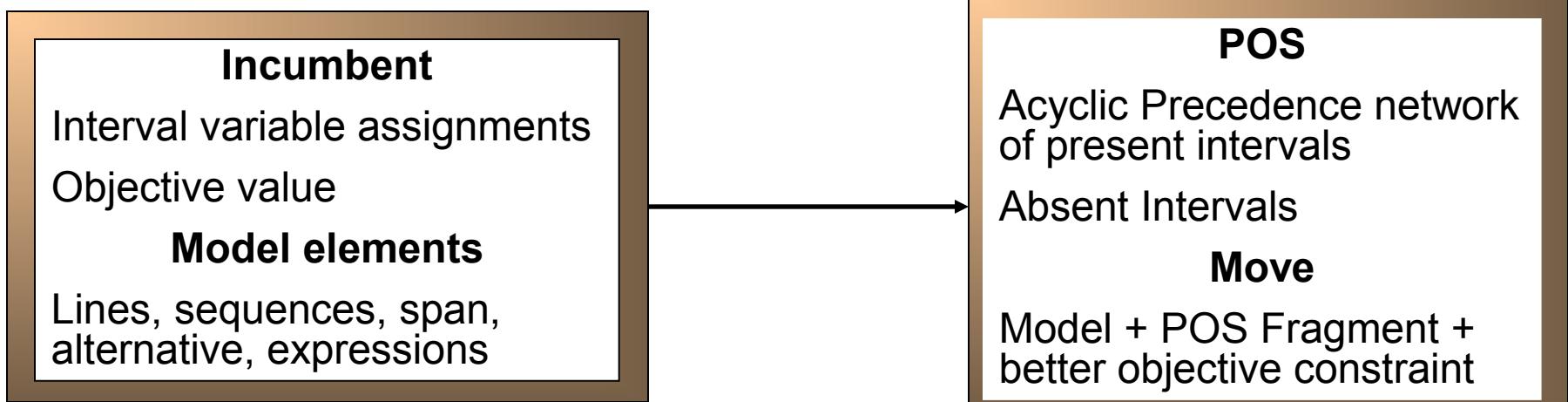


Resources and objective terms

- Resources are structures defined everywhere in time (between origin and horizon)
 - Lines (segmented integer functions)
 $\{([s_i, e_i], v_i) \text{ integers, } s_i \leq e_i, 0 \leq v_i, e_i \leq s_{i+1}\}$
 - Calendar, cumulative, state resources, oven, ...
 - Sequence of intervals
permutation of a set of intervals linked by precedences
 - Disjunctive resources, transportation systems, batches, campaign, ...
- Numerical expressions are functions of ranges of the intervals.
 - Example `startOf(a, v)` is the value of the start point of a if present, v if a is absent
 - Start, end, length, height of cumul, next or previous values on sequences, ...

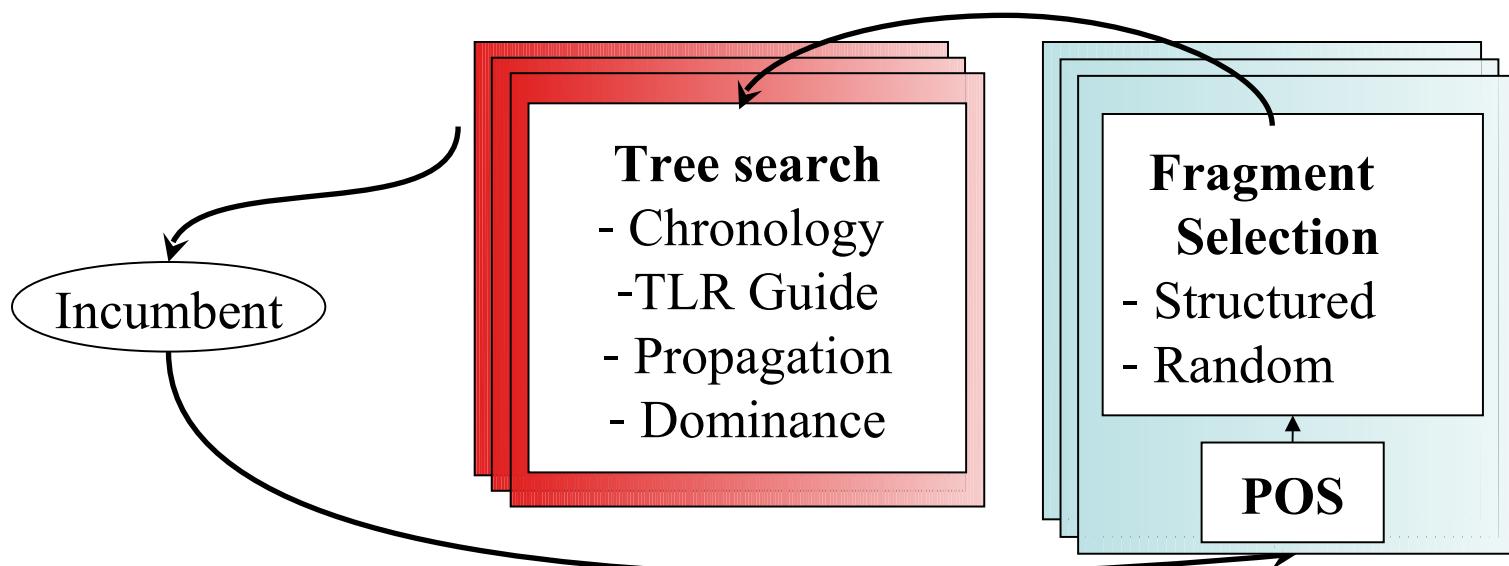
LNS in CPO temporal problems

- Issue is having a robust iterative procedure for scheduling problems including presence (alternative, optional tasks) costs, early tardy costs, and duration (or workload) costs,
 - Main hypothesis is the cost is dominated by the time placement.
 - The incumbent is relaxed in a **Preferred Ordered Set** (POS) given by a **acyclic precedence network**.
 - The **LNS** moves consist in selecting a fragment of the POS



TLR Guideline

- The tree search completion procedure generating solutions is essentially a **chronological decisions procedure** on the interval variables presence and end-points
- Eventually a **Temporal Linear Relaxation (TLR)** is computed at the fix point of the initial propagation of the move and solved by Linear Programming.
 - The TLR is used as a **guide for date decision selections and presence values.**



Notations for the TLR formulation

- Notations in the Linear model
 - Interval ranges define the range variables s , e , l in the LP
 - Interval range presence: x boolean
- A CP numerical range or expression defines a range variable v whose bound are set from the current domain in the propagation engine.

v in $[V_{\min}, V_{\max}]$

Principle of the TLR formulations

- The implication graph is stated by binary constraints:

$$x(a) \leq x(b)$$

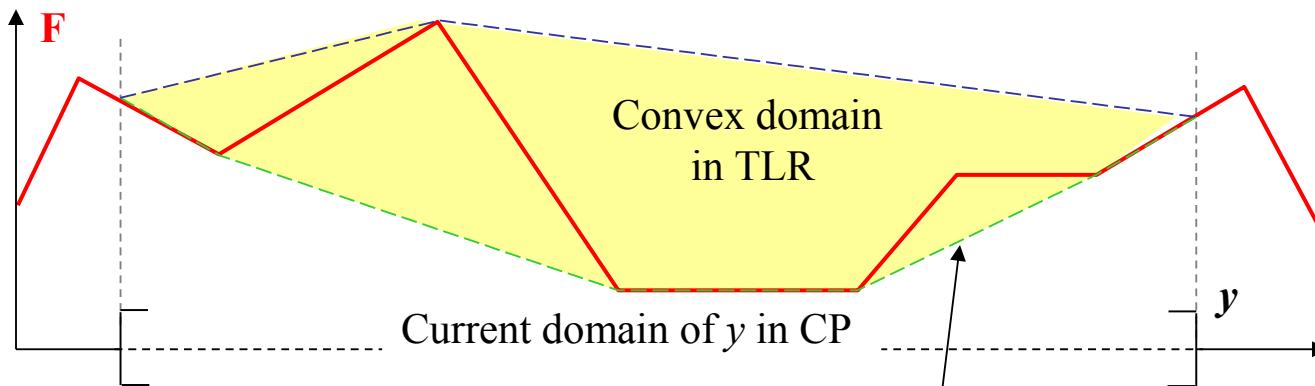
- The precedence graph is stated thanks to big-M constraints reified by known implications (y in start or end point) (including $e = s + l$ for an interval variable)

$$y_1 - y_2 + \text{delay} - M_1(1 - x_1) - M_2(1 - x_2) \leq 0$$

x_1 implies not x_2 x_1 implies x_2
 x_1 and x_2

- From the set of precedences for each ~~present~~ and l of an interval variable, we state Y_o in $[Y_{\min}, Y_{\max}]$ as the **value of y when the interval is absent** ($x = 0$) and such that Y_o tightens the big-M constraints.
- The TLR formulation is then calculated from objective terms introspection.

Unary constraints or expressions



- The expression is a numerical function of the range y (in s, e, l) of the interval. For example a **piecewise linear function**.
- The value of the expression is v_a when the interval is absent. Be v_p the linear relaxation of V when the interval is present

$$v = x^*v_p + (1-x)^*v_a$$

- The expression defines a range variable v in the TLR that is initialized from the current bounds of y in the propagation engine $[V_{\min}, V_{\max}]$
- The idea is to used the absence values in the precedence graph relaxation to state the big-M constraint.
- Be V_o the value of F for Y_o , the formulation is: $v = v_p + x(v_a - V_o)$

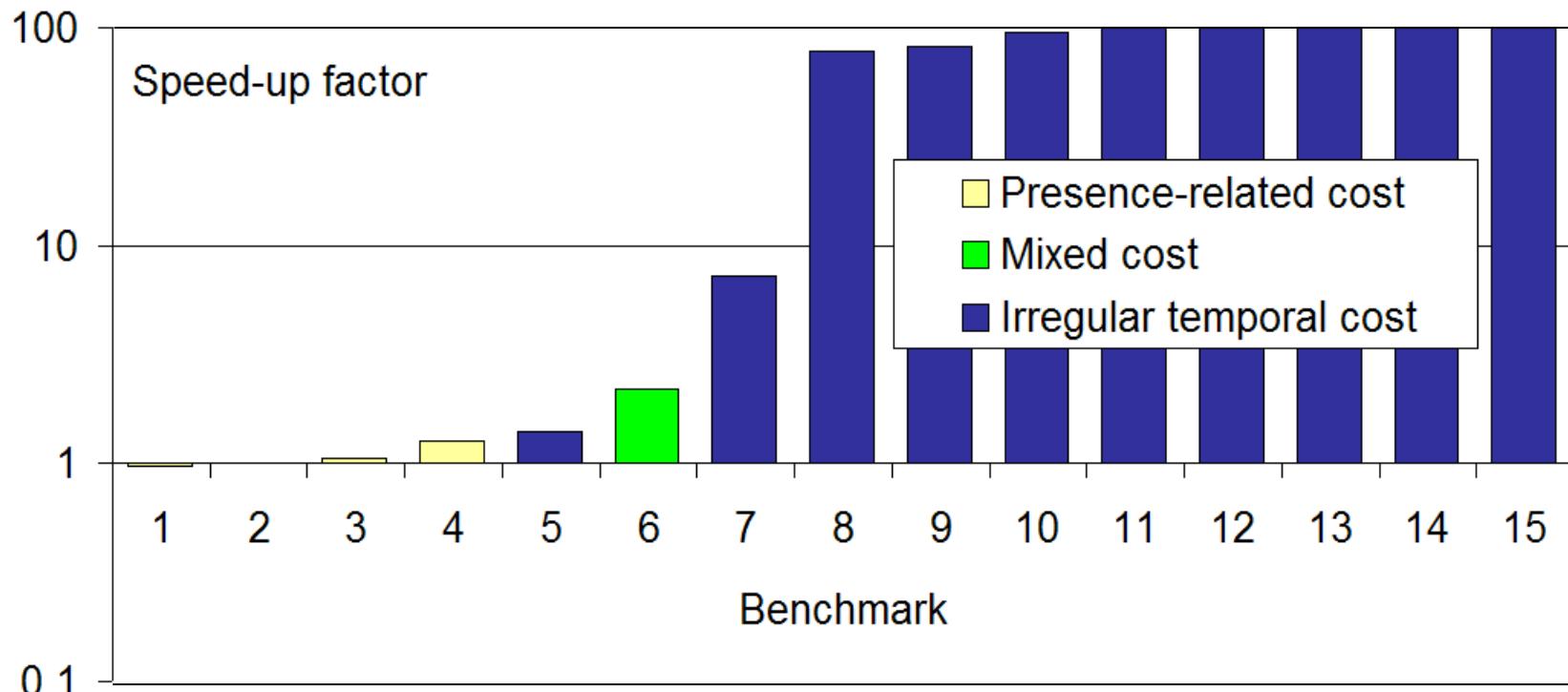
Binary constraints and expressions

- A full examples is shown in the paper for the overlap length of two intervals.
- The general formulation is $v = x_1^*x_2^*v_p + (1 - x_1^*x_2)^*v_a$
- From the implication graph (when possible e.g. x_1 is equivalent with x_2) we used a similar formulation as for unary expressions:
$$v = v_p(a_1, a_2) + x^*(v_a - V_o(a_1, a_2))$$
- v_p formulation:
 - Binary expressions usually expresses some temporal disjunction between the interval variables
 - The formulation of v_p is improved by cuts computed from
 - The current domain of a_1 and a_2 in the propagation engine.
 - Eventual precedence graph arcs involving a_1 and a_2 .

Other elements of the model

- Breakdown structures are composition of temporal constraints and logical constraints
 - an alternative is a SOS constraints on presence variables + synchronization of end-points of intervals
- Resource constraints are **still captured by frozen precedence arcs of the POS**
 - TLR formulations use rough energetic relaxations
- Numerical expression formulations use similar convex relaxations as the piecewise linear function formulations

Experimental Results



- 15 scheduling benchmarks comparing the iterative procedure with against without the TLR guide. The LP is ILOG IBM Cplex.
- Speed up is relative to the time at which the compared algorithms reached the same objective value under a global time limit.
- A capped 100 factor are problems such that the without TLR procedure never reaches even the first incumbent of the with TLR procedure.

Conclusion

- The TLR formulation as a guide is a very effective method for optimizing non regular scheduling costs
 - The TLR tends to be very informative as most of the resource constraints are still captured by frozen precedence arcs of the POS.
- Future works will concern both the control and the diversification of the temporal relaxation by:
 - Controlling the size and the content of the TLR depending on the characteristics of the problem and the current impact on the search,
 - Exploring stronger relaxations in the TLR, for instance by solving a MILP enforcing the Boolean presence status of interval variables,
 - Considering concurrent relaxations that exploits other dominant structures of the problem like capacity planning or sequencing sub-problems.