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# Similarity-based Lexicographic Method for Hierarchical Multiobjective Linear Programs

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## At High Level

We define the **concept of similarity** between intermediate linear programs (LPs) appearing while solving the hierarchical multiobjective linear program (**h-MOLP**) for the decision of whether we should solve the current LP from **scratch** or use the **available feasible solution** obtained from the **previous LP** solve.

## h-MOLP with bounded variables

$$\begin{aligned} &\text{lexmin } c^1 x, c^2 x, \dots, c^t x \\ &\text{subject to } Ax = b, \\ &\quad l \leq x \leq u. \end{aligned}$$

**lexmin:**  $c^1 x$  is more important than  $c^2 x$  which is, more important than  $c^3 x$ , and so on and,  $c^t x$  is of least importance.

## Lexicographic Methods

1.  $LP^k$  - Constraint-addition rule
2.  $\text{modLP}^k$  - Variable fixing rule

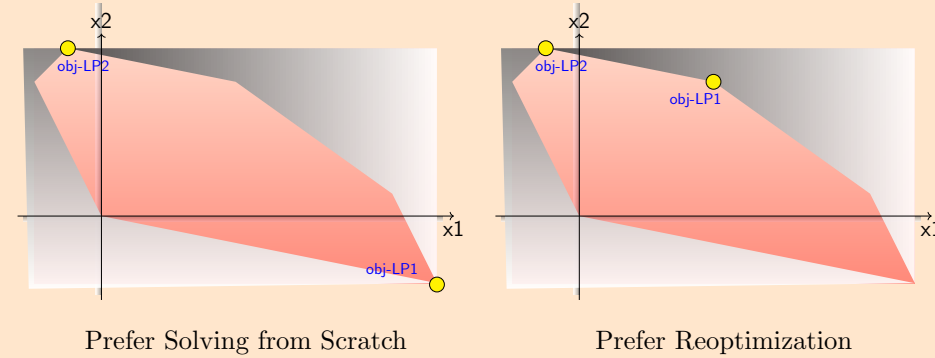
$$\begin{aligned} LP^k &:= \min c^k x & \text{modLP}^k &:= \min c^k x \\ \text{s.t. } Ax &= b, & \text{s.t. } Ax &= b, \\ c^i x &= y^i, \quad i = 1, \dots, k-1, & x_j &= f_j, \quad j \in J^k \subseteq \{1, \dots, n\}, \\ l &\leq x \leq u. & l &\leq x \leq u. \end{aligned}$$

## Remarks

1. Both provide Pareto-optimal solution
2. Most literature describes  $LP^k$  rule in the discussion of the method
3. Most solvers prefer implementing the  $\text{modLP}^k$
4. Assumption: h-MOLP is nontrivial - alternative optimal solutions while solving the underlying LPs, and we have conflicting objectives

## Motivation

- To obtain **one solution** point both need the solution of **many LPs**
- To **speed it up**, many solvers provide a feature of **Reoptimization**
- However, a rule is required to **selectively decide** when to use Reoptimization



## Notion of Similarity

$LP^1$  is **similar** to  $LP^2$ , if the solution of  $LP^1$  helps solve  $LP^2$  faster than just solving it from scratch.

**Definition 1.** Let  $LP^1 := \min\{c^1 x \mid Ax = b, l^1 \leq x \leq u^1\}$  and  $LP^2 := \min\{c^2 x \mid Ax = b, l^2 \leq x \leq u^2\}$  are two LPs. We assume that the feasible sets of both  $LP^1$  and  $LP^2$  are non-empty. With respect to  $p$  and  $B$ , the optimal solution and the optimal basis of  $LP^1$ , we say  $LP^1$  and  $LP^2$  are “similar” if

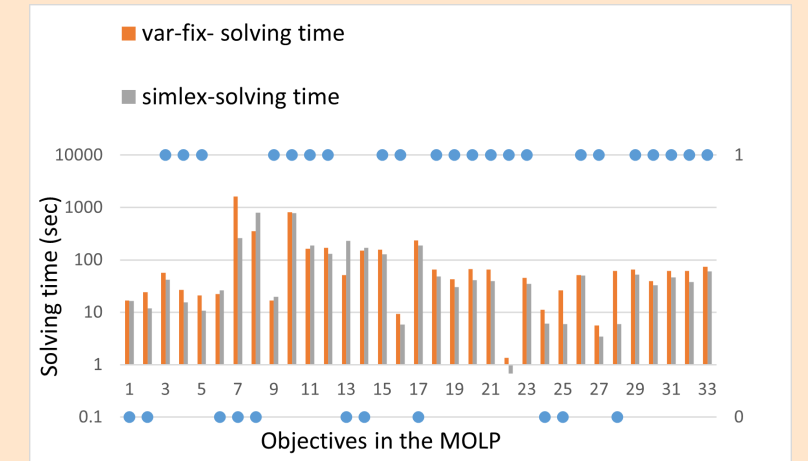
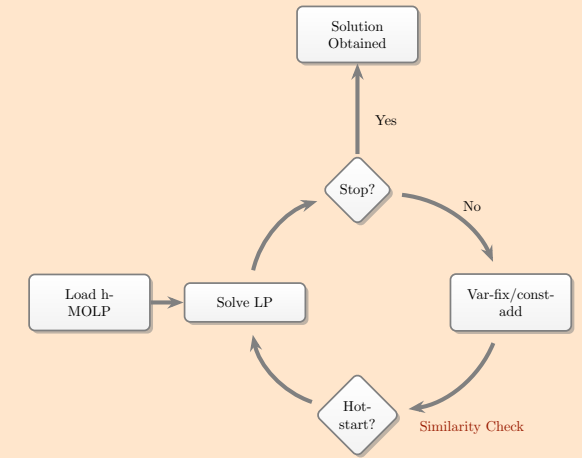
$$\left[1 - \left(\frac{\sum_{j=1}^n I_{p_j} |c_j^2 - c_j^1|}{\sum_{j=1}^n |c_j^2 - c_j^1|}\right)\right] \geq \kappa,$$

where  $0 \leq \kappa \leq 1$  is the given parameter.  $I_{p_j}$  the indicator function is equal to 1 if  $e_j \neq 0$  and  $\bar{c}_j^1 = 0$ , else is set to 0. Here, reduced cost of  $LP^1$  is  $\bar{c}^1 := c^1 - c_B^1 B^{-1} A$  and  $e := c^2 - c_B^2 B^{-1} A$  is assumed as the estimated reduced cost of  $LP^2$ .

## Result

Consider  $p$ ,  $B$  and  $\bar{c}^1$  are at optimality of  $LP^1$ . Consider a perturbed cost vector  $c^2 = c^1 + \delta$  where  $\delta \neq 0$  is given. Let  $e = c^2 - c_B^2 B^{-1} A$ . For any component  $p_i, i \in \{1, \dots, n\}$ , if  $\bar{c}_i^1 \cdot e_i \leq 0$  and,  $\bar{c}_i^1$  and  $e_i$  both can not together be zero, the basis  $B$  of  $LP^1$  is not the optimal basis of  $LP^1$  with perturbed cost vector  $c^2$ .

## SimLex: Similarity-based Decision to Reoptimize



**Performance comparison:** SimLex Vs Variable-fixing. Blue points at level 1 indicate solving with hotstart and at level 0 indicate solving from scratch

## Result and Future Work

1. Comparison is done over default CPLEX12.10 and other rules
2. Apply this idea with 1) **MOPLIB**, a library of benchmark multi objectives programs, and, 2) model of the **master production schedules** (MPS)
3. Reported **25%** (SGM-50) speedup from our rule over default CPLEX rule on obtaining MPS on **13** large scaled **consumer and products goods** (CPG) industry dataset
4. **Ideal parameter selection** of  $\kappa$  for similarity computation is not emphasized and is the main works that need to be done