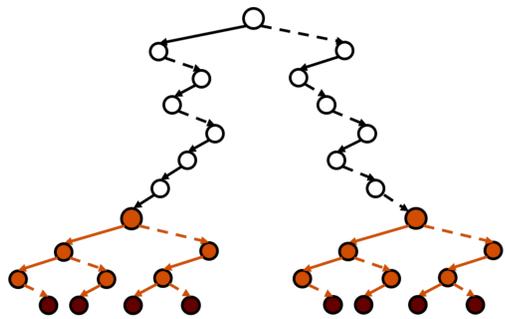


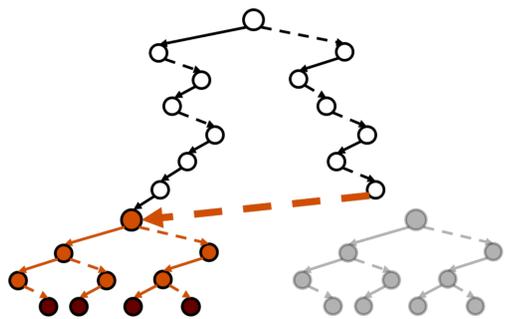
1. Collecting IP Solutions

We can collect near-optimal solutions of an integer program by branching to exhaustion, like in the one-tree approach [Danna et al., 2007] used by CPLEX solver [IBM Corp., 2013].

That entails **finding solutions one-by-one**, which **takes too long if there are too many**. Our goal is to remedy that by **identifying subproblems with the same completions**.



We might **add multiple solutions in a single step**, branch less, and possibly finish sooner.



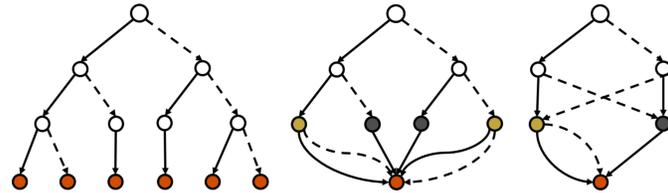
We assume subproblems on the same variables. Branching is not usually done that way, but we **save a set of k variables for branching last**.



In fact, **most equivalences are at the bottom**. For $n - k$ fixed binary variables, there are $O(2^{n-k})$ partial solutions. For k binary variables left, there are $O(2^k)$ completion sets.

2. Turning a Branching Tree into a Decision Diagram

A bottom-up pass suffices to merge all nodes with isomorphic completions [Bryant, 1986]:



We can check if nodes root isomorphic subtrees without branching again. After exploring

$$\sum_{i=1}^n a_i x_i \leq a_0, \quad a_0, \dots, a_n \in \mathbb{Z}, \quad (1)$$

we find **all values $[\beta, \gamma]$ equivalent to a_0** [Behle, 2007, Abío et al., 2012, Abío and Stuckey, 2014].

3. Equivalence for Single Inequalities

We generalize that to **additively separable inequalities with fractional coefficients**:

$$\sum_{i=1}^n f_i(x_i) \leq \rho \quad (2)$$

For a partial solution $\bar{x}_1, \dots, \bar{x}_{n-k}$, we have

$$\sum_{i=n-k}^n f_i(x_i) \leq \rho - \sum_{i=1}^{n-k} f_i(\bar{x}_i) \quad (3)$$

We can compute the **equivalent Right-Hand Side (RHS) values $[\beta, \gamma]$** for x_1 to x_{n-k} fixed, and later compare it with other subproblems:

$$\beta = \max \left\{ \sum_{i=n-k}^n f_i(x_i) \mid x \text{ is feasible for (3)} \right\} \quad (4)$$

$$\gamma = \min \left\{ \sum_{i=n-k}^n f_i(x_i) \mid x \text{ is infeasible for (3)} \right\} \quad (5)$$

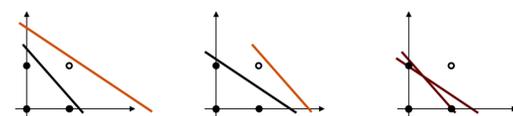
4. Equivalence for Multiple Inequalities

A partial solution on m inequalities as those above defines a subproblem of the form

$$\sum_{i=n-k}^n f_i^j(x_i) \leq \rho_j - \sum_{i=1}^{n-k} f_i^j(\bar{x}_i), \quad \forall j = 1, \dots, m \quad (6)$$

Equivalent RHS values are interdependent, but there is a **unique tightest RHS vector**:

$$\begin{array}{lll} 5x_1 + 4x_2 \leq 6 & \leq 10 & \leq 5 \\ 6x_1 + 10x_2 \leq 17 & \leq 11 & \leq 10 \\ x_1, x_2 \in \{0,1\} \end{array}$$



For the j -th inequality, the tightest RHS is

$$\beta_j = \max \left\{ \sum_{i=n-k}^n f_i^j(x_i) \mid x \text{ is feasible for (6)} \right\} \quad (7)$$

Given an unexplored node u with RHS ρ^u and an explored node v with tightest RHS β^v :

❶ **Does v define a subset of u ?**

Iff $\beta^v \leq \rho^u$

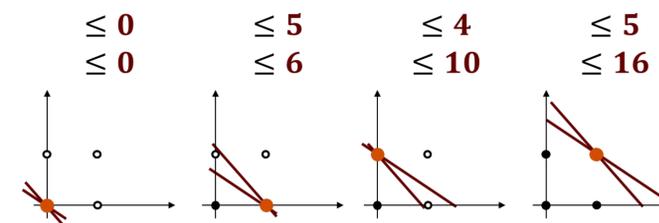
❷ **Does v define the dominating subset of u ?**

Iff ❶ & $\beta^v \geq \beta^{v'}$ for any node v' satisfying ❶

❸ **Does v define a subset that is not proper?**

Iff ❷ & explored nodes span all completion sets

We generate $O(2^k)$ nodes to span all those sets: one for **the tightest RHS of each completion**.



5. Results

Our approach is implemented on CPLEX through callbacks. For now, we compare it with one-tree upon fixed variable ordering.

Ours **branched less** and sometimes **ran faster**, with geometric means down by **40%** and **20%**.

Problem	Pool Gap	Sols	Branches		Runtime (s)	
			One-tree	$k = 9$	One-tree	$k = 9$
bm23	59	2,168	18,066	16,203	2.4	2.7
lseu	250	95,335	2,188,870	2,162,601	220.6	237.1
mod008	15	1,228	2,324,558	2,317,052	188.9	183.4
p0033	2112	10,746	20,475	7,861	1.7	1.0
p0040	7102	519,216	851,354	144,066	74.9	13.3
pipex	120	12,266	48,978	48,842	5.2	5.9
sentoy	160	3,513	550,555	473,010	41.5	38.7
stein15	6	2,809	2,011	830	0.21	0.19
stein27	9	367,525	325,275	43,909	37.8	39.0
stein45	1	19,603	874,245	870,796	89.3	96.8
Geometric mean			1.7×10^5	1.0×10^5	16.9	13.8

6. Next Steps

Find good choices for the k bottom variables; explore heuristics to choose where to branch next when changing CPLEX default choices.

Extend this approach to the mixed-integer case.

Mix it with branching dominance, which pays off at the top of the branching tree [Fischetti and Toth, 1988, Fischetti and Salvagnin, 2010].

7. References

I. Abío and Peter J. Stuckey. Encoding linear constraints into SAT. In B. O'Sullivan, editor, *Proceedings of CP*, pages 75–91. Springer, 2014.

I. Abío, R. Nieuwenhuis, A. Oliveras, E. Rodríguez-Carbonell, and V. Mayer-Eichberger. A new look at BDDs for pseudo-boolean constraints. *Journal of Artificial Intelligence Research*, 45:443–480, 2012.

M. Behle. *Binary Decision Diagrams and Integer Programming*. PhD thesis, Universität des Saarlandes, 2007.

R.E. Bryant. Graph-based algorithms for boolean function manipulation. *IEEE Trans. Comput.*, C-35(8):677–691, 1986.

E. Danna, M. Fenelon, Z. Gu, and R. Wunderling. Generating multiple solutions for mixed integer programming problems. In M. Fischetti and D. P. Williamson, editors, *Proceedings of IPCO*, pages 280–294. Springer, 2007.

M. Fischetti and D. Salvagnin. Pruning moves. *INFORMS Journal on Computing*, 22(1):108–119, 2010.

M. Fischetti and P. Toth. A new dominance procedure for combinatorial optimization problems. *OR Lett.*, 7(4):181–187, 1988.

IBM Corp. *IBM ILOG CPLEX Optimization Studio CPLEX Users Manual Version 12 Release 6*, 2013.