

LP Solution Polishing to improve MIP Performance

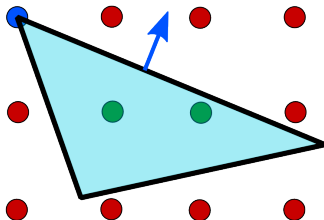
Matthias Miltenberger

Zuse Institute Berlin

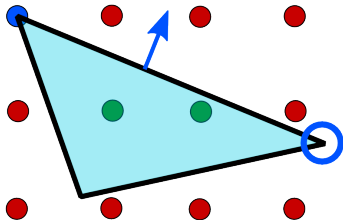
ICCOPT 2016 - Tokyo, August 9th 2016



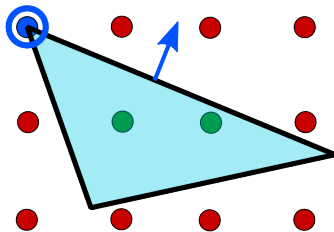
- ▶ Solving a MIP $\min\{c^T x \mid Ax = b, x_i \in \mathbb{Z} \text{ for } i \in I\}$ involves solving many LPs as linear relaxations
- ▶ LP solutions are rarely unique



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- ▶ LP solutions are rarely unique



- ▶ How to find the **best** one?

1. Introduction

Dual Degeneracy

Performance Variability

2. Related Work

3. Solution Polishing

Integrality of Variables

4. Computational Results

SCIP Optimization Suite

5. Conclusion and Outlook

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- ▶ Two types of degeneracy in LP:
 - ▶ primal: multiple bases defining one vertex of the polyhedron
 - ▶ dual: facet of the polyhedron parallel to the objective function
- ▶ Most (practical) problems are primal and dual degenerate
- ▶ Degeneracy is the most prominent cause of MIP performance variability

- ▶ Performance of a MIP solver may vary drastically when the data changes
 - ▶ change row and column order
 - ▶ use a different random seed
 - ▶ implement a different tie breaker
 - ▶ ...
- ▶ Several causes for variability
 - ▶ different LP optima are probably most influential
- ▶ Explained in
 - ▶ *Danna, E.:* **Performance variability in mixed integer programming** MIP Workshop (2008)
 - ▶ *Koch, T., et al.:* **MIPLIB 2010**, Math. Program. Comp. (2011)

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Improving branch-and-cut performance by random sampling

M. Fischetti, A. Lodi, M. Monaci, D. Salvagnin, A. Tramontani

Math. Program. Comp. (2016), Vol. 8

1. perform preprocessing on one core
 2. solve root LP on $k - 1$ cores with different random seeds
 - ▶ collect primal solutions and generated cuts
 3. complete solving process on one core with yet another random seed
-
- ▶ previously collected information helps to improve the performance
 - ▶ performance variability is reduced
 - ▶ contained in the latest CPLEX release for $k = 3$

Lexicography and degeneracy: Can a pure cutting plane algorithm work?

A. Zanette, M. Fischetti, E. Balas

Math. Program. (2011) Vol. 130

- ▶ answer: Yes, it can!
- ▶ ...when choosing the **correct** LP basis
 - ▶ cutting plane method adds many cuts (almost) parallel to objective
 - ▶ use the lexicographic dual simplex to deal with high dual degeneracy
 - ▶ or modify the objective to mimic the lexicographic behavior
- ▶ standard cutting plane approach suffers from bad numerical stability

LP relaxation modification and cut selection in a MIP solver

T. Achterberg

US Patent (2011)

- ▶ similar to k -Sample, another optimal LP basis is constructed
 - ▶ fix some non-basic variables and modify the objective
 - ▶ use new basis to collect more information, e.g. for cuts
- ▶ implemented in CPLEX

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- ▶ Dual simplex algorithm terminates at first primal feasible, optimal basis
- ▶ Perform additional **polishing steps** altering this basis
- ▶ **Reminder:**
 - ▶ Basic indices: B , non-basic indices: N
 - ▶ Nonbasic variables are on their bound:
 $x_N = 0$ or $x_N = u$
 - ▶ Basic variables can be between bounds (depending on x_N)
 $x_B = A_B^{-1}(b - A_N x_N)$

- ▶ Polishing steps are primal iterations (to preserve feasibility):
 1. find non-basic indices to enter the basis ($\hat{=}$ pricing step)
 - ▶ choose one with zero reduced costs to stay on optimal hyperplane
 2. try pivoting and check whether leaving index is **good** ($\hat{=}$ ratio test)
 3. repeat

1. Decrease fractionality $\hat{=}$ push integer variables **out of** basis
 - ▶ less branching candidates
 - ▶ hopefully *closer* to an integer feasible solution
2. Increase fractionality $\hat{=}$ push integer variables **into** basis
 - ▶ may generate better cuts (basis matrix contains less slack)

- ▶ Usually, LP solver has no knowledge of integrality
- ▶ **Unlucky scenario:**
 1. push **continuous** variable out of basis
 2. remaining basic **integer** variables are moved away from bounds
- ▶ **Remedy:**
 1. transfer information about integer variables to LP solver
 2. push only basic integer variables to their bounds

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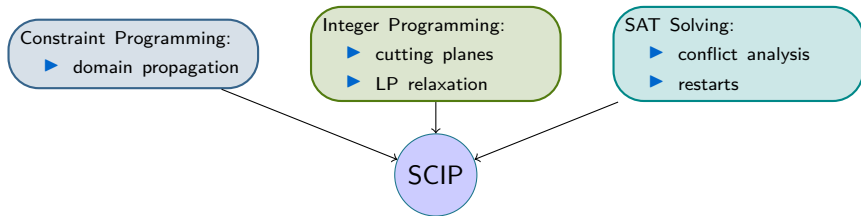
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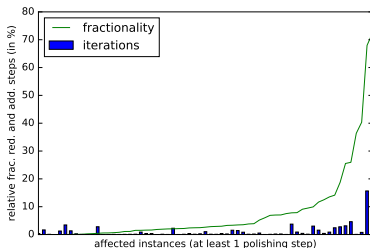
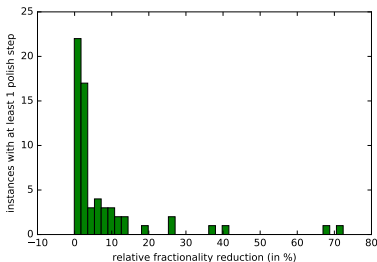
- ▶ Test set: MIPLIB 3 + MIPLIB 2003 + MIPLIB 2010, 168 instances
- ▶ All runs sequentially on one core
- ▶ SCIP Optimization Suite 3.2.1 with modifications



SCIP (Solving Constraint Integer Programs) ...

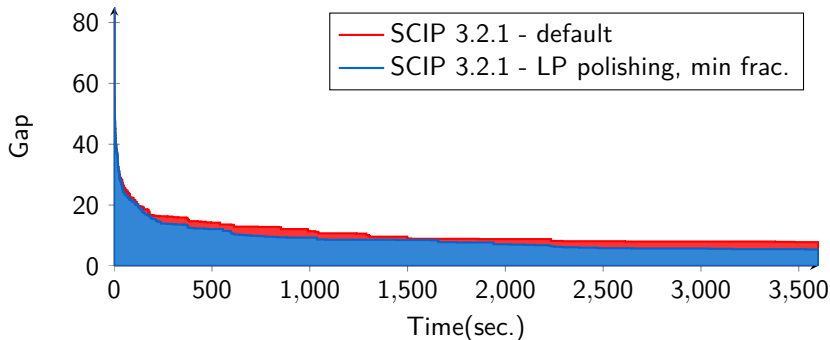
- ▶ has a modular structure via plugins,
- ▶ provides a full-scale global MINLP solver,
- ▶ part of the [SCIP Optimization Suite](#) (incl. SoPlex, ZIMPL, GCG, and UG),
- ▶ is free for academic purposes,
- ▶ and is available in source-code under <http://scip.zib.de>

- ▶ Compare fractionality before and after solution polishing
- ▶ Only root LP is solved
- ▶ With integrality information in SoPlex

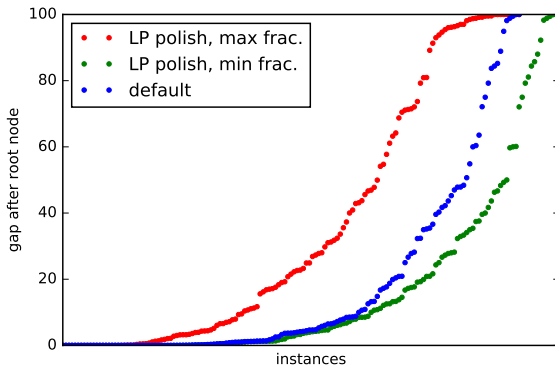


number of affected instances (of 168):	63
number of instances with a reduction of more than 5%:	22
mean percentage reduction of fractionality:	7.74
mean percentage of additional steps:	1.29

- ▶ Polishing reduces number of nodes by 2-3 %
- ▶ Transferring integrality information is expensive
- ▶ Mean primal integral improvement: 38481.0 → 31316.1



- ▶ Increasing fractionality leads to a high increase in nodes **and** deteriorates the root gap
- ▶ Reducing fractionality leads to a smaller root gap



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- ▶ Polished LP optimum is still not unique
 - ▶ maximum or minimum of fractionalities is not guaranteed

Possible improvements:

- ▶ Implement a (more expensive) technique to find the **best** basis
- ▶ Transfer of integrality information needs a more efficient implementation
 - ▶ use integrality information also in other parts of the LP solver
- ▶ Make use of several optimal bases

- ▶ Solution polishing is cheap to apply
 - ▶ ...when used to reduce fractionality
 - ▶ ...when transfer of integrality information is improved
- ▶ Does not modify the LP problem data
- ▶ Already provides promising results concerning fractionality and gap reduction
- ▶ No effect on reducing performance variability observed yet

- ▶ More refinement and tuning possible
 - ▶ especially regarding fractionality increase
 - ▶ polishing could be applied more selectively
- ▶ Reduce performance variability by LP solution polishing

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Thank you for your attention!
ご清聴ありがとうございました