

Large neighbourhood Benders' search

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Structured mixed integer programming

Basic idea: Minimise a linear objective function over a set of solutions satisfying a structured set of linear constraints.

$$\begin{aligned} \min \quad & c^\top x + d^\top y, \\ \text{subject to} \quad & Ax \geq b, \\ & Bx + Dy \geq g, \\ & x \in \mathbb{Z}_+^{p_1} \times \mathbb{R}_+^{n_1 - p_1}, \\ & y \in \mathbb{Z}_+^{p_2} \times \mathbb{R}_+^{n_2 - p_2}. \end{aligned}$$

Solving large scale optimisation problems

Aim: Embed Benders' decomposition within a state-of-the-art solver to provide effective tools to employ decomposition to solve large-scale optimisation problems.

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- ▶ Develop a general Benders' decomposition framework
- ▶ Harness the capabilities of state-of-the-art MIP solvers when using decomposition techniques
 - ▶ Integration of Benders' decomposition and large neighbourhood search heuristics.

Benders' decomposition

Original problem

$$\begin{aligned} \min \quad & c^\top x + d^\top y, \\ \text{subject to} \quad & Ax \geq b, \\ & Bx + Dy \geq g, \\ & x \in \mathbb{Z}_+^{p_1} \times \mathbb{R}_+^{n_1 - p_1}, \\ & y \in \mathbb{R}_+^{n_2}. \end{aligned}$$

Benders' decomposition

$$\begin{aligned} \min \quad & c^\top x + f(x), \\ \text{subject to} \quad & Ax \geq b, \\ & x \in \mathbb{Z}_+^{p_1} \times \mathbb{R}_+^{n_1 - p_1}. \end{aligned}$$

where

$$f(x) = \min_{y \in \mathbb{R}_+^{n_2}} \{d^\top y \mid Bx + Dy \geq g\}$$

Benders' decomposition

Master problem

$$\begin{aligned} \min \quad & c^\top x + \varphi, \\ \text{subject to} \quad & Ax \geq b, \\ & \varphi \geq u_\omega^\top (g - Bx) \quad \forall \omega \in \mathcal{O}, \\ & 0 \geq u_\omega^\top (g - Bx) \quad \forall \omega \in \mathcal{F}, \\ & \varphi \in \mathbb{R}_+, \\ & x \in \mathbb{Z}_+^{p_1} \times \mathbb{R}_+^{n_1 - p_1}. \end{aligned}$$

Subproblem

$$\begin{aligned} z(\hat{x}) = \min \quad & d^\top y, \\ \text{subject to} \quad & Dy \geq g - B\hat{x}, \\ & y \in \mathbb{R}_+^{n_2 - p_2}. \end{aligned}$$

Benders' decomposition

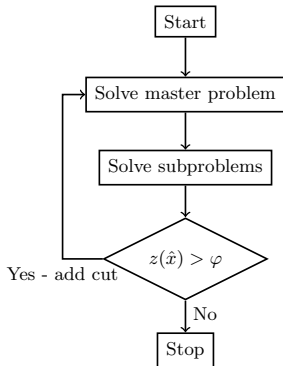
Master problem

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Standard Benders' implementation

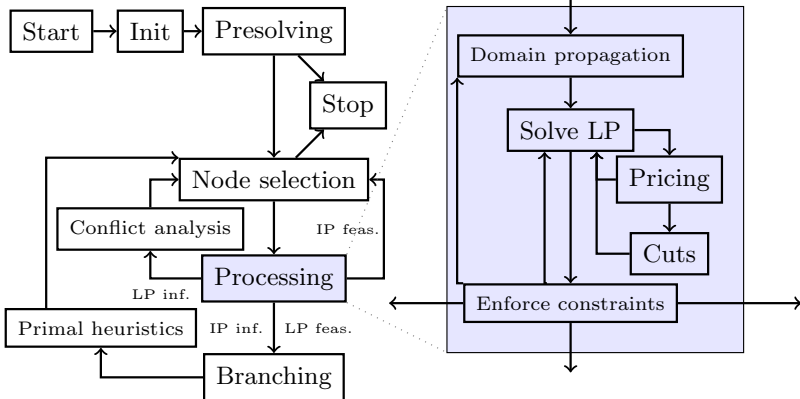


- ▶ Easy to understand and simple to implement.
- ▶ Not always effective, large overhead in repeatedly solving master problem.

Branch-and-check

- ▶ Modern solvers pass through a number of different stages during node processing.
- ▶ Some of these stages can be used to generate Benders' cuts.
- ▶ By interrupting node processing, Benders' cuts are generated during the tree search.

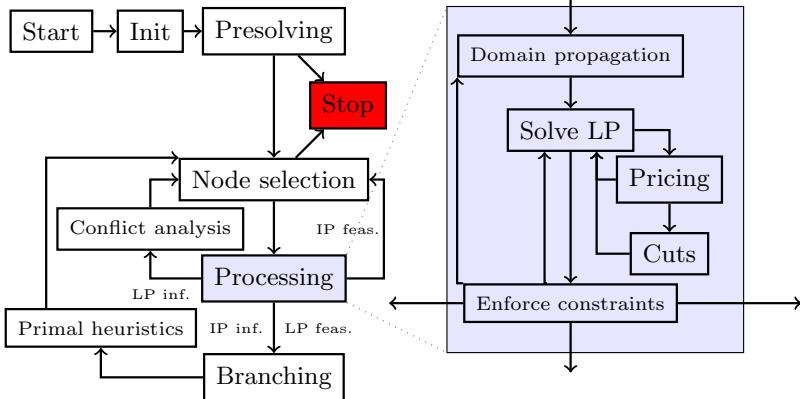
Solving process



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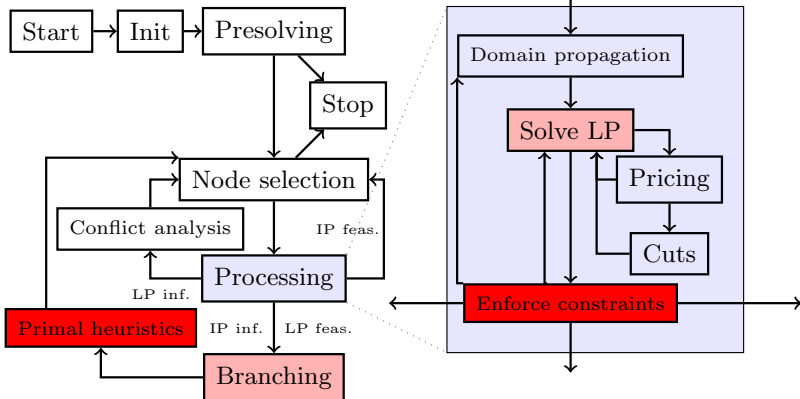
Cut generation - Standard Benders'



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Cut generation - Branch-and-check

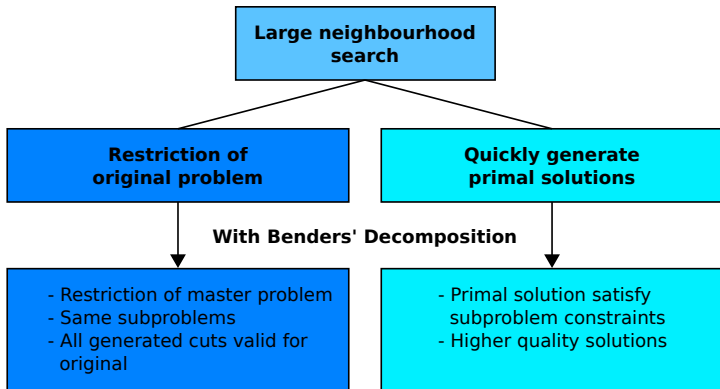


Large neighbourhood search heuristics

- ▶ *Concept*: Identify improved primal solutions by solving an auxiliary problem that is a restriction of the original problem.
- ▶ The auxiliary problem is typically formed by fixing variables or the addition of constraints.
- ▶ The restricted auxiliary problem is expected to be easier to solve than the original problem.
- ▶ State-of-the-art solvers employ many variants of large neighbourhood search heuristics
 - ▶ Crossover, DINS, Local branching, proximity search, RENS, ...
- ▶ Very effective in finding solutions to difficult optimisation problems.

Large Neighbourhood Benders' Search

- ▶ *Concept:* Using large neighbourhood search to improve the convergence of the Benders' decomposition algorithm.



- ▶ Builds upon successful trust region approaches.

Benders' decomposition everywhere

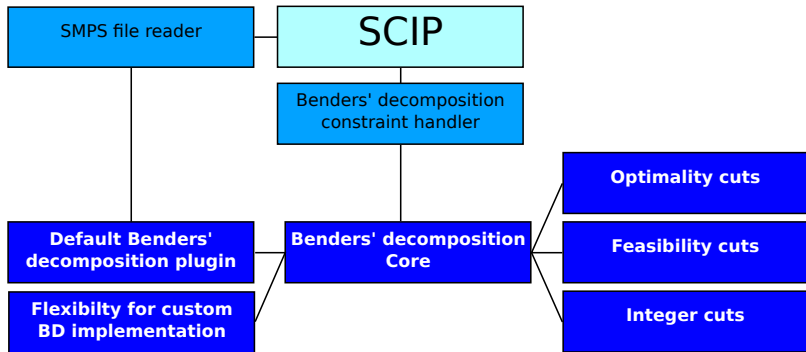
- ▶ Local branching (Rei et al. (2009)) and proximity search (Boland et al. (2015)) have demonstrated potential to enhance BD using LNS heuristics.
- ▶ Modern MIP solvers are endowed with vast array of LNS heuristics.
- ▶ LNS heuristics solve sub-MIP instances—can be solved by Benders' decomposition.

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Aim: Employ Benders' decomposition in all available LNS heuristics.

Generic Benders' Decomposition



Using the Benders' decomposition framework

- ▶ *First option:*
 - ▶ Provide an instance in SMPS (Stochastic MPS) format to SCIP.
 - ▶ The SMPS format consists of three components: core model, time file for stages and stochastic information.
- ▶ To use the Benders' decomposition algorithm, the following setting must be used

`reading/sto/usebenders = TRUE.`

Using the Benders' decomposition framework

- ▶ *Second option:*
 - ▶ `SCIPcreateBendersDefault`(master SCIP, array of subproblem SCIPs, number of subproblems)
- ▶ Master SCIP and Subproblem SCIP instances must be created by the user.
- ▶ Variables common between the master problem and subproblems must have the same name.
- ▶ All variable mappings between master and subproblems are generated automatically.

Using the Benders' decomposition framework

- ▶ *Third option:*
 - ▶ Implement a custom Benders' decomposition plugin: `benders_xyz.c` and `benders_xyz.h`
- ▶ Required callbacks:
 - ▶ Mapping between the master and subproblem variables,
 - ▶ Method to create each subproblem.
- ▶ Various optional callbacks:
 - ▶ Pre subproblem solving, a solving method for the subproblem, post solving, freeing subproblem, and usual SCIP callbacks (`init`, `initsol`, `exit`, `exitsol`, `copy`, ...)

Using Benders' decomposition in LNS

Advantages:

- ▶ Solutions are guaranteed to be optimal w.r.t the subproblems.
- ▶ Generated cuts can be transferred to the original problem.
- ▶ Flexible w.r.t SCIP development.
- ▶ Simple changes of parameters can provide aggressive use of LNS for Benders' decomposition.

Test problems

Stochastic Capacitated Facility Location Problem

- ▶ CAP instances from OR-Library with 25 or 50 facilities.
- ▶ Instances with 250 and 500 scenarios.
- ▶ 48 instances.

Stochastic Network Interdiction Problem

- ▶ The same set of instances as used by Bodour et al. (2017).
- ▶ Graph with 738 nodes and 2586 arcs, 320 possible sensor locations.
- ▶ 456 scenarios.

Stochastic Multiple Knapsack Problem

- ▶ Instances collected from SIPLIB.
- ▶ First stage: 240 binary variables, 50 knapsack constraints.
- ▶ Second stage: 120 binary variables, 5 knapsack constraints.
- ▶ 30 instances, each with 20 scenarios.

Experiments

Benders - Standard implementation of branch-and-check.

- ▶ Benders' cuts are generated from the LP, Relaxation, Pseudo and feasible solutions

LNS Check - BD with large neighbourhood Benders' search.

- ▶ Employing Benders' decomposition within each LNS heuristic

Transfer cuts - Extension of large neighbourhood Benders' search.

- ▶ All cuts generated during the large neighbourhood Benders' search are transferred to the original problem

Primal Integral (Berthold (2013))

Primal Gap

$$\gamma(Z^P, \hat{Z}^P) = \begin{cases} 0, & \text{if } |Z^P(t)| = |\hat{Z}^P| = 0, \\ 1, & \text{if } Z^P(t) \times \hat{Z}^P < 0, \\ \frac{|Z^P(t) - \hat{Z}^P|}{\max\{|Z^P(t)|, |\hat{Z}^P|\}}, & \text{otherwise.} \end{cases}$$

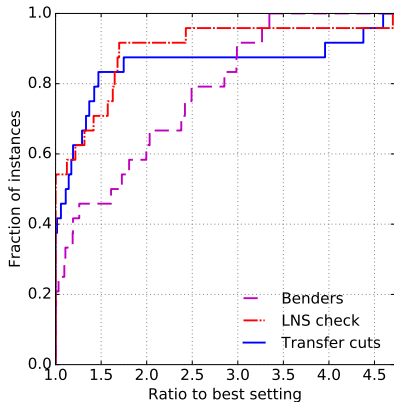
Integral

$$P(T) = \int_{t=0}^T \gamma(Z^P, \hat{Z}^P) dt$$

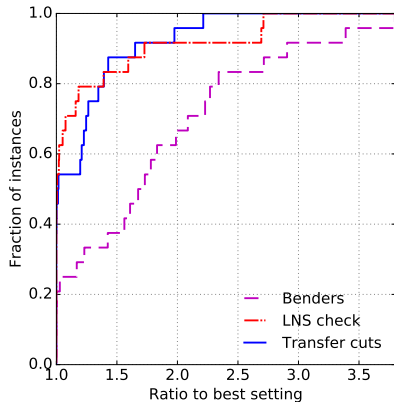
SCFLP

Performance profile of the primal integral

250 Scenarios

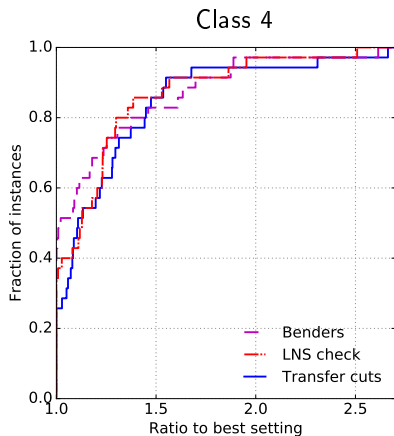
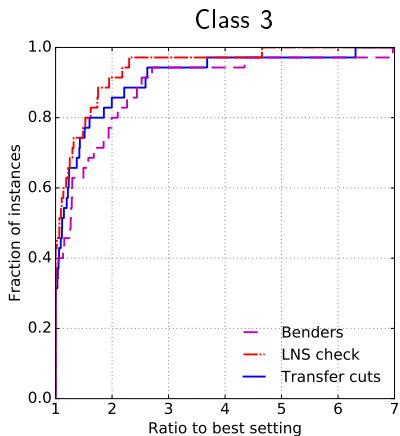


500 Scenarios



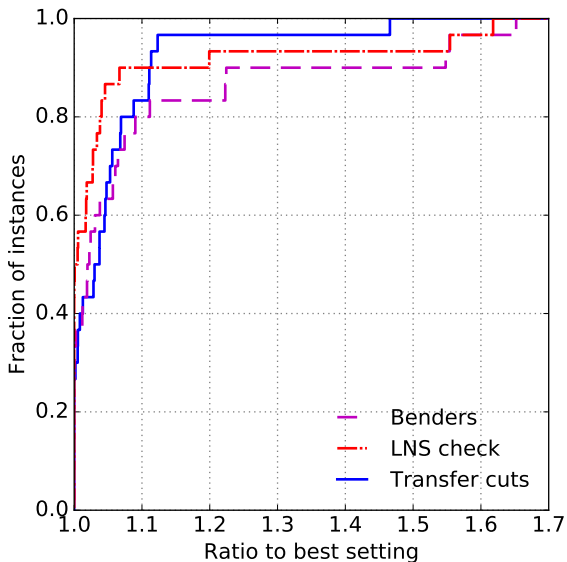
SNIP

Performance profile of the primal integral



SMKP

Performance profile of the primal integral



Key points

- ▶ SCIP has been extended with a generic implementation of Benders' decomposition.
- ▶ Benders' decomposition has been implemented as a branch-and-check algorithm.
- ▶ Functionality is available to use Benders' decomposition within LNS heuristics.
- ▶ Integrating Benders' decomposition and LNS heuristics can significantly enhance the primal bound improvement.