# Large-scale parallelisation for the Benders' decomposition framework in SCIP

Stephen J. Maher

University of Exeter,

@sj\_maher

s.j.maher@exeter.ac.uk

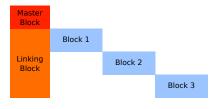
October 1, 2021

#### Structured mixed integer programming

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

min  $c^{\top}x + d^{\top}y$ , subject to  $Ax \ge b$ ,  $Bx + Dy \ge g$ ,  $x \ge 0$ ,  $y \ge 0$ ,  $x \in \mathbb{Z}^{p_1} \times \mathbb{R}^{n_1 - p_1}$ ,  $y \in \mathbb{Z}^{p_2} \times \mathbb{R}^{n_2 - p_2}$ .

# Decomposition methods for mixed integer programming



Linking variables

#### Variable decomposition

- Existence of a set of linking variables
- Exploits property of restriction, i.e. blocks are "easy" to solve after fixing variables
- > Parallelisation: each block can be solved in parallel.

# Benders' decomposition Original problem

min  $c^{\top}x + d^{\top}y$ , subject to  $Ax \ge b$ ,  $Bx + Dy \ge g$ ,  $x \ge 0$ ,  $y \ge 0$ ,  $x \in \mathbb{Z}^{p_1} \times \mathbb{R}^{n_1 - p_1}$ ,  $y \in \mathbb{R}^{n_2}$ .

$$egin{array}{lll} \min & c^ op x + f(x), \ {
m subject to} & {\cal A}x \geq b, \ & x \geq 0, \ & x \in \mathbb{Z}^{p_1} imes \mathbb{R}^{n_1 - p_1}. \end{array}$$

where

$$f(x) = \min_{y \ge 0} \{ d^\top y \mid Bx + Dy \ge g, \ y \in \mathbb{R}^{n_2} \}$$

$$\begin{array}{ll} \min & c^\top x + f(x),\\ \text{subject to} & Ax \geq b,\\ & x \geq 0,\\ & x \in \mathbb{Z}^{p_1} \times \mathbb{R}^{n_1 - p_1}. \end{array}$$

where

$$f(x) = \min_{y \ge 0} \{ d^\top y \, | \, Bx + Dy \ge g, \, y \in \mathbb{R}^{n_2} \}$$

equivalently, using the dual formulation we can define

$$f'(x) = \max_{u \ge 0} \{ u^{ op}(g - Bx) | D^{ op} u \ge d^{ op}, u \in \mathbb{R}^{m_2} \}$$
  
 $(f'(x) = f(x))$ 

Using the dual formulation of f(x), given by

$$f'(x) = \max_{u \ge 0} \{ u^{ op}(g - Bx) \, | \, D^{ op} u \ge d^{ op}, \, u \in \mathbb{R}^{m_2} \}$$

let

- $\mathcal{O}$  be the set of all extreme points of f'(x)
- $\mathcal{F}$  be the set of all extreme rays of f'(x)

an equivalent formulation of the original problem is

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

- $\blacktriangleright$  The sets  ${\mathcal O}$  and  ${\mathcal F}$  are exponential in size
- The reformulated original problem becomes intractable

- $\blacktriangleright$  The sets  ${\cal O}$  and  ${\cal F}$  are exponential in size
- The reformulated original problem becomes intractable

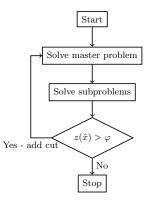
#### Need to use a delayed constraint generation algorithm

- $\blacktriangleright$  The sets  ${\cal O}$  and  ${\cal F}$  are exponential in size
- The reformulated original problem becomes intractable
- Need to use a delayed constraint generation algorithm

Cut generating LP  $\Leftrightarrow$  Benders' subproblem

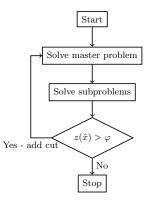
$$egin{aligned} &z(\hat{x}) = \min \quad d^{ op}y, \ & ext{subject to} \quad Dy \geq g - B\hat{x}, \ &y \geq 0, \ &y \in \mathbb{R}^{n_2}. \end{aligned}$$

# Standard Benders' implementation



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

# Standard Benders' implementation

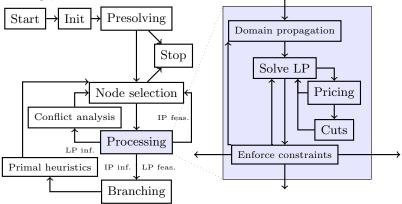


- Easy to understand and simple to implement.
- Not always effective, large overhead in repeatedly solving master problem.
- Easily parallelisable. All subproblems can be solved in parallel.
- Not always efficient—master problem is still solved sequentially.

# Branch-and-cut

- 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.

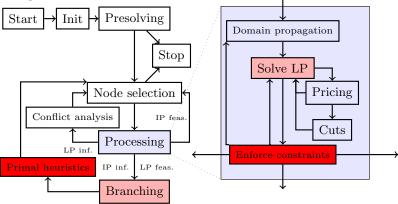
#### Solving process



# Branch-and-cut

- 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.

#### Cut generation - Branch-and-cut



# Branch-and-cut - parallelisation

- Many software implementations for the parallelisation of branch-and-cut, such as the UG framework.
- Useful if solving the subproblems sequentially is not time consuming.
- Addresses issues related to difficult master problem.

# Hybrid parallelisation

- Parallelise the branch-and-cut tree search using the UG framework (distributed memory).
- Parallelise the solving of the subproblems using OpenMP (shared memory).

# Hybrid parallelisation

- Parallelise the branch-and-cut tree search using the UG framework (distributed memory).
- Parallelise the solving of the subproblems using OpenMP (shared memory).

#### Benefits

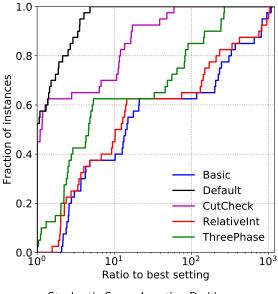
- Parallelisation of tree search avoids waiting for difficult master problem solves.
- Solving the subproblems in parallel takes advantage of available cores at each node.
- Able to balance the effort between master problem tree search and subproblem solving.

#### Current status and features

Both tree search and subproblem parallelisation available.

- Tree search parallelisation activated by enabling Benders' framework.
- Subproblem parallelisation enabled by setting the number of threads. SCIP must be built with OpenMP.
- Customisable sorting of subproblems for load balancing
  - Prioritise subproblems with less calls, then by average number of LP iterations.
- Parallelisation relies on transfer of Benders' cuts between solvers. By calling SCIPstoreBendersCut in custom Benders' cuts plugins, custom Benders' decomposition implementations can be parallelised.

# Current challenges



Stochastic Server Location Problem

# Design Issues

- Each Benders' subproblem is implemented as a SCIP instance.
- With a large number of subproblems memory consumption can be very high.
- When parallelising the tree search, the subproblems must be copied to every solver.
  - Large number of solvers and large number of subproblems results in a very high memory consumption.

# Design Issues

- Each Benders' subproblem is implemented as a SCIP instance.
- With a large number of subproblems memory consumption can be very high.
- When parallelising the tree search, the subproblems must be copied to every solver.
  - Large number of solvers and large number of subproblems results in a very high memory consumption.

### Memory saving mode

- Subproblems, especially in the context of stochastic programming, may have very similar structures.
  - differences only in the constraint matrix or objective function coefficients or in the RHS.
- Create a SCIP instance per each thread. Using a subproblem difference create each subproblem on the fly.

#### Disadvantage

- Only applicable when subproblems have similar structures
- Does not benefit from warm starting between subproblem solves
- Creating and destroying subproblems is time consuming

# Load balancing

- Current best balance of shared and distributed memory unclear
- Benders' cuts are not generated at all nodes in the tree, so reserving threads for subproblem solving may be inefficient.
- Automatic load balancing would allow for idle threads to be used for alternative purposes.

# Partial node processing

- Only solve a subset of subproblems at each node in the tree. Delaying the complete processing of the node.
- The node is not fully evaluated, but enough cuts may be generated to improve the bound.
- Identify the balance of the number of subproblems to solve to gain sufficient bound improvement.

# Key points

- Hybrid parallel implementation of Benders' decomposition available in SCIP
- Current version is available to solve smaller scale problems.
- Future development will reduce the memory consumption of the Benders' framework, enabling its use on large problems and large computational resources.
- Improved load balancing and partial node processing will be investigated.