

## Contributions

- Comparing the performance of the known classes of **valid inequalities** for **mixed integer bilevel linear optimization problems** (MIBLPs) within the **MibS** open source solver [6],[3]
- Suggestion of a **new method** that can be used to
  - strengthen** the known cuts by generating **supporting** valid inequalities
  - compare** the strength of various valid inequalities
  - develop** a **new family of cuts** for MIBLPs

## Problem Definition

The general form of an MIBLP is

$$\min \{cx + d^1y \mid x \in X, y \in \mathcal{P}_1(x) \cap \mathcal{P}_2(x) \cap Y, d^2y \leq \phi(b^2 - A^2x)\},$$

(MIBLP)

where

### First-level feasible region

$$\mathcal{P}_1(x) = \{y \in \mathbb{R}_+^{n_2} \mid A^1x + G^1y \geq b^1\}$$

### Second-level feasible region

$$\mathcal{P}_2(x) = \{y \in \mathbb{R}_+^{n_2} \mid G^2y \geq b^2 - A^2x\}$$

### Second-level value function

$$\phi(\beta) = \min \{d^2y \mid G^2y \geq \beta, y \in Y\} \quad \forall \beta \in \mathbb{R}^{m_2}. \quad (\text{VF})$$

- Integrality Constraints:** Sets  $X = \mathbb{Z}_+^{r_1} \times \mathbb{R}_+^{n_1-r_1}$  and  $Y = \mathbb{Z}_+^{r_2} \times \mathbb{R}_+^{n_2-r_2}$  represent integrality constraints.

Further related definitions are as below.

- Linking Variables:** The set of indices of first-level variables with non-zero coefficients in the second-level problem ( $x_L$ ).

### Bilevel feasible region

$$\mathcal{F} = \{(x, y) \in X \times Y \mid y \in \mathcal{P}_1(x) \cap \mathcal{P}_2(x), d^2y \leq \phi(b^2 - A^2x)\}$$

### Relaxation Problem

- Removing the **optimality constraint** of the second-level problem

$$\mathcal{S} = \{(x, y) \in \mathbb{R}_+^{n_1 \times n_2} \mid x \in X, y \in \mathcal{P}_1(x) \cap \mathcal{P}_2(x) \cap Y\}$$

- Removing the **optimality constraint** of the second-level problem and the **integrality constraints**

$$\mathcal{P} = \{(x, y) \in \mathbb{R}_+^{n_1 \times n_2} \mid y \in \mathcal{P}_1(x) \cap \mathcal{P}_2(x)\}$$

MibS uses  $\mathcal{P}$  as the relaxation problem.

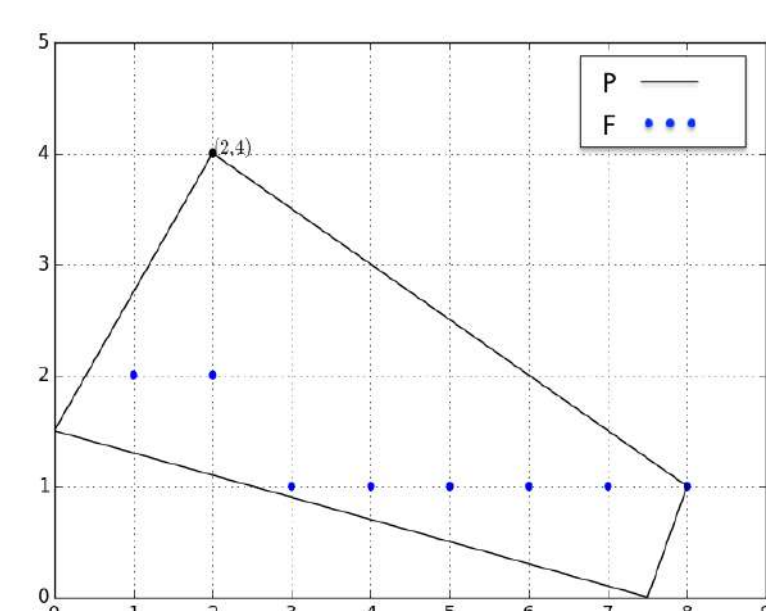


Figure: The feasible region of IBLP[5]

## Valid Inequalities

The set of valid inequalities for MIBLPs can be classified based on the goals of generating them.

### Feasibility cuts

- Goal:** Removing  $(\bar{x}, \bar{y}) \notin X \times Y$
- These valid inequalities are **violated** by  $(\bar{x}, \bar{y})$ , but they are **valid** for  $\mathcal{S}$ .
- This set includes all valid inequalities work for the mixed integer optimization problems(MIPs).

### Optimality cuts

- Goal:** Removing  $(\bar{x}, \bar{y}) \in X \times Y$ , but  $d^2\bar{y} > \phi(b^2 - A^2\bar{x})$
- These valid inequalities are **violated** by  $(\bar{x}, \bar{y})$ , but they are **valid** for  $\mathcal{F}$ .
- This set includes

#### 1 Benders cut

- Assumptions
- $-x_L \subseteq \mathbb{B}^L$ .
  - Corresponding to each linking variable  $x_i$ , there exists  $y_i$  so that  $x_i = 1$  results  $y_i = 0$ .
  - $-G^2 \leq 0$ .

#### 3 Intersection cut(types I and II)[4]

- Assumptions
- $-A^2x + G^2y - b^2 \in \mathbb{Z}$  for all  $(x, y) \in \mathcal{S}$ .
  - $-Y = \mathbb{Z}^{m_2}$  and  $d^2 \in \mathbb{Z}^{m_2}$ .

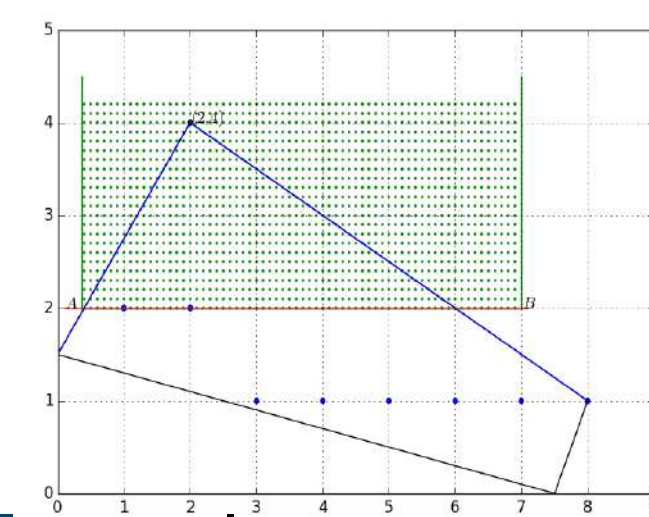


Figure: Intersection cut

#### 2 Increasing objective cut[1]

- Assumptions
- $-x_L \subseteq \mathbb{B}^L$ .
  - $-A^2 \leq 0$ .

#### 4 Integer no-good cut[2]

- Assumptions
- $-X = \mathbb{Z}^{n_1}$  and  $Y = \mathbb{Z}^{n_2}$ .
  - $-b^1, b^2, A^1, A^2, G^1$  and  $G^2$  are discrete.

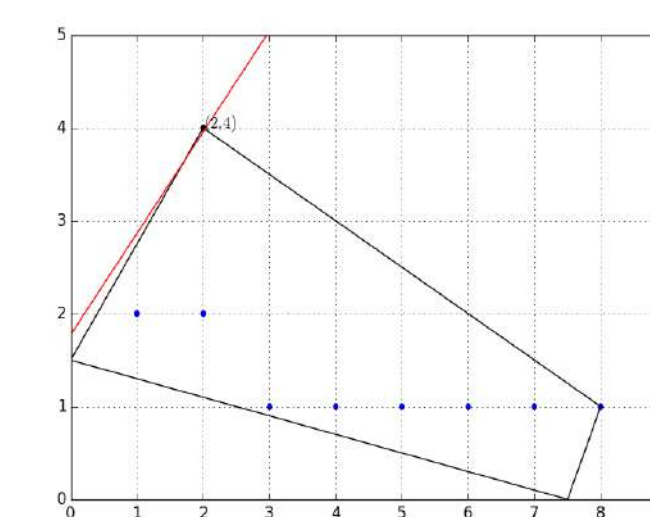


Figure: Integer no-good cut

#### 5 Watermelon intersection cut[4]

- Assumptions
- $-A^2x + G^2y - b^2 \in \mathbb{Z}$  for all  $(x, y) \in \mathcal{S}$ .
  - $-d^2 \in \mathbb{Z}^{m_2}$ .

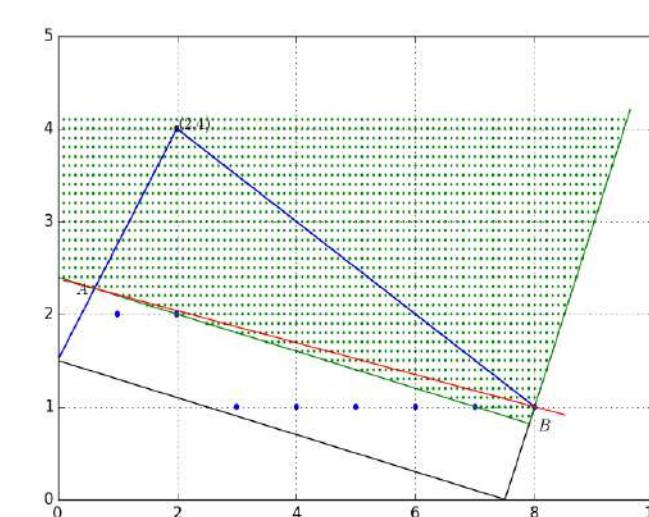


Figure: Watermelon intersection cut

#### 6 Hypercube intersection cut[4]

- Assumption
- $-x_L \subseteq \mathbb{Z}^L$ .

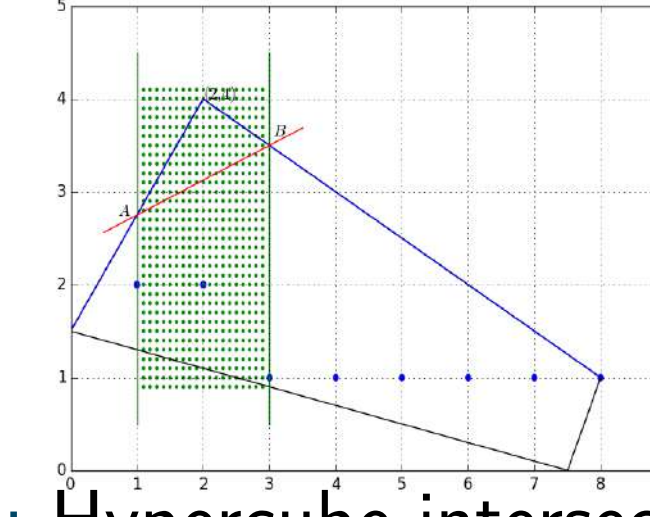


Figure: Hypercube intersection cut

### Projected optimality cuts

- Goal:** Removing all  $(x, y) \in \mathcal{P}$  with  $x_L = \lambda \in \mathbb{Z}^L$
- These valid inequalities are **violated** by  $(x, y) \in \mathcal{P}$  with  $x_L = \gamma \in \mathbb{Z}^L$ , but are **valid** for  $\text{conv}(\{(x, y) \in \mathcal{F} \mid cx + d^1y < U\})$ , where  $U$  represents the current incumbent.
- This set includes the **generalized no-good cut** which works for the problems with  $x_L \subseteq \mathbb{B}^L$ .

## Strengthening the Valid Inequalities

- Idea:** The valid inequality  $\alpha x \geq \beta$  can be **strengthened** by **increasing**  $\beta$ .
- The **best value** of rhs can be obtained by solving

$$\min_{(x,y) \in \mathcal{F}^t} \alpha x \quad (1)$$

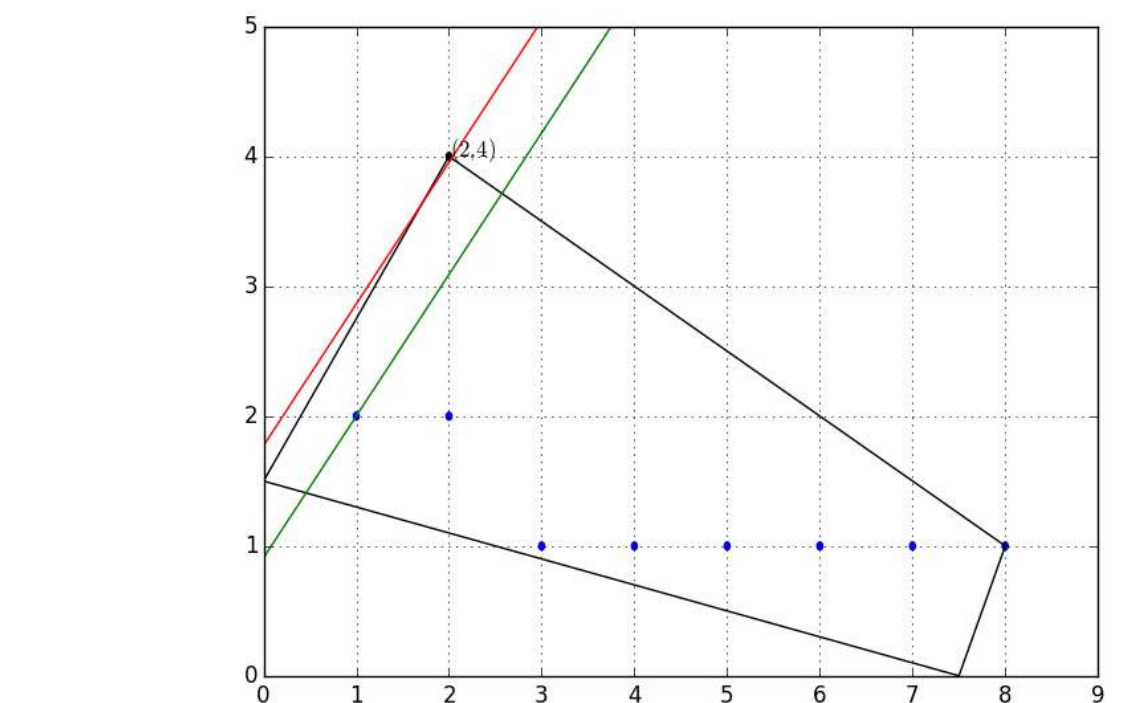


Figure: Improving integer no-good cut

- The strength of different cuts can be compared by measuring the **difference** between the rhs of each cut with its best value.
- Mixed integer no-good cut**
  - The **left-hand-side** employs the idea of int no-good cut.
  - The **rhs** is obtained by solving (1).
  - The **only** assumption is  $x_L \subseteq \mathbb{Z}^L$ .

## Computational Results

Two data sets were employed in experiments: INTERD-DEN, IBLP-DEN

### Performance of different cuts

- Data set: INTERD-DEN

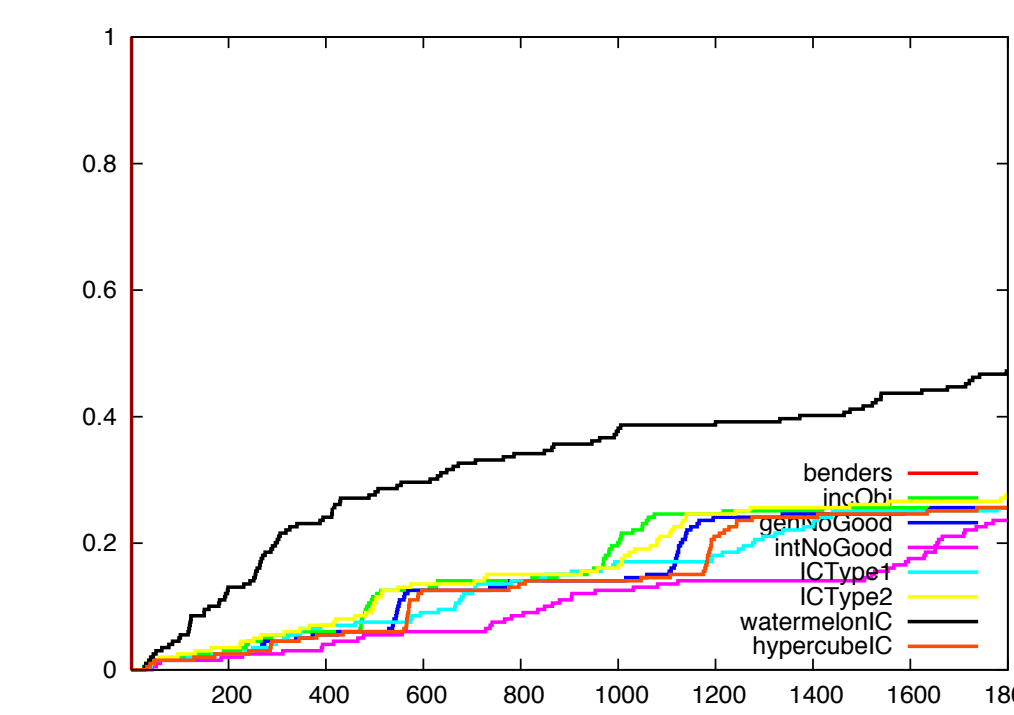


Figure: Perf profile for solution time

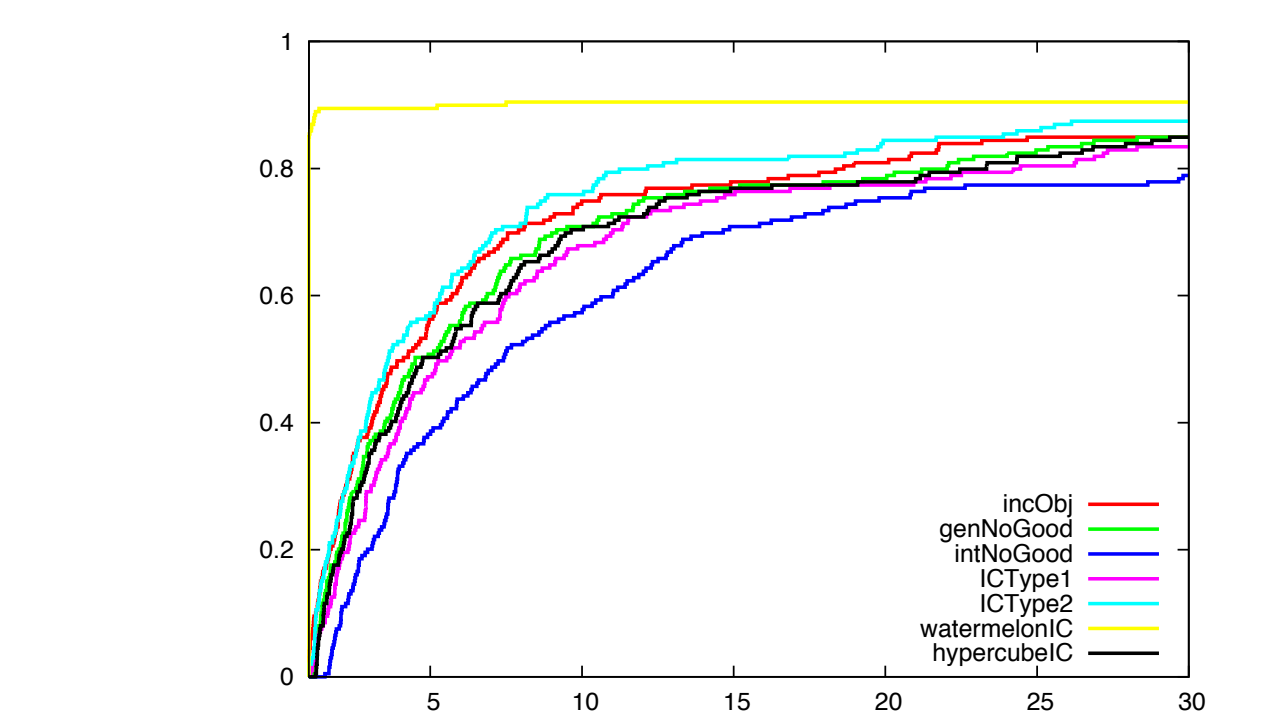


Figure: Perf profile for solution time (without benders cut)

- Data set: IBLP-DEN

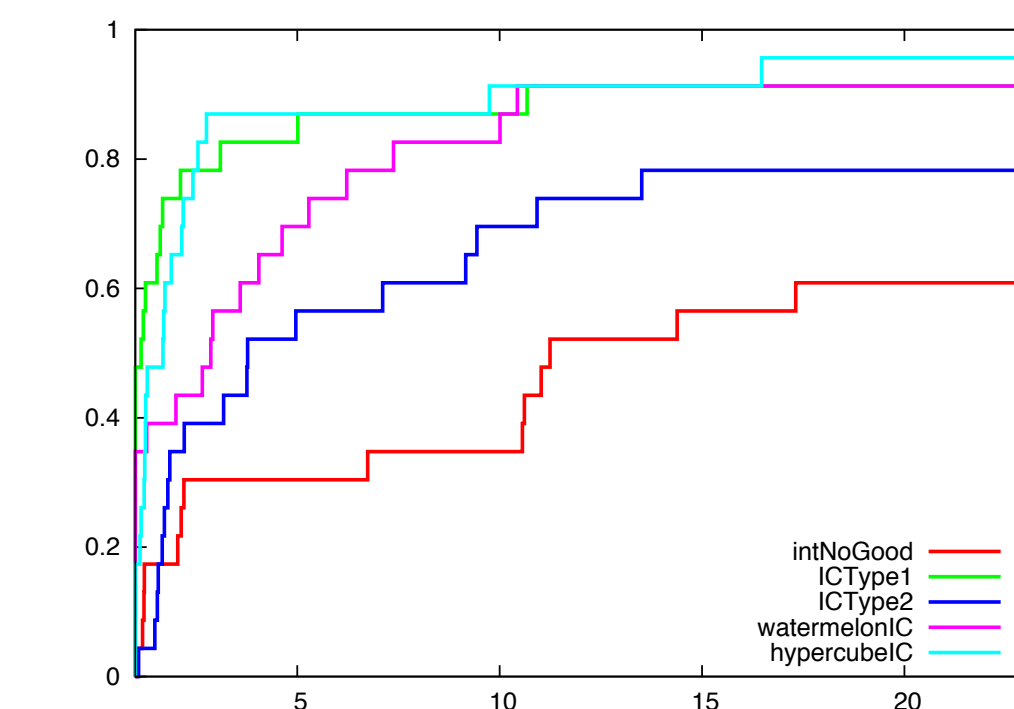


Figure: Perf profile for solution time

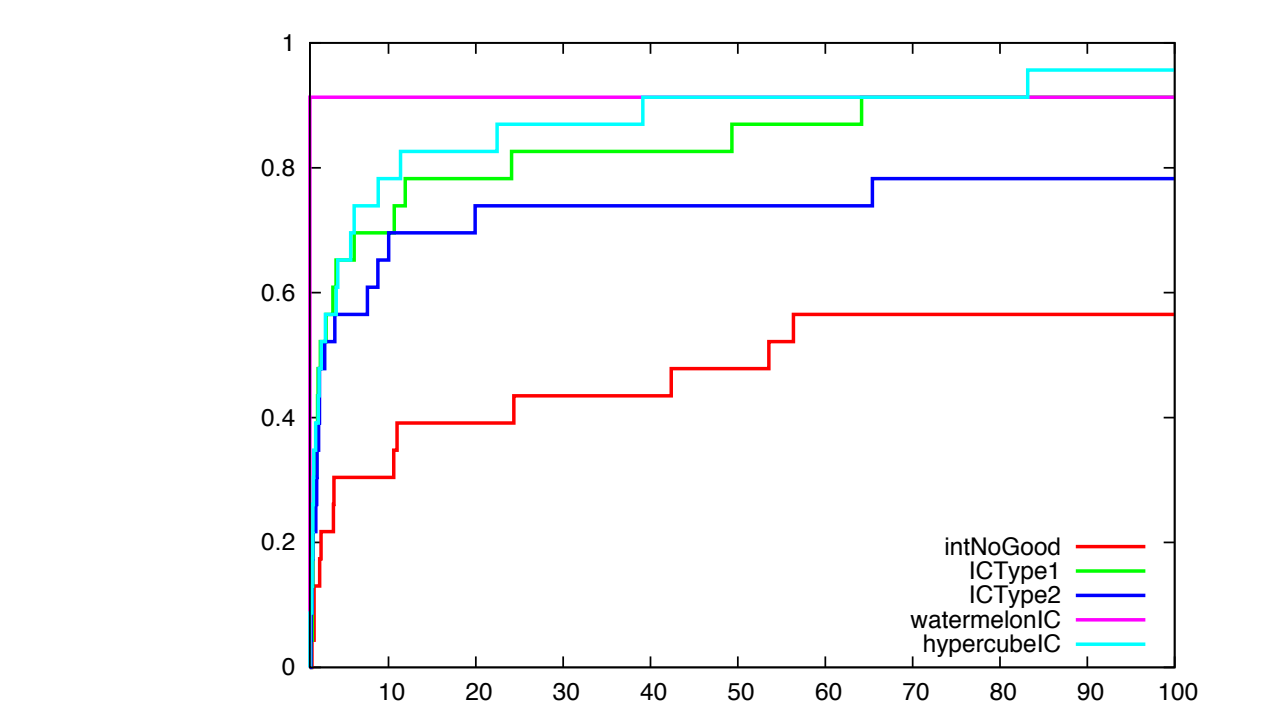


Figure: Perf profile for number of nodes

### Impact of improving the rhs of integer no-good cut

- Data set: INTERD-DEN

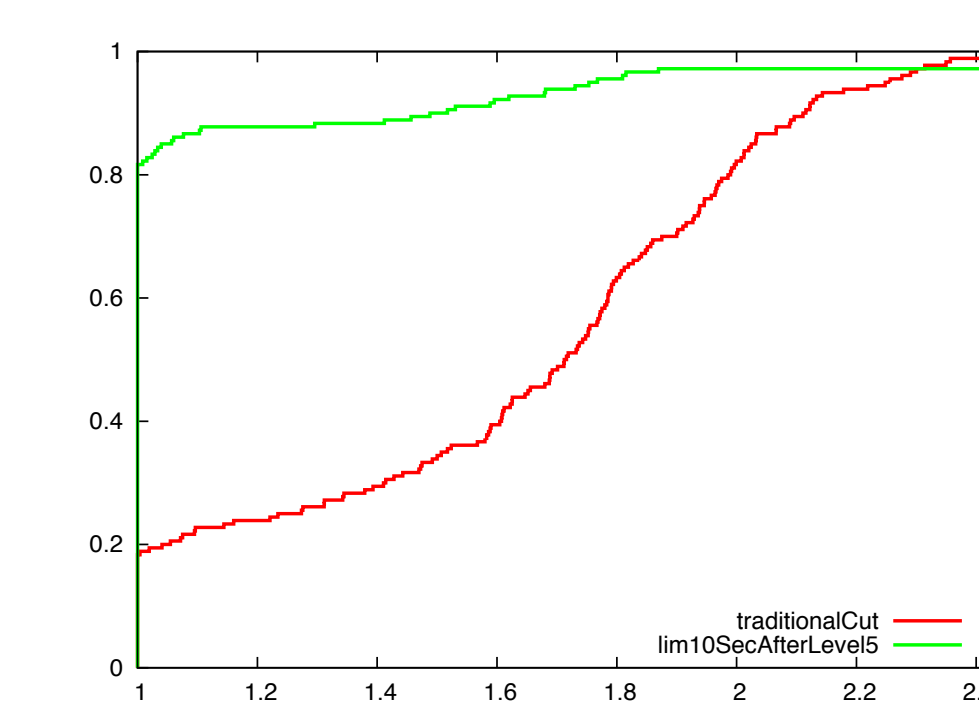


Figure: Perf profile for solution time

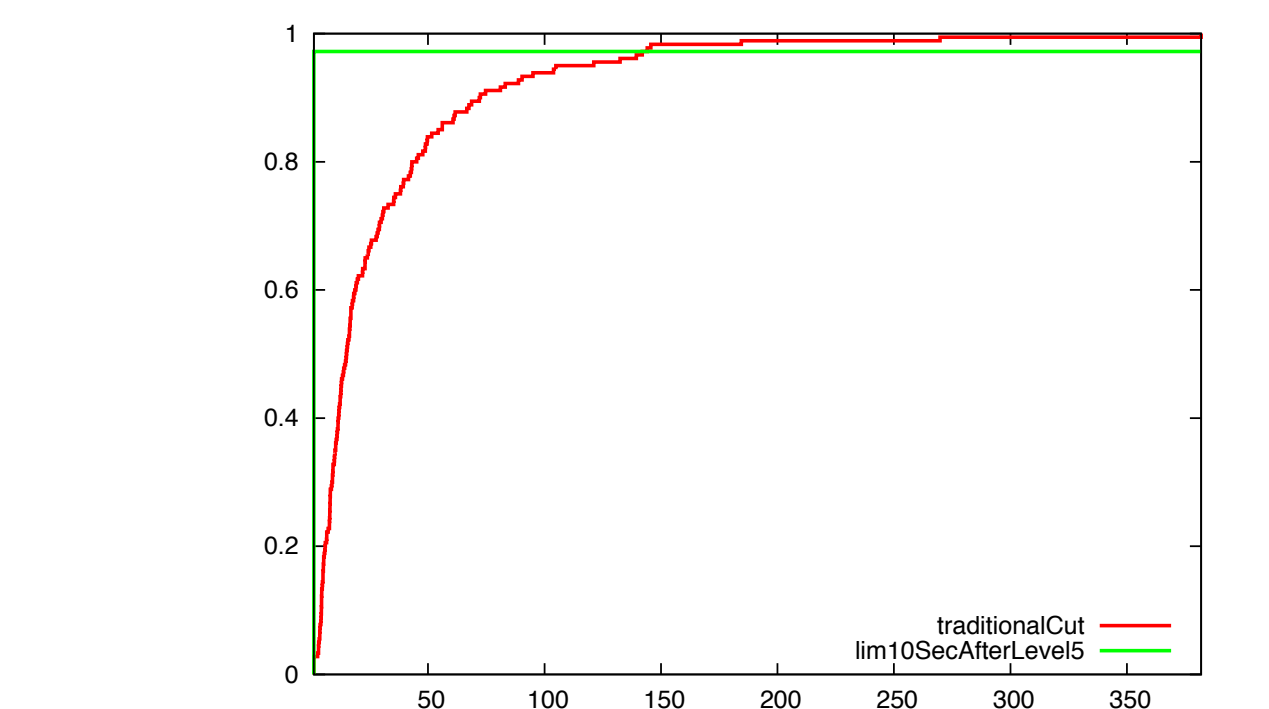


Figure: Perf profile for number of nodes

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