Benders' decomposition: Fundamentals and implementations

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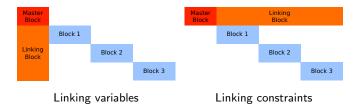
Part 1

Fundamentals

Mixed integer programming

$$\begin{array}{ll} \min & \bar{c}^{\top}\bar{x},\\ \text{subject to} & \bar{A}\bar{x}\geq\bar{b},\\ & \bar{x}\geq0,\\ & \bar{x}\in\mathbb{Z}^{p}\times\mathbb{R}^{n-p}, \end{array}$$

Decomposition methods for mixed integer programming



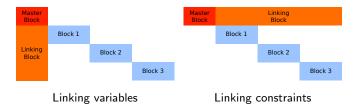
Constraint decomposition

Existence of a set of linking constraints

Variable decomposition

Existence of a set of linking variables

Decomposition methods for mixed integer programming



Constraint decomposition

- Existence of a set of linking constraints
- Exploits property of relaxation, i.e. blocks exhibit structure after relaxation, such as network flow or knapsack.

Variable decomposition

- Existence of a set of linking variables
- Exploits property of restriction, i.e. blocks are "easy" to solve after fixing variables

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

Solving structured mixed integer programs - Resources

- D. Bertsimas and J. N. Tsitsiklis. Introduction to Linear Optimization, 1997.
- J. F. Benders. Partitioning procedures for solving mixed-variables programming problems, Numerische Mathematik, 1962, 4, 238-252.
- R. Rahmaniani, T. G. Crainic, M. Gendreau, and W. Rei. The Benders decomposition algorithm: A literature review. European Journal of Operational Research, 2017, 259, 801-817.
- A. Maheo. A Short Introduction to Benders. https: //arthur.maheo.net/a-short-introduction-to-benders/.

Benders' decomposition Original problem

$$\begin{array}{ll} \min & c^{\top}x + d^{\top}y,\\ \text{subject to} & Ax \geq b,\\ & Bx + Dy \geq g,\\ & x \geq 0,\\ & y \geq 0,\\ & x \in \mathbb{Z}^{p_1} \times \mathbb{R}^{n_1 - p_1},\\ & y \in \mathbb{R}^{n_2}. \end{array}$$

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

where

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

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where

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

The dual formulation of f(x) is important for Benders' decomposition. Can you write down the dual formulation?

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

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where

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

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

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

an equivalent formulation of the original problem is

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

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an equivalent formulation of the original problem is

min
$$c^{\top}x + \varphi$$
,
subject to $Ax \ge b$,
 $\varphi \ge f'(x)$
 $x \ge 0$,
 $x \in \mathbb{Z}^{p_1} \times \mathbb{R}^{n_1 - p_1}$

Note that the feasible region of f'(x) does not depend on x,

only the objective function depends on the input value of x

- Thus, we can describe f'(x) as a set of extreme points and extreme rays.
 - Equivalently, we can describe f(x) as a set of dual extreme points and dual extreme rays

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)

Can you write down the expressions for the optimality and feasibility cuts?

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

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let

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an equivalent formulation of the original problem is

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

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

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Need to use a delayed constraint generation algorithm

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- The reformulated original problem becomes intractable
- Need to use a delayed constraint generation algorithm

Cut generating LP \Leftrightarrow Benders' subproblem

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

Benders' master problem

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

O is replaced by *O'* (which is a subset of *O*). *F* is replaced by *F'* (which is a subset of *F*).

Benders' subproblem

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

If \hat{x} induces

an infeasible instance, then the dual ray u is used to generate a feasibility cut

$$0 \ge u^{ op}(g - Bx)$$

a feasible instance, then the dual solution u is used to generate an optimality cut

$$\varphi \geq u^{\top}(g - Bx)$$

the auxiliary variable φ is an underestimator of the optimal subproblem objective value

Benders' subproblem – discrete variables (Note: master problem must be pure binary)

$$egin{aligned} & z(\hat{x}) = \min \quad d^{ op}y, \ & \text{subject to} \quad Dy \geq g - B\hat{x}, \ & y \geq 0, \ & y \in \mathbb{Z}^{p_2} imes \mathbb{R}^{n_2 - p_2}. \end{aligned}$$

Define the index set $\mathcal{B}^+ := \{i | \hat{x}_i = 1\}$. If \hat{x} induces

an infeasible instance, then the add no-good cut

$$\sum_{i\in\mathcal{B}^+}(1-x_i)+\sum_{i
otin\mathcal{B}^+}x_i\geq 1$$

a feasible instance, then add a Laporte and Louveaux optimality cut (L is a valid lower bound of z(x̂))

$$arphi \geq L + \left(\hat{z}(\hat{x}) - L
ight) \left(1 - \left(\sum_{i \in \mathcal{B}^+} (1 - x_i) + \sum_{i \notin \mathcal{B}^+} x_i
ight)
ight)$$

Exposes an iterative delayed constraint generation algorithm

- 1. Find an \hat{x}
- 2. Solving subproblem using \hat{x} as input
- 3. Add optimality/feasibility cut to master problem to eliminate \hat{x} .

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Key questions

- When to terminate?
- When to solve the Benders' subproblem to generate cuts?
- What solution \hat{x} should be used?
- How to best used MIP solvers to boost iterative algorithm performance?

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Key questions

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 \rightarrow Algorithm design decisions and enhancement techniques