Logic-based Benders Decomposition

A short Introduction

Martin Riedler AC Retreat



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2 Motivation



Basic idea

Benders Decomposition (BD) is applicable to LPs having the following shape (S is the feasible set and D_x , D_y are the domains of x and y):

min
$$f(x, y)$$
 (1)

 $(x, y) \in S$
 (2)

 $x \in D_x$
 (3)

 $y \in D_y$
 (4)

First compute a solution \overline{y} w.r.t. the *y*-variables. Then solve the subproblem solely defined on the *x*-variables:

$$\min f(x, \bar{y})$$
(5)
$$(x, \bar{y}) \in S$$
(6)
$$x \in D_x$$
(7)

- Fix part of the variables (master-problem)
- \Rightarrow split larger problem into easier solvable sub-problem(s)
- Solve sub-problem(s)
- Add Benders cut(s) to master-problem
- Go to 1 if cuts have been added
- Optimal solution found

Logic-based Benders Decomposition (LBBD)

• Classical BD [Benders62]

- Sub-problems need to be LPs
- $+\,$ Cuts can be derived via duality theory

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Logic-based Benders Decomposition (LBBD)

• Classical BD [Benders62]

- Sub-problems need to be LPs
- $+\,$ Cuts can be derived via duality theory
- Logic-based BD [Hooker&Ottosson03]
 - $+\,$ Sub-problems can also be IPs, CPs, \ldots
 - $-\,$ No systematic way to identify Benders cuts
 - $\Rightarrow \ {\sf Requires \ problem \ specific \ cuts}$

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A small example

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Machine Scheduling (minimizing makespan) [Hooker07]

Definition

We consider:

- machines $I = \{1, \ldots, m\}$ with capacities C_i
- jobs $J = \{1, \ldots, n\}$ with
 - processing times p_{ij}
 - ▶ resource consumption *c_{ij}*
 - ► due dates *d_j*
- goal: Schedule all jobs on the machines s.t. due dates are not exceeded, resource limits are respected while minimizing the time by which the last job finishes (makespan).

A small example: ILP

Machine Scheduling (minimizing makespan) [Hooker07]

$$\sum_{i \in I} \sum_{t \in \mathbb{N}_{0}: t \leq d_{j} - p_{ij}} (t + p_{ij}) x_{ijt} \leq M \qquad \qquad \forall j \in J \qquad (9)$$

$$\sum_{i \in I} \sum_{t \in \mathbb{N}_{0}: t \leq d_{j} - p_{ij}} x_{ijt} = 1 \qquad \qquad \forall j \in J \qquad (10)$$

$$\sum_{j \in J} \sum_{t' \in \mathbb{N}_{0}: t - p_{ij} < t' \leq t} c_{ij} x_{ijt'} \leq C_{i} \qquad \forall i \in I, \forall t \in \mathbb{N}_{0}: t < T_{max} \qquad (11)$$

$$x_{ijt} \in \{0, 1\} \qquad \qquad \forall t \in \mathbb{N}_{0}: t \leq d_{j} - p_{ij} \qquad (12)$$

 $\min M$

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(8)

A small example: Benders master problem

Machine Scheduling (minimizing makespan) [Hooker07]

$$\min M \tag{13}$$

$$\sum_{i \in I} x_{ij} = 1 \qquad \qquad \forall j \in J \qquad (14)$$

$$x_{ij} \in \{0,1\} \qquad \qquad \forall i \in I, j \in J \tag{15}$$

- We only assign jobs to machines
- The rest is done in the subproblems (for each machine individually)

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A small example: Benders subproblem

Machine Scheduling (minimizing makespan) [Hooker07]

Let J_i^h be the set of jobs assigned to machine *i* in iteration *h*. Then, the subproblem reads as follows:

$$\min M_{ih}$$
(16)
$$\sum_{t \in \mathbb{N}_0: t \le d_j - p_{ij}} (t + p_{ij}) x_{jt} \le M_{ih} \qquad \forall j \in J_i^h$$
(17)
$$\sum_{j \in J_i^h} \sum_{t' \in \mathbb{N}_0: t - p_{ij} < t' \le t} c_{ij} x_{jt'} \le C_i \qquad \forall t \in \mathbb{N}_0: t < T_{max}$$
(18)
$$x_{jt} \in \{0, 1\} \qquad \forall j \in J_i^h, \qquad (19)$$

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A small example: Benders master problem revisited **ac**

Machine Scheduling (minimizing makespan) [Hooker07]

In iteration H the master problem becomes the following:

$$\min M \tag{20}$$

$$\sum_{i \in I} x_{ij} = 1 \qquad \forall j \in J \tag{21}$$

$$M_{ih}^* - \left(\sum_{j \in J_i^h} (1 - x_{ij})p_{ij} + \Delta_i^h\right) \le M \qquad \forall i \in I, h = 1, \dots, H - 1 \tag{22}$$

$$\frac{1}{C_i} \sum_{j \in J} c_{ij}p_{ij}x_{ij} \le M \qquad \forall i \in I \tag{23}$$

$$x_{ij} \in \{0, 1\} \qquad \forall i \in I, j \in J \tag{24}$$

 $\left(\Delta_{i}^{h}=\max_{j\in J_{i}^{h}}\left\{d_{j}\right\}-\min_{j\in J_{i}^{h}}\left\{d_{j}\right\}\right)$

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When to consider LBBD?



• Sometimes, such formulations are somehow natural

- Vehicle Routing (cluster-first route-second)
- Scheduling (first assign jobs to machines, then solve per machine)
- Cutting and Packing (first select items to pack, then find the packing)
- By decomposition the individual problems usually get smaller \Rightarrow otherwise too large problems can be handled
- Sometimes decomposition allows to apply other algorithmic techniques for a certain part of the problem (e.g. CP)
- By dealing with the problem in two stages one can avoid modeling difficulties (e.g. quadratic constraints)



- Metaheuristics have been applied with great success for solving the master problem.
- Metaheuristics can also be used for solving difficult subproblems, helping in deriving Benders cuts.
- Utilize a two-stage approach:
 - first solve master or subproblems heuristically
 - ► then complete verification steps to preserve optimality

Thank you for your Attention! Questions?