

# Decomposition Part, Week 1

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## We will Cover:

- Dantzig-Wolfe decomposition and column generation in IP.
- 2 applications of DW decomposition: maximum independent set problem, multistage stochastic IP.
- Lagrangian relaxation (this week or next week).
- Formulation of Benders decomposition.
- Global optimization of nonconvex MINP with decomposable structures.
- Any suggestions? Additions? Not interesting?
- Volunteers?
- **Suggestions for the structure of the seminar?**

# Decomposition In General

- 1 DW Decomposition
- 2 Lagrangian relaxation
- 3 Benders decomposition

- All are large scale (models with large number of constraints/variables) optimization algorithms.
- We can find better bounds for branch and bound using these approaches.
- For the first 2, alternative approaches are cutting plane, variable redefinition.
- Basically, first 2 decomposition includes 3 steps:
  - Decompose system of inequalities into two parts.
  - Find the convex hull of the second system.
  - optimize first system over this convex hull.

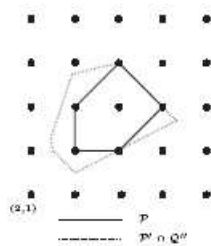
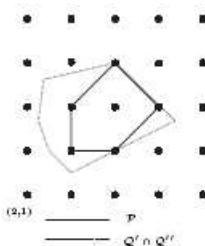
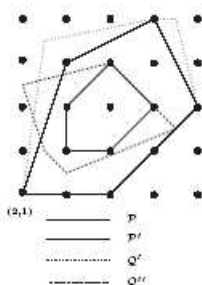
# Why decomposition can find a better bound?

- From T. Ralphs and M. Galati, Decomposition in Integer Programming, in Integer Programming: Theory and Practice, John Karlof, ed. (2005), 57.

$$\begin{aligned} & \min cx \\ & \text{s.t. } Ax \leq b \\ & \quad x \in \mathbb{Z}^n \end{aligned}$$

- Let  $Q = \{x \in \mathbb{R}^n \mid Ax \leq b\} = \{x \in \mathbb{R}^n \mid A''x \leq b'', A'x \leq b'\}$ .
- Let  $F = Q \cap \mathbb{Z}^n$  (feasible integer points) and  $P$  be the convex hull of  $F$ .
- Let  $Q' = \{x \in \mathbb{R}^n \mid A'x \leq b'\}$ ,  $F' = Q' \cap \mathbb{Z}^n$  (feasible integer points) and  $P'$  be the convex hull of  $F'$ .
- Let  $Q'' = \{x \in \mathbb{R}^n \mid A''x \leq b''\}$ .
- $ZLP = \min_{x \in Q} \{cx\}$ .
- $ZDECOMP = \min_{x \in P'} \{cx \mid A''x \leq b''\} = \min_{x \in P' \cap Q''} \{cx\}$ .

# Why decomposition can find a better bound?



Source: T. Ralphs and M. Galati, Decomposition in Integer Programming, in Integer Programming: Theory and Practice, John Karlof, ed. (2005), 57.

# Dantzig Wolfe Decomposition

- Assume  $P'$  is bounded. Let  $E \subseteq F'$  be the set of *extreme* points of  $P'$ .

$$P' = \{x \in \mathbb{R}^n \mid x = \sum_{s \in \mathcal{E}} s\lambda_s, \sum_{s \in \mathcal{E}} \lambda_s = 1, \lambda_s \geq 0 \forall s \in \mathcal{E}\}.$$

- Dantzig Wolfe formulation and bound:

$$z_{DW} = \min_{x \in \mathbb{R}^n} \{c^T x \mid A''x \geq b'', x = \sum_{s \in \mathcal{E}} s\lambda_s, \sum_{s \in \mathcal{E}} \lambda_s = 1, \lambda_s \geq 0 \forall s \in \mathcal{E}\}.$$

- Substitute new variables:

$$z_{DW} = \min_{\lambda \in \mathbb{R}_+^{\mathcal{E}}} \{c^T (\sum_{s \in \mathcal{E}} s\lambda_s) \mid A''(\sum_{s \in \mathcal{E}} s\lambda_s) \geq b'', \sum_{s \in \mathcal{E}} \lambda_s = 1\}.$$

- $|E|$  may be very large, thus  $E$  should be generated dynamically.

Source: T. Ralphs and M. Galati, Decomposition in Integer Programming, in Integer Programming: Theory and Practice, John Karlof, ed. (2005), 57.

## Cutting Strips Problem

$$\begin{aligned} \min \quad & \sum_{k=1}^K L^k \left( W^k y^k - \sum_{i=1}^p w_i z_i^k \right) \\ \text{[PCS]} \quad \text{s.t.} \quad & \end{aligned} \quad (2)$$

$$\sum_{k=1}^K L^k z_i^k \geq d_i, \quad i = 1, \dots, p,$$

$$\sum_{i=1}^p w_i z_i^k \leq W^k y^k, \quad k = 1, \dots, K, \quad (3)$$

$$y^k \in \{0, 1\}, \quad k = 1, \dots, K, \quad (4)$$

$$z_i^k \in \mathbb{N}, \quad i = 1, \dots, p, \quad k = 1, \dots, K. \quad (5)$$

Source: Vanderbeck F., On Dantzig-Wolfe Decomposition in Integer Programming and ways to Perform Branching in a Branch-and-Price Algorithm, Oper. Res. 48-1, 2000, pg. 111.

## Integer Points in Second System

- Let  $Q^k$  be the set of feasible cutting patterns for sheet  $k$ ,  $\lambda_q^k$  be the number of times pattern  $q^k$  is selected.

$$\begin{aligned}
 X^{cs} &= \left\{ (y^k, z^k)_{k=1, \dots, K} \in \mathbb{N}^{K(1+p)} : \right. \\
 &\quad \left. \sum_{i=1}^p w_i z_i^k \leq W^k y^k \text{ for } k = 1, \dots, K \right\} \\
 &= \left\{ (y^k, z^k)_{k=1, \dots, K} \in \mathbb{R}^{K(1+p)} : \right. \\
 &\quad \left. y^k = \sum_{q^k \in Q^k} \lambda_{q^k}^k, \quad z^k = \sum_{q^k \in Q^k} q^k \lambda_{q^k}^k, \right. \\
 &\quad \left. \sum_{q^k \in Q^k} \lambda_{q^k}^k \leq 1 \quad \forall k, \quad \lambda_{q^k}^k \in \{0, 1\} \quad \forall q^k \in Q^k, k \right\}.
 \end{aligned}$$



## Dantzig Wolfe Decomposition of Cutting Strips Problem

$$\min \sum_{k=1}^K \sum_{q^k \in Q^k} c_q^k \lambda_q^k$$

$[M^{CS}]$  s.t.

$$\sum_{k=1}^K \sum_{q^k \in Q^k} L^k q_i^k \lambda_q^k \geq d_i, \quad i = 1, \dots, p, \quad (6)$$

$$\sum_{q^k \in Q^k} \lambda_q^k \leq 1, \quad k = 1, \dots, K,$$

$$\lambda_q^k \in \{0, 1\}, \quad q^k \in Q^k, k = 1, \dots, K.$$

# Column Generation

- Master problem may include infinitely many cutting patterns:

$$Q^k = \{q^k \in \mathbb{N}^p : \sum_i^p w_i q_i^k \leq W\}$$

- Start with “a” set of feasible cutting patterns  $\rightarrow$  **RMP**.
- Generate the column with most negative reduced cost to improve the bound.
- Let  $\pi_i$  be the dual variable for demand constraint  $i$ ,  $\mu_k$  be the dual variable for convex comb. constr. Subproblem:

$$\zeta^k(\pi, \mu) = \min L^k \left( W^k - \sum_{i=1}^p (w_i + \pi_i) z_i \right) + \mu_k$$

s.t.

$$\sum_{i=1}^p w_i z_i \leq W^k,$$

$$z_i \in \mathbb{N}, \quad i = 1, \dots, p,$$

# Lower bound for DW Decomposition LP

If column generation subproblem is solved to optimality:

$$z_{DW}(LP)^{LB} = \text{(obj. value of dual master)} \\ + \text{(obj. function value of optimal column generation subproblem)}$$

Let  $q$  be the optimal solution to the column generation problem.

$$z_{DW}(LP)^{LB} = \left( \sum_{i=1..p} d_i \pi_i + \sum_{k=1..K} \mu_k \right) + \left( \sum_{k=1..K} c_q q_k - \sum_{k=1..K, i=1..p} L^k q_i^k - \sum_{k=1..K} \mu_k \right)$$

# Some Points

- Lagrangian relaxation and Dantzig Wolfe decomposition is used to find better bounds for the MIP.
- Lagrangian relaxation, Dantzig Wolfe decomposition, cutting plane, variable redefinition can be used to get the same bound.
- Cutting plane: convex hull of second system  $P'$  is defined by finding the facet defining inequalities of the system.
- Master LP (from Dantzig Wolfe Decomp.) = dual formulation of the Lagrangian dual that results from dualizing the  $A''x \leq b''$ .
- Dantzig Wolfe decomposition leads to models with large number of variables which requires column generation algorithm.
- Variable redefinition: develop an alternative formulation  $Z$  for the polyhedron  $P'$ .

# Discretization vs. Convexification

- The first DW example is **convexification**.
- Cutting Strip problem uses **discretization**.
- Convexification:  $\lambda_s > 0$ ,
- Discretization:  $\lambda_s \in \{0, 1\}$  if it is an extreme point,  $\lambda_s \in \mathbb{N}$  if it is an extreme ray.
- Convexification:  $x = \sum_s s\lambda_s$  does not imply  $x$  is integer. To express integrality, must return to original variables. Branching should be in original variables.
- Discretization:  $x = \sum_s s\lambda_s$  results integer variables. Can do branching or write cuts in terms of  $\lambda_s$ .
- Both give the same LP relaxation of master.
- Both has the same IP master if variables are binary.
- Discretization  $\neq$  convexification if some variables are general integer.

## Section 1 (Vanderbeck 2000): Decomposition of IP

- General form of master problem.
- $\sum_s \lambda_s = 1$  can be changed with  $\sum_s \lambda_s \leq 1$  if 0 vector is a feasible solution for the problem.
- **Independent subsystems:**  $Dx \leq d \rightarrow D^k x \leq d^k, k = 1, \dots, K.$

$$\sum_{q \in Q(k)} \lambda_q = 1, k = 1, \dots, K.$$

- **Identical independent subsystems:**  $Dx \leq d \rightarrow D^k x \leq d^k, k = 1, \dots, K,$  but  $D^k = \bar{D}, d^k = \bar{d}$  for  $k = 1..K.$

$$\sum_{q \in \bar{Q}} \lambda_q = K.$$

# Equivalence Between Original and Master Formulation

- Let  $P$  be the original problem,  $M$  be the master problem.
- Both have the same set of feasible integer points.
- Since the representation of solution is different in both formulations, a solution  $x$  for  $P$  may not result a unique solution  $\lambda$  for  $M$ .
- Cases when the solutions do not corresponds to a unique solution is given in the paper.
- $Z_{LP}(P) \leq Z_{LP}(M) \leq Z_{IP}$ .
- $Z_{LP}(P) < Z_{LP}(M)$  if subsystem does not have integrality property.

# Practical Issues with Column Generation

- At each branch and bound node, a **feasible** LP solution is required. Especially with branching it becomes difficult to maintain feasibility. One way is to use artificial variables that will always result feasible solutions.
- Designing **branching rules**: Branching rule must be incorporated into subproblem (column generation problem).
- **RMP**: Optimize RMP to get dual variables. Dual solution is not **unique** if primal is degenerate. Dual variables effect the generated column.
  - Initialization is necessary with simplex and for other methods it can reduce **heading-in effect** of initially producing bad duals which may cause irrelevant columns.
  - Solution with primal, primal-dual simplex, barrier?
  - Simplex based column generation has **tailing off effect** (poor convergence).