

1 Outline

In this lecture, we study

- Dantzig-Wolfe decomposition based on the Lagrangian dual,
- Dantzig-Wolfe decomposition for binary programs,
- Dantzig-Wolfe decomposition for models with block diagonal structure,
- Column generation for the Dantzig-Wolfe reformulation.

2 Dantzig-Wolfe decomposition

Let us consider a mixed integer program

$$\begin{aligned} z_I &= \max && c^\top x \\ &\text{s.t.} && Ax \leq b \\ &&& Ex \leq f \\ &&& x \in \mathbb{Z}_+^d \times \mathbb{R}_+^p. \end{aligned} \tag{MIP}$$

We will learn the **Dantzig-Wolfe decomposition** framework for solving the mixed-integer program.

2.1 Dantzig-Wolfe decomposition based on the Lagrangian dual

Let Q be defined as

$$Q = \left\{ x \in \mathbb{Z}_+^d \times \mathbb{R}_+^p : Ax \leq b \right\}.$$

Assume that Q is nonempty and that A, b have rational entries. Let m be the number of rows of E , and take $\lambda \in \mathbb{R}_+^m$. Remember that we define the **Lagrangian relaxation** of (MIP) with respect to λ as follows.

$$\begin{aligned} z_{\text{LR}}(\lambda) &= \max && c^\top x + \lambda^\top (f - Ex) \\ &\text{s.t.} && Ax \leq b \\ &&& x \in \mathbb{Z}_+^d \times \mathbb{R}_+^p. \end{aligned} \tag{LR}$$

Moreover, recall that the **Lagrangian dual** of the mixed integer program (MIP) is defined as

$$z_{\text{LD}} = \min \{ z_{\text{LR}}(\lambda) : \lambda \geq 0 \}. \tag{LD}$$

We learned that (MIP) and (LD) are related according to the following characterization of (LD).

$$z_{\text{LD}} = \max \left\{ c^\top x : Ex \leq f, x \in \text{conv}(Q) \right\}.$$

Furthermore, by the Minkowski-Weyl theorem, $\text{conv}(Q)$ can be expressed as

$$\text{conv}(Q) = \text{conv}\{v^1, \dots, v^n\} + \text{cone}\{r^1, \dots, r^\ell\}$$

where v^1, \dots, v^n are the extreme points of $\text{conv}(Q)$ and r^1, \dots, r^ℓ are the extreme rays of $\text{conv}(Q)$. Then any point x in $\text{conv}(Q)$ can be written as

$$x = \sum_{k \in [n]} \alpha_k v^k + \sum_{h \in [\ell]} \beta_h r^h$$

for some $\alpha \in \mathbb{R}_+^k$ and $\beta \in \mathbb{R}_+^\ell$ such that

$$\sum_{k \in [n]} \alpha_k = 1.$$

Based on this, it follows that

$$\begin{aligned} z_{\text{LD}} &= \max && \sum_{k \in [n]} (c^\top v^k) \alpha_k + \sum_{h \in [\ell]} (c^\top r^h) \beta_k \\ &\text{s.t.} && \sum_{k \in [n]} (E v^k) \alpha_k + \sum_{h \in [\ell]} (E r^h) \beta_k \leq f \\ &&& \sum_{k \in [n]} \alpha_k = 1 \\ &&& \alpha \in \mathbb{R}_+^k, \beta \in \mathbb{R}_+^\ell. \end{aligned} \tag{DW1}$$

Remember that the Lagrangian dual (**LD**) is a relaxation of (**MIP**). Hence, we refer to (**DW1**) as the **Dantzig-Wolfe relaxation** of (**MIP**). Moreover, we have

$$z_I = \max \left\{ c^\top x : E x \leq f, x \in \text{conv}(Q), x_j \in \mathbb{Z} \forall j \in [d] \right\}.$$

Therefore, we deduce

$$\begin{aligned} z_I &= \max && \sum_{k \in [n]} (c^\top v^k) \alpha_k + \sum_{h \in [\ell]} (c^\top r^h) \beta_k \\ &\text{s.t.} && \sum_{k \in [n]} (E v^k) \alpha_k + \sum_{h \in [\ell]} (E r^h) \beta_k \leq f \\ &&& \sum_{k \in [n]} \alpha_k = 1 \\ &&& \alpha \in \mathbb{R}_+^k, \beta \in \mathbb{R}_+^\ell \\ &&& \sum_{k \in [n]} \alpha_k v_j^k + \sum_{h \in [\ell]} \beta_h r_j^h \in \mathbb{Z}, \quad j \in [d]. \end{aligned} \tag{DW2}$$

Here, the formulation (**DW2**) is referred to as the **Dantzig-Wolfe reformulation** of (**MIP**).

2.2 Dantzig-Wolfe decomposition as the dual of the Lagrangian dual

Recall that the Dantzig-Wolfe decomposition is given by

$$\begin{aligned}
\max \quad & \sum_{k \in [n]} (c^\top v^k) \alpha_k + \sum_{h \in [\ell]} (c^\top r^h) \beta_h \\
\text{s.t.} \quad & \sum_{k \in [n]} (E v^k) \alpha_k + \sum_{h \in [\ell]} (E r^h) \beta_h \leq f \\
& \sum_{k \in [n]} \alpha_k = 1 \\
& \alpha \in \mathbb{R}_+^k, \beta \in \mathbb{R}_+^\ell.
\end{aligned}$$

is the equivalent representation of the Lagrangian dual. Let us take its dual. We use dual variable λ for the inequality constraint and dual variable μ for the equality constraint. Then we deduce

$$\begin{aligned}
\min \quad & \lambda^\top f + \mu \\
\text{s.t.} \quad & \mu + (E v^k)^\top \lambda \geq c^\top v^k, \quad k \in [n] \\
& (E r^h)^\top \lambda \geq c^\top r^h, \quad h \in [\ell] \\
& \lambda \geq 0
\end{aligned}$$

Note that this is equivalent to

$$\begin{aligned}
\min \quad & \lambda^\top f + \mu \\
\text{s.t.} \quad & \mu \geq \max_{k \in [n]} \left\{ (c - E^\top \lambda)^\top v^k \right\} \\
& \lambda \in \text{dom}(z_{\text{LR}})
\end{aligned}$$

because

$$\text{dom}(z_{\text{LR}}) = \left\{ \lambda : (c - E^\top \lambda)^\top r^h \leq 0 \quad \forall h \in [\ell], \lambda \geq 0 \right\}.$$

Eliminating the variable μ , we obtain

$$\begin{aligned}
\min \quad & \lambda^\top f + \max_{k \in [n]} \left\{ (c - E^\top \lambda)^\top v^k \right\} \\
\text{s.t.} \quad & \lambda \in \text{dom}(z_{\text{LR}}).
\end{aligned}$$

This is equivalent to

$$\begin{aligned}
& \min_{\lambda \in \text{dom}(z_{\text{LR}})} \max_{k \in [n]} \left\{ \lambda^\top f + (c - E^\top \lambda)^\top v^k \right\} \\
& = \min_{\lambda \in \text{dom}(z_{\text{LR}})} \underbrace{\max_{k \in [n]} \left\{ c^\top v^k + \lambda^\top (f - E v^k) \right\}}_{z_{\text{LR}}(\lambda)} \\
& = \min \{ z_{\text{LR}}(\lambda) : \lambda \in \text{dom}(z_{\text{LR}}) \} \\
& = z_{\text{LD}}.
\end{aligned}$$

2.3 Dantzig-Wolfe decomposition for pure binary programs

Let us consider a pure binary integer program as follows.

$$\begin{aligned}
 z_I &= \max && c^\top x \\
 &\text{s.t.} && Ax \leq b \\
 &&& Ex \leq f \\
 &&& x \in \{0, 1\}^d.
 \end{aligned} \tag{BP}$$

We define Q as

$$Q = \{x \in \{0, 1\}^d : Ax \leq b\}.$$

Since Q is bounded and finite,

$$Q = \{v^1, \dots, v^n\}.$$

Then any point x in Q can be expressed as

$$x = \sum_{k \in [n]} \alpha_k v^k, \quad \sum_{k \in [n]} \alpha_k = 1, \quad \alpha \in \{0, 1\}^n.$$

Then it follows that

$$\begin{aligned}
 z_I &= \max && \sum_{k \in [n]} (c^\top v^k) \alpha_k \\
 &\text{s.t.} && \sum_{k \in [n]} (E v^k) \alpha_k \leq f \\
 &&& \sum_{k \in [n]} \alpha_k = 1 \\
 &&& \alpha \in \{0, 1\}^n.
 \end{aligned}$$

This formulation is the Dantzig-Wolfe reformulation of (BP). Then the Dantzig-Wolfe relaxation of (BP) is

$$\begin{aligned}
 \max &&& \sum_{k \in [n]} (c^\top v^k) \alpha_k \\
 \text{s.t.} &&& \sum_{k \in [n]} (E v^k) \alpha_k \leq f \\
 &&& \sum_{k \in [n]} \alpha_k = 1 \\
 &&& \alpha \geq 0
 \end{aligned}$$

2.4 Problems with block diagonal structure

We consider the following optimization model

$$\begin{aligned}
\max \quad & c^{1\top} x^1 + c^{2\top} x^2 + \cdots + c^{p\top} x^p \\
\text{s.t.} \quad & A^1 x^1 \leq b^1 \\
& A^2 x^2 \leq b^2 \\
& \vdots \\
& A^p x^p \leq b^p \\
& E^1 x^1 + E^2 x^2 + \cdots + E^p x^p \leq f \\
& x^j \in \{0, 1\}^{d_j}, \quad j \in [p].
\end{aligned} \tag{22.1}$$

For $j \in [p]$, let Q_j be defined as

$$Q_j = \left\{ x^j \in \{0, 1\}^{d_j} : A^j x^j \leq b^j \right\}.$$

Here, Q_j is bounded and finite, so any point x^j in Q_j can be written as

$$x^j = \sum_{v \in Q_j} \alpha_v^j v, \quad \sum_{v \in Q_j} \alpha_v^j = 1, \quad \alpha_v^j \in \{0, 1\}^{|Q_j|}.$$

Therefore, the Dantzig-Wolfe reformulation of (22.1) is given by

$$\begin{aligned}
\max \quad & \sum_{v \in Q_1} (c^{1\top} v) \alpha_v^1 + \sum_{v \in Q_2} (c^{2\top} v) \alpha_v^2 + \cdots + \sum_{v \in Q_p} (c^{p\top} v) \alpha_v^p \\
\text{s.t.} \quad & \sum_{v \in Q_1} (E^1 v) \alpha_v^1 + \sum_{v \in Q_2} (E^2 v) \alpha_v^2 + \cdots + \sum_{v \in Q_p} (E^p v) \alpha_v^p \leq f \\
& \sum_{v \in Q_j} \alpha_v^j = 1, \quad j \in [p] \\
& \alpha_v^j \in \{0, 1\}^{|Q_j|}, \quad j \in [p].
\end{aligned}$$

Then the Dantzig-Wolfe relaxation of (22.1) is given by

$$\begin{aligned}
\max \quad & \sum_{v \in Q_1} (c^{1\top} v) \alpha_v^1 + \sum_{v \in Q_2} (c^{2\top} v) \alpha_v^2 + \cdots + \sum_{v \in Q_p} (c^{p\top} v) \alpha_v^p \\
\text{s.t.} \quad & \sum_{v \in Q_1} (E^1 v) \alpha_v^1 + \sum_{v \in Q_2} (E^2 v) \alpha_v^2 + \cdots + \sum_{v \in Q_p} (E^p v) \alpha_v^p \leq f \\
& \sum_{v \in Q_j} \alpha_v^j = 1, \quad j \in [p] \\
& \alpha_v^j \geq 0, \quad j \in [p].
\end{aligned}$$

Let us consider the special case where

- $c^1 = \cdots = c^p = c$,
- $E^1 = \cdots = E^p = E$,

- $Q^1 = \dots = Q^p = Q$.

Then in the Dantzig-Wolfe relaxation, we may set

$$\alpha = \alpha^1 + \alpha^2 + \dots + \alpha^p.$$

As a result, the Dantzig-Wolfe relaxation becomes

$$\begin{aligned} \max \quad & \sum_{v \in Q} (c^\top v) \alpha_v \\ \text{s.t.} \quad & \sum_{v \in Q} (Ev) \alpha_v \leq f \\ & \sum_{v \in Q} \alpha_v = p \\ & \alpha \geq 0. \end{aligned}$$

3 Column generation for solving the Dantzig-Wolfe reformulation

The Dantzig-Wolfe relaxation (DW2) has variables $\alpha_1, \dots, \alpha_n$ for the extreme points of Q and variables $\beta_1, \dots, \beta_\ell$ for the extreme rays of Q . Therefore, n and ℓ are potentially very large. In this case, we may apply the column generation technique. Recall that the dual of (DW2) is given by

$$\begin{aligned} \min \quad & \lambda^\top f + \mu \\ \text{s.t.} \quad & \mu + (Ev^k)^\top \lambda \geq c^\top v^k, \quad k \in [n] \\ & (Er^h)^\top \lambda \geq c^\top r^h, \quad h \in [\ell] \\ & \lambda \geq 0. \end{aligned}$$

The column generation procedure works as follows. We start with $N \subseteq [n]$ and $L \subseteq [\ell]$. Then we have the master problem

$$\begin{aligned} \max \quad & \sum_{k \in N} (c^\top v^k) \alpha_k + \sum_{h \in L} (c^\top r^h) \beta_h \\ \text{s.t.} \quad & \sum_{k \in N} (Ev^k) \alpha_k + \sum_{h \in L} (Er^h) \beta_h \leq f \\ & \sum_{k \in N} \alpha_k = 1 \\ & \alpha \in \mathbb{R}_+^n, \beta \in \mathbb{R}_+^\ell. \end{aligned}$$

Given the corresponding dual solution (λ, μ) , then the associated subproblem is given by

$$\max \left\{ \max_{k \in [n]} \left\{ (c - E^\top \lambda)^\top v^k - \mu \right\}, \max_{h \in [\ell]} \left\{ (c - E^\top \lambda)^\top r^h \right\} \right\}.$$

If the value of the subproblem is strictly positive, then there exists $k \in [n] \setminus N$ or $h \in [\ell] \setminus L$ whose associated constraint in the dual is violated. Then we can add the corresponding variable. In fact, the subproblem can be equivalently solved by

$$\max \left\{ (c - E^\top \lambda)^\top x - \mu : x \in \text{conv}(Q) \right\} = \max \left\{ (c - E^\top \lambda)^\top x - \mu : x \in \text{conv}(Q) \right\}.$$

If this optimization problem is unbounded, then there must exist an extreme ray r^h for some $h \in [\ell] \setminus L$ such that $(Er^h)^\top \lambda < c^\top r^h$. If it has a strictly positive finite optimum, then there exists an extreme point v^k for some $k \in [n] \setminus N$ such that $\mu + (Ev^k)^\top \lambda < c^\top v^k$.