

## 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

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

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

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

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  $\lambda$  as follows.

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

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  $r^1, \dots, r^\ell$  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

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

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

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

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  $\lambda$  for the inequality constraint and dual variable  $\mu$  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

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

Eliminating the variable  $\mu$ , 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

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

### 2.3 Dantzig-Wolfe decomposition for pure binary programs

Let us consider a pure binary integer program as follows.

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

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

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

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

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

## 2.4 Problems with block diagonal structure

We consider the following optimization model

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

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 (DW1) has variables  $\alpha_1, \dots, \alpha_n$  for the extreme points of  $\text{conv}(Q)$  and variables  $\beta_1, \dots, \beta_\ell$  for the extreme rays of  $\text{conv}(Q)$ . Therefore,  $n$  and  $\ell$  are potentially very large. In this case, we may apply the column generation technique. Recall that the dual of (DW1) 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

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

Given the corresponding dual solution  $(\lambda, \mu)$ , 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

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

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  $\mu + (Ev^k)^\top \lambda < c^\top v^k$ .