

1. Let x^* solve (P), y^* , u^* solve (D).

$$\text{Then } \theta(u^*) = \min_x c^T x + u^{*T}(b - Ax) \\ \text{s.t. } Ax = b, x \geq 0$$

$$= u^{*T} b + \max_y b^T y$$

$$\text{s.t. } A^T y \leq c - H^T u^*$$

$$\geq u^{*T} b + b^T y^*, \text{ since } y^* \text{ feasible here}$$

$$= c^T x^*$$

We've told $\theta(u) \leq c^T x$ for any feasible x, u .

So $\theta(u^*)$ must achieve the maximum for $\theta(u)$,

$$\text{and } \theta(u^*) = c^T x^* = z.$$

$$2. \quad \theta(\pi) = \min_x c^T x + \pi^T (b - Hx)$$

$$\text{s.t.} \quad Ax = b \\ x \geq 0$$

$$\leq c^T \bar{x} + \pi^T (b - H\bar{x}) \quad \text{since } \bar{x} \text{ feasible in here}$$


This is the valid linear constraint.

It holds at equality at $\bar{u} = \bar{u}$.

3. Then (LD) has the form:

$$\max \quad \theta \quad (\text{LD})$$

$$\text{s.t. } -(h - Hx^k)^T \pi + \theta \leq c^T x^k \quad \text{for extreme points } x^k.$$

Taking the dual, we obtain

$$\min \quad \sum_k \lambda_k c^T x^k$$

$$\text{s.t. } -\sum_k \lambda_k (h - Hx^k) = 0$$

$$\sum_k \lambda_k = 1$$

$$\lambda_k \geq 0 \quad \forall k$$

Rewriting the first constraint, and exploiting $\sum \lambda_k = 1$,

we obtain:

$$\min \quad \sum_k \lambda_k c^T x^k$$

$$\text{s.t. } \sum_k \lambda_k Hx^k = h$$

$$\sum_k \lambda_k = 1$$

$$\lambda_k \geq 0 \quad \forall k.$$

This is exactly the Dantzig-Wolfe master problem.

The D-W subproblem is exactly the LP given by $\mathcal{J}(\pi)$.

$$4. \quad \begin{array}{ll} \min & c^T x \\ \text{s.t.} & Ax = b \\ & Hx = h \\ & x \geq 0 \end{array} \quad (P) \quad \begin{array}{ll} \max & b^T y + h^T w \\ \text{s.t.} & A^T y + H^T w \leq c \end{array} \quad (D)$$

Subproblem, and its dual:

$$\begin{array}{ll} \min & (c - A^T \bar{\pi})^T x \\ \text{s.t.} & Hx = h \\ & x \geq 0 \end{array} \quad (SP) \quad \begin{array}{ll} \max & h^T \bar{v} \\ \text{s.t.} & H^T \bar{v} \leq c - A^T \bar{\pi} \end{array} \quad (SD)$$

Let \bar{v} be optimal in (SD). Then $s^T = h^T \bar{v}$.

Also, $y = \bar{\pi}$, $w = \bar{v}$ is feasible in (D).

So, $c^T x \geq b^T \bar{\pi} + h^T \bar{v} = s^T + b^T \bar{\pi}$ for any feasible x .

This gives a lower bound on the optimal value of (P).

It also gives a possible termination criterion:

stop if duality gap small enough.

5. Look at the dual pair:

$$\begin{array}{ll} \min & b^T y \\ \text{s.t.} & A^T y = 0 \\ & y \geq 0 \end{array}$$

(P)

$$\begin{array}{ll} \max & 0 \\ \text{s.t.} & Ax \leq b \end{array}$$

(D)

① Assume $b^T y \geq 0$ is redundant.

Then (P) has optimal value 0 (since 0 is feasible)

So (D) has optimal value 0.

So $Ax \leq b$ is consistent.

② Assume $b^T y \geq 0$ is not redundant.

So $\exists \bar{y}$ with $A^T \bar{y} = 0$, $\bar{y} \geq 0$, but $b^T \bar{y} < 0$.

Then (P) is unbounded below (consider $\alpha \bar{y}$, let $\alpha \rightarrow +\infty$)

So (D) is infeasible

So $Ax \leq b$ is inconsistent.