

APPLIED INTEGER PROGRAMMING, MODELING AND SOLUTION

CHAPTERS 14.4-14.15

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CHAPTER 14.5: LAGRANGIAN DUAL

LAGRANGIAN RELAXATION

$$\min \quad \mathbf{c}\mathbf{x}$$

$$s.t. \quad \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 \leftarrow \text{Complicated constraints.}$$

$$\mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \leftarrow \text{Easy constraints.}$$

$$\mathbf{x} \in \mathbf{Z}_+^n$$

(IP)

LAGRANGIAN RELAXATION

$$\begin{array}{ll}\min & \mathbf{c}\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 \\ & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n\end{array}$$

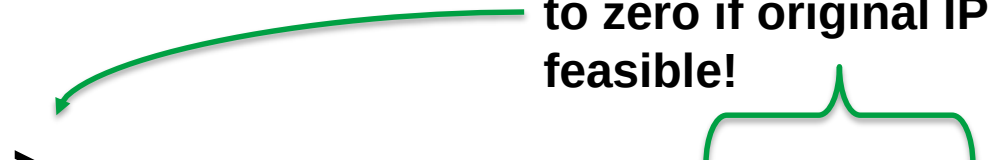
(IP)

$$\begin{array}{ll}\min & \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x}) \\ \text{s.t.} & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n\end{array}$$

(LRIP)

LAGRANGIAN RELAXATION

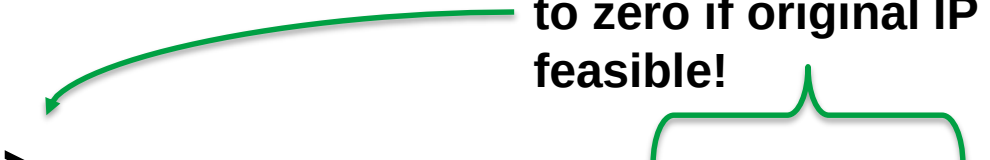
NOTE: Less than or equal to zero if original IP feasible!


$$\begin{array}{ll} \min & \mathbf{c}\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 \\ & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n \end{array} \quad \geq \quad \begin{array}{ll} \min & \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x}) \\ \text{s.t.} & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n \end{array}$$

(IP) (LRIP)

LAGRANGIAN RELAXATION

NOTE: Less than or equal to zero if original IP feasible!


$$\begin{array}{ll} \min & \mathbf{c}\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 \\ & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n \end{array} \geq \begin{array}{ll} \min & \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x}) \\ \text{s.t.} & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n \end{array}$$

So the Lagrangian relaxation optimal value can be used as a **lower bound** to the original IP. But we want a tight bound...

LAGRANGIAN DUAL PROBLEM

...and to find the largest lower bound we solve the Lagrangian dual problem.


$$\begin{aligned} \max_{\mathbf{u} \geq 0} \min_{\mathbf{x} \in \mathbf{Z}_+^n} \quad & \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x}) \\ \text{s.t.} \quad & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \end{aligned}$$

LAGRANGIAN DUAL PROBLEM

...and to find the largest lower bound we solve the Lagrangian dual problem.


$$\max_{\mathbf{u} \geq 0} \min_{\mathbf{x} \in \mathbf{Z}_+^n} \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x})$$

$$s.t. \quad \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2$$



Dual variable aka
Lagrange multiplier

LAGRANGIAN DUAL PROBLEM

...and to find the largest lower bound we solve the Lagrangian dual problem.


$$\max_{\mathbf{u} \geq 0} \min_{\mathbf{x} \in \mathbf{Z}_+^n} \mathbf{c}\mathbf{x} + \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_1\mathbf{x})$$

$$s.t. \quad \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2$$



Dual variable aka
Lagrange multiplier

$$\left(\begin{array}{l} \max_{u \geq 0} \quad \mathbf{u}\mathbf{b}_1 + \min_{\mathbf{x} \in \mathbf{Z}_+^n} (\mathbf{c} - \mathbf{u}\mathbf{A}_1)\mathbf{x} \\ s.t. \quad \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \end{array} \right)$$

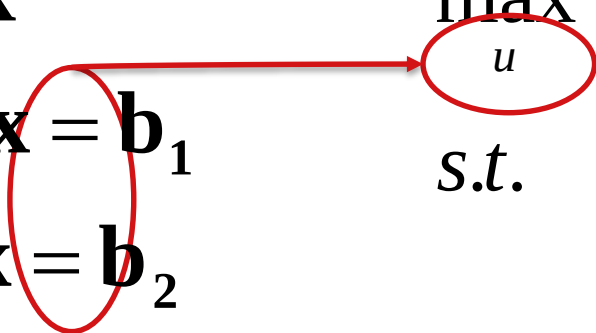
LAGRANGIAN DUAL PROBLEM

Proposition 14.8: Given a specific $\mathbf{u} \in R_+^{m_1}$, if the following two conditions are satisfied, then the optimal vector $\mathbf{x}^*(\mathbf{u})$ of LRIP($\mathbf{u}$) is optimal for IP:

1. $\mathbf{A}_1 \mathbf{x}^*(u) \geq \mathbf{b}_1$
2. $(\mathbf{A}_1 \mathbf{x}^*(u) \geq \mathbf{b}_1)_i = \mathbf{b}_{1i}$ whenever $\mathbf{u}_i > 0$

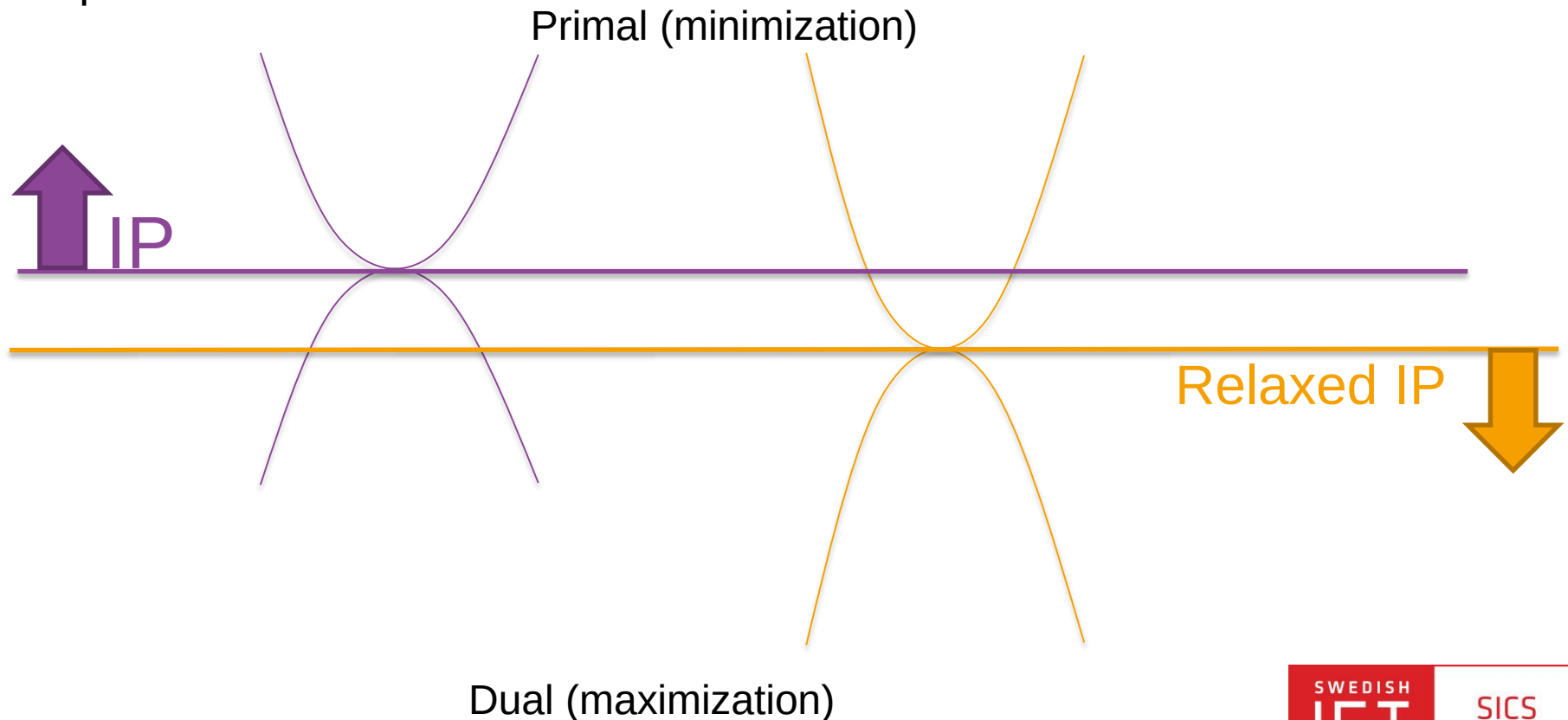
EQUALITY INSTEAD OF INEQUALITY

No sign restriction on u !

$$\begin{array}{ll} \min & \mathbf{c}\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} = \mathbf{b}_1 \\ & \mathbf{A}_2\mathbf{x} = \mathbf{b}_2 \\ & \mathbf{x} \in \mathbf{Z}_+^n \end{array}$$
$$\begin{array}{ll} \max & \mathbf{u}\mathbf{b}_1 + \min_{\mathbf{x} \in \mathbf{Z}_+^n} (\mathbf{c} - \mathbf{u}\mathbf{A}_1)\mathbf{x} \\ \text{s.t.} & \mathbf{A}_2\mathbf{x} \geq \mathbf{b}_2 \end{array}$$


LP DUAL OF IP

Proposition 14.5: The integer program (IP) and the dual of its linear programming relaxation (Relaxed IP) form a weak dual pair.



CHAPTER 14.6: PRIMAL –DUAL SOLUTION VIA BENDERS' PARTITIONING

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

$$\min \mathbf{c}_1 \mathbf{x} + \mathbf{c}_2 \mathbf{y}$$

$$s.t. \quad \mathbf{A}_1 \mathbf{x} + \mathbf{A}_2 \mathbf{y} \geq \mathbf{b}_1$$

$$\mathbf{D} \mathbf{y} \geq \mathbf{b}_2$$

$$\mathbf{x} \geq 0, \mathbf{y} \geq 0 \text{ and integer}$$

MIPs are hard. Let's try to partition it.

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

$$\begin{array}{ll}\min & \mathbf{c}_1\mathbf{x} + \mathbf{c}_2\mathbf{y} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} + \mathbf{A}_2\mathbf{y} \geq \mathbf{b}_1 \\ & \mathbf{D}\mathbf{y} \geq \mathbf{b}_2 \\ & \mathbf{x} \geq 0, \mathbf{y} \geq 0 \text{ and integer}\end{array}$$

MASTER

$$\begin{array}{ll}\min & \mathbf{c}_2\mathbf{y} + q(\mathbf{y}) \\ \text{s.t.} & \mathbf{D}\mathbf{y} \geq \mathbf{b}_2 \\ & \mathbf{y} \geq 0 \text{ and integer}\end{array}$$

SUB-PROBLEM

$$\begin{array}{ll}q(\mathbf{y}) = \min & \mathbf{c}_1\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 - \mathbf{A}_2\mathbf{y} \\ & \mathbf{x} \geq 0\end{array}$$

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

$$\begin{array}{ll}\min & \mathbf{c}_1\mathbf{x} + \mathbf{c}_2\mathbf{y} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} + \mathbf{A}_2\mathbf{y} \geq \mathbf{b}_1 \\ & \mathbf{D}\mathbf{y} \geq \mathbf{b}_2 \\ & \mathbf{x} \geq 0, \mathbf{y} \geq 0 \text{ and integer}\end{array}$$

MASTER

$$\begin{array}{ll}\min & \mathbf{c}_2\mathbf{y} + q(\mathbf{y}) \\ \text{s.t.} & \mathbf{D}\mathbf{y} \geq \mathbf{b}_2 \\ & \mathbf{y} \geq 0 \text{ and integer}\end{array}$$

SUB-PROBLEM

$$\begin{array}{ll}q(\mathbf{y}) = \min & \mathbf{c}_1\mathbf{x} \\ \text{s.t.} & \mathbf{A}_1\mathbf{x} \geq \mathbf{b}_1 - \mathbf{A}_2\mathbf{y} \\ & \mathbf{x} \geq 0\end{array}$$

It'd be great if the $\mathbf{y}$'s weren't in the constraints of $q(\mathbf{y})$.

DUAL!

DUAL OF SUB-PROBLEM

SUB-PROBLEM

$$q(y) = \min \mathbf{c}_1 \mathbf{x}$$

$$s.t. \quad \mathbf{A}_1 \mathbf{x} \geq \mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}$$

$$\mathbf{x} \geq 0$$

$$\max \quad \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}')$$

$$s.t. \quad \mathbf{u} \mathbf{A}_1 \leq \mathbf{c}_1$$

$$\mathbf{u} \geq 0$$

DUAL OF SUB-PROBLEM

SUB-PROBLEM

$$\begin{aligned} q(y) = \min \quad & \mathbf{c}_1 \mathbf{x} \\ \text{s.t.} \quad & \mathbf{A}_1 \mathbf{x} \geq \mathbf{b}_1 - \mathbf{A}_2 \mathbf{y} \\ & \mathbf{x} \geq 0 \end{aligned}$$

Avoid unboundedness.

$$\begin{aligned} \max \quad & \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}') \\ \text{s.t.} \quad & \mathbf{u} \mathbf{A}_1 \leq \mathbf{c}_1 \\ & \mathbf{u} \mathbf{E} \leq \mathbf{M} \\ & \mathbf{u} \geq 0 \end{aligned}$$

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

1. For every $\mathbf{y}'$ we will pick out a vertex $\mathbf{u}_i$ of the dual feasible solution space. From weak duality:

$$q(\mathbf{y}) \geq \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

Dual of sub-problem

$$\max \quad \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y}')$$

$$\text{s.t.} \quad \mathbf{u}\mathbf{A}_1 \leq \mathbf{c}_1$$

$$\mathbf{u}\mathbf{E} \leq \mathbf{M}$$

$$\mathbf{u} \geq 0$$

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

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$$q(\mathbf{y}) \geq \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

2. So, we have that the master objective

$$z = \mathbf{c}_2\mathbf{y} + q(\mathbf{y}) \geq \mathbf{c}_2\mathbf{y} + \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

Dual of sub-problem

$$\max \quad \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y}')$$

$$\text{s.t.} \quad \mathbf{u}\mathbf{A}_1 \leq \mathbf{c}_1$$

$$\mathbf{u}\mathbf{E} \leq \mathbf{M}$$

$$\mathbf{u} \geq 0$$

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

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$$z = \mathbf{c}_2\mathbf{y} + q(\mathbf{y}) \geq \mathbf{c}_2\mathbf{y} + \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

3. ADD CONSTRAINT TO MASTER!

Dual of sub-problem

$$\max \quad \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y}')$$

$$\text{s.t.} \quad \mathbf{u}\mathbf{A}_1 \leq \mathbf{c}_1$$

$$\mathbf{u}\mathbf{E} \leq \mathbf{M}$$

$$\mathbf{u} \geq 0$$

Master

$$\min \quad \mathbf{c}_2\mathbf{y} + q(\mathbf{y})$$

$$\text{s.t.} \quad \mathbf{D}\mathbf{y} \geq \mathbf{b}_2$$

$$\mathbf{y} \geq 0 \text{ and integer}$$

PRIMAL-DUAL SOLUTION VIA BENDERS' PARTITIONING

For every $\mathbf{y}'$ we will pick out a vertex $\mathbf{u}_i$ of the dual feasible solution space. From weak duality:

$$q(\mathbf{y}) \geq \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

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$$z = \mathbf{c}_2\mathbf{y} + q(\mathbf{y}) \geq \mathbf{c}_2\mathbf{y} + \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y})$$

3. ADD CONSTRAINT TO MASTER!

$$\min z$$

$$\text{s.t. } z \geq \mathbf{c}_2\mathbf{y} + \mathbf{u}_i(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y}) \quad \forall i = 1 \dots T$$

$$\mathbf{D}\mathbf{y} \geq \mathbf{b}_2$$

$$\mathbf{y} \geq 0 \text{ and integer}$$

Dual of sub-problem

$$\max \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2\mathbf{y}')$$

$$\text{s.t. } \mathbf{u}\mathbf{A}_1 \leq \mathbf{c}_1$$

$$\mathbf{u}\mathbf{E} \leq \mathbf{M}$$

$$\mathbf{u} \geq 0$$

Master

$$\min \mathbf{c}_2\mathbf{y} + q(\mathbf{y})$$

$$\text{s.t. } \mathbf{D}\mathbf{y} \geq \mathbf{b}_2$$

$$\mathbf{y} \geq 0 \text{ and integer}$$

THE MASTER

$$\min z$$
$$s.t. \quad z \geq \mathbf{c}_2 \mathbf{y} + \mathbf{u}_i (\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}) \quad \forall i = 1 \dots T$$
$$\mathbf{D} \mathbf{y} \geq \mathbf{b}_2$$
$$\mathbf{y} \geq 0 \text{ and integer}$$

The number of vertices (T) can be very large, so instead of solving the Master with all T , start with one and then **generate** more constraints as needed.

BENDERS' ALGORITHM FOR MIP

Step 0: Initialization

Set $t = 1, B_u = +\infty$ and select some ε (convergence criterion). Select some $\mathbf{u}^1$ that is feasible for the dual sub-problem.

Step1: Solve master in iteration t

Solve the relaxed pure integer program:

$$\begin{aligned} \min \quad & z \\ \text{s.t.} \quad & z \geq \mathbf{c}_2 \mathbf{y} + \mathbf{u}_i (\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}) \quad \forall i = 1 \dots t \\ & \mathbf{D} \mathbf{y} \geq \mathbf{b}_2 \\ & \mathbf{y} \geq 0 \text{ and integer} \end{aligned}$$

Let z^t and $\mathbf{y}^t$ be the solution. If z is unbounded from below, take $\mathbf{y}^t$ to be some value that gives z^t some arbitrarily large negative value.

BENDERS' ALGORITHM FOR MIP

Step 2: Solve sub-problem in iteration t

Generate the most violated constraint of IP by solving the linear program:

$$\begin{array}{ll}\max & \mathbf{u}(\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y}') \\ \text{s.t.} & \mathbf{u} \mathbf{A}_1 \leq \mathbf{c}_1 \\ & \mathbf{u} \mathbf{E} \leq \mathbf{M} \\ & \mathbf{u} \geq 0\end{array}$$

Let the solution of this LP be optimal value u_0^{t+1} at $\mathbf{u}^{t+1}$.

Step 3: Optimality check

Check convergence criterion. Set $B_u = \min\{B_u, u_0^{t+1} + \mathbf{c}_1 \mathbf{y}^t\}$. If $z^t > B_u - \varepsilon$, stop; the optimal solution has been reached. Otherwise, add the constraint $z \geq \mathbf{c}_2 \mathbf{y} + \mathbf{u}^{t+1}(\mathbf{b}_1 - \mathbf{A}_2 \mathbf{y})$ to the master problem. Set $t = t + 1$. Return to step 1.