DM545/DM871

Linear and Integer Programming

Lecture 13
Branch and Bound

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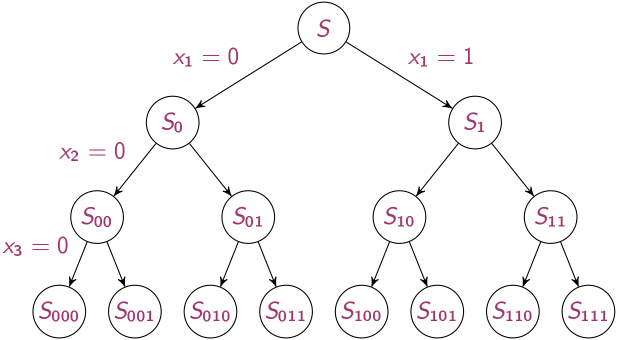
Outline

Outline

Branch and Bound

- Consider the problem $z = \max\{c^T x : x \in S\}$
- Divide and conquer: let $S = S_1 \cup \ldots \cup S_k$ be a decomposition of S into smaller sets, and let $z^k = \max\{c^T x : x \in S_k\}$ for $k = 1, \ldots, K$. Then $z = \max_k z^k$

For instance if $S \subseteq \{0,1\}^3$ the enumeration tree is:

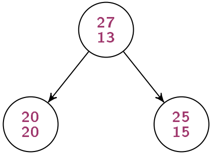


Bounding

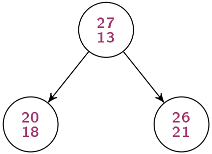
Let's consider a maximization problem

- Let $\overline{z}^k$ be an upper bound on z^k (dual bound)
- Let $\underline{z}^k$ be a lower bound on z^k (primal bound)
- $(\underline{z}^k \leq z^k \leq \overline{z}^k)$
- $\underline{z} = \max_k \underline{z}^k$ is a lower bound on z
- $\overline{z} = \max_k \overline{z}^k$ is an upper bound on z

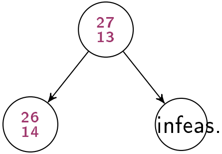
Pruning



$\overline{z} = 25$
$\underline{z} = 20$
pruned by optimality

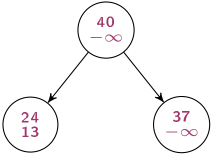


$\overline{z} = 26$
$\underline{z} = 21$
pruned by bounding



$\overline{z} = 26$
$\underline{z} = 14$
pruned by infeasibility

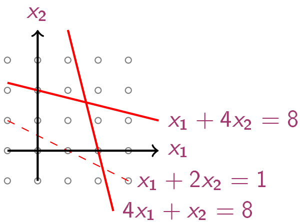
Pruning



$\overline{z} = 37$
$\underline{z} = 13$
nothing to prune

Example

$$
\begin{aligned}
\max\ \ x_1 &+ 2x_2 \\
x_1 &+ 4x_2 \le 8 \\
4x_1 &+ \ x_2 \le 8 \\
& x_1, x_2 \ge 0, \text{integer}
\end{aligned}
$$



- Solve LP

	x1	x2	x3	x4	-z	b
	1	4	1	0	0	8
	4	1	0	1	0	8
	1	2	0	0	1	0

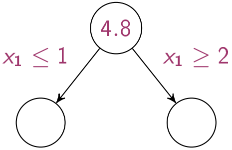
	x1	x2	x3	x4	-z	b
I'=I-II'	0	15/4	1	-1/4	0	6
II'=1/4II	1	1/4	0	1/4	0	2
III'=III-II'	0	7/4	0	-1/4	0	-2

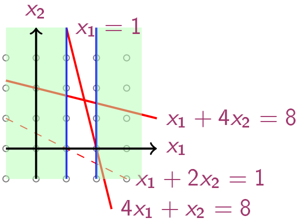
- continuing

	x1	x2	x3	x4	-z	b
I'=4/15I	0	1	4/15	-1/15	0	24/15
II'=II-1/4I'	1	0	-1/15	4/15	0	24/15
III'=III-7/4I'	0	0	-7/15	-3/5	1	-2-14/5

$x_2 = 1 + 3/5 = 1.6$
$x_1 = 8/5$
The optimal solution will not be more than $2 + 14/5 = 4.8$

- Both variables are fractional, we pick one of the two:





- Let's consider first the left branch:

	x1	x2	x3	x4	x5	-z	b
	1	0	0	0	1	0	1
	0	1	4/15	-1/15	0	0	24/15
	1	0	-1/15	4/15	0	0	24/15
	0	0	-7/15	-3/5	0	1	-24/5

	x1	x2	x3	x4	x5	b	-z
I'=I-III	0	0	1/15	-4/15	1	0	-9/15
	0	1	4/15	-1/15	0	0	24/15
	1	0	-1/15	4/15	0	0	24/15
	0	0	-7/15	-3/5	0	1	-24/5

	x1	x2	x3	x4	x5	b	-z
I'=-15/4I	0	0	-1/4	1	-15/4	0	9/4
II'=II-1/4I	0	1	15/60	0	-1/4	0	7/4
III'=III+I	1	0	0	0	1	0	1
	0	0	-37/60	0	-9/4	1	-90/20

always a b term negative after branching:

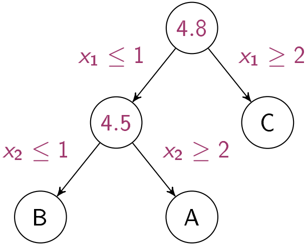
$b_1 = \lfloor \bar{b}_3 \rfloor$

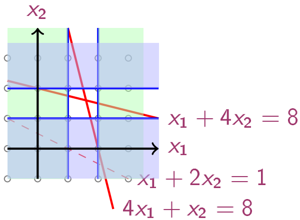
$\bar{b}_1 = \lfloor \bar{b}_3 \rfloor - b_3 < 0$

Dual simplex:

$\min_j \{ |\frac{c_j}{a_{ij}}| \; : \; a_{ij} < 0 \}$

- Let's branch again





We have three open problems. Which one we choose next?
Let's take A.

	x1	x2	x3	x4	x5	x6	b	-z
	0	-1	0	0	0	1	0	-2
	0	0	-1/4	1	-15/4		0	9/4
	0	1	15/60	0	-1/4		0	7/4
	1	0	0	0	1		0	1
	0	0	-37/60	0	-9/4		1	-9/2

	x1	x2	x3	x4	x5	x6	b	-z
III+I	0	0	1/4	0	-1/4	1	0	-1/4
	0	0	-1/4	1	-15/4		0	9/4
	0	1	15/60	0	-1/4		0	7/4
	1	0	0	0	1		0	1
	0	0	-37/60	0	-9/4		1	-9/2

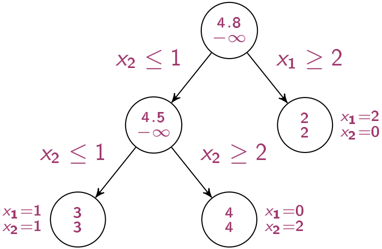
continuing we find:

$x_1 = 0$

$x_2 = 2$

$OPT = 4$

The final tree:



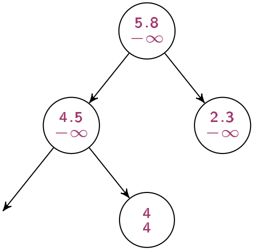
The optimal solution is 4.

Pruning

Pruning:

1. by optimality: $z^k = \max\{c^T x : x \in S^k\}$

2. by bound $\overline{z}^k \leq \underline{z}$
   Example:



3. by infeasibility $S^k = \emptyset$

B&B Components

Bounding:

1. LP relaxation
2. Lagrangian relaxation
3. Combinatorial relaxation
4. Duality

Branching:

$$S_1 = S \cap \{x : x_j \leq \lfloor \bar{x}_j \rfloor\}$$
$$S_2 = S \cap \{x : x_j \geq \lceil \bar{x}_j \rceil\}$$

thus the current optimum is not feasible either in S_1 or in S_2.
Which variable to choose?
Eg: Most fractional variable $\arg\max_{j \in C} \min\{f_j, 1 - f_j\}$

Choosing Node for Examination from the list of active (or open):

- Depth First Search (a good primal sol. is good for pruning + easier to reoptimize by just adding a new constraint)
- Best Bound First: (eg. largest upper: $\overline{z}^s = \max_k \overline{z}^k$ or largest lower - to die fast)
- Mixed strategies

Reoptimizing: dual simplex

Updating the Incumbent: when new best feasible solution is found:

$$\underline{z} = \max\{\underline{z}, 4\}$$

Store the active nodes: bounds + optimal basis (remember the revised simplex!)

Enhancements

- Preprocessor: constraint/problem/structure specific
  tightening bounds
  redundant constraints
  variable fixing: eg: $\max\{\mathrm{c}^T\mathrm{x} : A\mathrm{x} \leq \mathrm{b}, \mathrm{l} \leq \mathrm{x} \leq \mathrm{u}\}$
  fix $x_j = l_j$ if $c_j < 0$ and $a_{ij} > 0$ for all i
  fix $x_j = u_j$ if $c_j > 0$ and $a_{ij} < 0$ for all i

- Priorities: establish the next variable to branch

- Special ordered sets SOS (or generalized upper bound GUB)

$$\sum_{j=1}^{k} x_j = 1 \qquad x_j \in \{0, 1\}$$

instead of: $S_0 = S \cap \{\mathrm{x} : x_j = 0\}$ and $S_1 = S \cap \{\mathrm{x} : x_j = 1\}$
$\{\mathrm{x} : x_j = 0\}$ leaves $k - 1$ possibilities
$\{\mathrm{x} : x_j = 1\}$ leaves only 1 possibility
hence tree unbalanced
here: $S_1 = S \cap \{\mathrm{x} : x_{j_i} = 0, i = 1..r\}$ and $S_2 = S \cap \{\mathrm{x} : x_{j_i} = 0, i = r+1, .., k\}$,
$r = \min\{t : \sum_{i=1}^{t} x^*_{j_i} \geq \frac{1}{2}\}$

- Cutoff value: a user-defined primal bound to pass to the system.
- Simplex strategies: simplex is good for reoptimizing but for large models interior points methods may work best.
- Strong branching: extra work to decide more accurately on which variable to branch:
  1. choose a set C of fractional variables
  2. reoptimize for each of them (in case for limited iterations)
  3. $\bar{z}_j^{\downarrow}, \bar{z}_j^{\uparrow}$ (dual bound of down and up branch)

  $$j^* = \arg\min_{j \in C} \max\{\bar{z}_j^{\downarrow}, \bar{z}_j^{\uparrow}\}$$

  ie, choose variable with largest decrease of dual bound, eg UB for max

There are four common reasons because integer programs can require a significant amount of solution time:

1. There is lack of node throughput due to troublesome linear programming node solves.
2. There is lack of progress in the best integer solution, i.e., the primal bound.
3. There is lack of progress in the best dual bound.
4. There is insufficient node throughput due to numerical instability in the problem data or excessive memory usage.

For 2) or 3) the gap best feasible-dual bound is large:

$$\text{gap} = \frac{|\text{Primal bound} - \text{Dual bound}|}{\text{Primal bound} + \epsilon} \cdot 100$$

- heuristics for finding feasible solutions (generally NP-complete problem)
- find better lower bounds if they are weak: addition of cuts, stronger formulation, branch and cut
- Branch and cut: a B&B algorithm with cut generation at all nodes of the tree. (instead of reoptimizing, do as much work as possible to tighten)

  Cut pool: stores all cuts centrally
  Store for active node: bounds, basis, pointers to constraints in the cut pool that apply at the node

Relative Optimality Gap

In CPLEX:

$$\text{gap} = \frac{|\text{best dual bound} - \text{best integer}|}{|\text{best integer} + 10^{-11}|}$$

In SCIP and MIPLIB standard:

$$\text{gap} = \frac{pb - db}{\inf\{|z|, z \in [db, pb]\}} \cdot 100 \qquad \text{for a minimization problem}$$

(if $pb \geq 0$ and $db \geq 0$ then $\frac{pb-db}{db}$)
if $db = pb = 0$ then $\text{gap} = 0$
if no feasible sol found or $db \leq 0 \leq pb$ then the gap is not computed.

Last standard avoids problem of non decreasing gap if we go through zero

```
   3186  2520     -666.6217  4096       956.6330     -667.2010  1313338  169.74%
   3226  2560     -666.6205  4097       956.6330     -667.2010  1323797  169.74%
   3266  2600     -666.6201  4095       956.6330     -667.2010  1335602  169.74%
Elapsed real time = 2801.61 sec. (tree size = 77.54 MB, solutions = 2)
*  3324+ 2656                          -125.5775     -667.2010  1363079  431.31%
   3334  2668     -666.5811  4052      -125.5775     -667.2010  1370748  431.31%
   3380  2714     -666.5799  4017      -125.5775     -667.2010  1388391  431.31%
   3422  2756     -666.5791  4011      -125.5775     -667.2010  1403440  431.31%
```

Advanced Techniques

We did not treat:

- LP: Dantzig Wolfe decomposition
- LP: Column generation
- LP: Delayed column generation
- IP: Branch and Price
- LP: Benders decompositions
- LP: Lagrangian relaxation

Summary

1. Branch and Bound