Exact Methods

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Exact methods for CO

- Combinatorial optimization problem solved to optimality
 (Q) max {w^Tx: x ∈ S⊆ {0,1}ⁿ} = v(Q) 0-1 linear program
- Consider $T \supseteq S$ and let (R) max $\{w^T x: x \in T\} \ge v(Q)$
- Suppose the <u>optimal solution</u> x_R to (R) can be found efficiently e.g. T = P polyhedron and $S = P \cap \{0,1\}^n$
- Let $x^* \in S$ (incumbent solution) and $LB = w^T x^* \leq v(Q)$
- LB = -∞ if no incumbent solution is known
 Solving (R) = Solving (Q)?

Only if some conditions are verified.

Solving (R) instead of (Q)

- 1. *Infeasibility.* T is empty \rightarrow S is empty
- 2. Optimality. $x_{R} \in S \rightarrow x_{R}$ optimal for (Q). $v(Q) \ge w^{T}x_{R} = v(R) \ge v(Q)$
- 3. Value Dominance. $w^T x_R \leq LB = w^T x^* \rightarrow x^*$ optimal for (Q). $w^T x^* \geq v(R) \geq v(Q) \geq w^T x^*$
- If none of this conditions is satisfied we do *divide-et-impera*
- Partition S and decompose (Q) into a number of smaller subproblems $(Q(u)) \max \{w^T x: x \in S(u)\}, \bigcup_u S(u) = S$

Solving Q(*u*)

• (Q(u)) max $\{w^T x: x \in S(u)\}$ still difficult but "smaller" than Q



$$(Q(u)) \max \{w^T x: x \in S(u)\}$$

 $v(Q) = \max_u v(Q(u))$

Partitions can be built recursively by fixing variables to 0 or 1.



• Each leaf of the <u>complete</u> tree corresponds to a specific 0,1 vector!

Divite-et-impera

• We can try to solve each (Q(u)) by solving a relaxation (R(u)) $(Q(u)) \max \{w^T x: x \in S(u)\}$ $(R(u)) \max \{w^T x: x \in T(u) \supseteq S(u)\}$



x(u) optimal solution to R(u) $z(u) = w^T x(u)$ optimal value of R(u) x^* overall incumbent, $z_L = w^T x^*$

1. *Infeasibility.* T(u) is empty $\rightarrow S(u)$ is empty

3. Value Dominance. $z(u) \le LB = w^T x^*$ no use to solve Q(u)

 $w^T x^* \ge z(u) \ge v(Q(u))$ no solution in S(u) better than incumbent x^*

2. Optimality. $x_{R(u)} \in S(u) \rightarrow x_{R(u)}$ optimal for (Q(u)).

if $w^T x_{R(u)} > LB$ we can set the incumbent $x^* = x_{R(u)}$

Branch-and-bound

Branch-and-Bound algorithm for 0,1 programming

- Step 1. (Initialization) Let $V_n = \{v_r\}, z_L = -\infty$.
- Step 2. (Termination.) If $V_n = \emptyset$, terminate (x^* is optimal).
- Step 3. (Node selection and solution)
 Select u in V_n. Set V_n := V_n \ {u}. Solve the LP relaxation (R(u)).

• Step 4. (Pruning.) (i) If (R(u)) is infeasible Goto 2. (ii) If $x(u) \in S(u)$ and $w^T x(u) > z_L$, let $x^* = x(u)$, $z_L = w^T x(u)$. Goto 2. (iii) If $z(u) = w^T x(u) \le z_L$, Goto 2.

Step 5. (Branching.) Choose a variable x_i . Let $S(u_0) = \{x \in S \cap \{x_i = 0\}\}$. $S(u_1) = \{x \in S \cap \{x_i = 1\}\}$. Add u_0 and u_1 to V_{n_i} Goto 2.

$$\max - 5x_1 - 9x_2 - 7x_3 - 5x_4$$
$$\begin{cases} 4x_1 + 5x_2 + 3x_3 + 2x_4 \ge 7\\ x \in \{0,1\}^4 \end{cases}$$

$$\hat{x}_1 = 1$$

 $\hat{x}_2 = 3/5 = 0,6$
 $\hat{x}_3 = 0$
 $\hat{x}_4 = 0$

UB = -10,4

LB = -14
$$\Rightarrow \begin{cases} \hat{x}_1 = \hat{x}_2 = 1 \\ \hat{x}_3 = \hat{x}_4 = 0 \end{cases}$$

$$\begin{aligned}
x_1 &= 1 \\
\max &-5x_1 - 9x_2 - 7x_3 - 5x_4 \\
\begin{cases}
4x_1 + 5x_2 + 3x_3 + 2x_4 \ge 7 \\
x \in \{0,1\}^4
\end{aligned}$$

$$\begin{aligned}
x_1 &= 1 \\
\hat{x}_2 &= 3/5 = 0,6 \\
\hat{x}_4 &= 0 \\
\hat{x}_3 &= 0
\end{aligned}$$

$$x_2 = 0$$

 $x_2 = 1$

 $\max -5x_1 - 7x_3 - 5x_4$ $4x_1 + 3x_3 + 2x_4 \ge 7$

 $-9 + \max - 5x_1 - 7x_3 - 5x_4$ $4x_1 + 3x_3 + 2x_4 \ge 2$

 $x_2 = 0$ $x_2 = 1$ $-9 + \max - 5x_1 - 7x_3 - 5x_4$ $\max -5x_1 - 7x_3 - 5x_4$ $4x_1 + 3x_3 + 2x_4 \ge 2$ $4x_1 + 3x_3 + 2x_4 \ge 7$ $\begin{cases} \hat{x}_2 = 1 \\ \hat{x}_1 = 1/2 = 0,5 \quad \text{UB} = -11,5 \\ \hat{x}_3 = \hat{x}_4 = 0 \\ x_1 = 1 \end{cases}$ $x_1 = 0$ $-9 + \max - 7x_3 - 5x_4$ $-14 + \max -7x_3 - 5x_4$ $3x_{3} + 2x_{4} \ge 2$ $3x_3 + 2x_4 \ge -2$ UB ≤–14 =LB Similar

$$x_{2} = 0$$

$$\max -5x_{1} - 7x_{3} - 5x_{4}$$

$$4x_{1} + 3x_{3} + 2x_{4} \ge 7$$

$$\begin{cases} \hat{x}_{1} = \hat{x}_{3} = 1 \\ \hat{x}_{2} = \hat{x}_{4} = 0 \end{cases}$$

$$LB = -12$$

$$x \text{ binary}$$

$$x_{1} = 0$$

$$x_{1} = 1$$

$$x_{2} = 1$$

$$\hat{x}_{2} = 1$$

$$\hat{x}_{1} = 1/2 = 0,5 \quad UB = -11,5$$

$$\hat{x}_{3} = \hat{x}_{4} = 0$$

$$x_{1} = 1$$

$$x_{1} = 1$$

$$y_{1} = 1$$

$$x_{2} = 1$$

$$x_{1} = 1$$

$$x_{2} = 1$$

$$x_{1} = 1$$

$$x_{1} = 1$$

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$$x_{1} = 1$$

$$x_{2} = 1$$

$$x_{2} = 1$$

$$x_{3} = 2$$

Open choices in B&B

- Problem Selection: how to choose next problem?
 Depth First Search, Best Bound, ...
 - Branching: how to choose next branching variable?
 Most "fractional" variable
- Relaxation: how to generate good formulations?

Find new inequalities (cutting planes) to add to the initial formulation to make it stronger.

Valid Inequalities

Valid Inequality

Let $P \subseteq \mathbb{R}^n$, $a \in \mathbb{R}^n$ e $b \in \mathbb{R}$. The linear inequality $a^T x \le b$ is *valid* for *P* if it is satisfied by every point of *P*





Valid inequalities for polyhedra

- Let $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ polyhedron $(A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m)$
- Any conic combination of the constraints defining P is valid for P



Gomory Cuts

Let $P \subseteq \{x \in \mathbb{R}^n : Ax \le b, A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m\}$ and $u \in \mathbb{R}_+^m$, then $\lfloor u^T A \rfloor x \le \lfloor u^T b \rfloor$ is a *Gomory-cut* and is valid for $P \cap \{0,1\}^n$

A Gomory cut



Example: matching

- G = (V, E) undirected graph.
- Matching $M \subseteq E$: subset of edges *meeting* each vertex <u>at most</u> once
- $x \in \{0,1\}^E$ incidence vector of matching in $G \to x(\delta(v)) \le 1$ for all $v \in V$



• Maximum Cardinality Matching = $\max\{1^T x: A_G x \le 1, x \in \{0,1\}^E\}$ **II** *G* bipartite $\rightarrow A_G$ is TU $\implies \max\{1^T x: A_G x \le 1, x \ge 0\}$ LP!

Example: non-bipartite matching

• What if G = (V, E) is non bipartite?



OBS: maximum cardinality matching value: 2 but ...

 $\max\{1^T x: A_G x \le 1, x \ge 0\} = 2.5 \ (x_i = \frac{1}{2}, i = 1, ..., 5)$

S incidence vectors of matching of G

• A_G non totally unimodular $P = \{x \in \mathbb{R}^E : A_G x \le 1, x \ge 0\} \neq \text{conv}(S)$

Odd-cycle inequalities

 $\mathsf{P} = \{x \in \mathbb{R}^E \colon A_G x \le 1, \ x \ge 0\} \neq \operatorname{conv}(S)$





A	_G X	\leq	1
	G		

C odd-cycle (odd number of vertices) *M* matching

 $|M \cap C| \le (|C|-1) / 2$

$$\sum_{e \in C} x_e \leq |C| / 2$$

must be satisfied by every incidence vector of a matching of G

δ(5)



valid for conv(S) (but not for P! Why?)

 $x_4 + x_5 \le 1$

Odd-cycle inequalities and Gomory cuts

$\mathsf{P} = \{x \in R^E \colon A_G x \le 1, \ x \ge 0\} \neq \operatorname{conv}(S)$

Consider the edges of a cycle $C = \{e_1, ..., e_k\}$ For every pair of consecutive edges e_r , e_{r+1}

 $x_r + x_{r+1} \le 1$ is valid for *P* (why? Hint obtain it as a conic combination of inequalities defining *P*.)

Summing up all the |C| inequalities associated to C

$$\sum_{e \in C} 2x_e \leq |C|$$

Dividing by 2 and rounding down we finally obtain



 $x_{1} + x_{5} \le 1$ $x_{1} + x_{2} \le 1$ $x_{2} + x_{3} \le 1$ $x_{3} + x_{4} \le 1$ $x_{4} + x_{5} \le 1$

$$\sum_{e \in C} 2x_e \le 5$$
$$\sum x_e \le 2$$

$$\sum_{e \in C} x_e \leq |C| / 2_{-}$$

Branch & Cut

- Branch-and-Bound + cutting planes = Branch-and-Cut
- A *cut* is an inequality which is valid for $P \cap \{0,1\}^n$ but not for P
- While solving current problem we try to strengthen the current relaxation by adding valid inequalities
- This is done in a dynamic-simplex-method fashion: the current solution is input to an oracle which tries to find a violated cut.
- Two types of cuts
 - Template cuts: cuts with a given pattern (e.g. odd-cycles)
 - General cuts.