

Rank 1 Facets of the Set Packing Polyhedron as Limited Memory Cuts

Teobaldo Bulhões

Universidade Federal da Paraíba, Brazil

tbulhoes@ci.ufpb.br



Autumn school on Advanced BCP Tools: VRPSolver and Coluna

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Set Packing Problem

Formulation

$$\begin{aligned} \min \quad & \sum_{j=1}^m c_j x_j \\ \text{s.t.} \quad & Ax \leq \mathbb{1}, \\ & x \in \{0, 1\}^m. \end{aligned}$$

$$\begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} \leq \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Set Packing Problem

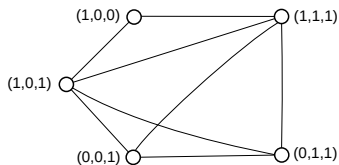


Figure 1: Intersection graph

$$\begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{bmatrix} \leq \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Set Packing Polytope

The set packing polytope associated with a matrix A is:

$$SPP_{\leq}(A) = \text{Conv} \{Ax \leq \mathbb{1}, x \in \{0, 1\}^m\}$$

Also denoted as $SPP_{\leq}(G)$, where G is the intersection graph derived from A

Set Packing Polytope

$SPP_{\leq}(G)$ is a classical polytope, studied since the seventies.

- Facet-inducing inequalities associated with specific graphs:
 - ▶ cliques (Padberg, 1973)
 - ▶ odd holes (Padberg, 1973)
 - ▶ odd anti-holes (Nemhauser and Trotter, 1974)
 - ▶ webs and anti-webs (Trotter, 1975)
 - ▶ $K_{1,3}$ -free graphs (Giles and Trotter, 1981)
 - ▶ wheels (Cheng and Cunningham, 1997)
 - ▶ antiweb-wheels (Cheng and Vries, 2002)
 - ▶ grilles (Cánovas et al., 2000).

Set Packing Polytope

- Complete characterizations of $SPP_{\leq}(G)$ are known for:
 - ▶ perfect graphs (Chvátal, 1975).
 - ▶ series-parallel graphs (Mahjoub, 1988; Barahona and Mahjoub, 1994a).
 - ▶ graphs that do not have a 4-wheel as a minor (Barahona and Mahjoub, 1994b).

Set Packing Polytope

- Facet-generating procedures (lifting, composition of graphs, and others) for $SPP_{\leq}(G)$ were described by:
 - ▶ Nemhauser and Trotter (1974).
 - ▶ Chvátal (1975).
 - ▶ Wolsey (1976).
 - ▶ Padberg (1977).
 - ▶ Balas and Zemel (1977).
 - ▶ Barahona and Mahjoub (1994a).
 - ▶ Cánovas et al. (2000).
 - ▶ Rossi and Smriglio (2001).
 - ▶ Rebennack et al. (2011).
 - ▶ S. Xavier and Campêlo (2011).
 - ▶ Corrêa et al. (2017).

Set Packing Cuts in Column Generation

- Many applications can be modeled as set packing/partitioning problems with a very large number of columns.
- Explicit representation of the coefficient matrix is unpractical.
- Any cutting plane should have a well-defined coefficient for every possible column.

Complete Set Packing Polytope

Complete set packing polytope:

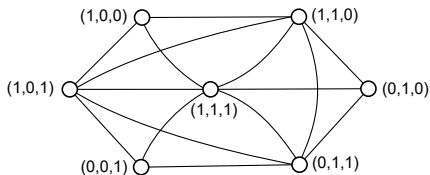
$$CSPP_{\leq}(n) = \text{Conv} \left\{ \sum_{j=1}^{2^n-1} b^j \lambda_j \leq \mathbb{1}, \lambda \in \{0, 1\}^{2^n-1} \right\}$$

where b^j is the column associated with the binary representation of j .

Complete Set Packing Polytope

For example, if $n = 3$ we have:

$$\begin{bmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \lambda_3 \\ \lambda_4 \\ \lambda_5 \\ \lambda_6 \\ \lambda_7 \end{bmatrix} \leq \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$



Why not partitioning?

- For several problems, one could study the set partitioning polytope to derive cuts.
 - ▶ Capacitated Vehicle Routing Problem
 - ▶ Heterogeneous Fleet Vehicle Routing Problem

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- For several problems, one could study the set partitioning polytope to derive cuts.
 - ▶ Capacitated Vehicle Routing Problem
 - ▶ Heterogeneous Fleet Vehicle Routing Problem
- As shown in (Bulhões et al., 2018), there is no significant difference between the complete set packing and the complete set partitioning polytopes!

Why not partitioning?

Theorem 1 (Bulhões et al., 2018)

Let

$$\sum_{v \in K} \lambda_v = 1, \forall K \in \mathcal{K}_=, \quad (0.1)$$

$$\sum_{v \in K} \lambda_v \leq 1, \forall K \in \mathcal{K}_\leq, \quad (0.2)$$

$$H\lambda \leq h, \quad (0.3)$$

$$\lambda_j \geq 0, \forall j \in \{1, \dots, 2^n - 1\} \setminus \{2^0, \dots, 2^{n-1}\}, \quad (0.4)$$

be a minimal description of $\text{CSPP}_=(n)$ where inequalities (0.3) are of the form described in Lemma 3. Then,

$$\sum_{v \in K} \lambda_v \leq 1, \forall K \in \mathcal{K}_= \cup \mathcal{K}_\leq, \quad (0.5)$$

$$H\lambda \leq h, \quad (0.6)$$

$$\lambda_j \geq 0, \forall j \in \{1, \dots, 2^n - 1\}, \quad (0.7)$$

is a minimal description of $\text{CSPP}_\leq(n)$.

Rank 1 Cuts

Rank 1 Cuts

$$0\lambda_1 + 1\lambda_2 + 0\lambda_3 + 1\lambda_4 + 0\lambda_5 \leq 1$$

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λ integer

multipliers vector $u = (0, \frac{1}{2}, 0, \frac{1}{2}, \frac{1}{2}, 0, 0)^\top$

Rank 1 Cuts

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Rank 1 Cuts

$$0(0\lambda_1 + 1\lambda_2 + 0\lambda_3 + 1\lambda_4 + 0\lambda_5 \leq 1)$$

$$\frac{1}{2}(1\lambda_1 + 1\lambda_2 + 1\lambda_3 + 0\lambda_4 + 0\lambda_5 \leq 1)$$

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Summing up:

$$1.5\lambda_1 + \lambda_2 + 0.5(\lambda_3 + \lambda_4 + \lambda_5) \leq 1.5$$

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Rounding
down LHS:

$$\lambda_1 + \lambda_2 \leq 1.5$$

Rounding
down RHS:

$$\lambda_1 + \lambda_2 \leq 1$$

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Subset Row Cuts

- R1Cs with $u = (\frac{1}{k}, \frac{1}{k}, \dots, \frac{1}{k})$
- Examples:
 - ▶ $(\frac{1}{2}, \frac{1}{2}, \dots, \frac{1}{2})$
 - ▶ $(\frac{1}{3}, \frac{1}{3}, \dots, \frac{1}{3})$
- Used in (Jepsen et al., 2008), (Baldacci et al., 2011) and (Contardo and Martinelli, 2014)
- Potentially very effective

The Breakthrough

Pecin et al., 2014 proposed the **the limited memory** technique for greatly reducing the impact of SRCs in the pricing and could obtain the full benefit of those cuts.

- In CVRP: from 150 to 360 customers!

Diego Pecin, Artur Pessoa, Marcus Poggi, and Eduardo Uchoa. Improved branch-and-cut-and-price for capacitated vehicle routing. In Integer Programming and Combinatorial Optimization, pages 393-403, Springer, 2014.

Finding the best multipliers

What are the best multipliers?

Computational study of Pecin et al. (2017) for $n \leq 5$.

- $n = 3 : \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$
- $n = 4 : \left(\frac{2}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right), \left(0, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$
- $n = 5 : \left(0, 0, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right), \left(0, \frac{2}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right), \left(\frac{2}{4}, \frac{2}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}\right), \left(\frac{3}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4}\right),$
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What are the best multipliers?

Theoretical study of Bulhões et al. (2018) for arbitrary n .

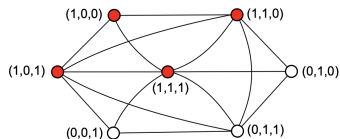
Theorem 2

The following multipliers induce rank 1 facets of $CSPP_{\leq}(n)$.

- (a) $\left(\frac{n-2}{n-1}, \frac{1}{n-1}, \frac{1}{n-1}, \dots, \frac{1}{n-1}\right), n \geq 4$
- (b) $\left(\frac{n-3}{n-1}, \frac{2}{n-1}, \frac{1}{n-1}, \frac{1}{n-1}, \dots, \frac{1}{n-1}\right), n \geq 5$
- (c) $\left(\frac{n-2}{n}, \frac{2}{n}, \frac{2}{n}, \frac{1}{n}, \dots, \frac{1}{n}\right), n \geq 5$
- (d) $\left(\frac{1}{2}, \frac{1}{2}, \dots, \frac{1}{2}\right), \text{ if } n \text{ is odd}$
- (e) $\left(\frac{1}{3}, \frac{1}{3}, \dots, \frac{1}{3}\right), \text{ if } n = 3k + 2, k \in \mathbb{N}, k \geq 1$
- (f) $\left(\frac{n-3}{n-2}, \frac{n-3}{n-2}, \frac{2}{n-2}, \frac{1}{n-2}, \dots, \frac{1}{n-2}\right), n \geq 5$
- (g) $\left(\frac{n-2}{n-1}, \frac{n-2}{n-1}, \frac{2}{n-1}, \frac{2}{n-1}, \frac{1}{n-1}, \dots\right), n \geq 5$

What are the best multipliers?

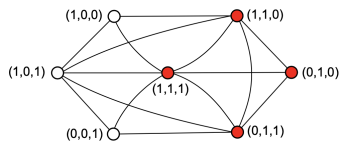
- Clique cuts



$$\begin{bmatrix} 1 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \end{bmatrix} \leq \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

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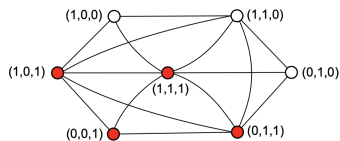
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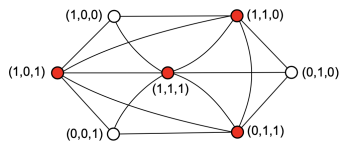
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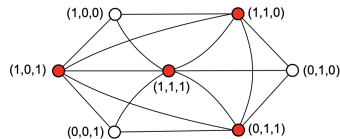
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What are the best multipliers?

- Clique cuts



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The corresponding cut

$$x_3 + x_5 + x_6 + x_7 \leq 1$$

is a R1C with $u = (\frac{1}{2}, \frac{1}{2}, \frac{1}{2})^T$.

What are the best multipliers?

Theoretical study of Bulhões et al. (2018) for arbitrary n

Lemma 1

Any set of multipliers $u = (\frac{\alpha_1}{\beta}, \frac{\alpha_2}{\beta}, \dots, \frac{\alpha_n}{\beta})$, with $\alpha_1, \alpha_2, \dots, \alpha_n$ and β integers, such that $\sum_{i=1}^n \alpha_i = 2\beta - 1$ induces a clique inequality.

Lemma 2

Every rank 1 clique inequality $a^T \lambda \leq 1$ can be generated by multipliers $u = (\frac{\alpha_1}{\beta}, \frac{\alpha_2}{\beta}, \dots, \frac{\alpha_n}{\beta})$, with $\alpha_1, \alpha_2, \dots, \alpha_n$ and β integers, such that $\sum_{i=1}^n \alpha_i = 2\beta - 1$.

Limited Memory Rank 1 Cuts in VRPSolver

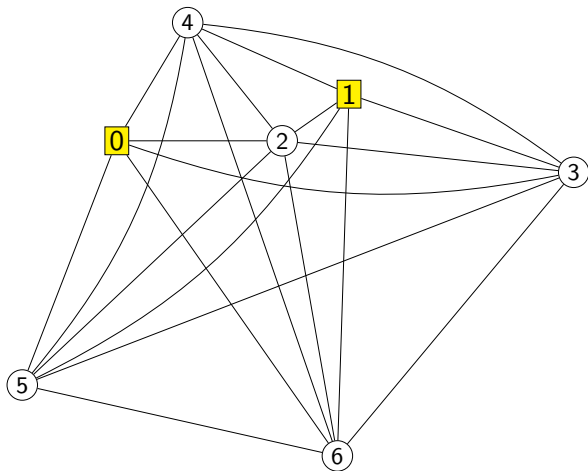
Packing sets concept

Let $\mathcal{B}^V \subset 2^V$ be a collection of mutually disjoint subsets of V such that the constraints:

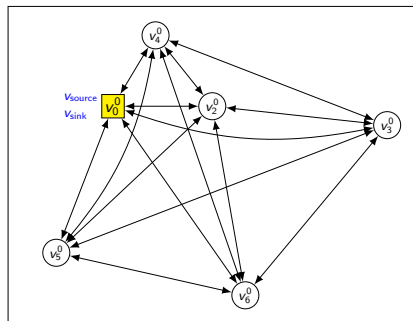
$$\sum_{p \in P} \left(\sum_{v \in B} h_v^p \right) \lambda_p \leq 1, \quad B \in \mathcal{B}^V,$$

are satisfied by at least one optimal solution. In those conditions, we say that the elements of \mathcal{B}^V are *packing sets on vertices*.

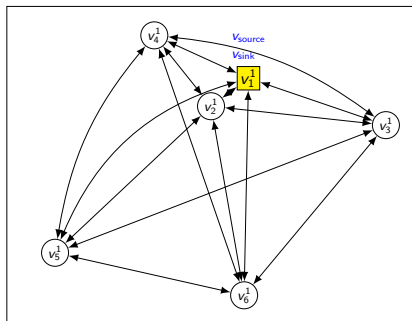
Multi-Depot VRP



Multi-Depot VRP

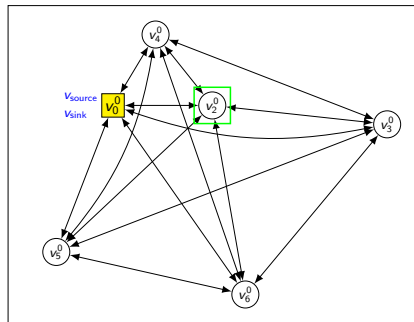


Graph G^0

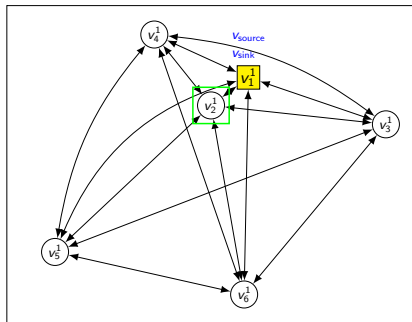


Graph G^1

Multi-Depot VRP



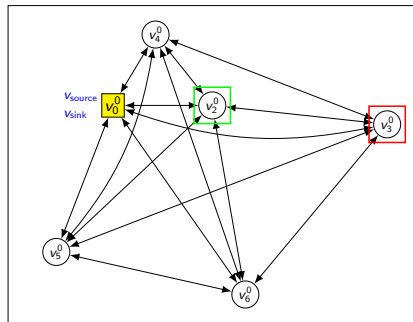
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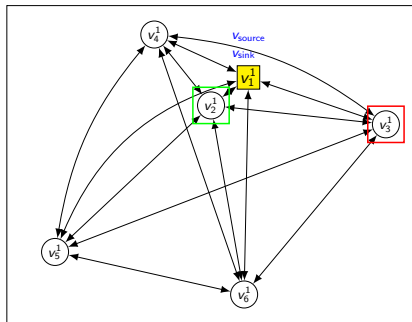
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$$\mathcal{B}^v = \{ \{v_2^0, v_2^1\} \}$$

Multi-Depot VRP



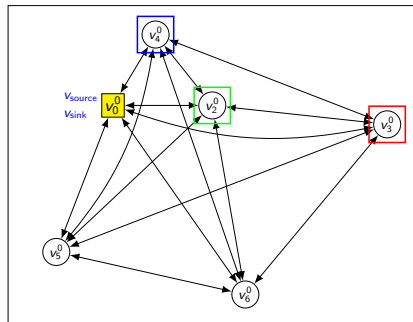
Graph G^0



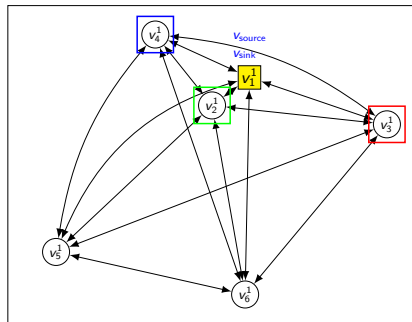
Graph G^1

$$\mathcal{B}^v = \{ \{v_2^0, v_2^1\}, \{v_3^0, v_3^1\} \}$$

Multi-Depot VRP



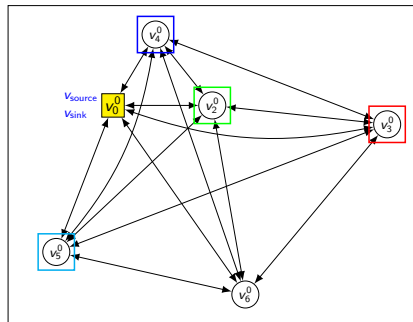
Graph G^0



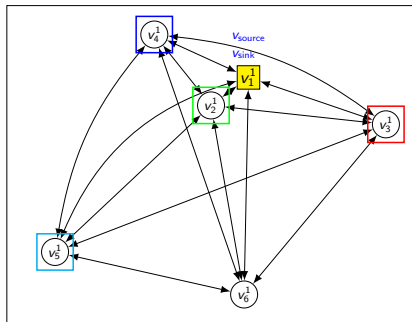
Graph G^1

$$\mathcal{B}^v = \{ \{v_2^0, v_2^1\}, \{v_3^0, v_3^1\}, \{v_4^0, v_4^1\} \}$$

Multi-Depot VRP



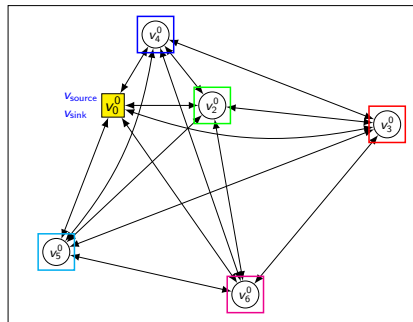
Graph G^0



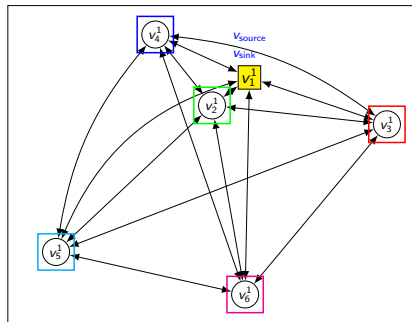
Graph G^1

$$\mathcal{B}^v = \{ \{v_2^0, v_2^1\}, \{v_3^0, v_3^1\}, \{v_4^0, v_4^1\}, \{v_5^0, v_5^1\} \}$$

Multi-Depot VRP



Graph G^0



Graph G^1

$$\mathcal{B}^v = \{ \{v_2^0, v_2^1\}, \{v_3^0, v_3^1\}, \{v_4^0, v_4^1\}, \{v_5^0, v_5^1\}, \{v_6^0, v_6^1\} \}$$

Packing sets: Examples

Capacitated Vehicle Routing Problem:

$$\mathcal{B}^{\mathcal{V}} = \cup_{i \in V_+} \{\{i\}\}$$

Heterogeneous Fleet Vehicle Routing Problem:

$$\mathcal{B}^{\mathcal{V}} = \cup_{i \in V_+} \{\{v_i^k : k \in K\}\}$$

Team Orienteering Problem:

$$\mathcal{B}^{\mathcal{V}} = \cup_{i \in V_+} \{\{i\}\}$$

Concluding remarks

- The limited memory Rank 1 Cuts are a key feature of state-of-the-art exact solvers for several VRPs
- VRPSolver will automatically add them using the packing sets defined by the user
- Future works on these cuts include:
 - ▶ Investigating better separation algorithms
 - ▶ Investigating the impact of the newly proposed multipliers

Thank you for your attention!

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