

# Hyperbolic set covering problems with competing ground-set elements

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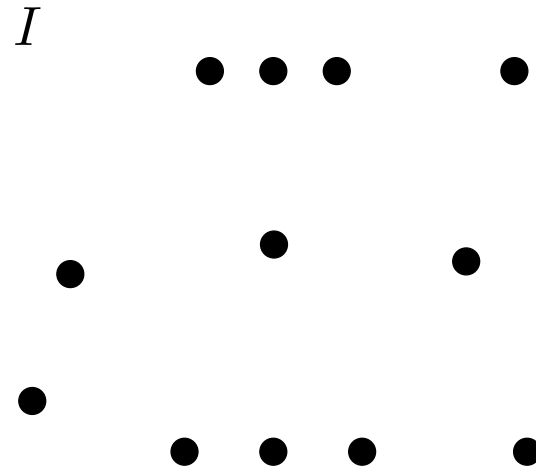
XI Workshop on Combinatorial Optimization, Aussois, 2007

# Outline

- Problems definition
- The motivating application: Wireless Local Area Network design
- Hyperbolic integer programming formulation
- Complexity and Approximability results
- Linearizations and Lagrangean Relaxation
- Ongoing work and concluding remarks

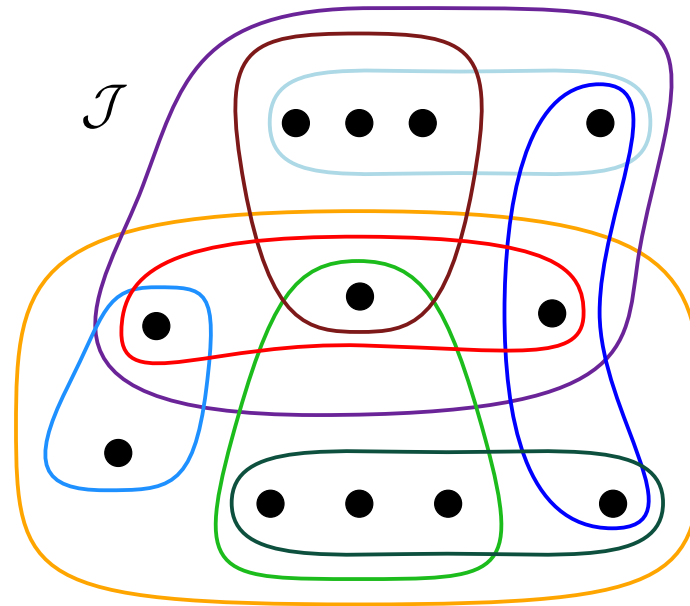
# Set Covering notation

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$I$ : a finite groundset

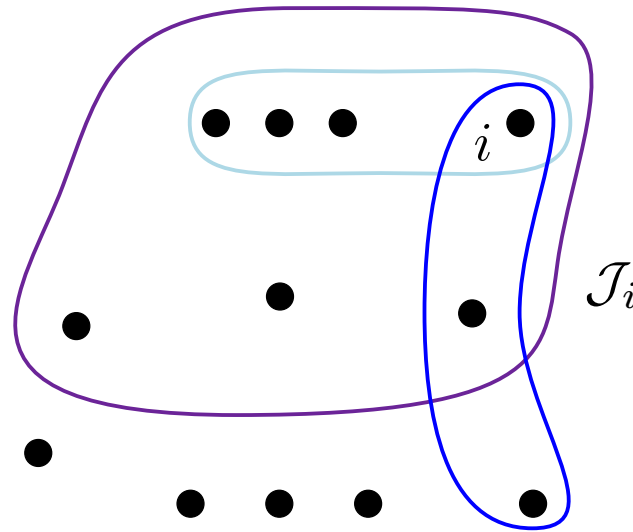
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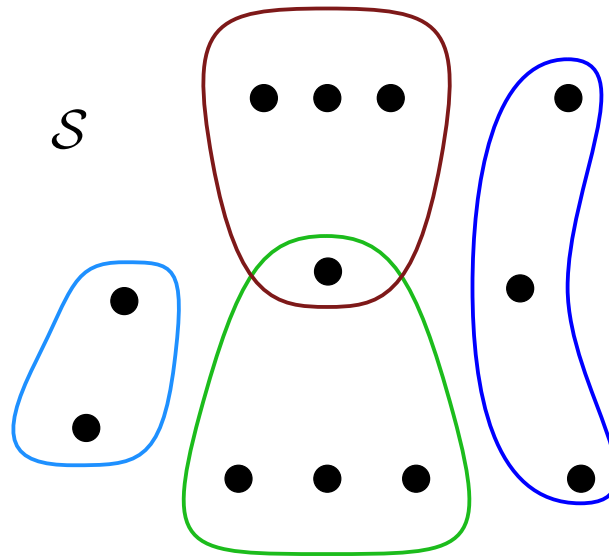


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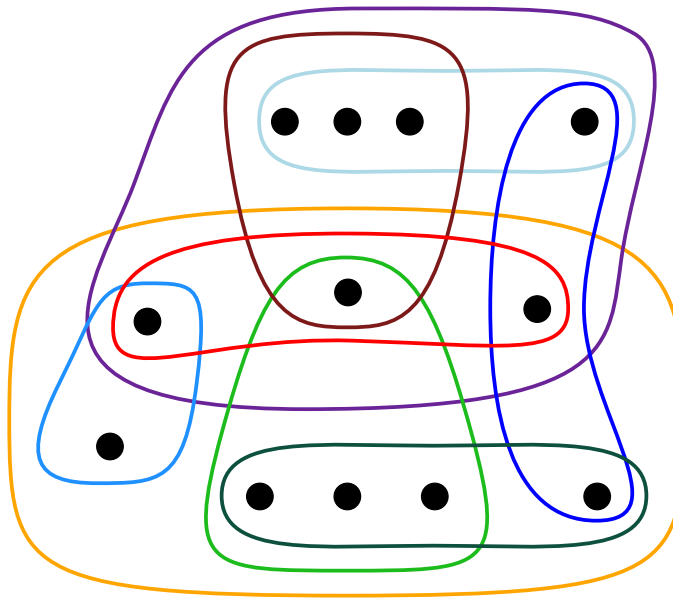
$\mathcal{J}_i \subseteq \mathcal{J}$ : subcollection of the subsets **covering an element**  $i \in I$

**cover**  $S$ : a subcollection indexed by  $S \subseteq J$  such that  $\bigcup_{j \in S} I_j = I$

# Set Covering problems

## Classical Set Covering Problem (SCP):

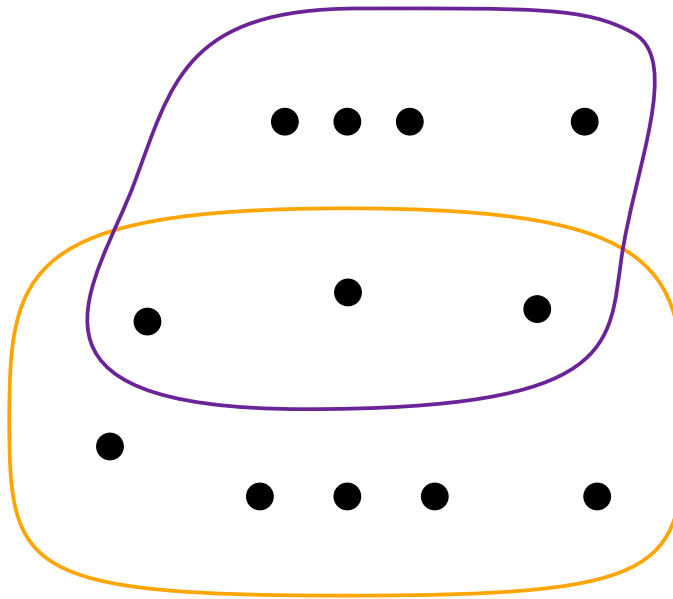
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**Set Partitioning**      **forbidden** overlap

**Set Multicover**      **required** overlap

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Also:

- Quadratic objective functions
- Maximum coverage
- ...

# Coverage share

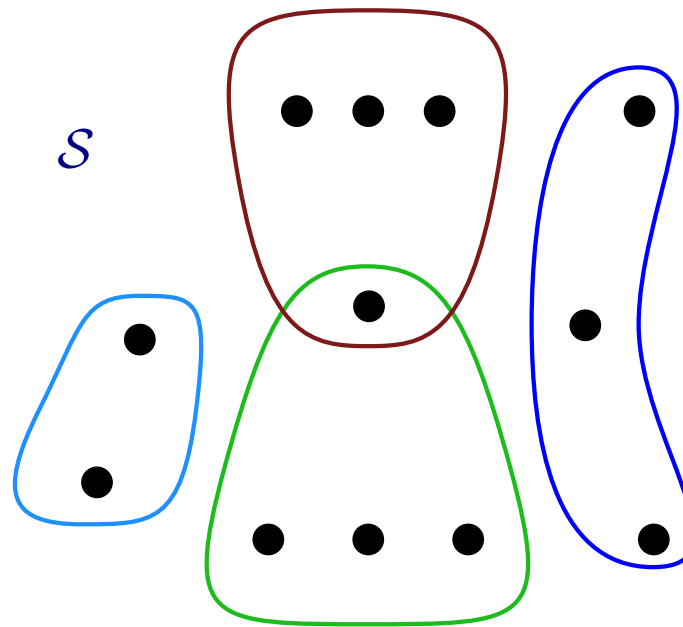
Given a covering instance  $(I, \mathcal{J})$ , a cover  $\mathcal{S}$  and an element  $i \in I$

coverage share: 
$$r(\mathcal{S}, i) = \frac{1}{1 + |N_i(\mathcal{S})|}$$

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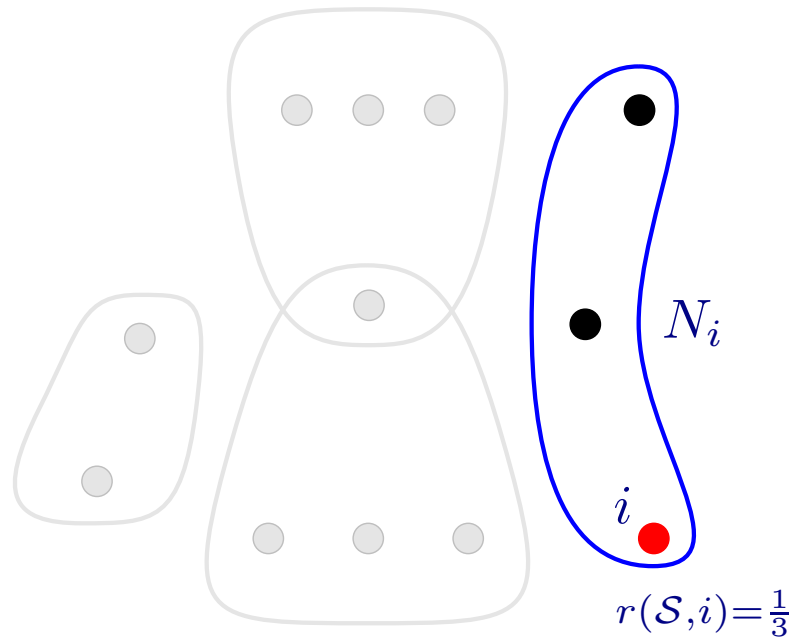
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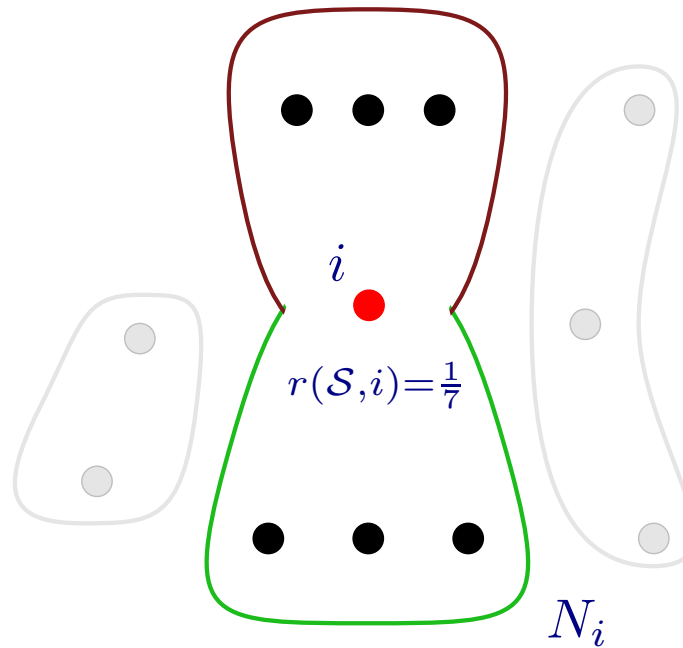
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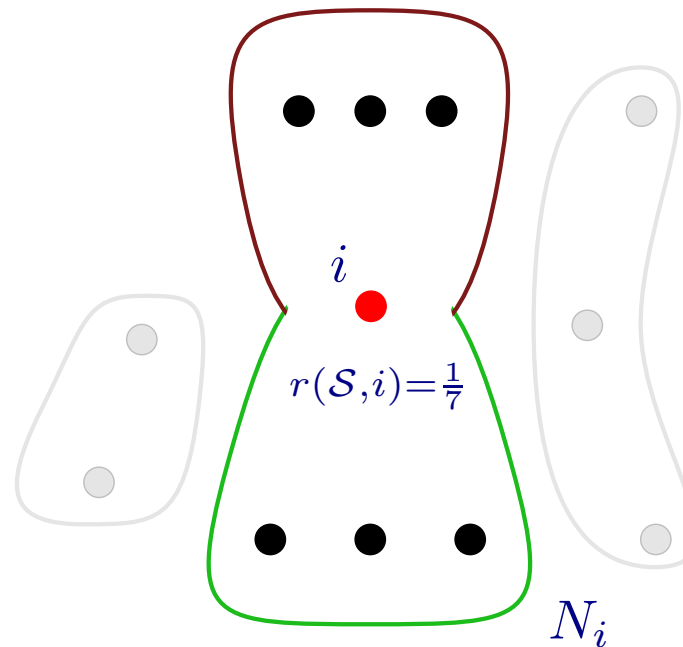
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Fraction of resource received by  $i$  assuming **fair** allocation  
among the **competing elements** (neighbors of  $i$ )

# Coverage share problems

## Maximum Total Coverage Share Problem (TCSP):

Given an instance  $(I, \mathcal{J})$ , find a **cover**  $\mathcal{S}$  that **maximizes**

$$f_t(\mathcal{S}) = \sum_{i \in I} \frac{1}{1 + |N_i(\mathcal{S})|}$$

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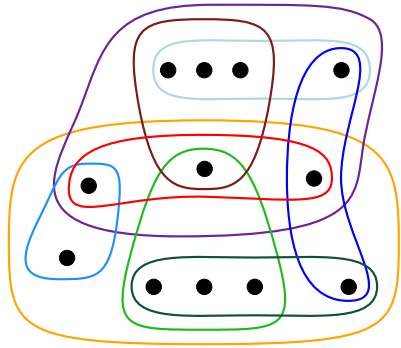
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Set covering problems with **competing** ground-set elements

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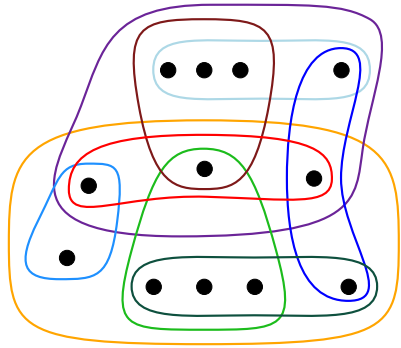
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Instance

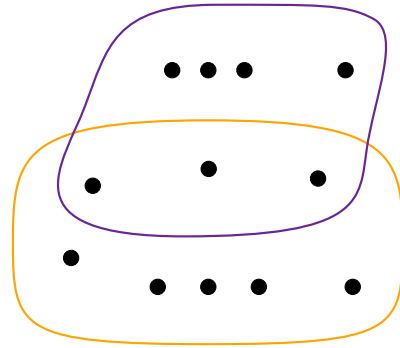


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SCP opt



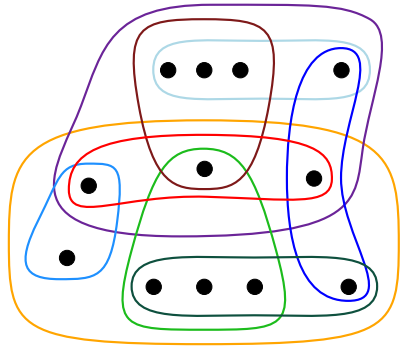
$$|\mathcal{S}| = 2$$

$$f_t(\mathcal{S}) = 1.40$$

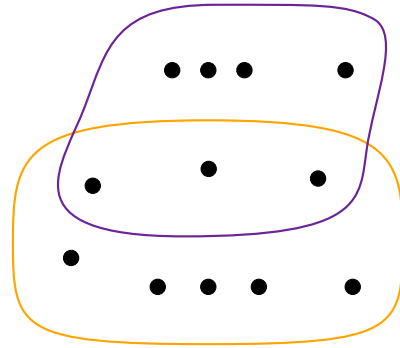
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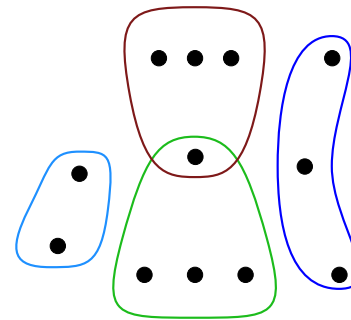


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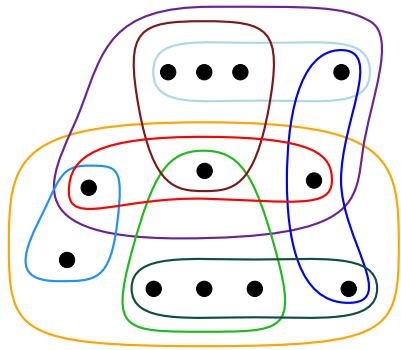
$$|\mathcal{S}| = 4$$

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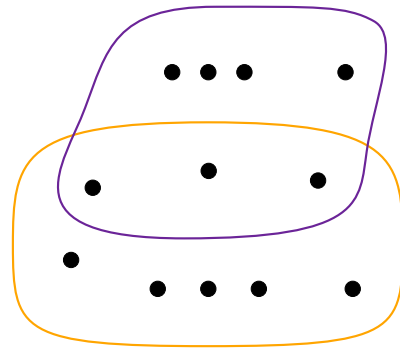
$$f_m(\mathcal{S}) = \frac{1}{7}$$

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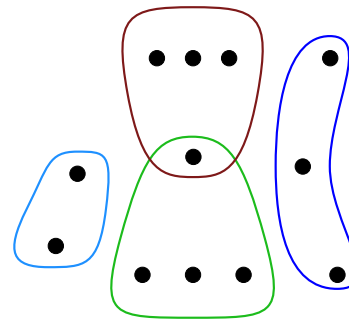


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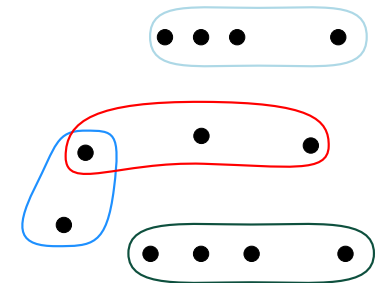


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MCSP opt



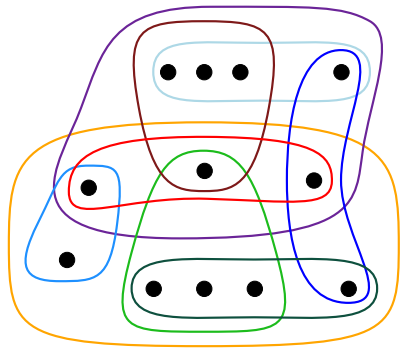
$$|\mathcal{S}| = 4$$

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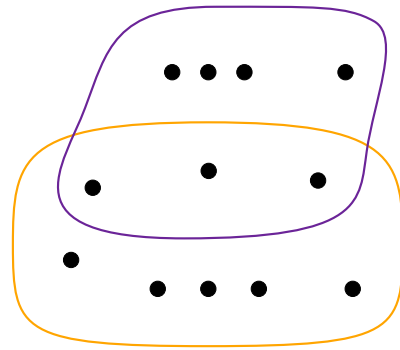
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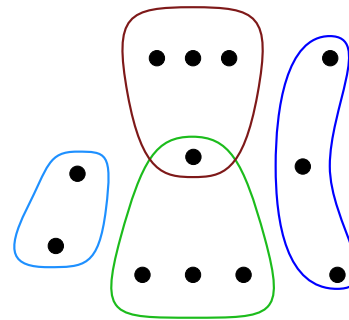
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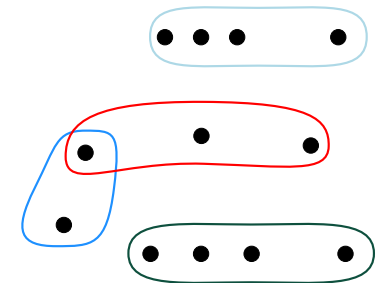
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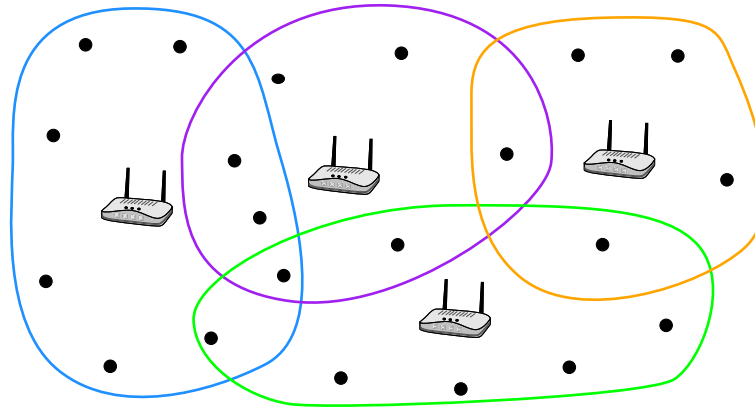
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Privilege covers whose subsets have **small cardinality** and **limited overlaps**.

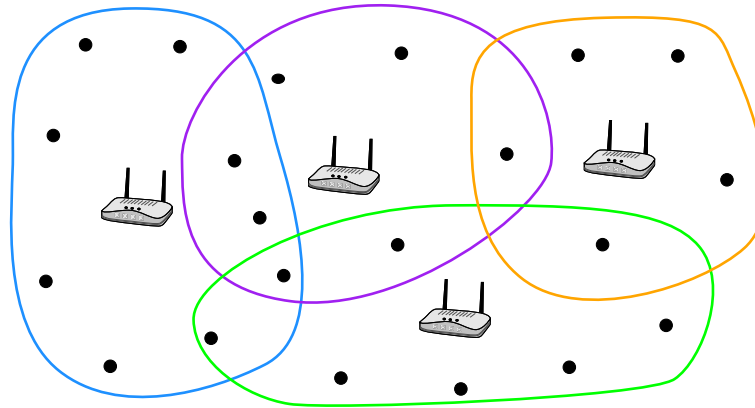
# Wireless Local Area Network

IEEE 802.11 WLAN: a set of **Access Points** each able of serving a set of **users**



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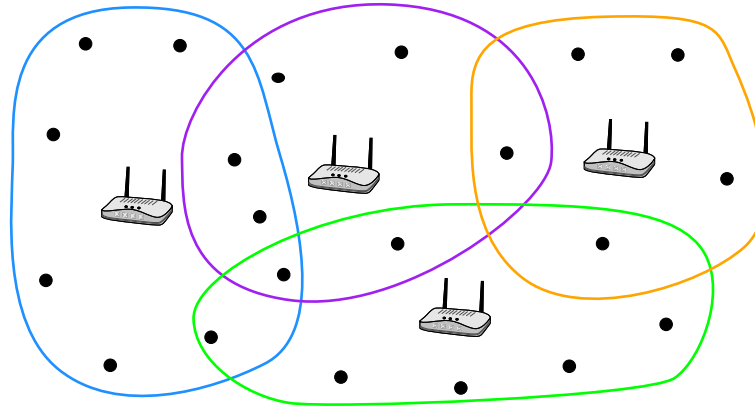
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WLANs are becoming pervasive in airports, trains and train stations, private companies, universities, hotels, ...

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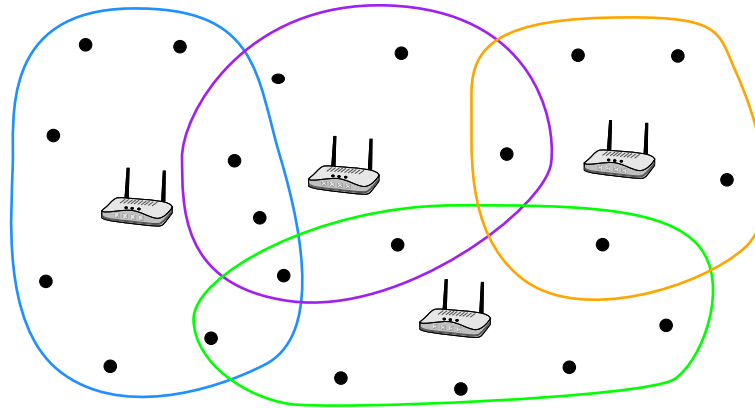


Medium Access Control (MAC) Protocol:

A user can access the network if and only if **no other user**  
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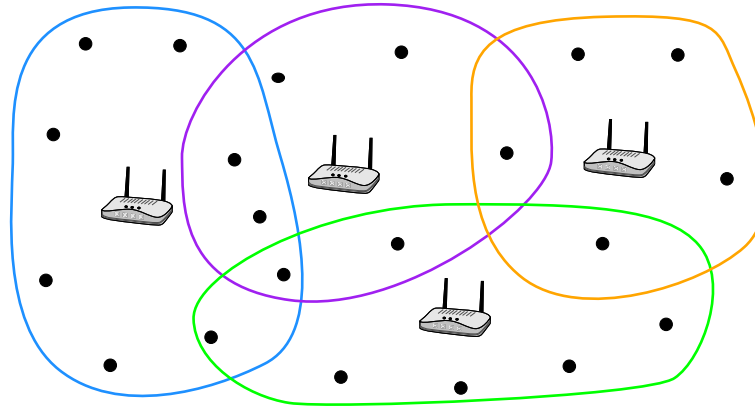
Medium Access Control (MAC) Protocol:

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Assuming uniform **peak** traffic and **fair** access after collision, **coverage share** of element  $i \approx$  **fraction of time** used by user  $i$

# Wireless Local Area Network

IEEE 802.11 WLAN: a set of **Access Points** each able of serving a set of **users**



Due to protocol issues, increasing sizes of deployed WLANs and limited resources,  
and optimization models and methods can be  
**very useful** to support the planning decisions.

# Previous and related work

## WLAN design:

- Large-scale WLAN design (Hills 01, ...)
- Max average signal quality in test points (Rodrigues, Mateus and Loureiro 00/01)
- Max coverage level (Kamenetsky and Unbehaun 02)
- Max capacity based on constraint satisfaction (Prommak et al. 02)
- ...
- First hyperbolic model and heuristics (Amaldi, Capone, Cesana and Malucelli 04)

# Integer programming formulations

$$\max \sum_{i \in I} \frac{1}{1 + |N_i(\mathcal{S})|}$$

(TCSP)  $s.t.$   $\bigcup_{j \in \mathcal{S}} I_j = I$  complete coverage

$\mathcal{S} \subseteq \mathcal{J}$  select subcollection

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Variables:

$x_j = 1$  if subset  $I_j$  is selected (0 otherwise)

$y_{ih} = 1$  if elements  $i$  and  $h$  are neighbors (0 otherwise)

# Integer programming formulations

$$\begin{aligned} & \max \sum_{i \in I} \frac{1}{1 + \sum_{h \in N_i} y_{ih}} \\ \text{(TCSP)} \quad & s.t. \sum_{j \in J_i} x_j \geq 1 && i \in I \\ & y_{ih} \geq x_j && i \in I, h \in N_i, j \in J_i \cap J_h \\ & y_{ih} \leq \sum_{j \in J_i \cap J_h} x_j && i \in I, h \in N_i \\ & x_j \in \{0, 1\} && j \in J \\ & y_{ih} \in \{0, 1\} && i \in I, h \in N_i \end{aligned}$$

# Integer programming formulations

$$\max \sum_{i \in I} \frac{1}{1 + \sum_{h \in N_i} y_{ih}} \quad \rightarrow \quad 0\text{-1 hyperbolic sum problem}$$

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# Integer programming formulations

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 \text{(MCSP)} \quad \max \quad \min_{i \in I} \frac{1}{1 + \sum_{h \in N_i} y_{ih}} = \frac{1}{1 + \min_{i \in I} \max_{h \in N_i} \sum y_{ih}} \\
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# Connection with Quadratic SCP

## Quadratic Set Covering Problem (QSCP):

Given  $(I, \mathcal{J})$ ,  $Q = \{q_{j\ell} \in \mathbb{R} : j, \ell \in J\}$  (wlog symmetric with zero diagonal) and  $\mathbf{c} = \{c_j \in \mathbb{R} : j \in J\}$ , find a **cover**  $\mathcal{S} \subseteq \mathcal{J}$  that **maximizes**

$$q(\mathcal{S}) = \frac{1}{2} \sum_{j \in \mathcal{S}} \sum_{\ell \in \mathcal{S}} q_{j\ell} + \sum_{j \in \mathcal{S}} c_j$$

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Choice:

$$c_j = \sum_{i \in I_j} \frac{1}{|I_j|} \quad q_{j\ell} = \sum_{i \in I_j \cap I_\ell} \left( \frac{1}{|I_j \cup I_\ell|} - \frac{1}{|I_j|} - \frac{1}{|I_\ell|} \right) \quad (\text{for } j \neq \ell)$$

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Then we can verify that:

$f_t(\mathcal{S}) = q(\mathcal{S})$  if at most two subsets overlap

$f_t(\mathcal{S}) \geq q(\mathcal{S})$  otherwise (overestimated penalty)

# Previous and related work

## Unconstrained 0-1 Hyperbolic Programming

- Single-ratio: NP-hard, poly with positive denominator (Hammer and Rudeanu 68)
- Multiple-ratio: NP-hard; tackled by SA, Tabu, and decomposing into independent polynomial single-ratio problems (Hansen, Poggi de Aragao and Ribeiro 90/91)

## Constrained 0-1 Hyperbolic Programming

- Single-ratio (Stancu-Minasian 97)
- Multiple-ratio: MILP convex reformulations (Tawarmalani, Ahmed and Sahinidis 02)

## Quadratic Set Covering Problem

- Various application oriented works (Bazaraa 75, Boros, Hammer et al. 00, ...)
- Generic: not  $2^{p(|I|)}$ -approximable for any polynomial  $p()$  (Escoffier and Hammer 05)
- Convex: approximable within  $O(\ln^2(|I|))$  but not within  $\rho \ln^2(|I|)$

# Complexity and Approximability (TCSP)

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Generic instances

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- Strongly NP-hard (Amaldi et al. 04)
- Not approximable within  $\rho (\sqrt{|I|})^{1-\varepsilon}$  or  $\rho (\sqrt{|\mathcal{J}|})^{1-\varepsilon}$  for a given  $\rho > 0$  and any  $\varepsilon > 0$ , unless  $\text{NP} = \text{ZPP}$

Reduction from Max Independent Set

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## Euclidean 2D Instances

- Strongly NP-hard (does not admit a FPTAS unless  $\text{P} = \text{NP}$ )
- Under a reasonable restriction, admits a PTAS

Adapting and extending a reduction for Disc-Cover (Fowler et al. 81)  
Using the shifting lemma

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## Euclidean 1D Instances (or instances with C1C covering matrix)

# Complexity and Approximability (TCSP)

## Generic instances

- Strongly NP-hard (Amaldi et al. 04)
- Not approximable within  $\rho (\sqrt{|I|})^{1-\varepsilon}$  or  $\rho (\sqrt{|\mathcal{J}|})^{1-\varepsilon}$  for a given  $\rho > 0$  and any  $\varepsilon > 0$ , unless  $\text{NP} = \text{ZPP}$

Reduction from Max Independent Set

## Euclidean 2D Instances

- Strongly NP-hard (does not admit a FPTAS unless  $\text{P} = \text{NP}$ )
- Under a reasonable restriction, admits a PTAS

Adapting and extending a reduction for Disc-Cover (Fowler et al. 81)

Using the shifting lemma

## Euclidean 1D Instances (or instances with C1C covering matrix)

- Polynomial-time solvable

Longest path on an appropriate directed acyclic digraph

# Complexity and Approximability (MCSP)

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Generic instances

# Complexity and Approximability (MCSP)

## Generic instances

- Strongly NP-hard (Amaldi et al. 04)
- Polynomial-time solvable if  $|I_j| = 2$

Reduction to perfect b-matching

# Complexity and Approximability (MCSP)

## Generic instances

- Strongly NP-hard (Amaldi et al. 04)
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## Euclidean 2D Instances

# Complexity and Approximability (MCSP)

## Generic instances

- Strongly NP-hard (Amaldi et al. 04)
- Polynomial-time solvable if  $|I_j| = 2$   
Reduction to perfect b-matching

## Euclidean 2D Instances

- Strongly NP-hard (does not admit a FPTAS unless  $P = NP$ )  
Adapting the reduction for TCSP
- Not approximable within  $3/2$  unless  $P = NP$   
Consequence of the above reduction
- Under a reasonable restriction, approximable within a factor 3  
Tiling with hexagons

# Linearization

For each ratio  $\frac{1}{1 + \sum_{h \neq i} y_{ih}}$  is introduced a variable  $r_i \geq 0$  and the quadratic constraint

$$r_i = \frac{1}{1 + \sum_{h \neq i} y_{ih}} \quad \equiv \quad r_i + \sum_{h \neq i} r_i y_{ih} = 1$$

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$r_i \cdot y_{ih}$  is standardly linearized with a variable  $z_{ih} \geq 0$  and the constraints

$$z_{ih} \leq u_i y_{ih}$$

$$z_{ih} \geq l_i y_{ih}$$

$$z_{ih} \geq r_i + u_i (y_{ih} - 1)$$

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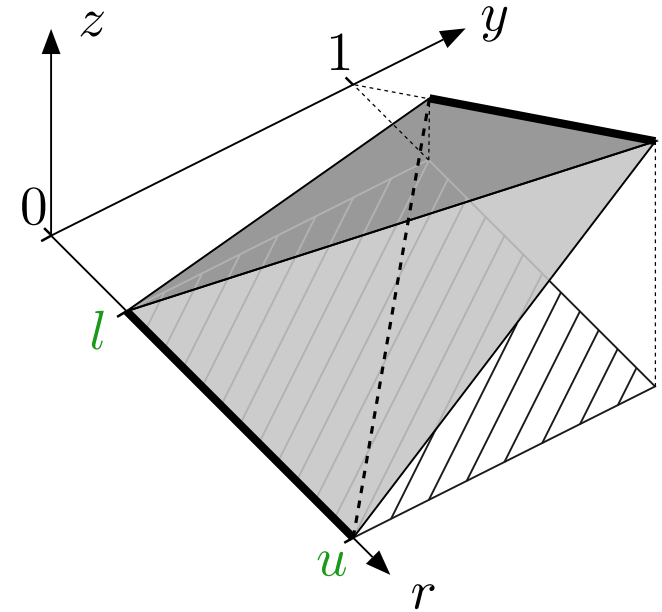
$$z_{ih} \leq r_i + l_i(y_{ih} - 1)$$

NB:  $r_i$  is continuous and bounded, and  $y$  is binary

# Tightening linearization of bilinear terms

$$Z = \{(r, y, z) : z = r \cdot y, r \in [l, u], y \in \{0, 1\}\}$$

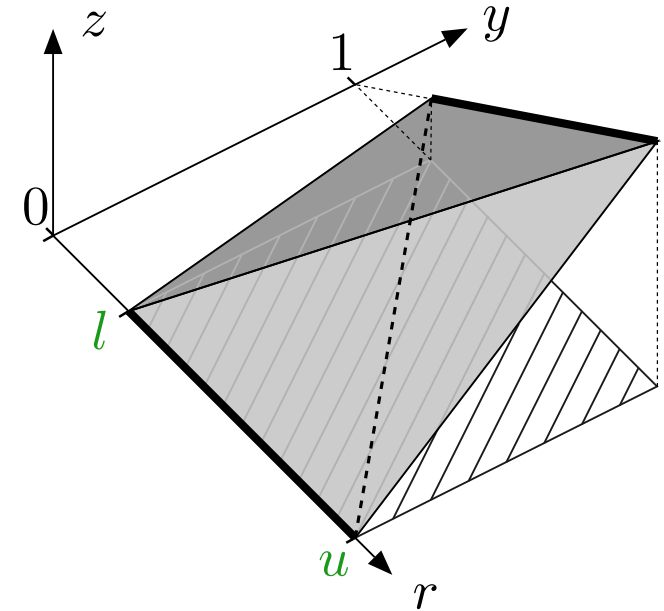
$$\text{conv}(Z) = \left\{ \begin{array}{ll} z \geq ly, & z \geq r + u(y - 1), \\ z \leq uy, & z \leq r + l(y - 1) \end{array} \right\}$$



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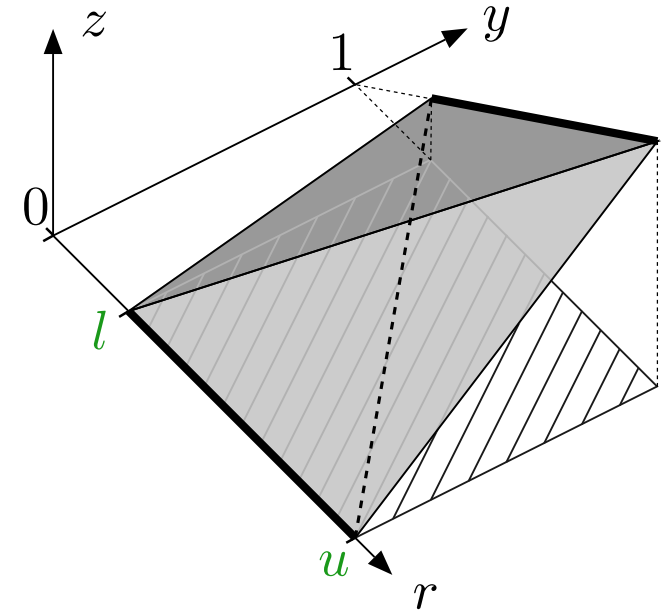


$$Z = \{(r, y, z) : z = 0, r \in [l, u], y = 0\} \\ \cup \{(r, y, z) : z = r, r \in [l, u], y = 1\}$$

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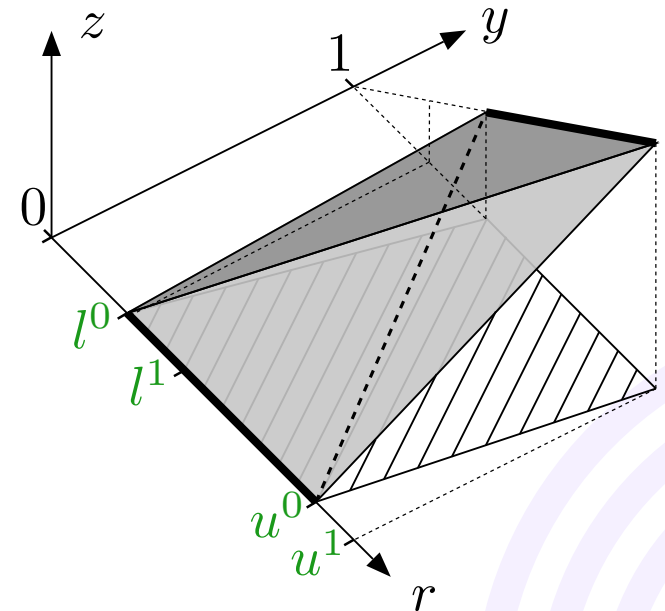
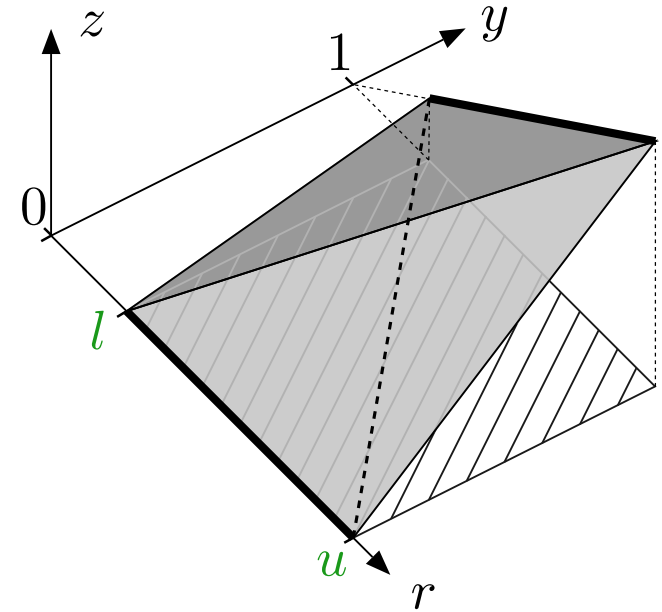
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# Lagrangian relaxation

By applying Lagrangian relaxation to an appropriate reformulation the problem is decomposed into **smaller** and **easier** subproblems.

# Lagrangian relaxation

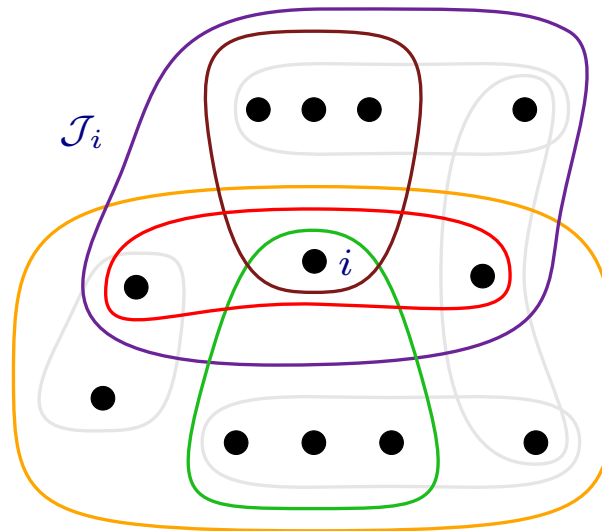
By applying Lagrangian relaxation to an appropriate reformulation the problem is decomposed into **smaller** and **easier** subproblems.

Expanded formulation obtained by adding for each  $i \in I$  a vector  $\chi_i = \{\chi_{ij} : j \in J_i\}$  of binary variables, one for each covering subset. Incidence vector of a **local covering solution** for  $i$ .

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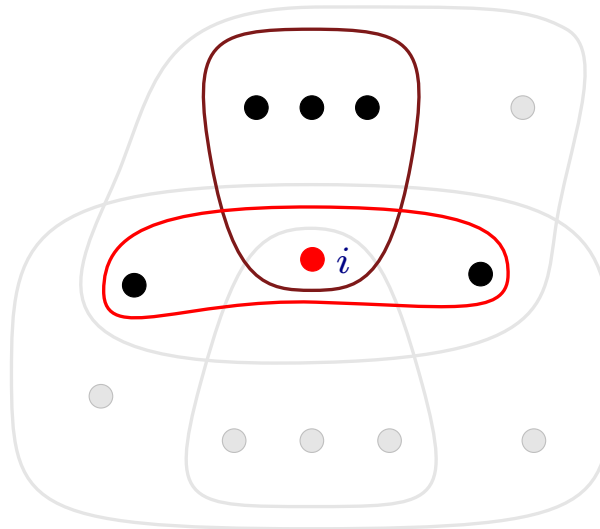
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# Lagrangean relaxation

Expanded formulation:

$$\begin{array}{ll}
 \max & \sum_{i \in I} \frac{1}{1 + \sum_{h \in N_i} y_{ih}} \\
 \text{s.t.} & \sum_{j \in J_i} x_j \geq 1 \quad i \in I \\
 & y_{ih} \geq x_j \quad i \in I, h \in N_i, j \in J_i \cap J_h \\
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 & x_j, \chi_{ij}, y_{ih} \in \{0, 1\}
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Several possibilities, depending on which constraints are deleted/dualized

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Without (2), (6) and (7): one SCP and  $|I|$  independent hyperbolic subproblems

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$\text{LAG}_a$ : remove (2) and dualize (6) and (7)  $\rightarrow$  NP-hard hyperbolic subproblems

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LAG<sub>b</sub>: remove (5), (6) and dualize (2), (7) → polynomial hyperbolic subproblems

# Lagrangian subproblem for $LAG_b$

Problem for a given element  $i$ :

$$\max \quad \frac{1}{1 + \sum_{h \in N_i} y_h} + \sum_{h \in N_i} c_h y_h$$

$$s.t. \quad \sum_{j \in J_i} \chi_j \geq 1$$

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Fix one variable  $\chi_\ell$  to 1 (try all). This covers all  $h \in I_\ell$ .

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Remains an unconstrained problem with hyperbolic+linear o.f..

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1) sort  $c_h$  coefficients in nonincreasing order.

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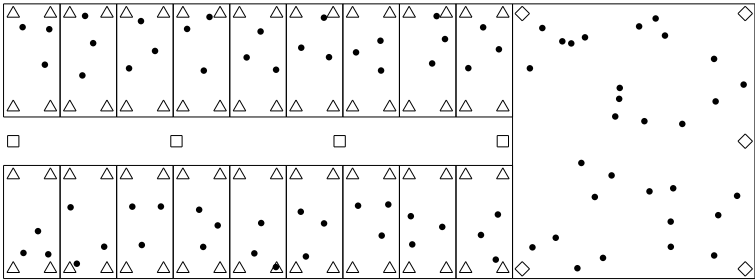
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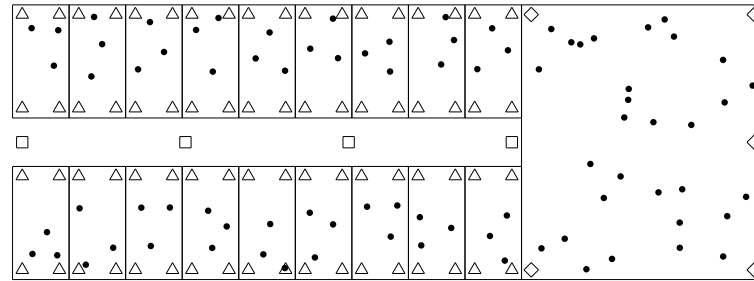
- 1) sort  $c_h$  coefficients in nonincreasing order.
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# Comparison - our department

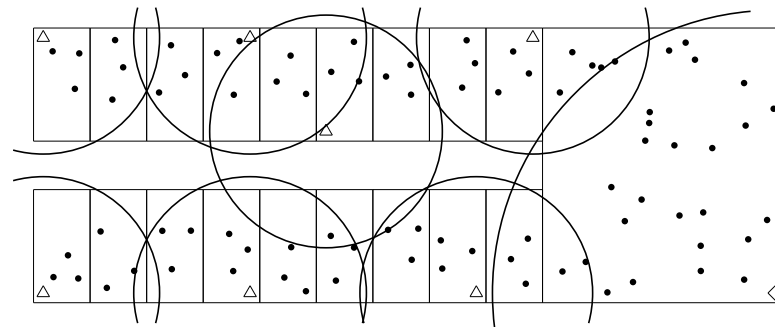


Our Department

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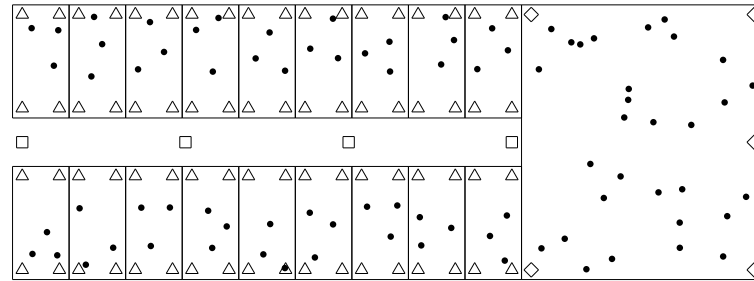


TCSP best solution

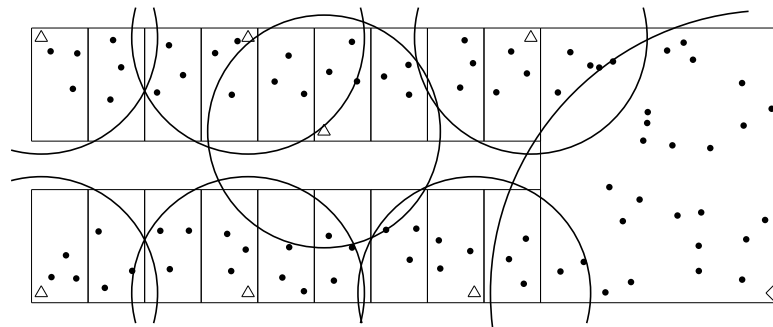
			Standard Linearization		Improved Linearization		LAG <sub>b</sub>	
J	I	den	gap	time	gap	time	gap	time
			(%)	(sec)	(%)	(sec)	(%)	(sec)
81	84	13.7	9.28	—	5.08	—	0.87	237.2

— : time limit exceeded

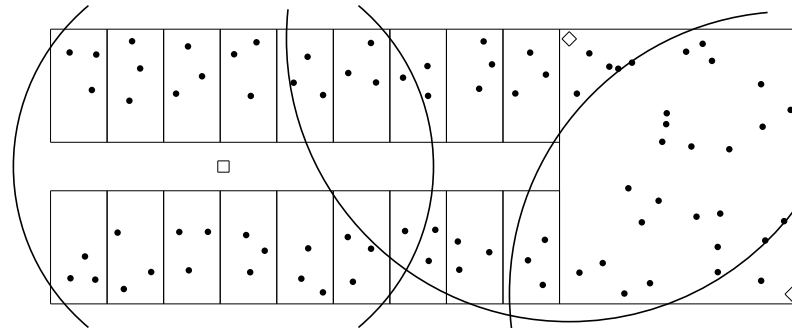
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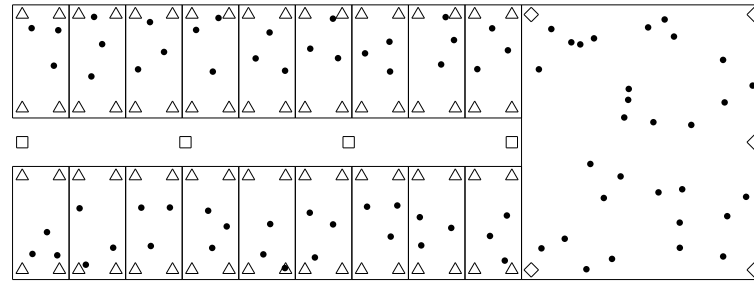


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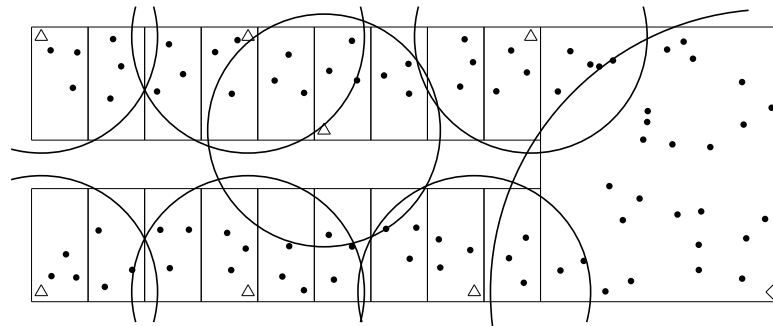


SCP optimal solution

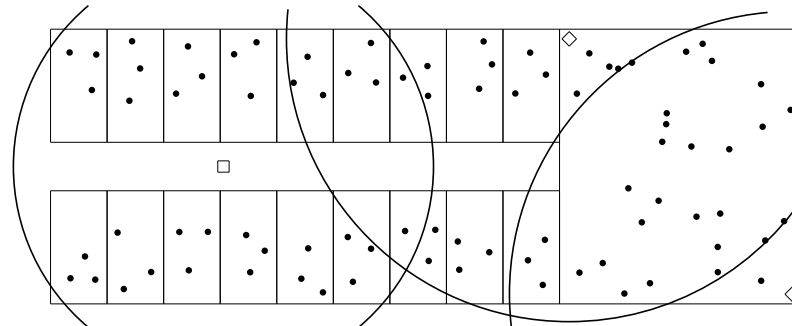
# Comparison - our department



Our Department



TCSP best solution



SCP optimal solution

Tests with a WLAN simulator (ns-2):

2.58 Mb/s for SCP solution, 15.8 Mb/s for TCSP solution

# Comparison - synthetic instances

J	I	den	Standard Linearization				Improved Linearization				LAG <sub>b</sub>			
			gap	stdev	time	stdev	gap	stdev	time	stdev	gap	stdev	time	stdev

## GEOMETRIC INSTANCES (LOW DENSITY)

50	50	6.6	*		0.4	0.3	*		0.1	0.1	*		0.5	0.3
50	100	6.4	*		7.2	4.1	*		4.1	4.1	0.11	0.15	5.3	2.8
100	100	5.3	0.79	1.51	1826.6	1639.3	*		177.1	203.1	0.28	0.28	24.8	13.6
100	200	5.1	9.42	2.57	—		3.07	1.88	3363.0	530.0	0.37	0.22	409.6	129.5
50	300	6.3	*		561.6	557.5	*		395.3	353.9	*		621.8	137.9

## GEOMETRIC INSTANCES (HIGH DENSITY)

50	50	10.5	*		15.5	20.3	*		6.1	5.7	0.17	0.32	2.0	1.7
50	100	10.3	*		798.6	452.7	*		313.6	188.9	0.10	0.14	52.5	29.8
100	100	10.8	27.76	3.42	—		12.26	3.43	—		1.89	0.61	215.0	24.8
100	200	10.6	33.17	3.50	—		18.96	1.08	—		2.26	1.14	1111.2	44.5
50	300	11.1	27.12	7.28	—		26.43	6.69	—		0.98	0.69	1494.4	140.7

## STANDARD SCP INSTANCES (CLASS SCP4\*)

1000	200	2.0	11.53	1.69	—		6.13	1.65	—		0.12	0.09	3304.6	403.6
------	-----	-----	-------	------	---	--	------	------	---	--	------	------	--------	-------

## STANDARD SCP INSTANCES (CLASS SCPE\*)

500	50	20.0	71.22	4.69	—		35.75	16.62	—		6.14	0.78	1765.9	148.1
-----	----	------	-------	------	---	--	-------	-------	---	--	------	------	--------	-------

- \* : the primal-dual gap is zero (proven optimality)  
 — : time limit exceeded for all instances of the class

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