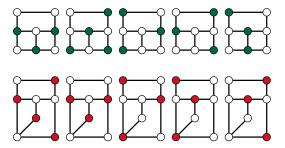
Local approximation algorithms for scheduling problems in sensor networks

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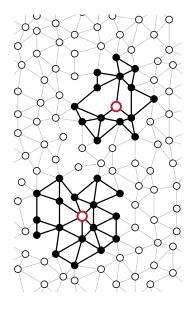
Algosensors 14 July 2007



Local algorithms

- Operation of a node only depends on input within its constant-size neighbourhood
- Extreme scalability: constant amount of communication, memory and computation per node
- Weak model: 3-colouring a cycle impossible (Linial 1992)

Our result: local algorithms can be used to approximate nontrivial **scheduling problems**

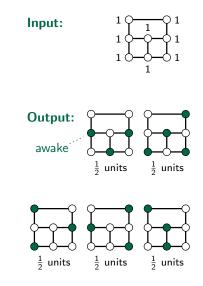


Sleep scheduling

Input: redundancy graph, battery capacities

- Set of awake nodes = dominating set of redundancy graph
- Associate a time period with each dominating set
- Maximise total length
- Obey battery constraints

Motivation: maximising lifetime of a sensor network (pairwise redundancy)

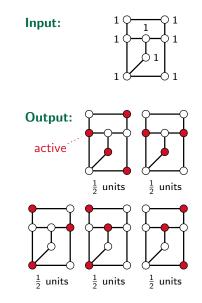


Activity scheduling

Input: **conflict graph**, activity requirements

- Set of active nodes = independent set of conflict graph
- Associate a time period with each independent set
- Minimise total length
- Fulfil activity requirements

Motivation: minimising makespan of radio transmissions (pairwise interference)

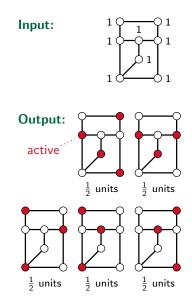


Scheduling problems

Sleep scheduling: generalisation of fractional domatic partition

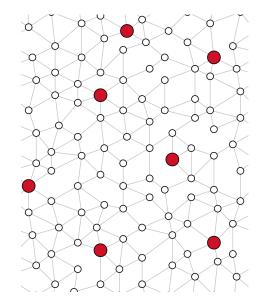
Activity scheduling: generalisation of fractional graph colouring

- Linear programs
- The size of the LP can be exponential in the size of the graph
- Hard to solve and approximate in general graphs



Solution

- Hard problems
- Weak model of computation
- Solution: markers
- 1. Markers break symmetry
- Characterisation of marker distribution constrains the family of graphs

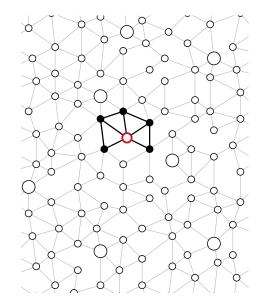


Marked graphs

 $(\Delta, \ell_1, \ell_\mu, \mu)$ -marked graph:

- ► Degree $\leq \Delta$
- ► ≥ 1 marker within ℓ₁ hops from any node
- ► ≤ µ markers within ℓ_µ hops from any node

Intuition: bounded growth, symmetrybreakers nearby

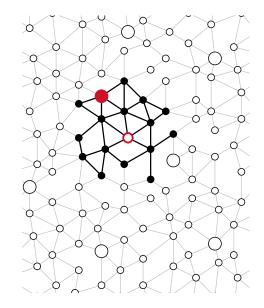


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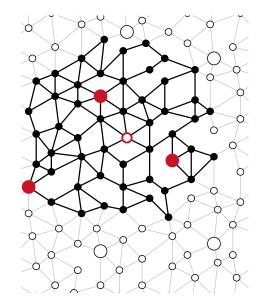


Marked graphs

 $(\Delta, \ell_1, \ell_\mu, \mu)$ -marked graph:

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- $\leq \mu$ markers within ℓ_{μ} hops from any node

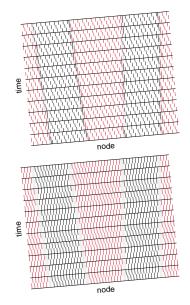
Intuition: bounded growth, symmetrybreakers nearby



Local $(1 + \epsilon)$ -approximation algorithm for sleep scheduling in $(\Delta, \ell_1, \ell_\mu, \mu)$ -marked graphs for any $\epsilon > 4\Delta / \lfloor (\ell_\mu - \ell_1) / \mu \rfloor$

Local $1/(1 - \epsilon)$ -approximation algorithm for activity scheduling in $(\Delta, \ell_1, \ell_\mu, \mu)$ -marked graphs for any $\epsilon > 4/\lfloor (\ell_\mu - \ell_1)/\mu \rfloor$

- Markers are enough: no coordinates needed
- Markers are necessary
- Cannot improve ϵ by factor 9

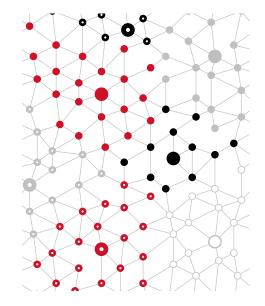


Algorithm sketch

Several partitions of communication graph

- Configuration 0: Voronoi cells for markers
- Configuration 1: shift cell borders
- Configuration i: shift i units

Solve the scheduling problem locally for each cell, interleave the solutions

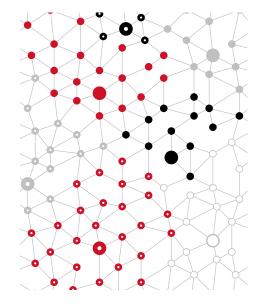


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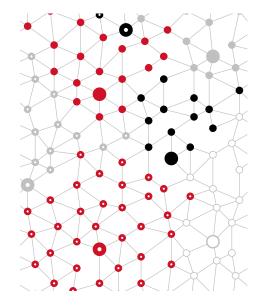


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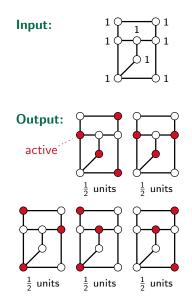
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Summary

- Local approximation scheme: constant effort per node
- Fractional scheduling problems, both packing and covering
- Can be extended beyond pairwise redundancy/conflicts as long as there is "locality"
- Markers are enough, coordinates not needed
- Constants are not practical, more work needed

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Appendix: Examples of marked graphs

- 2-dimensional grid of nodes
 - Use a sparser grid to place the markers
 - "Local approximation scheme": any approximation ratio by using a sparse enough grid (cost: higher computational complexity)
- "Coarse grids", graphs quasi-isometric to 2-dimensional grids
 - Arbitrary small-scale structure
- Cutting parts of coarse grids, with L + 1 hop margins
 - Arbitrary small-scale and large-scale structure
 - Medium-scale structure has similarities with low-dimensional grids

Appendix: Sleep scheduling LP

Input:

- communication graph ${\cal G}$
- redundancy graph \mathcal{R} , subgraph of \mathcal{G}
- battery capacity $b(v) \geq 0$ for each node $v \in V_{\mathcal{R}}$

Task:

maximise $\sum_D x(D)$ subject to $\sum_D D(v)x(D) \le b(v)$ and $x(D) \ge 0$

v ranges over $V_{\mathcal{R}}$ D ranges over dominating sets of $\mathcal R$

D(v) = 1 if $v \in D$ and D(v) = 0 if $v \notin D$ x(D) = the length of the time period associated with D

Appendix: Activity scheduling LP

Input:

- communication graph ${\cal G}$
- conflict graph ${\mathcal C}$, subgraph of ${\mathcal G}$
- activity requirement $a(v) \geq 0$ for each node $v \in V_{\mathcal{C}}$

Task:

minimise $\sum_{I} x(I)$

subject to $\sum_{I} I(v)x(I) \ge a(v)$ and $x(I) \ge 0$

```
v ranges over V_{\mathcal{C}} l ranges over independent sets of \mathcal C
```

I(v) = 1 if $v \in I$ and I(v) = 0 if $v \notin I$ x(I) = the length of the time period associated with I