

# Lifetime Maximization of Monitoring Sensor Networks

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Institute for Theoretical Informatics - Algorithms II



# Introduction

## Motivation

- fixed area
  - stationary sensor nodes with
    - circular monitoring areas,
    - limited power supply
- monitor entire region  
as long as possible
- schedule node activation



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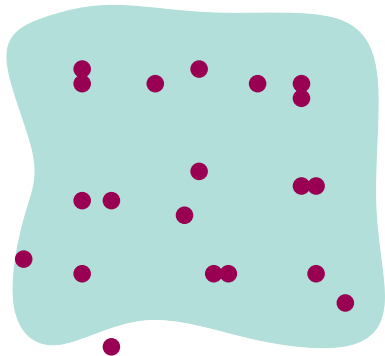


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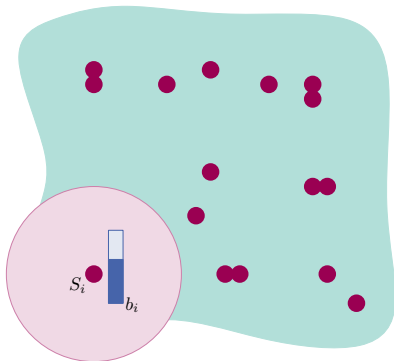
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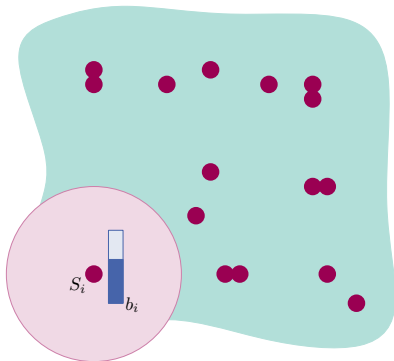
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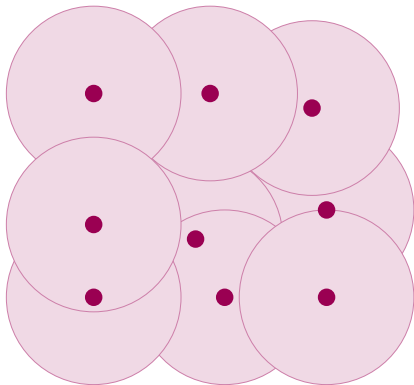
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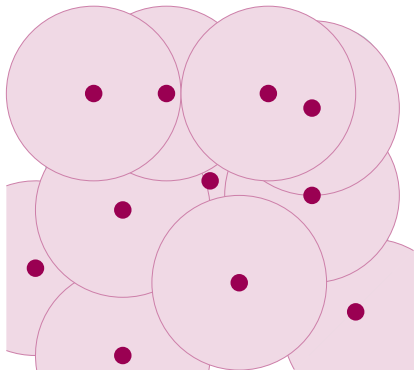
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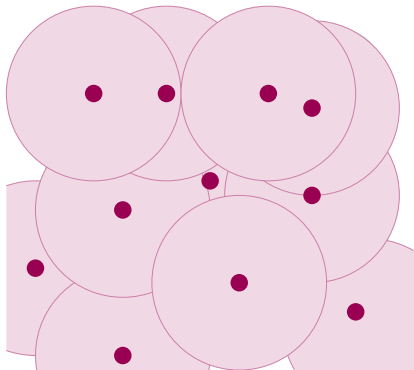


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### Extensive Previous Work

- [Cardei, Wu 05] area monitoring equals target monitoring
- [Slijepcevic, P. 05] target monitoring, uniform energy, disjoint sets
- [Cardei, Wu 06] non-disjoint sets, superlinear approximation algorithm
- [Berman et al. 06] general problem, log approximation in superlinear time
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- pseudo-linear time dual approximation scheme
- proof of NP-completeness (see paper)

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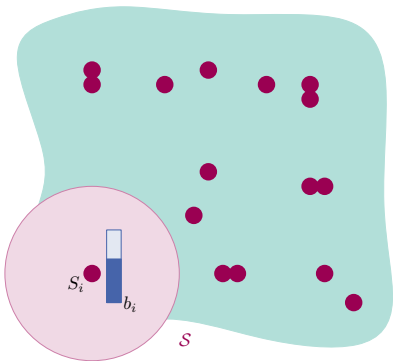
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# Model and Problem Definition

## Sensor Network Model

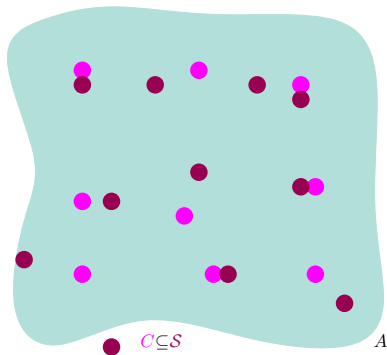
- sensor network  $\mathcal{S} = \{S_1, \dots, S_n\}$   
with  $S_i = (x_i, y_i, b_i)$ 
  - $(x_i, y_i)$ : coordinates
  - $b_i$ : battery capacity
- $C \subseteq \mathcal{S}$  is a cover of area  $A$ ,
  - if the union of disks centered at each  $S \in C$  contains area  $A$
  - disk radii equal sensing range  $R$
- $\mathcal{C}$  set of all possible covers of  $A$



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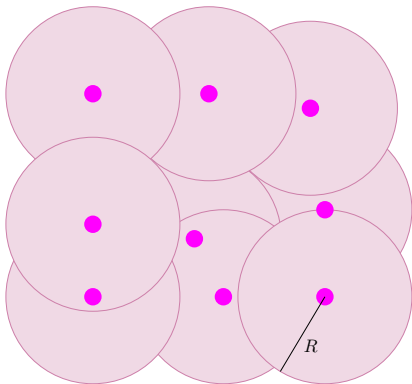
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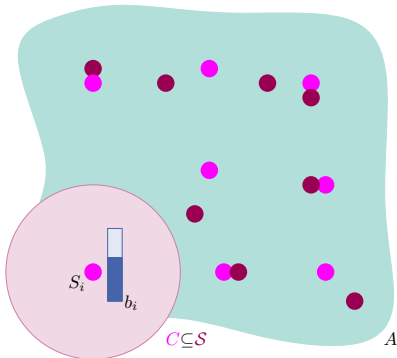




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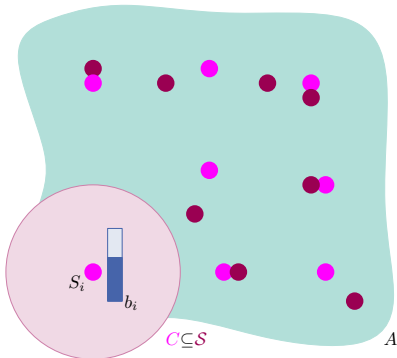
- find **schedule**  $(\underline{C}, \underline{t})$ ,
  - covers  $\underline{C} = \{C_1, \dots, C_m\} \subseteq \mathcal{C}$
  - durations  $\underline{t} = \{t_1, \dots, t_m\}$
- maximizing **lifetime**  $T = \sum_{j=1}^m t_j$ 
  - s.t.  $\sum_{i: S_j \in C_i} t_i \leq b_j \forall S_j \in \mathcal{S}$  (1)
  - i.e. node  $S_j$  cannot consume more than  $b_j$  units of energy
- **problem instance**  $(\mathcal{S}, \mathcal{A}, \mathcal{R})$ 
  - **solution** is any schedule  $(\underline{C}, \underline{t})$
  - solution  $(\underline{C}, \underline{t})$  **feasible** if (1) holds
  - lifetime of a solution  $T(\mathcal{S}, \mathcal{A}, \mathcal{R})$



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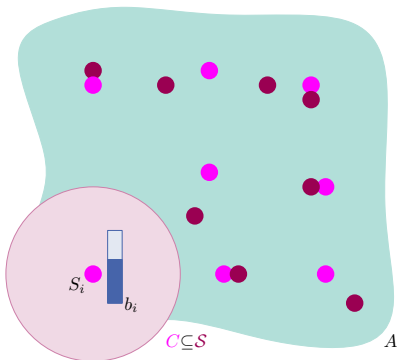
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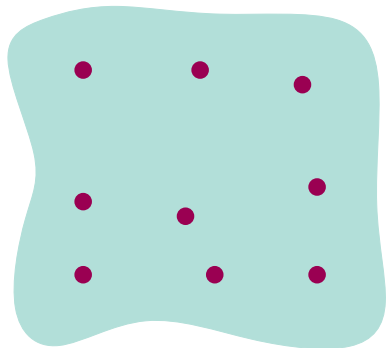
### Sensor Network Lifetime Problem (SNLP)

- **problem instance**  $(S, A, 1)$ 
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# Model and Problem Definition

## Further Considerations

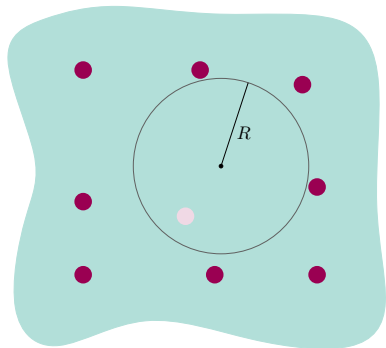
- problem reinterpretation
  - area monitoring with min. guaranteed resolution
  - resolution
    - $\hat{=}$  max. distance of any point in the area to one sensor
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- central algorithm sufficient
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  - providing upper bounds for distributed algorithms



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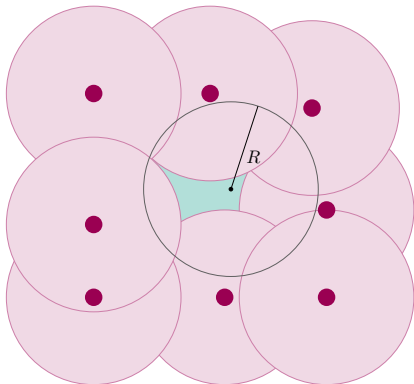
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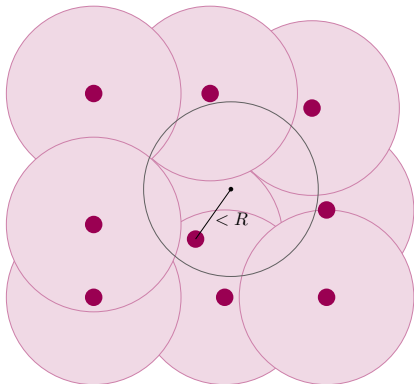
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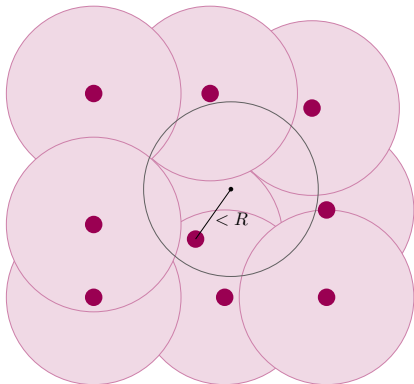




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- combination of two approximation techniques
  - consider simpler problem instances
  - assume  $f$ -approximate algorithm  $\mathcal{A}$  exists to solve these instances
  - transform problem instances so that  $\mathcal{A}$  can solve them
  - solutions are feasible for the original instance and near-optimal
- discretizing positions
  - nodes restricted to points on a grid
- area partitioning
  - area is divided into squared areas
  - subproblems restrained to these squares are solved and combined

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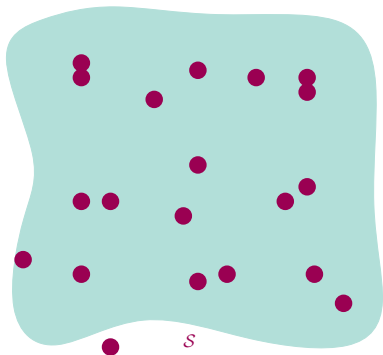
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### Algorithm 1

*in:* instance  $(\mathcal{S}, 1)$ , algorithm  $\mathcal{A}$ ,  
parameter  $\delta \in [0, 1]$

*out:* schedule  $(\underline{C}, \underline{t})$

- Define grid of width  $\delta/2$ .
- Move every node in  $\mathcal{S}$  to the closest point on the grid  $\rightarrow \tilde{\mathcal{S}}$ .
- Solve  $(\tilde{\mathcal{S}}, 1 + \delta/2) \rightarrow (\underline{C}, \underline{t})$ .  
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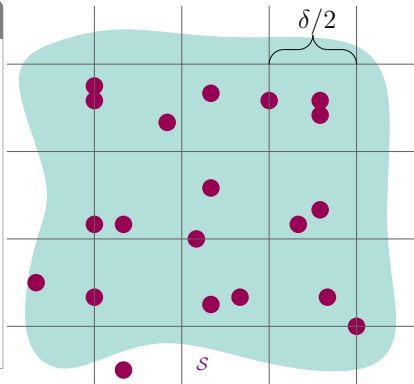
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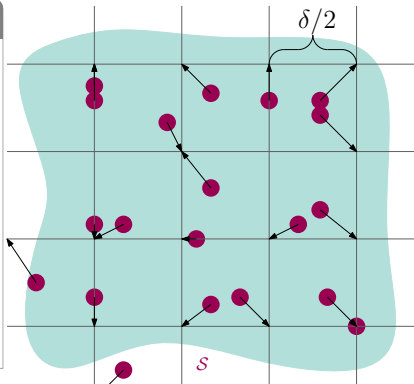
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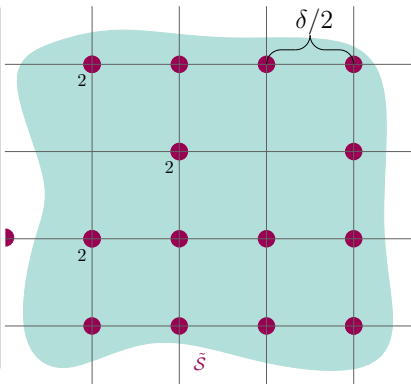
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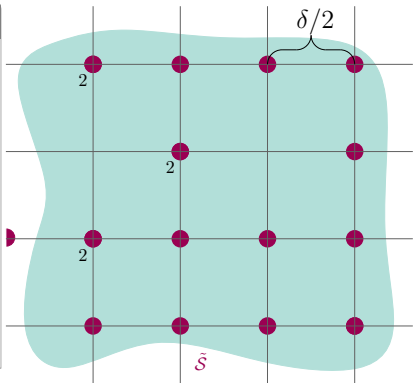
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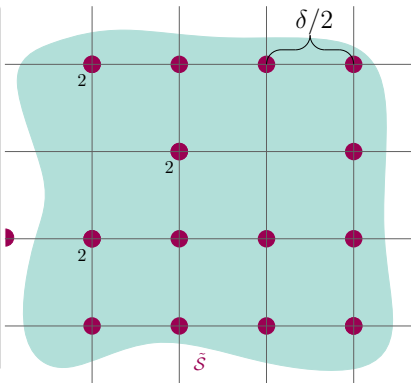
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Let  $\delta \in [0, 1]$ . *Algorithm 1* yields a feasible solution to  $(\mathcal{S}, 1 + \delta)$  with lifetime  $T\langle \mathcal{S}, 1 + \delta \rangle \geq f \cdot T_{\text{opt}}\langle \mathcal{S}, 1 \rangle$ .

#### *Correctness*

- solution to  $(\mathcal{S}, 1)$  is solution to  $(\tilde{\mathcal{S}}, 1 + \frac{\delta}{2})$   
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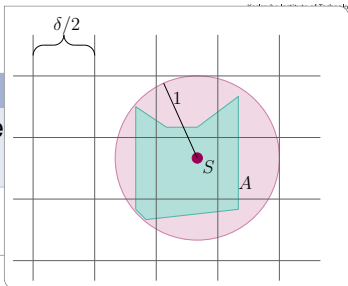
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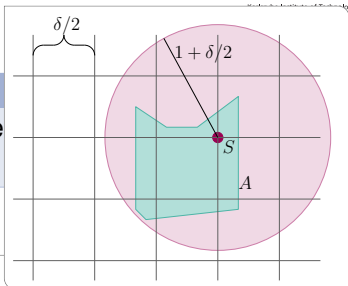
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- solution to  $(\tilde{S}, 1 + \frac{\delta}{2})$  is solution to  $(S, 1 + \delta)$   
 $\rightarrow T\langle S, 1 + \delta \rangle \geq T\langle \tilde{S}, 1 + \frac{\delta}{2} \rangle \geq f \cdot T_{\text{opt}}\langle S, 1 \rangle$

### Lemma 1

Let  $\delta \in [0, 1]$ . *Algorithm 1* yields a feasible solution to  $(\mathcal{S}, 1 + \delta)$  with lifetime  $T\langle \mathcal{S}, 1 + \delta \rangle \geq f \cdot T_{\text{opt}}\langle \mathcal{S}, 1 \rangle$ .

### Correctness

- solution to  $(\mathcal{S}, 1)$  is solution to  $(\tilde{\mathcal{S}}, 1 + \frac{\delta}{2})$   
 $\rightarrow T_{\text{opt}}\langle \tilde{\mathcal{S}}, 1 + \frac{\delta}{2} \rangle \geq T_{\text{opt}}\langle \mathcal{S}, 1 \rangle$
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 $\rightarrow T\langle \tilde{\mathcal{S}}, 1 + \frac{\delta}{2} \rangle \geq f \cdot T_{\text{opt}}\langle \tilde{\mathcal{S}}, 1 + \frac{\delta}{2} \rangle \geq f \cdot T_{\text{opt}}\langle \mathcal{S}, 1 \rangle$
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# Area Partitioning

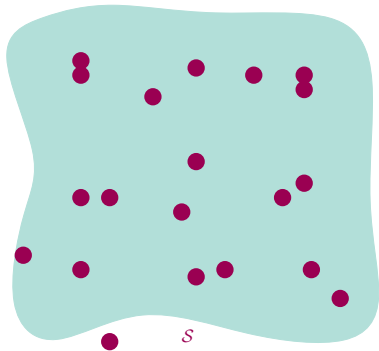
## Procedure

### Algorithm 2

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*out:* schedule  $(\underline{C}, \underline{t})$

- Define  $k$  partitions  $\mathcal{T}^i$ ,  
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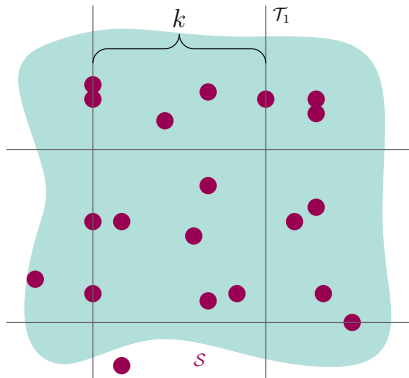
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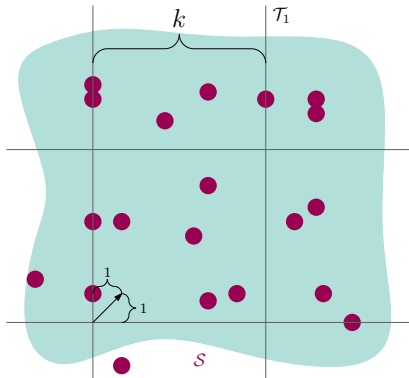
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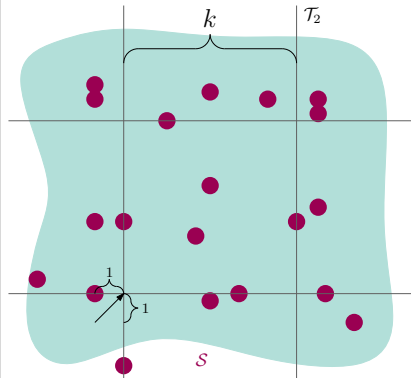
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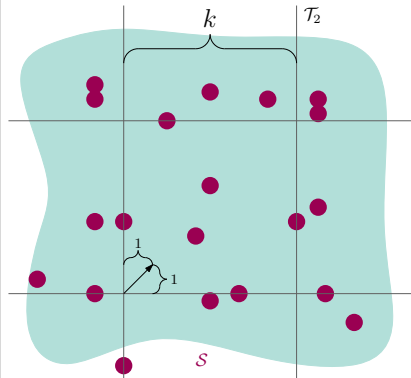
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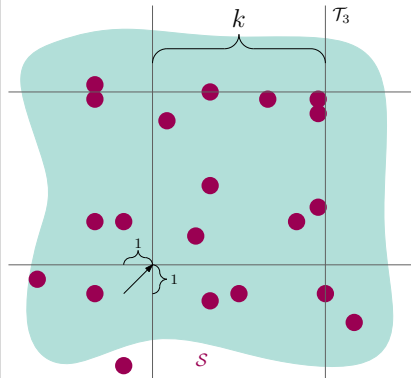
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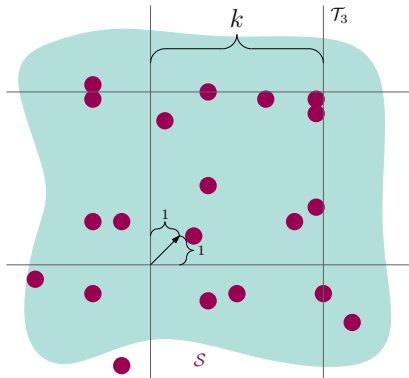
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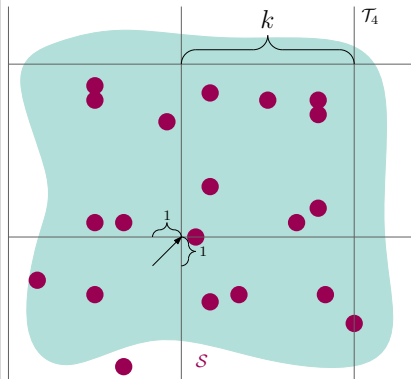
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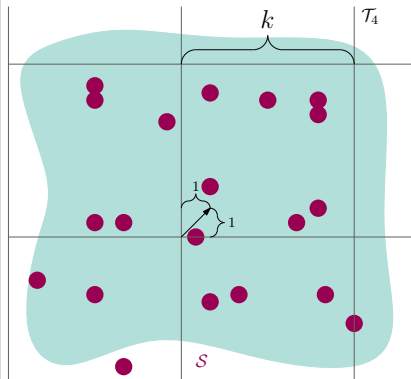
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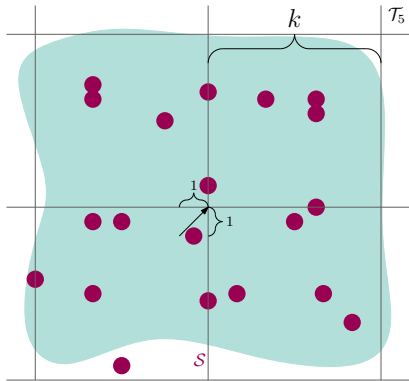
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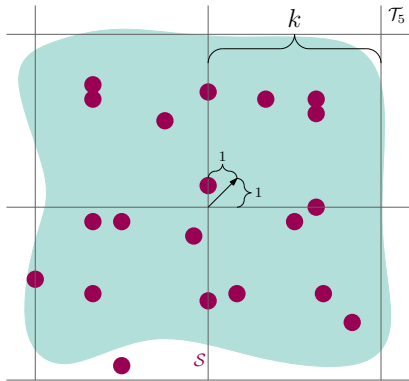
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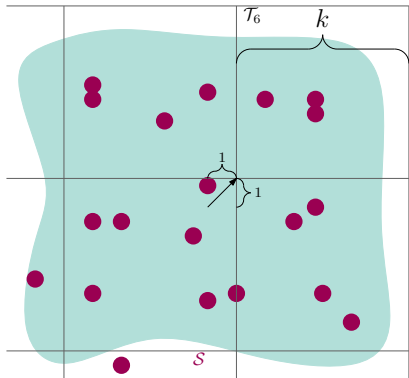
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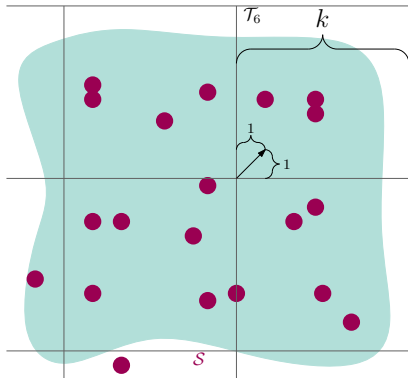
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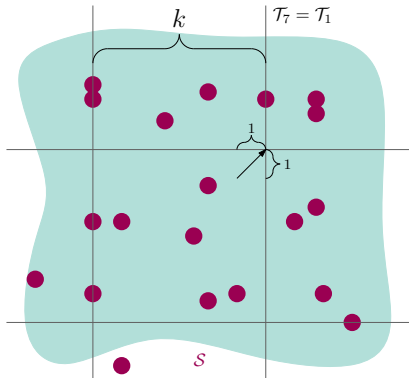
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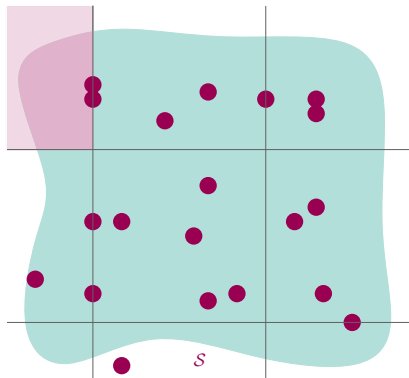
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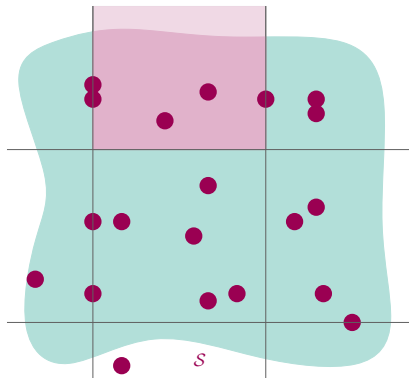
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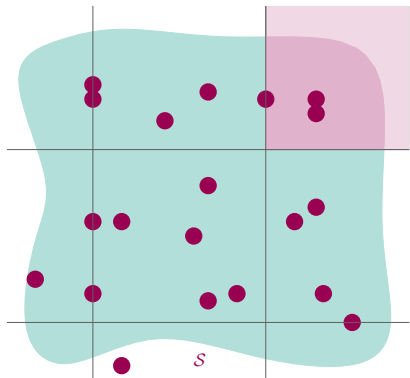
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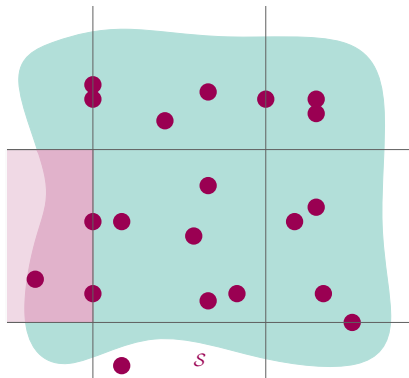
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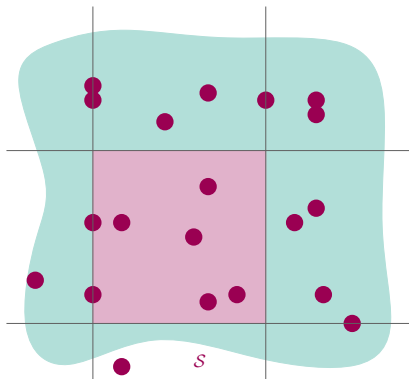
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# Area Partitioning

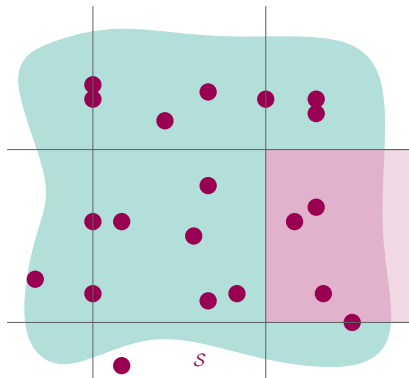
## Procedure

### Algorithm 2

*in:* instance  $(S, 1)$ , algorithm  $\mathcal{A}$ ,  
parameter  $\epsilon \in (0, 1]$

*out:* schedule  $(\underline{C}, \underline{t})$

- Define  $k$  partitions  $\mathcal{T}^i$ ,  
 $k = \lceil \frac{10}{\epsilon} \rceil, i \in \mathbb{Z}_k = \{1, \dots, k\}$ .
- For each partition  $\mathcal{T}^i$ ,
  - solve  $(S, 1)$  restricted to each  
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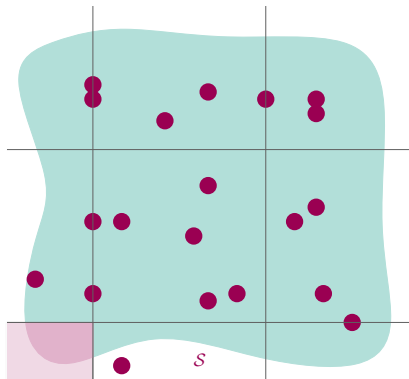
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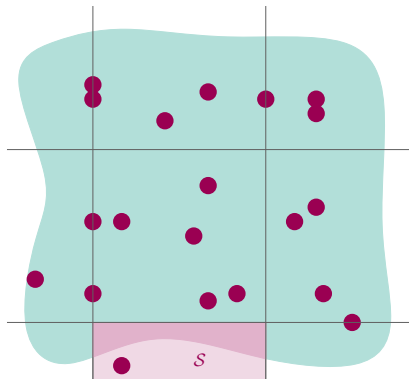
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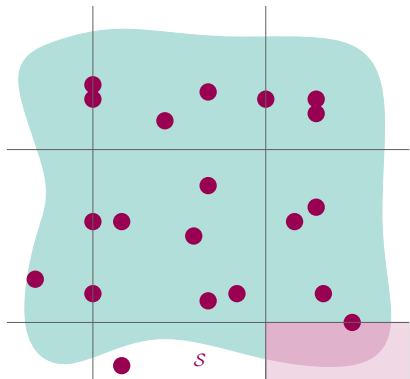
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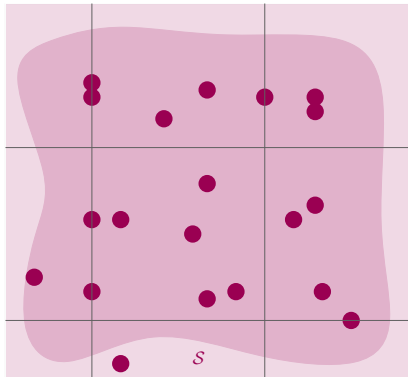
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# Area Partitioning

## Procedure

### Algorithm 2

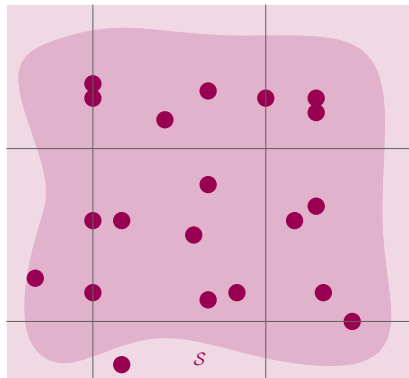
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$$(\underline{C}, \underline{t}) = \left( \bigcup_{i \in \mathbb{Z}_k} \underline{C}^i, \frac{(1-\epsilon)}{k} \cdot \bigcup_{i \in \mathbb{Z}_k} \underline{t}^i \right).$$



### Lemma 2

Let  $\epsilon \in (0, 1]$ . *Algorithm 2* yields a feasible solution to  $(S, 1)$  with lifetime  $T\langle S, 1 \rangle \geq f \cdot (1 - \epsilon) \cdot T_{\text{opt}}\langle S, 1 \rangle$ .

#### Lifetime

$$\begin{aligned} \blacksquare T\langle S, 1 \rangle &= \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} T\langle S, 1 \rangle^i \geq \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} f \cdot T_{\text{opt}}\langle S, 1 \rangle \\ &= f \cdot (1 - \epsilon) \cdot T_{\text{opt}}\langle S, 1 \rangle \end{aligned}$$

#### Feasibility

$$\blacksquare \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} \sum_{j: S_n \in C_j^i} t_j^i \leq \frac{1-\epsilon}{k} \left( (k-2) \cdot 1 + 2 \cdot 4 \right) \cdot b_n \leq b_n \quad \forall S_n \in S.$$

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$$T\langle S, 1 \rangle^i \triangleq \sum_{j=1}^{|\mathcal{C}^i|} t_j^i$$

#### Feasibility

$$\blacksquare \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} \sum_{j: S_n \in \mathcal{C}_j^i} t_j^i \leq \frac{1-\epsilon}{k} \left( (k-2) \cdot 1 + 2 \cdot 4 \right) \cdot b_n \leq b_n \quad \forall S_n \in S.$$

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$$T\langle S, 1 \rangle^i \geq f \cdot T_{\text{opt}}\langle S, 1 \rangle$$

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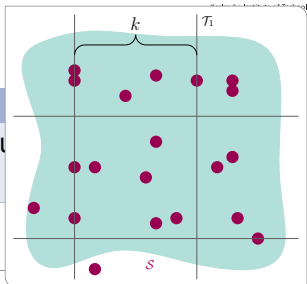


# Area Partitioning

## Proofs

### Lemma 2

Let  $\epsilon \in (0, 1]$ . Algorithm 2 yields a feasible solution  $T\langle S, 1 \rangle \geq f \cdot (1 - \epsilon) \cdot T_{\text{opt}}\langle S, 1 \rangle$ .



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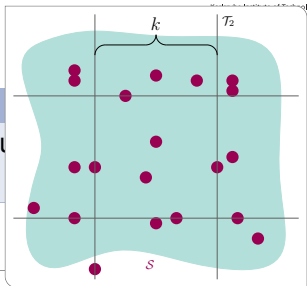
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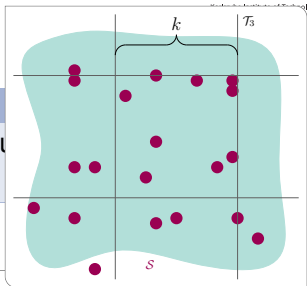
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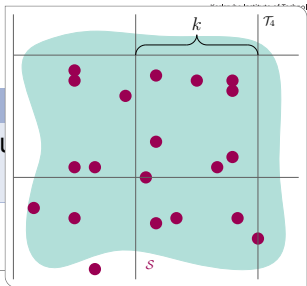
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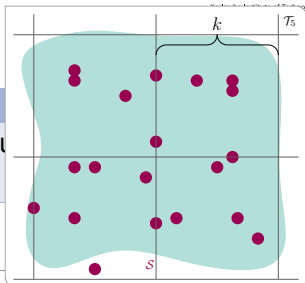
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# Area Partitioning

## Proofs

### Lemma 2

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### Lifetime

$$\begin{aligned} \blacksquare T(\mathcal{C}, 1) &= \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} T(\mathcal{S}, 1)^i \geq \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} f \cdot T_{\text{opt}}(\mathcal{S}, 1) \\ &= f \cdot (1 - \epsilon) \cdot T_{\text{opt}}(\mathcal{S}, 1) \end{aligned}$$

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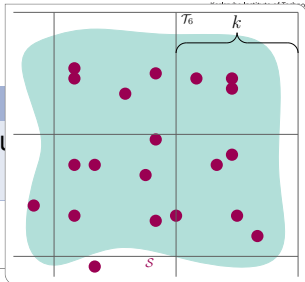
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# Area Partitioning

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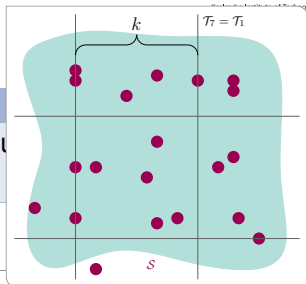
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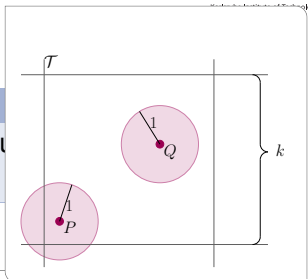
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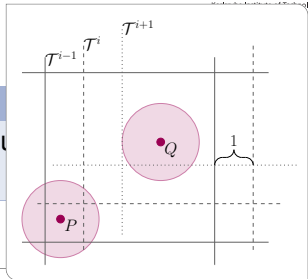


# Area Partitioning

## Proofs

### Lemma 2

Let  $\epsilon \in (0, 1]$ . Algorithm 2 yields a feasible solution  $T\langle S, 1 \rangle \geq f \cdot (1 - \epsilon) \cdot T_{\text{opt}}\langle S, 1 \rangle$ .



### Lifetime

$$\begin{aligned} \blacksquare T\langle S, 1 \rangle &= \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} T\langle S, 1 \rangle^i \geq \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} f \cdot T_{\text{opt}}\langle S, 1 \rangle \\ &= f \cdot (1 - \epsilon) \cdot T_{\text{opt}}\langle S, 1 \rangle \end{aligned}$$

### Feasibility

$$\blacksquare \frac{1-\epsilon}{k} \sum_{i \in \mathbb{Z}_k} \sum_{j: S_n \in C_j^i} t_j^i \leq \frac{1-\epsilon}{k} \left( (k-2) \cdot 1 + 2 \cdot 4 \right) \cdot b_n \leq b_n \quad \forall S_n \in S.$$

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# Complete Approximation Algorithm

## Procedure

### Complete Algorithm:

*input:* instance  $(\mathcal{S}, 1)$ , algorithm  $\mathcal{A}$ , parameters  $\delta \in [0, 1], \epsilon \in (0, 1]$

*output:* schedule  $(\underline{\mathcal{C}}, \underline{t})$

- Define grid of width  $\delta/2$ .
- Move every node in  $\mathcal{S}$  to the closest point on the grid  $\rightarrow \tilde{\mathcal{S}}$ .
- Define  $k$  partitions  $\mathcal{T}^i$ ,  
 $k = \lceil \frac{10}{\epsilon} \rceil, i \in \mathbb{Z}_k = \{1, \dots, k\}$ .
- For each partition  $\mathcal{T}^i$ ,
  - solve  $(\tilde{\mathcal{S}}, 1 + \frac{\delta}{2})$  for each square of  $\mathcal{T}^i$ , (using algorithm  $\mathcal{A}$ )
  - combine solutions  $\rightarrow (\underline{\mathcal{C}}^i, \underline{t}^i)$ .
- Return  $(\underline{\mathcal{C}}, \underline{t}) = (\bigcup_{i \in \mathbb{Z}_k} \underline{\mathcal{C}}^i, \frac{(1-\epsilon)}{k} \cdot \bigcup_{i \in \mathbb{Z}_k} \underline{t}^i)$ .

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## Algorithm 2

# Complete Approximation Algorithm

## Procedure

### Complete Algorithm:

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- For each partition  $\mathcal{T}^i$ ,
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### Algorithm 1

# Complete Approximation Algorithm

## Theorem

### Theorem 1

Let  $\delta \in [0, 1]$  and  $k = \lceil 10/\epsilon \rceil$  with  $\epsilon \in (0, 1]$ . *Complete Algorithm* yields a feasible solution  $(\underline{C}, \underline{t})$  to  $(\mathcal{S}, 1 + \delta)$  with lifetime

$$T\langle \mathcal{S}, 1 + \delta \rangle \geq (1 - \epsilon) \cdot f \cdot T_{\text{opt}}\langle \mathcal{S}, 1 \rangle.$$

Its runtime complexity is bounded by

$$O\left(|\mathcal{S}| + \epsilon |\mathcal{S}| \cdot g_{\mathcal{A}}(O(1/\delta^2 \epsilon^2))\right) = O(|\mathcal{S}|)$$

with  $g_{\mathcal{A}}(|\mathcal{S}|)$  the runtime of algorithm  $\mathcal{A}$ .

# Complete Approximation Algorithm

## Proofs - 1

### Theorem 1 - (part a)

Let  $\delta \in [0, 1]$  and  $k = \lceil 10/\epsilon \rceil$  with  $\epsilon \in (0, 1]$ . *Complete Algorithm* yields a feasible solution  $(\underline{C}, \underline{t})$  to  $(S, 1 + \delta)$  with lifetime

$$T\langle S, 1 + \delta \rangle \geq (1 - \epsilon) \cdot f \cdot T_{\text{opt}}\langle S, 1 \rangle.$$

### *Feasibility*

- follows directly from Lemma 1 & 2

### *Approximation Guarantee*

- node discretization  $\rightarrow T\langle S, 1 + \delta \rangle \geq f \cdot T_{\text{opt}}\langle S, 1 \rangle$  for all squares
- combining solutions to all tiles  $\rightarrow$  additional factor  $(1 - \epsilon)$

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### Theorem 1 - (part b)

The runtime complexity of *Complete Algorithm* is bounded by

$$\mathcal{O}\left(|\mathcal{S}| + 1/\epsilon|\mathcal{S}| \cdot g_{\mathcal{A}}(\mathcal{O}(1/\delta^2\epsilon^2))\right) = \mathcal{O}(|\mathcal{S}|)$$

with  $g_{\mathcal{A}}(|\mathcal{S}|)$  runtime of algorithm  $\mathcal{A}$  with respect to number of nodes.

### Runtime

- $|\mathcal{S}|$  : discretizing nodes
- $\mathcal{O}(1/\epsilon|\mathcal{S}|)$  : squares to be computed
- $g_{\mathcal{A}}(\mathcal{O}(1/\delta^2\epsilon^2))$ : runtime of algorithm  $\mathcal{A}$  for each square



# Complete Approximation Algorithm

## Proofs - 2

### Theorem 1 - (part b)

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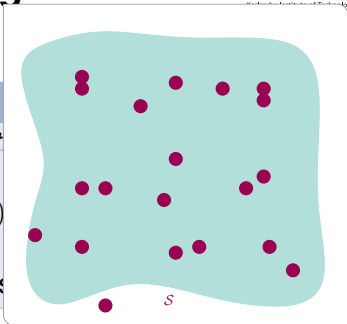
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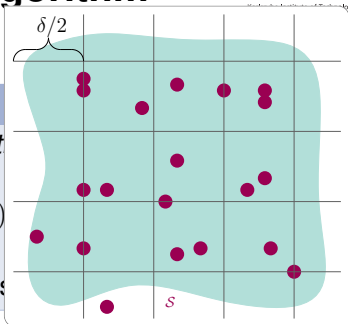
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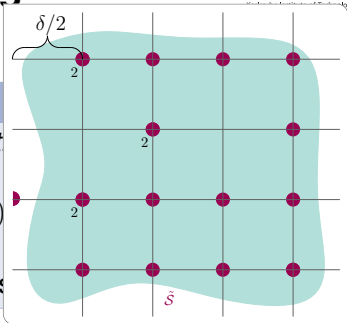
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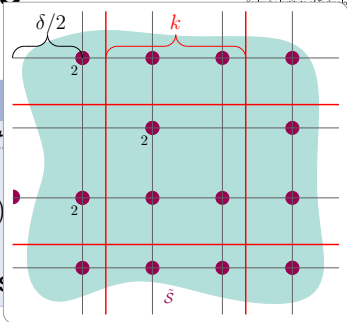
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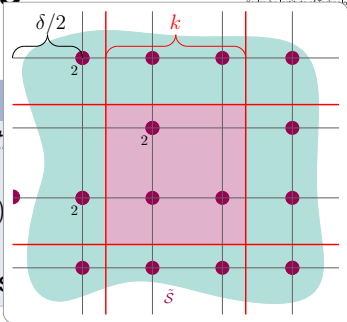
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## Contribution

- pseudo-linear time dual approximation scheme
  - $(1 - \epsilon)$  approximation, if sensing ranges are allowed to grow by  $\delta$
  - runtime dependent on  $\delta$ ,  $\epsilon$ , number of nodes
- proof of NP-completeness
  - respecting the geometric structure of the problem

## Future Work

- enhance model (non-uniform sensing ranges, obstacles, ...)  
→ extension to low-dimensional metrics
- implementation using an exact solver as algorithm  $\mathcal{A}$
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# Thank you for your attention!



## time for questions