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# Approximation Algorithms

## Chapter 6: Review and Relay Placement

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Sándor P. Fekete

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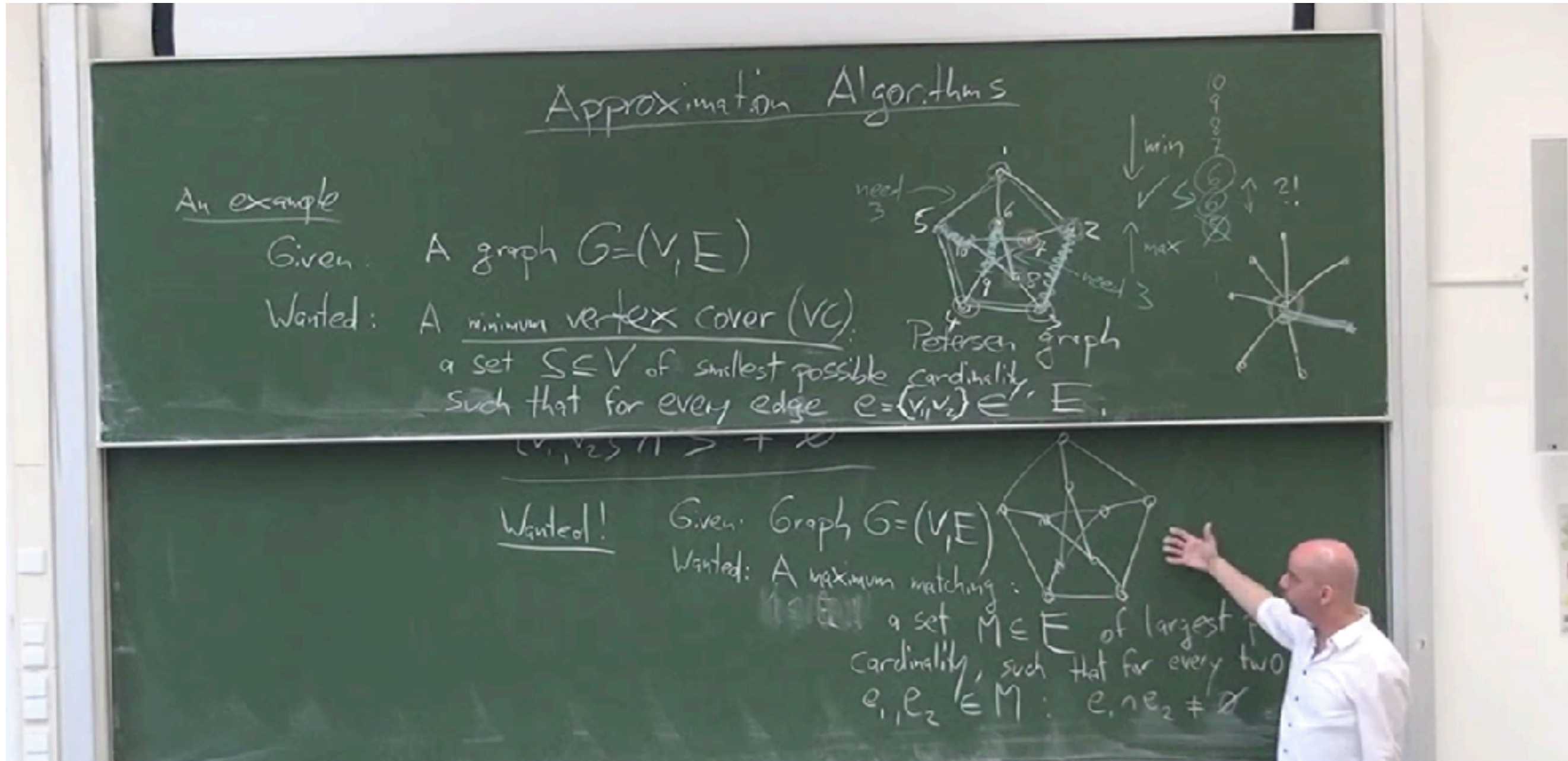


- 1. Introduction**
- 2. Review**
- 3. Extra Packing: Dispersion**
- 4. Extra Tours: Lawn Mowing**
- 5. Relay Placement**
- 6. Coordinated Motion Planning**

1. Introduction
2. Review
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# Review: Matching & Vertex Cover

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# Review: Approximation Algorithms

# Review: Approximation Algorithms

Given: An NP-complete (or NP-hard) problem  $\Pi$

Wanted: An approximation algorithm for  $\Pi$ :

(1) Runtime is polynomial.

(2) There is a constant  $c$ , such that for any instance  $I$  of  $\Pi$ , the algorithm computes a solution of value within  $c$  of the optimum value.

For minimization:  $\text{APPROX}(I) \leq c \cdot \text{OPT}(I) \rightarrow c \geq 1$

For maximization:  $\text{APPROX}(I) \geq c \cdot \text{OPT}(I) \rightarrow c \leq 1$

To check:  
(0)  
(1)  
(2)



Veget a VC!  
factor

Wanted: An AA for VC!

Idea: Consider a maximal matching. Pick both vertices for each edge.

# Review: Set Cover



# Review: Set Cover

Set cover

Idea: Consider the relative cost (cost per element) for the currently uncovered elements

4Z ← cost per element

Approximation algorithm for SETCOVER: Greedy

Algorithm 2.2 (Greedy set cover)

1.  $C \leftarrow \emptyset$  //  $C$ : covered elements
2. WHILE ( $C \neq U$ ) DO
  - Find the most cost-efficient set in the current iteration, say  $S$
  - Let  $\alpha = \frac{\text{cost}(S)}{|S-C|}$ , i.e., the cost-efficiency of  $S$ .

• Pick  $S$ , and for each  $e \in S$ , set price

# Review: Maximum Coverage

# Review: Maximum Coverage

Now: GREEDY is never worse than an  $H_n$ -approximation.

Important step:

Lemma 2.3

For each  $k \in \{1, \dots, n\}$  in the sequence  $e_1, \dots, e_n$  in which elements are covered by GREEDY, we have  $\text{price}(e_k) \leq \frac{\text{OPT}}{n-k+1}$

THEOREM 2.4

The GREEDY algorithm is an  $H_n$ -approximation algorithm for SET COVER.

Proof:

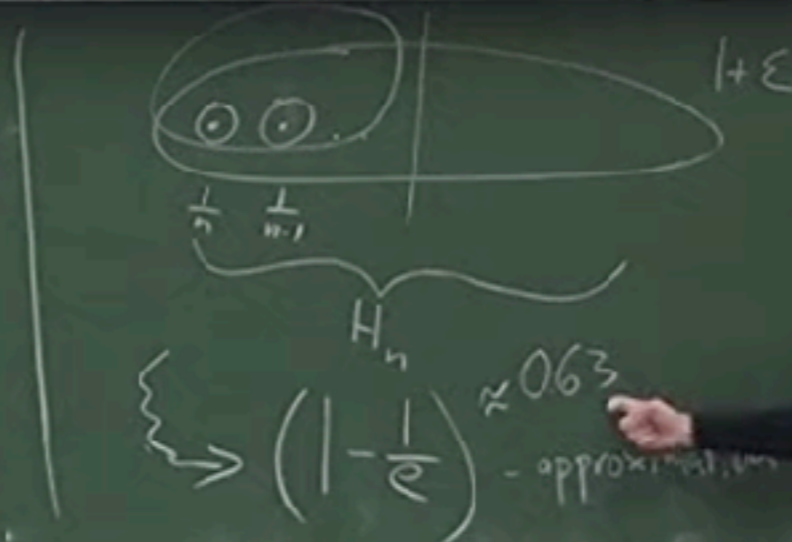
$$\begin{aligned} \text{The total cost for GREEDY is } & \sum_{k=1}^n \text{cost}(e_k) \\ & \leq \sum_{k=1}^n \frac{\text{OPT}}{n-k+1} = \text{OPT} \cdot \sum_{k=1}^n \frac{1}{k} = \text{OPT} \cdot H_n \end{aligned}$$

Different but related:

PROBLEM 2.4 $\frac{1}{2}$  (Maximum Coverage)

Int: Same as for SET COVER, plus a budget (e.g. a number  $k$  that tells us how many sets we can pick)

Goal: Cover as many elements as possible



# Review: Shortest Superstring

# Review: Shortest Superstring

LEMMA 2.11  $OPT \leq OPT_{\phi} \leq 2OPT$

Consider an optimal solution:

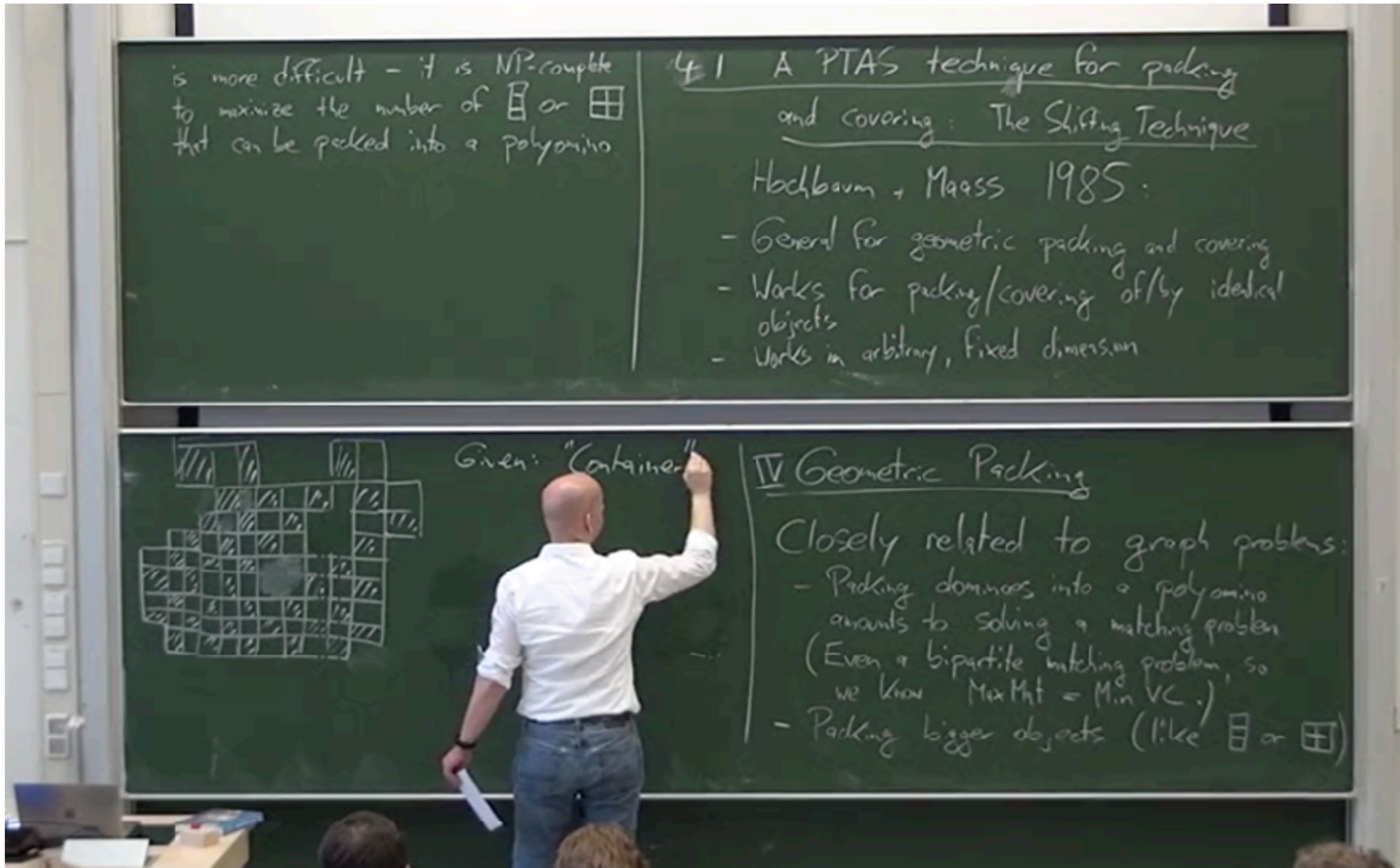
Contained Substrings

Sketch:  
Consider "blocks" of overlapping substrings, shown in color on the left. Each block corresponds to a  $G_{ij}$ , so if we can argue that no three blocks overlap in  $OPT$ , we are done. Assuming that we have an overlap between red, yellow, and blue, we must have

an overlap between the last red and the first blue substring, so we must have an overlap between the first yellow and the first blue (because the first yellow is right of the last red). That is a contradiction to the assumption that the blue substring is not part of the yellow block.

# Review: Packing

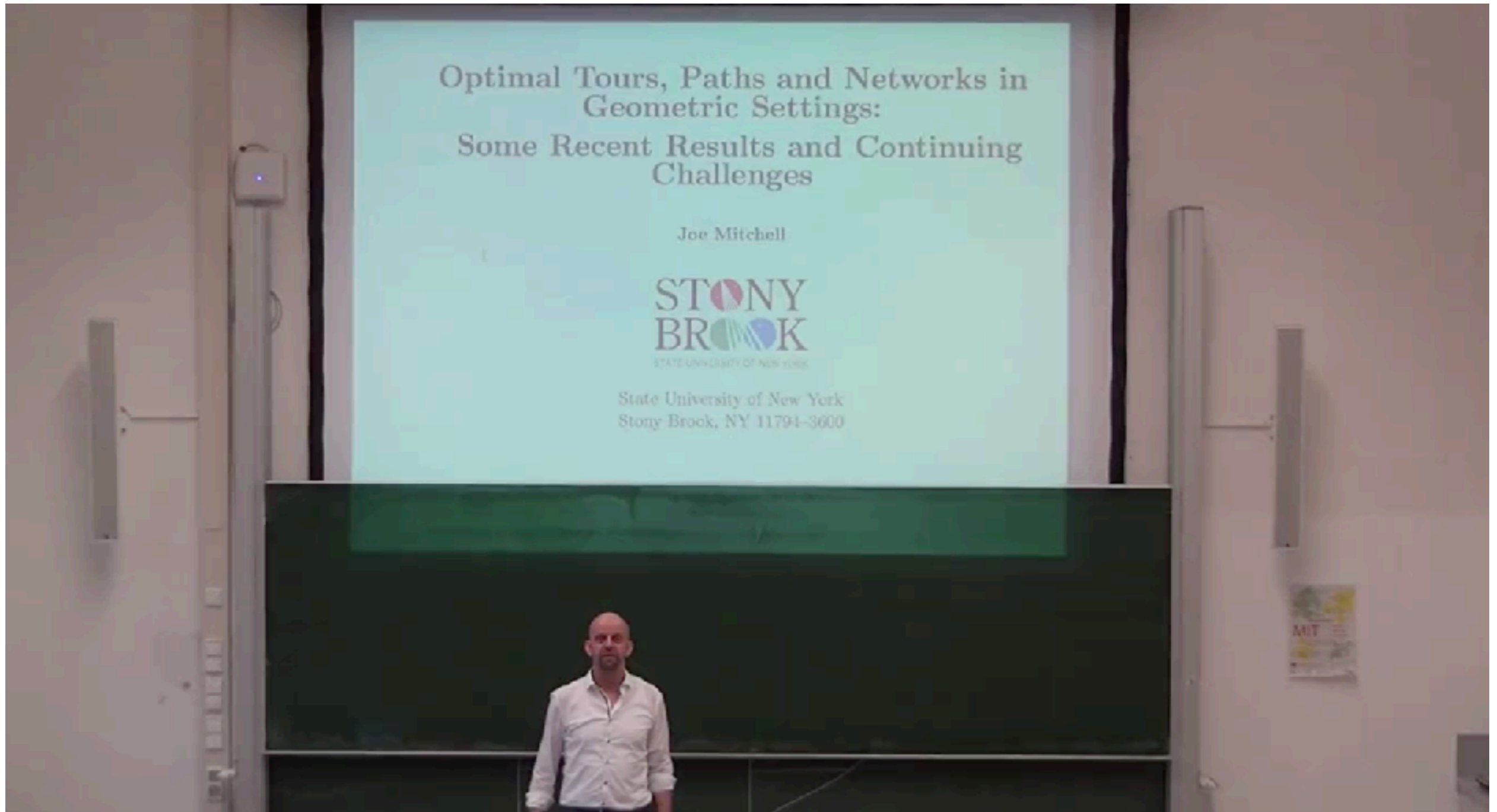
# Review: Packing



# Review: Tour Problems

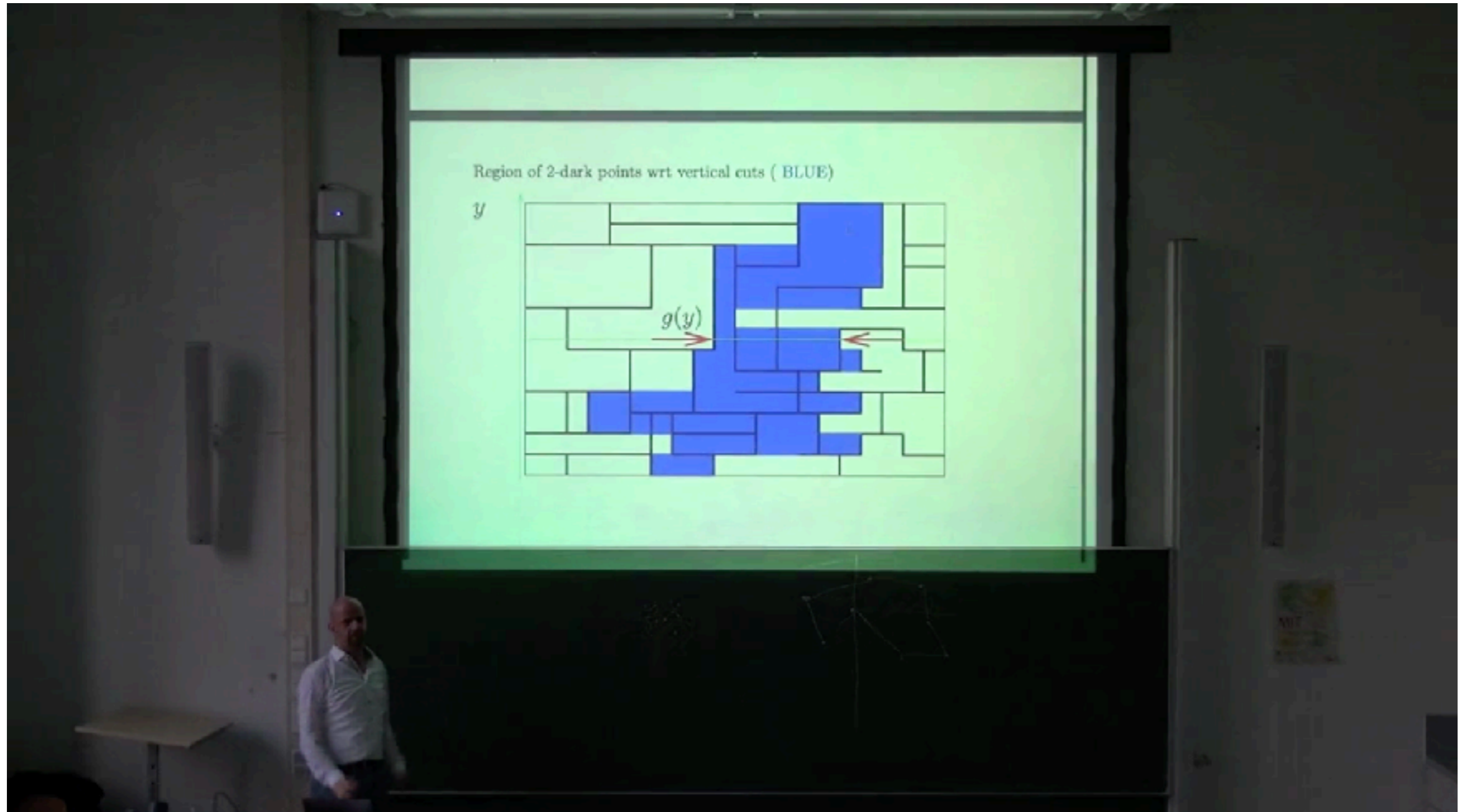


# Review: Tour Problems



# Review: Tour Problems

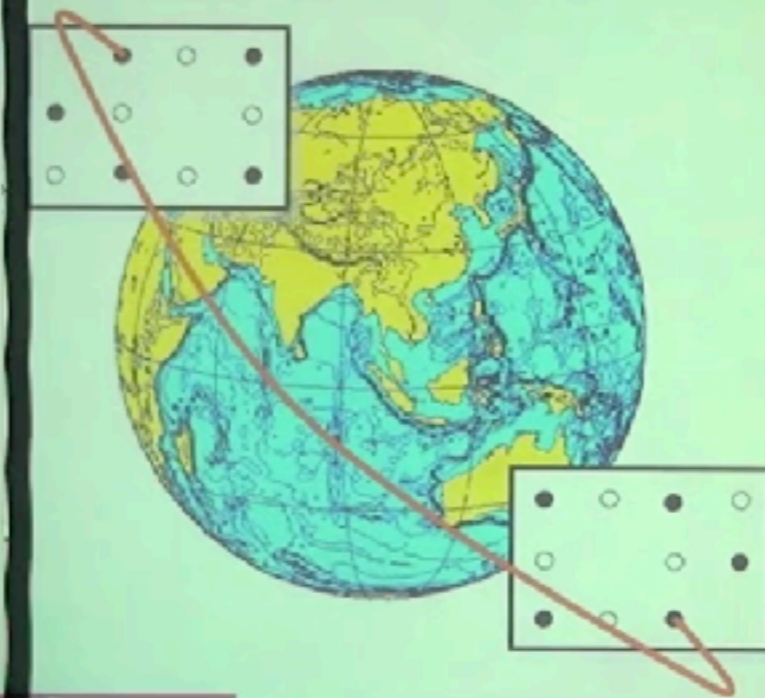
# Review: Tour Problems



# Review: Tour Problems

# Review: Tour Problems

**Think Globally**



- Consider a grid graph embedded on a sphere.
- Add second copy at the antipodes.
- Longest connections are between antipodal points.

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17.12.2019 | Sándor P. Fekete | Combinatorial Optimization Meets Computational Geometry

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$$\begin{array}{l} N_{++} \\ N_{+-} \\ N_{-+} \\ N_{--} \end{array}$$
$$\begin{array}{l} N_{++} + N_{--} = N_{+-} + N_{-+} \\ N_{++} + N_{+-} = N_{-+} + N_{--} \\ \hline 2N_{++} + (N_{+-} + N_{-+}) = (N_{+-} + N_{-+}) + 2N_{--} \end{array}$$

# The Freeze-Tag Problem: How to Wake Up a Swarm of Robots

Estie Arkin<sup>1</sup>

Michael Bender<sup>1</sup>

**Sándor Fekete**<sup>2</sup>

Joe Mitchell<sup>1</sup>

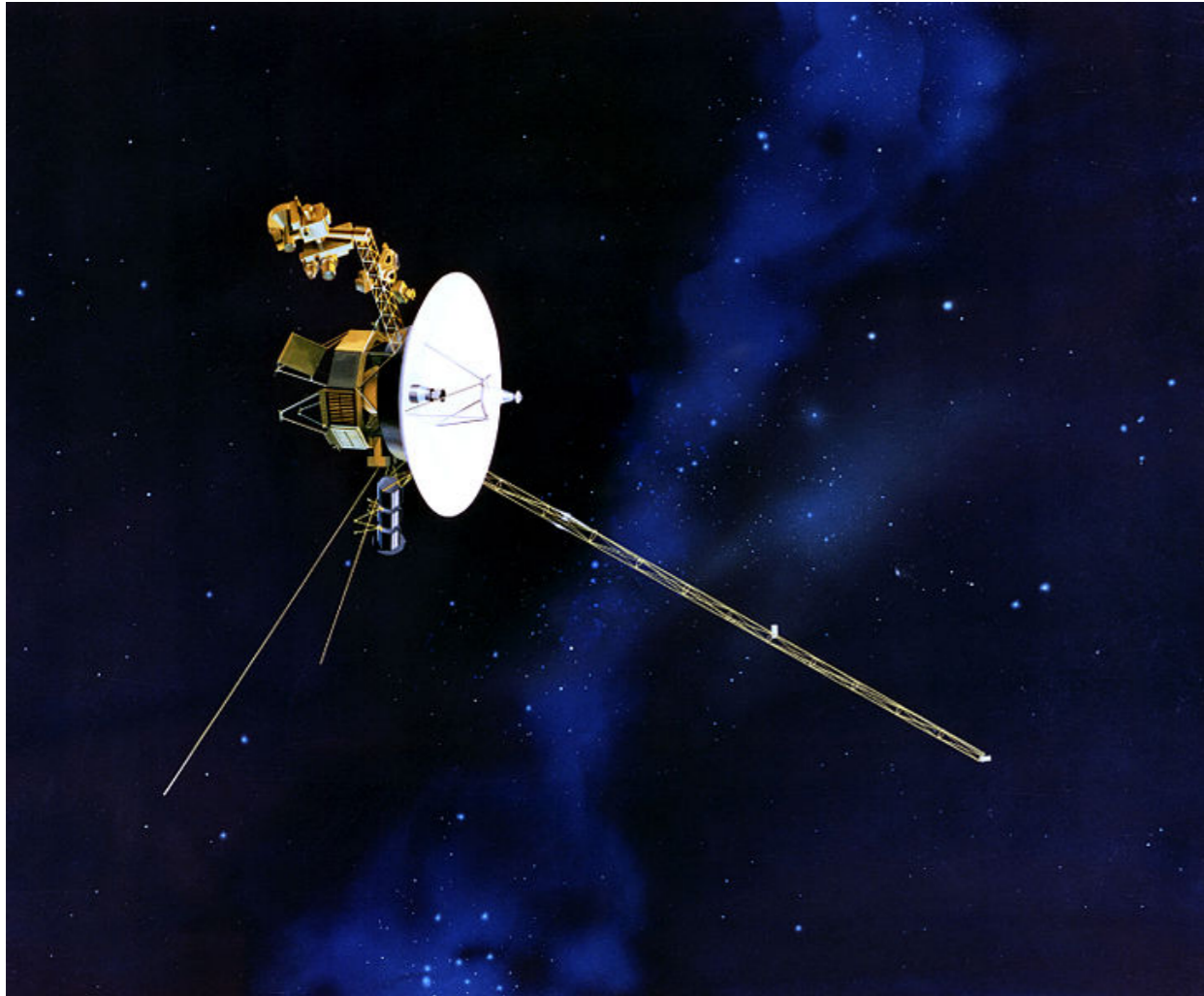
Martin Skutella<sup>3</sup>

<sup>1</sup> University at Stony Brook

<sup>2</sup> TU Braunschweig

<sup>3</sup> TU Berlin

# Motivation

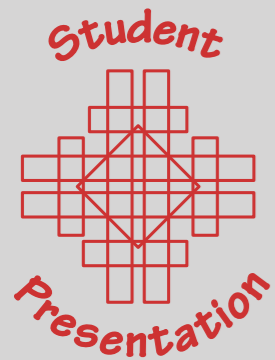


- Highly focused antennas.
- Expensive rotations.

**How can we quickly distribute information from one satellite to all others?**



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Universität  
Braunschweig



# Minimum Scan Cover with Angular Transition Costs



**Sándor Fekete**



**Linda Kleist**



**Dominik Krupke**



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# Dispersion (1)

Algorithmica (2001) 30: 451–470  
DOI: 10.1007/s00453-001-0022-x

Algorithmica

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## Approximation of Geometric Dispersion Problems<sup>1</sup>

C. Baur<sup>2</sup> and S. P. Fekete<sup>3</sup>

**Abstract.** We consider problems of distributing a number of points within a polygonal region  $P$ , such that the points are “far away” from each other. Problems of this type have been considered before for the case where the possible locations form a discrete set. Dispersion problems are closely related to packing problems. While Hochbaum and Maass [20] have given a polynomial-time approximation scheme for packing, we show that geometric dispersion problems cannot be approximated arbitrarily well in polynomial time, unless  $P = NP$ . A special case of this observation solves an open problem by Rosenkrantz et al. [31]. We give a  $\frac{2}{3}$  approximation algorithm for one version of the geometric dispersion problem. This algorithm is strongly polynomial in the size of the input, i.e., its running time does not depend on the area of  $P$ . We also discuss extensions and open problems.

**Key Words.** Packing, Dispersion, Location problems, Geometric optimization, Bounds on approximation factors.

# Dispersion (2)

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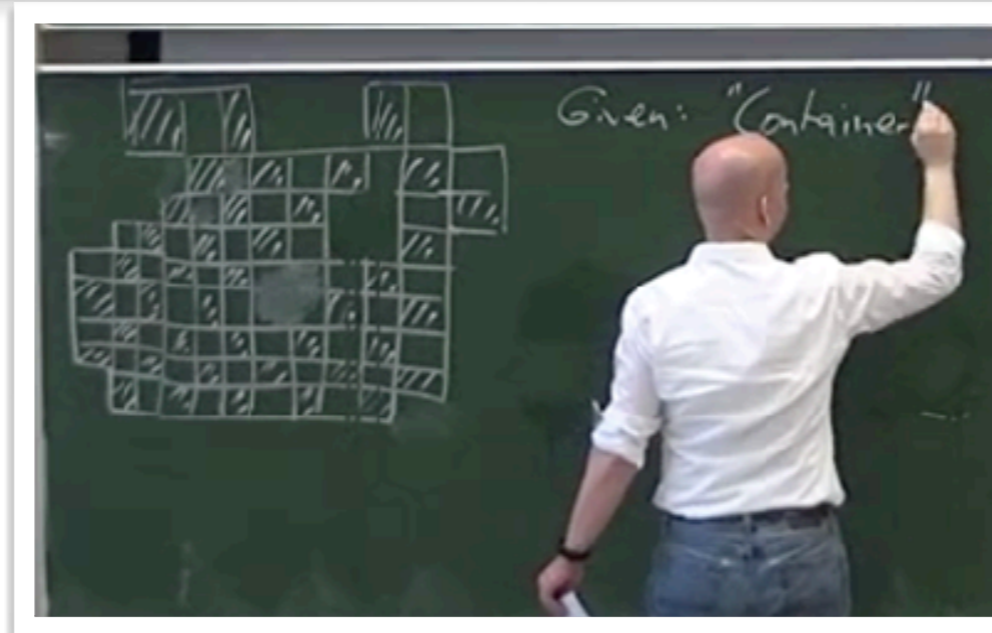


# Dispersion (2)

$\text{PACK}(k, L)$

*Input:* A polygonal region  $P$  with  $n$  vertices, a parameter  $k$ , a parameter  $L$ .

*Question:* Can  $k$  many  $L$ -squares be packed into  $P$ ?



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*Input:* A polygonal region  $P$  with  $n$  vertices.

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$\max_L \text{PACK}(k)$

*Input:* A polygonal region  $P$  with  $n$  vertices.

*Task:* Pack  $k$  many  $L \times L$  squares into  $P$ , such that  $L$  is as big as possible.





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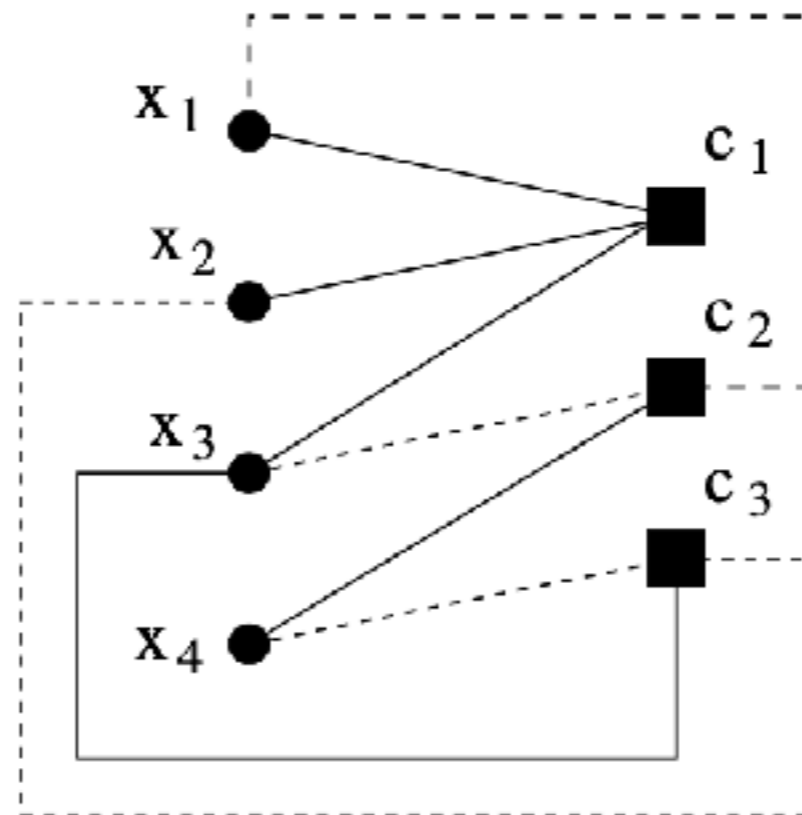
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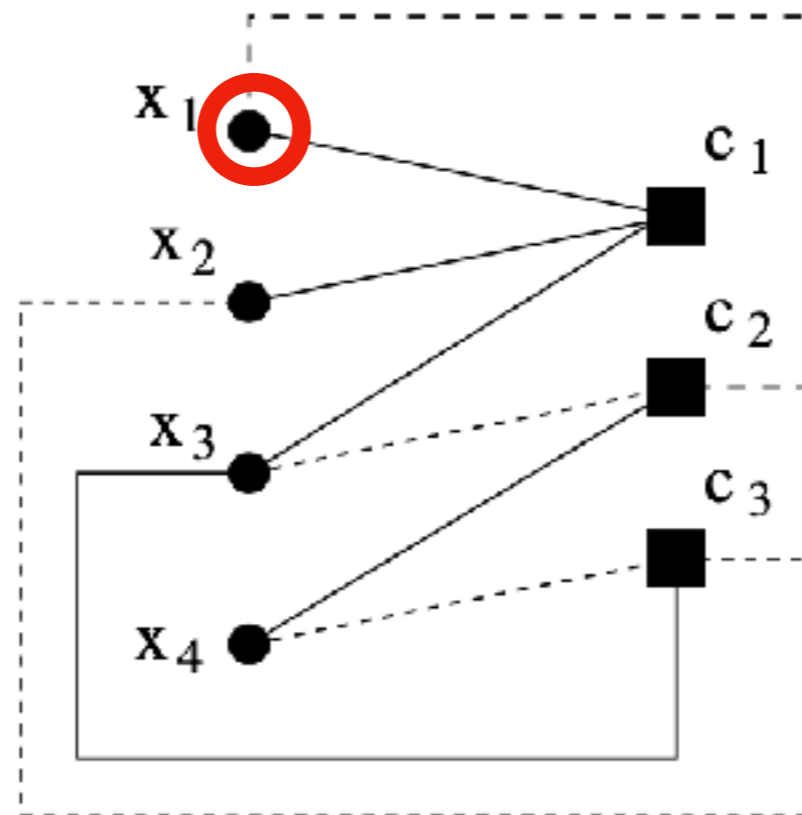
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**Fig. 1.** The graph  $G_I$  representing the PLANAR 3SAT instance  $(x_1 \vee x_2 \vee x_3) \wedge (\bar{x}_1 \vee \bar{x}_3 \vee x_4) \wedge (\bar{x}_2 \vee x_3 \vee \bar{x}_4)$ . Edges are distinguished according to the logical parity of the corresponding literals.

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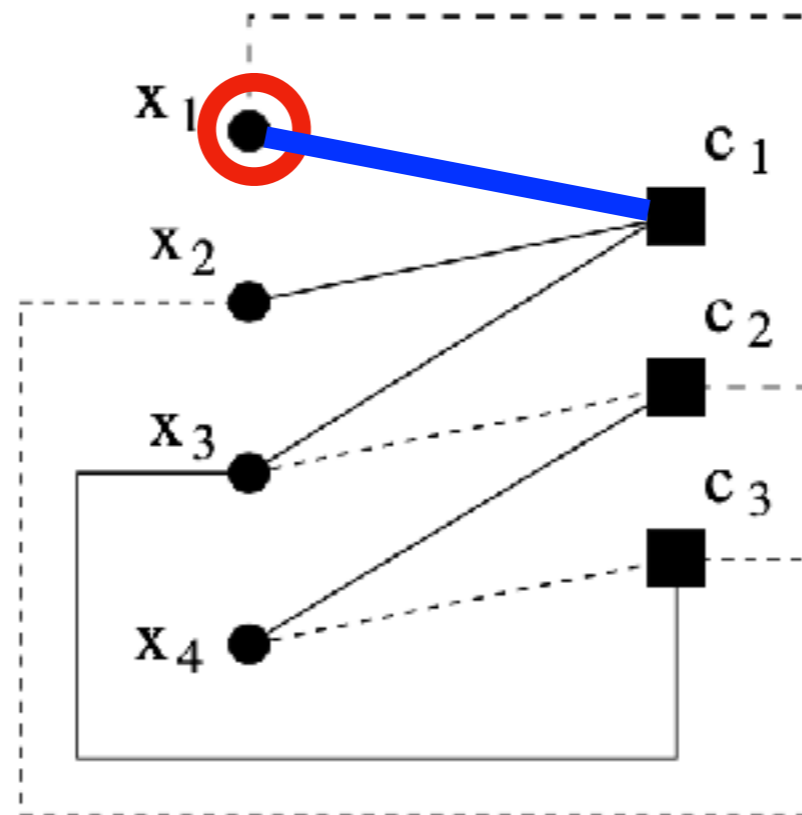
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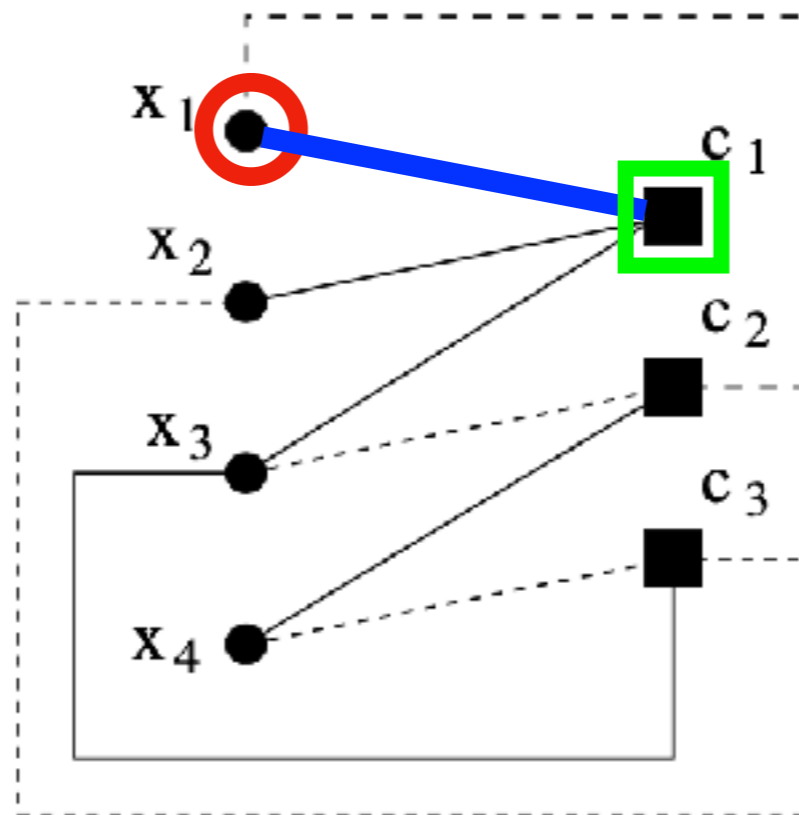
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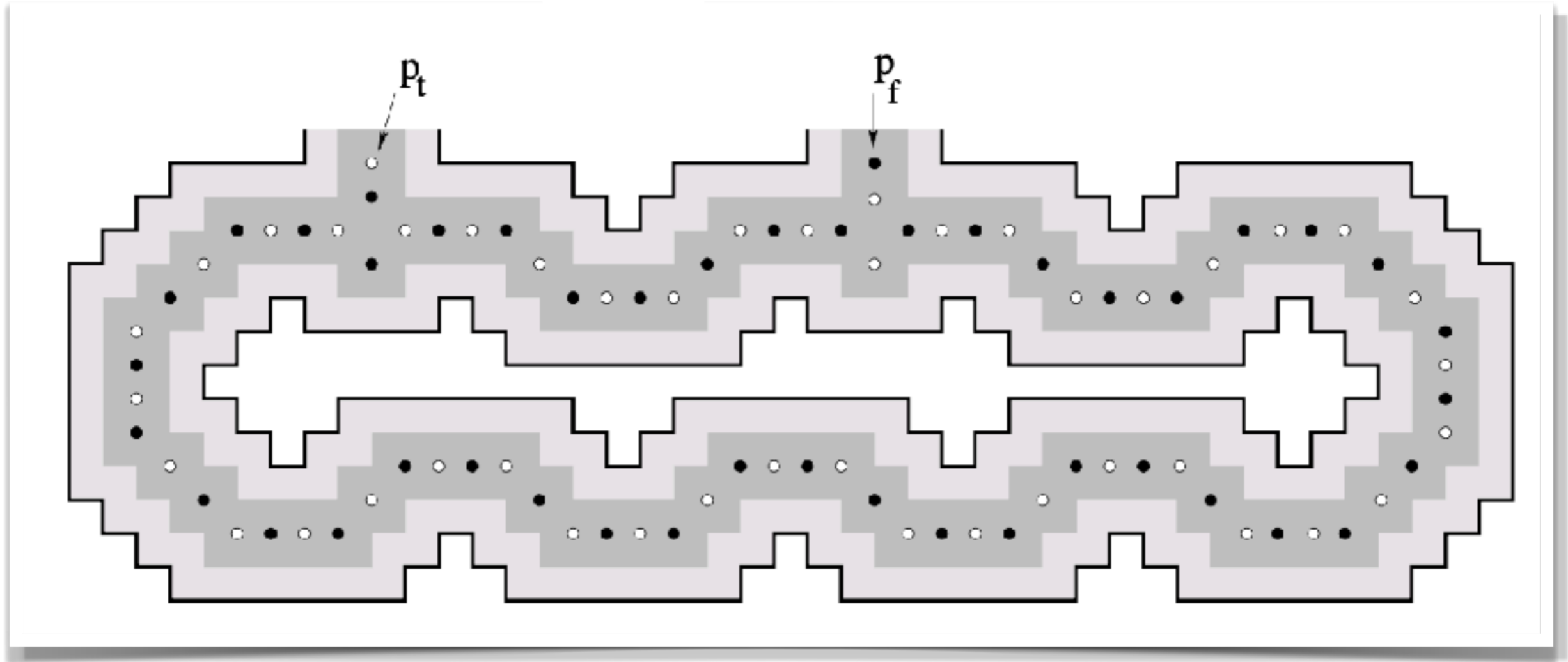
# ○ Variables (1)



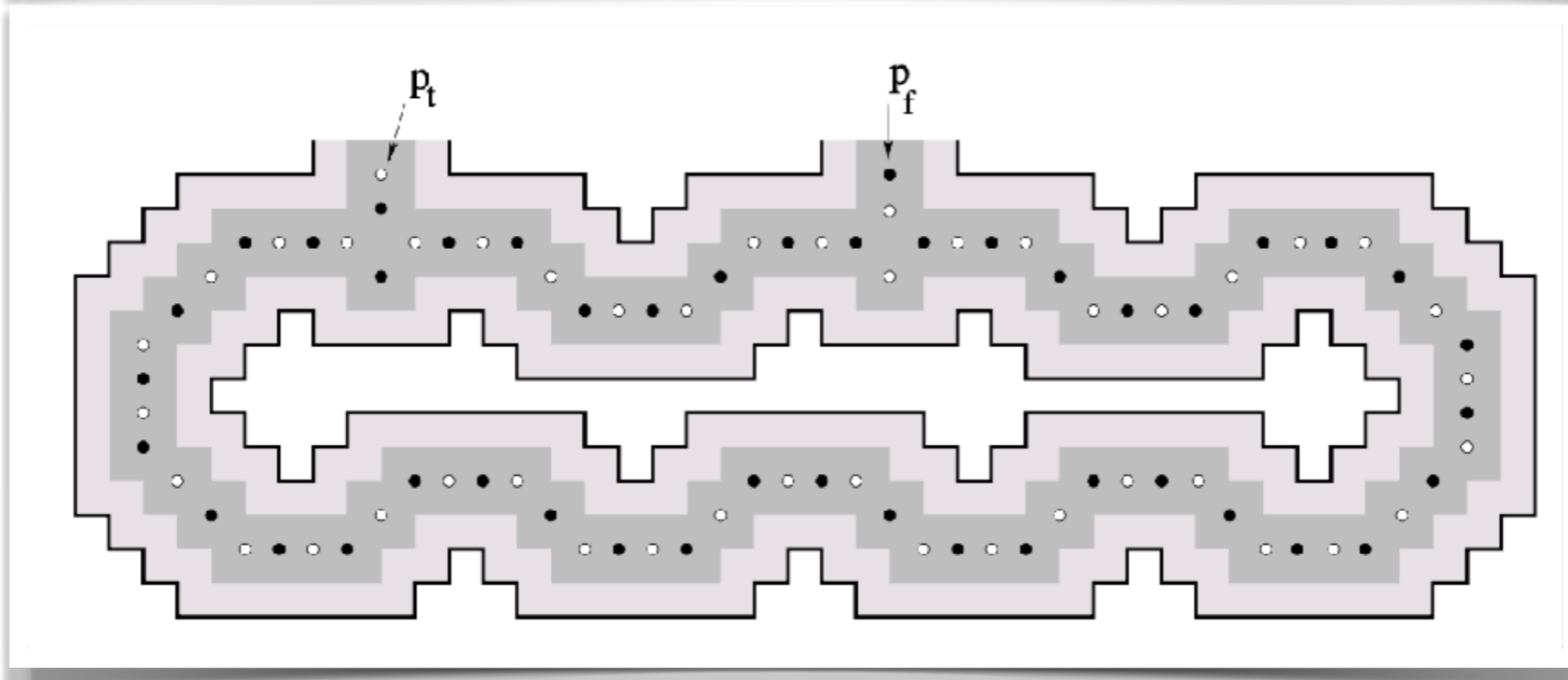
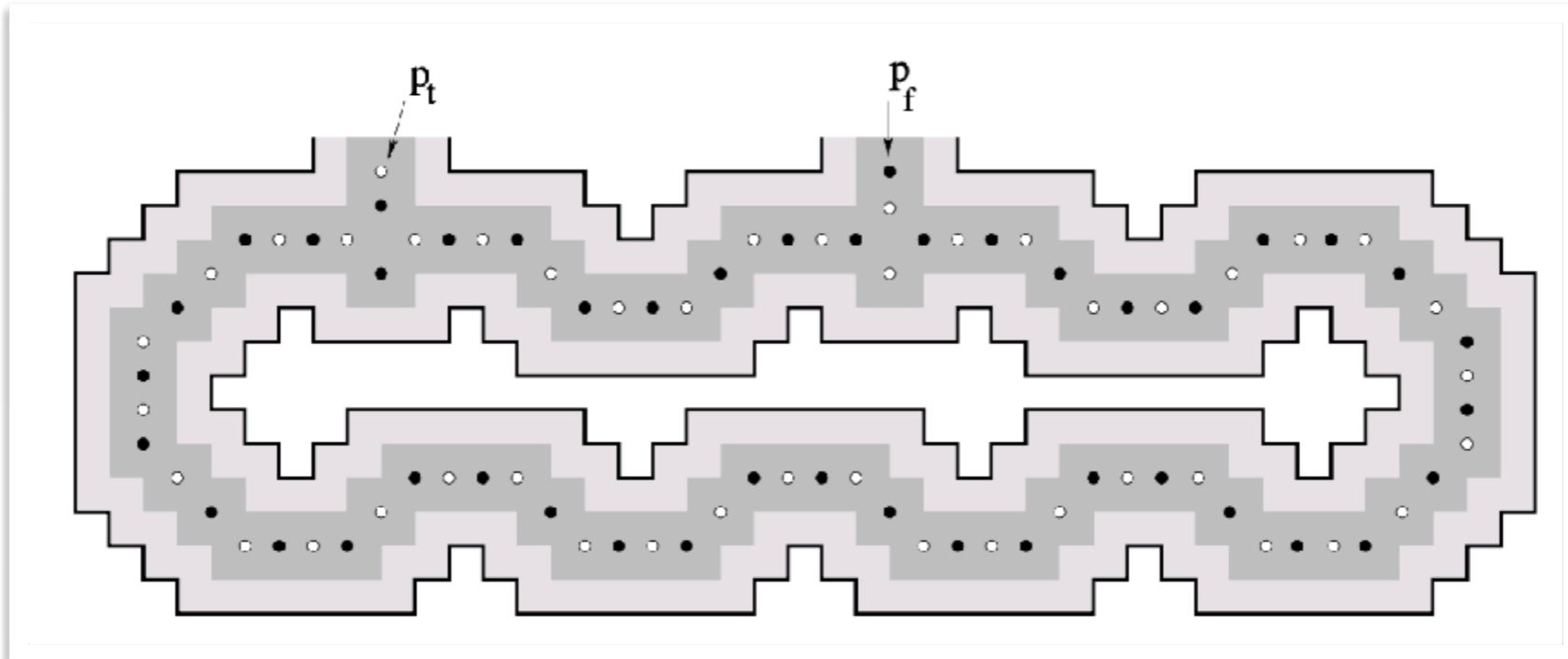
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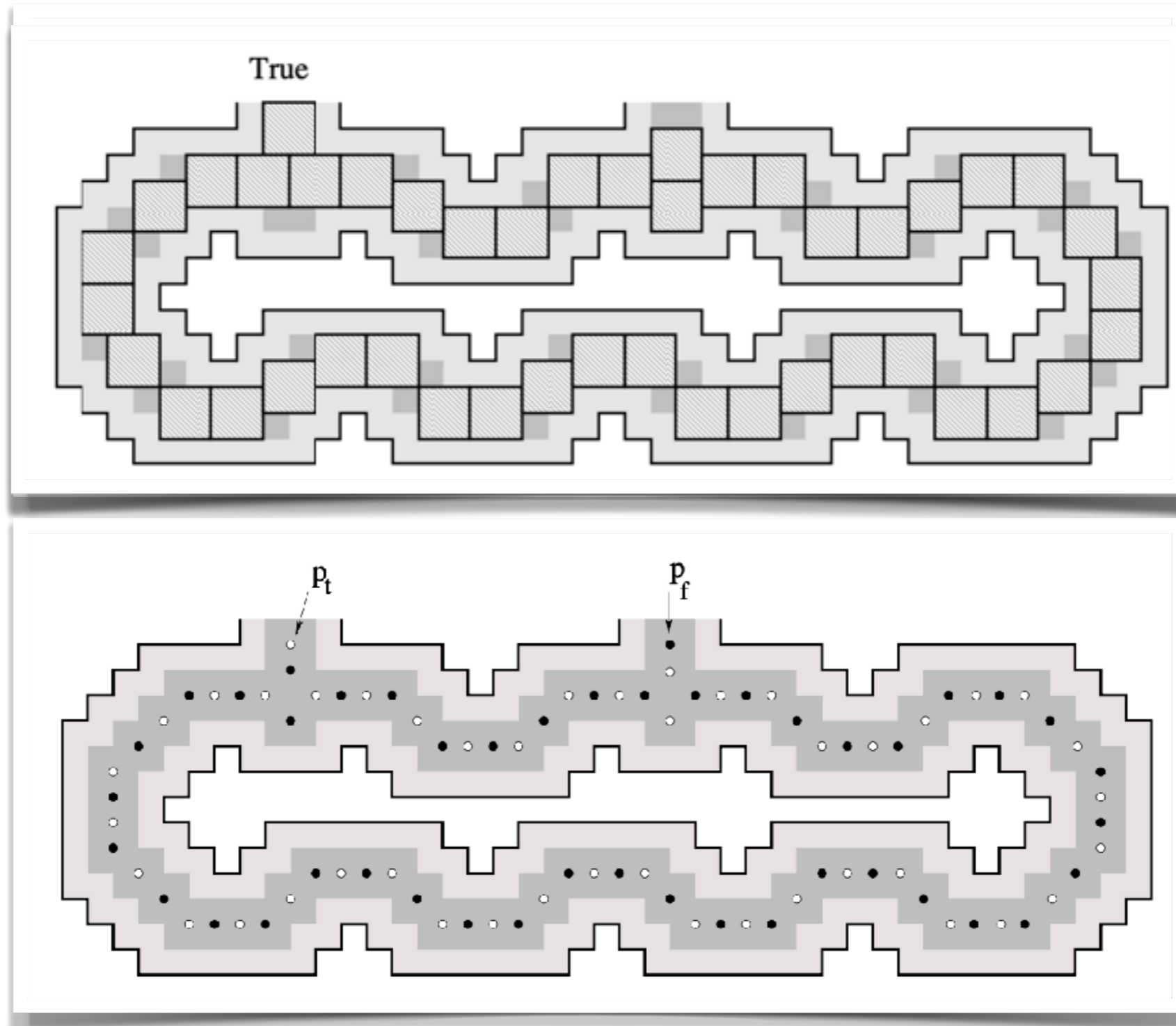
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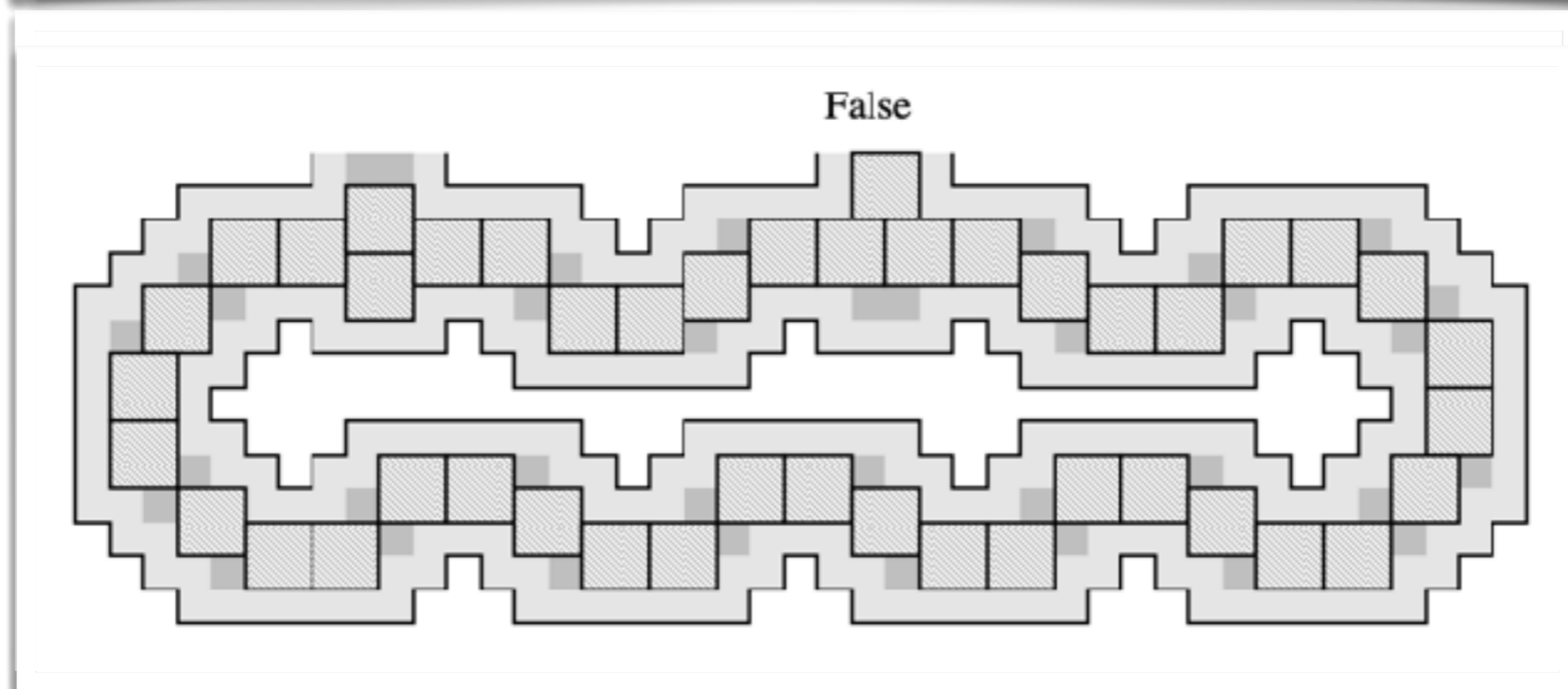
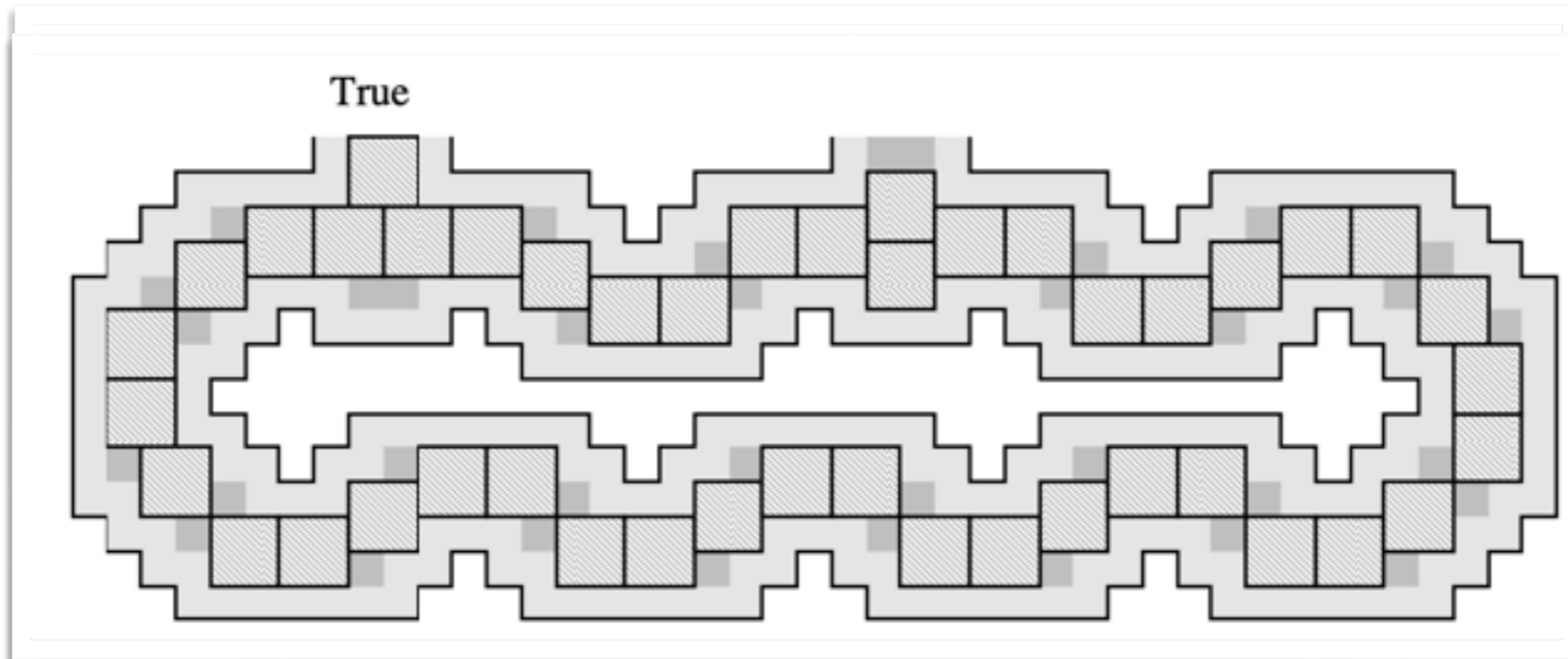
# ○ Variables (2)



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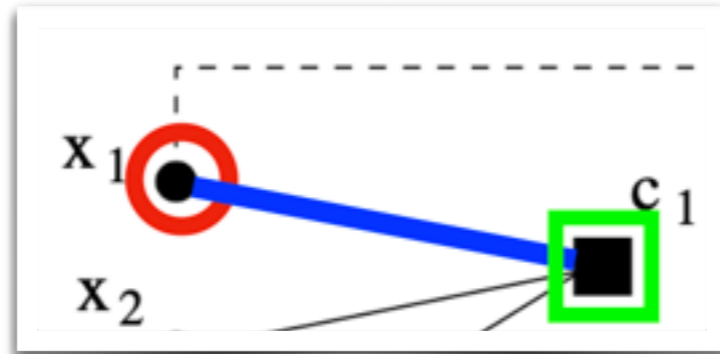


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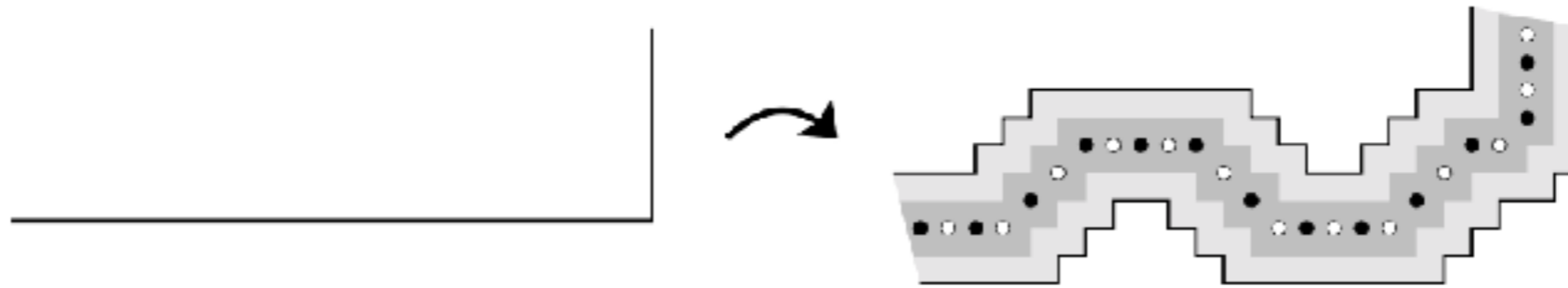
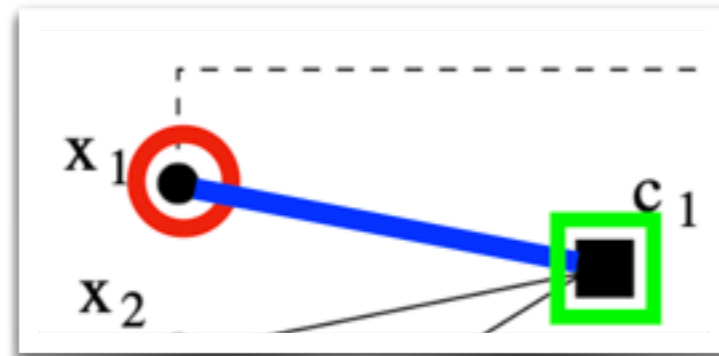


# — Connectors

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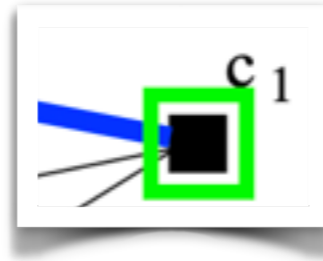


**Fig. 3.** Part of a connector component (right) for representing an edge between variable nodes and clause nodes (left).

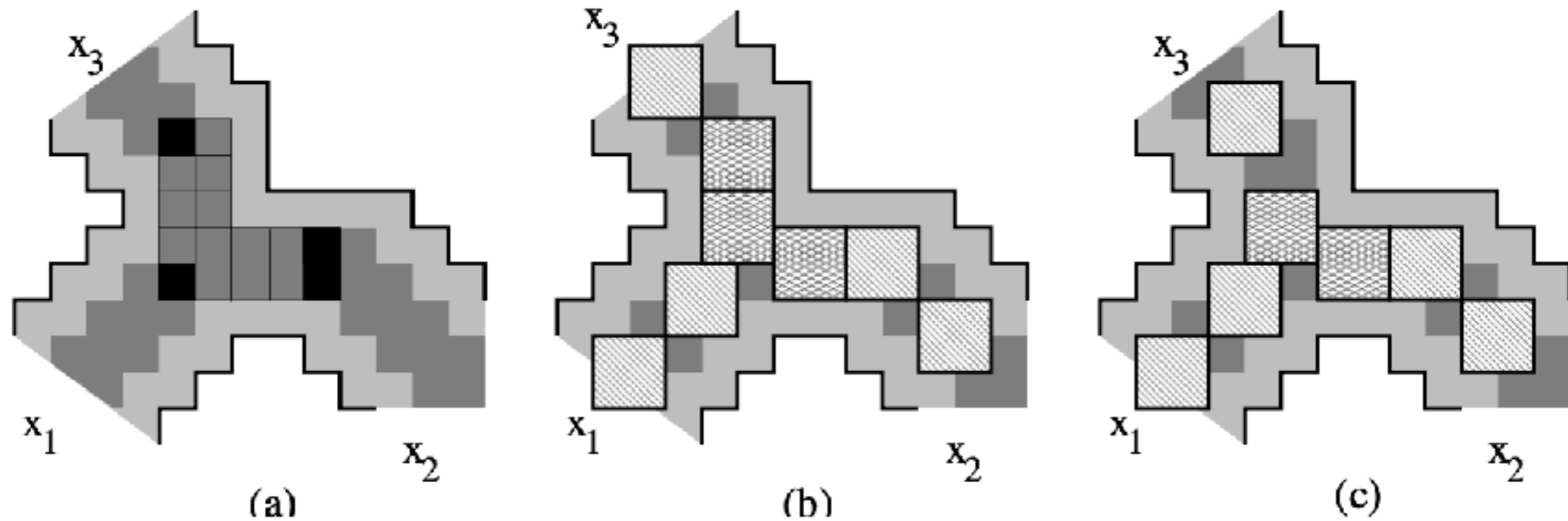
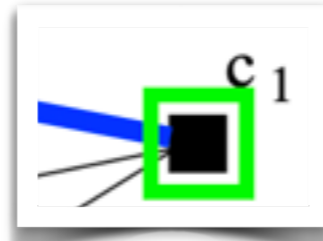


# Clauses

# □ Clauses



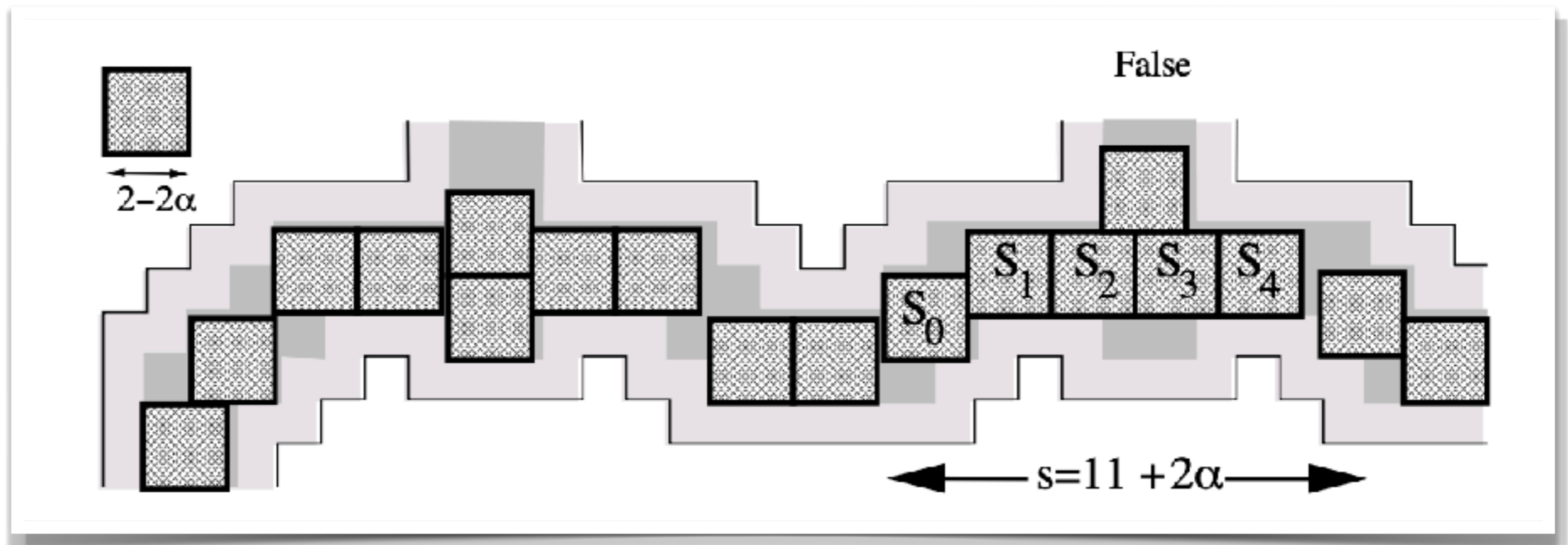
# □ Clauses



**Fig. 4.** A clause component for dispersion with boundaries and its receptor region (a); a satisfying placement (b); and an unsatisfying placement (c).

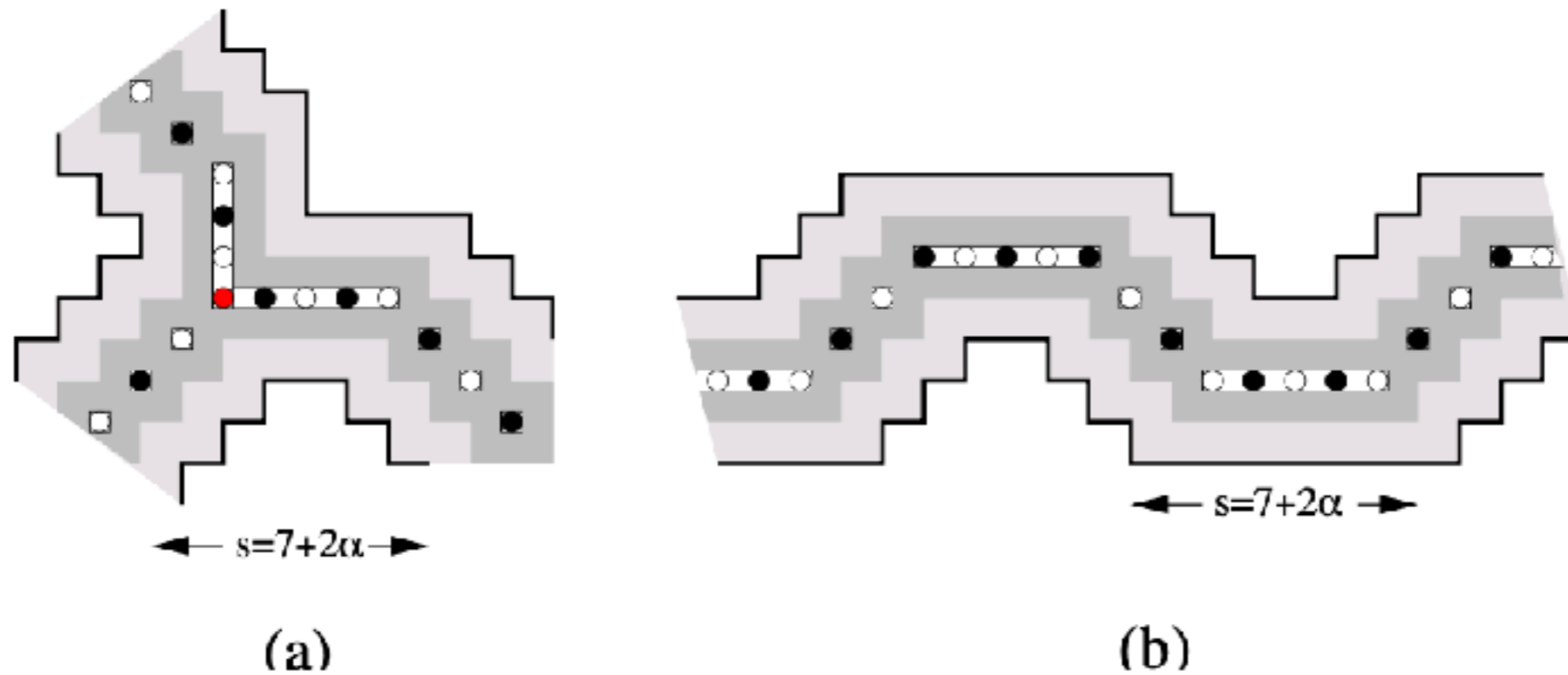
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# Smaller Squares (2)

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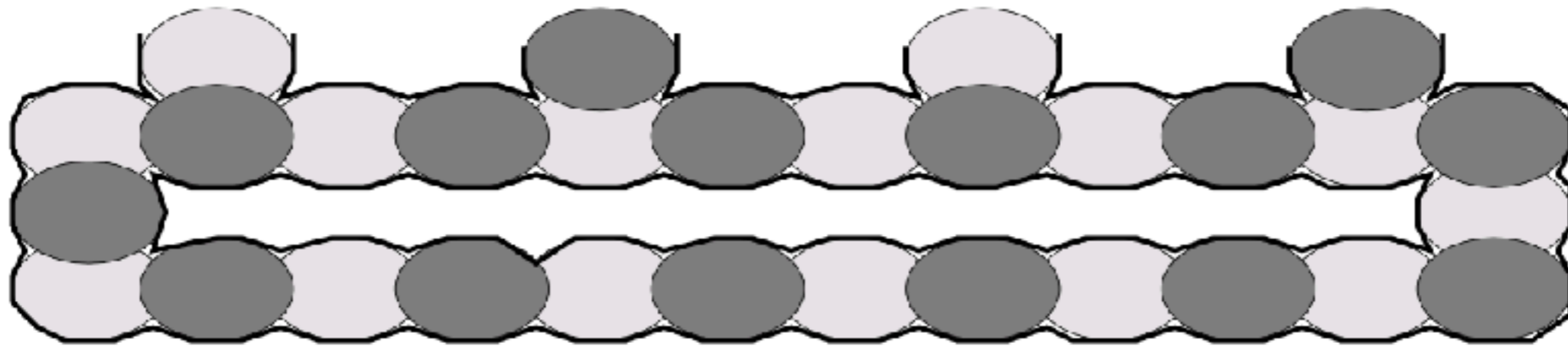


**Fig. 7.** An upper bound on the approximation factor: clause components (a) and connector components (b).

# Other Metrics



# Other Metrics



**Fig. 9.** A variable gadget for the case where the unit ball is an ellipse.

# Approximation

# Approximation

**THEOREM 9.** *For rectilinear geometric dispersion with boundaries of  $k$  locations in a rectilinear polygon  $P$  with  $n$  vertices, Algorithm 8 computes a solution  $A_{Dis}(P, k)$ , such that*

$$A_{Dis}(P, k) \geq \frac{2}{3}OPT(P, k).$$

*The running time is strongly polynomial.*

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**ALGORITHM 8.**

*Input:* rectilinear polygon  $P$ , positive integer  $k$ .

*Output:* a set of  $k$  locations, such that  $A_{Dis}(P, k) := d$  is the minimum  $L_\infty$  distance between a location and the boundary, or between two locations.

1. **For all**  $(e_i, e_j) \in Par(P)$  **do**
  - (a) Perform binary search for the smallest integer  $m$ ,  $2 \leq m \leq k + 1$ , with the following property:
    - For  $d_{ijm} := Dist(e_i, e_j)/m$ ,  $AS(P - d_{ijm}/2, d_{ijm}, 6)$  returns a feasible solution for at least  $k$  locations at distance  $d_{ijm}$ .
  - (b) Let  $d_{ij}$  be the distance  $d_{ijm}$  for the critical value  $m$ .
2. Let  $d$  be the maximum  $d_{ij}$  for any  $(e_i, e_j)$ .

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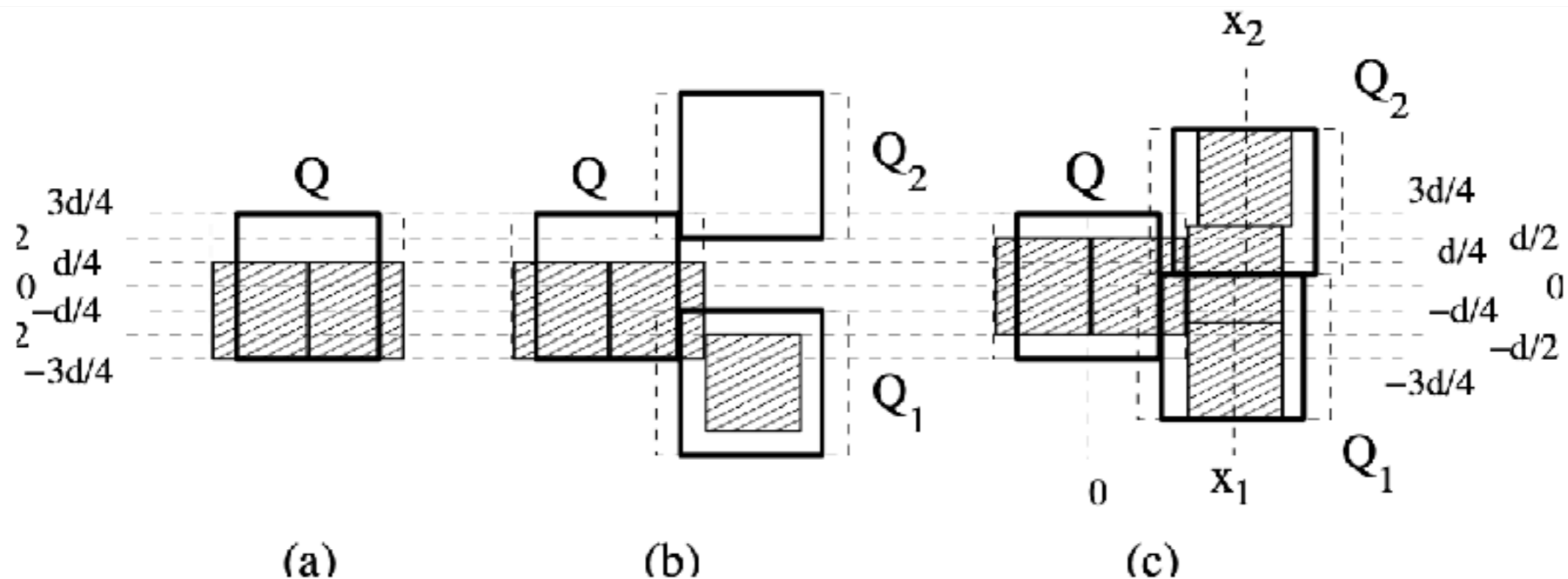
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The running time is strongly polynomial.



**Fig. 12.** Constructing a packing of  $d$ -squares.

# Average Distance

# Average Distance

Algorithmica (2004) 38: 501–511  
DOI: 10.1007/s00453-003-1074-x

Algorithmica

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## Maximum Dispersion and Geometric Maximum Weight Cliques<sup>1</sup>

Sándor P. Fekete<sup>2</sup> and Henk Meijer<sup>3</sup>

**Abstract.** We consider a facility location problem, where the objective is to “disperse” a number of facilities, i.e., select a given number  $k$  of locations from a discrete set of  $n$  candidates, such that the average distance between selected locations is maximized. In particular, we present algorithmic results for the case where vertices are represented by points in  $d$ -dimensional space, and edge weights correspond to rectilinear distances. Problems of this type have been considered before, with the best result being an approximation algorithm with performance ratio 2. For the case where  $k$  is fixed, we establish a linear-time algorithm that finds an optimal solution. For the case where  $k$  is part of the input, we present a polynomial-time approximation scheme.

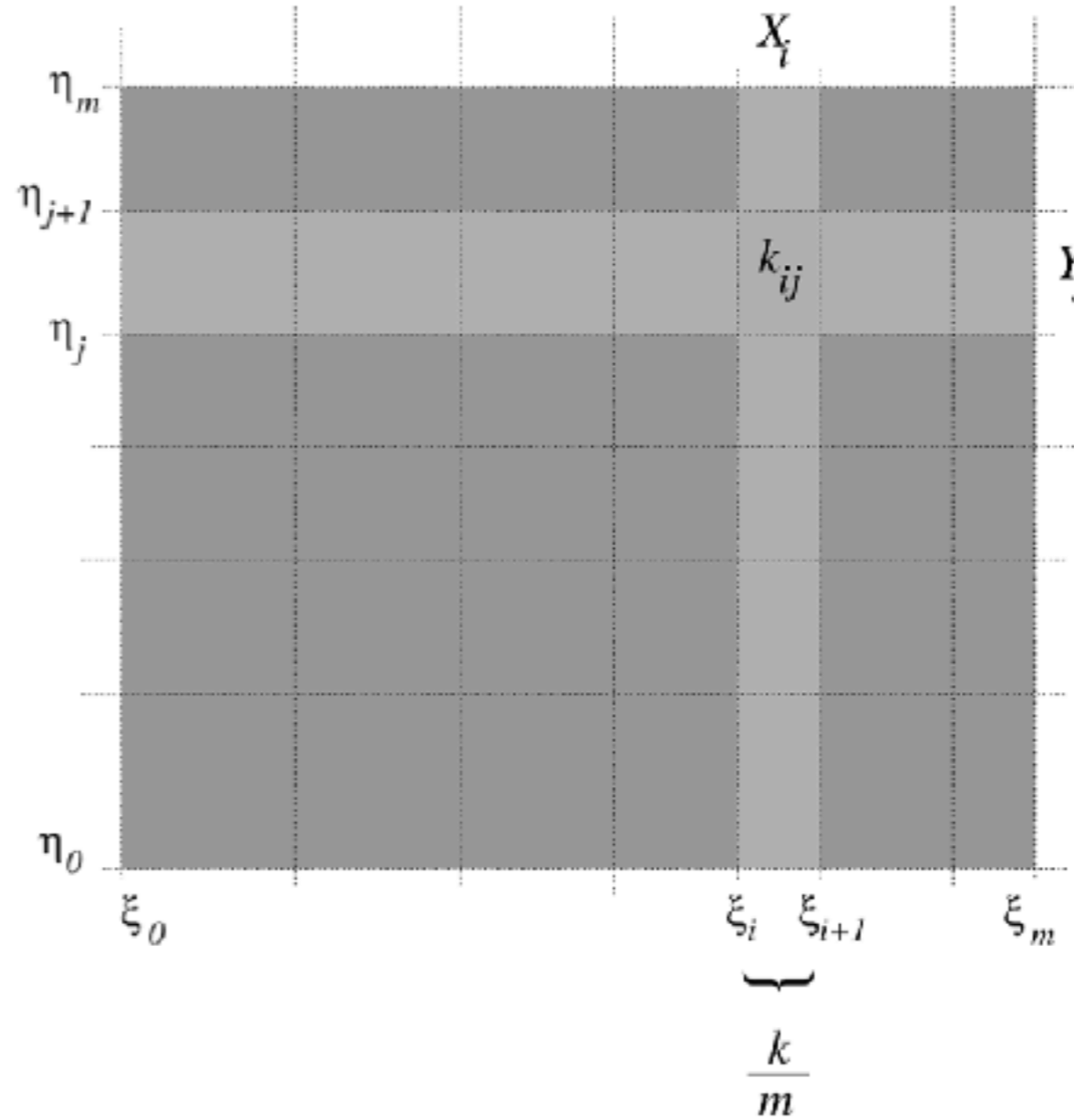
**Key Words.** Dispersion, Facility location, Maximum weight cliques, Remote clique, Heaviest subgraph, Geometric optimization, Approximation, PTAS.



# Average Distance

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**Fig. 1.** Subdividing the plane into cells.

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# Lawn Mowing (1)

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Computational Geometry 17 (2000) 25–50

Computational  
Geometry

Theory and Applications

[www.elsevier.nl/locate/comgeo](http://www.elsevier.nl/locate/comgeo)

## Approximation algorithms for lawn mowing and milling <sup>☆</sup>

Esther M. Arkin <sup>a,1</sup>, Sándor P. Fekete <sup>b,\*</sup>, Joseph S.B. Mitchell <sup>a,2</sup>

<sup>a</sup> *Department of Applied Mathematics and Statistics, SUNY Stony Brook, NY 11794-3600, USA*

<sup>b</sup> *Department of Mathematics, TU Berlin, 10623 Berlin, Germany*

Communicated by K. Mehlhorn; received 18 March 1997; accepted 12 August 1999

### Abstract

We study the problem of finding shortest tours/paths for “lawn mowing” and “milling” problems: Given a region in the plane, and given the shape of a “cutter” (typically, a circle or a square), find a shortest tour/path for the cutter such that every point within the region is covered by the cutter at some position along the tour/path. In the milling version of the problem, the cutter is constrained to stay within the region. The milling problem arises naturally in the area of automatic tool path generation for NC pocket machining. The lawn mowing problem arises in optical inspection, spray painting, and optimal search planning.

# Lawn Mowing (1)



Computational Geometry 17 (2000) 25–50

Computational  
Geometry

Theory and Applications

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## Approximation algorithms for lawn mowing and milling <sup>☆</sup>

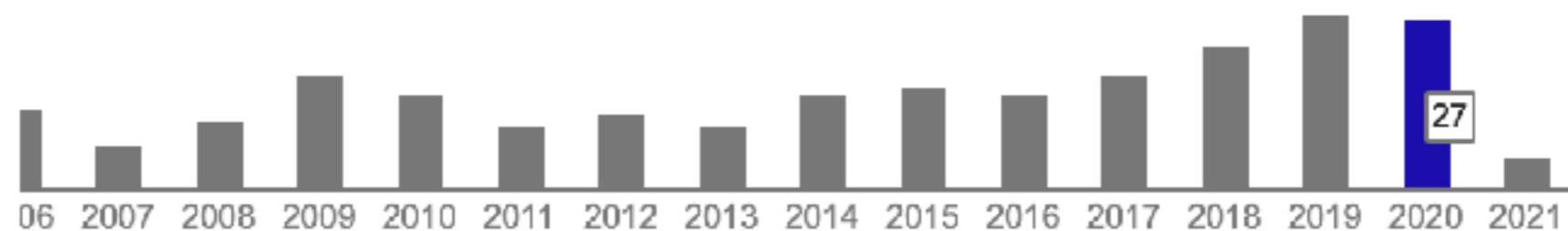
Esther M. Arkin <sup>a,1</sup>, Sándor P. Fekete <sup>b,\*</sup>, Joseph S.B. Mitchell <sup>a,2</sup>

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Total citations Cited by 283



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*E.M. Arkin et al. / Computational Geometry 17 (2000) 25–50*

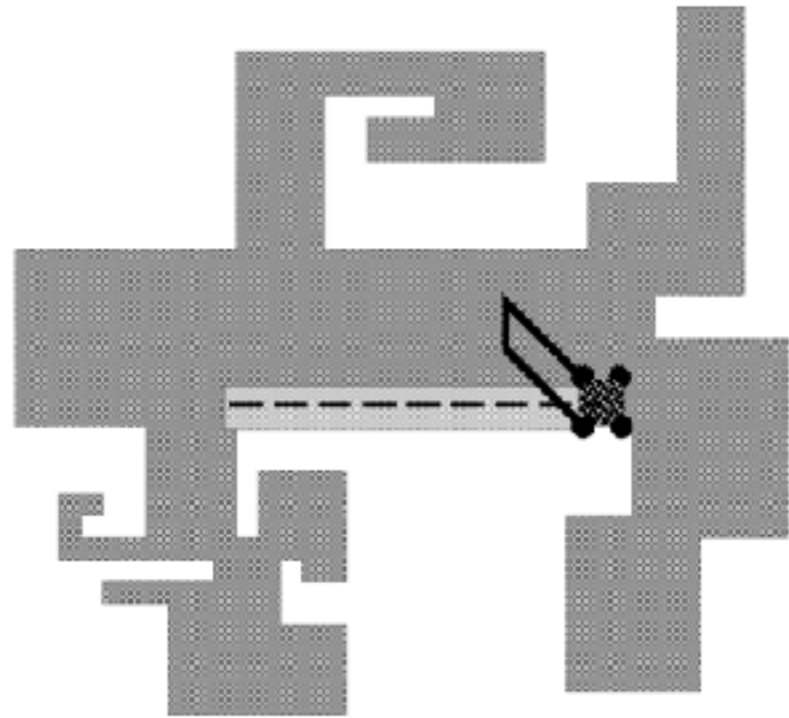


Fig. 1. The lawn mowing problem.



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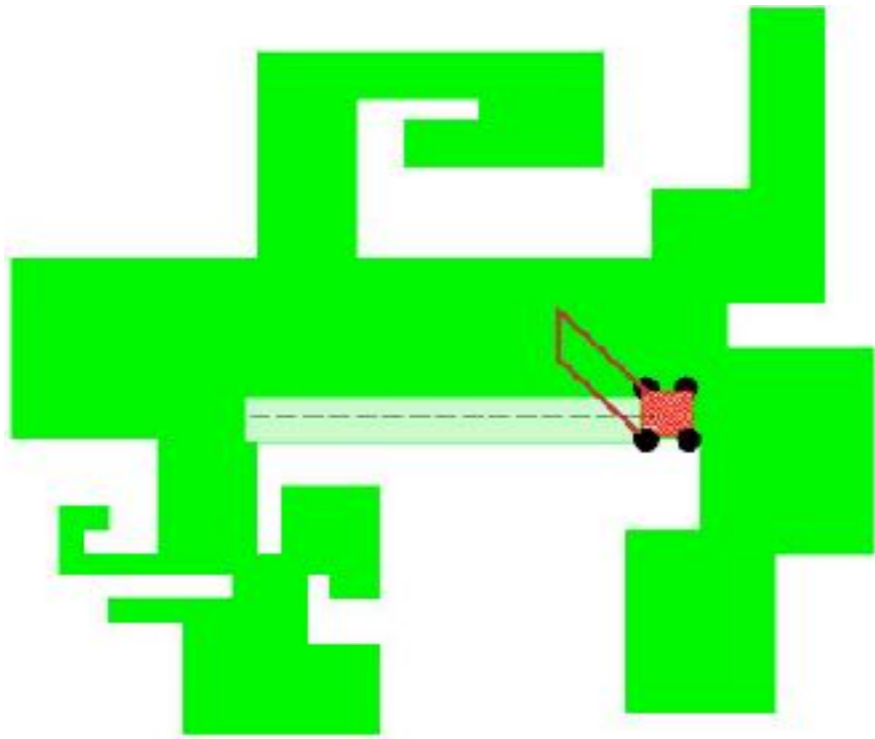


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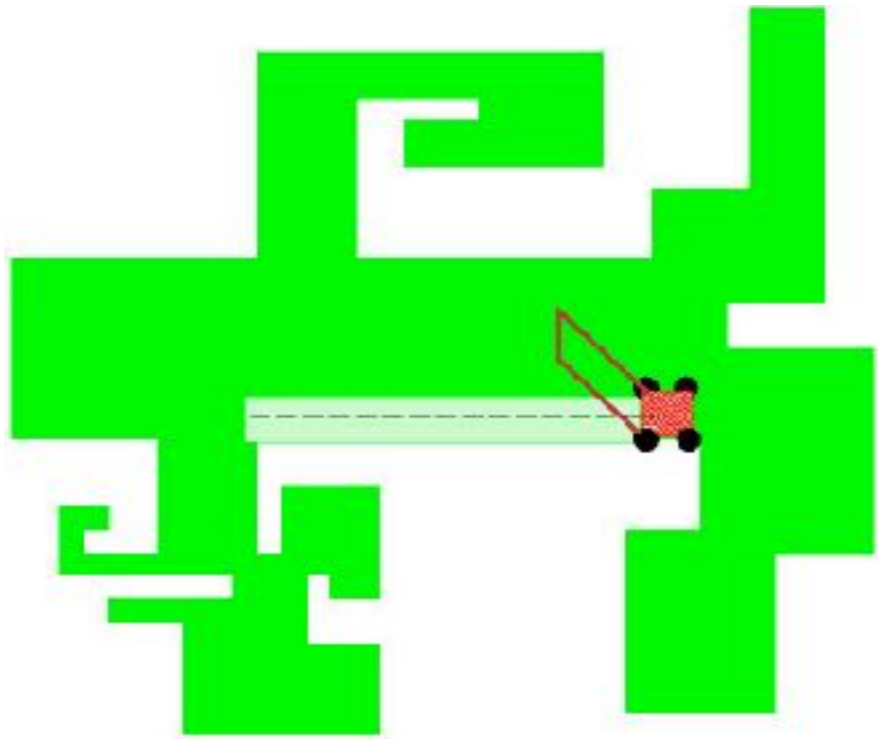
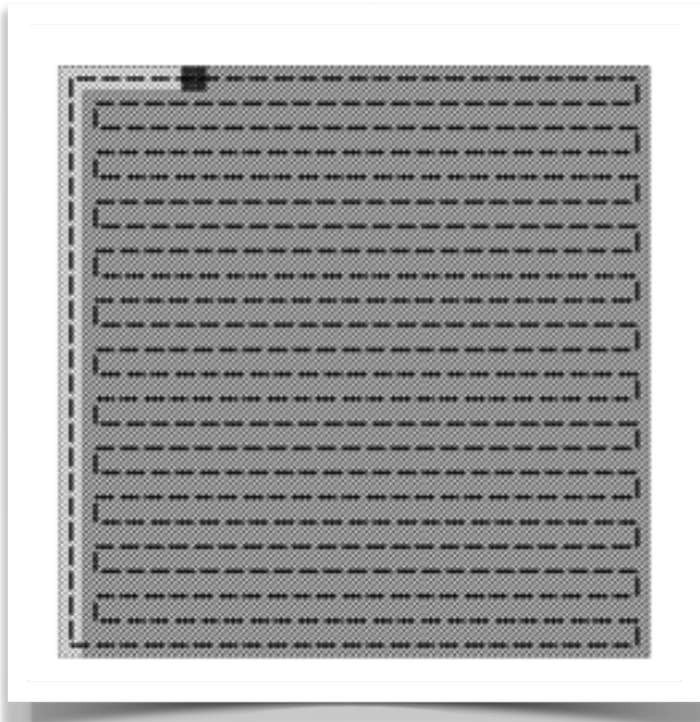


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# Lawn Mowing (3)



*E.M. Arkin et al. / Computational Geometry 17 (2000) 25–50*

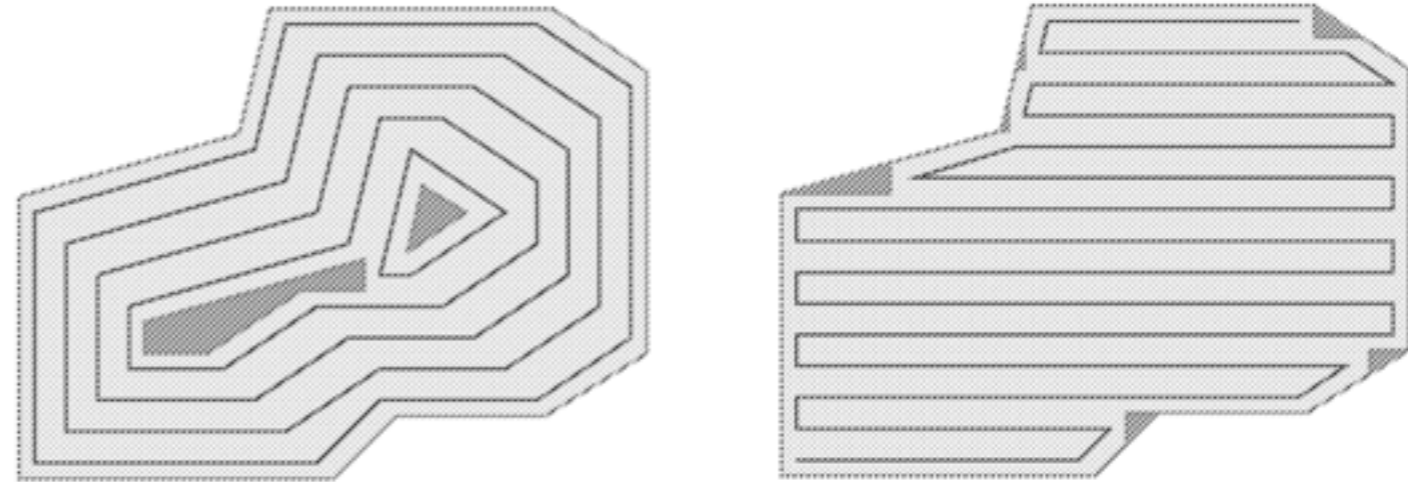


Fig. 4. Left: Contour-parallel milling. Right: Axis-parallel milling.

# Lawn Mowing (4)

**Theorem 1.** *The lawn mowing problem for a connected polygonal region is NP-hard for the case of an aligned unit square cutter  $\chi$ .*

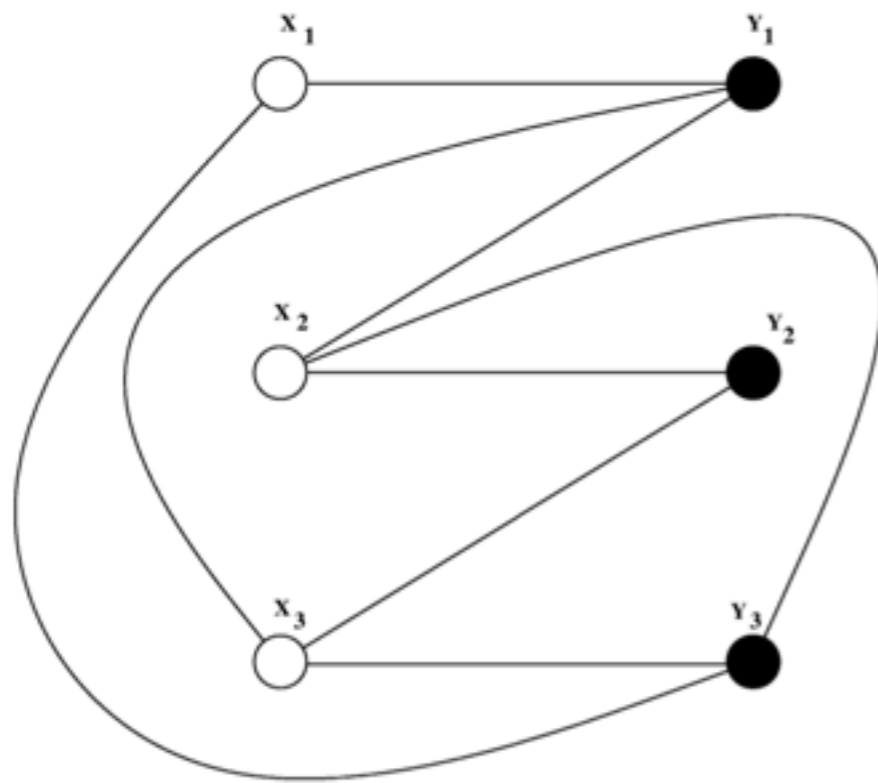
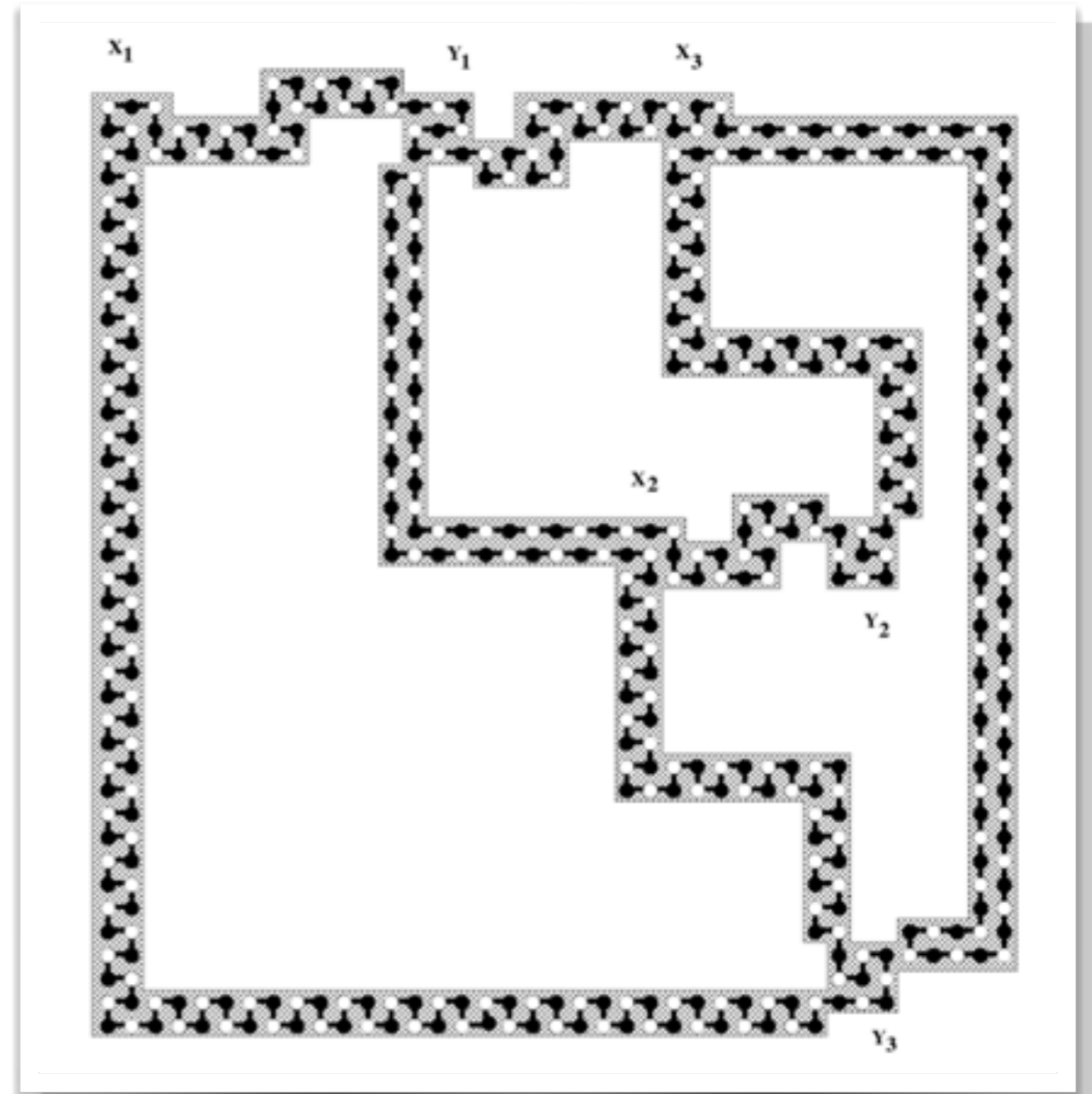


Fig. 5. A planar bipartite graph  $G$ , with maximum vertex degree 3.



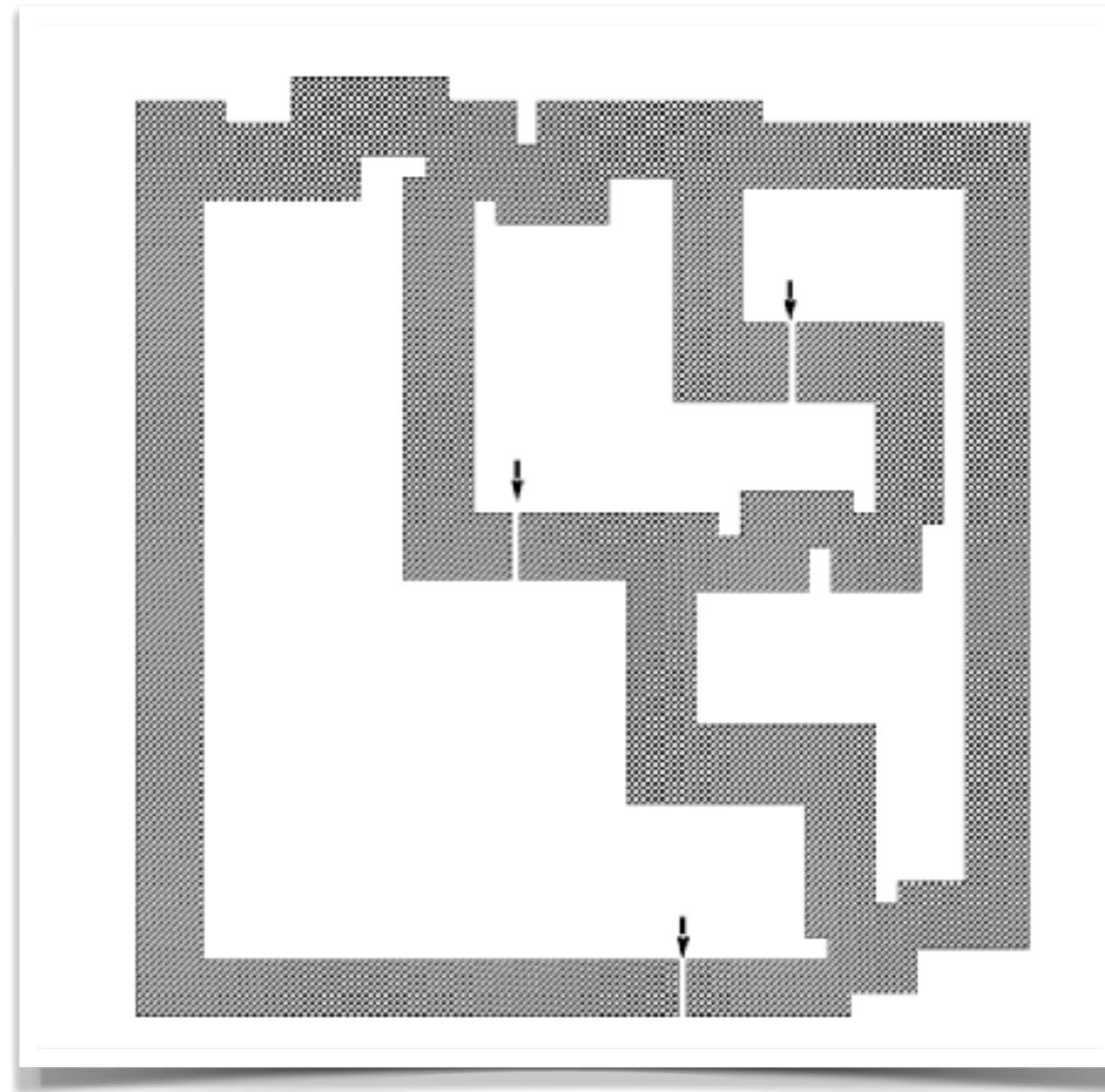
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**Theorem 2.** *Finding a rectilinear TSP approximation (with factor  $\alpha_{\text{TSP}}$ ) on the set of centerpoints  $S$  yields a lawn mowing tour of length at most  $\alpha_{\text{TSP}}(3\ell^* + 6)$ , where  $\ell^*$  is the length of an optimal lawn mowing tour.*

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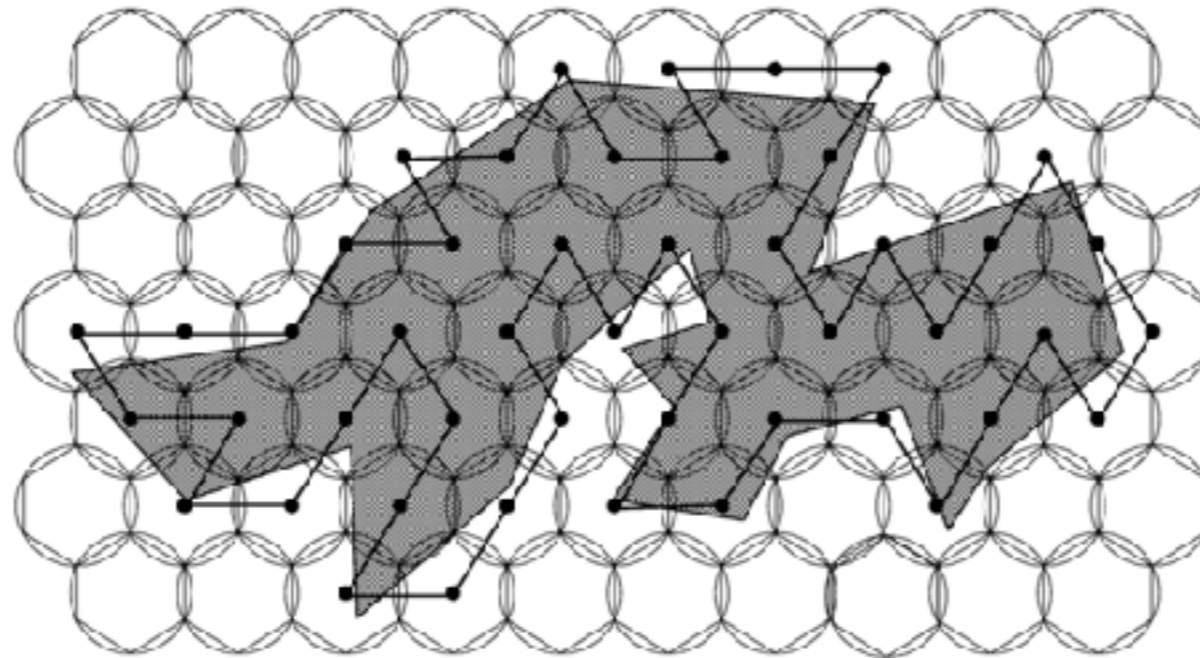
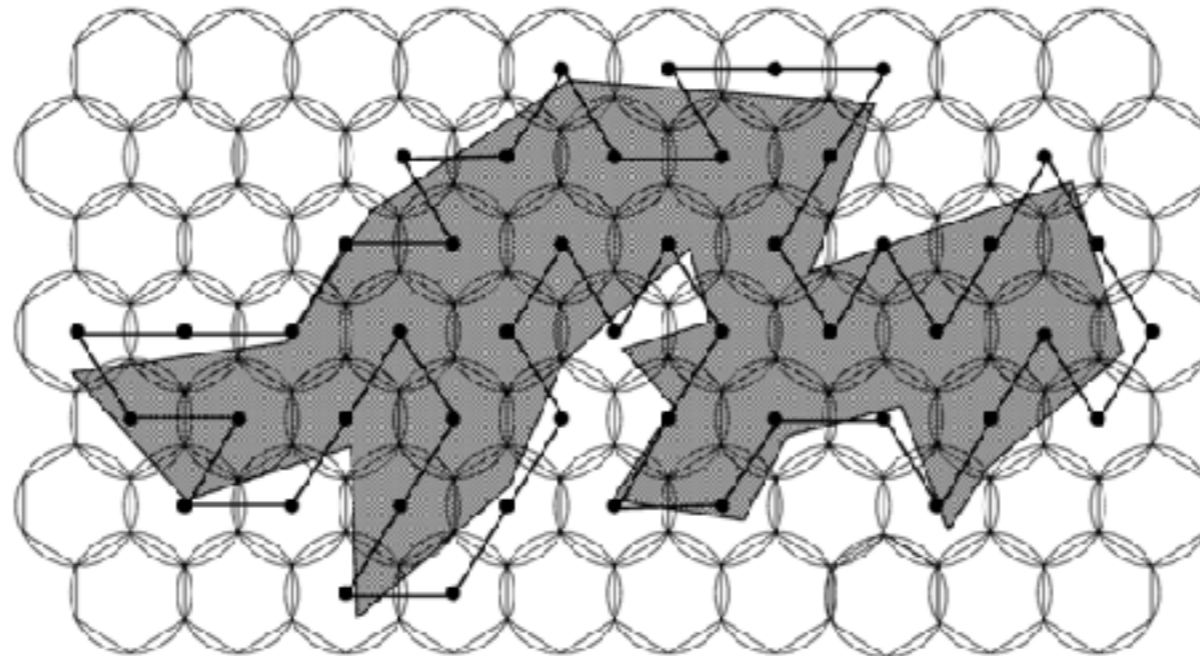


Fig. 10. Approximation for the case of a circular cutter  $\chi$ .

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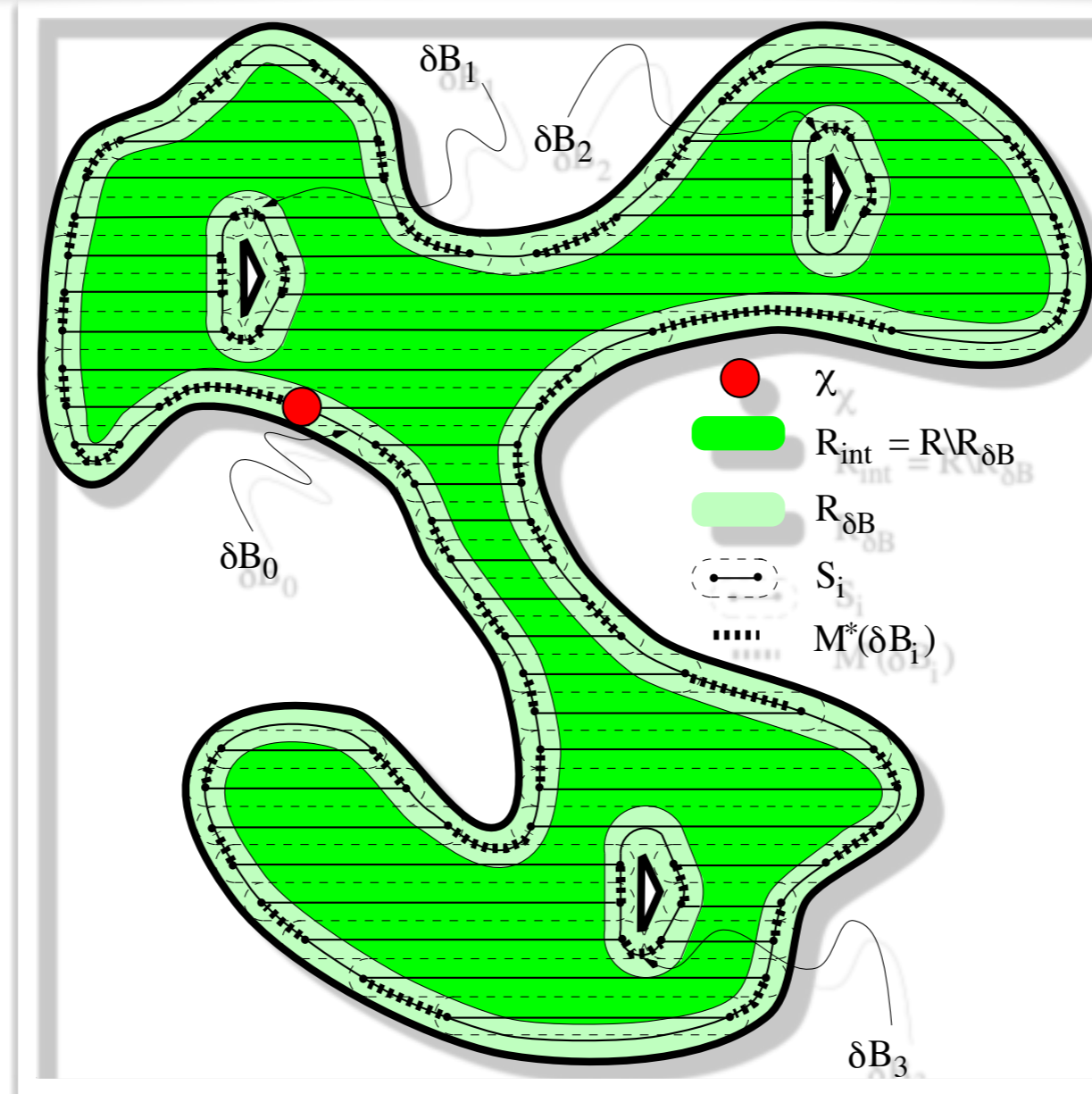
**Theorem 3.** The lawn mowing problem has a constant-factor approximation algorithm that runs in polynomial time (dependent on the TSP heuristic employed). For the case of an aligned unit square cutter, the approximation factor is  $3\alpha_{\text{TSP}}$  for rectilinear motion, and is  $3\beta\alpha_{\text{TSP}}$  for arbitrary translational motion. For the case of a unit circular cutter, with arbitrary motion, the approximation factor is  $3\gamma\alpha_{\text{TSP}}$ . Here,  $\beta = \frac{2}{\sqrt{2+\sqrt{2}}} \approx 1.08$  and  $\gamma = \frac{2\sqrt{3}}{3} \approx 1.15$ .

# Lawn Mowing (7)

**Theorem 4.** *In time  $O(n \log n)$ , one can decide whether a (multiply) connected region with  $n$  sides (straight or circular arc) can be milled by a unit disk or unit square, and, within the same time bound, one can construct a tour of length at most  $2\frac{1}{2}$  times the length of an optimal milling tour.*

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# Lawn Mowing (8)

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**Theorem 5.** *Let  $G$  be a simple grid graph, having  $N$  nodes at the centerpoints,  $V$ , of pixels within a simple rectilinear polygon,  $R$ , having  $n$  (integer-coordinate) sides. Assume that  $G$  has no cut vertices. Then, in time  $O(n)$ , one can find a representation of a tour,  $T$ , that visits all  $N$  nodes of  $G$ , of length at most  $\frac{6N-4}{5}$ .*

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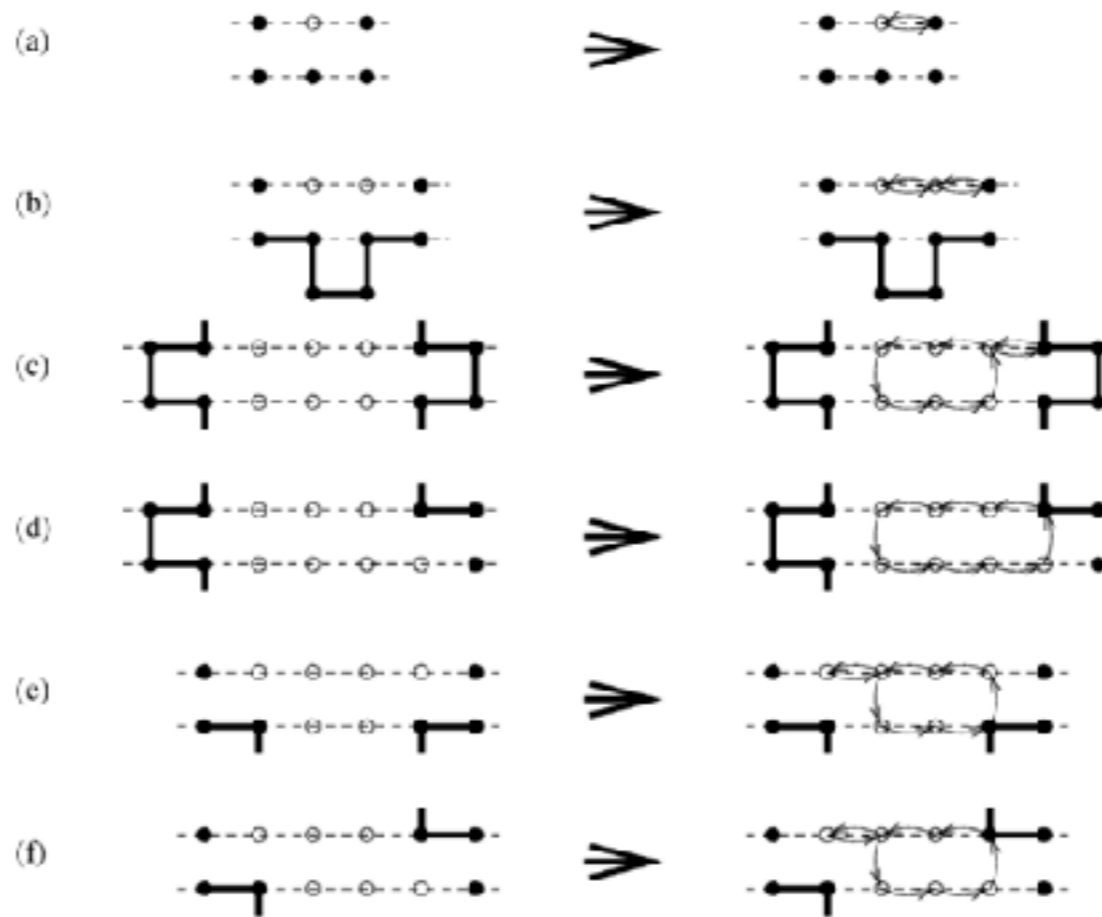


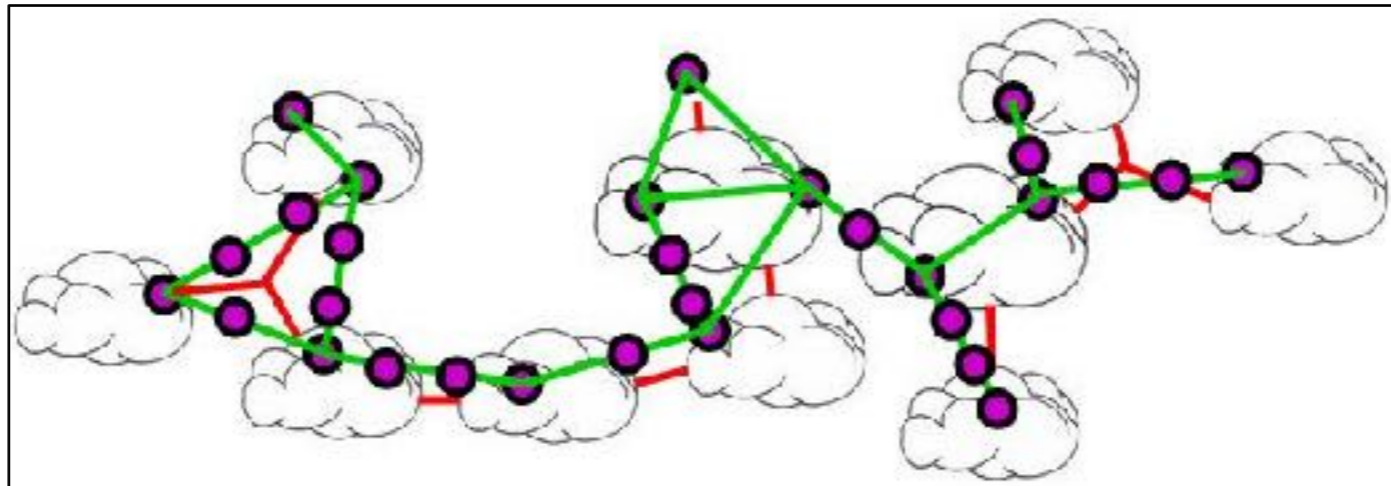
Fig. 15. Six cases for incorporating internal nodes into the modified contour tour. Hollow circles denote internal nodes,  $V'_i$ , and solid circles denote nodes of  $C'$ . Solid edges are drawn where there *must* be edges of  $C'$ .



- 1. Introduction**
- 2. Review**
- 3. Extra Packing: Dispersion**
- 4. Extra Tours: Lawn Mowing**
- 5. Relay Placement**
- 6. Coordinated Motion Planning**

# Approximation Algorithms for Relay Placement in the Plane

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Alon Efrat, Sándor P. Fekete, Joe Mitchell, Valentin Polishchuk,  
G. R. Poornananda, and Jukka Suomela

# Relay Placement

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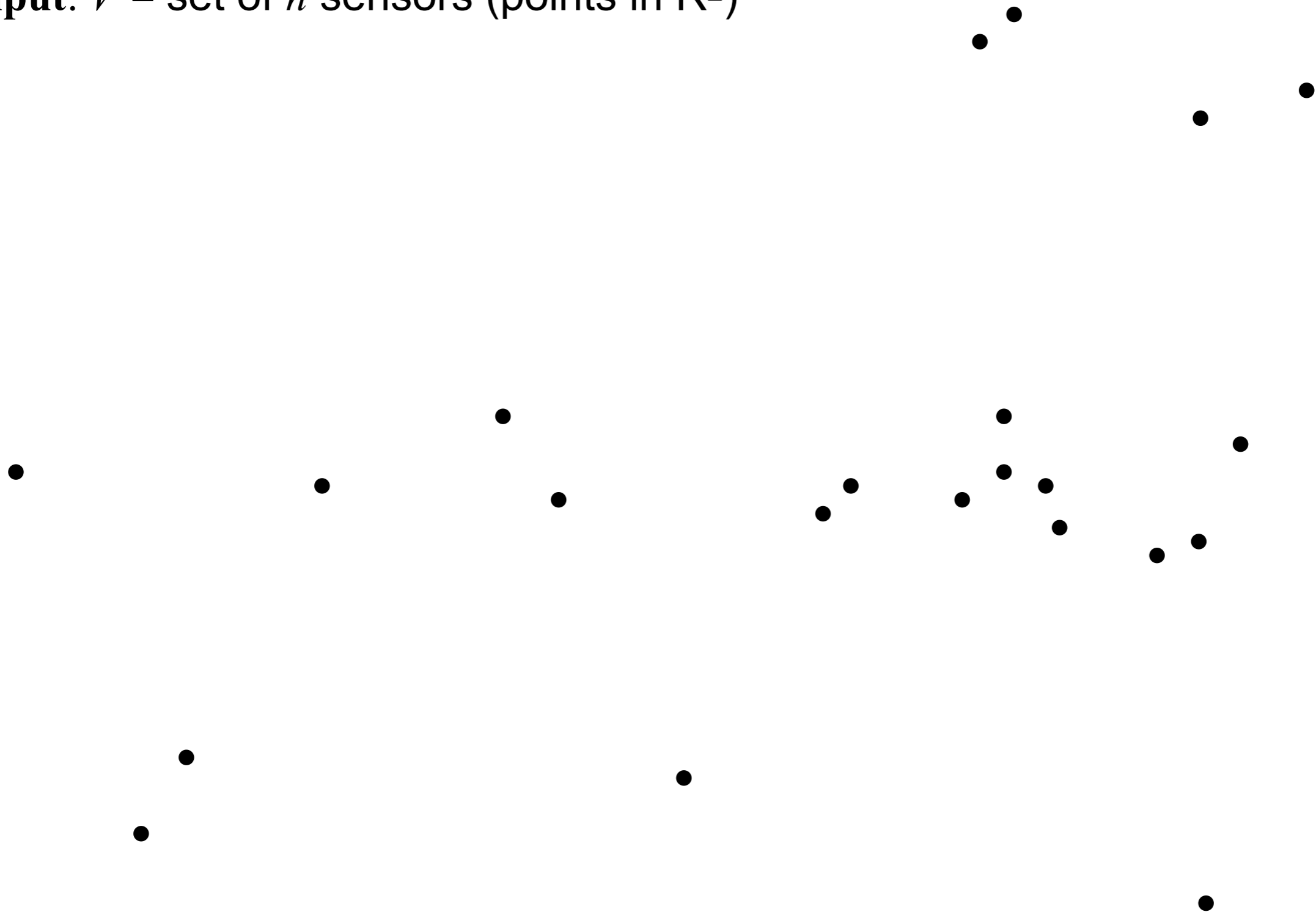
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**Input:**  $V$  = set of  $n$  sensors (points in  $\mathbb{R}^2$ )

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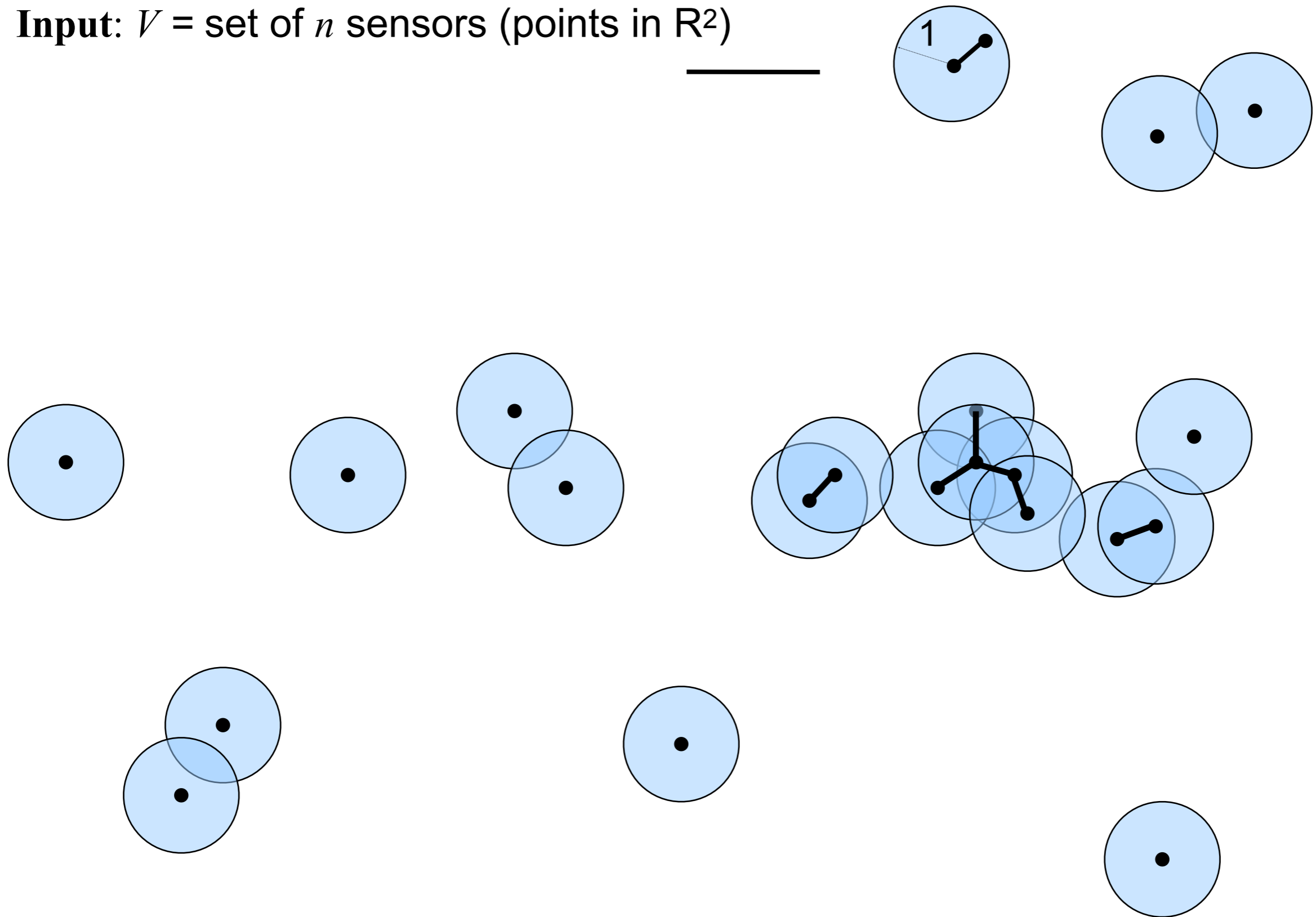
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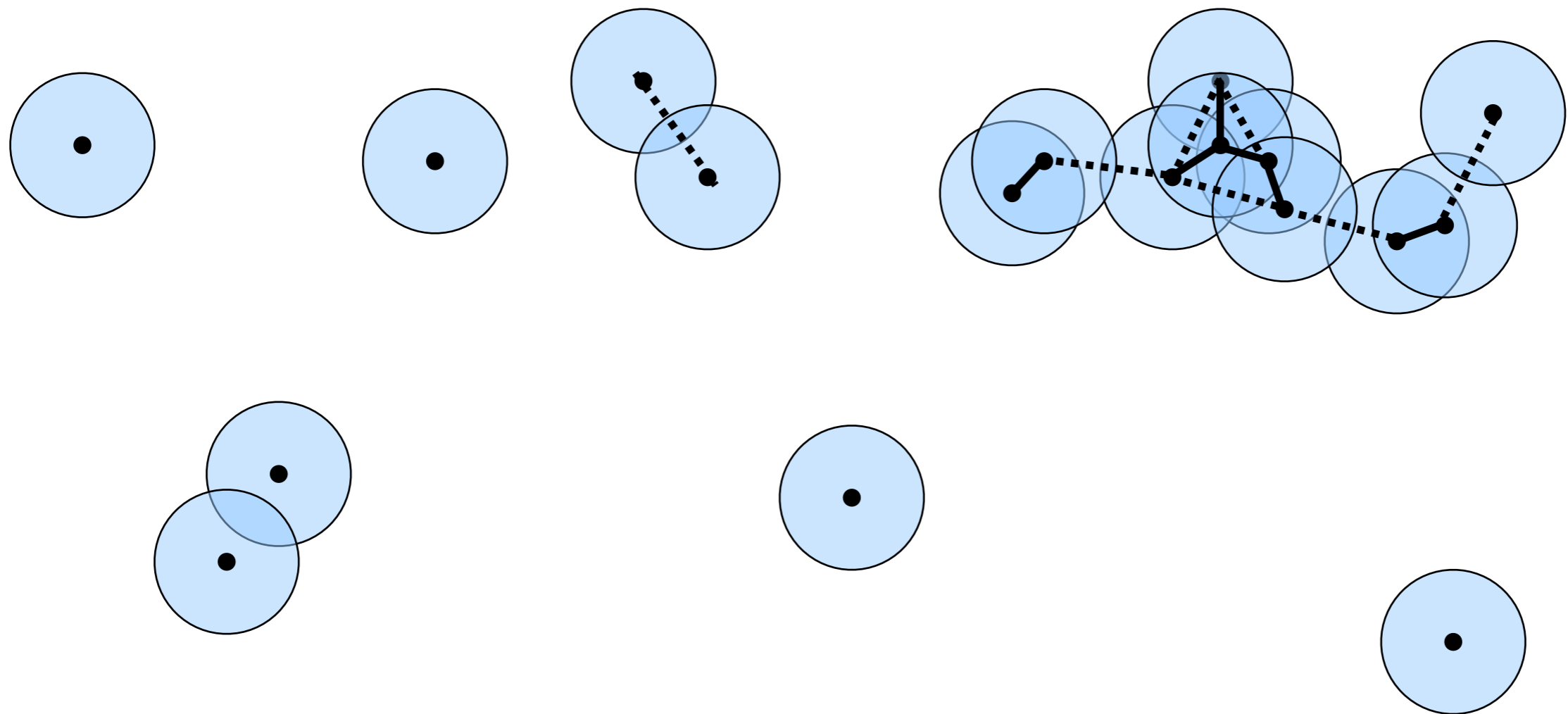
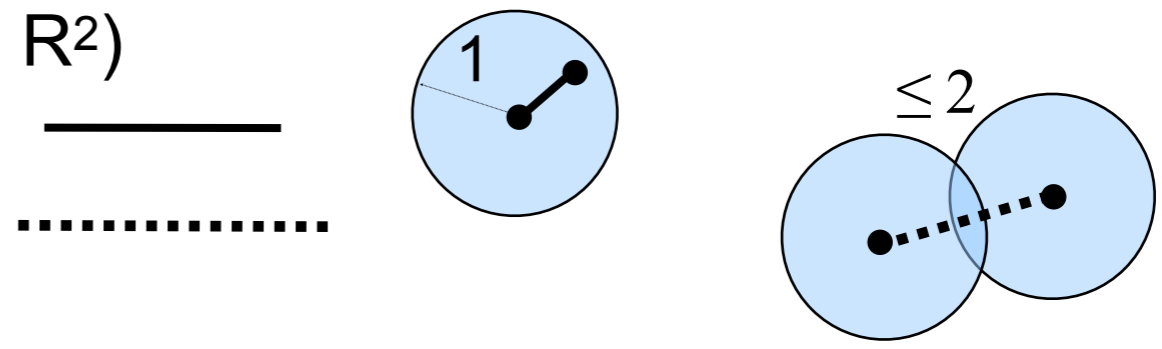
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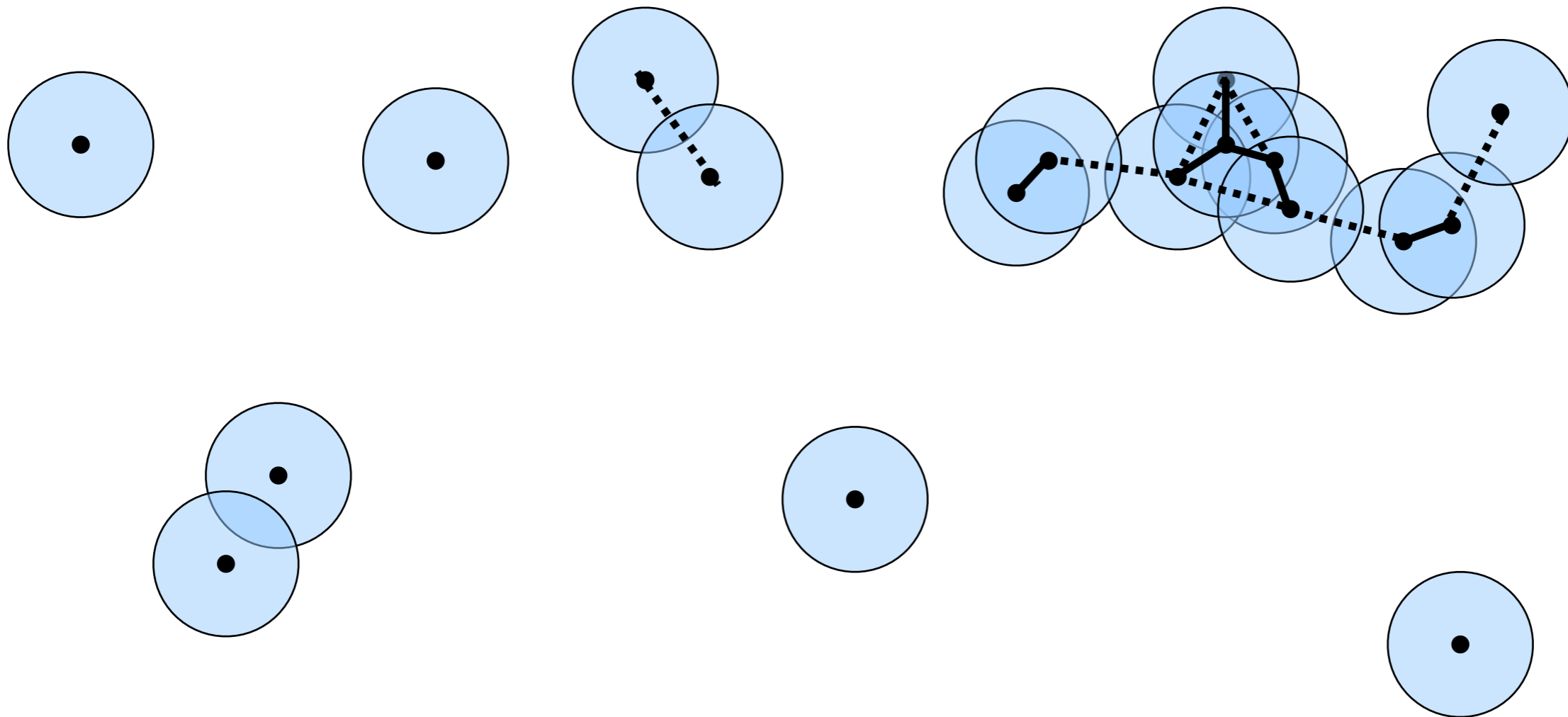
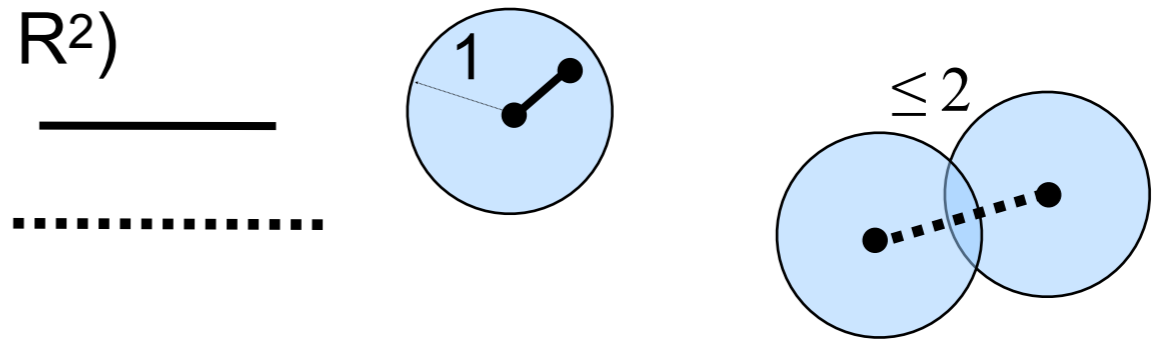


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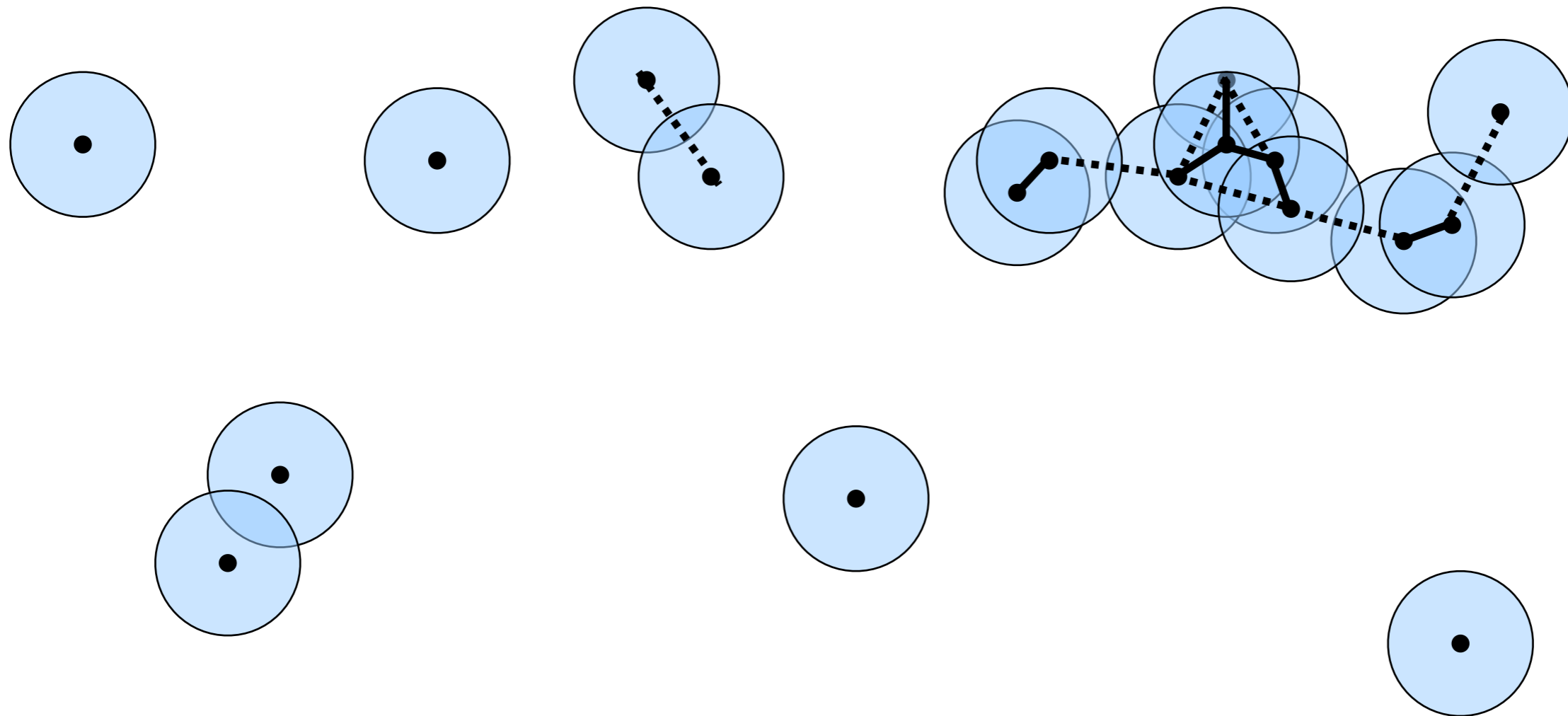
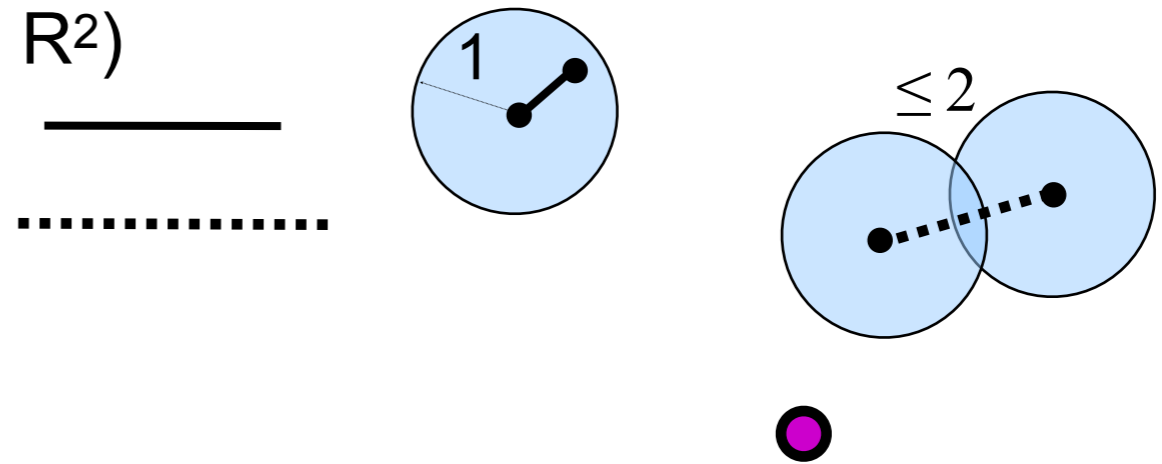


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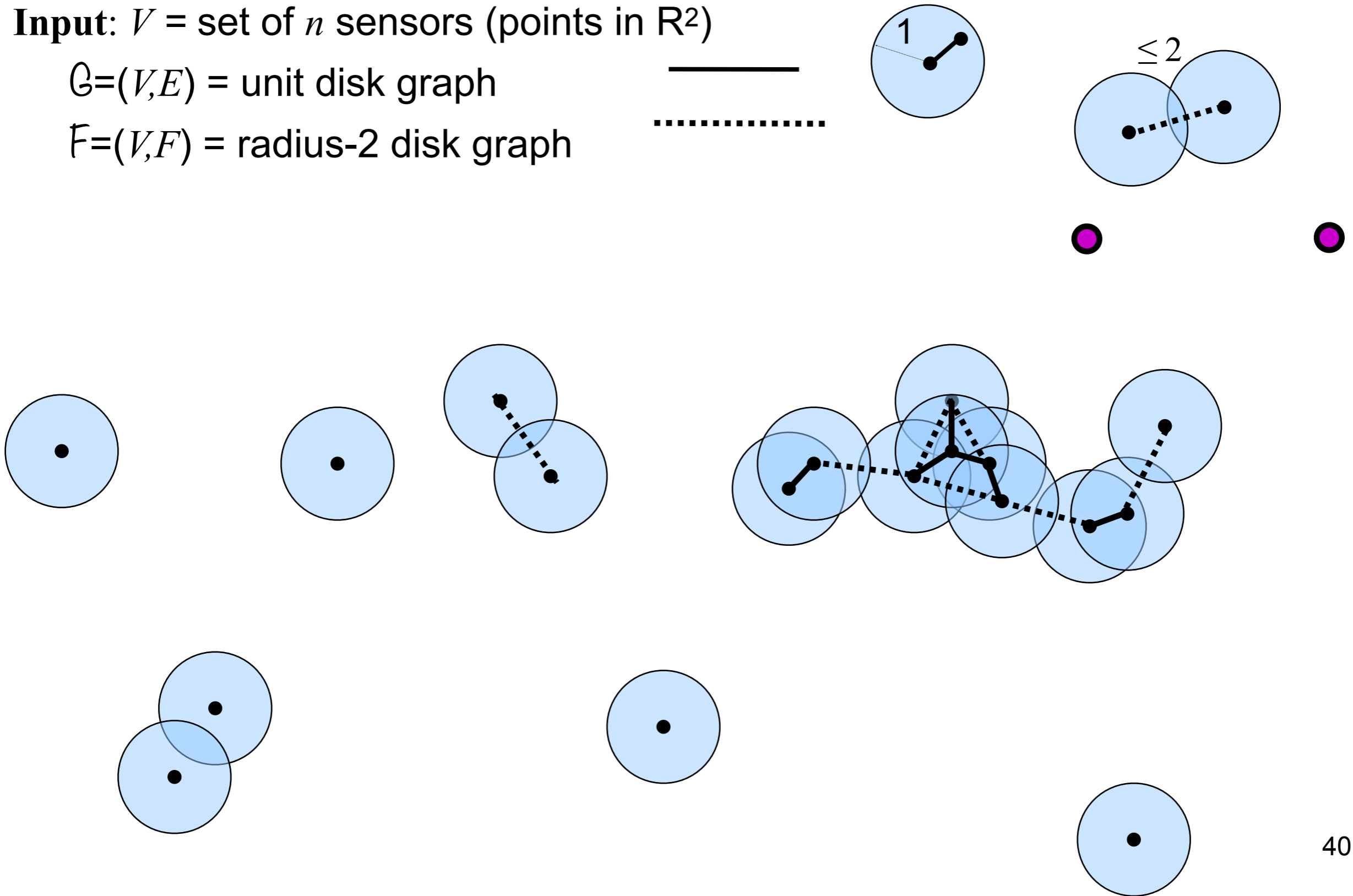


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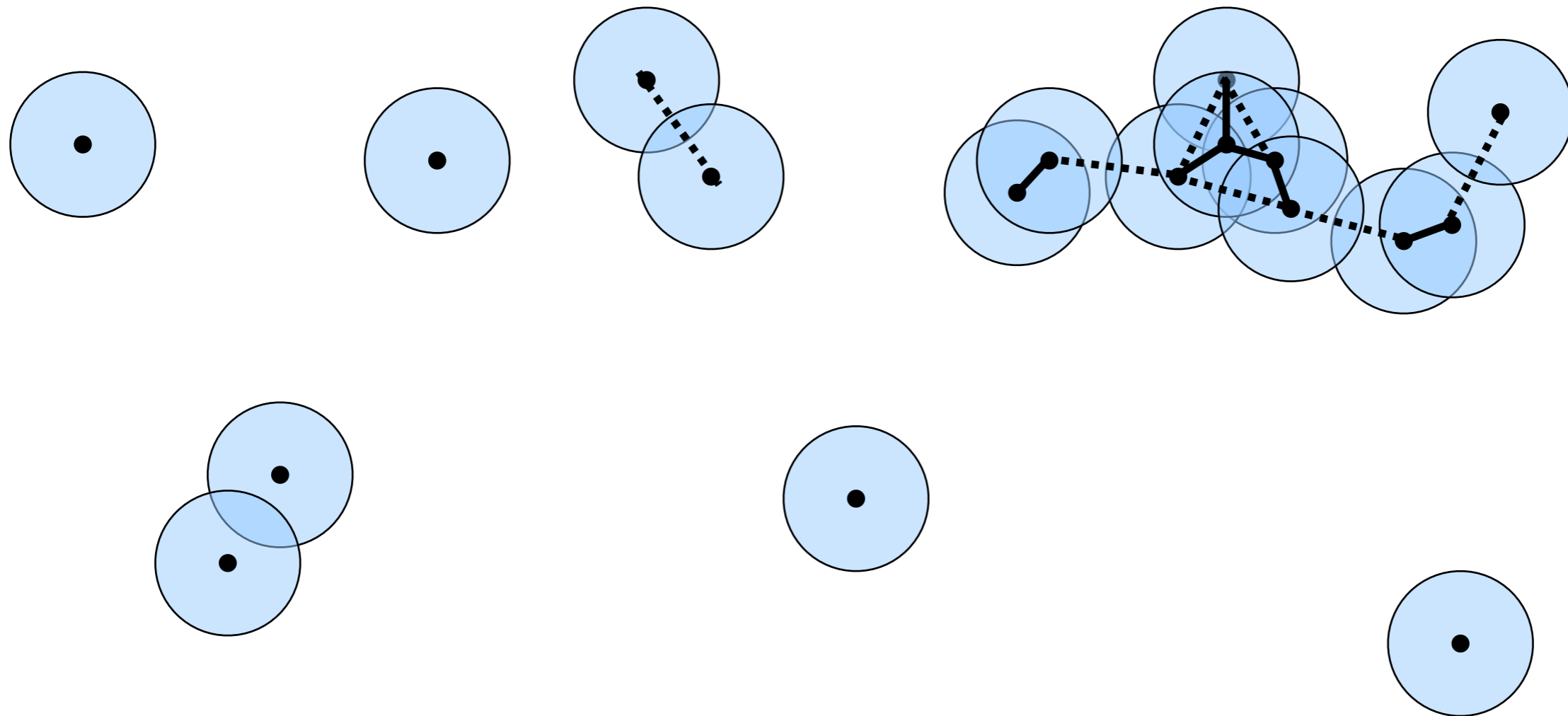
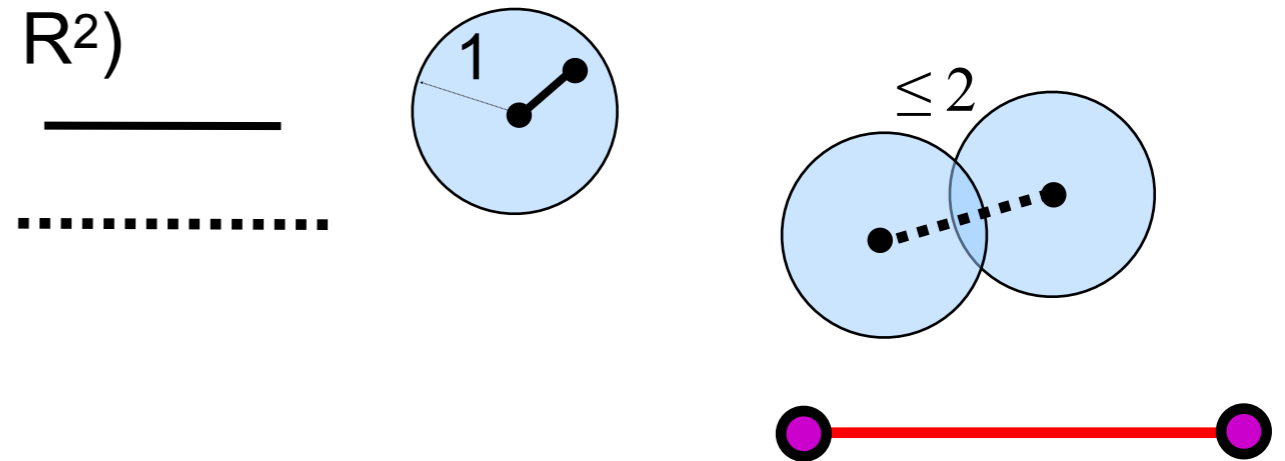


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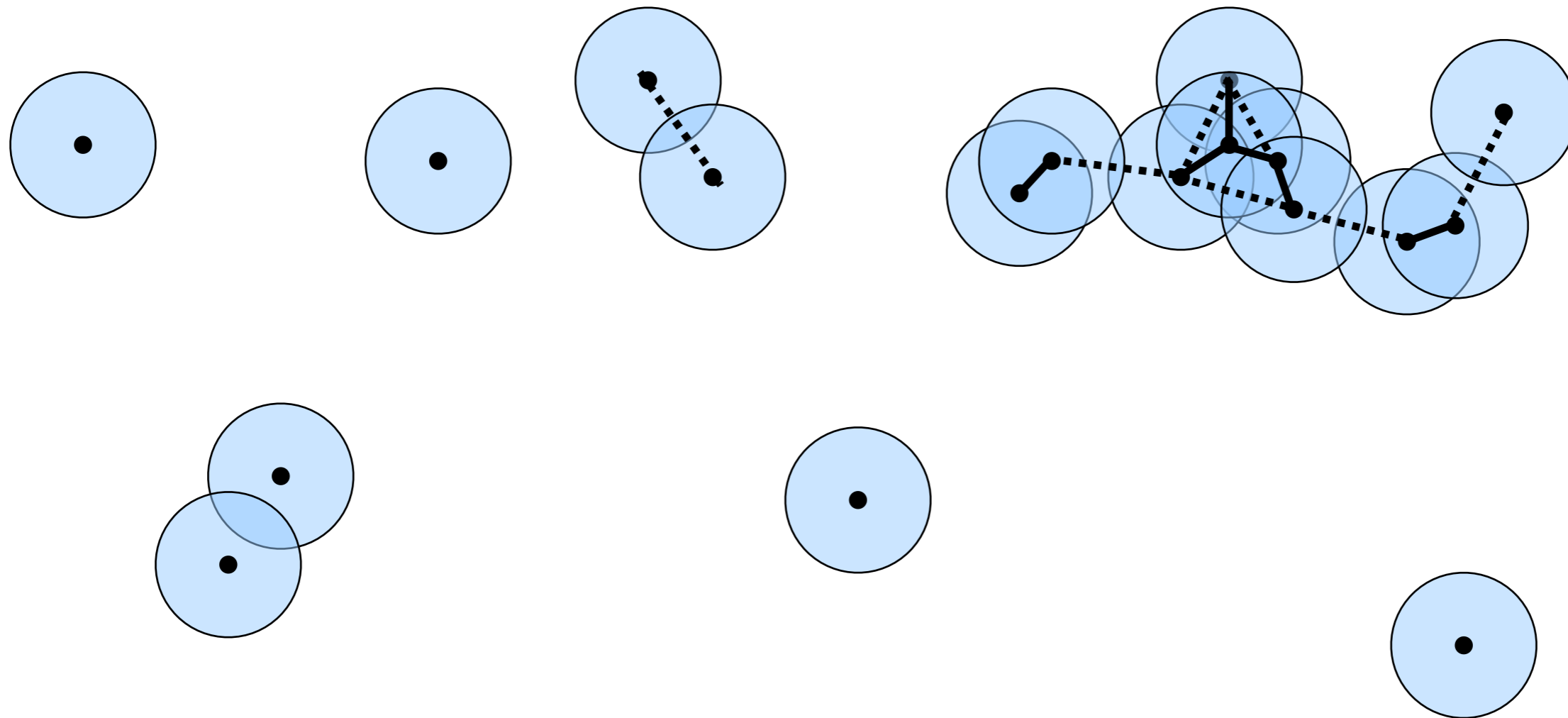
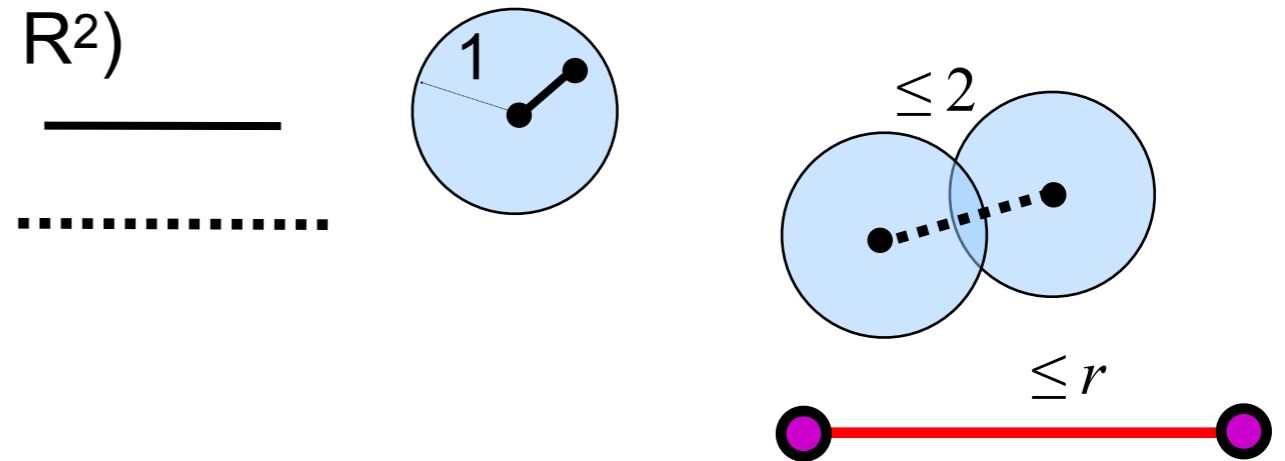


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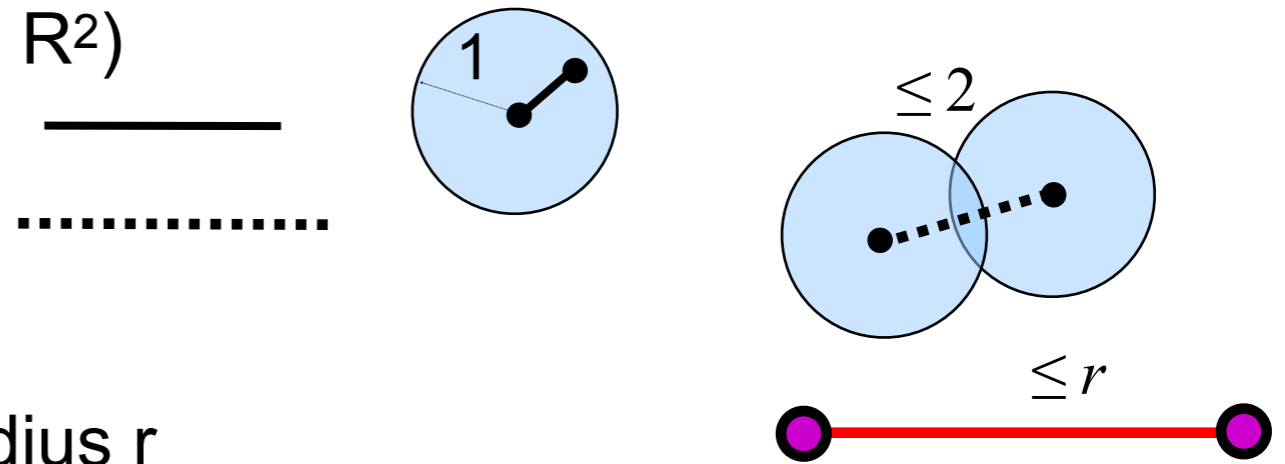


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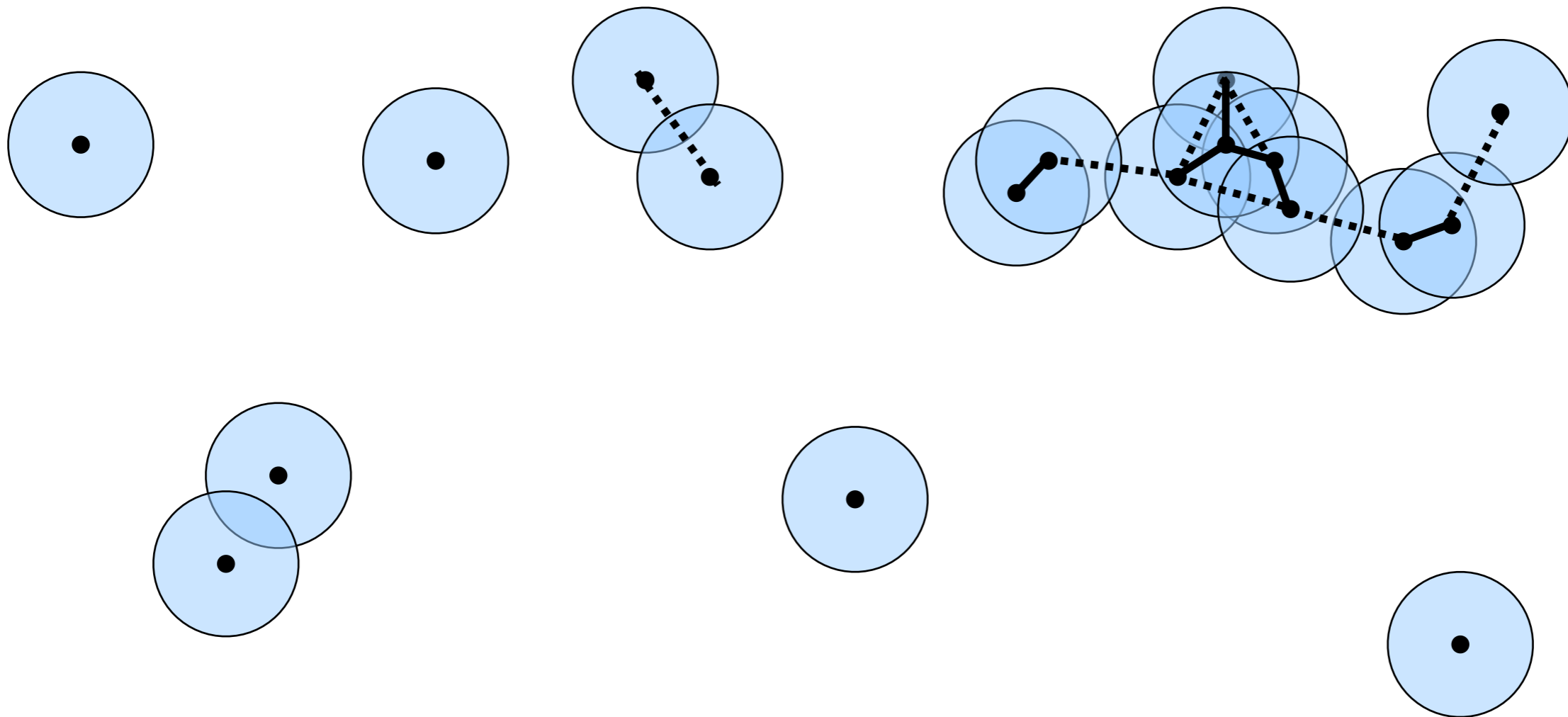
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**Add:** relays with communication radius  $r$

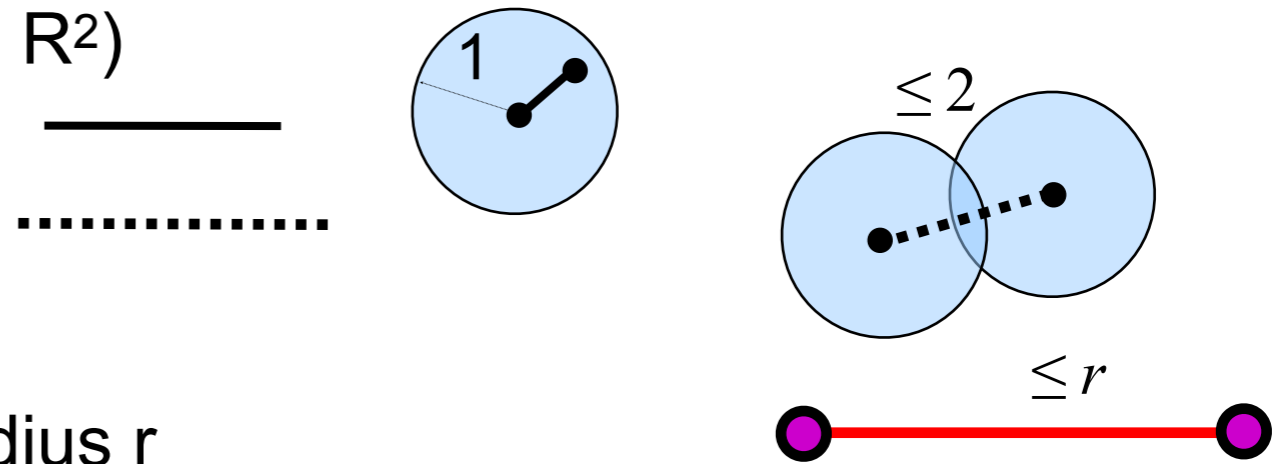


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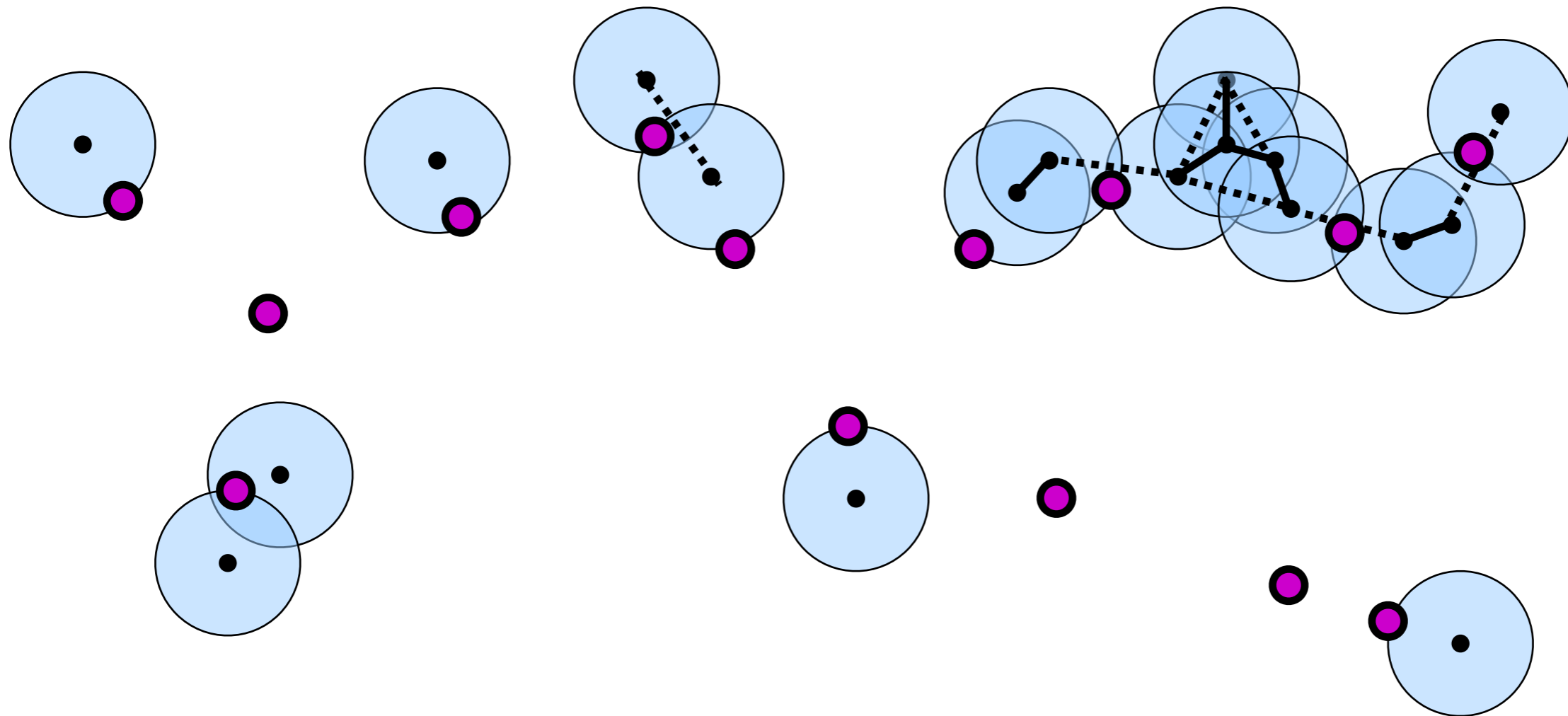
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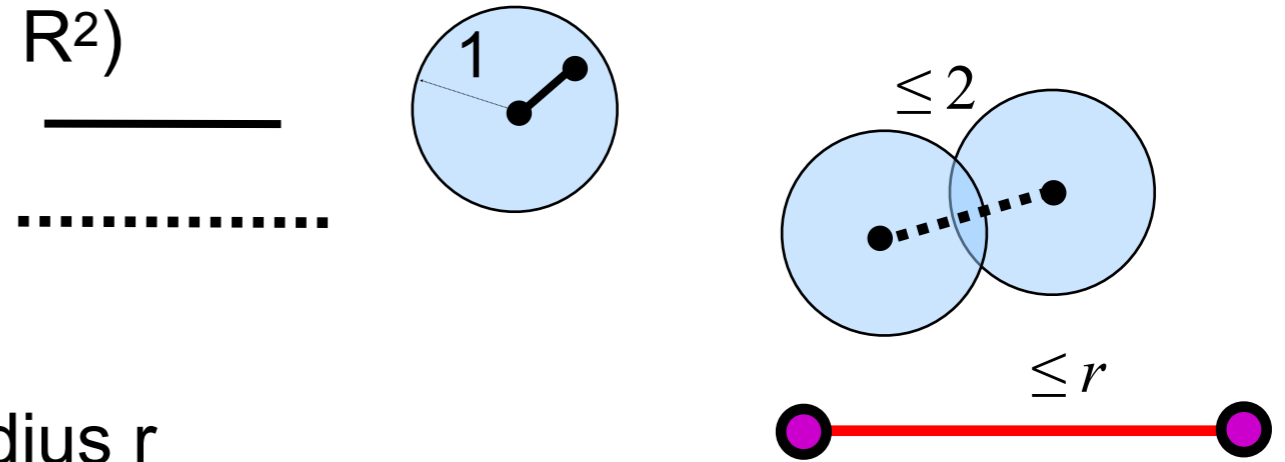


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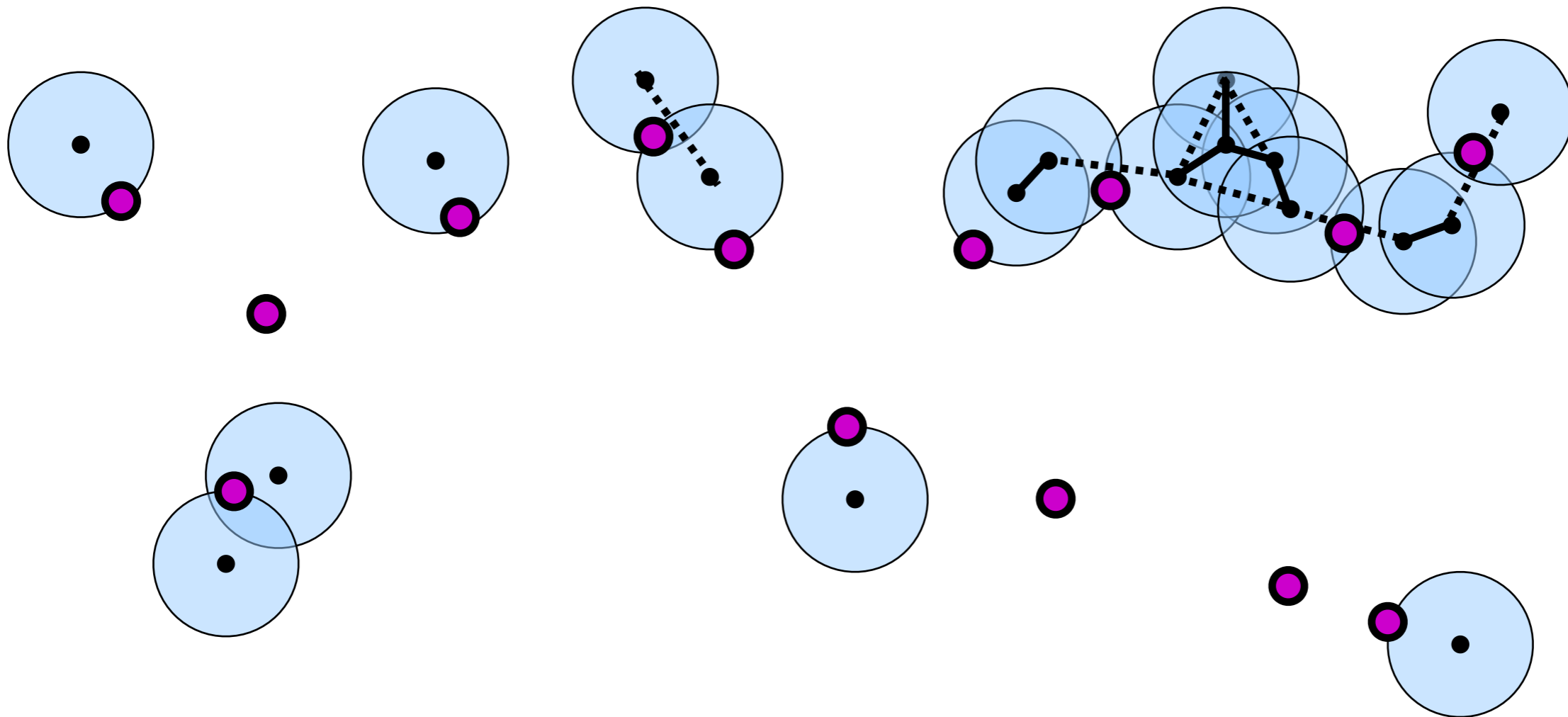
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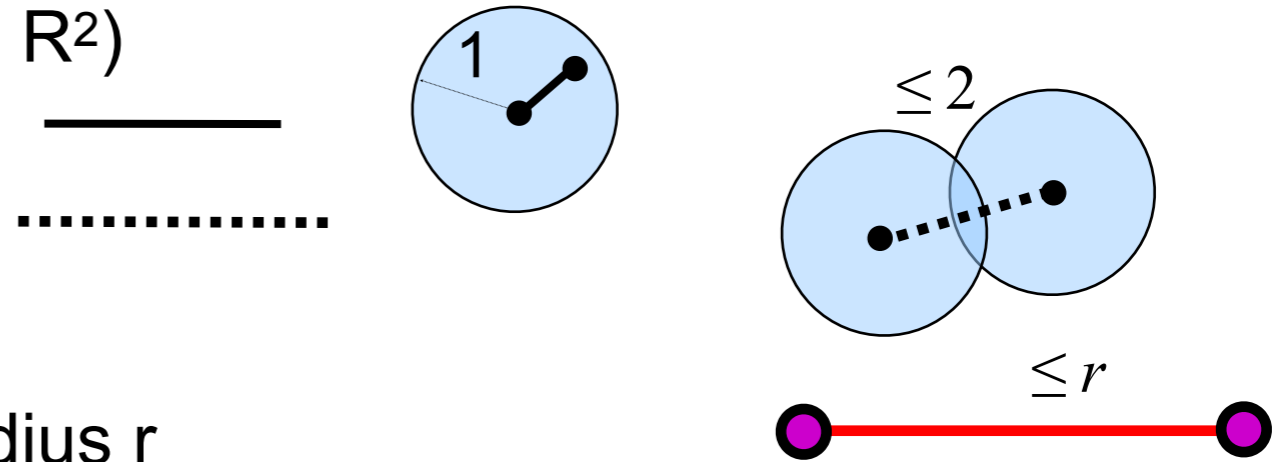


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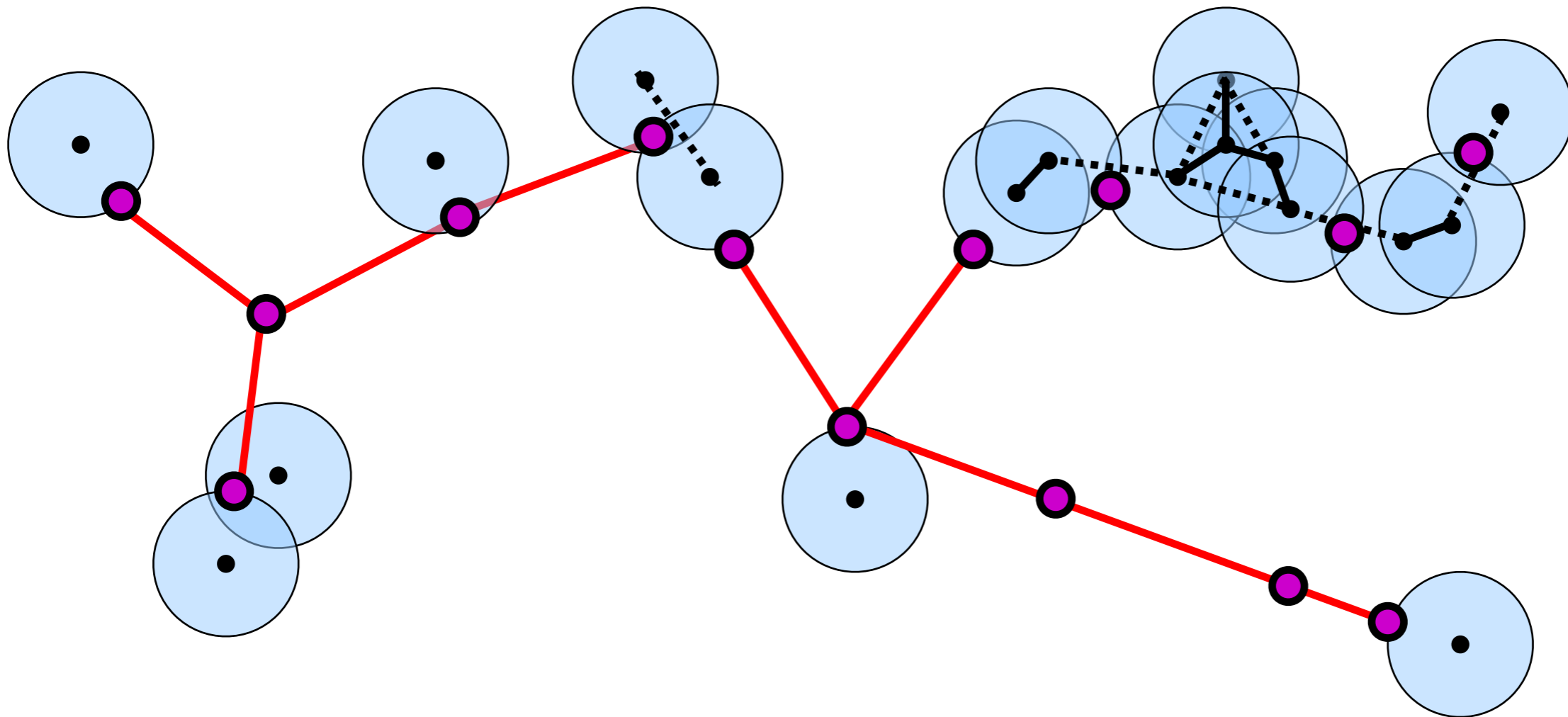
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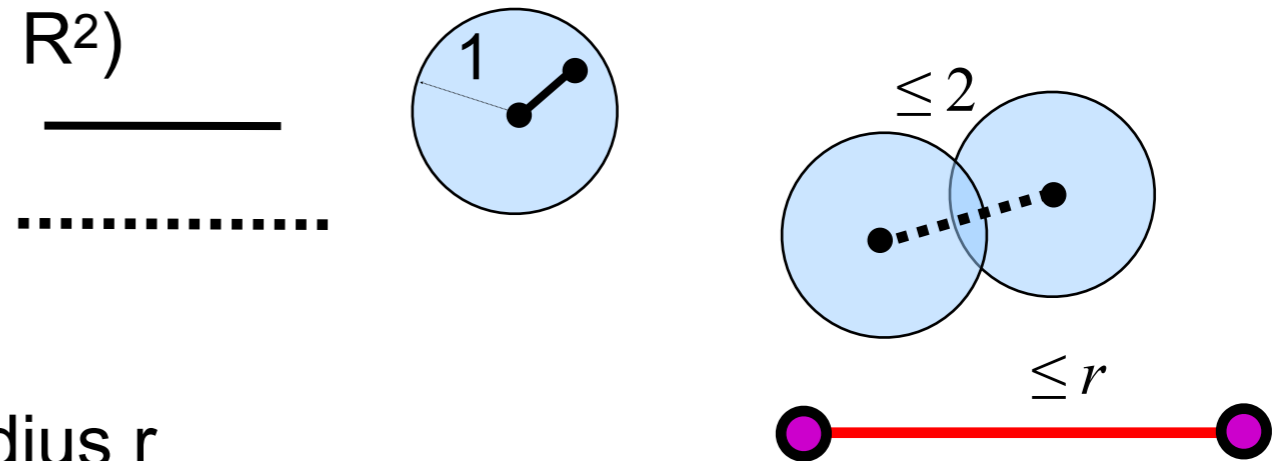


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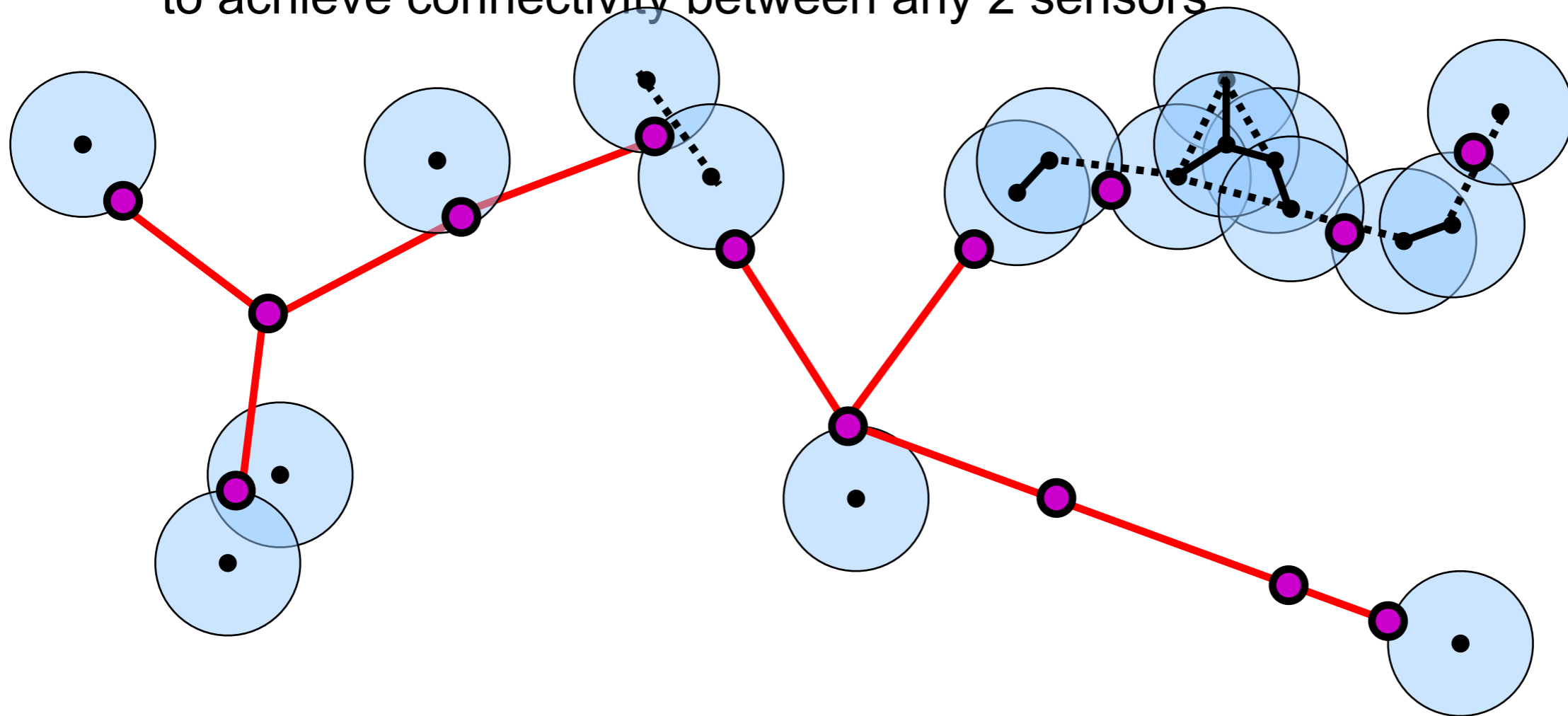
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to achieve connectivity between any 2 sensors



# Model of Communication

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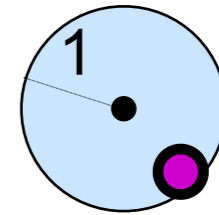
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- Sensor and relay: communicate if  $\text{dist} \leq 1$

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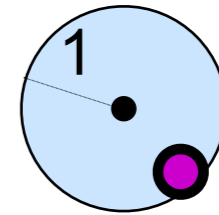
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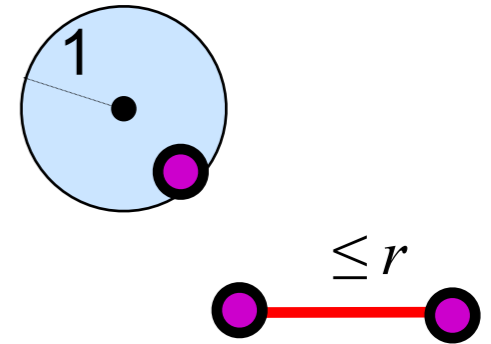
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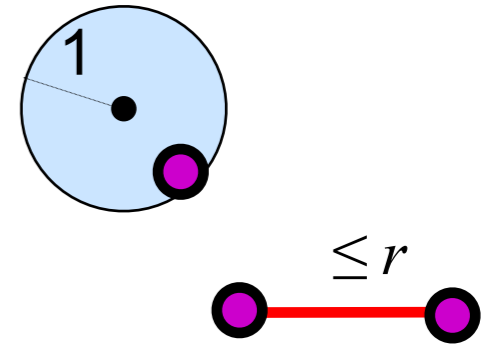
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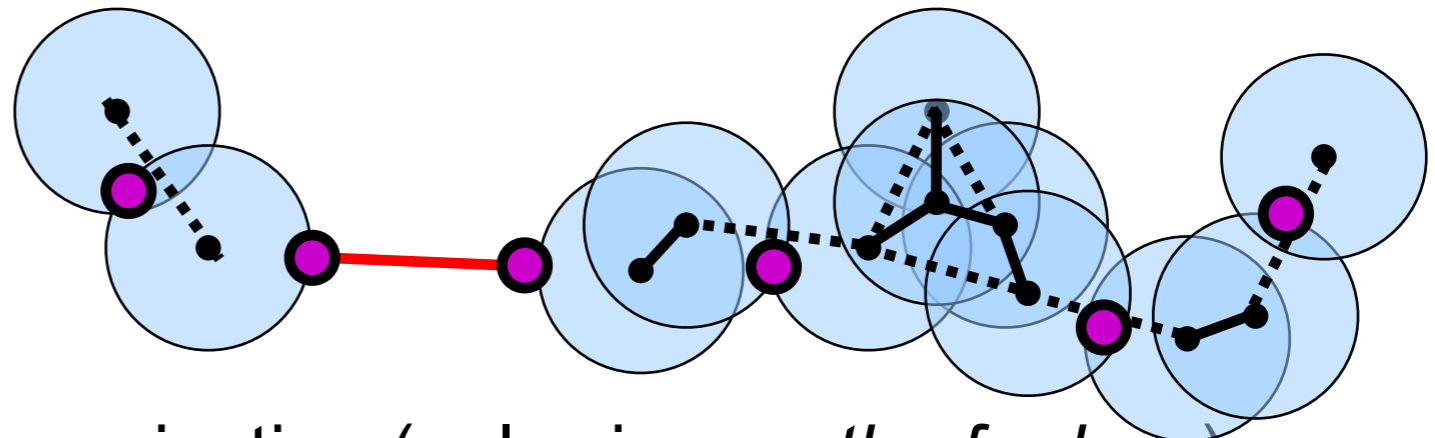
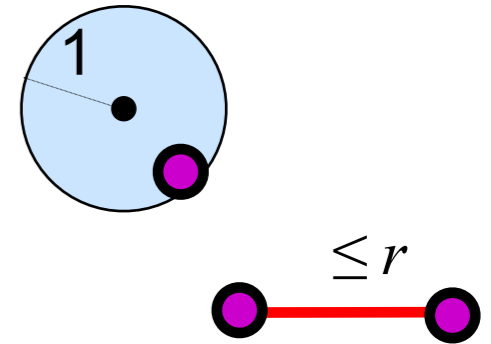
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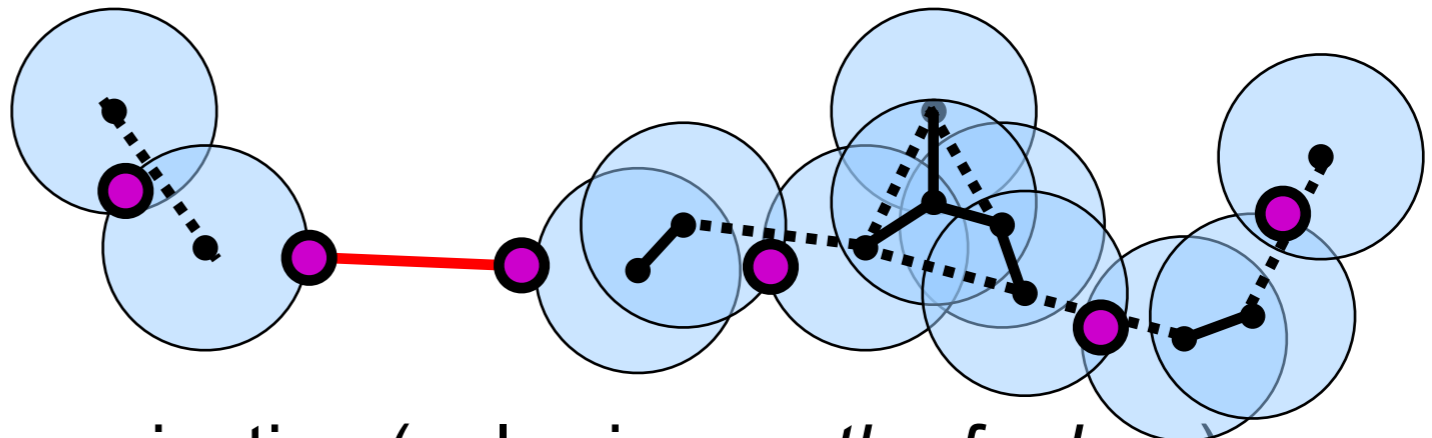
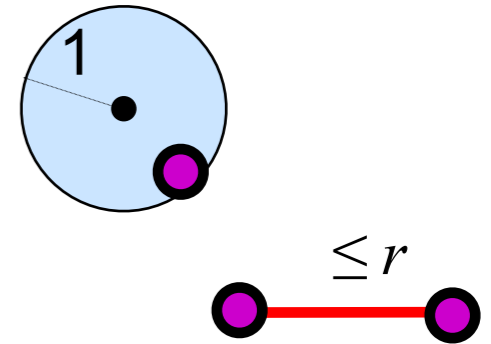


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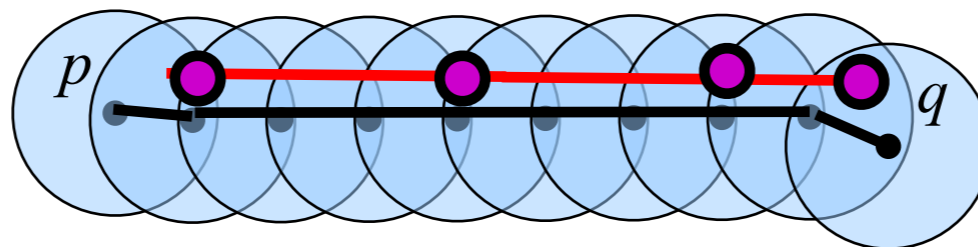


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- **$(4+\varepsilon)$ -approx for two-tier, any  $r \geq 2$**



# Prior Related Work

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- **Steiner trees with min # Steiner points, edge length  $\leq 1$**   
(same as one-tier,  $r=1$ ) :
  - NP-hard, 5-approx [Lin, Xue '99]
  - [Lin,Xue] is actually a 4-approx; also give 3-approx [Chen et al '00]
  - Faster 3-approx, randomized 2.5-approx [Cheng et al '07]
- **7-approx for one-tier, arbitrary  $r$**  (time  $O(n \log n)$ )  
[Lloyd, Xue '07]
- **$(5+\varepsilon)$ -approx for two-tier, any  $r \geq 1$**  [Lloyd, Xue '07]
- **$(4+\varepsilon)$ -approx for two-tier, any  $r \geq 2$**   
[Srinivas, Zussman, Modiano '06]

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# Terminology: Blobs and Clouds

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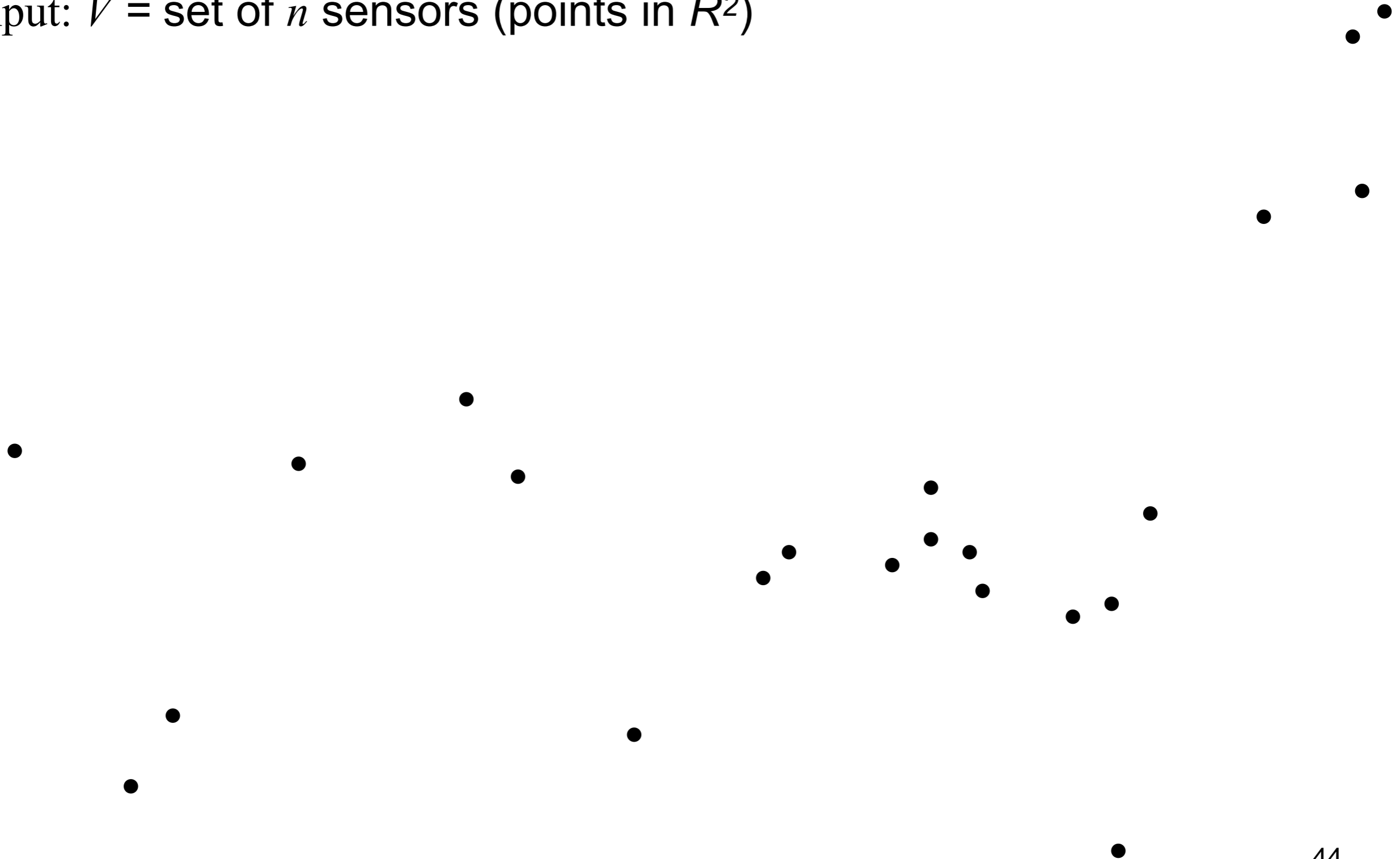
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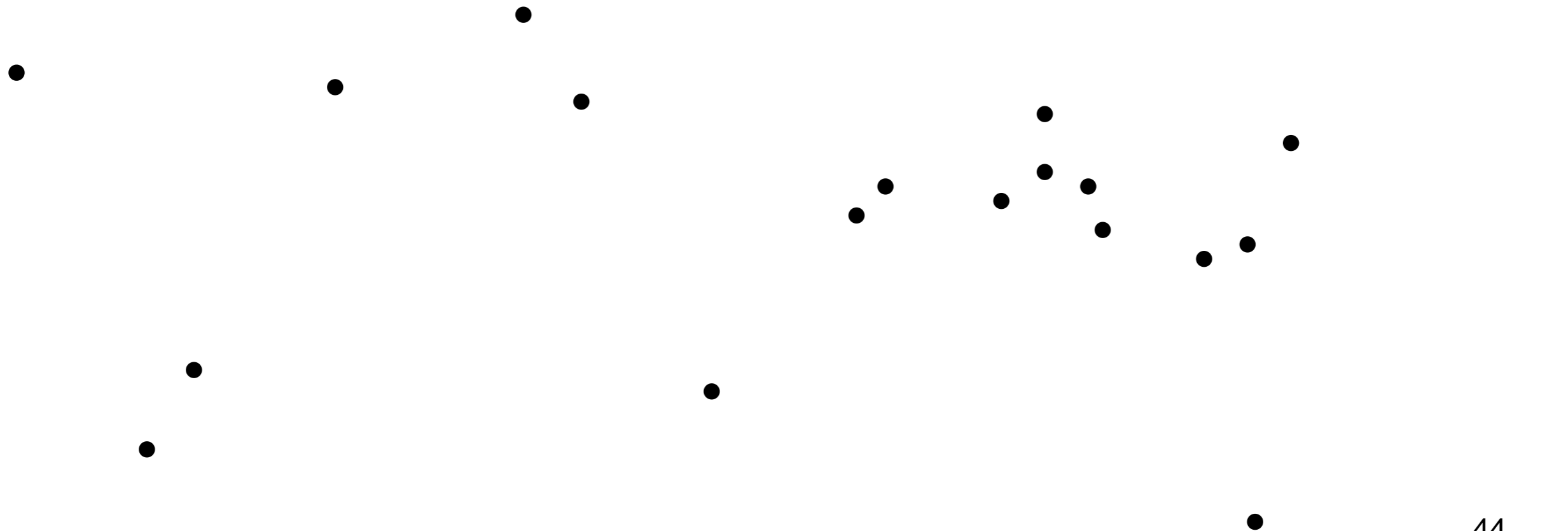


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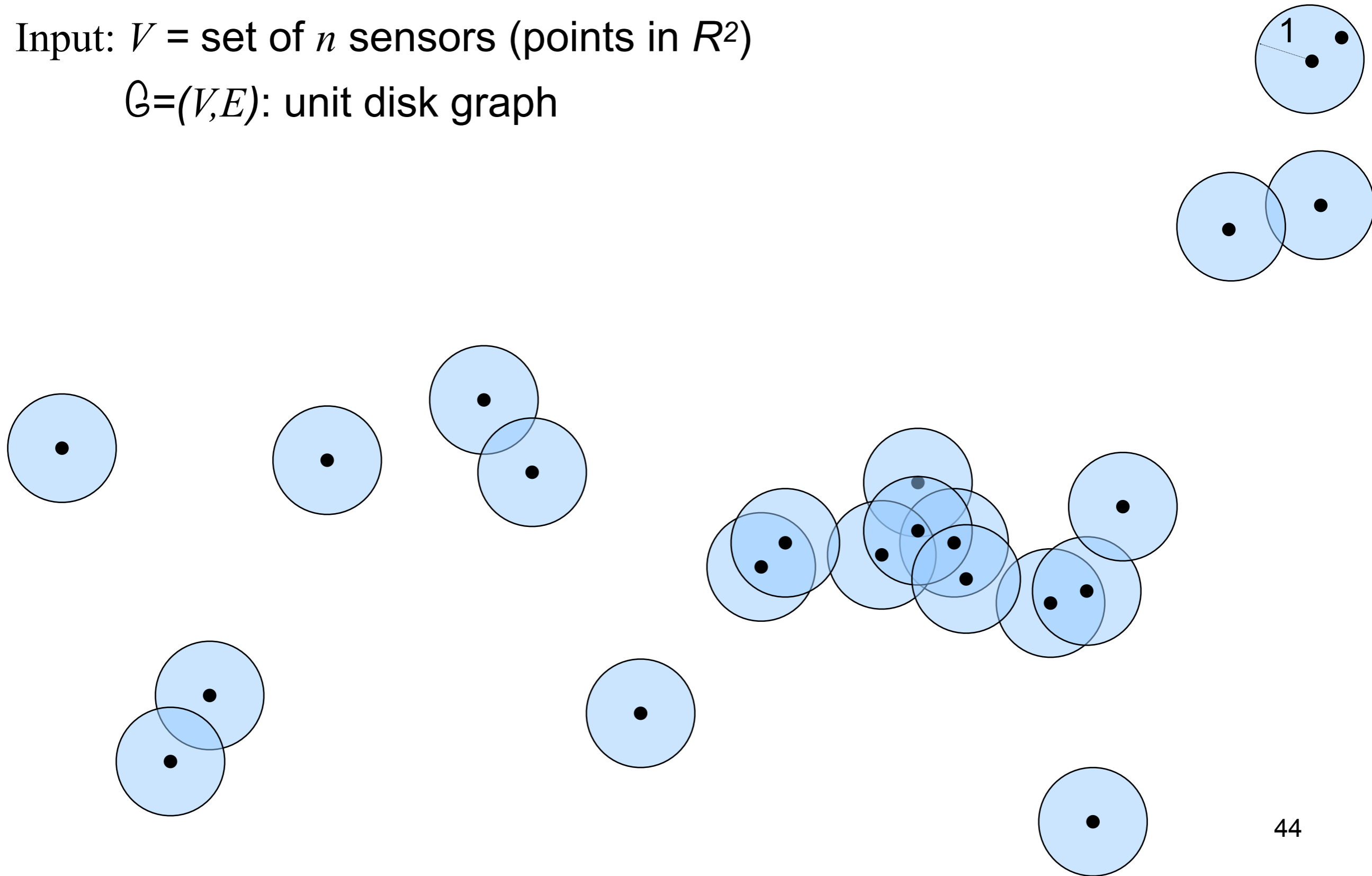


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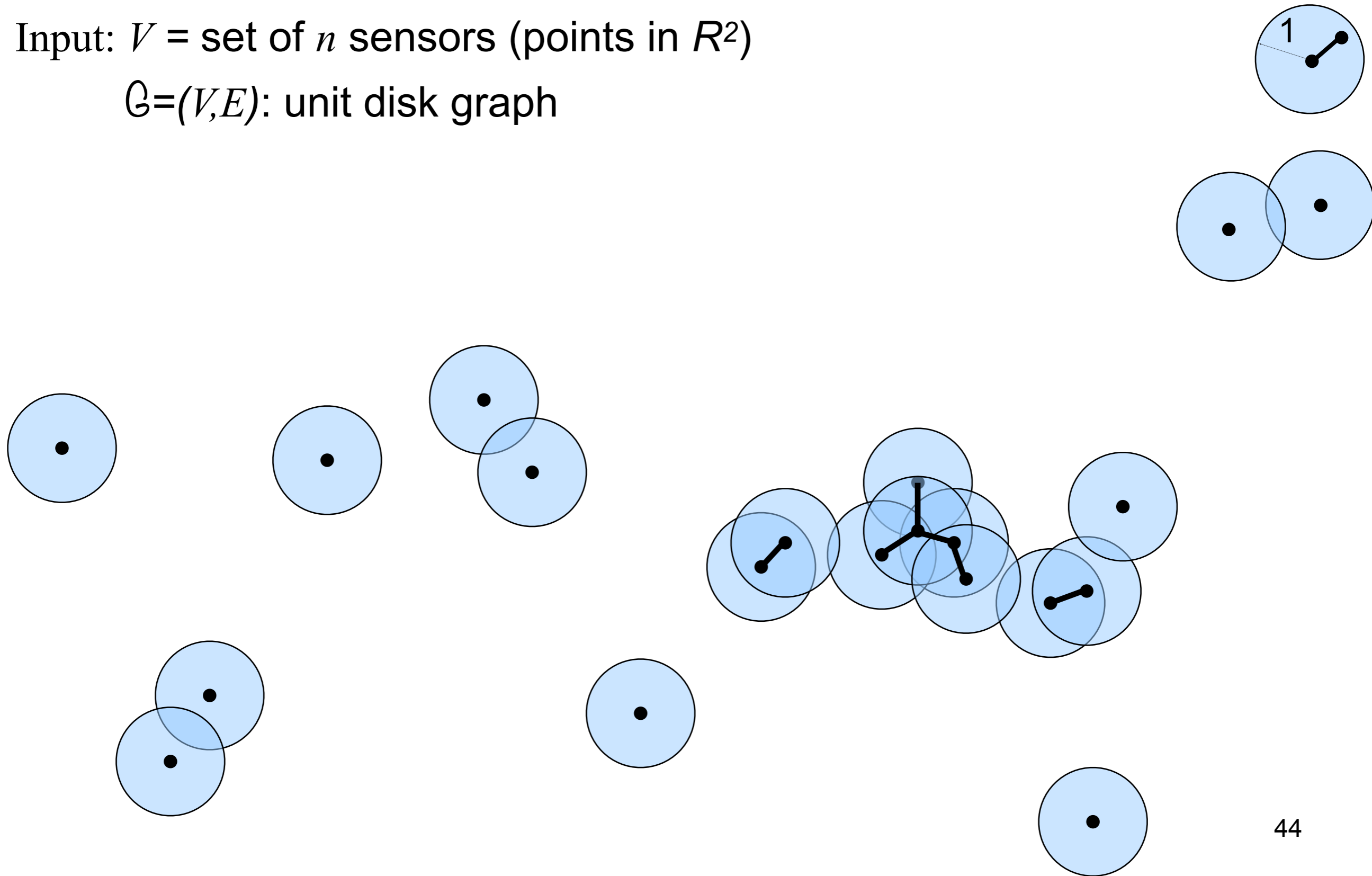


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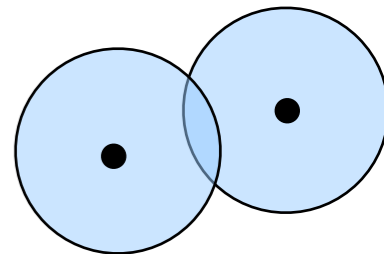
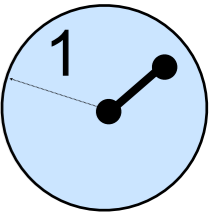
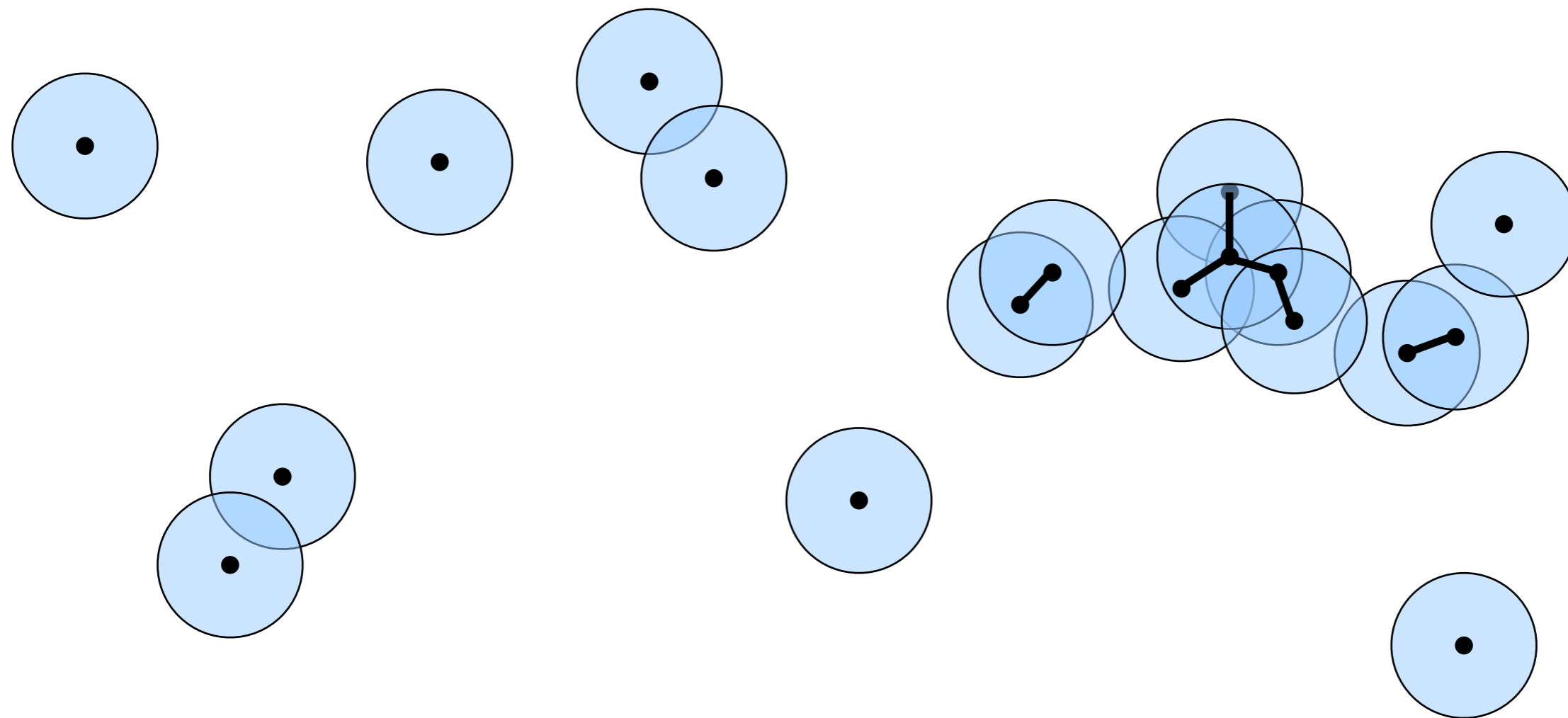
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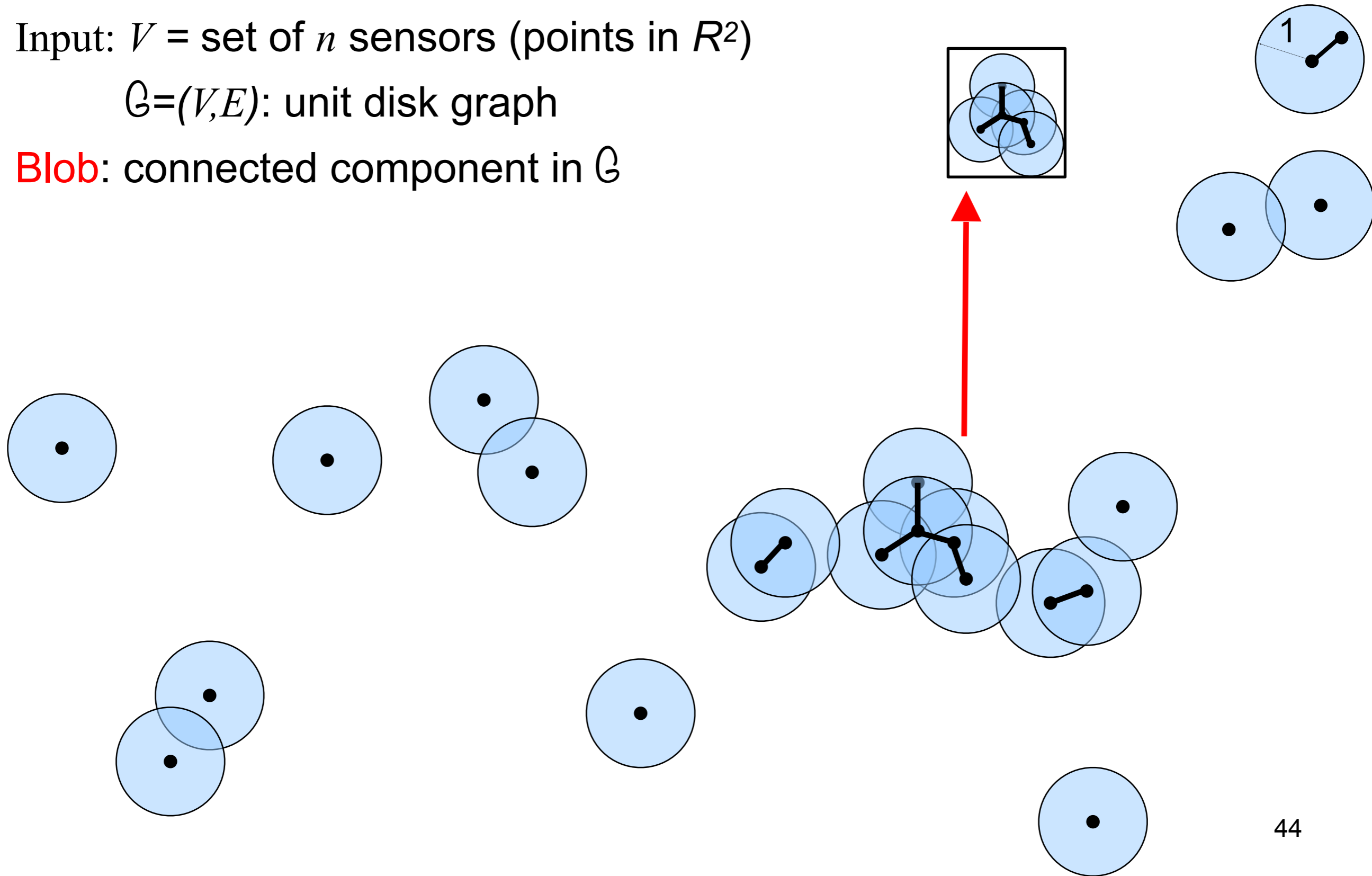


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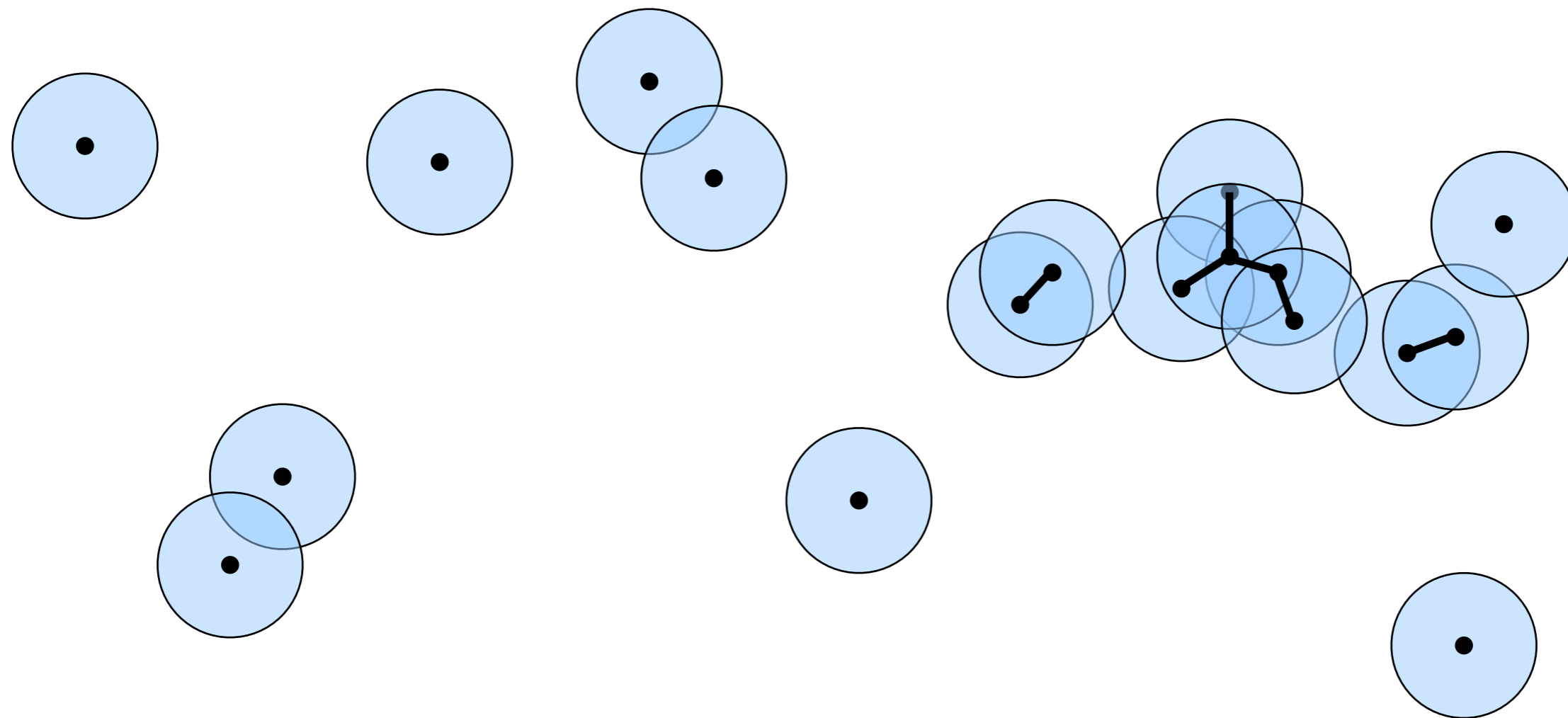
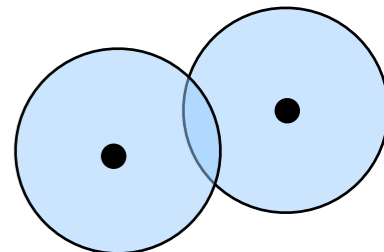
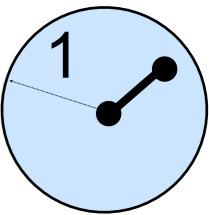
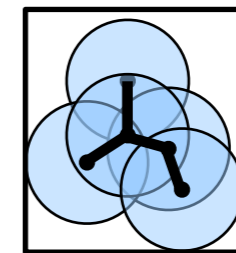


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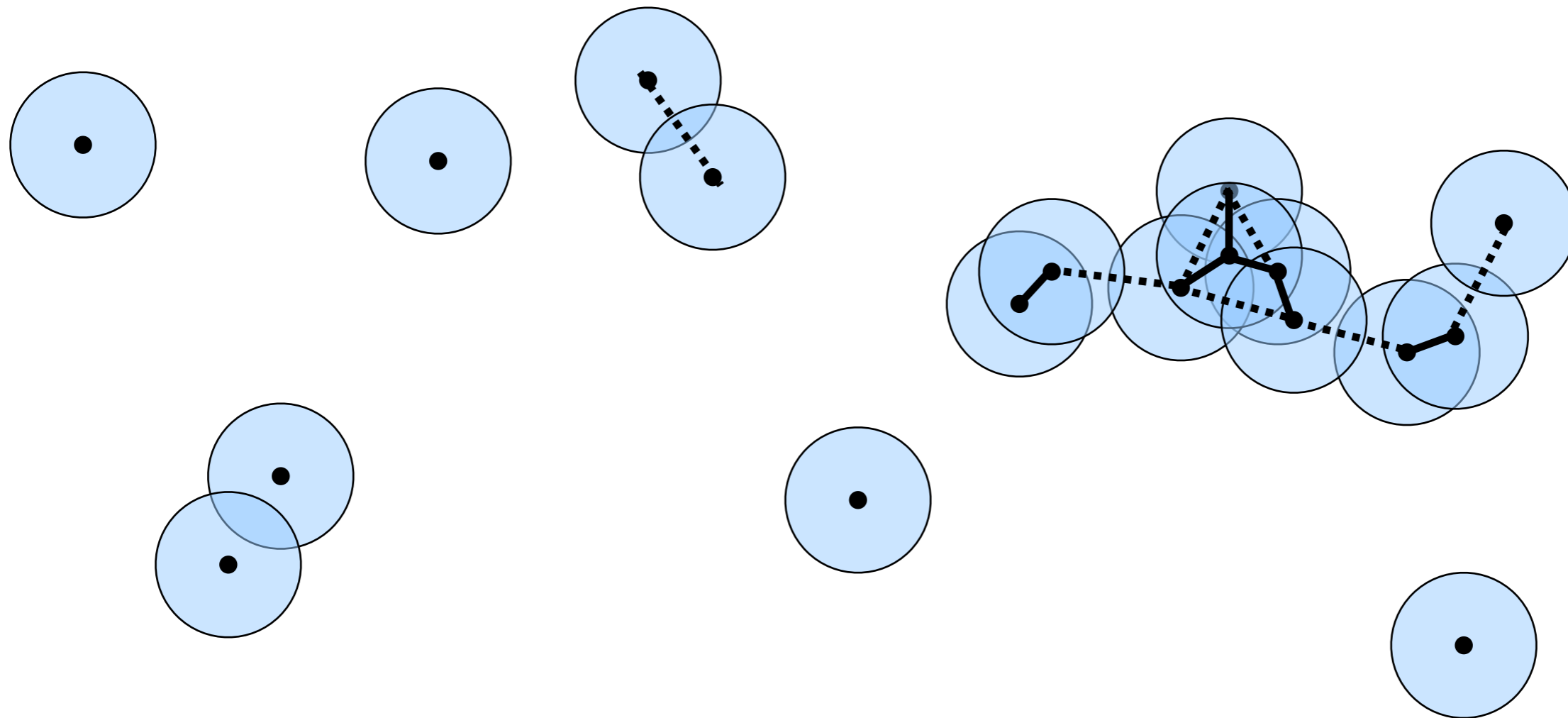
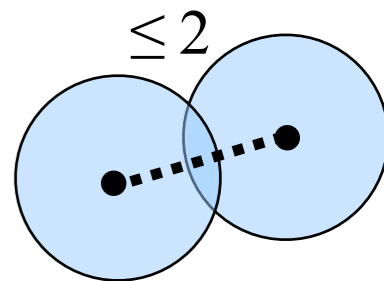
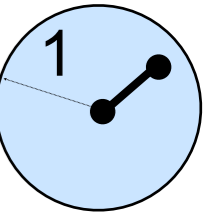
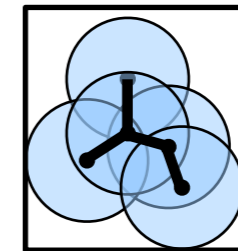
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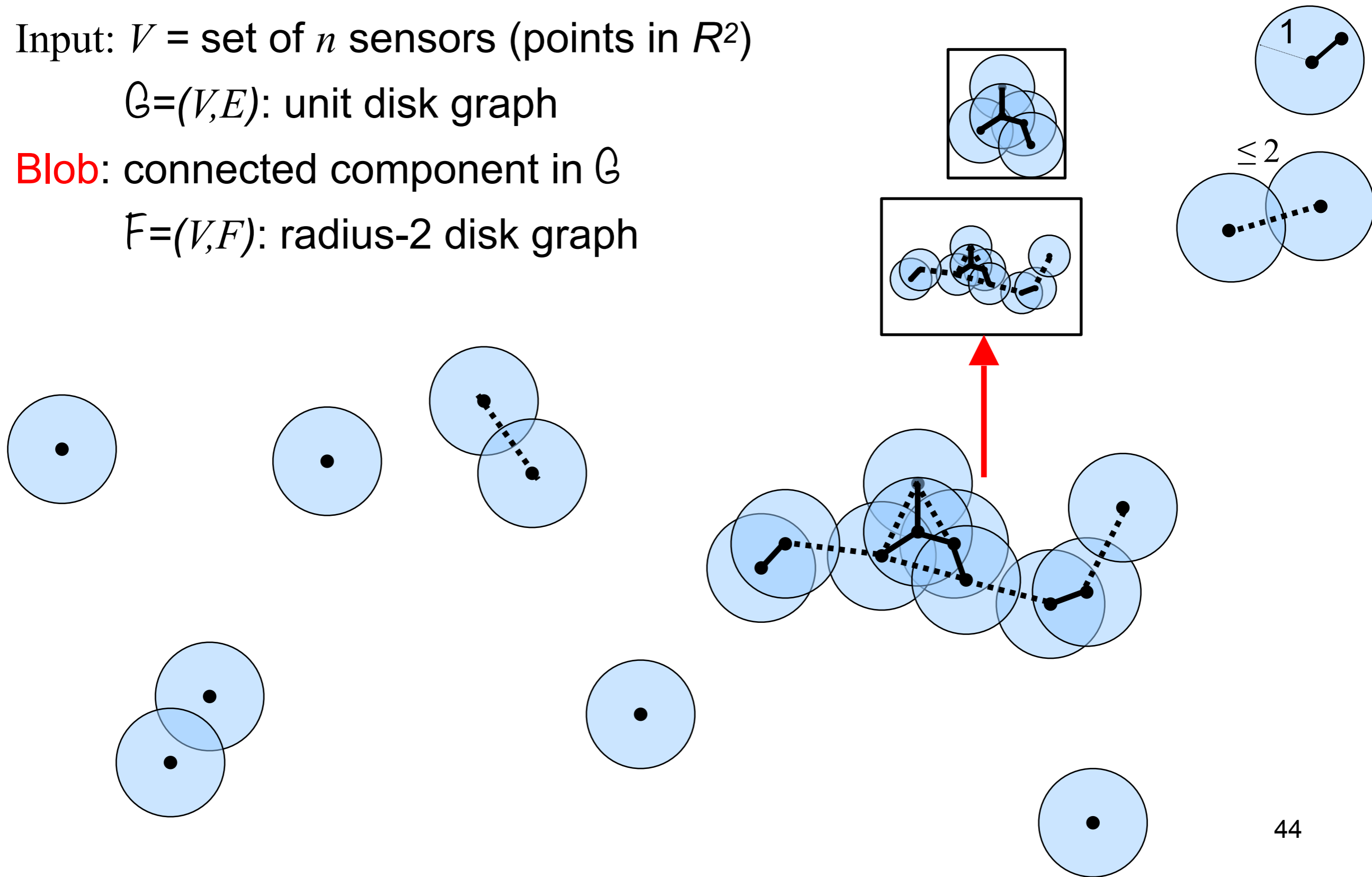
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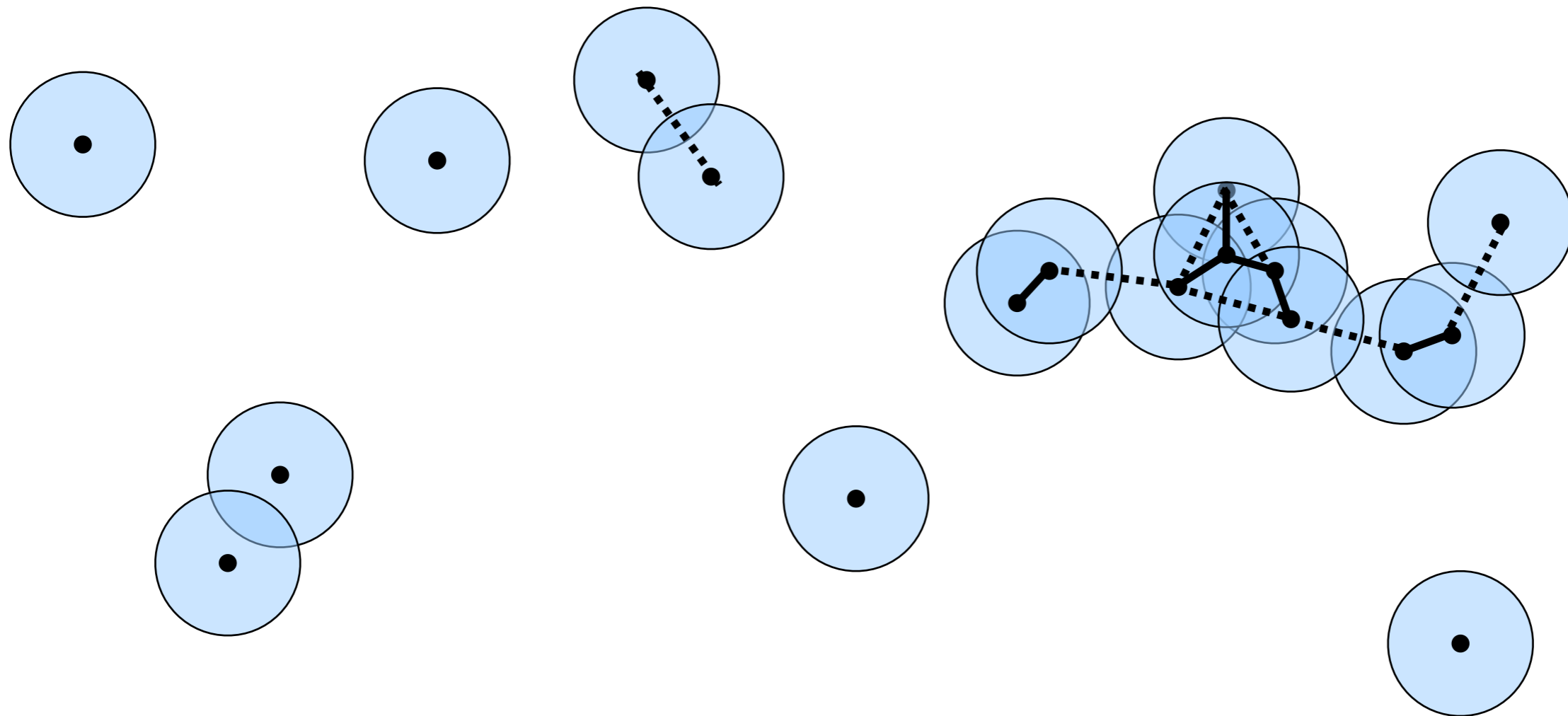
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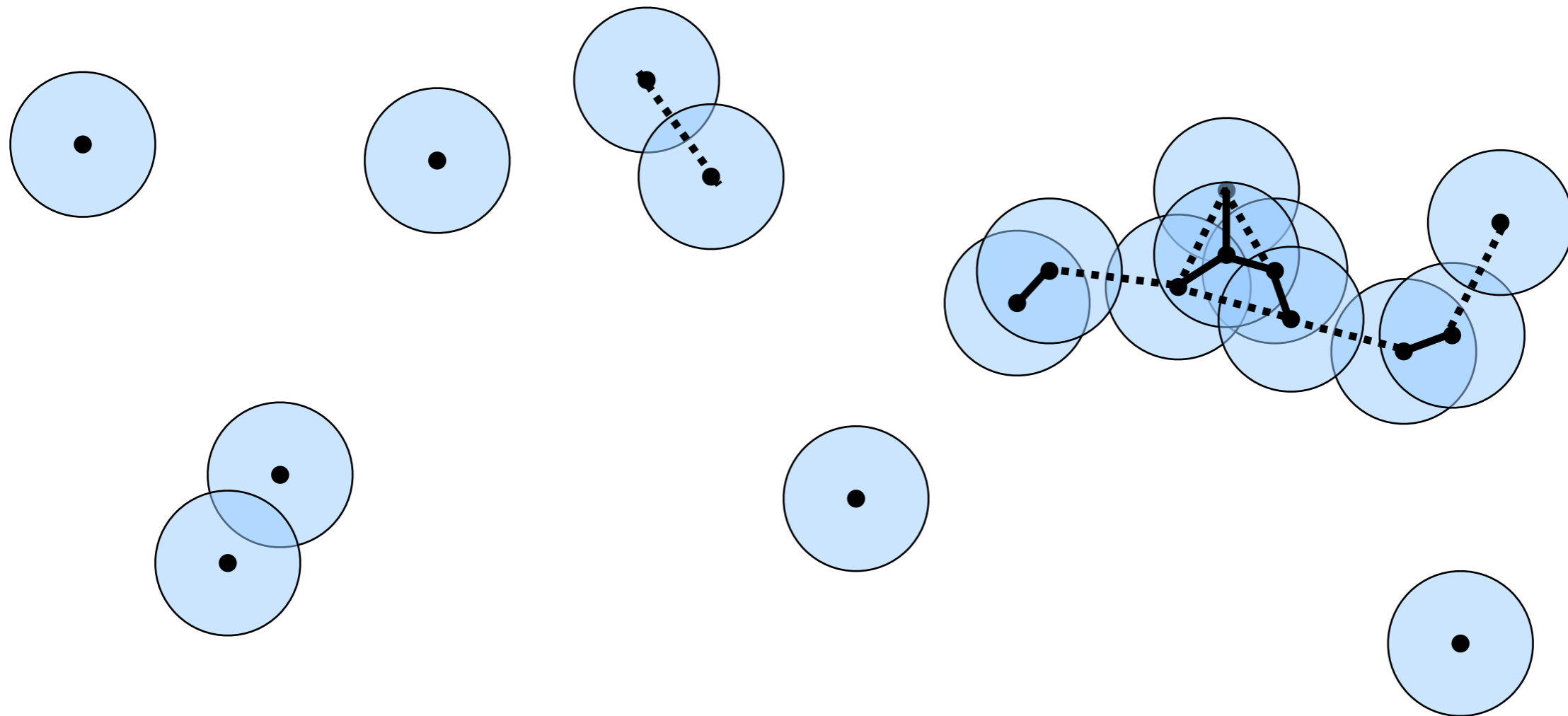
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**Cloud**: connected component in  $\mathcal{F}$



# Lemma 1

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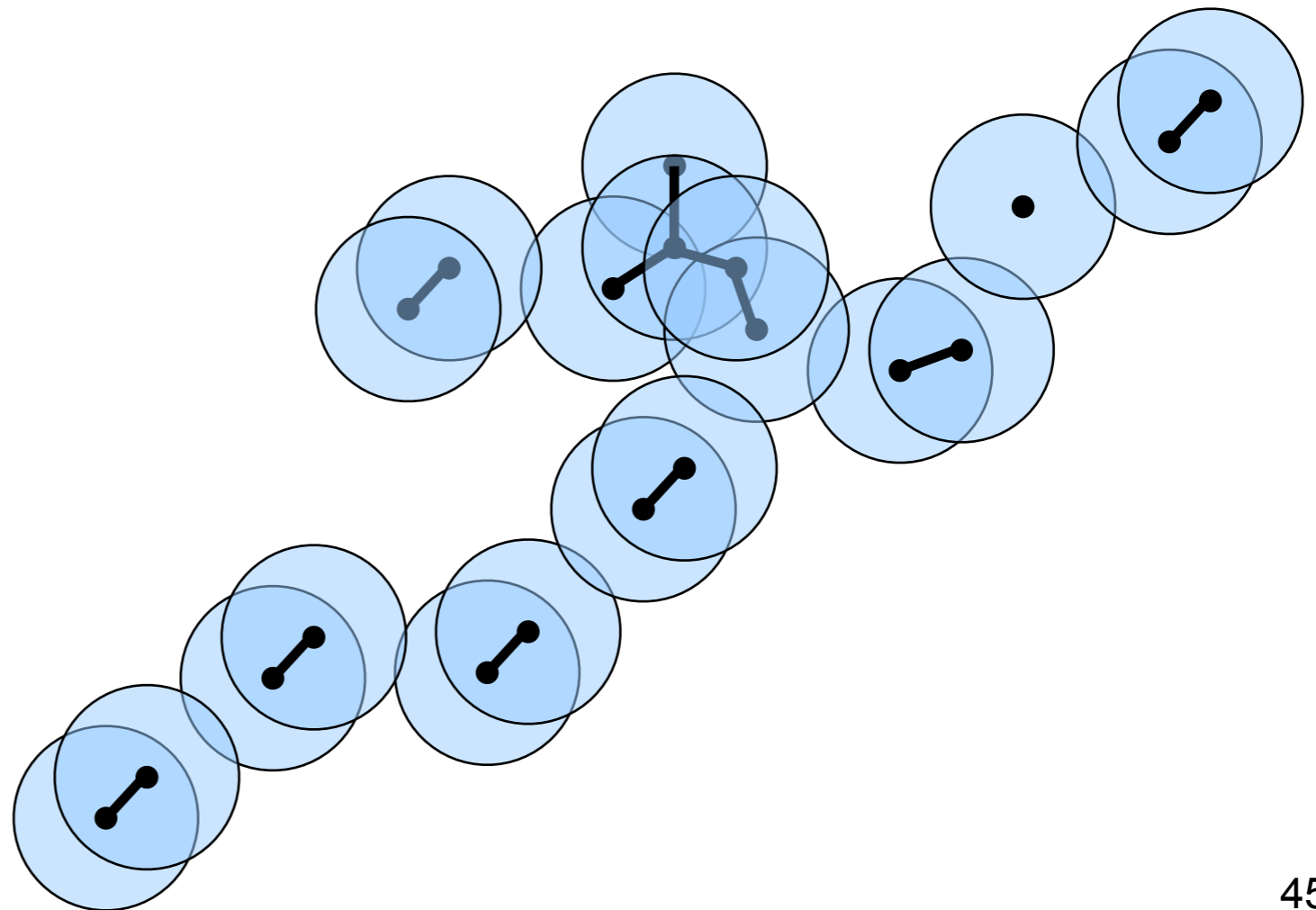
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**Lemma 1:** To connect all sensors within a cloud  $C$ ,

# Lemma 1

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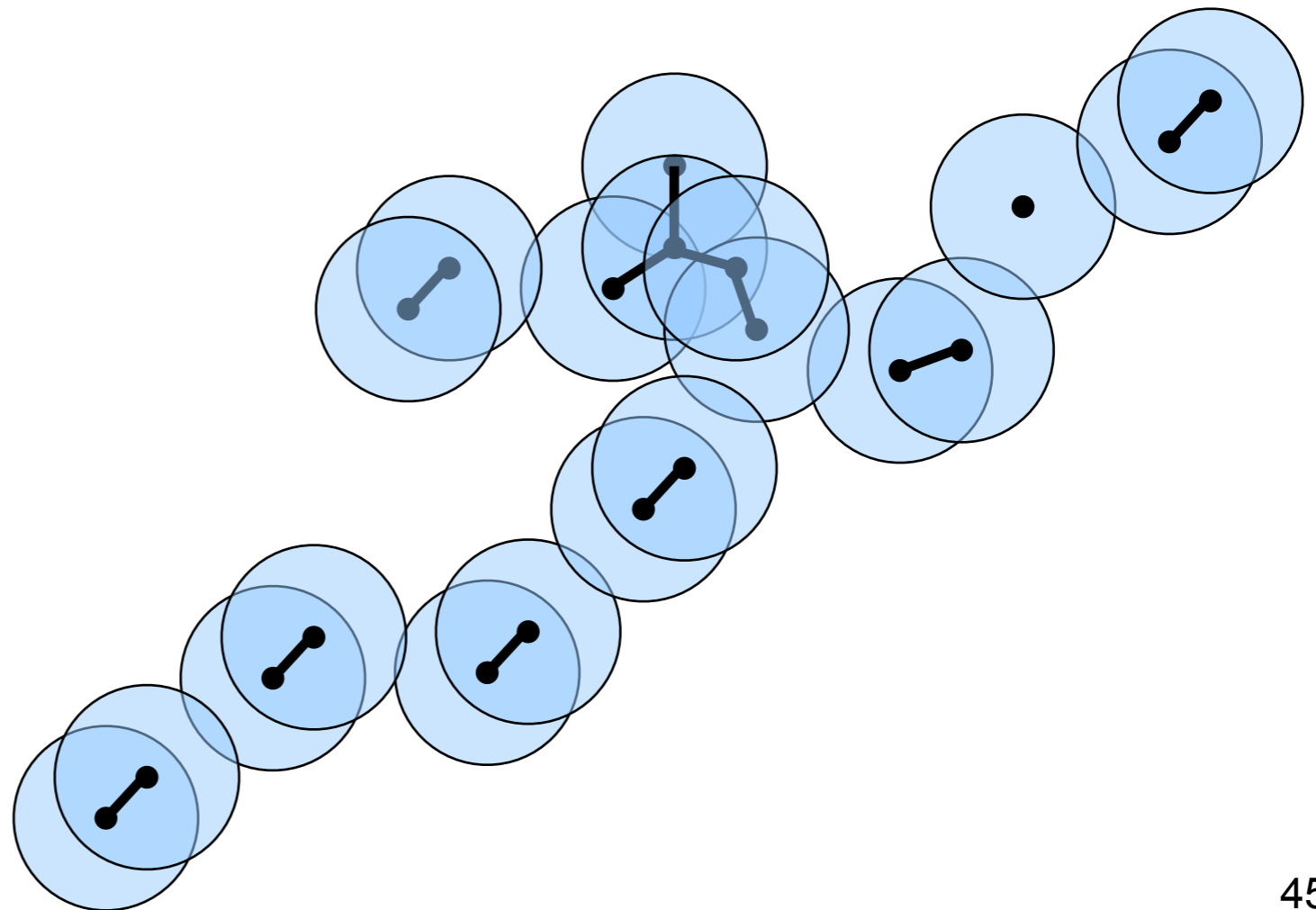
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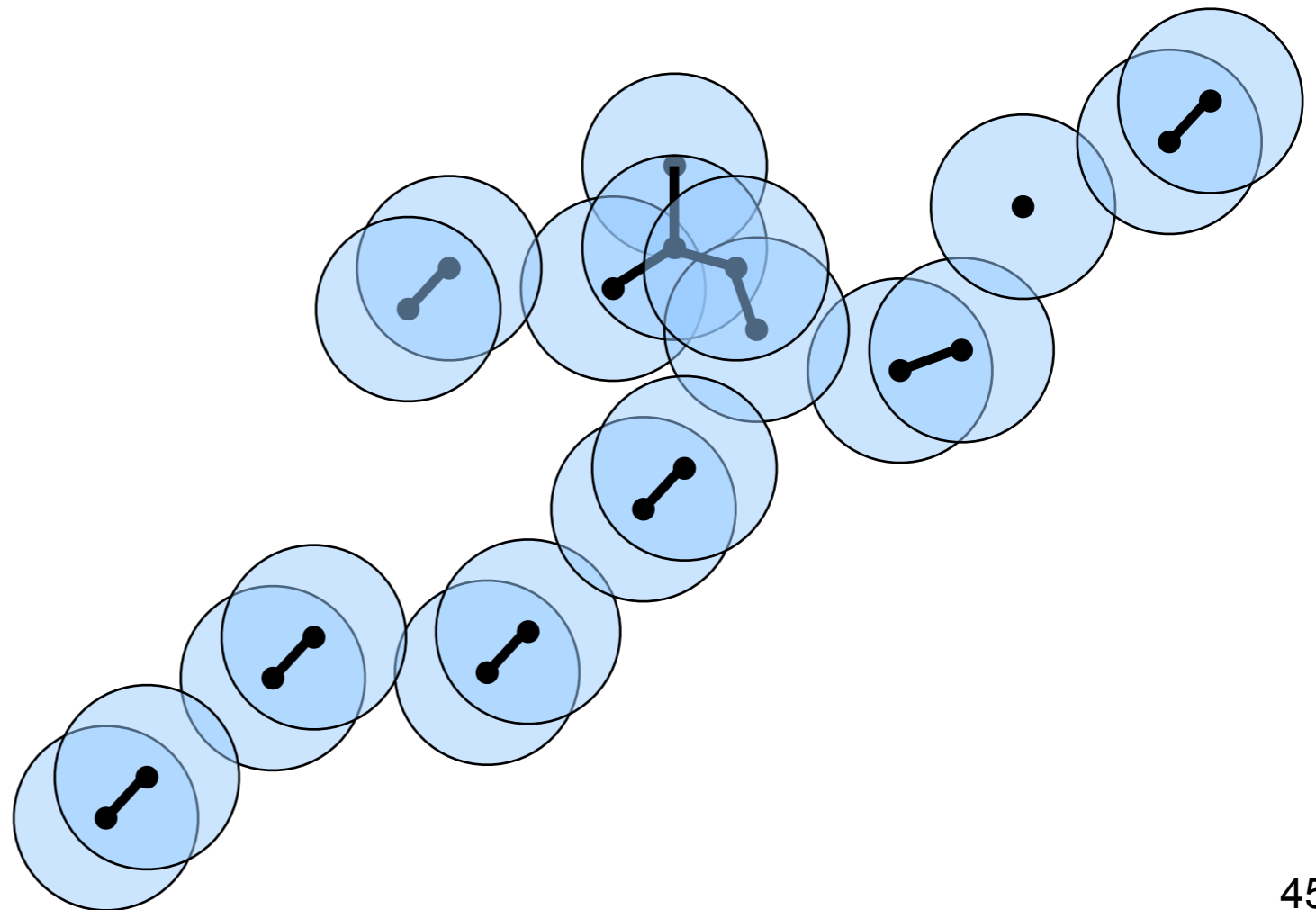


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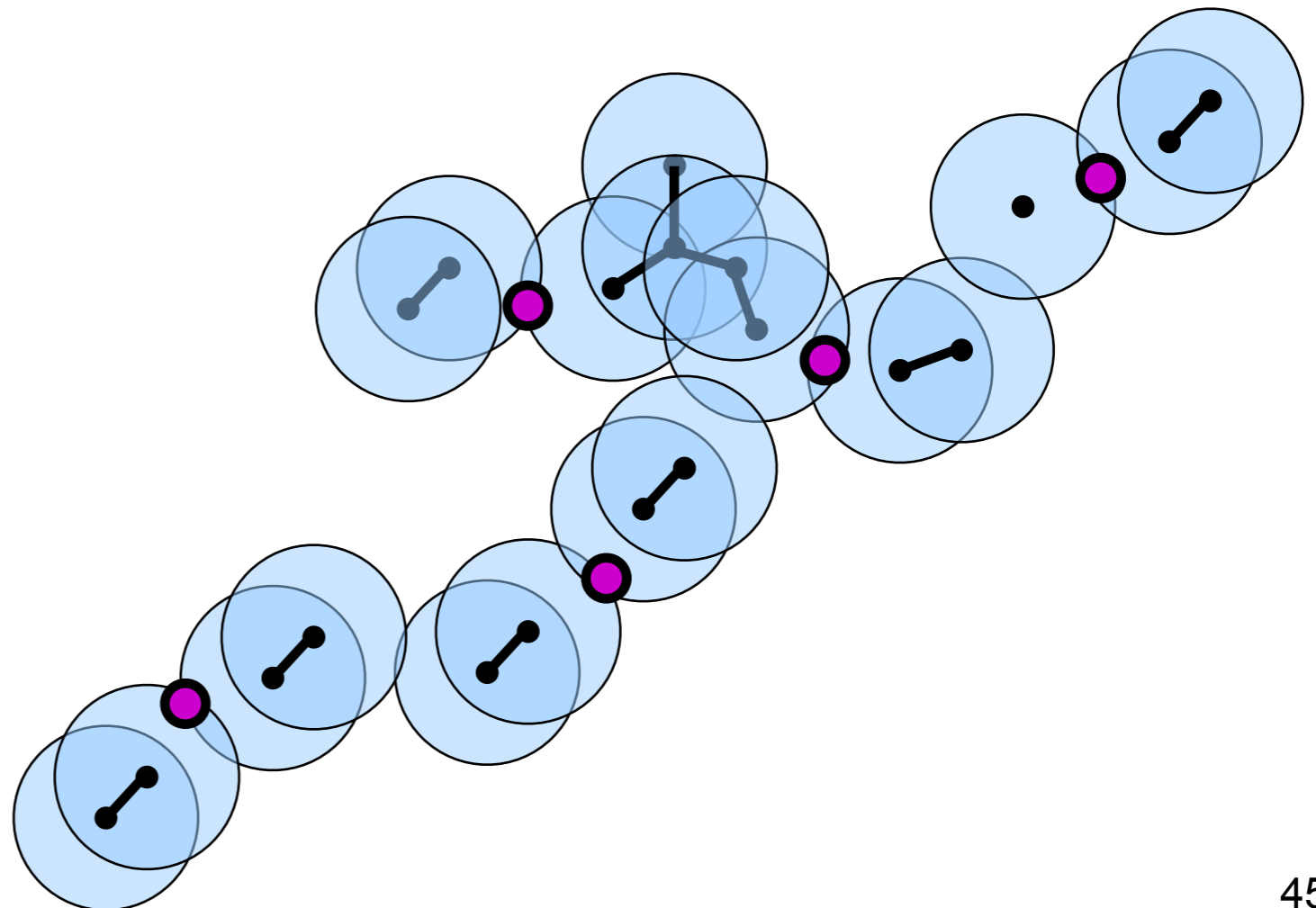


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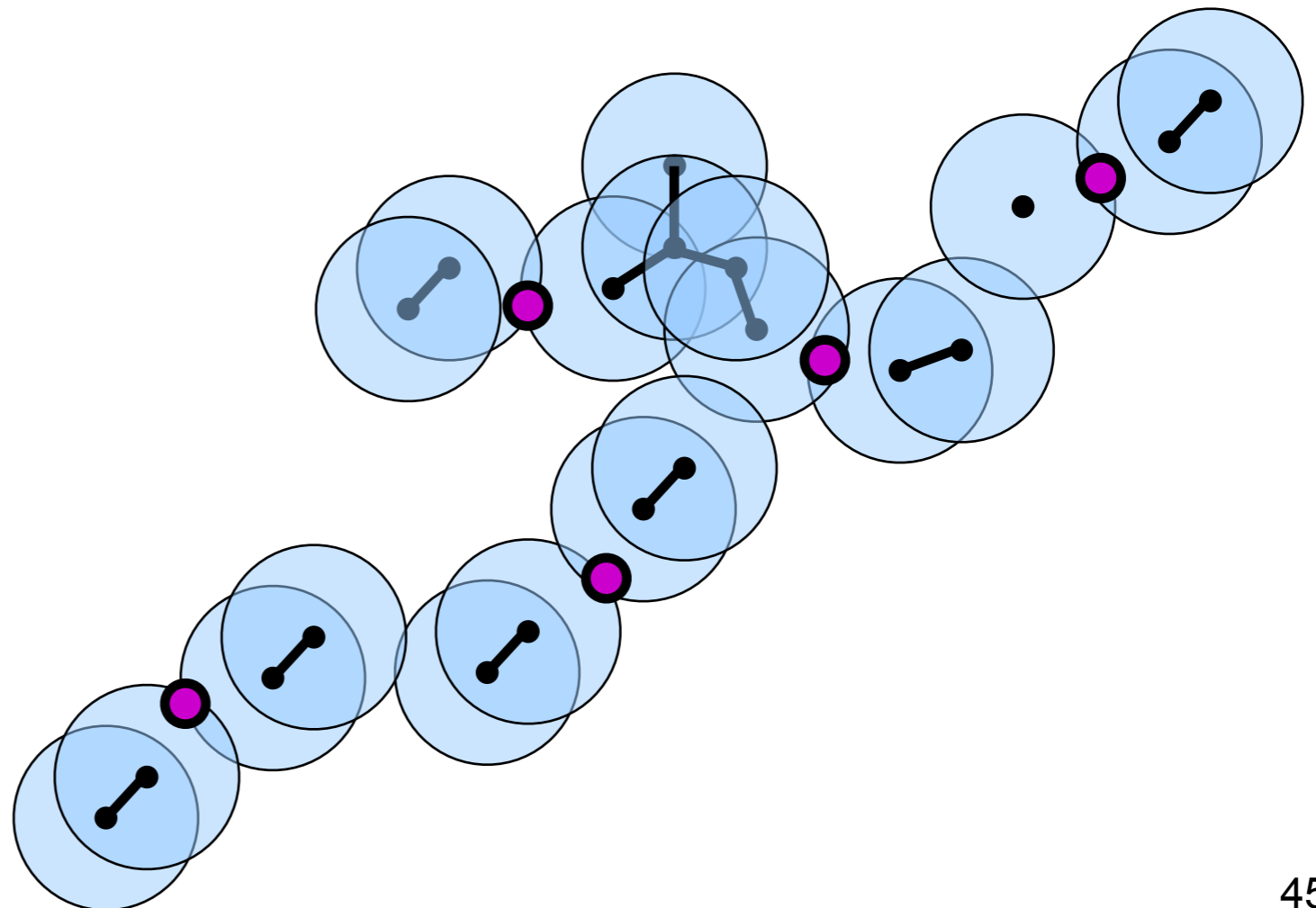


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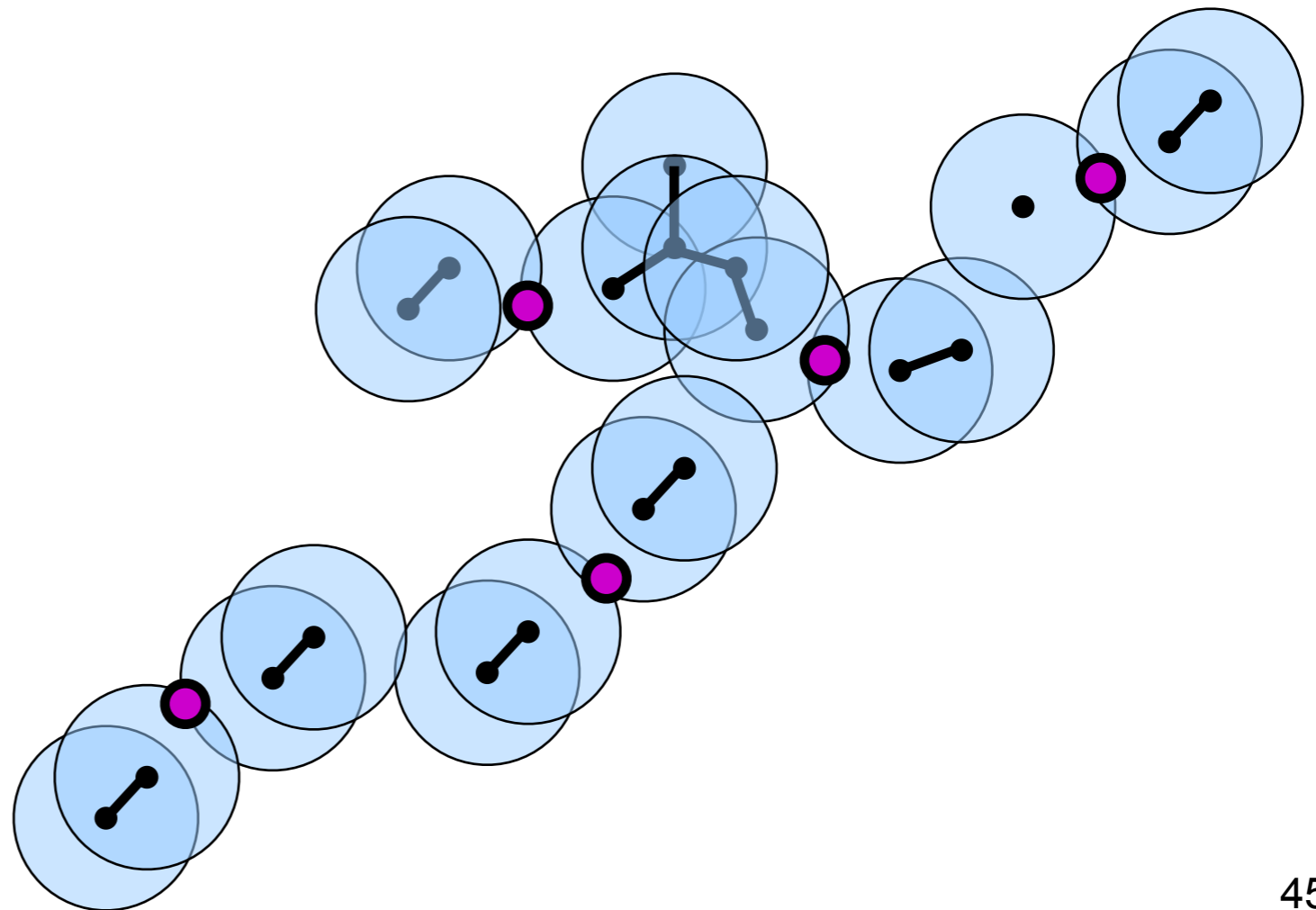


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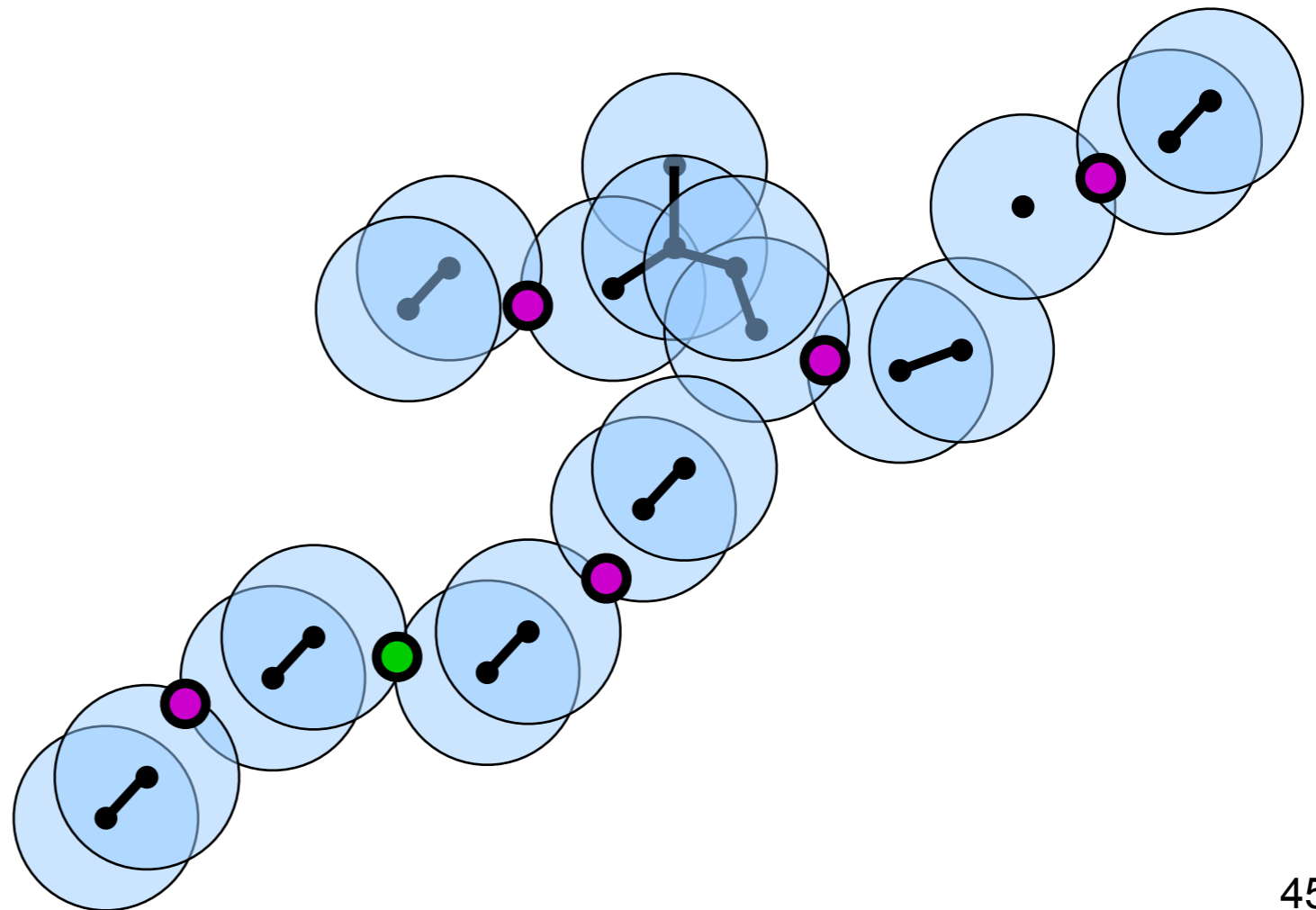


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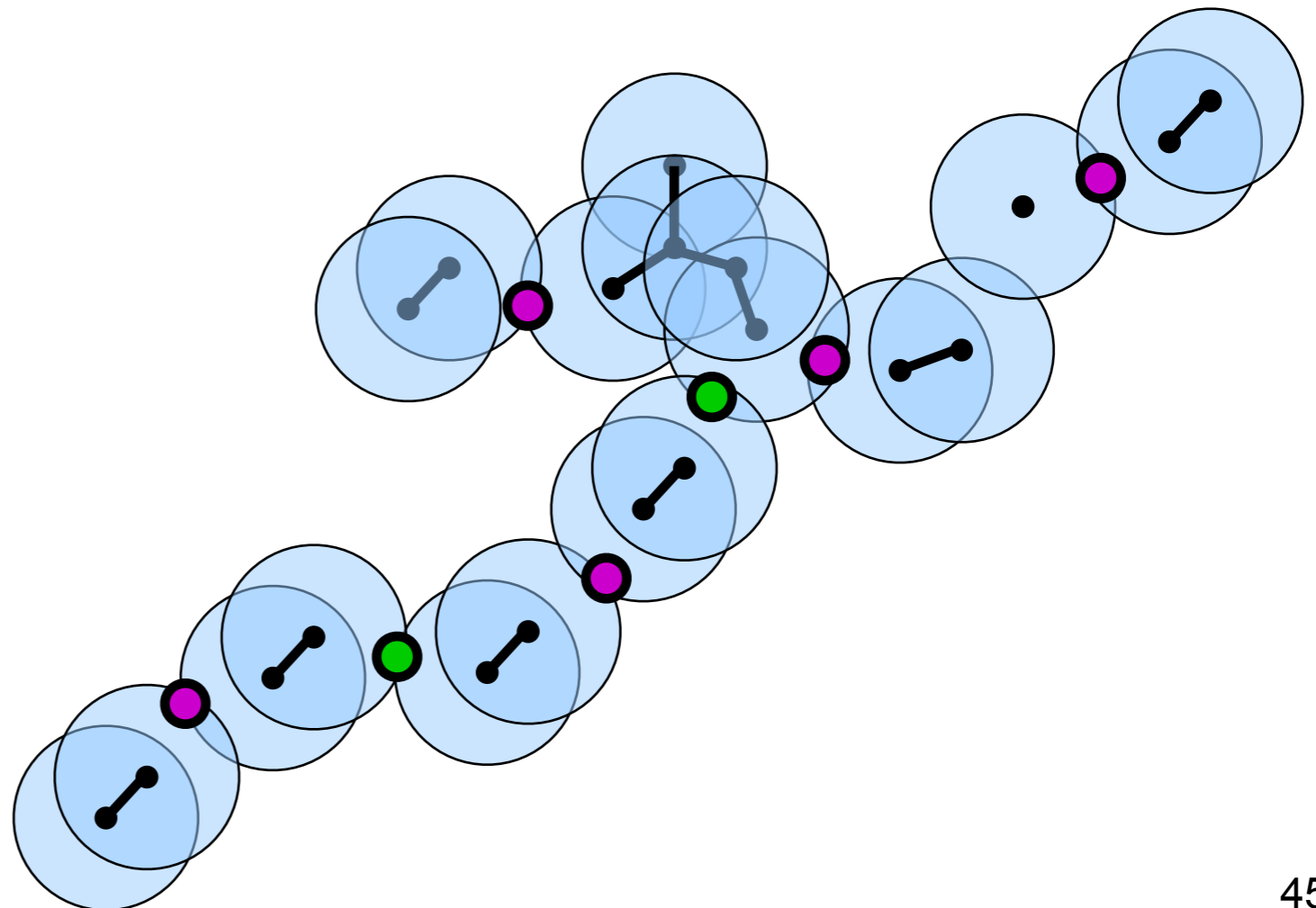


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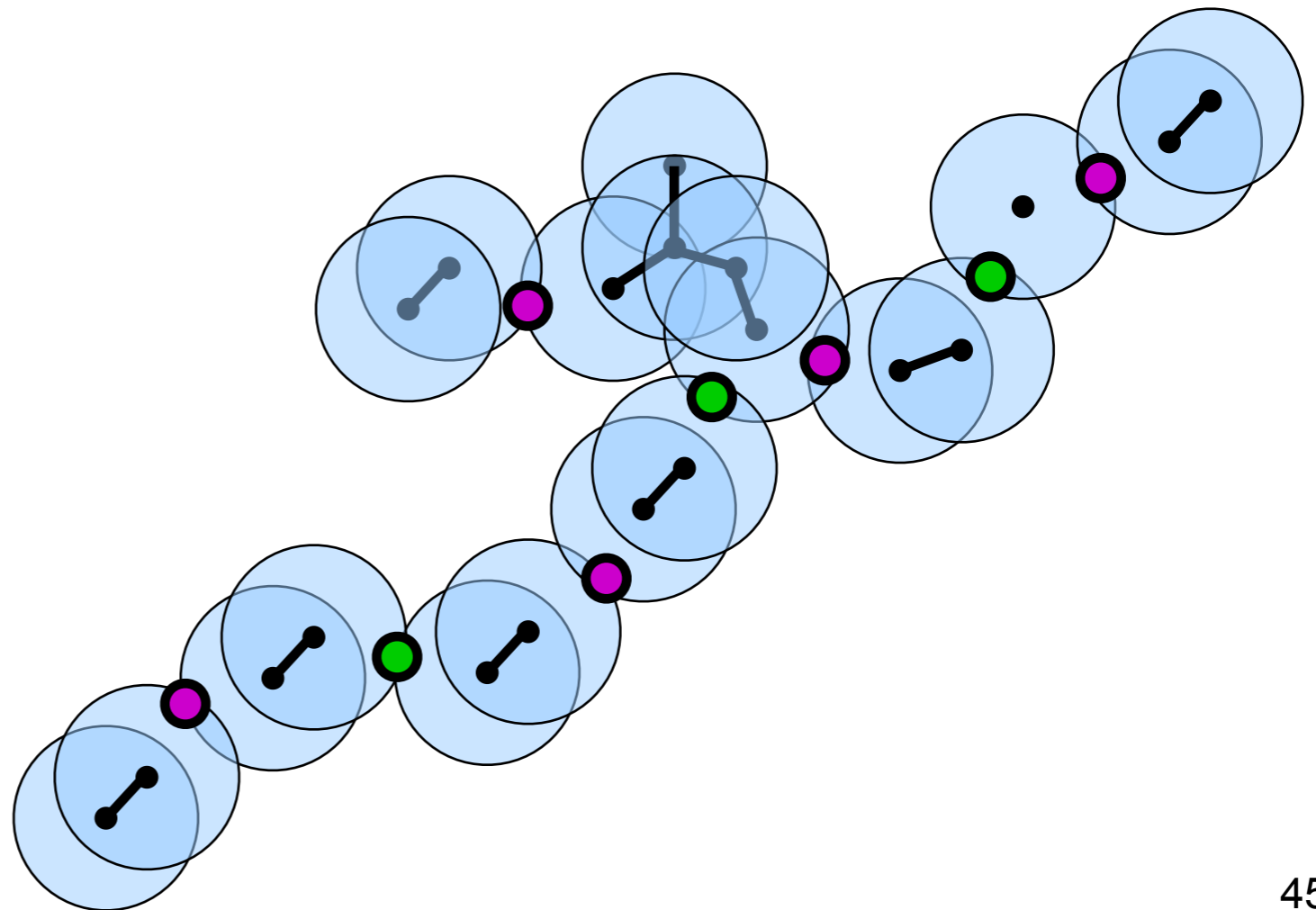


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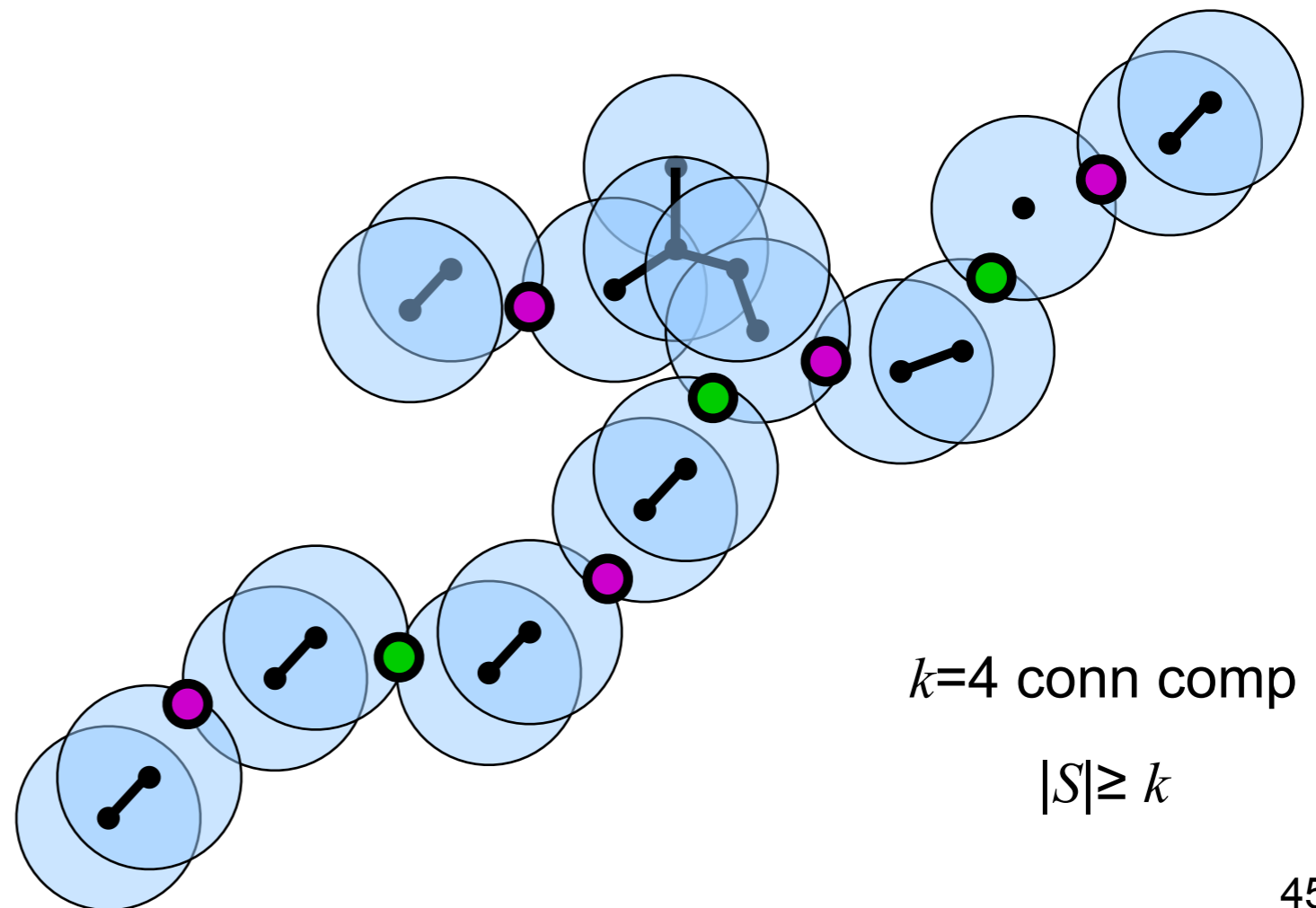
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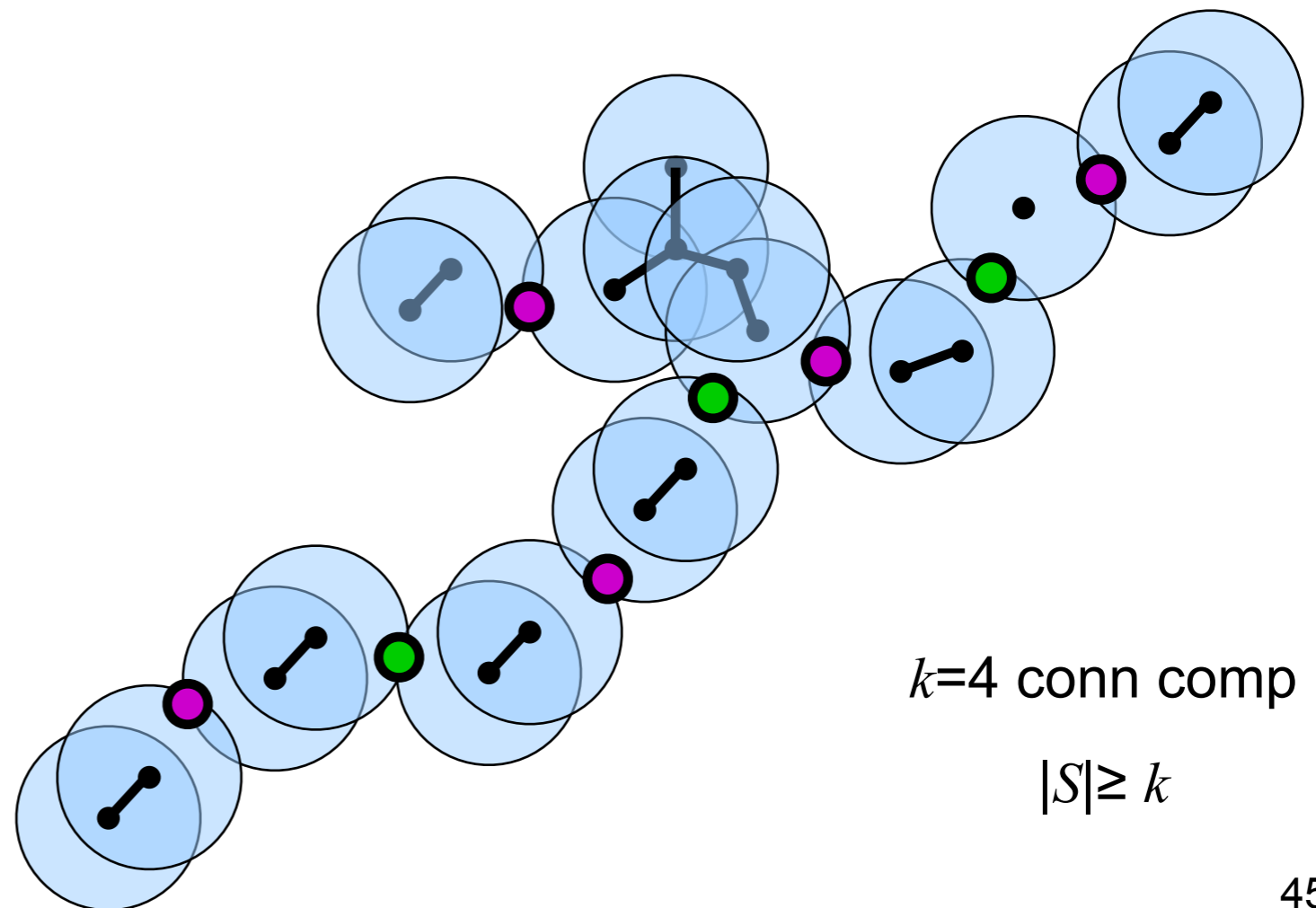
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Each added relay reduces the number of connected components by 1

Thus, at most  $k-1 \leq |S|-1$  additional relays needed to connect the relays within  $C$



# Steiner Forests and Spanning Trees with Neighborhoods

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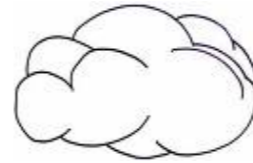
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Let  $\mathcal{P}$  = family of planar subsets (neighborhoods)

# Steiner Forests and Spanning Trees with Neighborhoods

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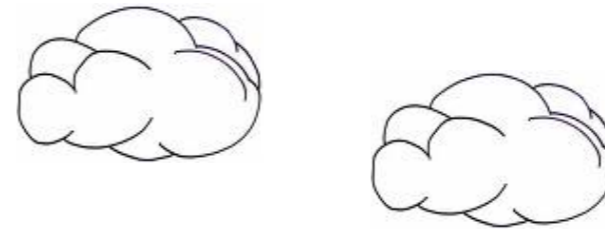
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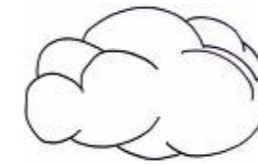
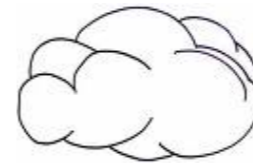
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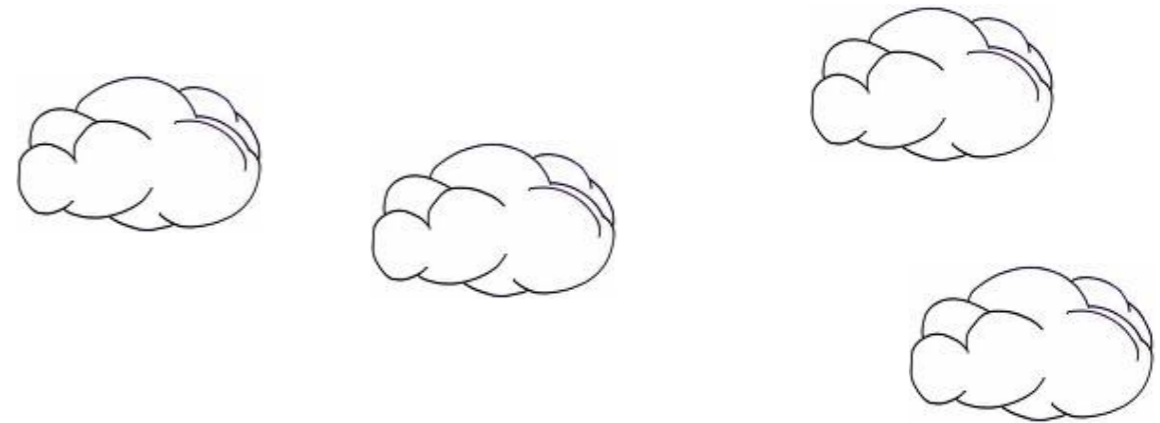
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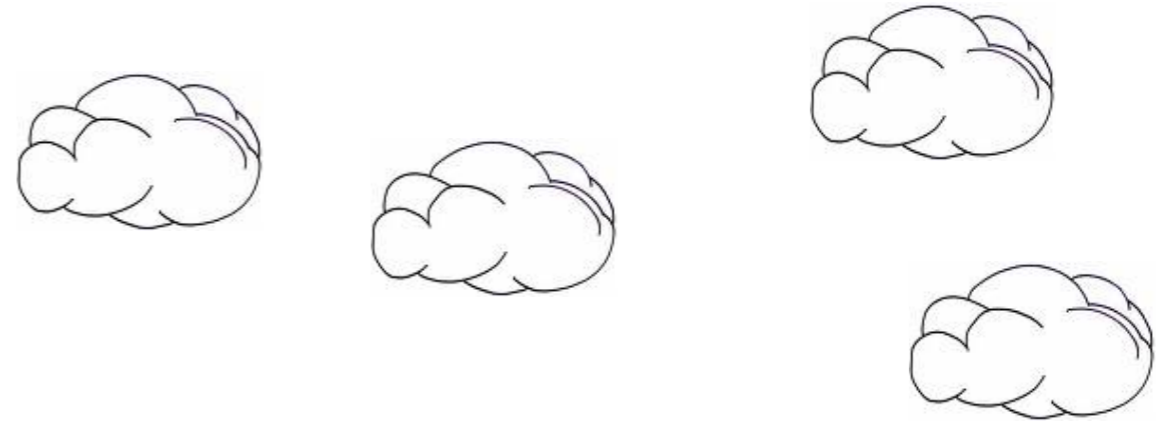


# Steiner Forests and Spanning Trees with Neighborhoods

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**Minimum Steiner Forest for the Neighborhoods in  $\mathcal{P}$**





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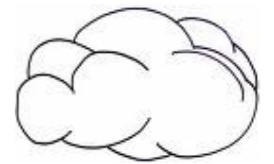
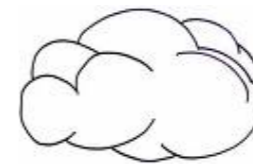
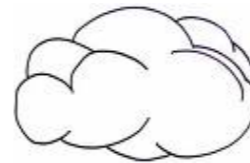
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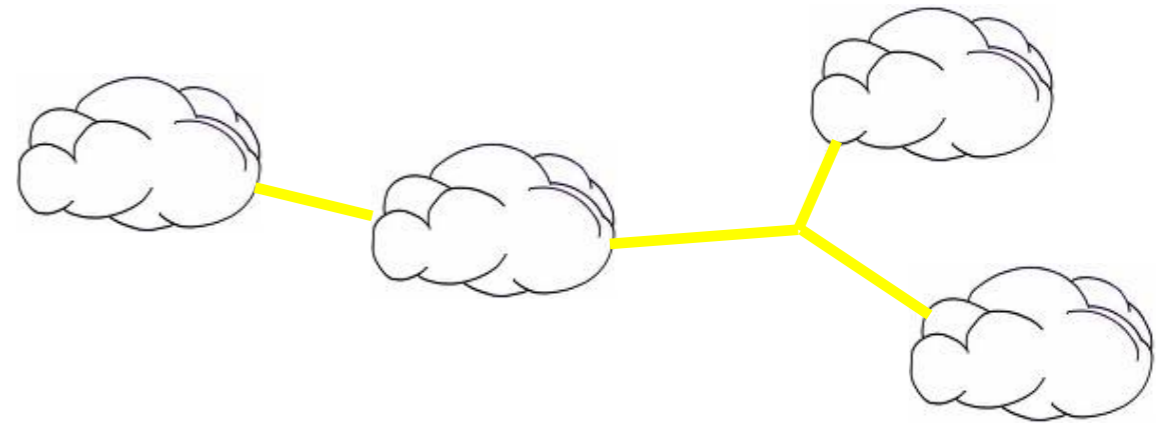
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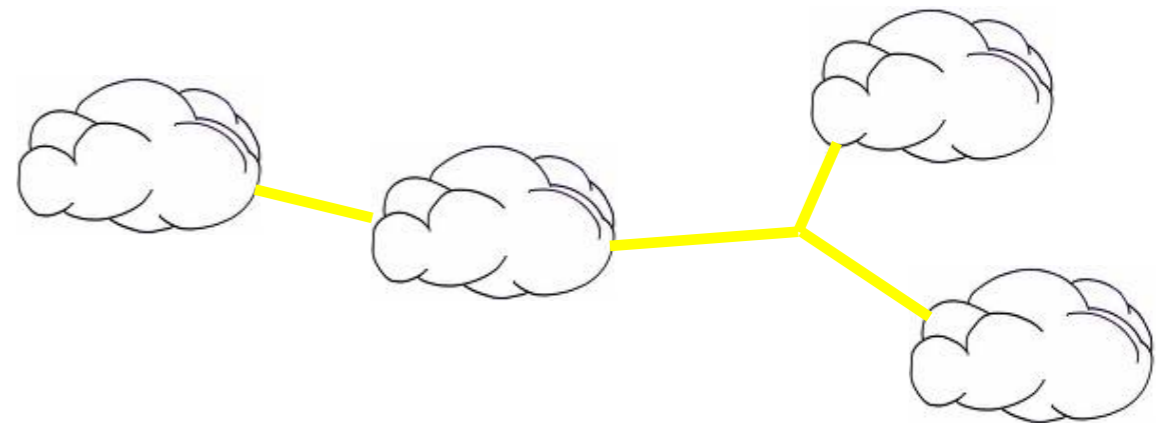
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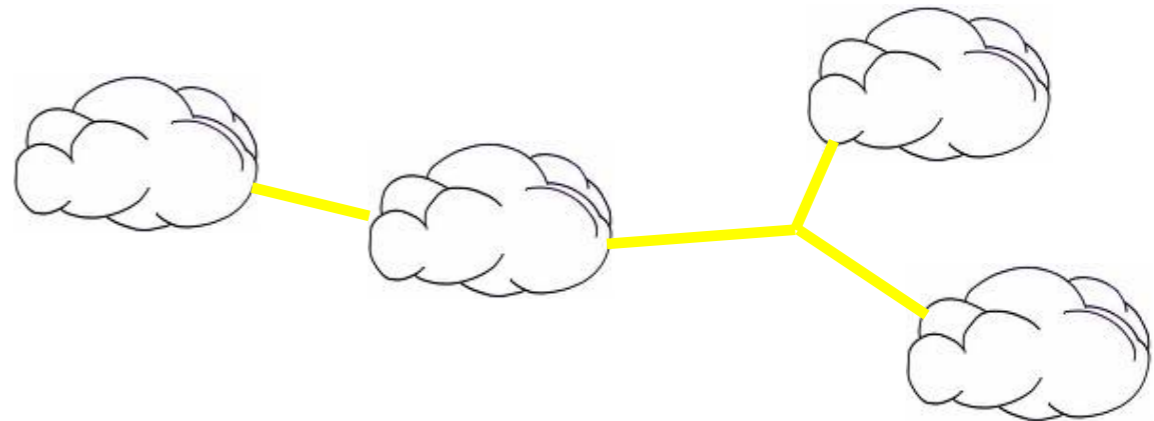
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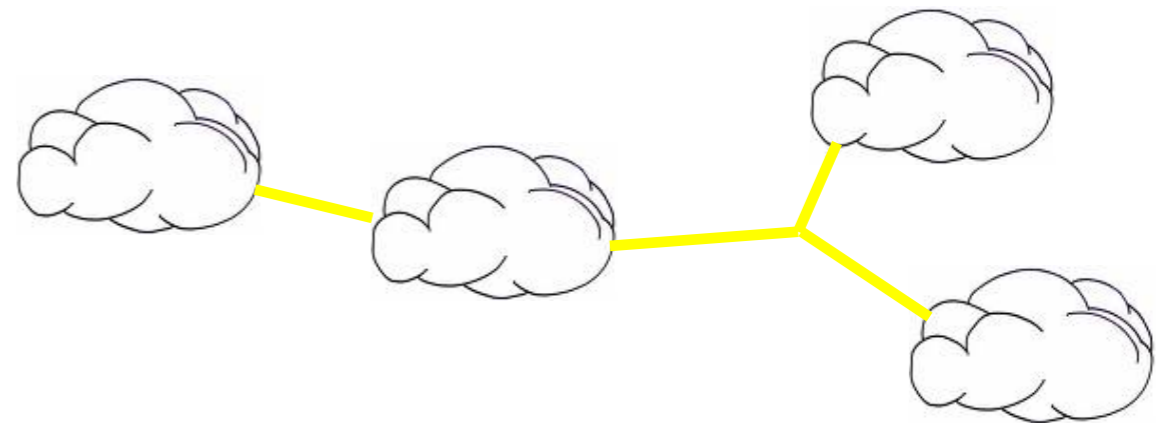
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## Minimum Spanning Forest for the Neighborhoods in $\mathcal{P}$

# Steiner Forests and Spanning Trees with Neighborhoods

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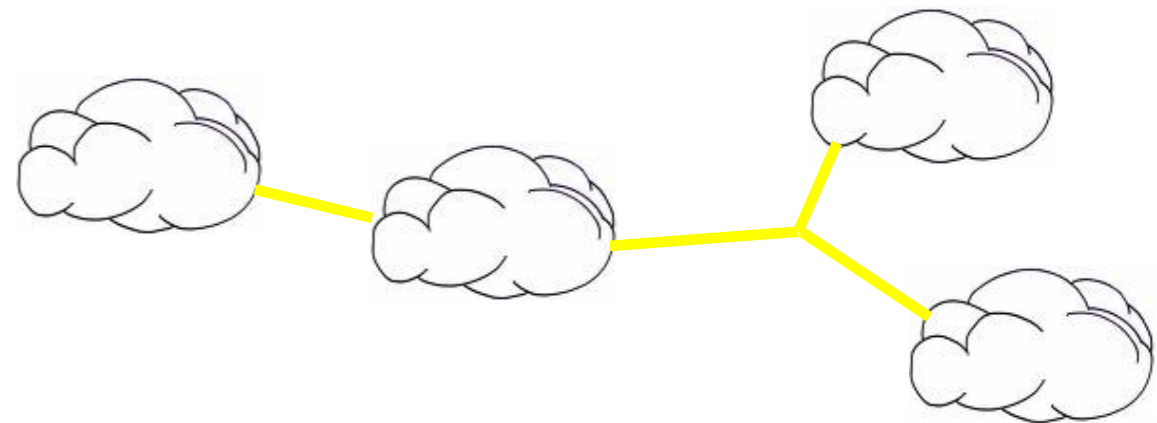
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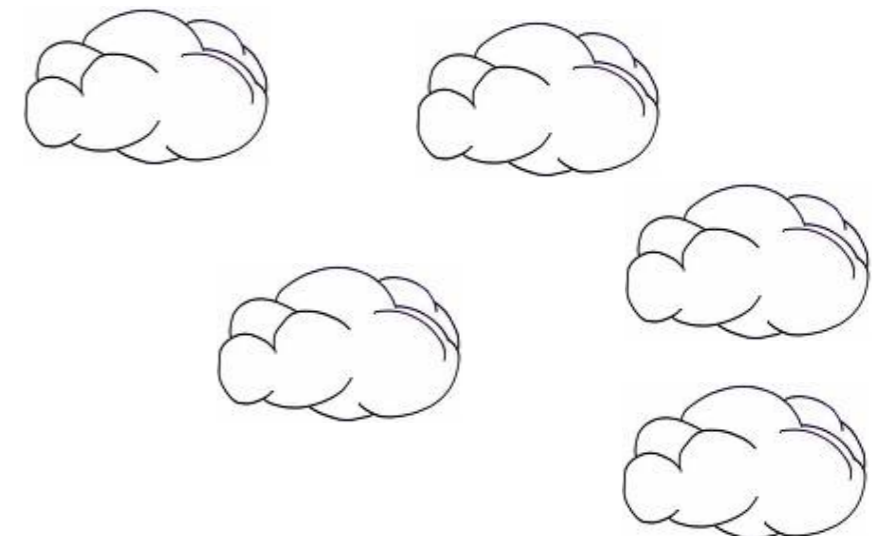
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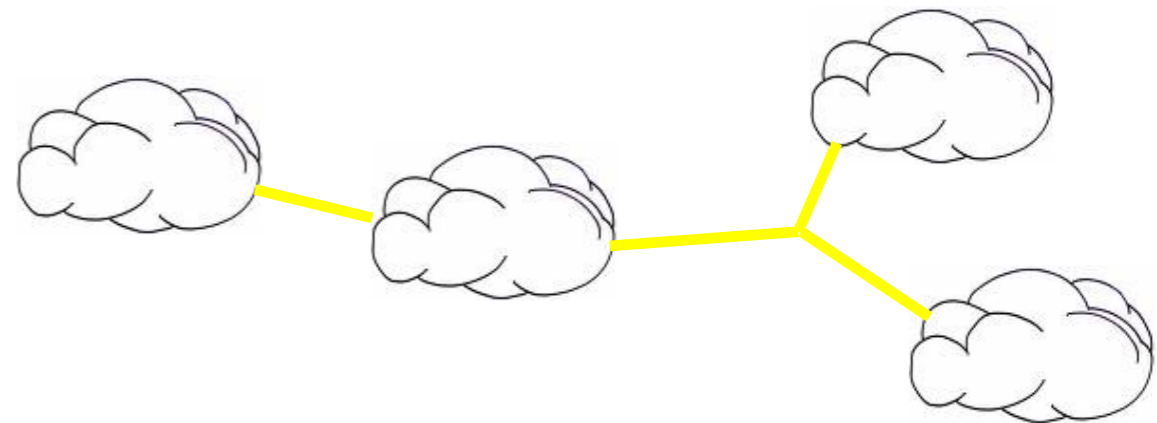
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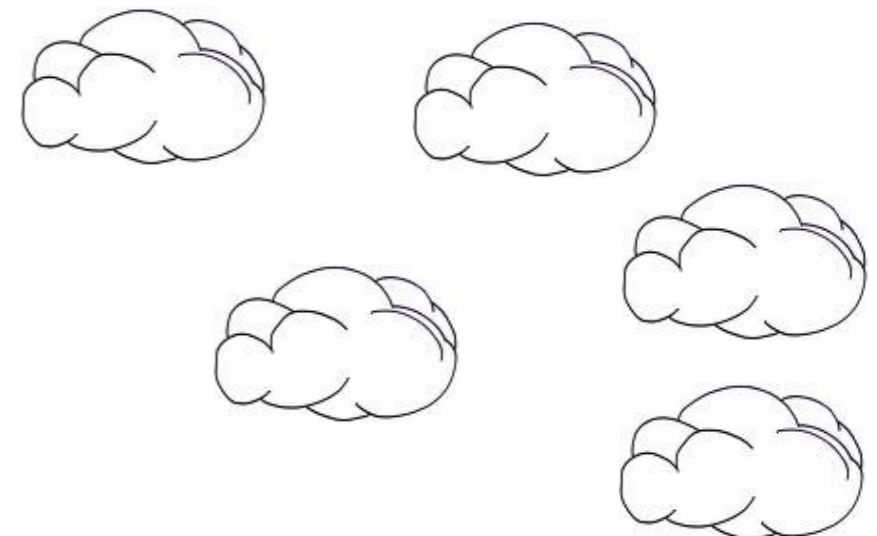
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## Minimum Spanning Forest for the Neighborhoods in $\mathcal{P}$

MSFN( $\mathcal{P}$ ) : min spanning tree in graph on  $\mathcal{P}$ ,





# Steiner Forests and Spanning Trees with Neighborhoods

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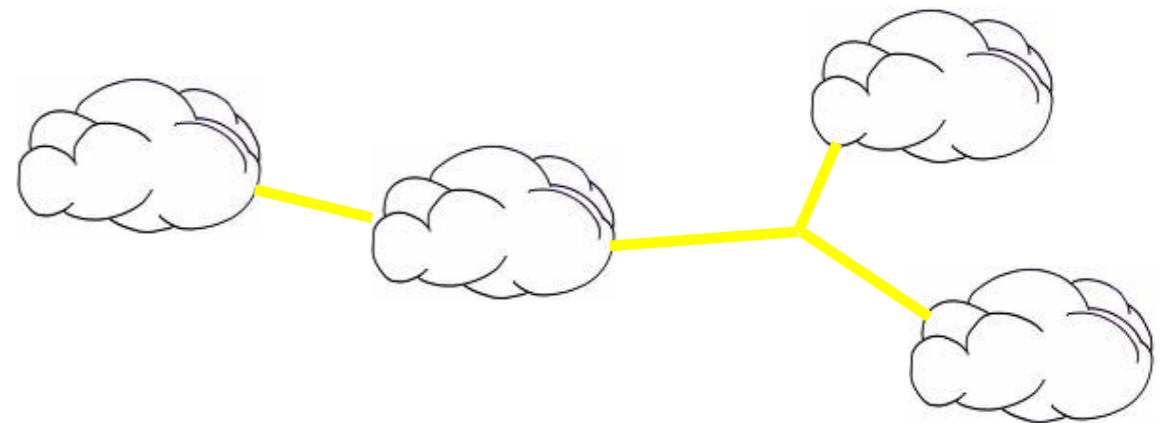
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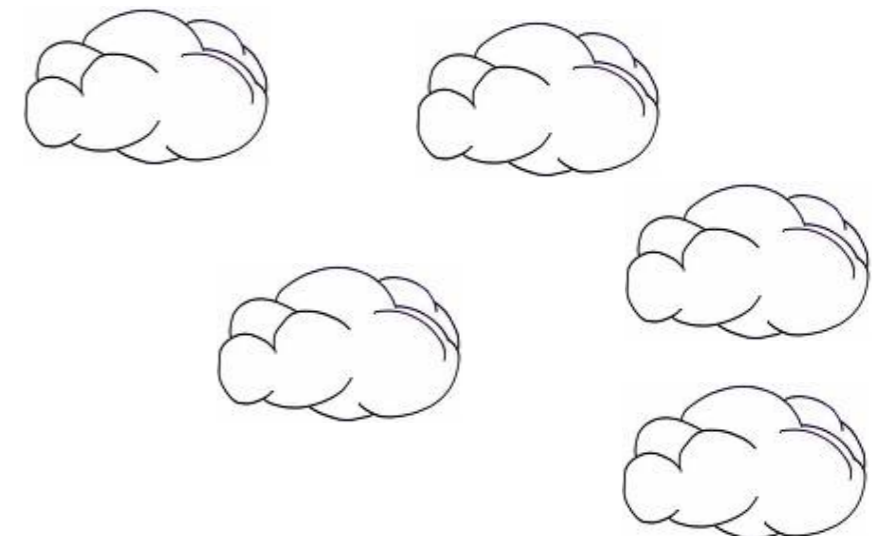
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edge weights = distance between sets



# Steiner Forests and Spanning Trees with Neighborhoods

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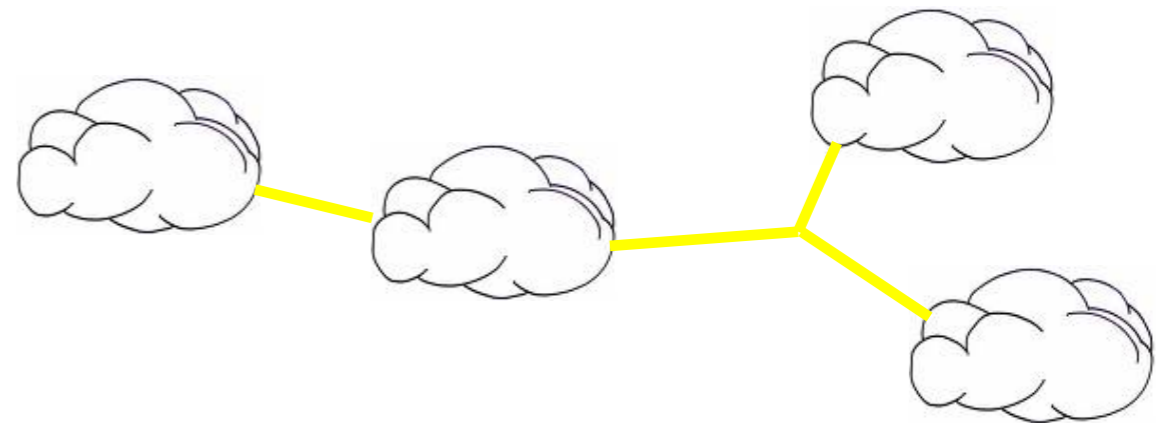
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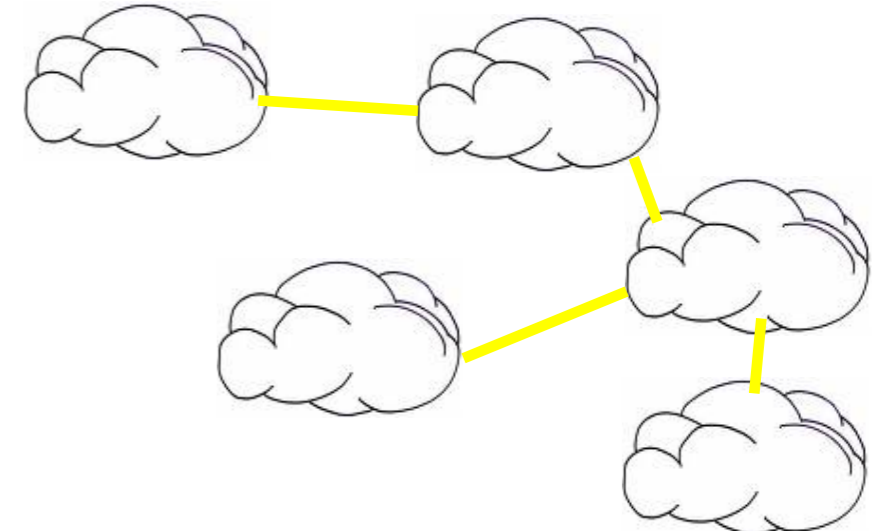
$G_{\mathcal{P}} = (\mathcal{P}, E(G))$  is connected

Counts only length *outside*  $\mathcal{P}$   
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MSFN( $\mathcal{P}$ ) : min spanning tree in graph on  $\mathcal{P}$ ,  
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# Steiner Forests and Spanning Trees with Neighborhoods

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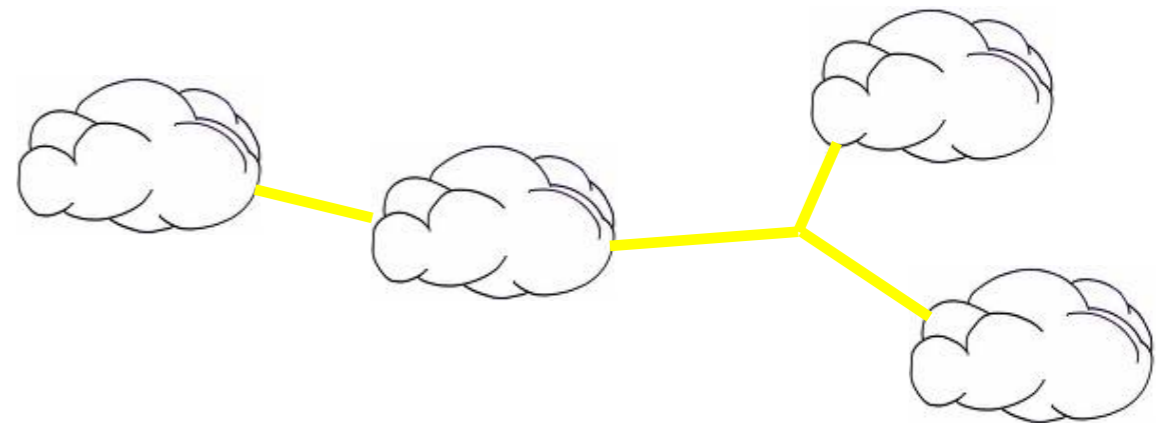
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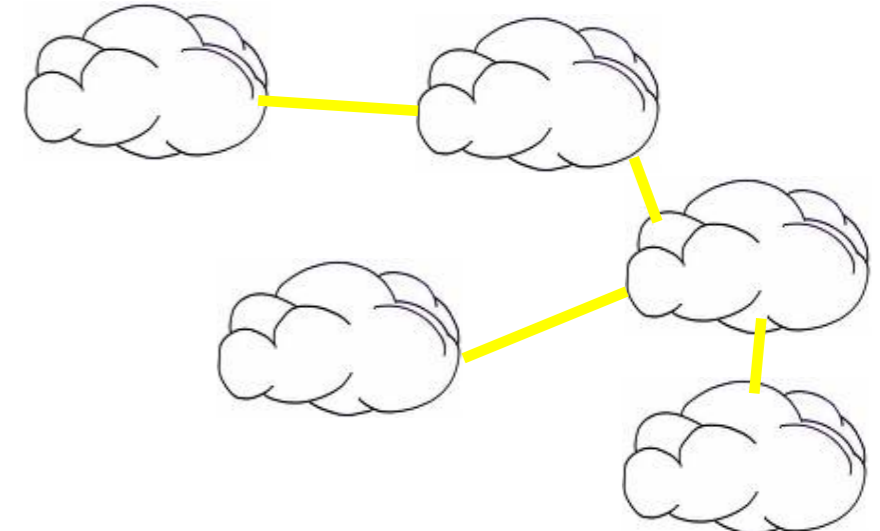
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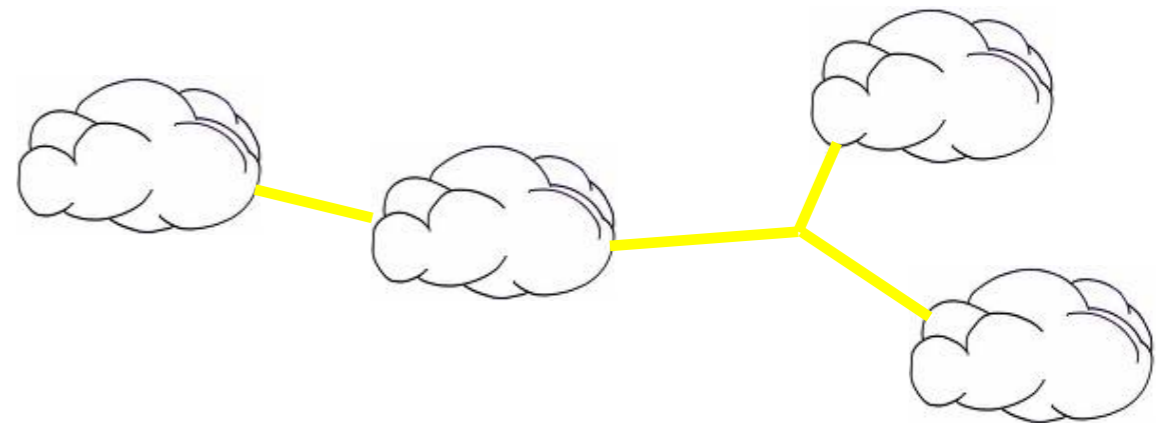
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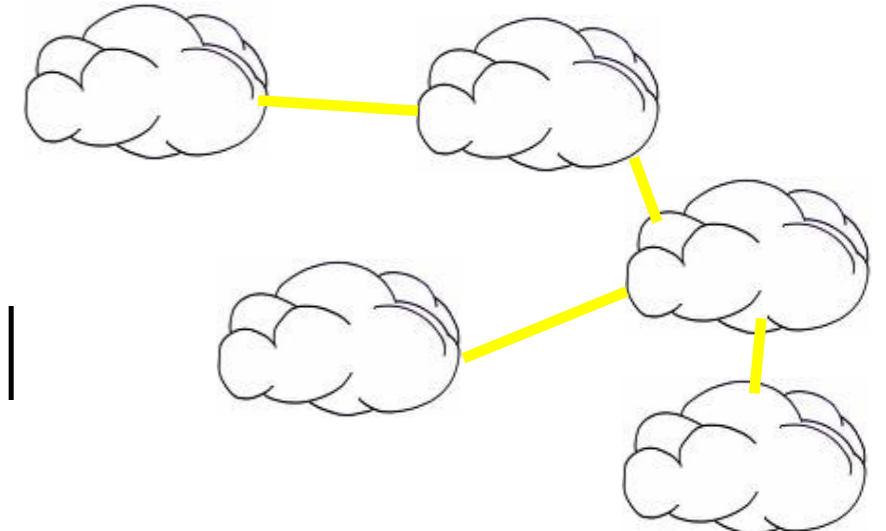
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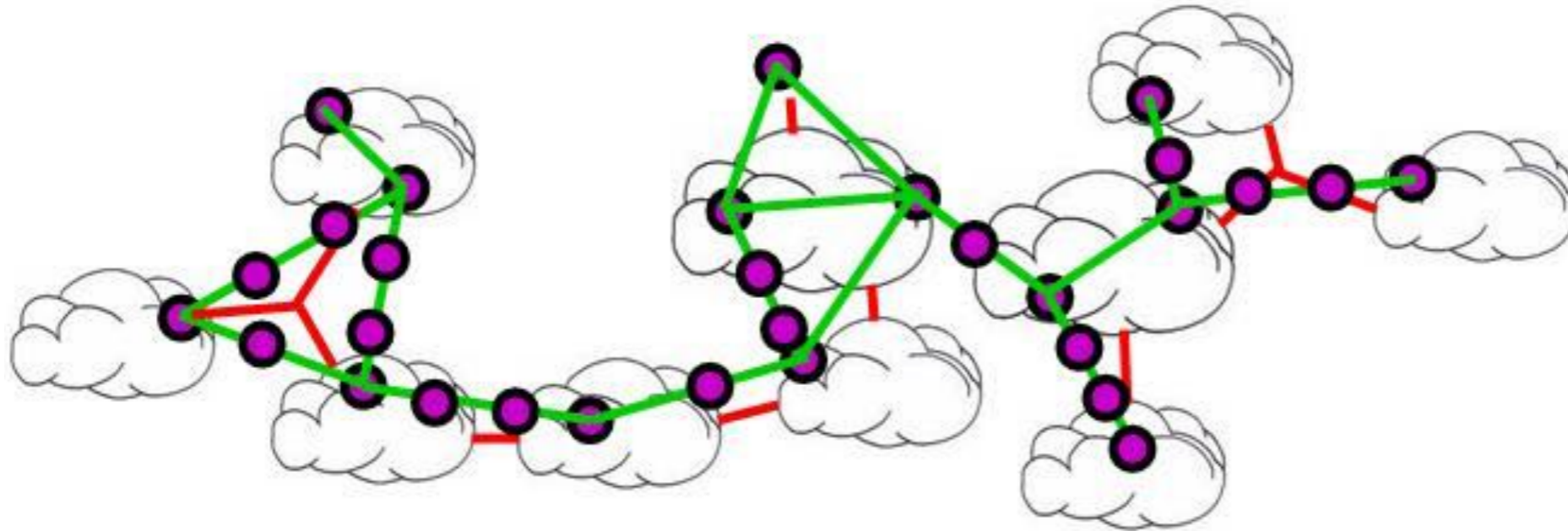
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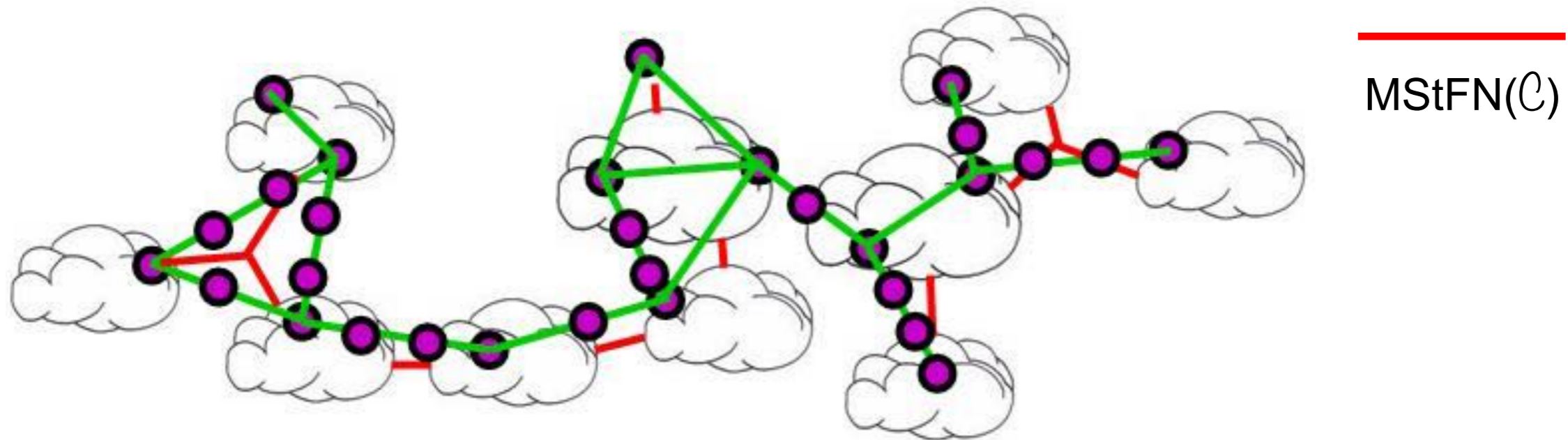
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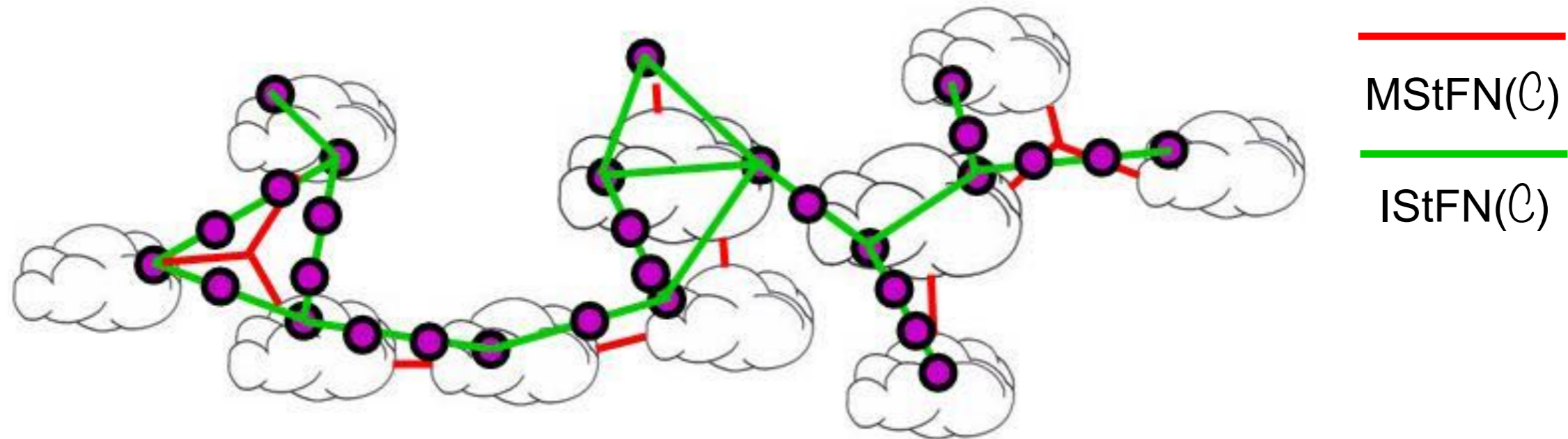




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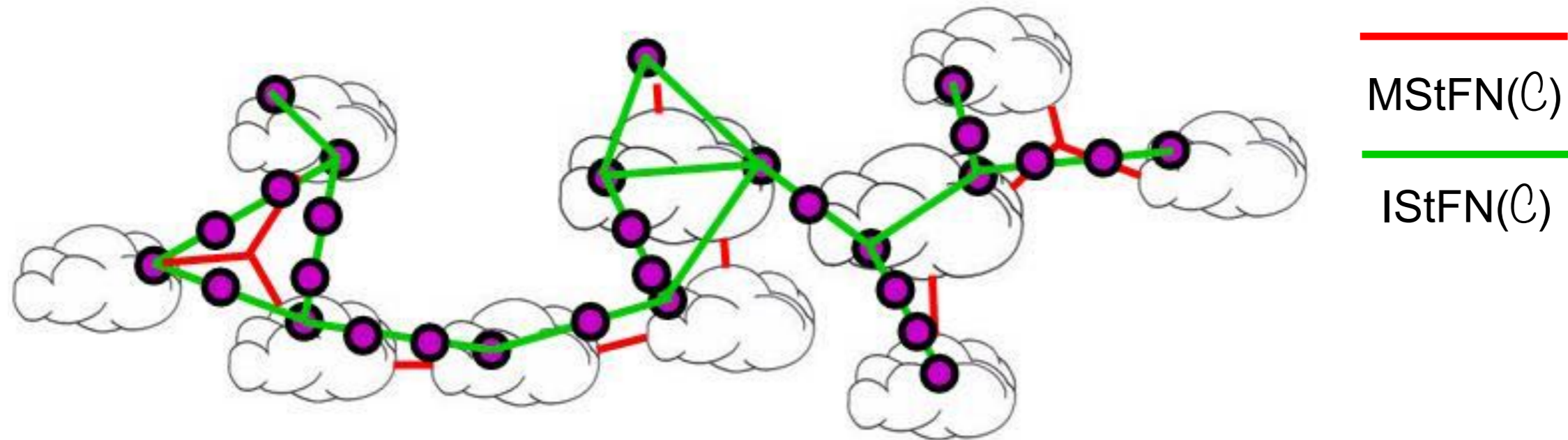
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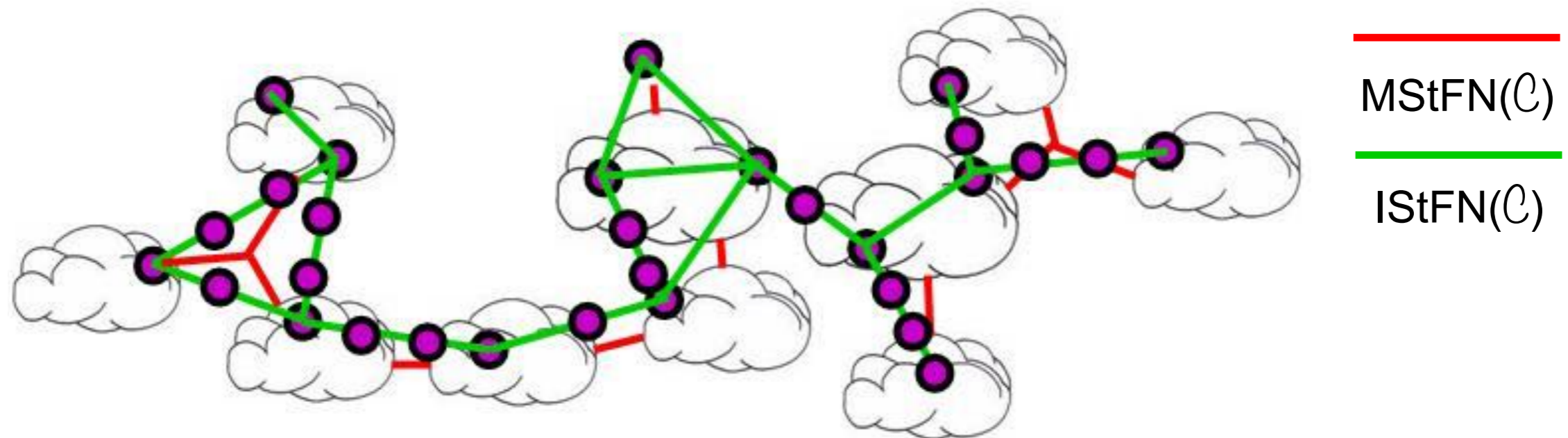
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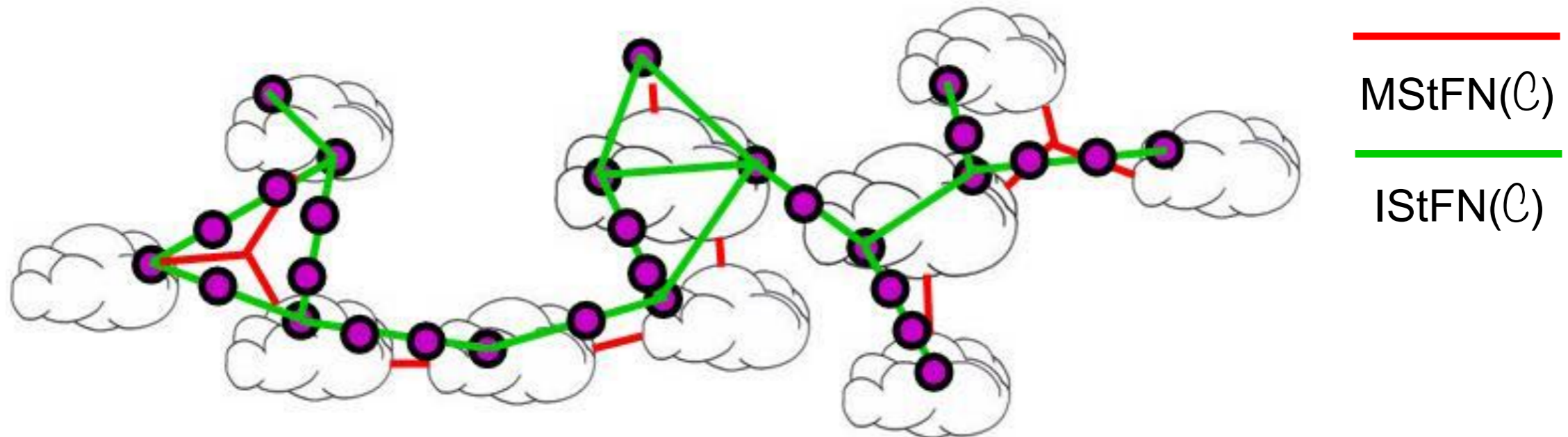


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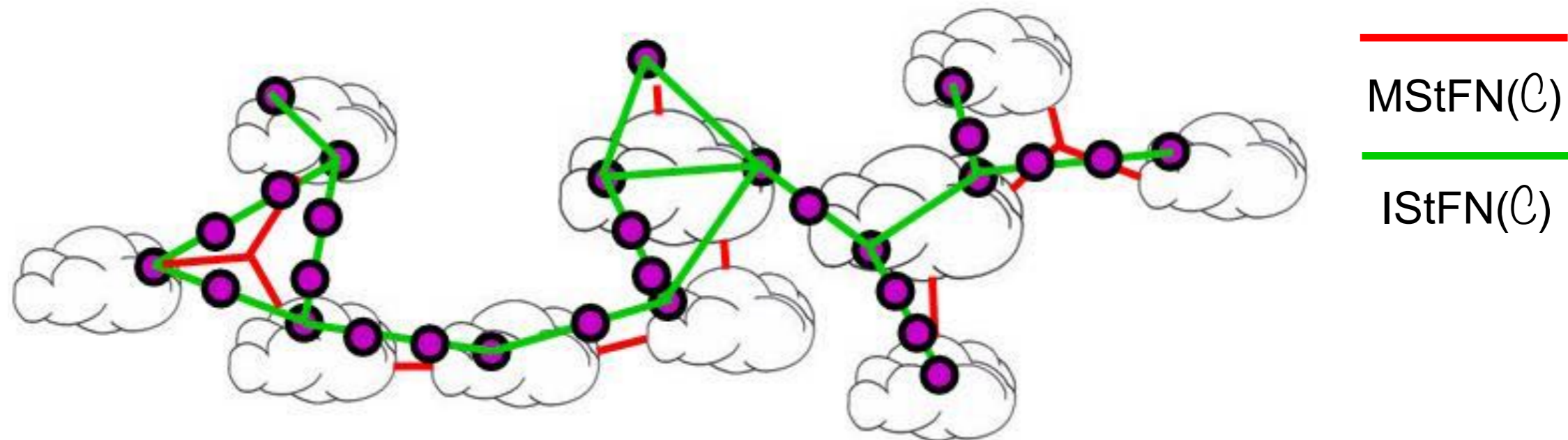


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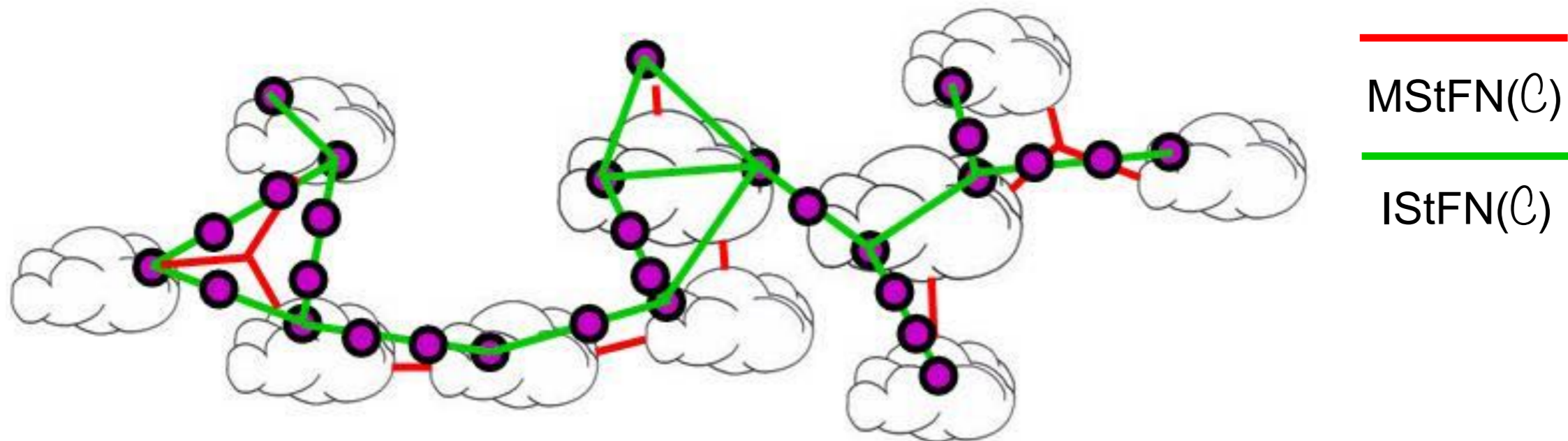
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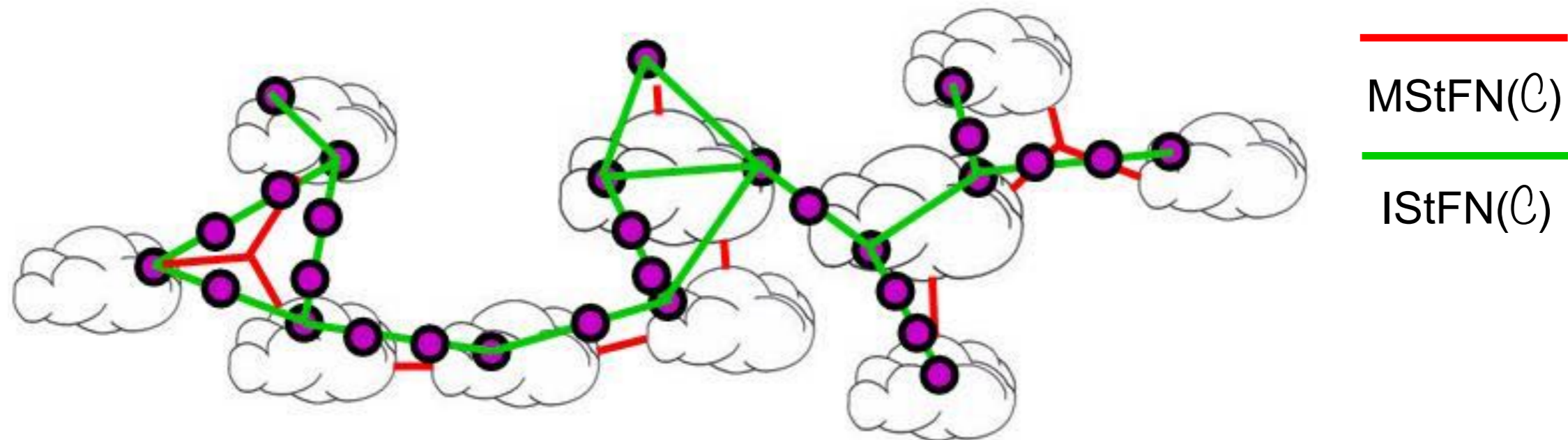
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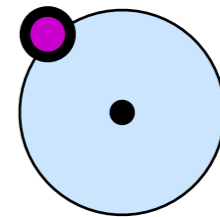
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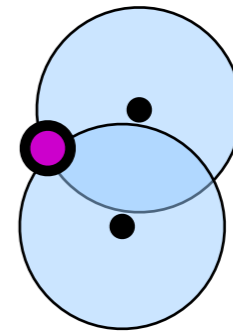
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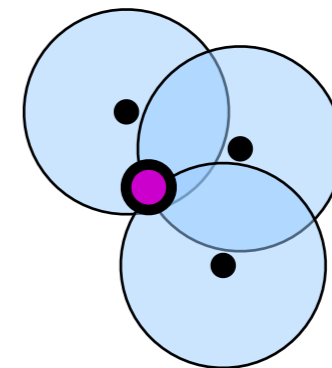
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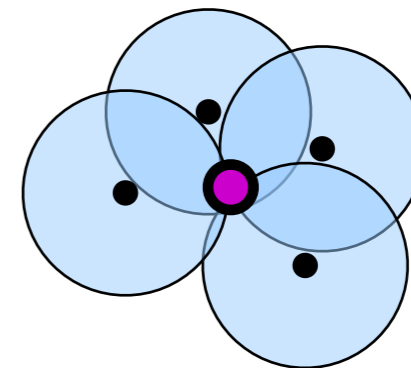




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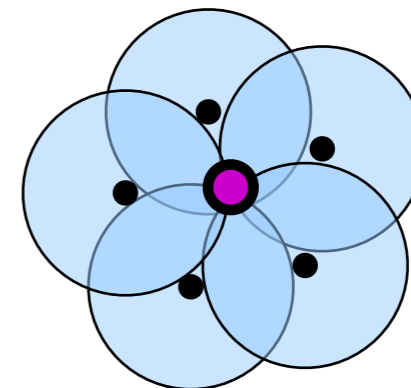
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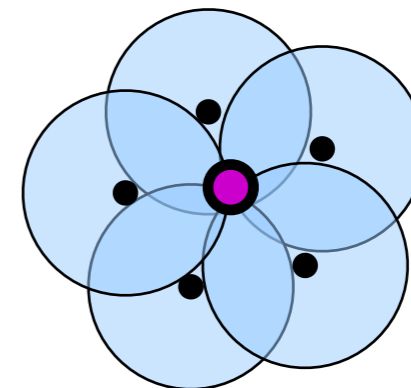
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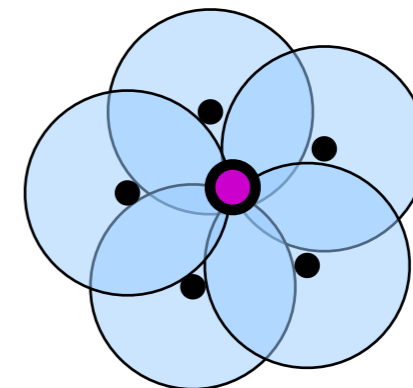
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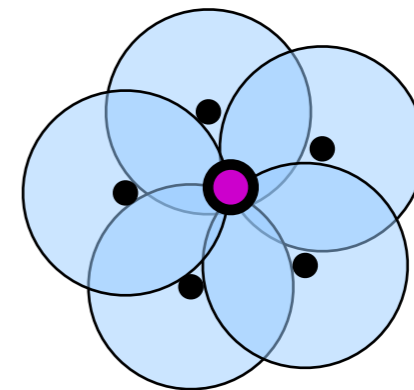


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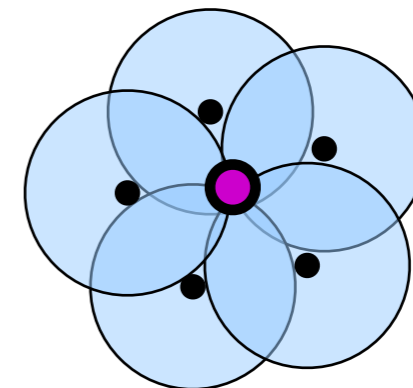


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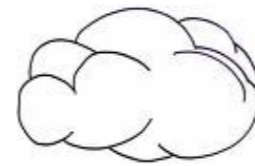
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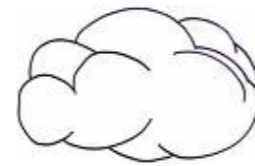
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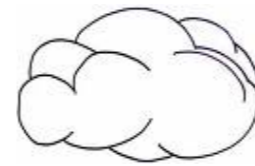


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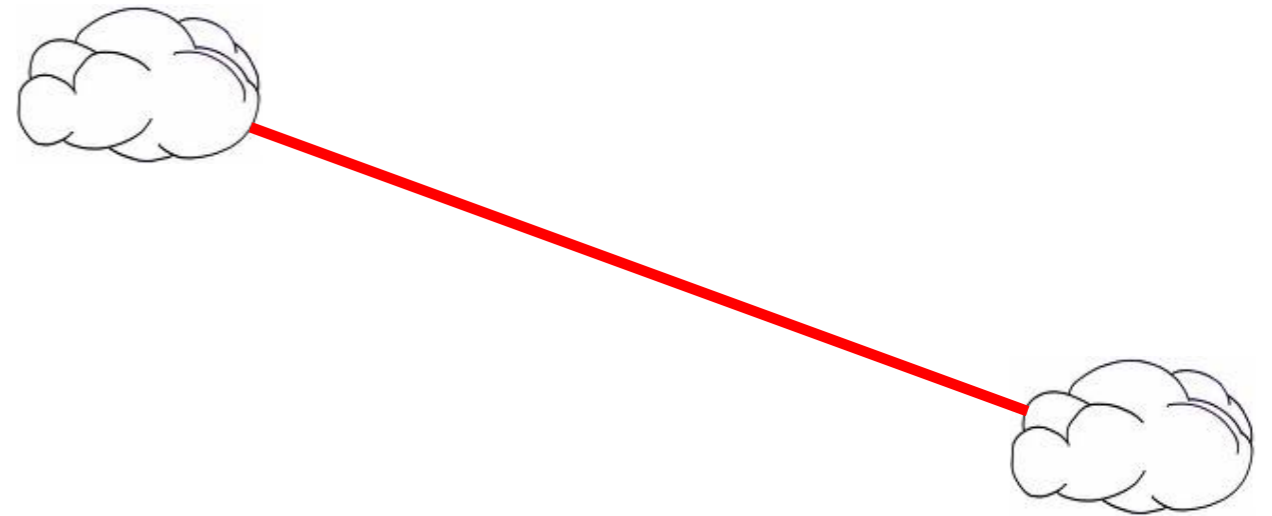
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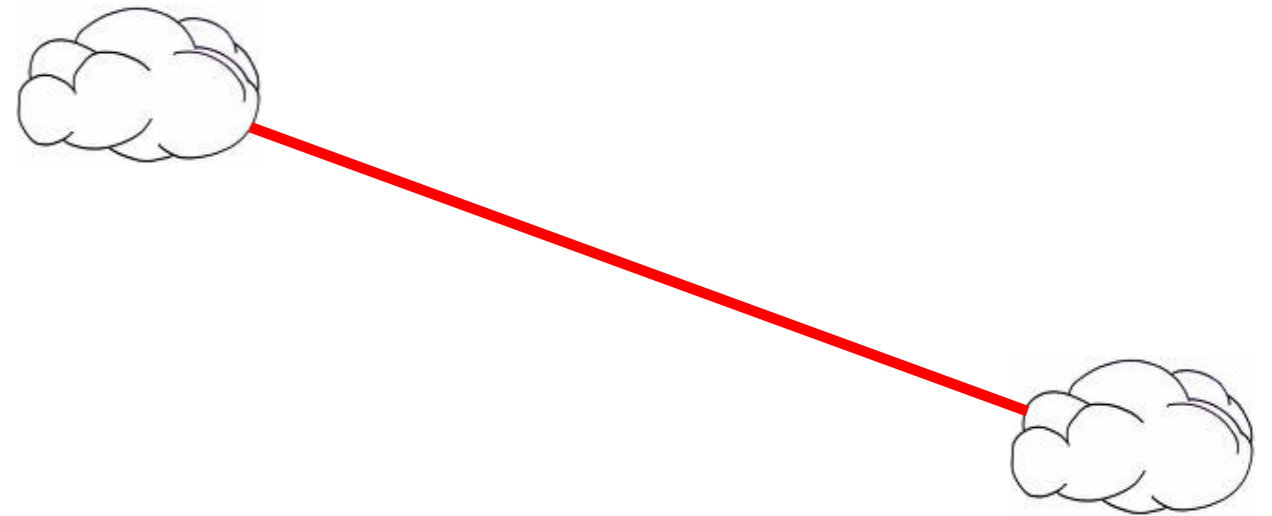
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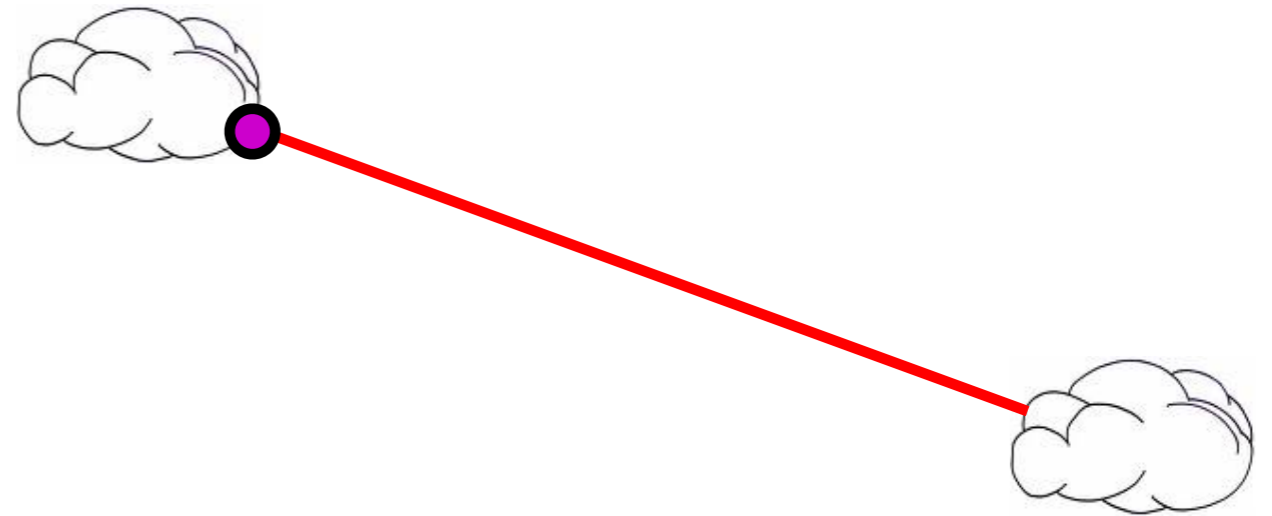
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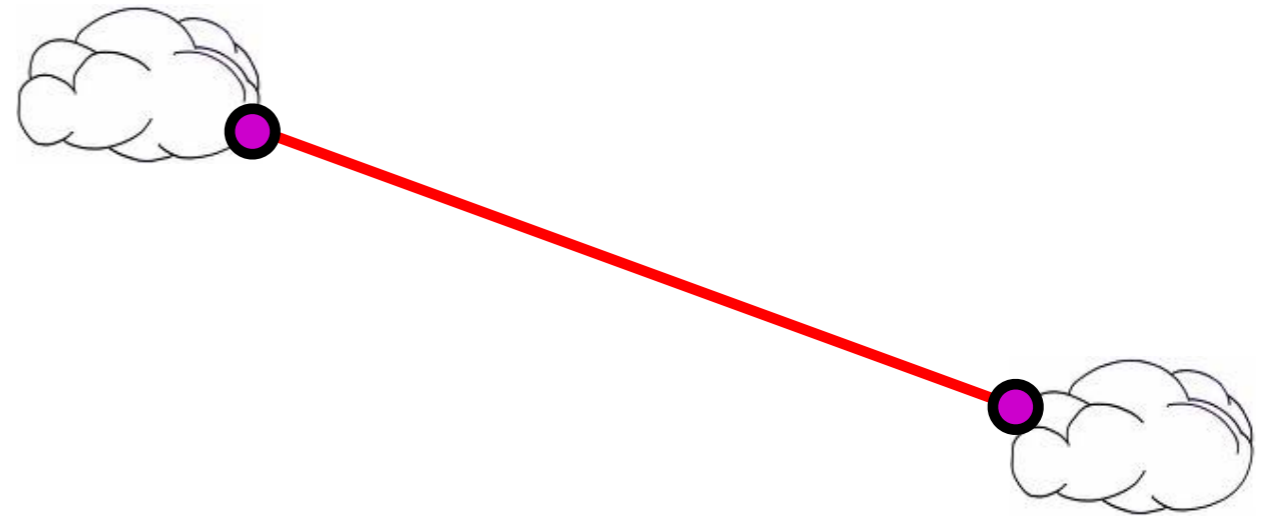
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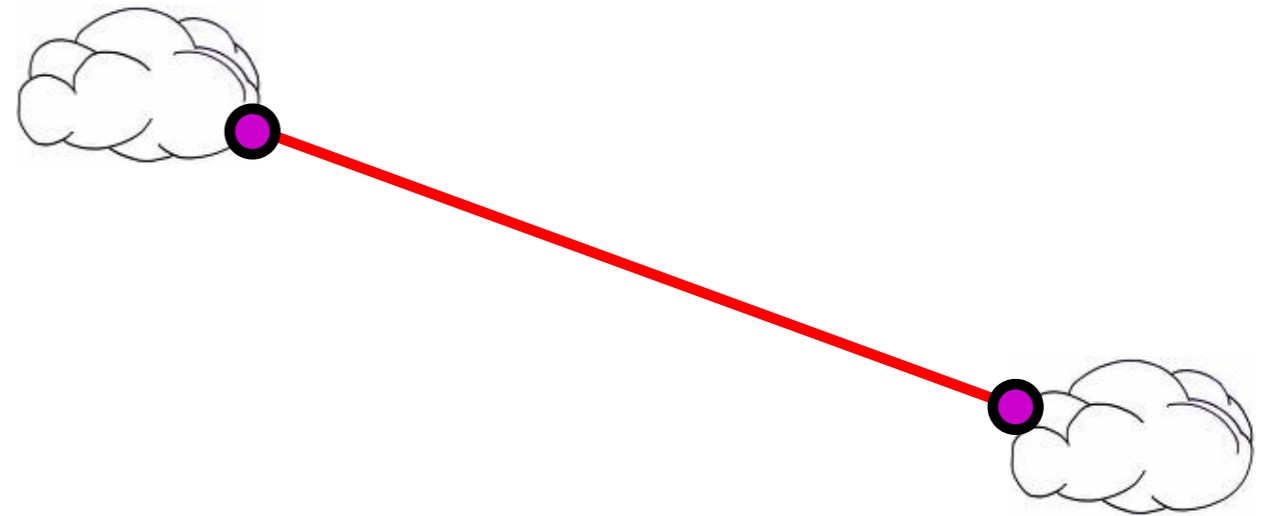
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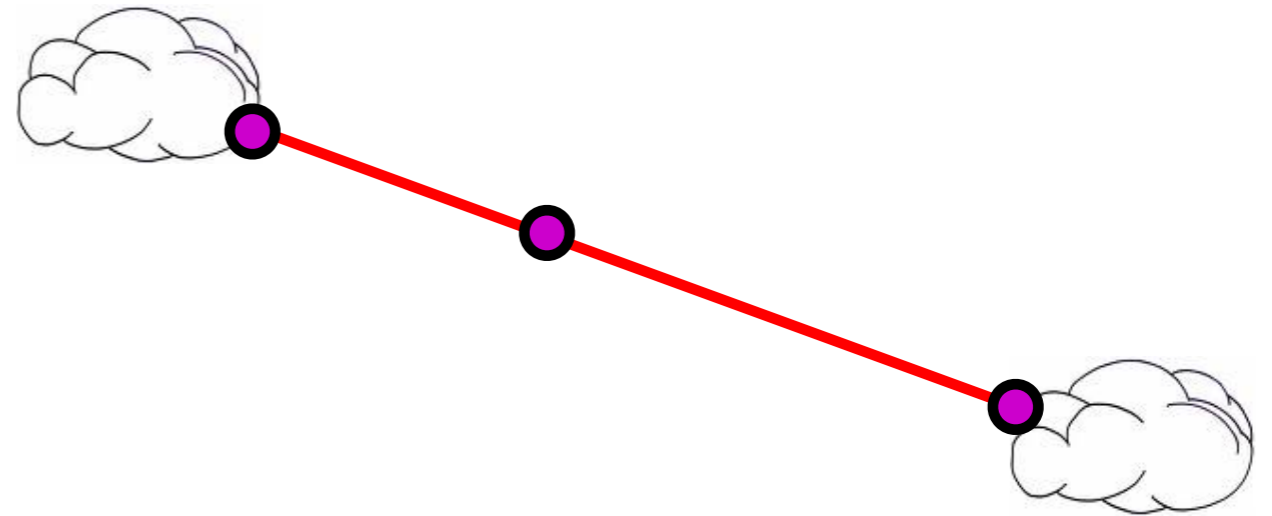
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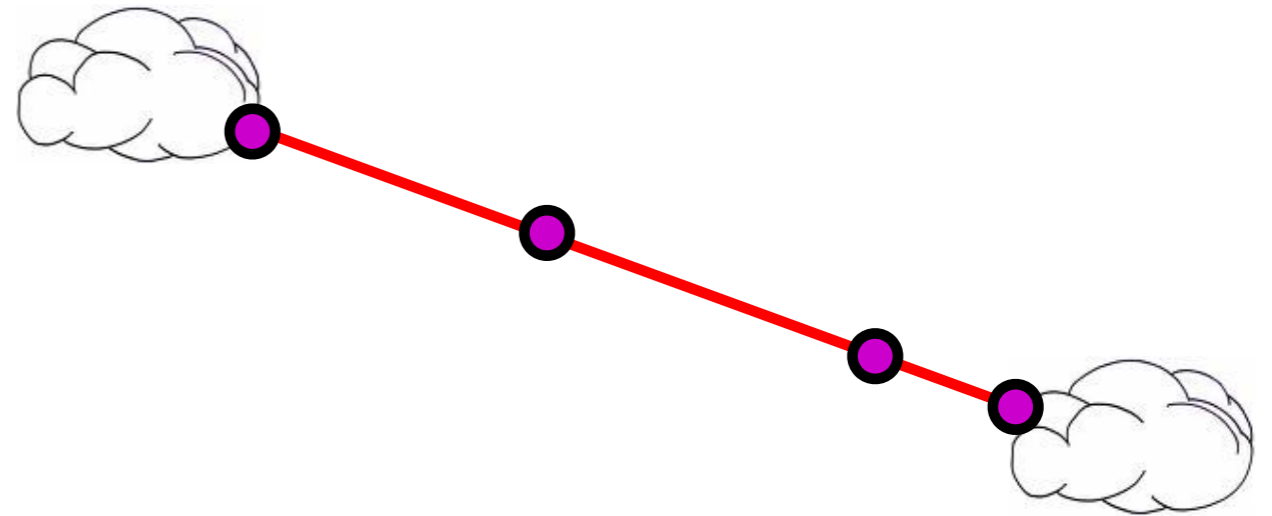
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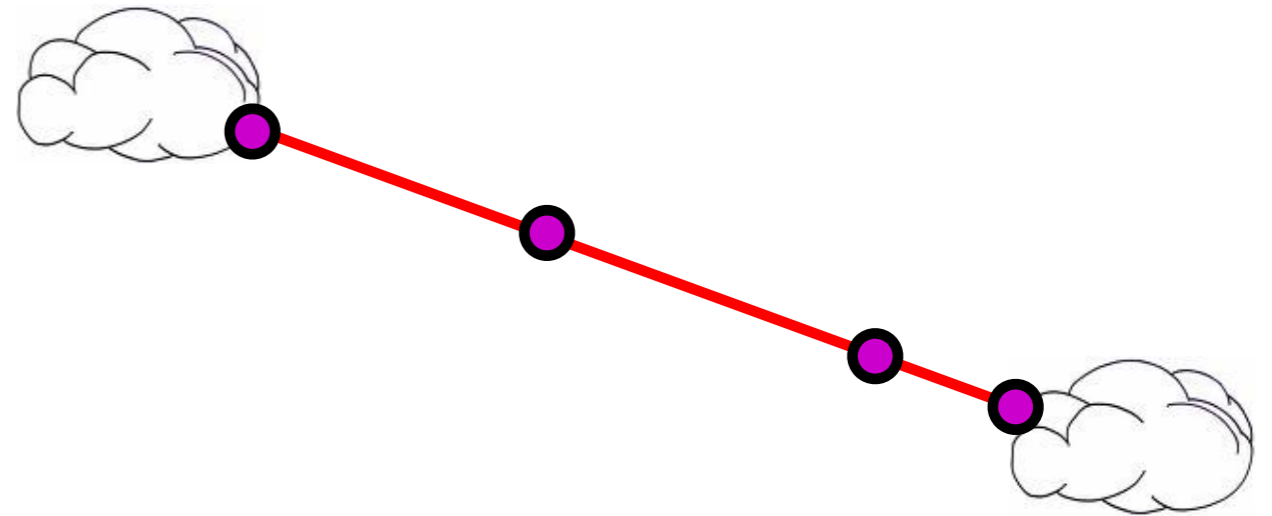
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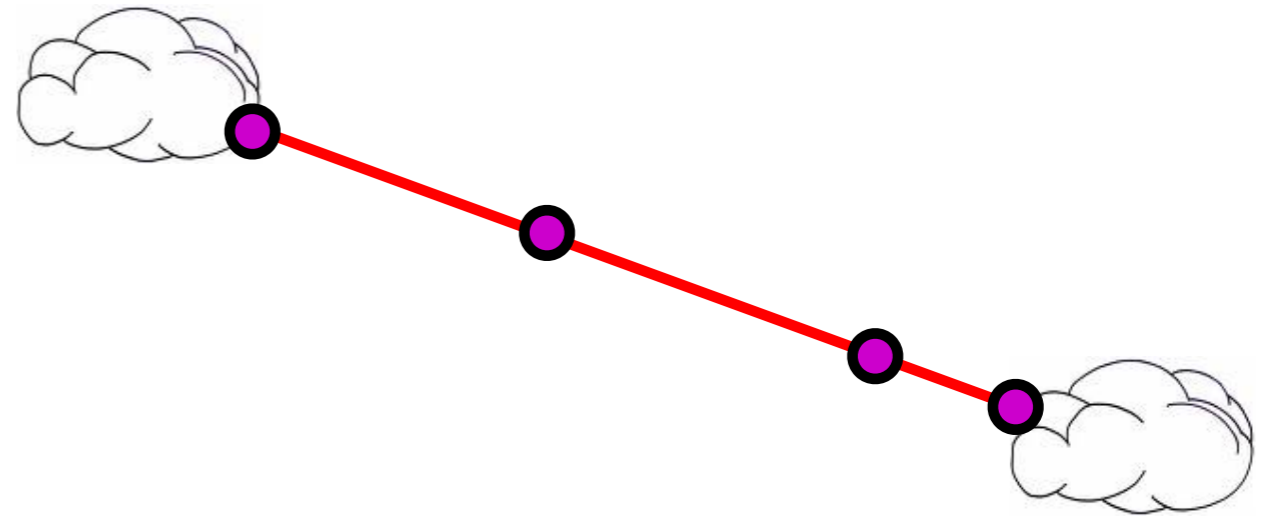
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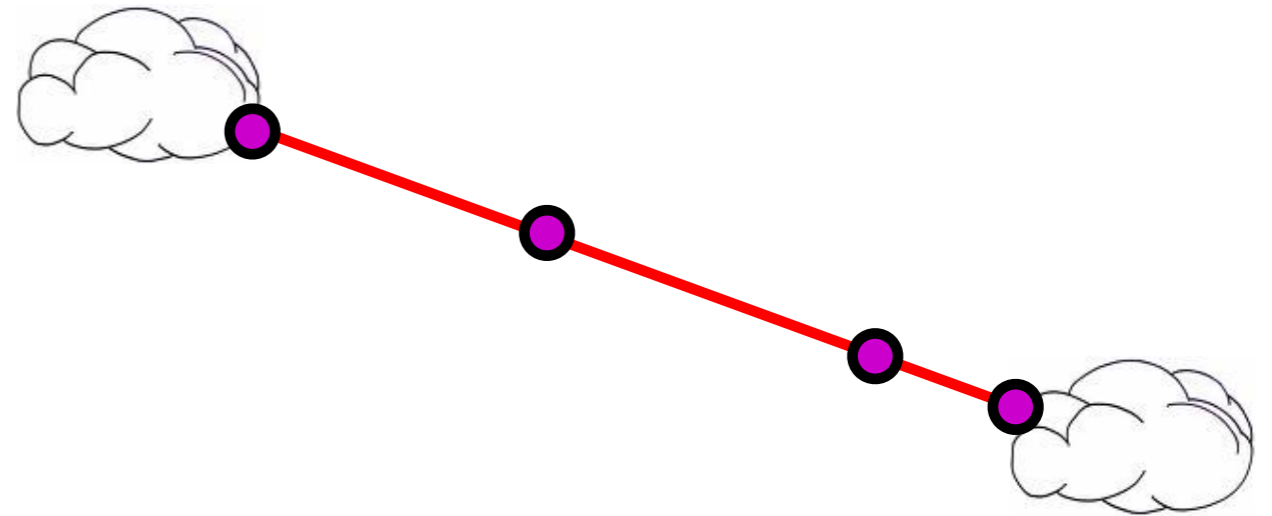
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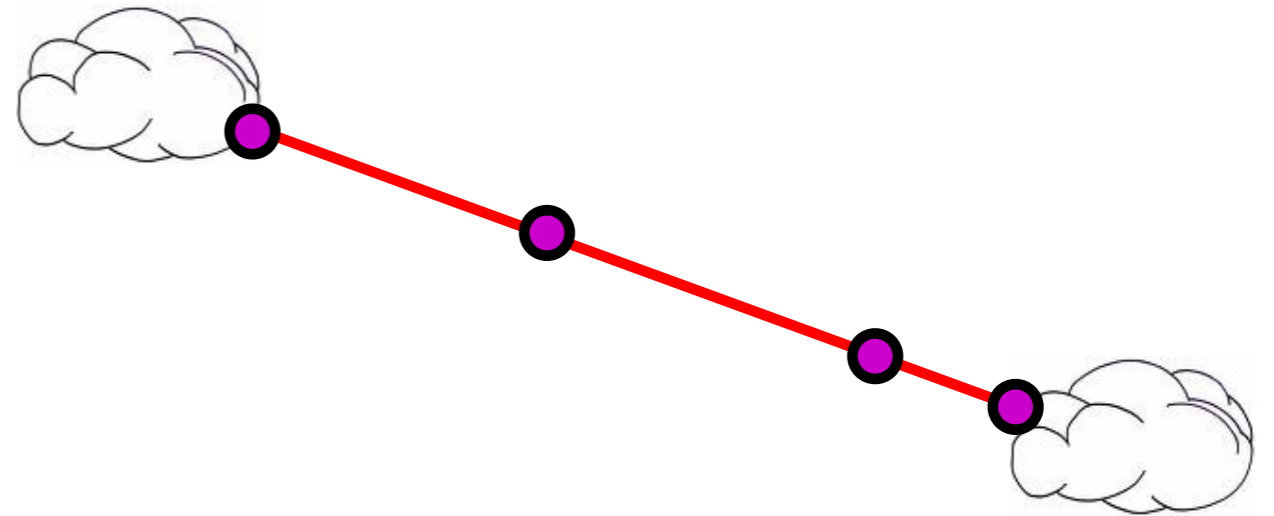
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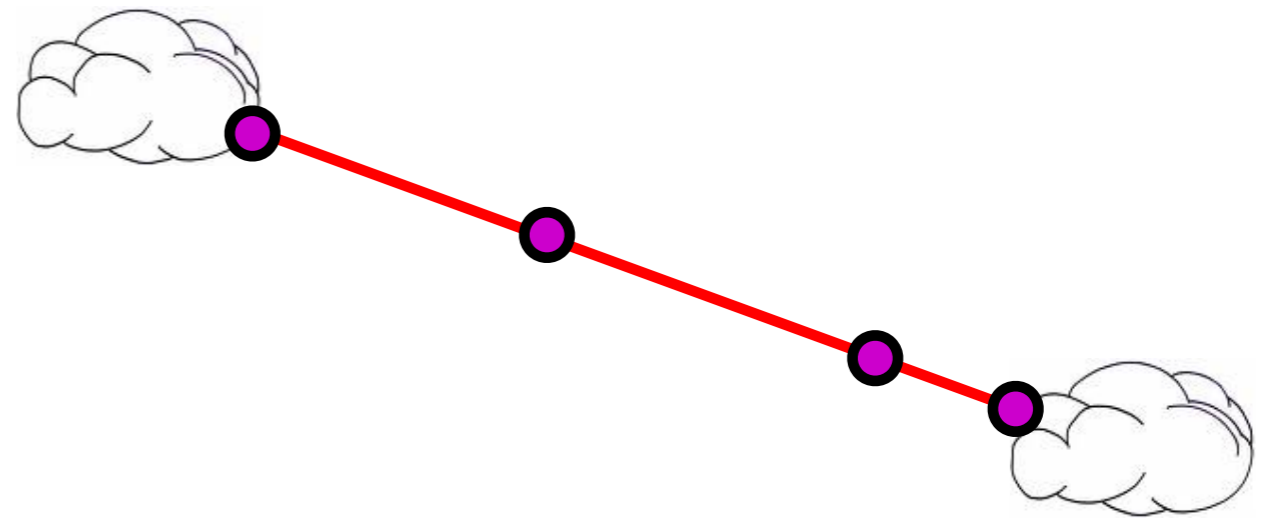
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$$\begin{aligned} |R''| &\leq 2(|\mathcal{C}|-1) + \sum_e \left\lfloor \frac{|e|}{r} \right\rfloor \\ &\leq 2|\mathcal{C}| + \frac{|\text{MSFN}(\mathcal{C})|}{r} \\ &\leq 2|\mathcal{C}| + \frac{2}{\sqrt{3}} \cdot \frac{|\text{MStFN}(\mathcal{C})|}{r} \end{aligned}$$

# Simple 6.73-Approximation (III)

- Compute  $\text{MSFN}(\mathcal{C})$ .
- Put relays  $R''$  along edges:
  - 2 relays at end of each edge
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$$\begin{aligned} |R''| &\leq 2(|\mathcal{C}|-1) + \sum_e \left\lfloor \frac{|e|}{r} \right\rfloor \\ &\leq 2|\mathcal{C}| + \frac{|\text{MSFN}(\mathcal{C})|}{r} \\ &\leq 2|\mathcal{C}| + \frac{2}{\sqrt{3}} \cdot \frac{|\text{MStFN}(\mathcal{C})|}{r} \end{aligned}$$

# **Simple 6.73-Approximation (IV)**



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$$|R'| + |R''| \leq$$

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$$|R'| + |R''| \leq \text{[redacted]}$$

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$$|R'| + |R''| \leq \text{[red box]} \quad \text{[blue box]}$$

# Simple 6.73-Approximation (IV)

$$|R'| + |R''| \leq \frac{137}{30} |\text{Stab}(B)| - |C| + 2|C| + \frac{2}{\sqrt{3}} \cdot \frac{|\text{MStFN}(C)|}{r}$$

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$$\begin{aligned} |R'| + |R''| &\leq \frac{137}{30} |\text{Stab}(B)| - |C| + 2|C| + \frac{2}{\sqrt{3}} \cdot \frac{|\text{MStFN}(C)|}{r} \\ &\leq \left( \frac{137}{30} + 1 + \frac{2}{\sqrt{3}} \right) |R^*| \end{aligned}$$

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$$\begin{aligned} |R'| + |R''| &\leq \frac{137}{30} |\text{Stab}(B)| - |C| + 2|C| + \frac{2}{\sqrt{3}} \cdot \frac{|\text{MStFN}(C)|}{r} \\ &\leq \left( \frac{137}{30} + 1 + \frac{2}{\sqrt{3}} \right) |R^*| \\ &< 6.73 |R^*| \end{aligned}$$



# **Simple 6.73-Approximation (V)**

---

# Simple 6.73-Approximation (V)

- Running time:
  - Build Delaunay of  $n$  sensors  $O(n \log n)$
  - Identify blobs and clouds  $O(n)$
  - Greedy to stab blobs ( $\leq 5$  per relay)  $O(n)$
  - MSFN(C) is subgraph of Delaunay  $O(n \log n)$
  
  - Total:  $O(n \log n)$

# A 3.11-Approximation

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**Overview:**

# A 3.11-Approximation

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- Compute *optimal* stabbings for clouds of blobs that can be stabbed with few ( $\leq k$ , constant) relays  
Complete connectivity of these clouds (Lemma 1)

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# Clouds with Few Stabs

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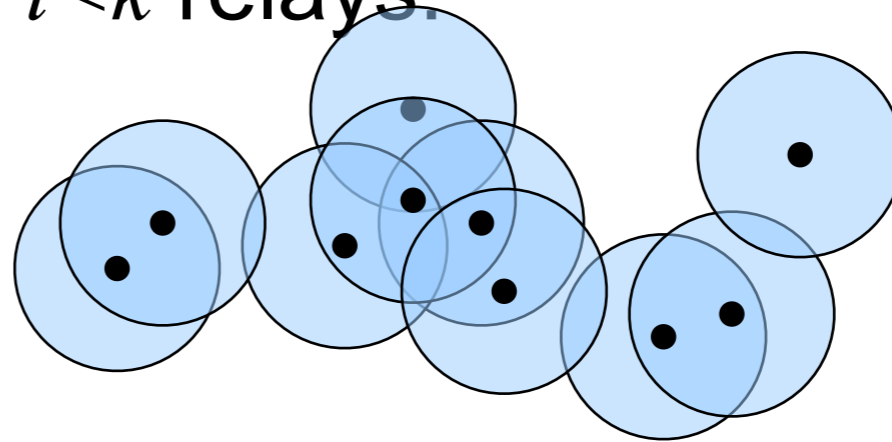
For any constant  $k$ , determine those clouds  $C \in \mathcal{C}_i$  whose blobs can be stabbed with  $i < k$  relays:



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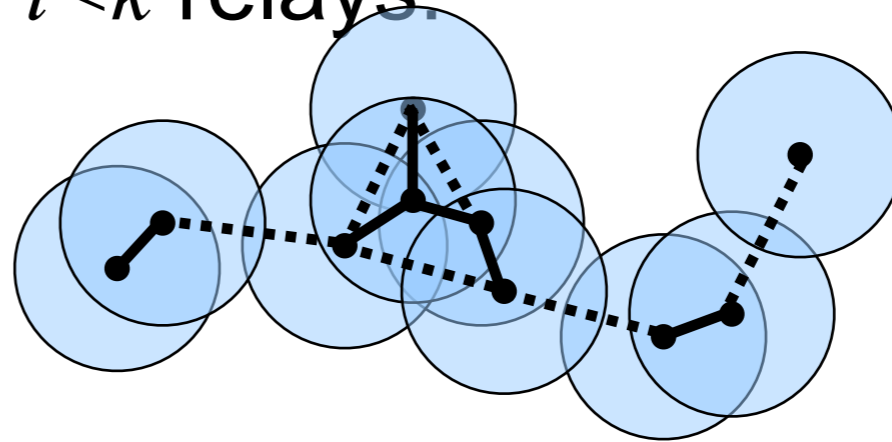
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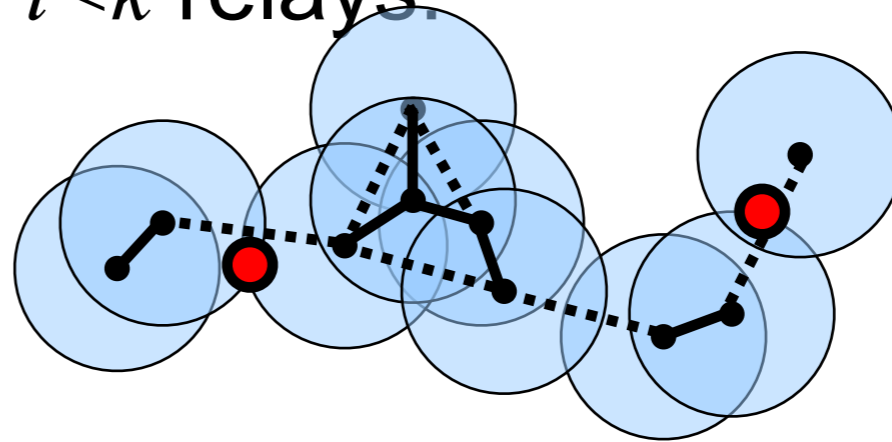
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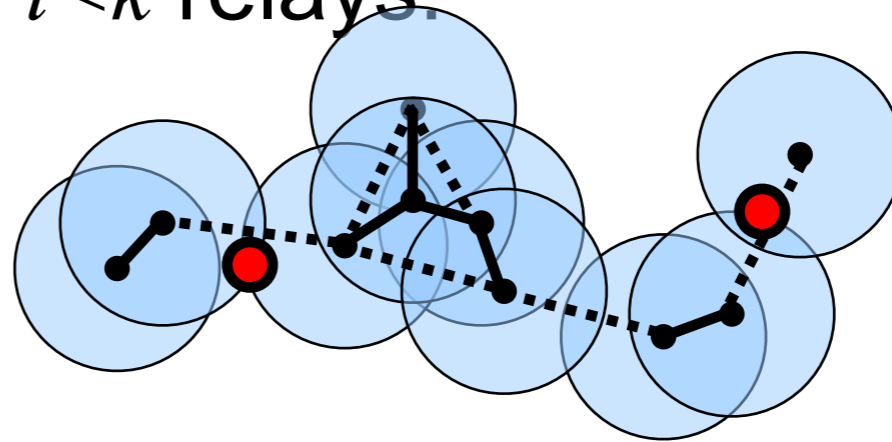
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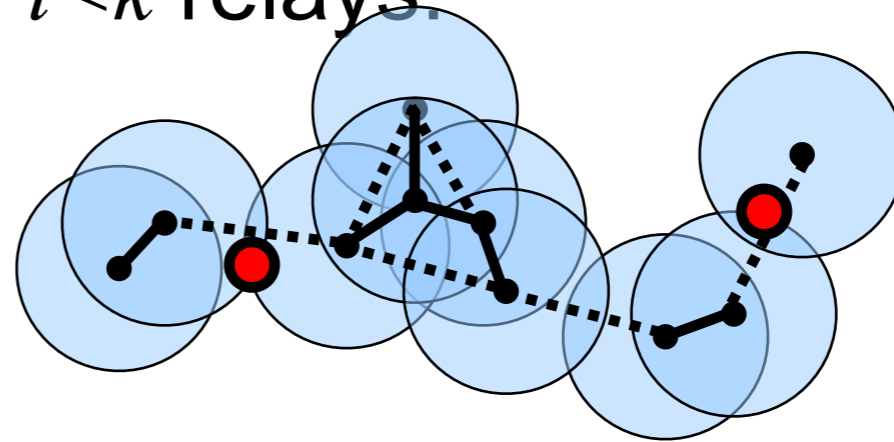


Then use **Lemma 1** to complete connection:

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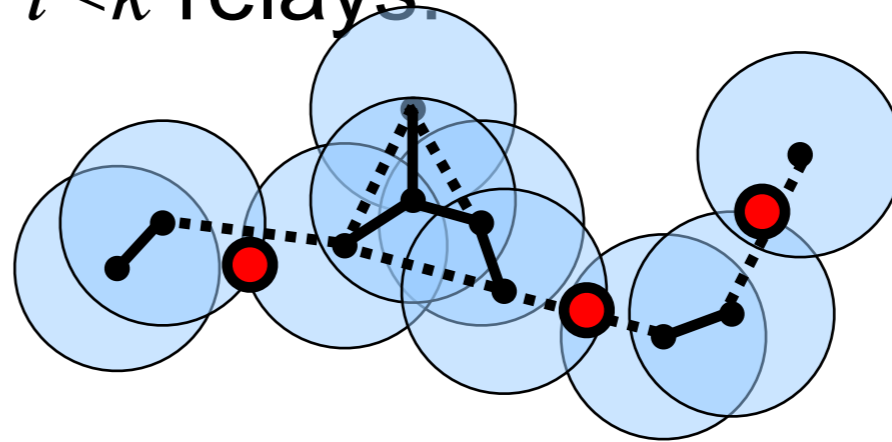
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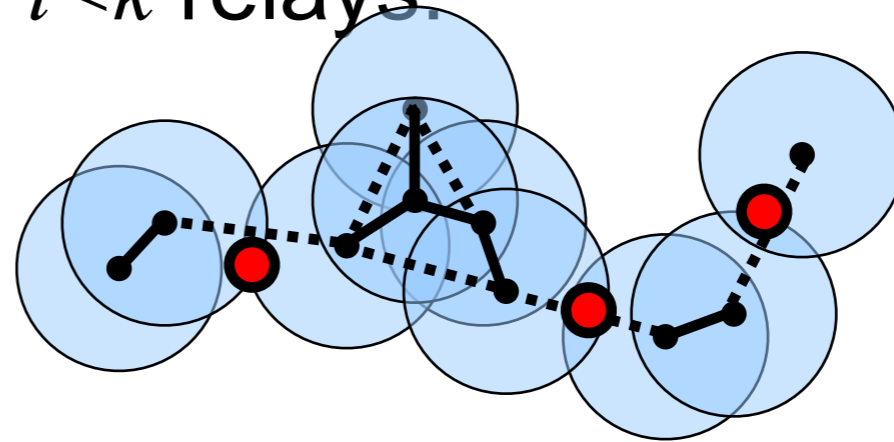
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Let  $\mathcal{C}_{k+}$  be the set of clouds requiring  $\geq k$  stabs

# Stitching a Cloud From $\mathcal{C}^{k+}$

---



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Consider clouds  $C \in \mathcal{C}^{k+}$  that need more than  $k$  stabs.

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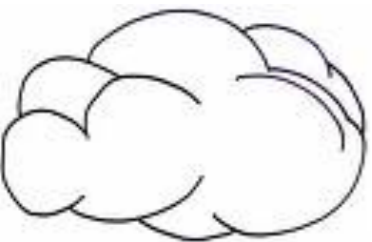
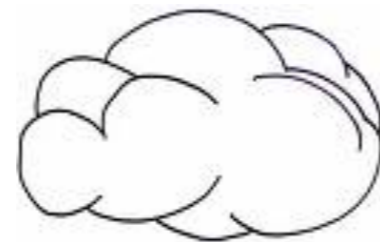
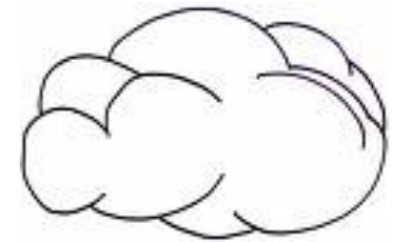
**Proof:** Omitted. Note that  $1 + \frac{1}{4} + \frac{1}{3} + \frac{1}{2} + \frac{1}{1} = \frac{37}{12}$ .

# Cloud Clusters: Green Relays

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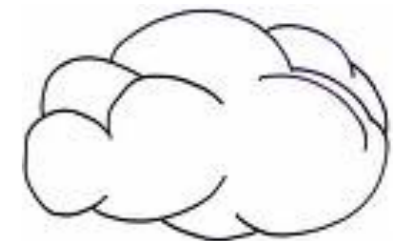
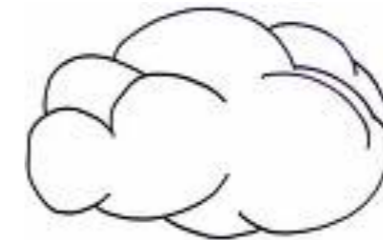
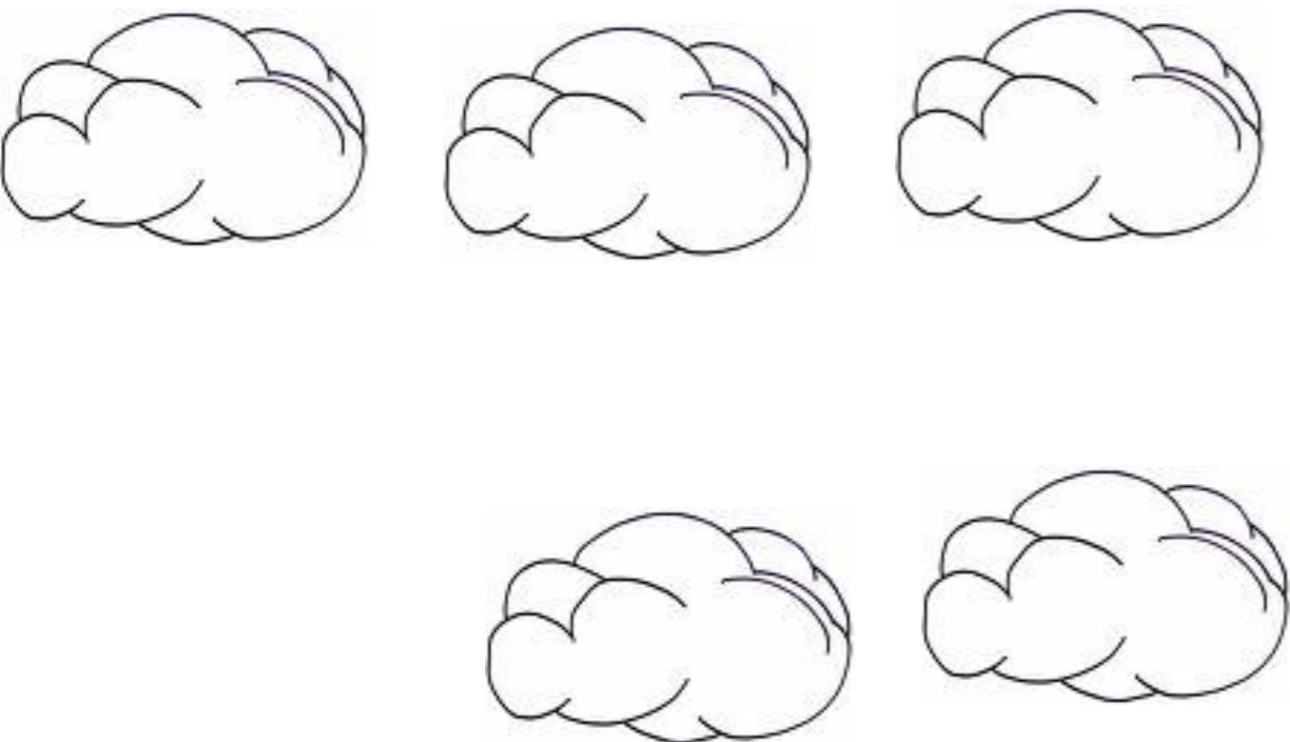




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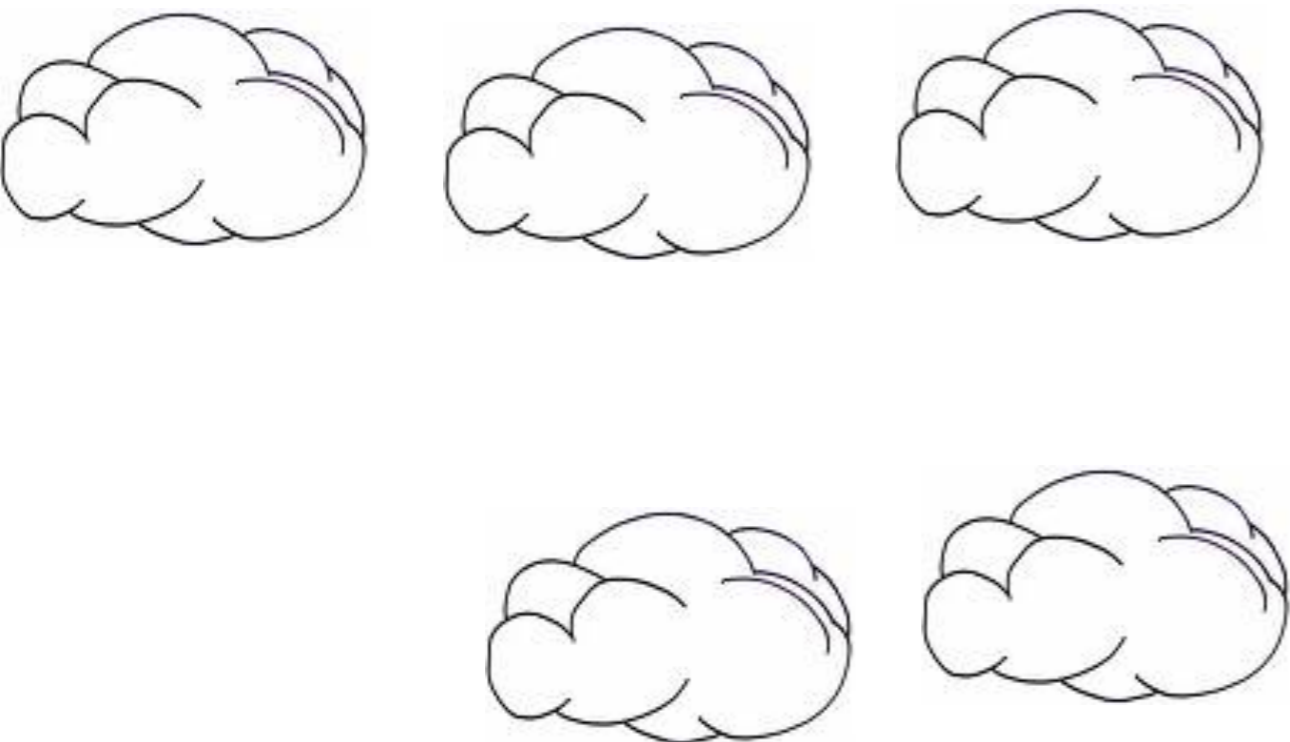
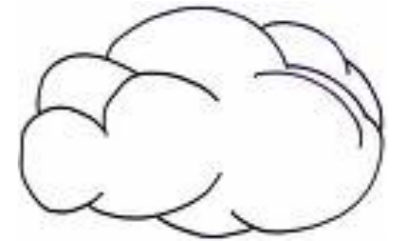
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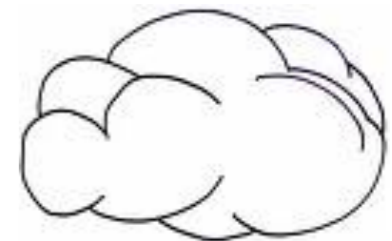
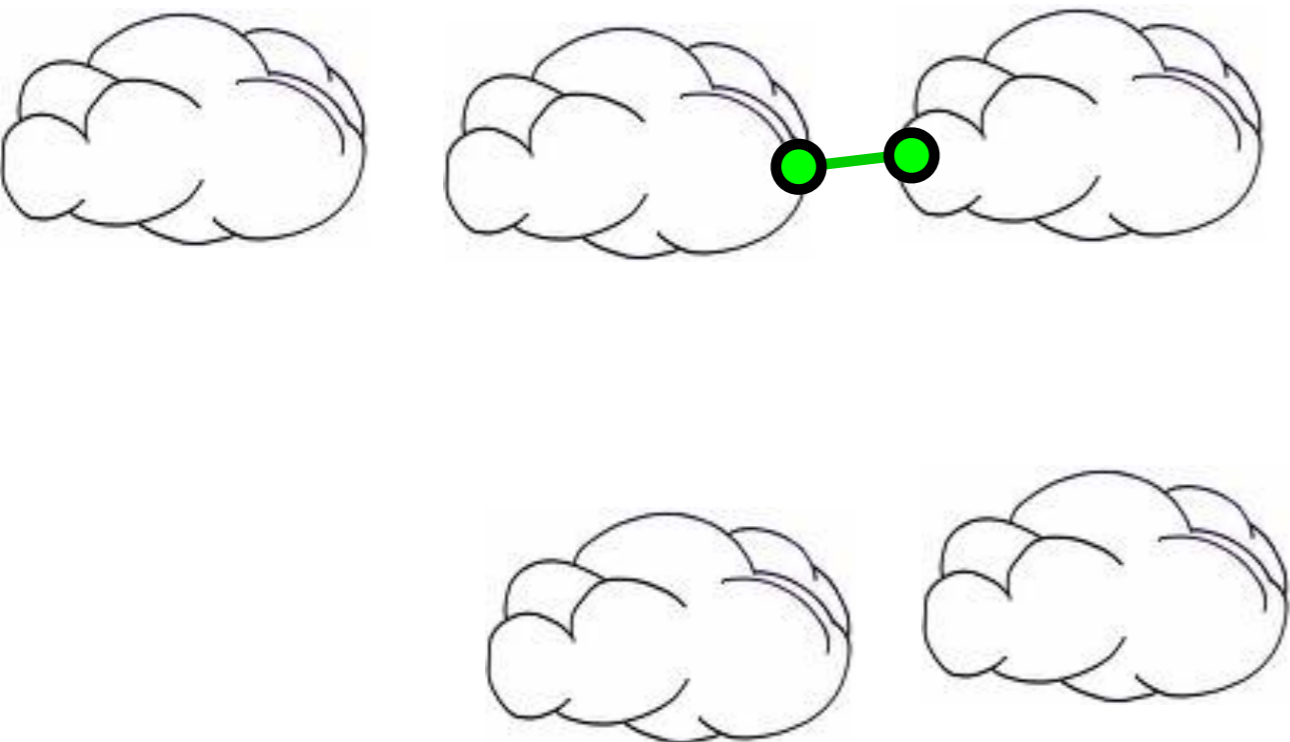
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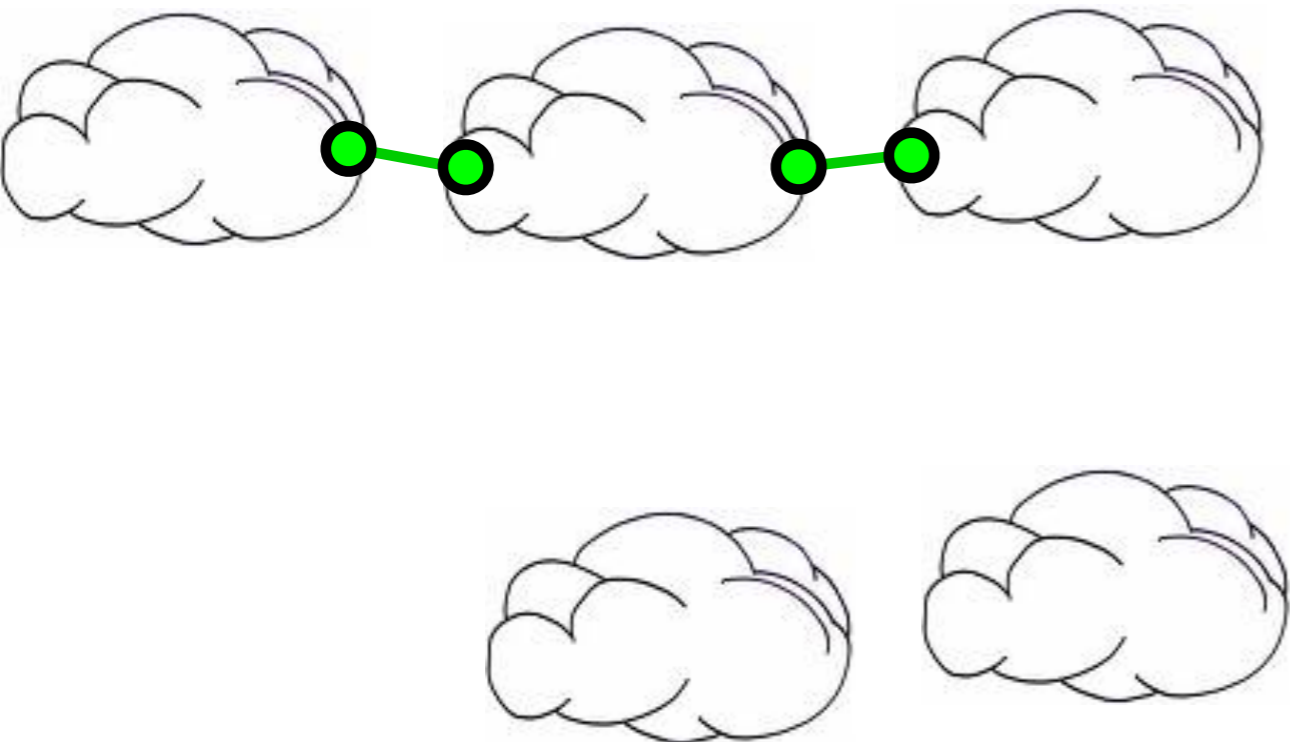
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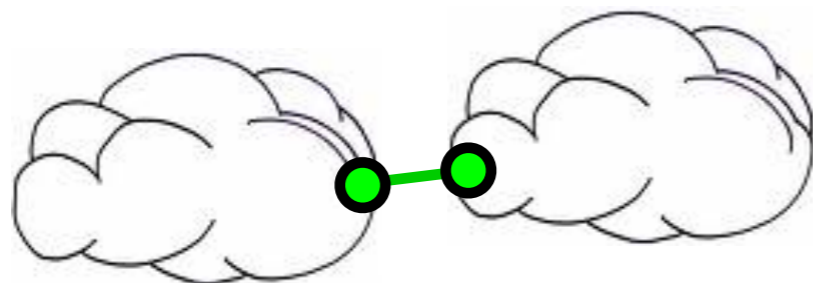
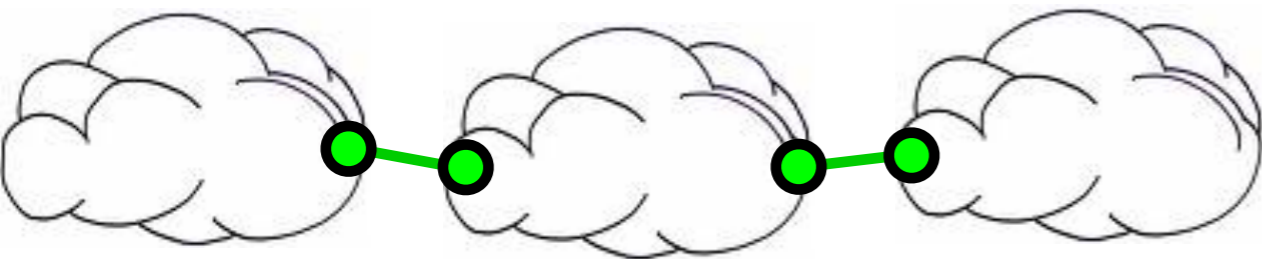
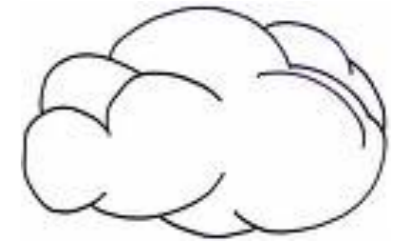
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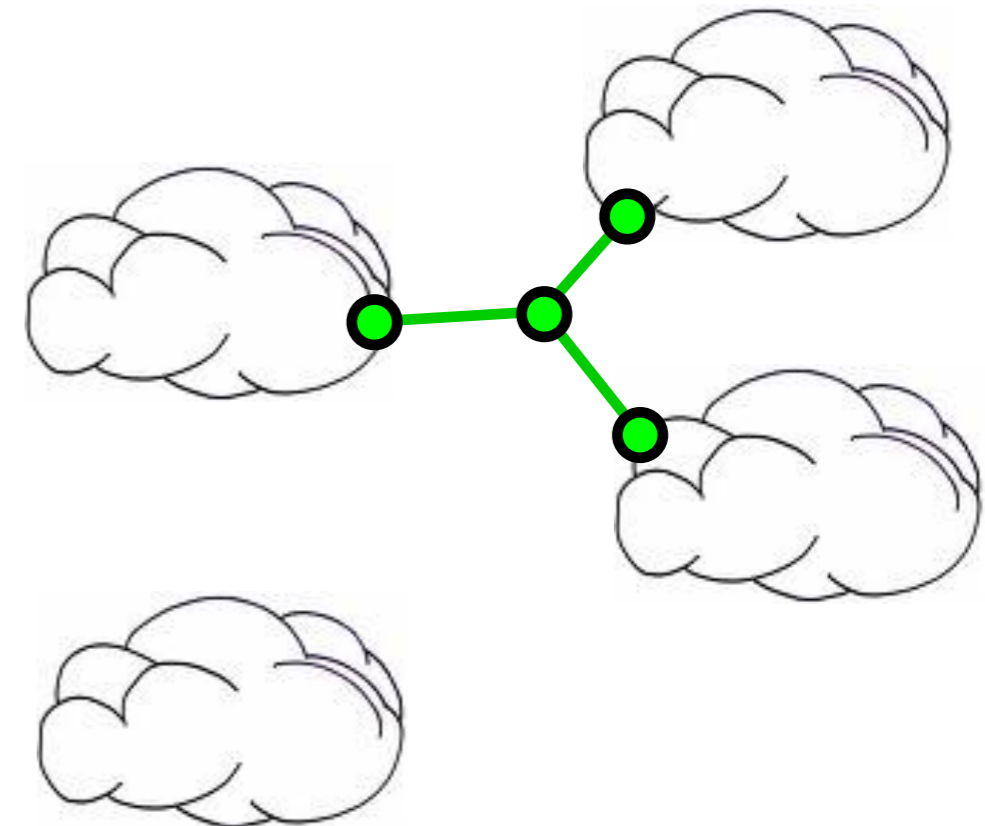
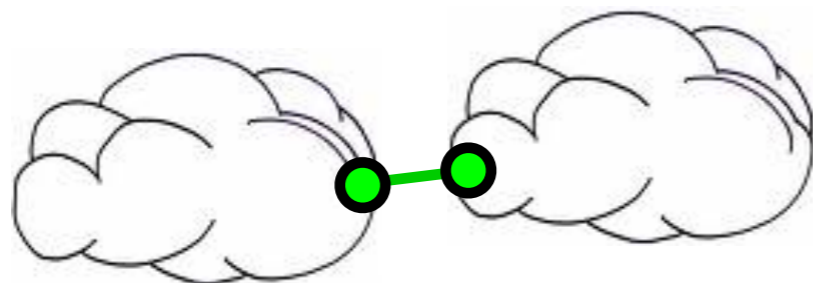
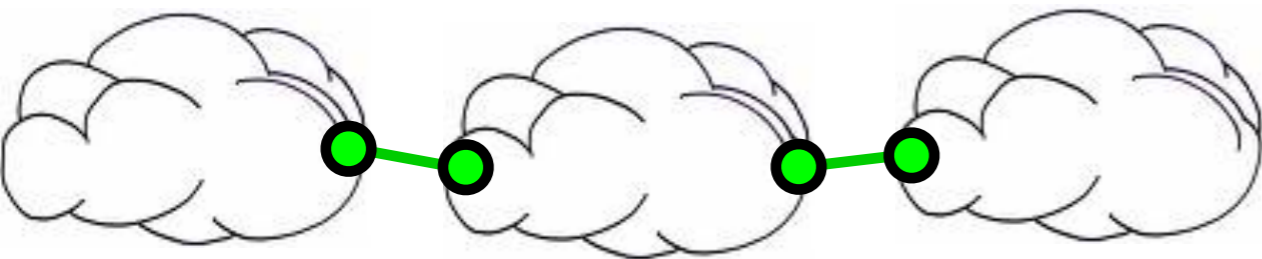
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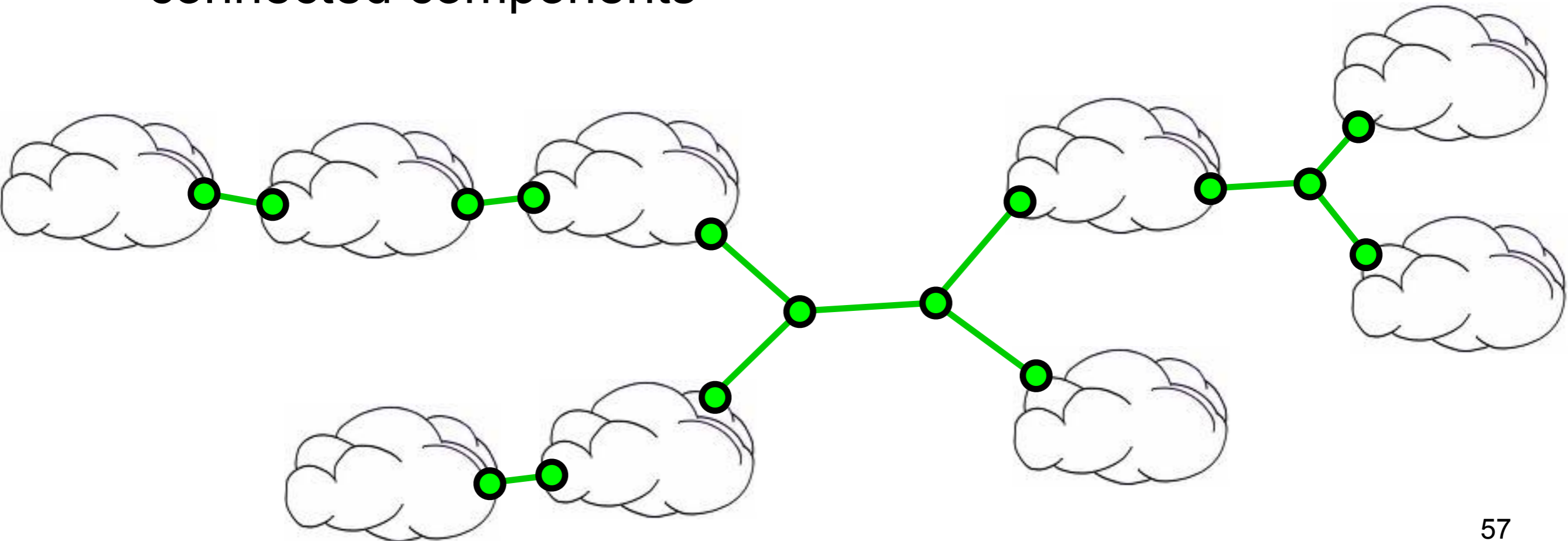
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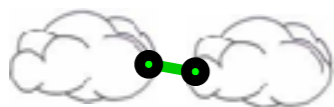
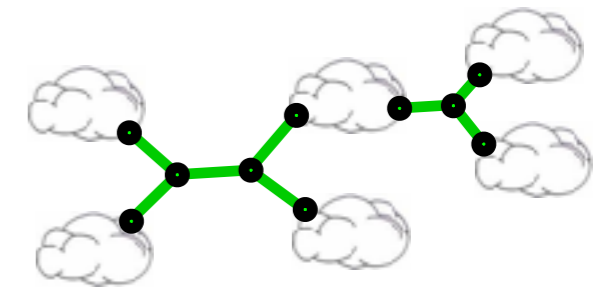
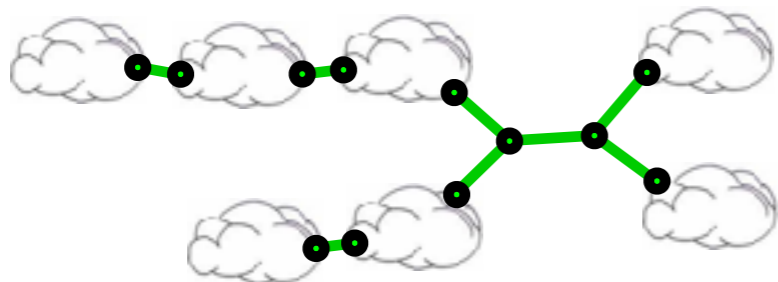
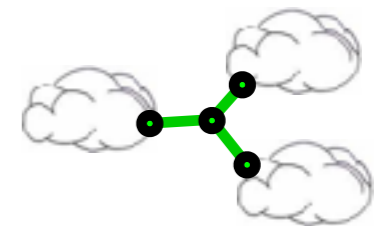
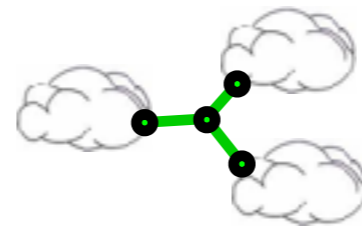
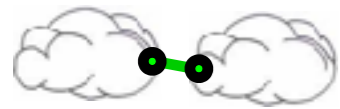
# Interconnecting the Clusters

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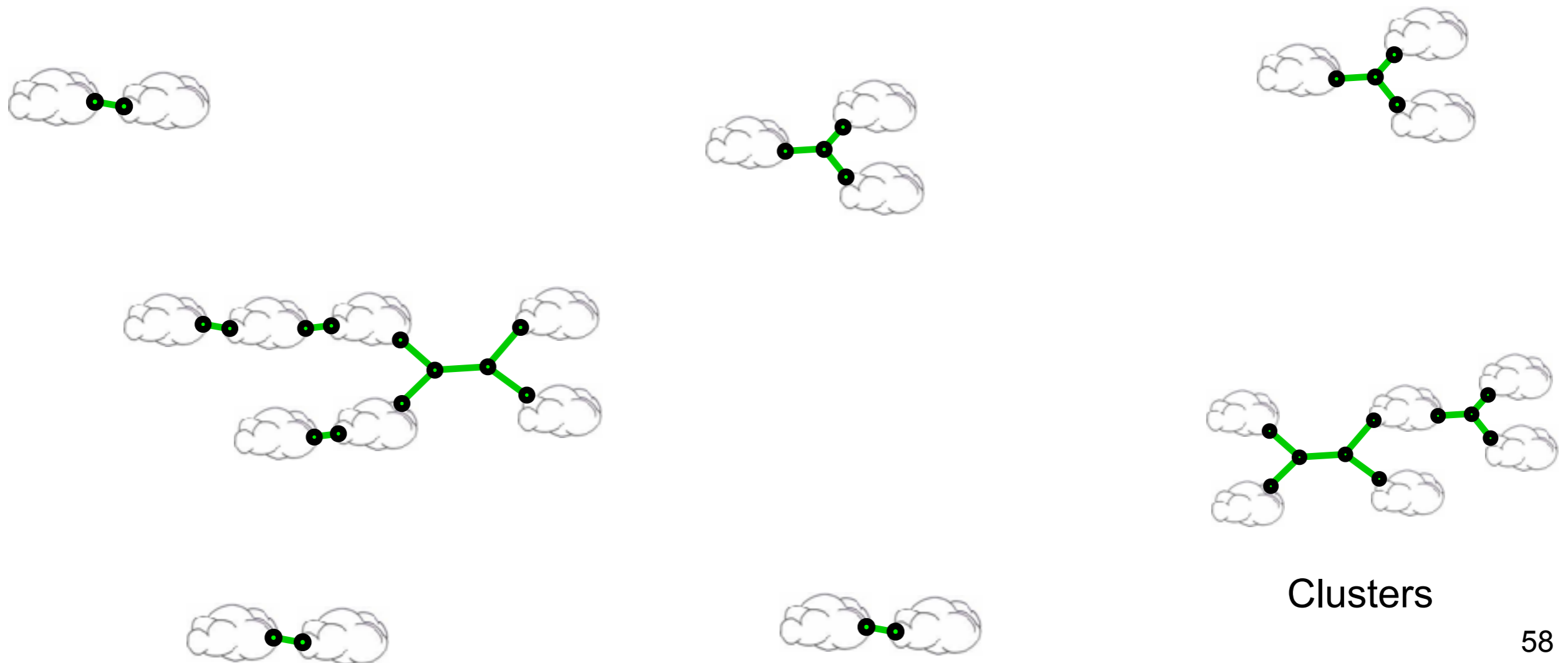


Clusters

# Interconnecting the Clusters

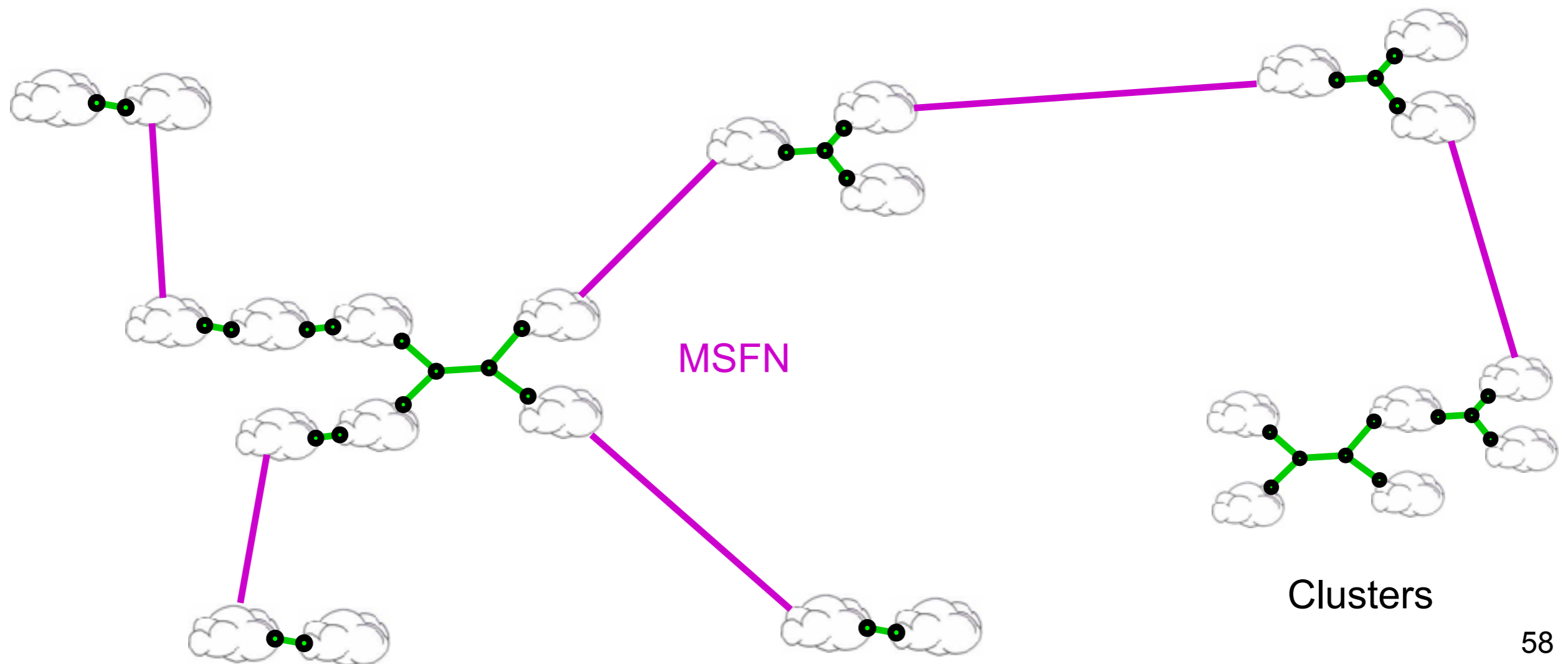
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- Compute MSFN on set of *clusters*



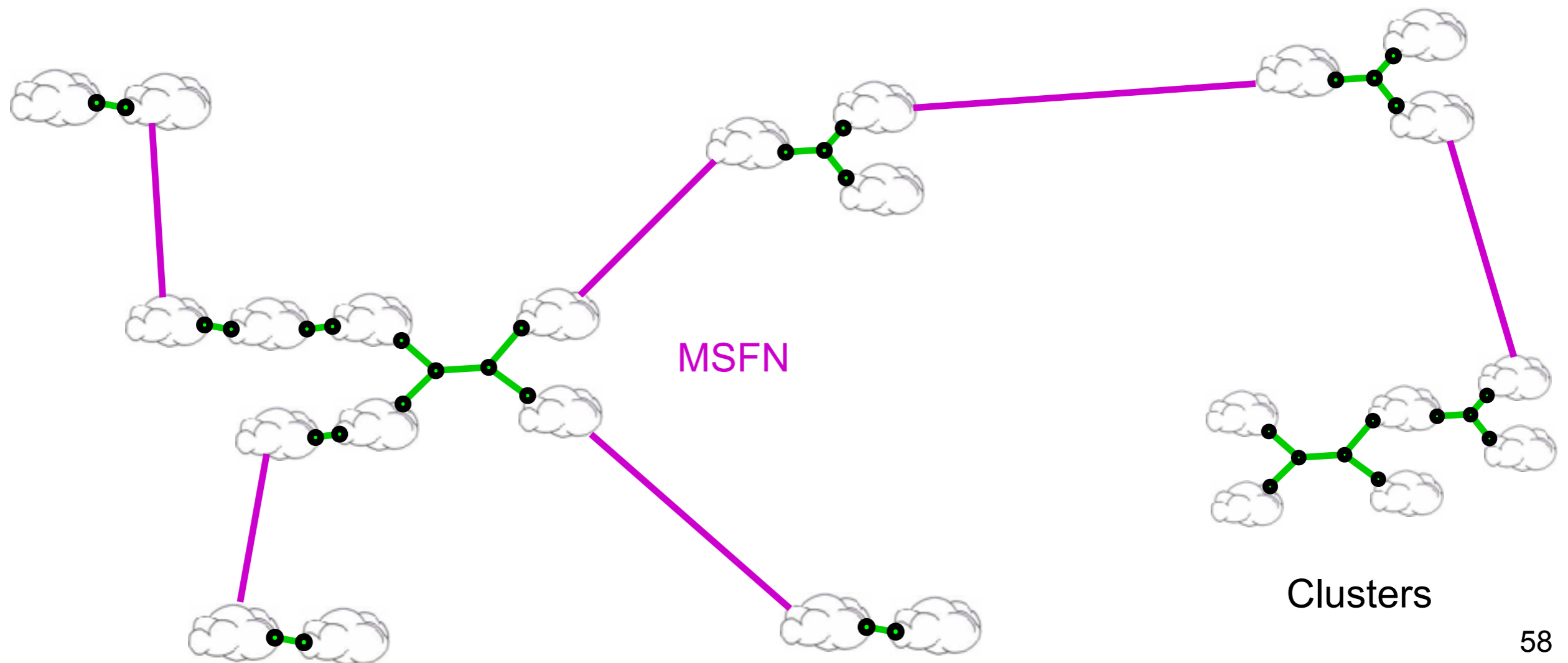
# Interconnecting the Clusters

- Compute MSFN on set of *clusters*



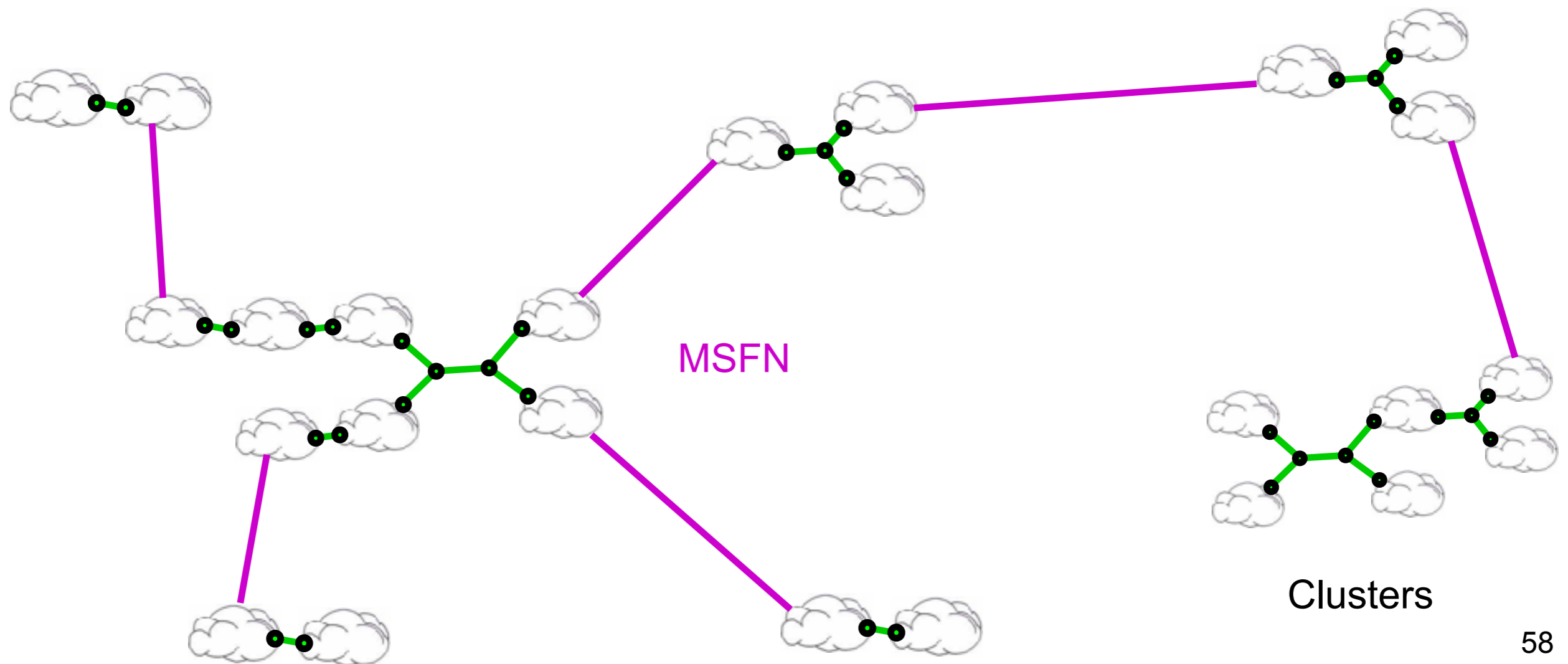
# Interconnecting the Clusters

- Compute MSFN on set of *clusters*
- Put relays along edges:



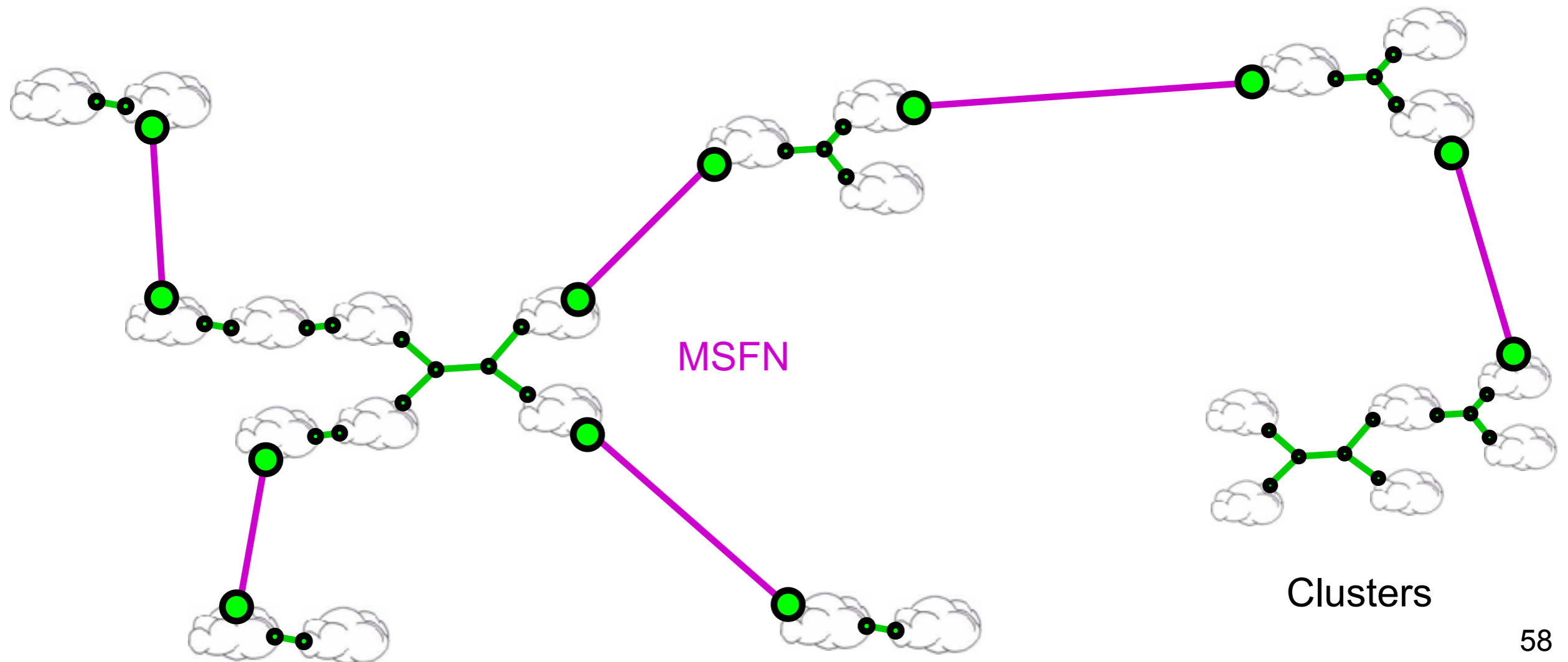
# Interconnecting the Clusters

- Compute MSFN on set of *clusters*
- Put relays along edges:
  - 2 green relays at ends of each edge



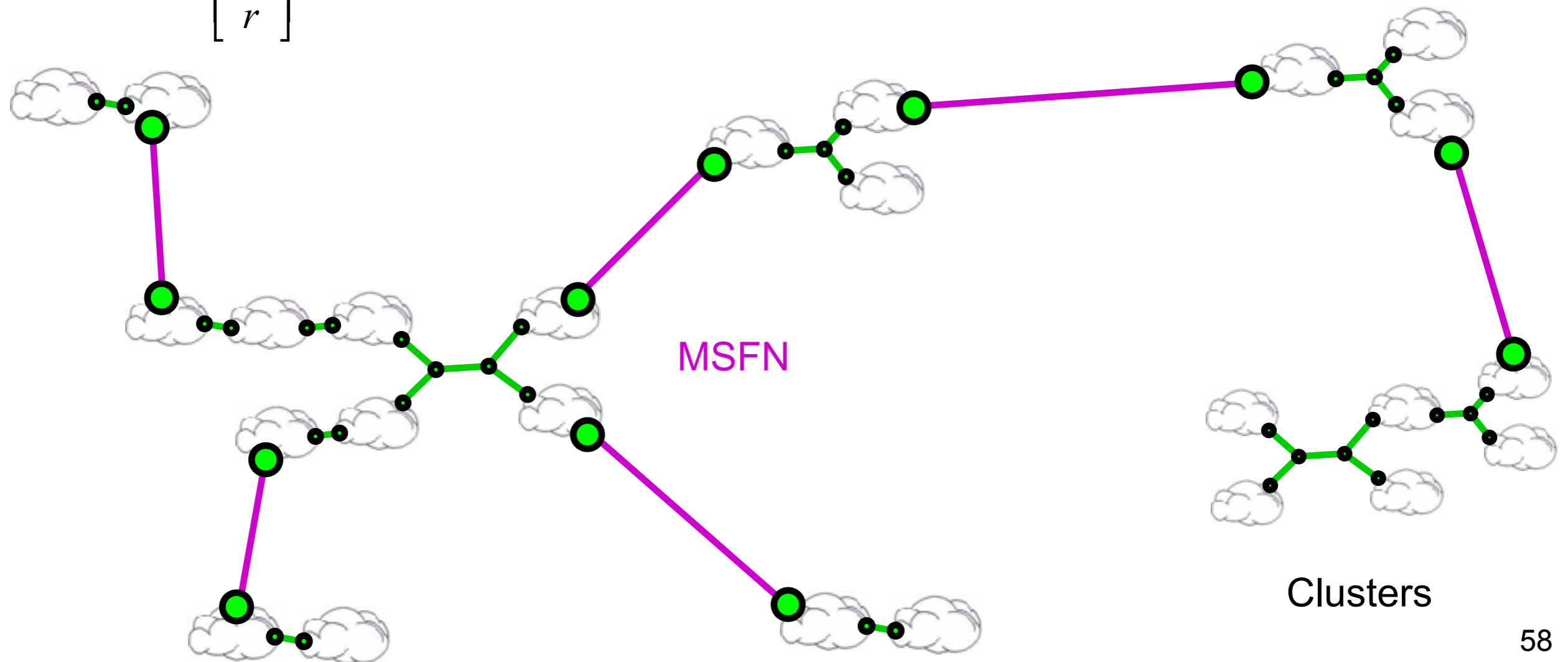
# Interconnecting the Clusters

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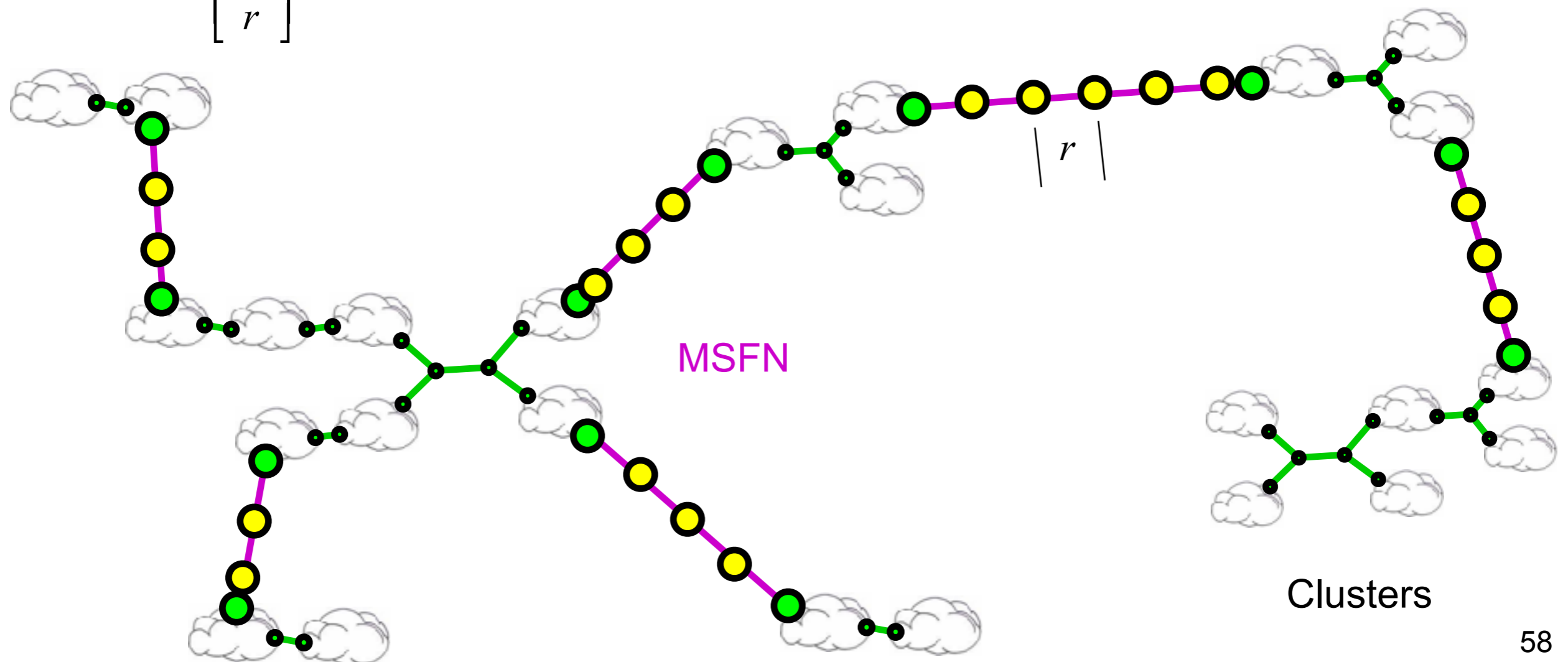
# Interconnecting the Clusters

- Compute MSFN on set of *clusters*
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  - $\left\lceil \frac{|e|}{r} \right\rceil$  yellow relays along each edge  $e$



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# **Analysis: Red and Green Relays**

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dark relays (inside clouds)  
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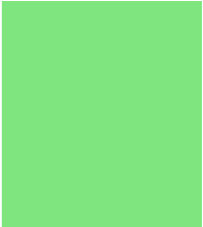
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# **Analysis: Red and Green Relays**

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$$|A_r| + |A_g| \cong \sum_{C \in \mathcal{C}^{k+1}} \left( \frac{37}{12} |R_d^* \cap C| - 1 \right) + \sum_{i=1}^{k-1} (2i-1) |C^i| + 2|C|$$

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 |A_r| + |A_g| &\stackrel{\cong}{=} \sum_{C \in \mathcal{C}^{k+}} \left( \frac{37}{12} |R_d^* \cap C| - 1 \right) + \sum_{i=1}^{k-1} (2i-1) |C^i| + 2|C| \\
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$$|A_r| + |A_g| \cong \sum_{C \in \mathcal{C}^{k+1}} \left( \frac{37}{12} |R_d^* \cap C| - 1 \right) + \sum_{i=1}^{k-1} (2i-1) |C^i| + 2|C|$$

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$$\frac{37}{12} i > (2i+1)$$

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**Proof:** Omitted.

(Combines Steiner tree ratio with degree arguments.)

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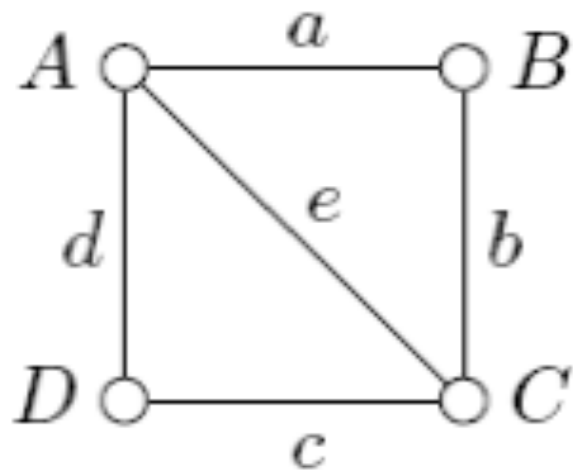
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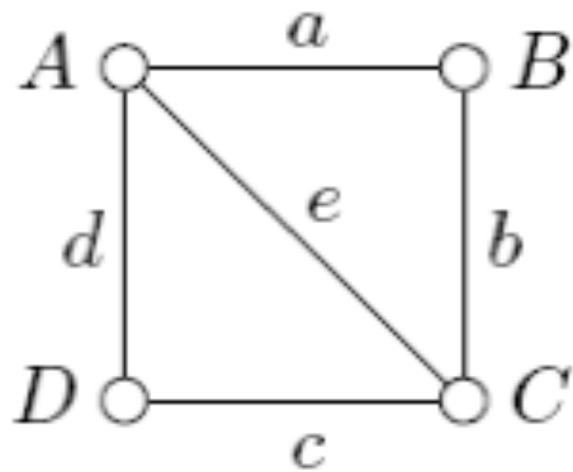
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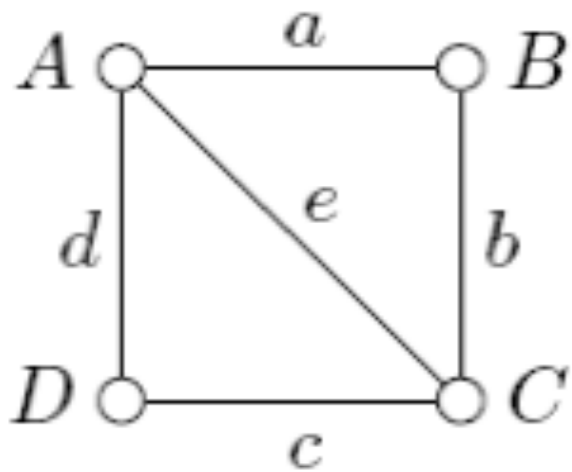
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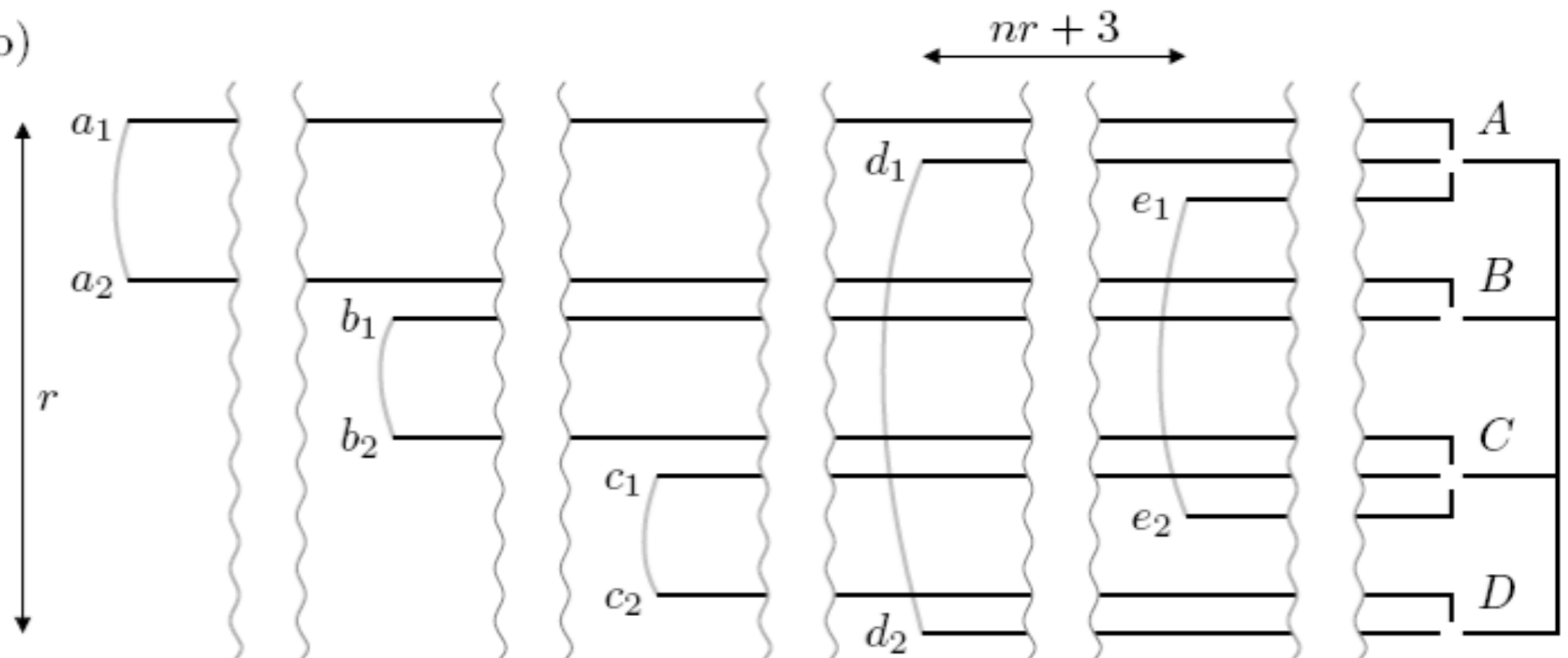
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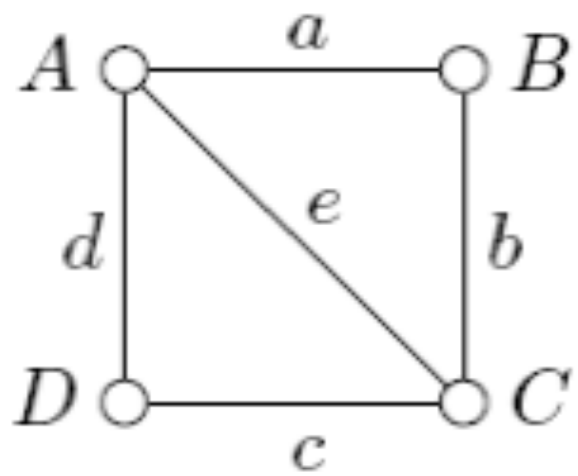
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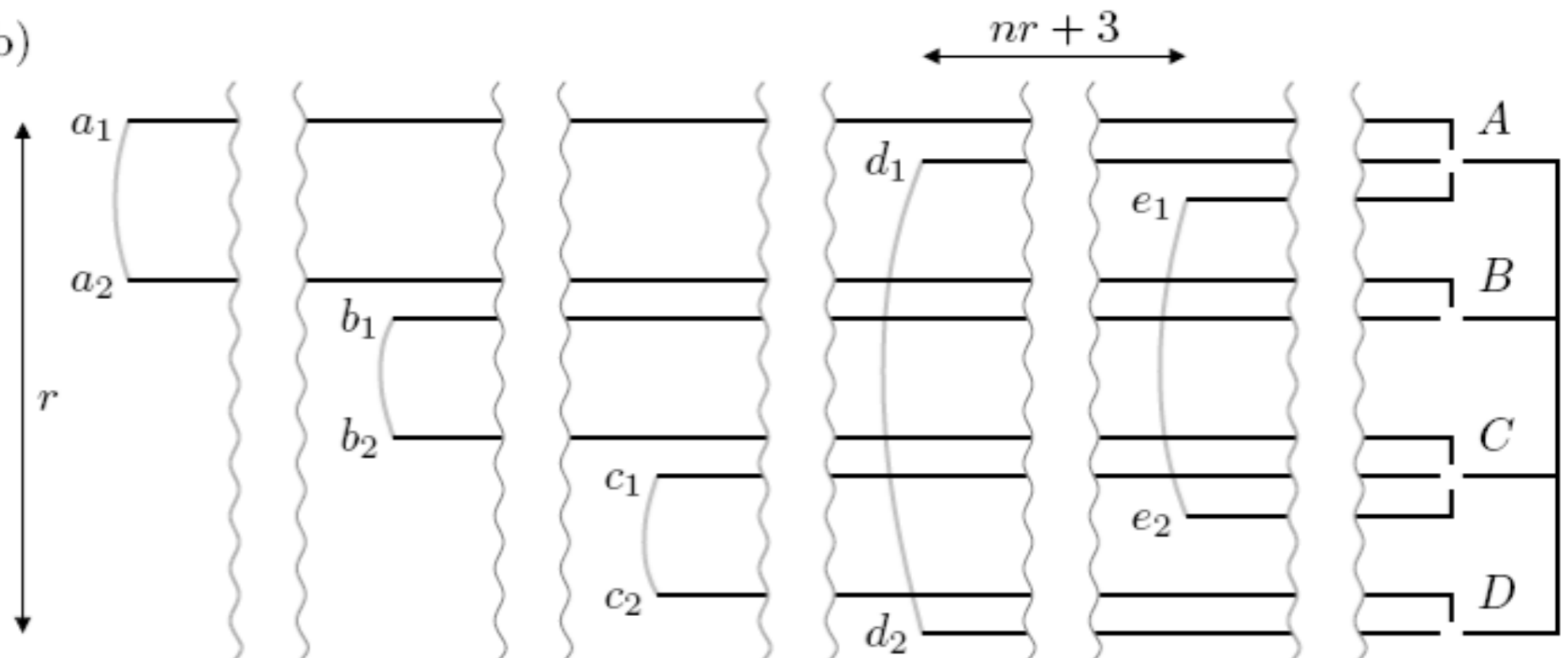


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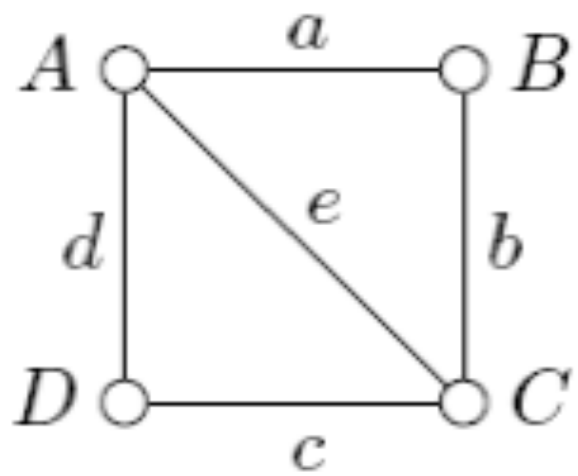


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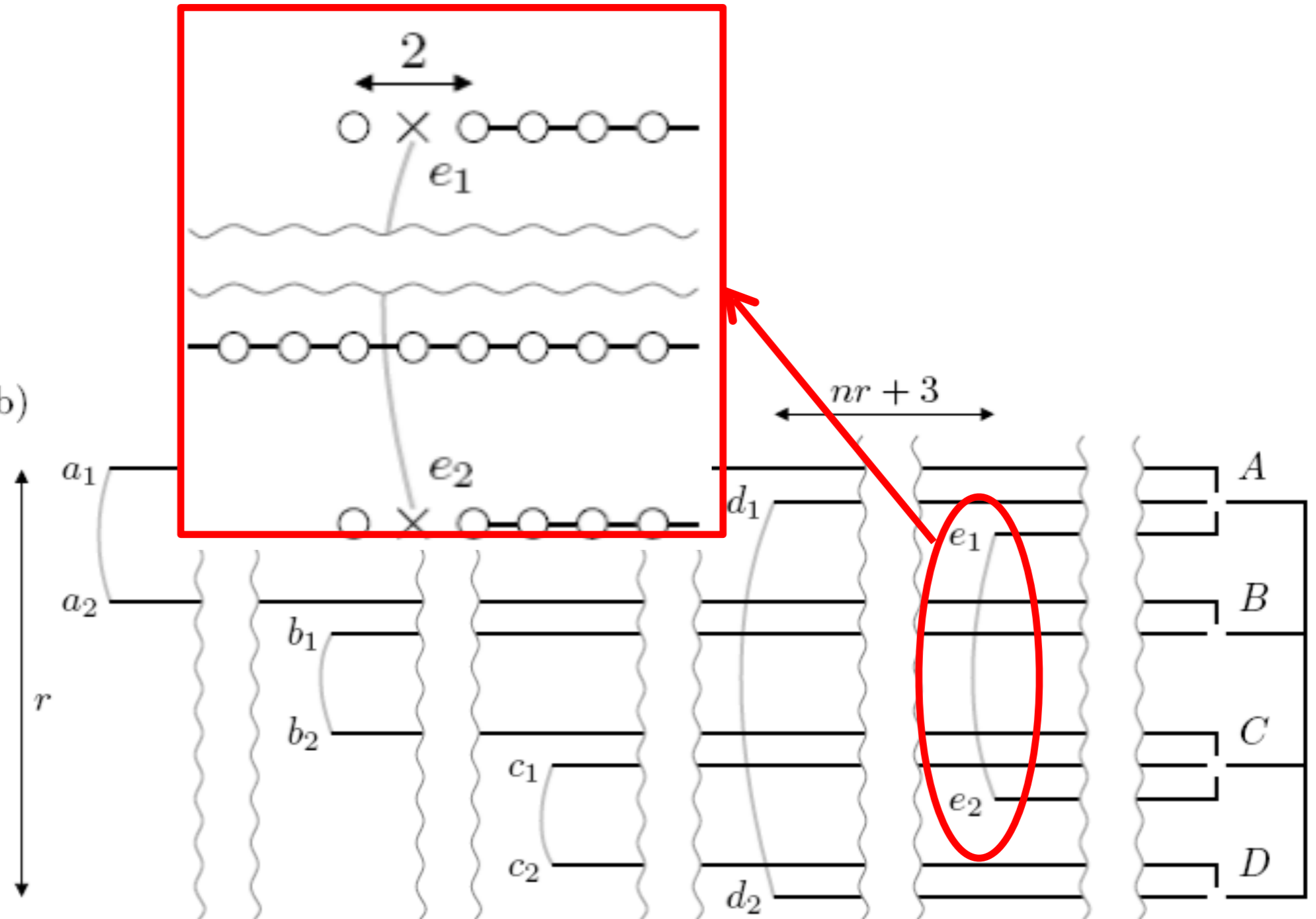


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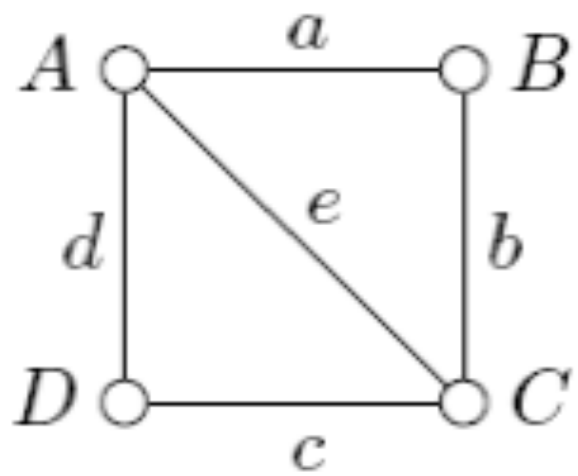
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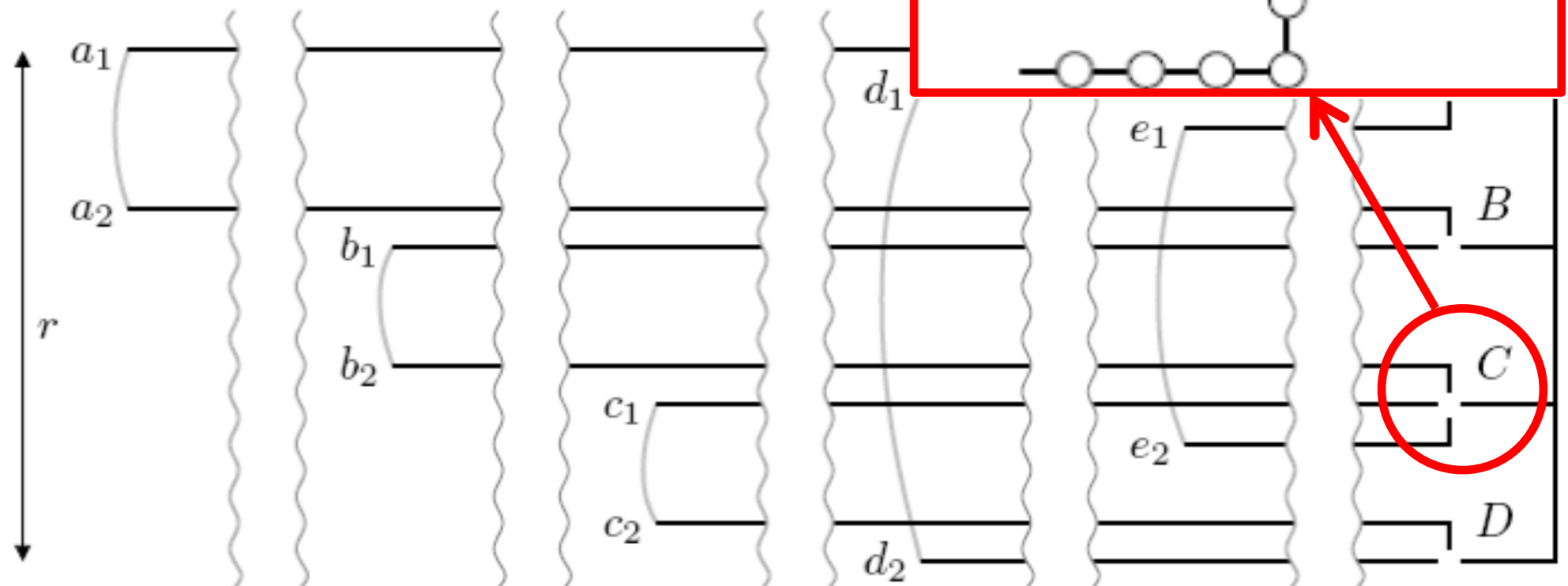


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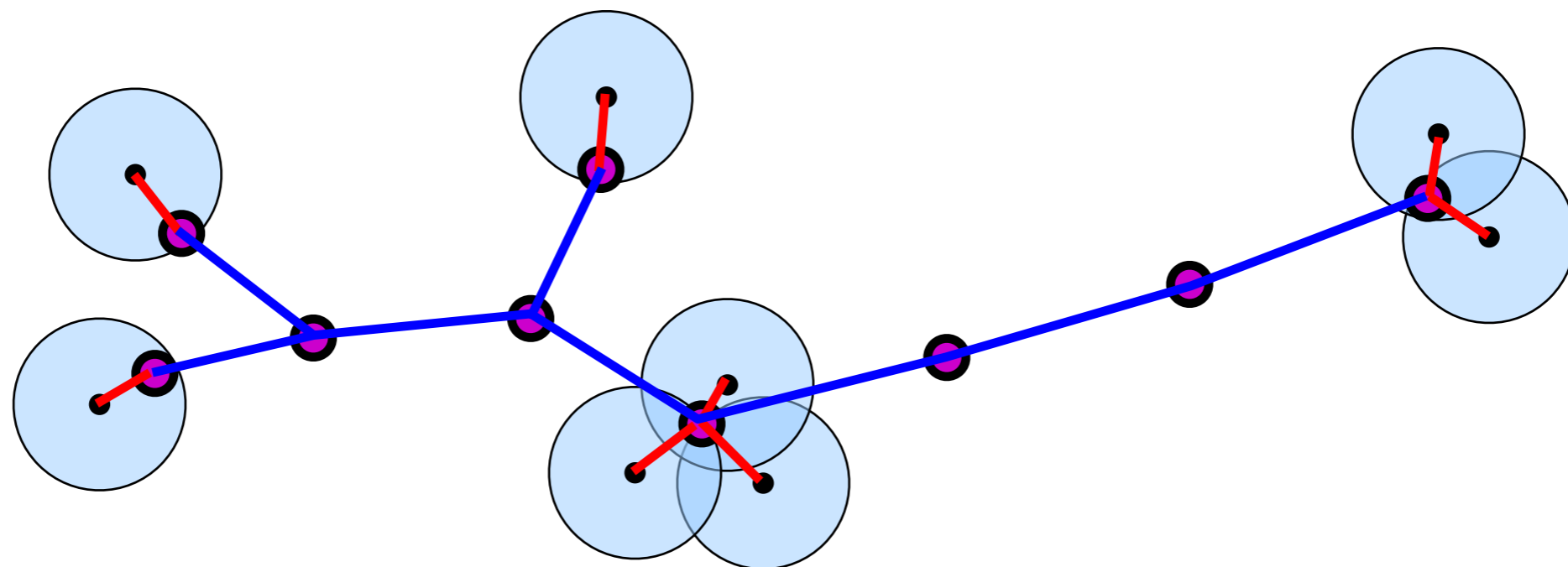


# **PTAS: Two-Tier Relay Placement**

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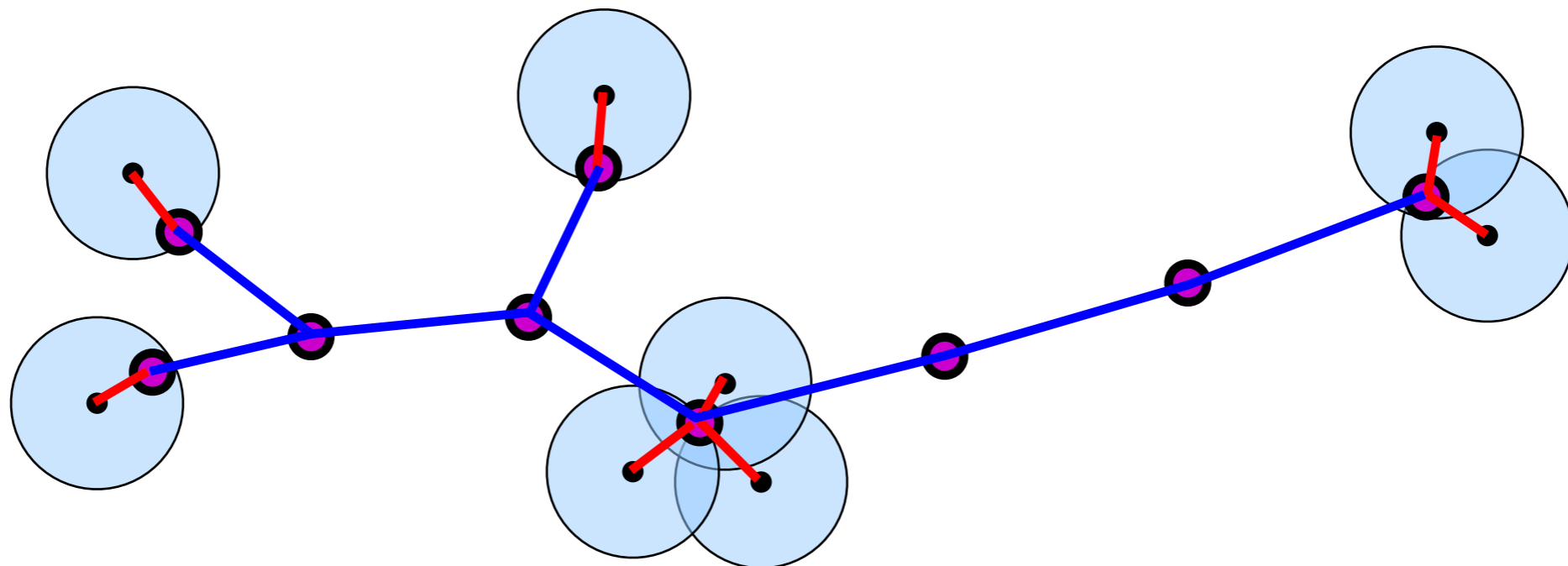
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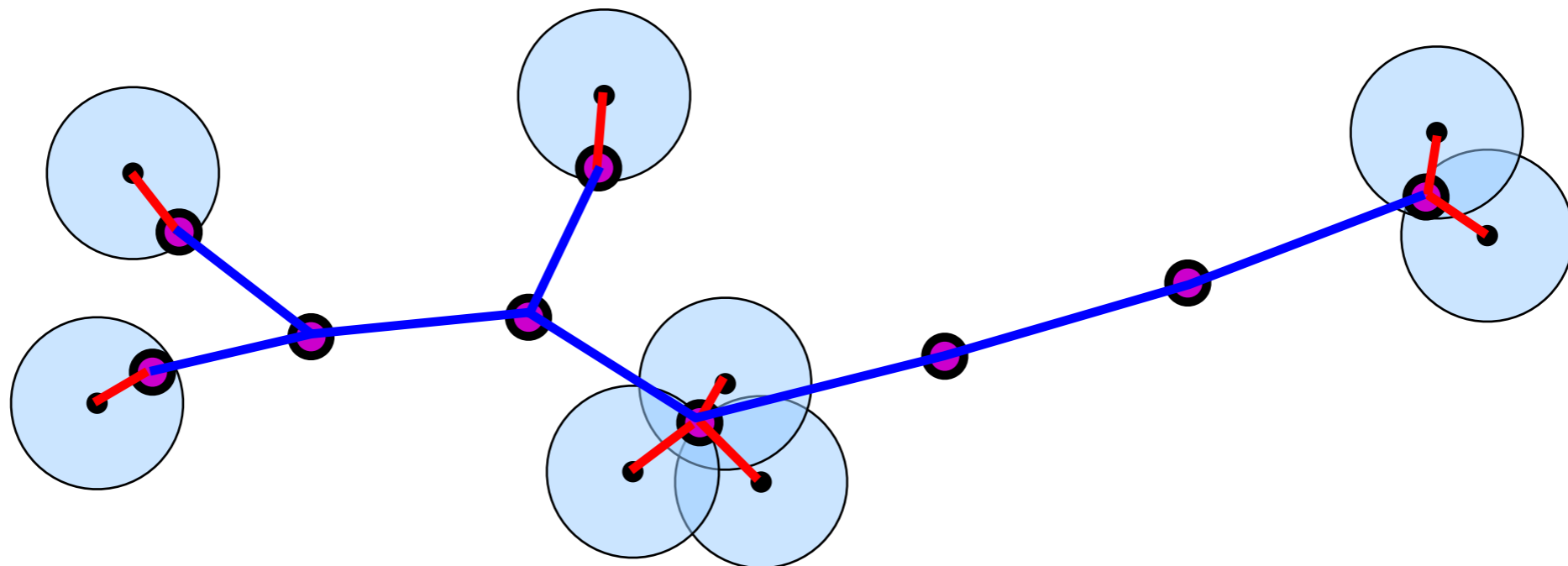
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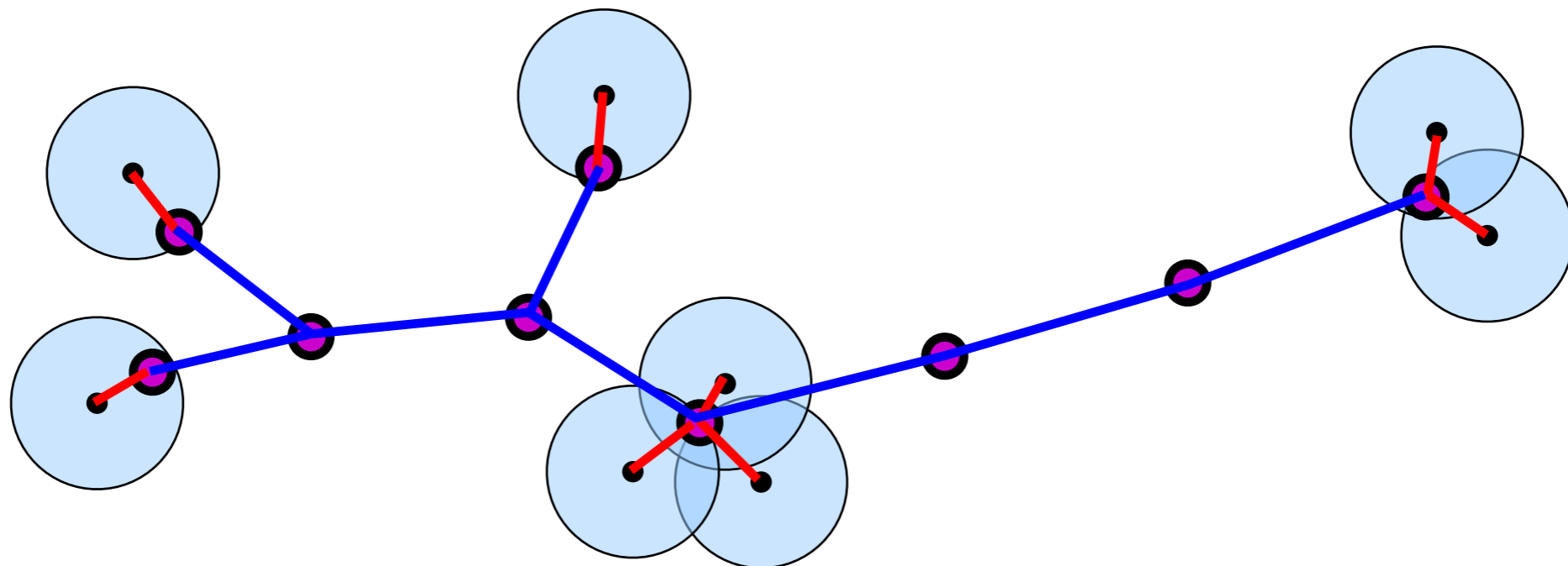
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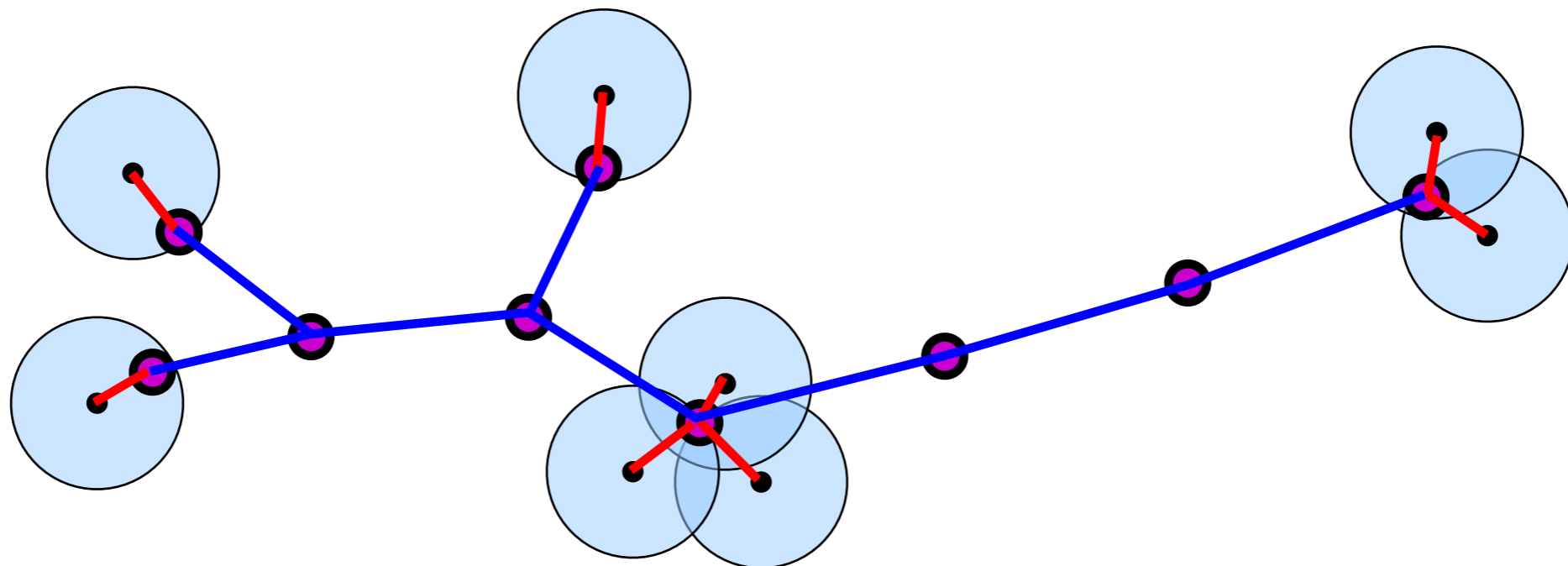
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 $\leq 1$  between a relay and a sensor
- A sensor has degree 1 (cannot relay data)
- Goal: Min # Steiner points (relays)



# PTAS: Outline

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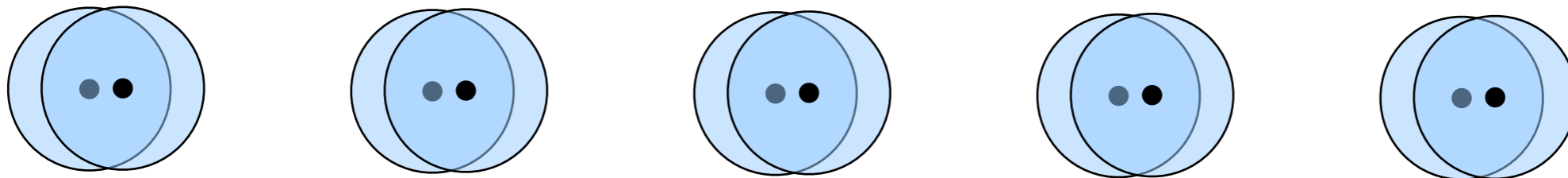
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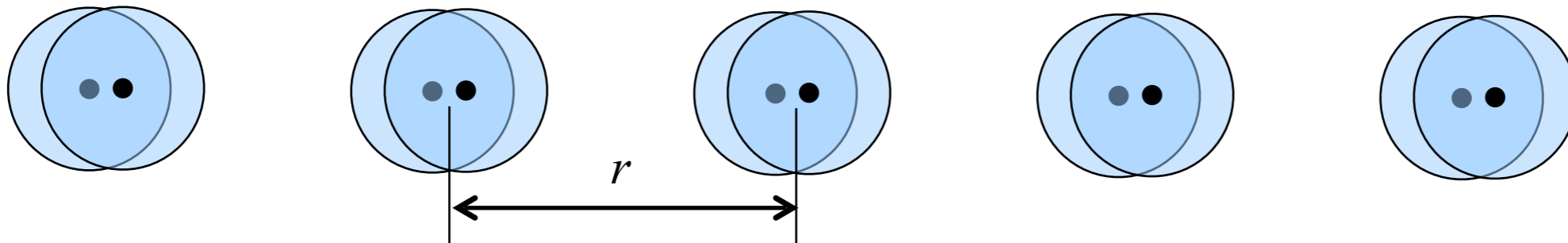
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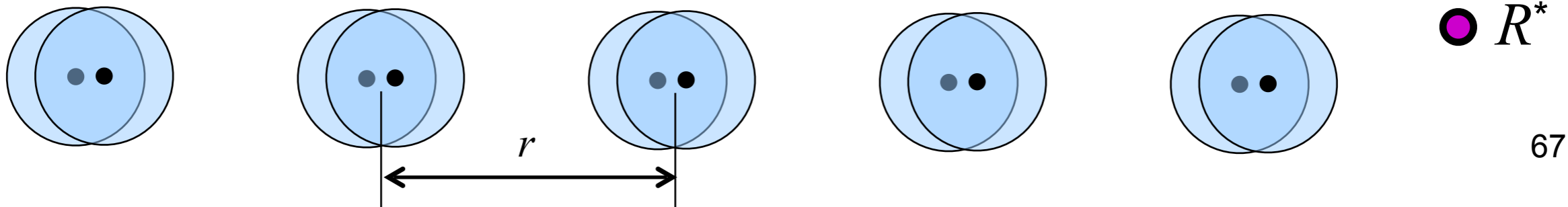
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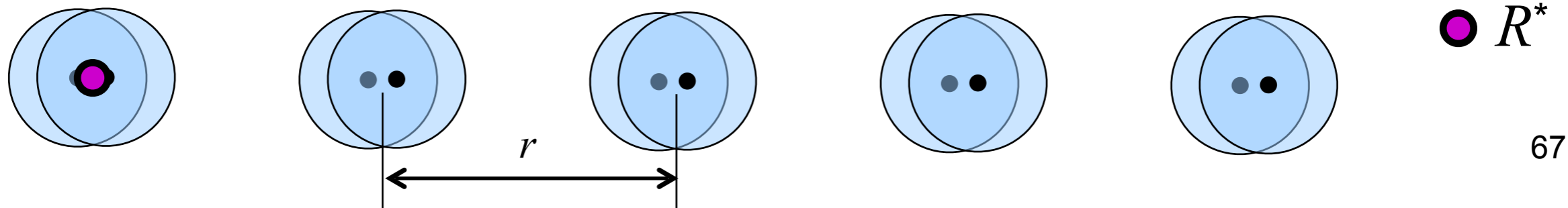




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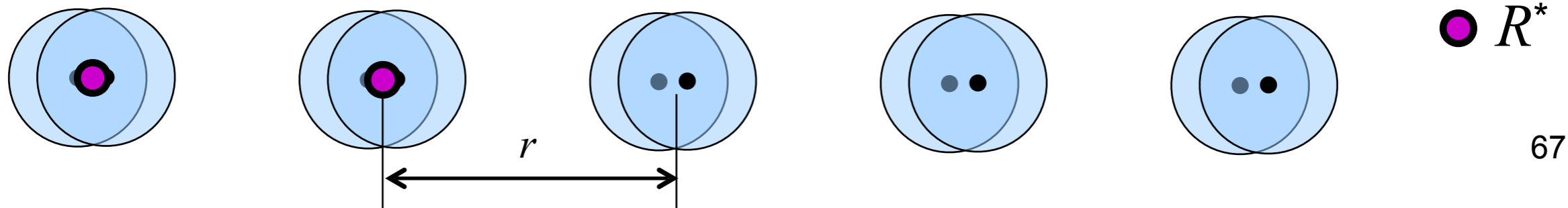
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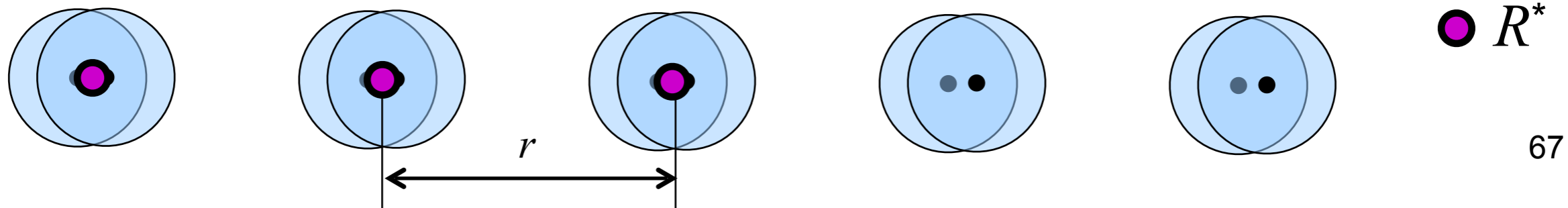
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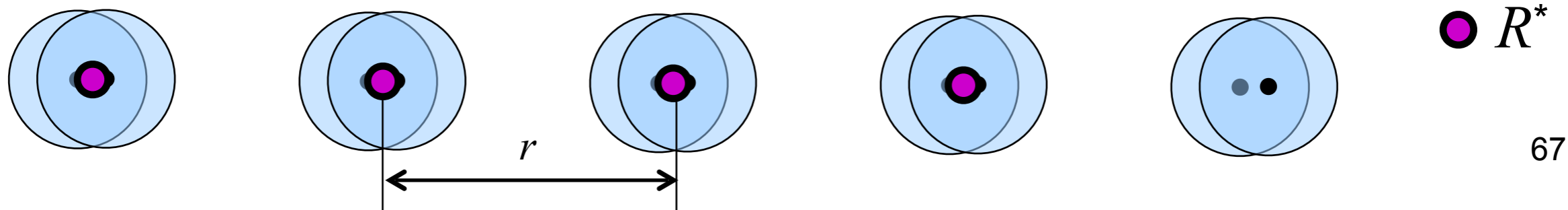
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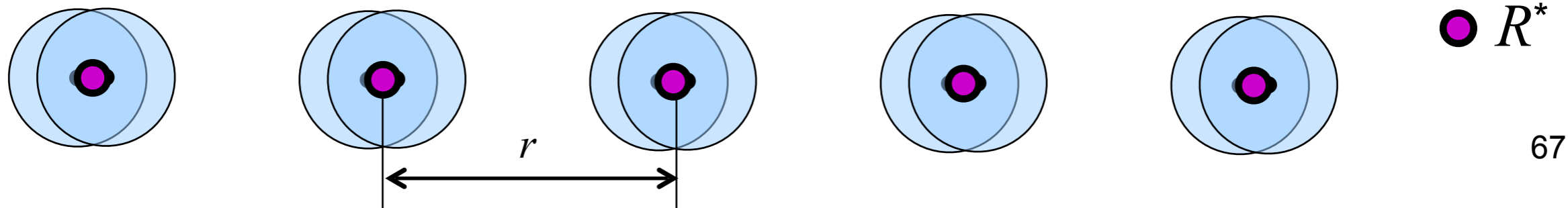
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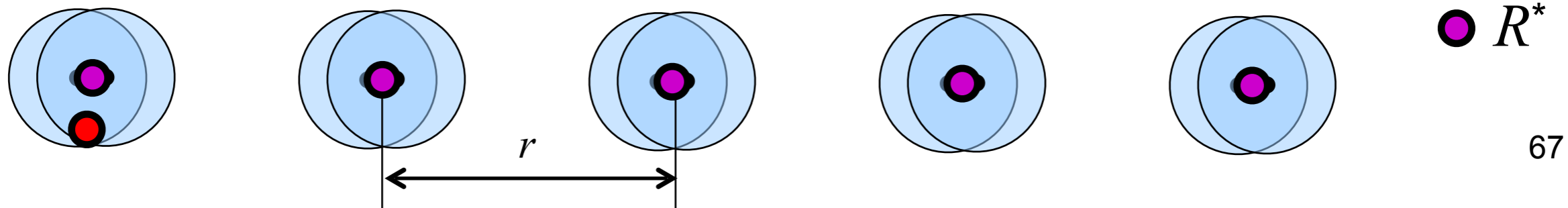
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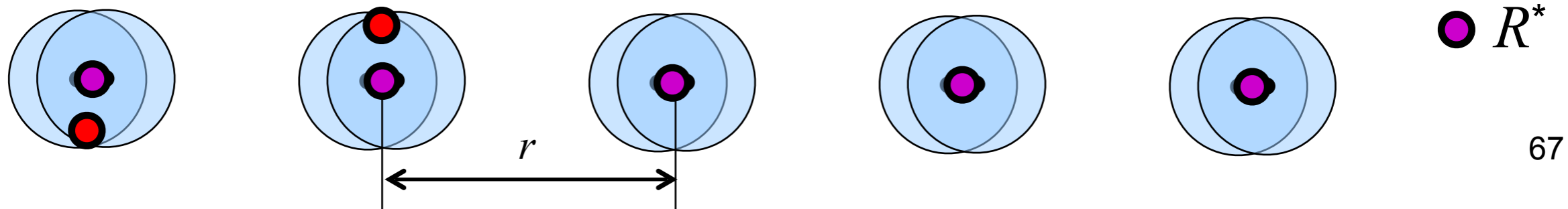
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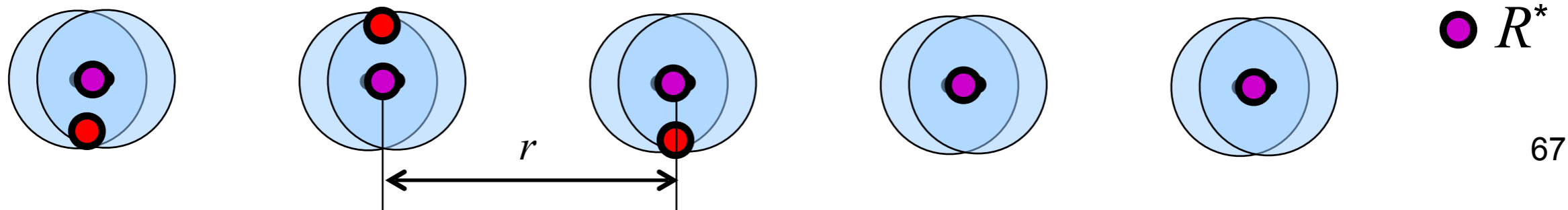
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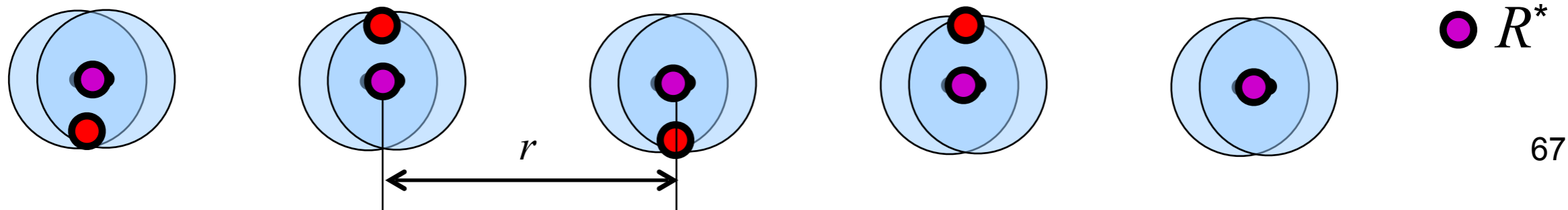




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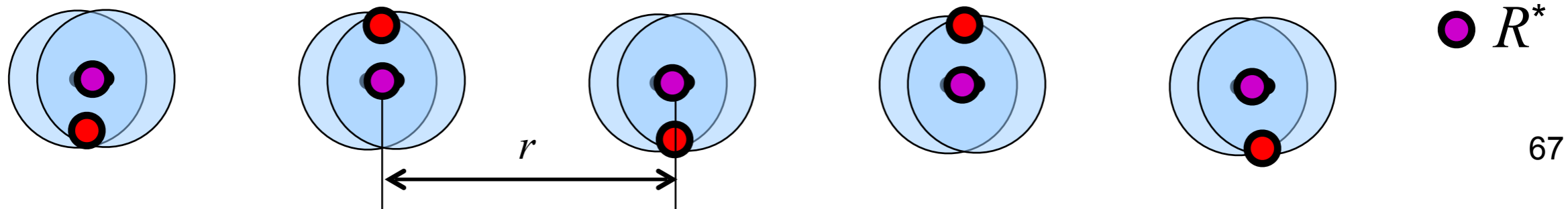
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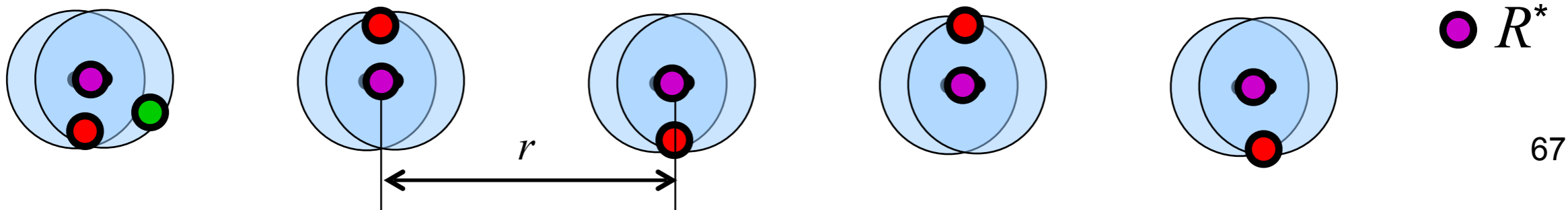
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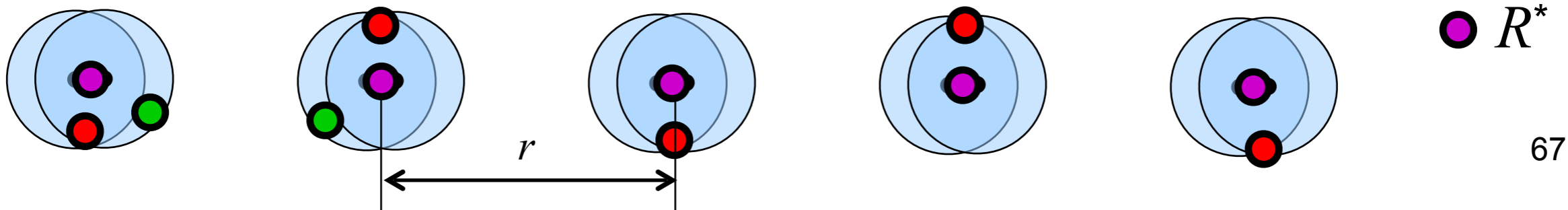
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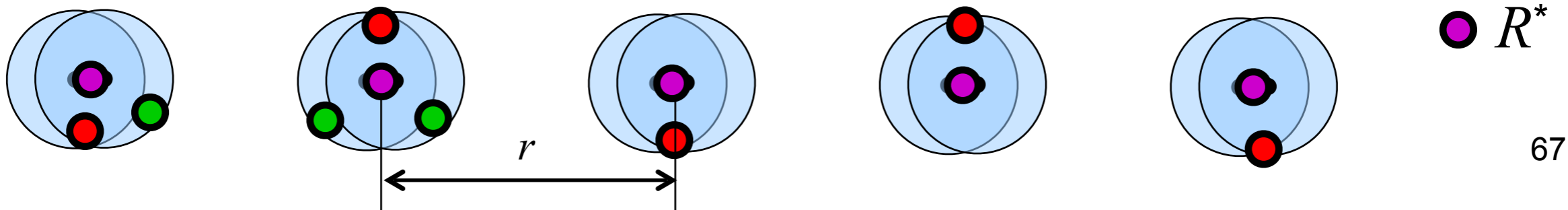
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# Additional Remarks

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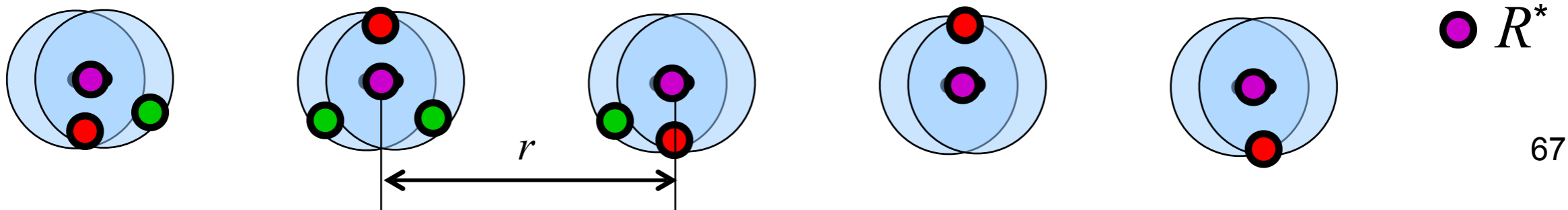
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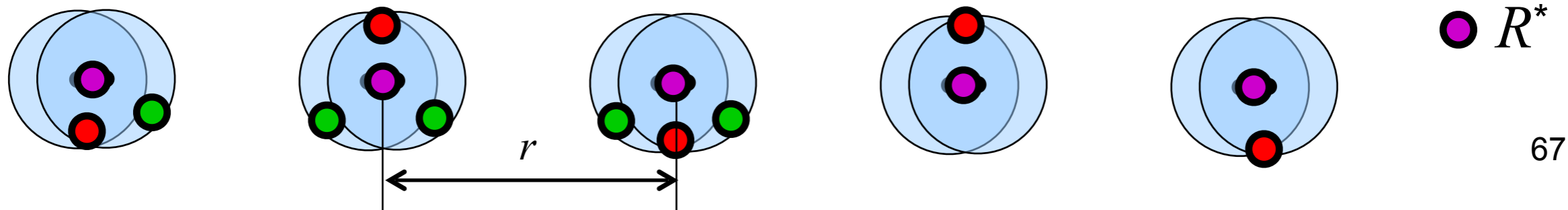
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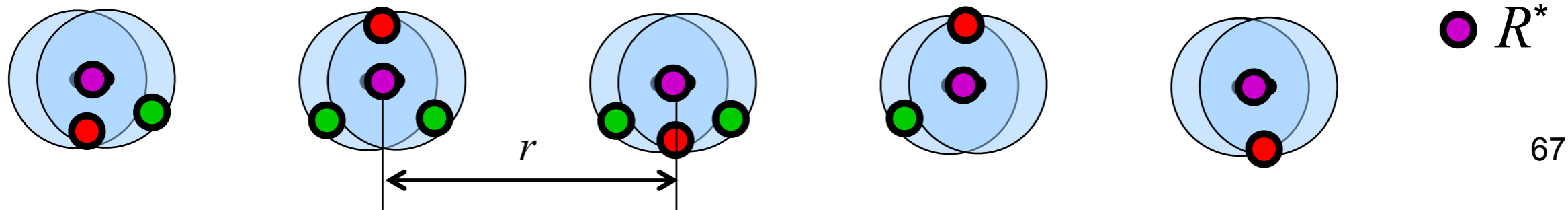
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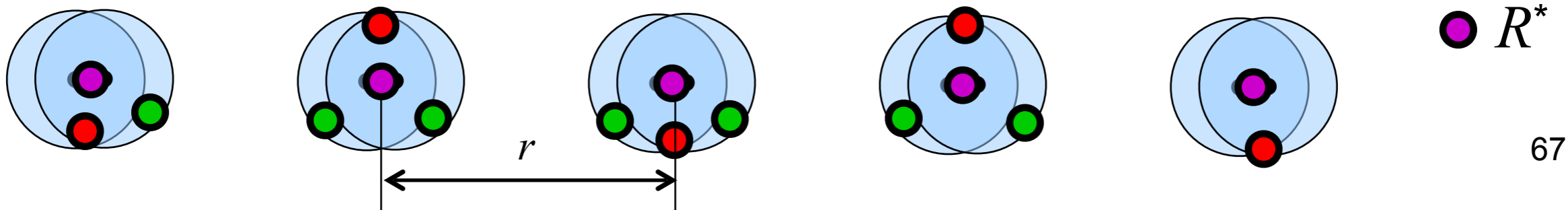




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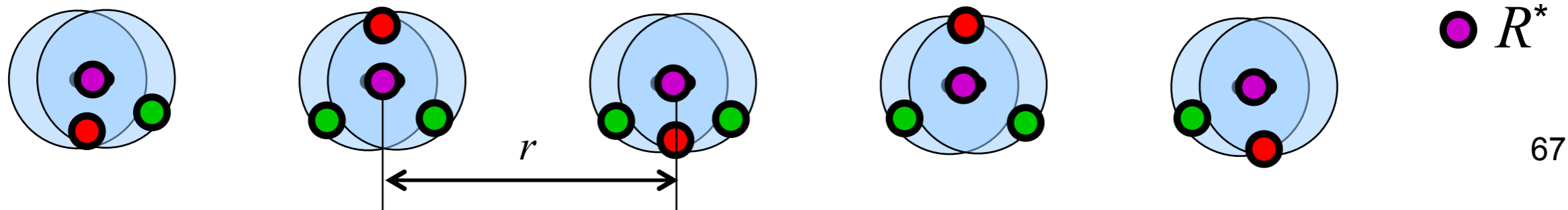
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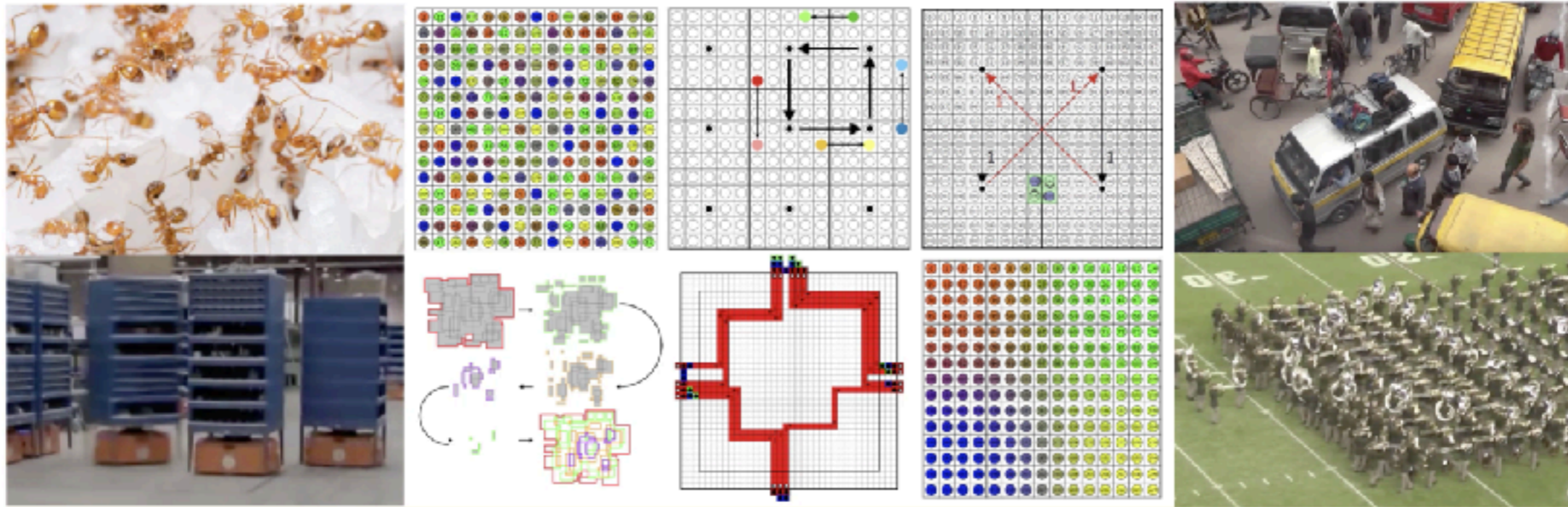
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- Fault tolerance:  $k$ -connectivity
  - [Bredin, Demaine, Hajiaghayi, Rus, MobiHoc'05, Zhang, Xue, Misra]

- 1. Introduction**
- 2. Review**
- 3. Extra Packing: Dispersion**
- 4. Extra Tours: Lawn Mowing**
- 5. Relay Placement**
- 6. Coordinated Motion Planning**

# Video!



## Coordinated Motion Planning: The Video

Aaron Becker, Sándor P. Fekete, Phillip Keldenich,  
Matthias Konitzny, Lillian Lin, Christian Scheffer





**Thank you!**