

SECTION 11.2

11. APPROXIMATION ALGORITHMS

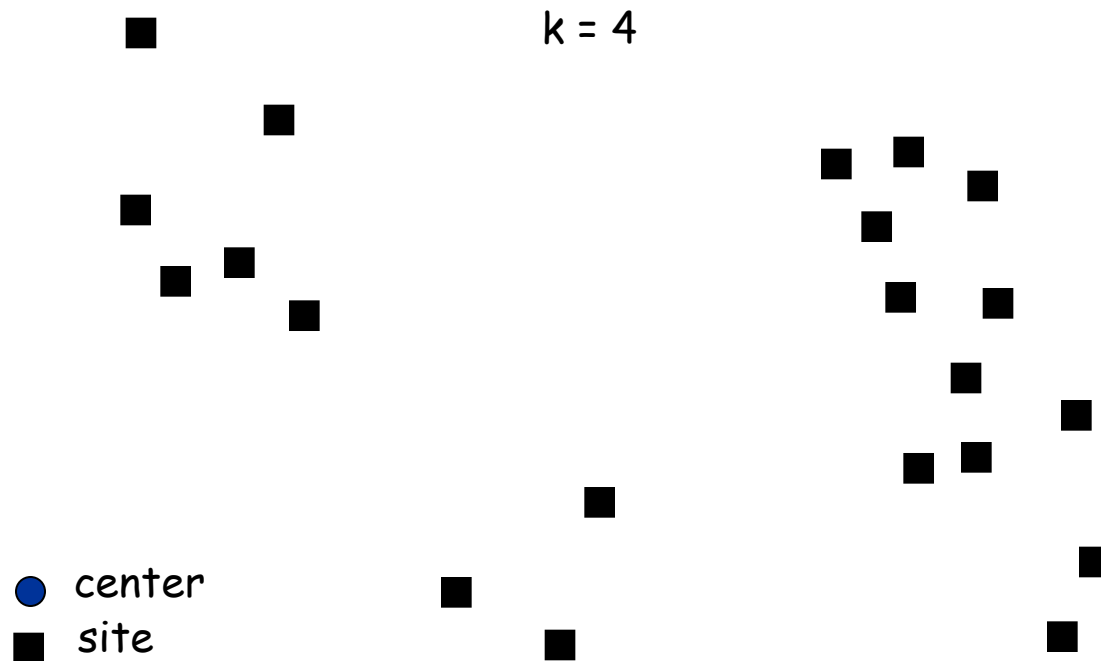
- *load balancing*
- *center selection*

k-Center Problem

k-Center Problem

Input. Set of n sites s_1, \dots, s_n and integer $k > 0$.

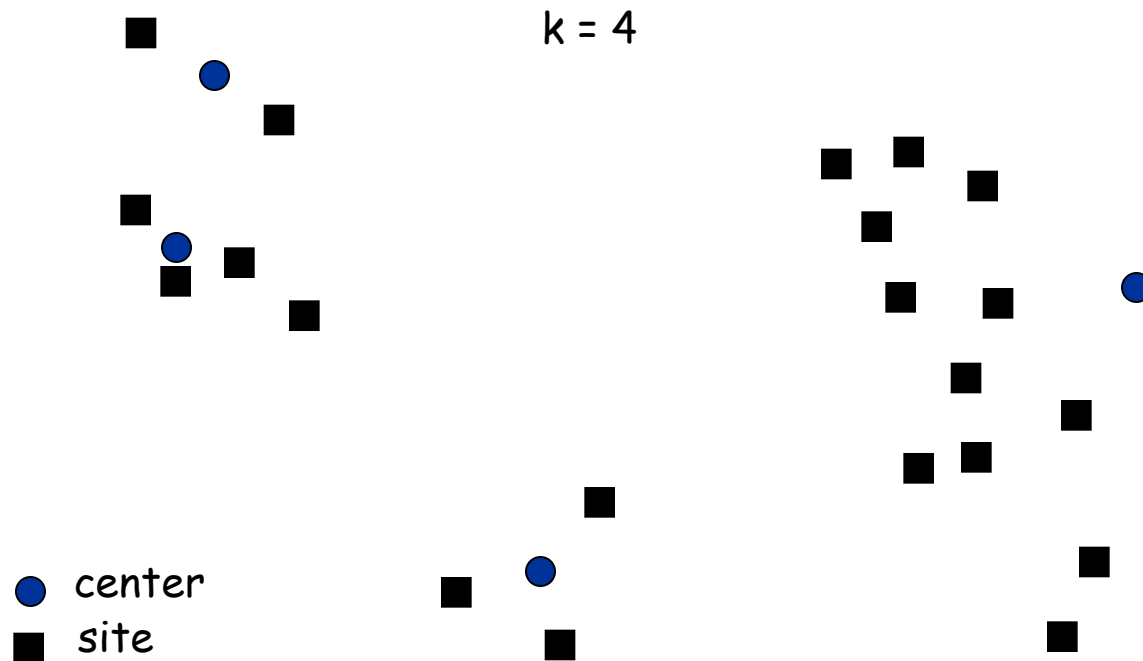
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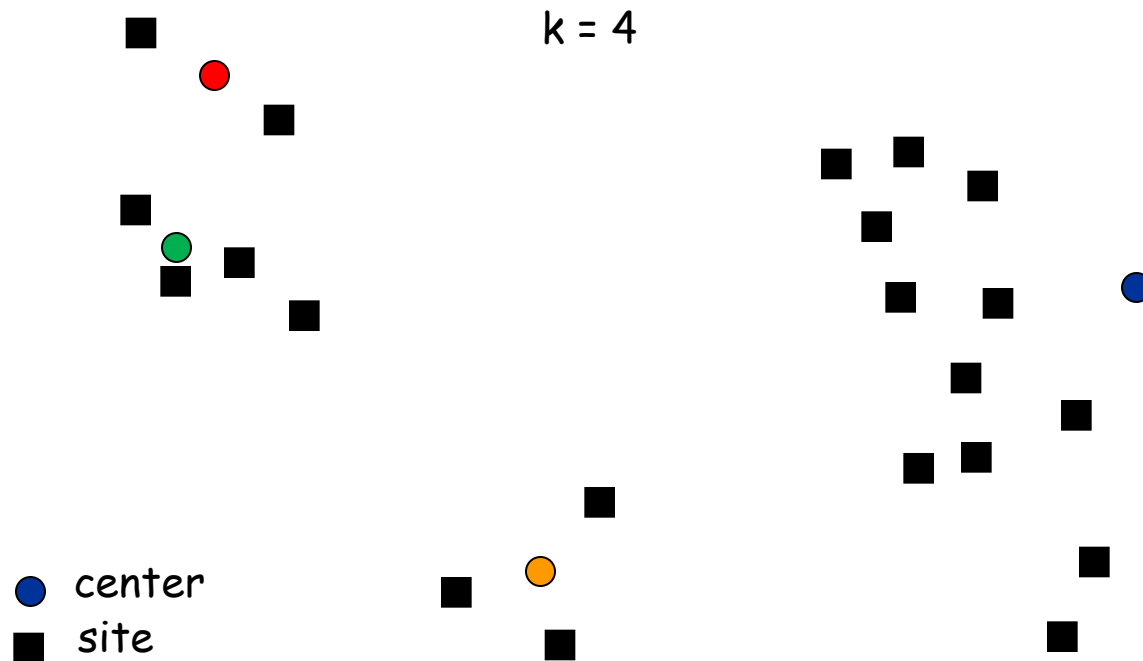
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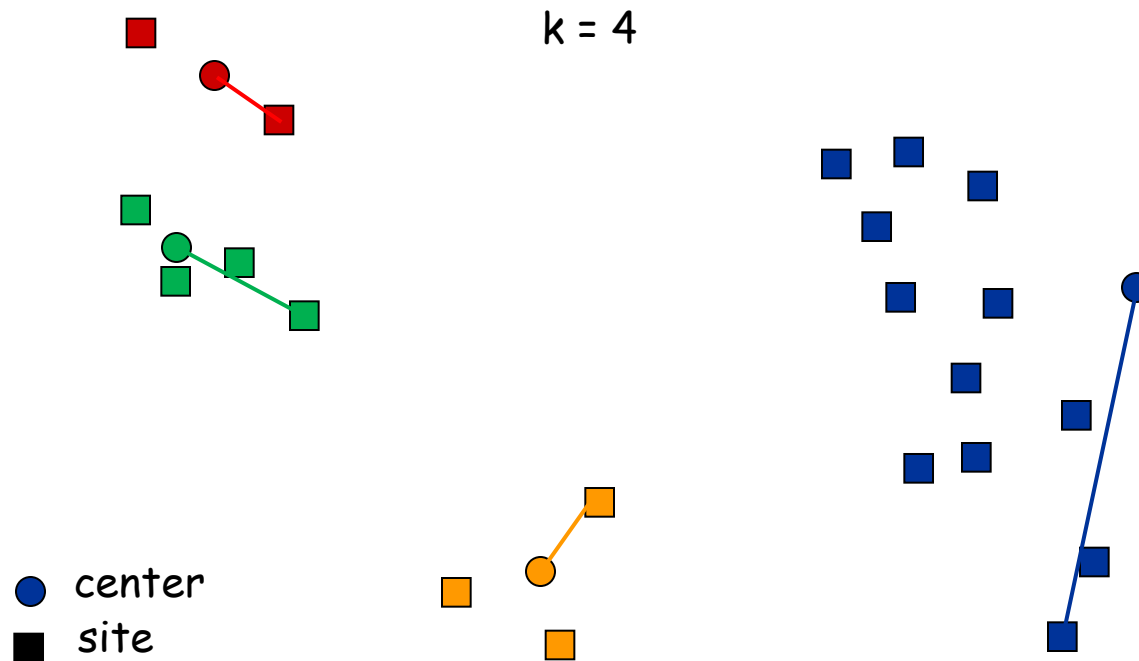
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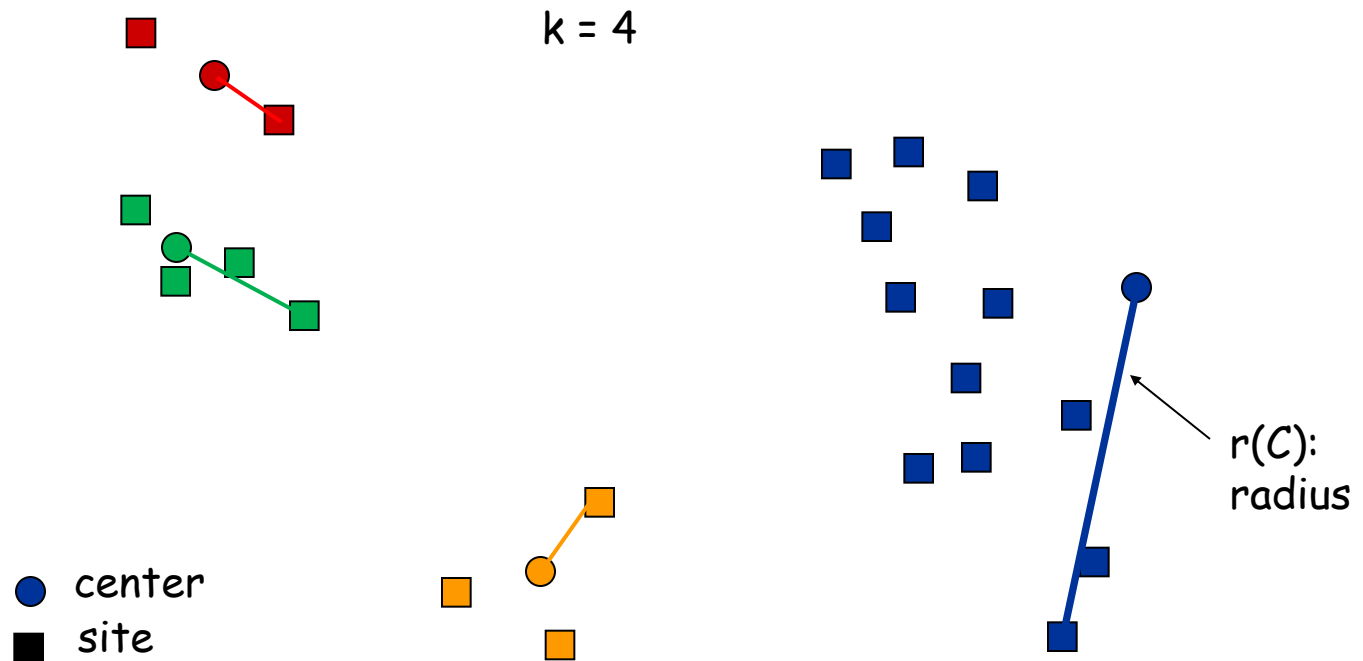
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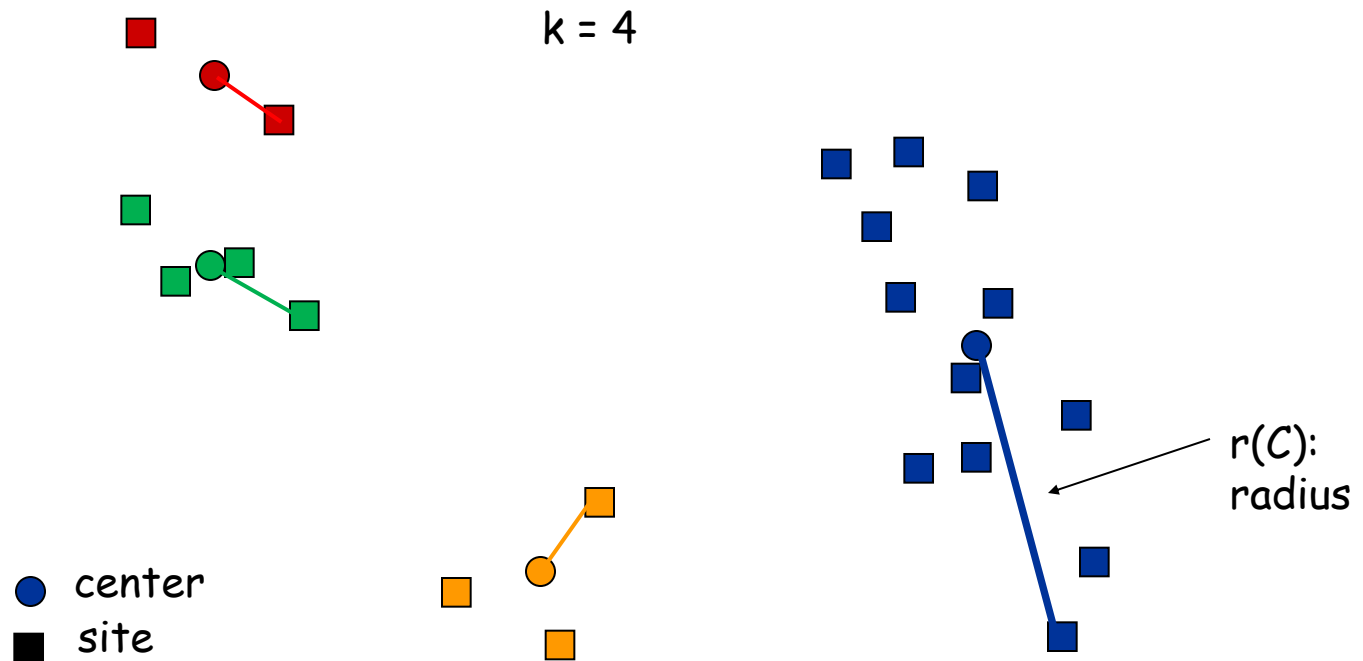
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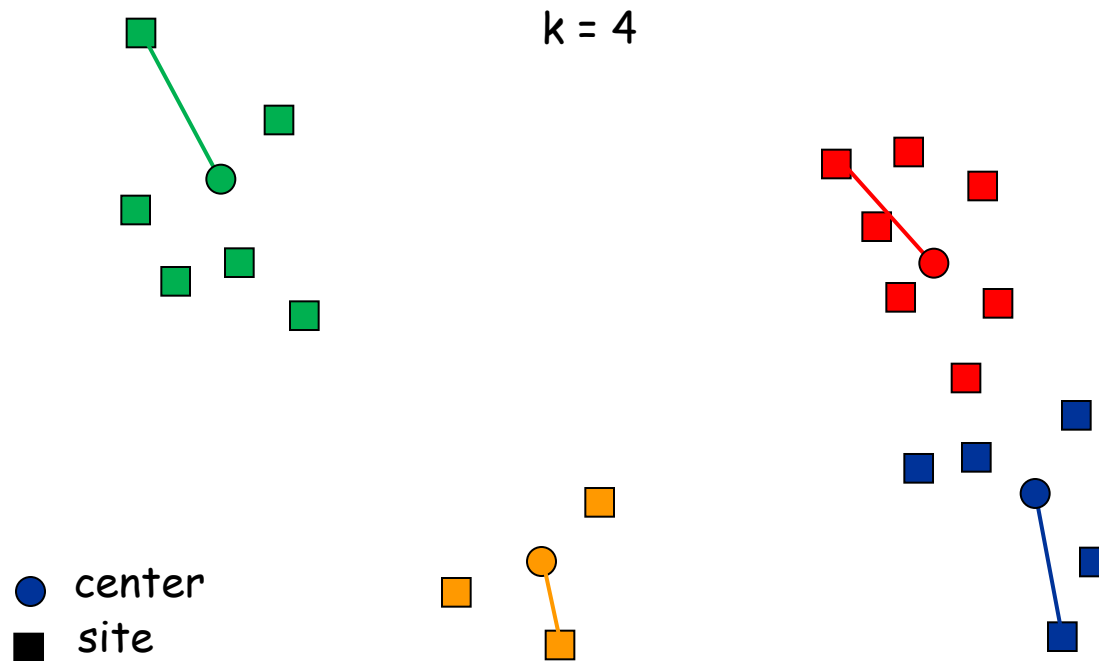
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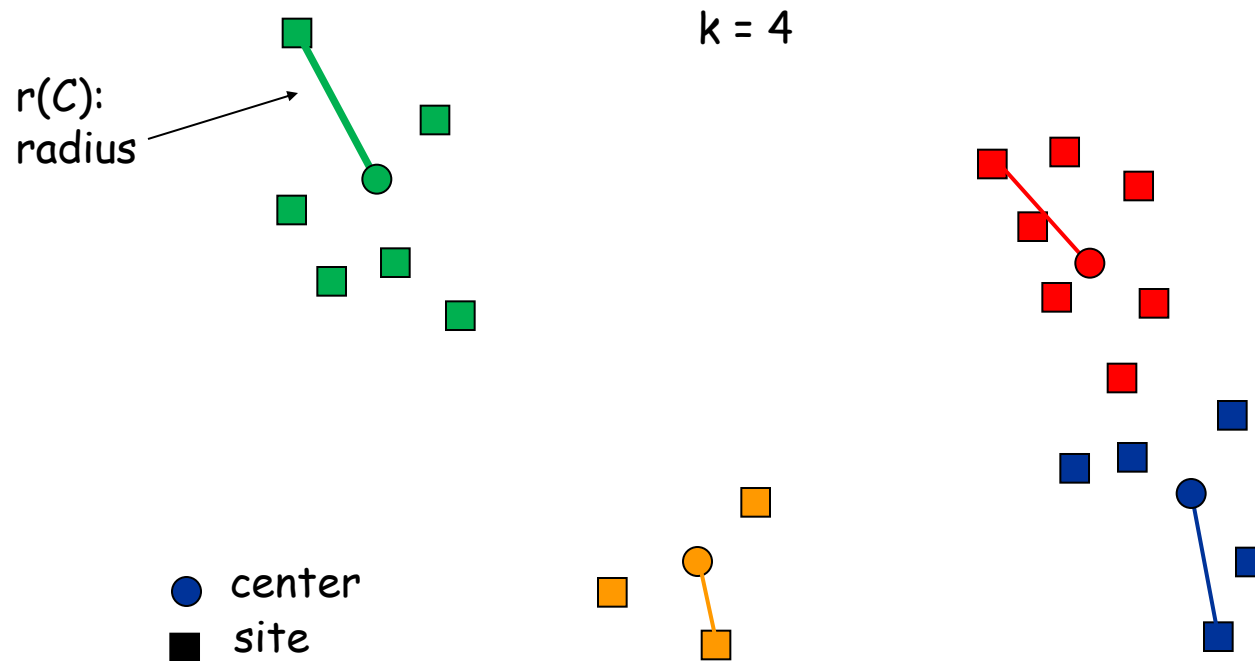
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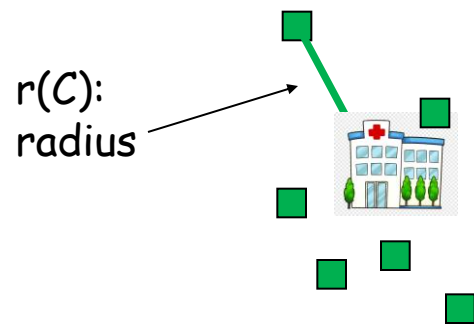
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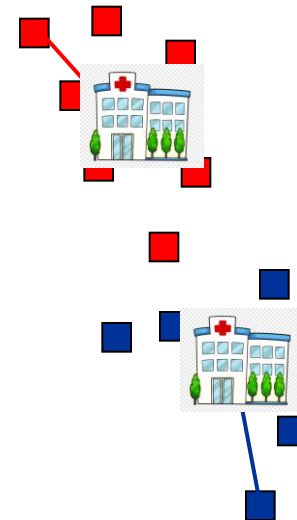
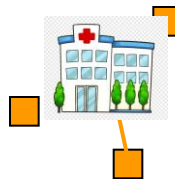
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$k = 4$

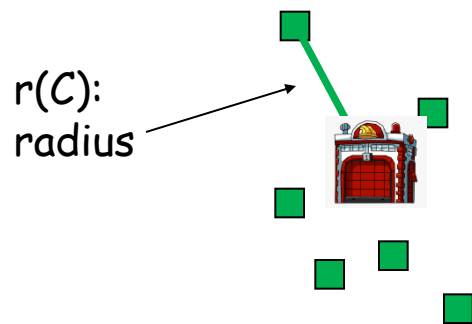
● center
■ site



k-Center Problem

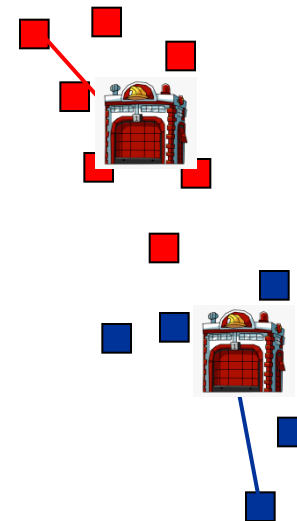
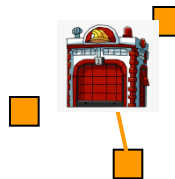
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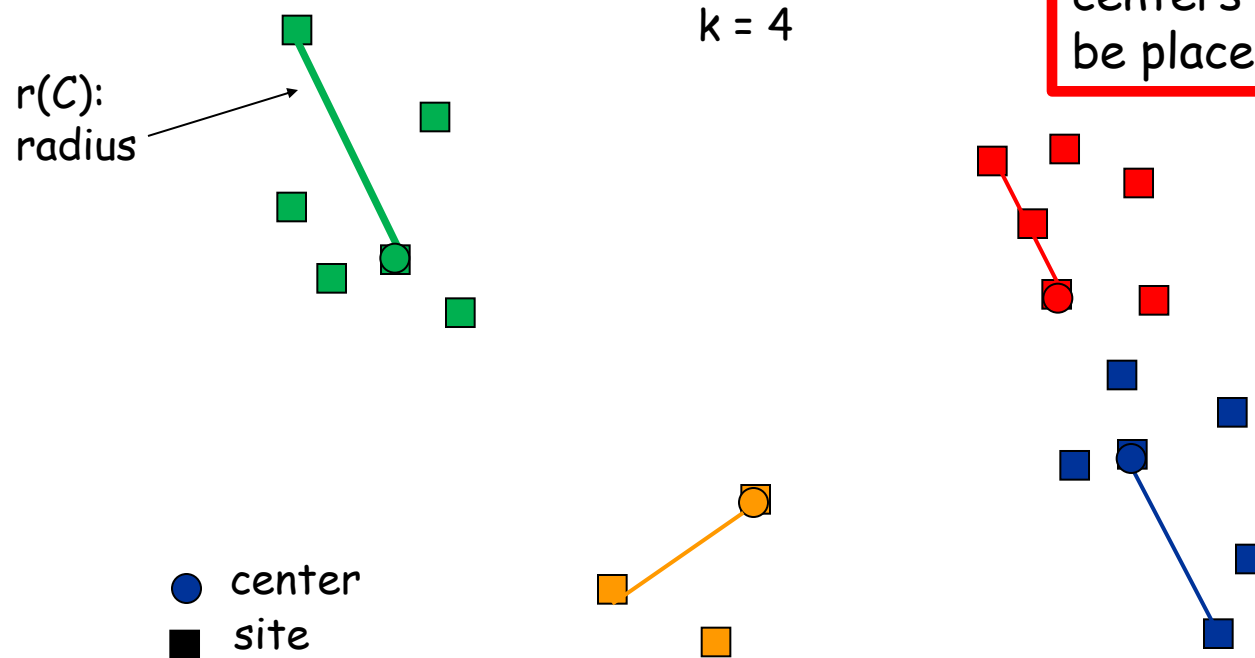
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Notation.

- $\text{dist}(x, y)$ = distance between x and y .
- $\text{dist}(s_i, C) = \min_{c \in C} \text{dist}(s_i, c)$ = distance from s_i to closest center.
- $r(C) = \max_i \text{dist}(s_i, C)$ = smallest covering radius.

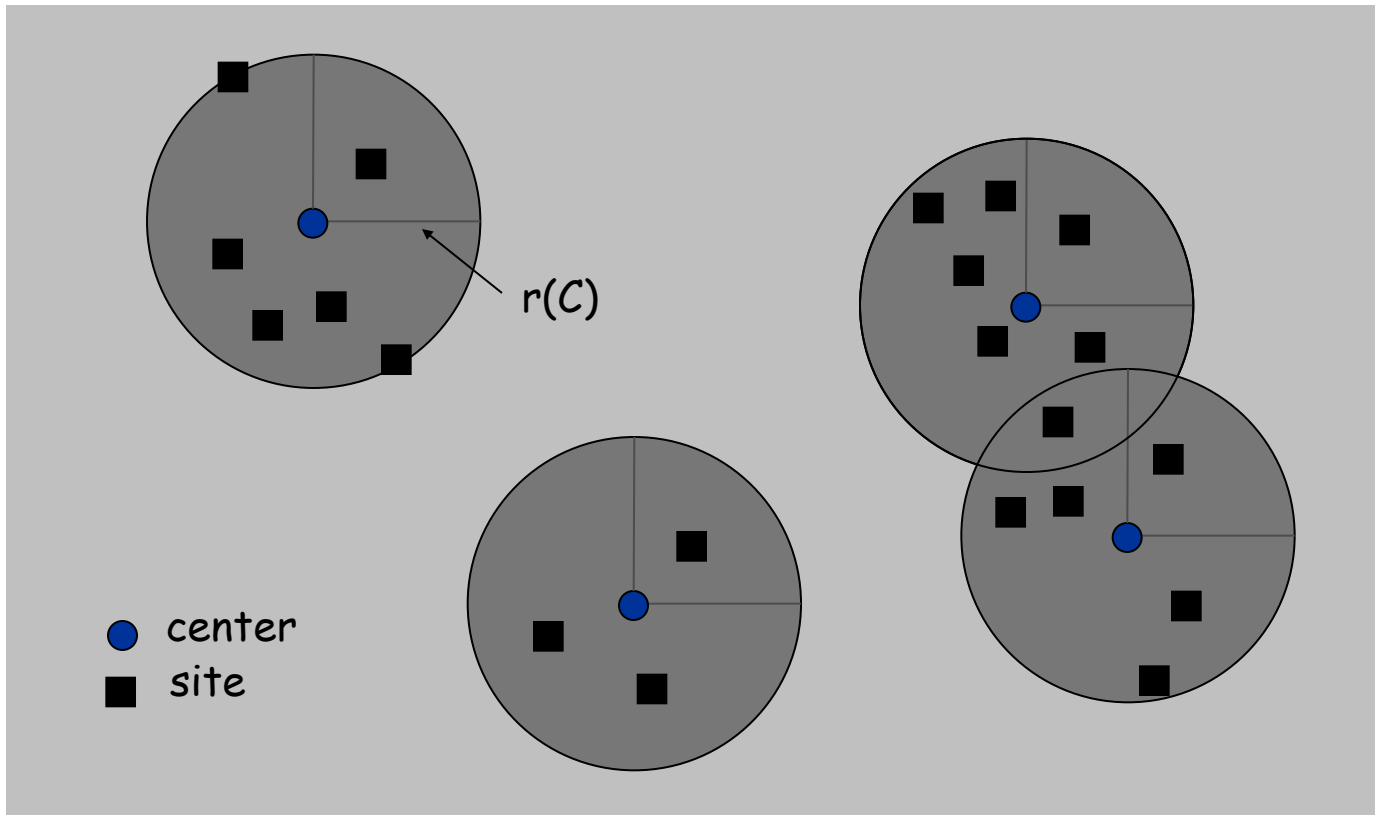
Goal. Find set of centers C that minimizes $r(C)$, subject to $|C| = k$.

Distance function properties.

- $\text{dist}(x, x) = 0$ (identity)
- $\text{dist}(x, y) = \text{dist}(y, x)$ (symmetry)
- $\text{dist}(x, y) \leq \text{dist}(x, z) + \text{dist}(z, y)$ (triangle inequality)

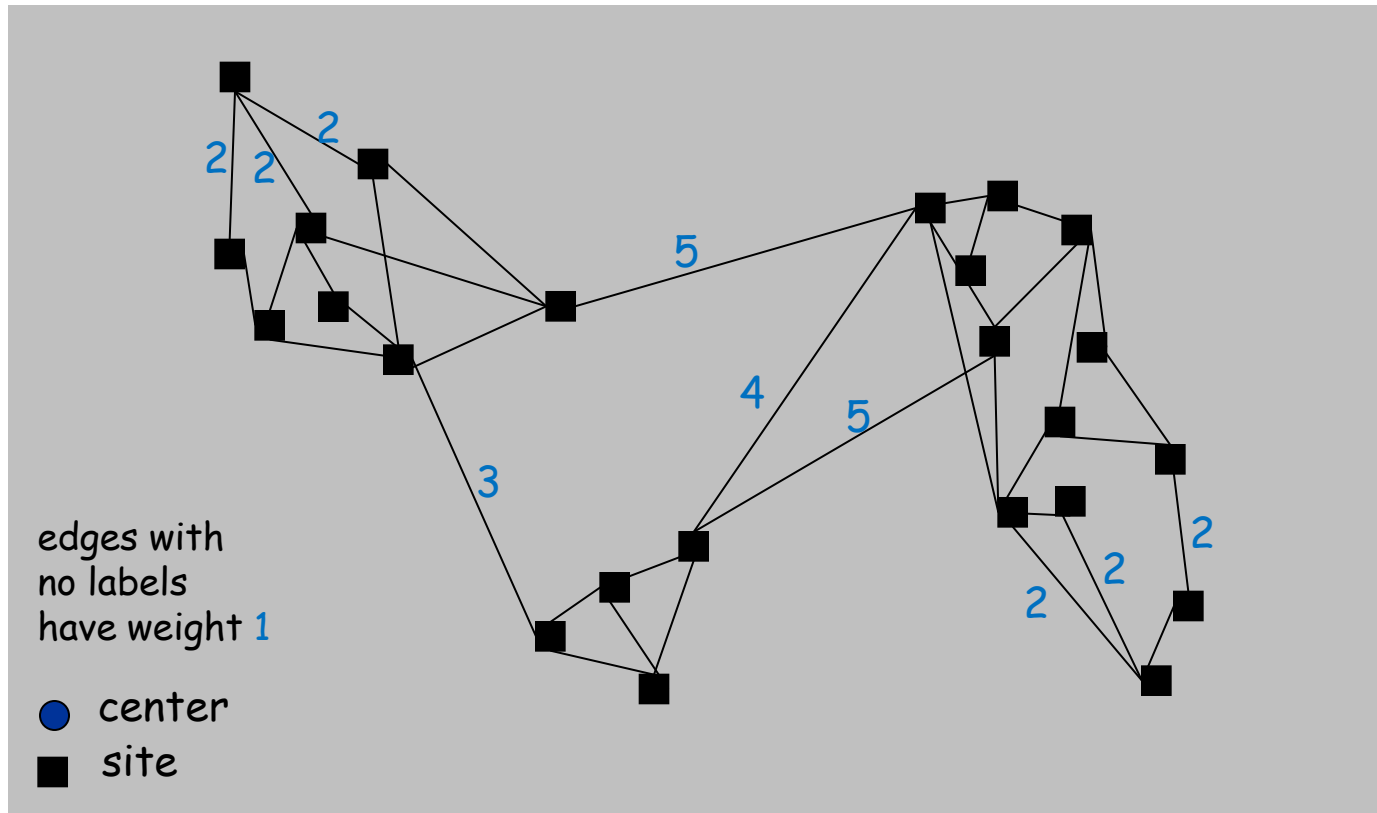
k-Center Problem

Ex: each site is a point in the plane, a center can be any point in the plane, $\text{dist}(x, y) = \text{Euclidean distance}$.



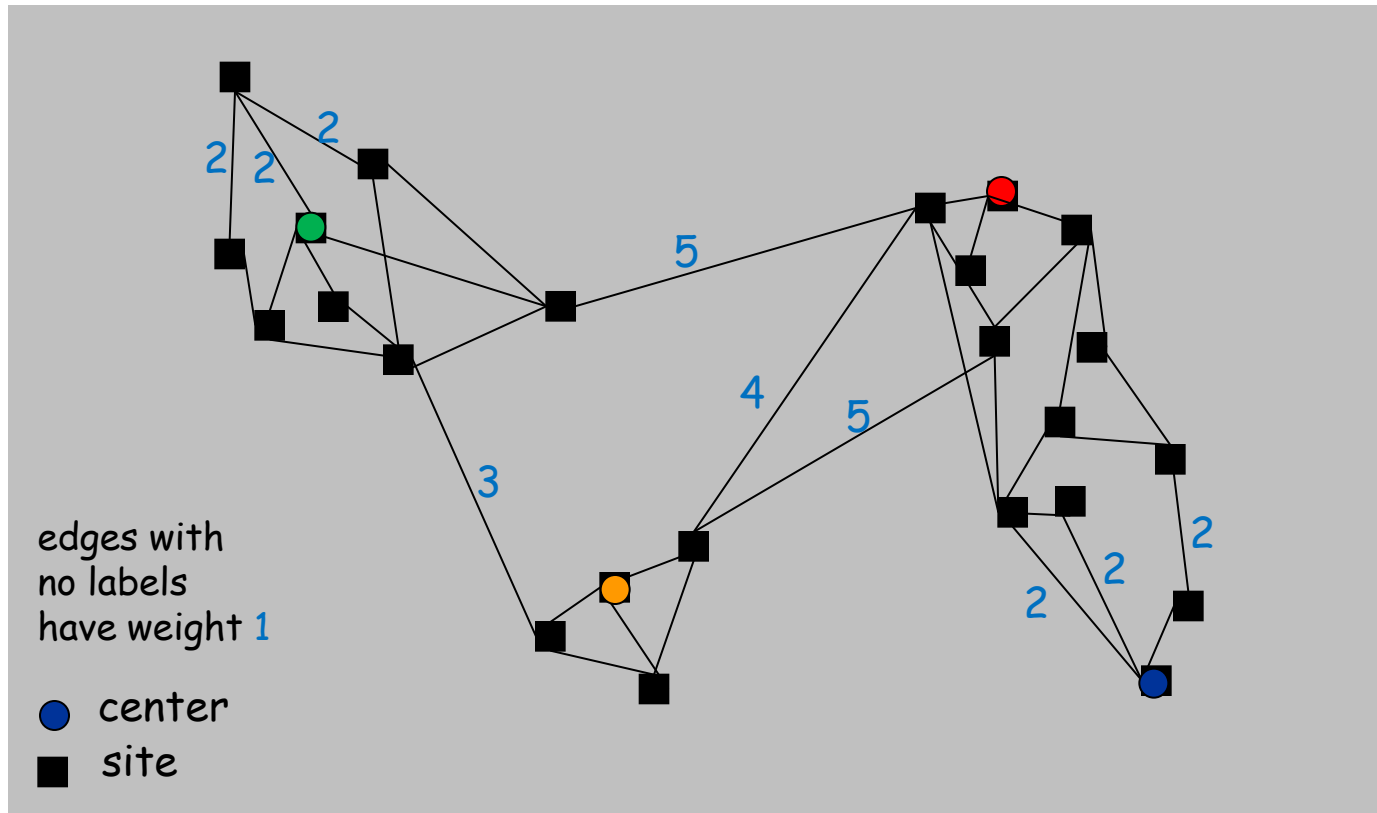
k-Center Problem

Ex: each site is a vertex in undirected weighted graph, a center can be any vertex, $\text{dist}(x, y) = (\text{weighted})$ distance in G between x and y .



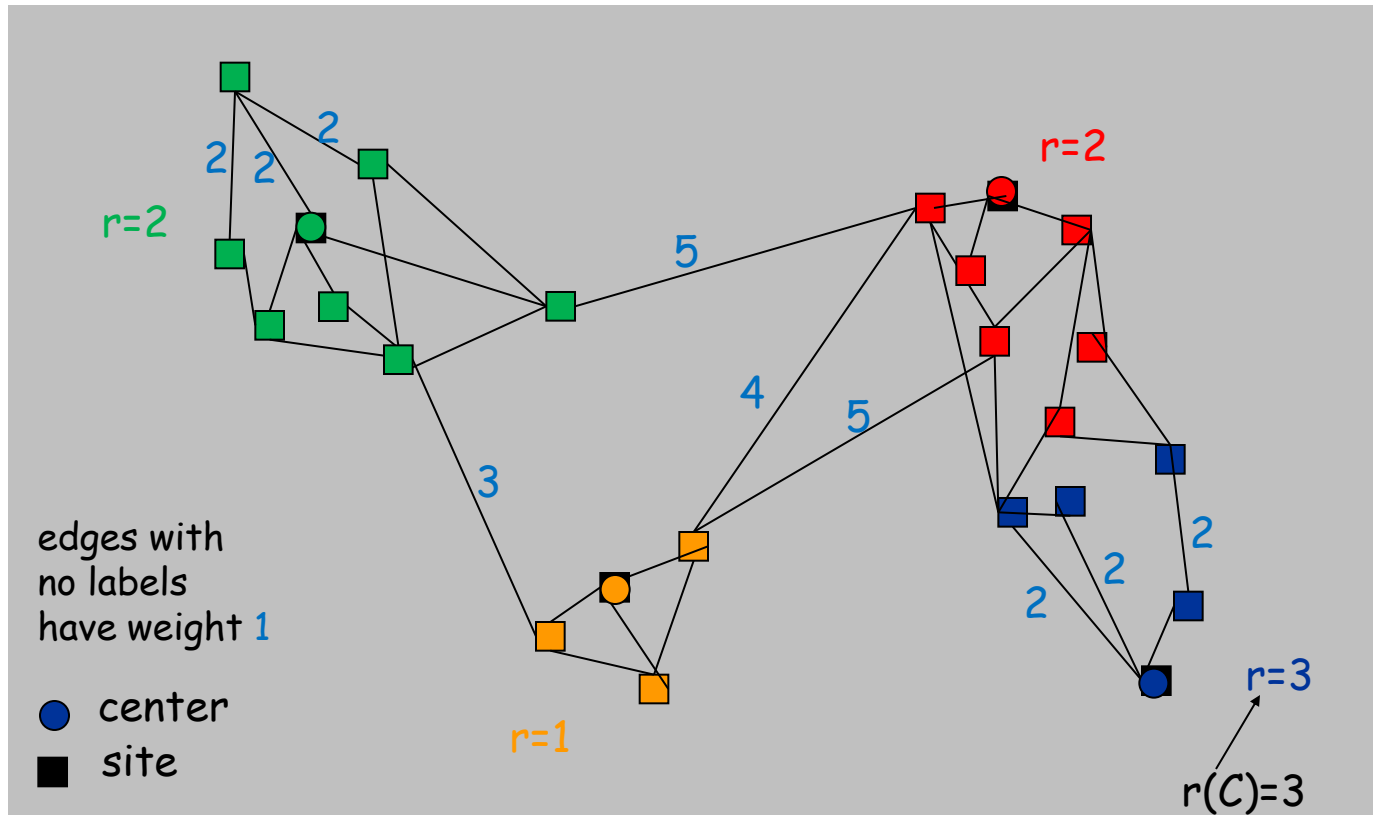
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Center Selection: Greedy Algorithm

Greedy algorithm. Repeatedly choose the next center to be the site **farthest** from any existing center.

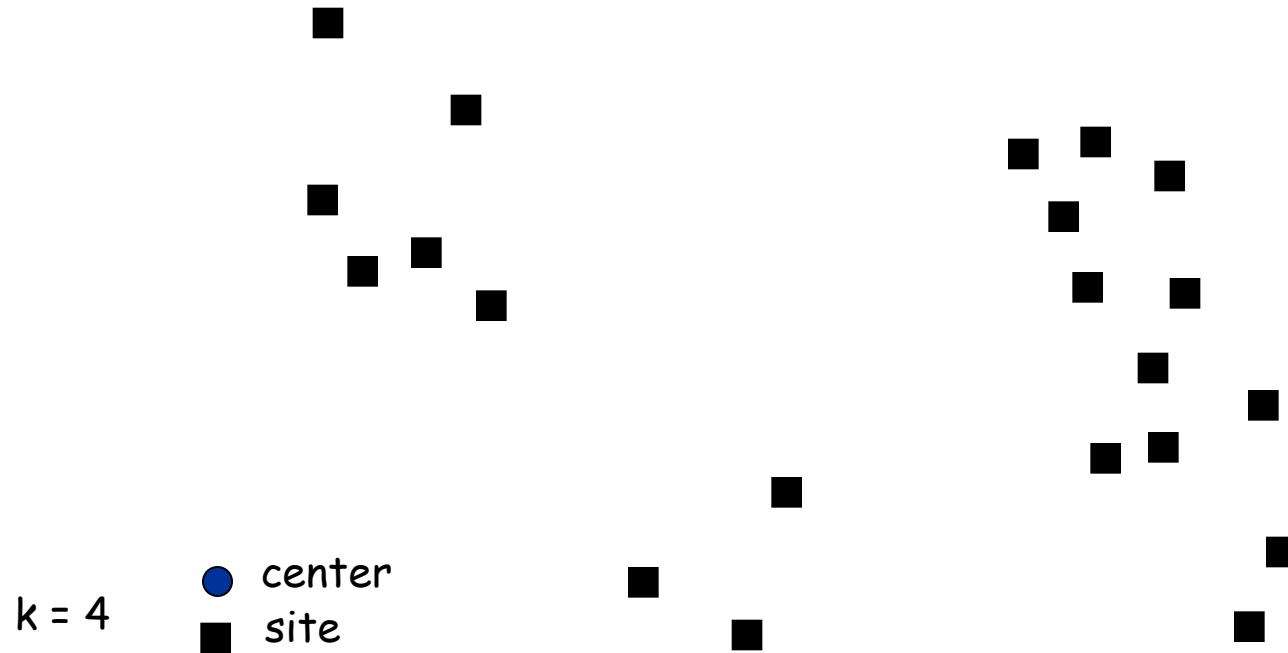
```
Greedy-Center-Selection(k, n, s1, s2, ..., sn) {  
  
    C =  $\phi$   
    repeat k times {  
        Select a site si with maximum dist(si, C)  
        Add si to C  
    }  
    return C  
}
```

↑
site farthest from any center

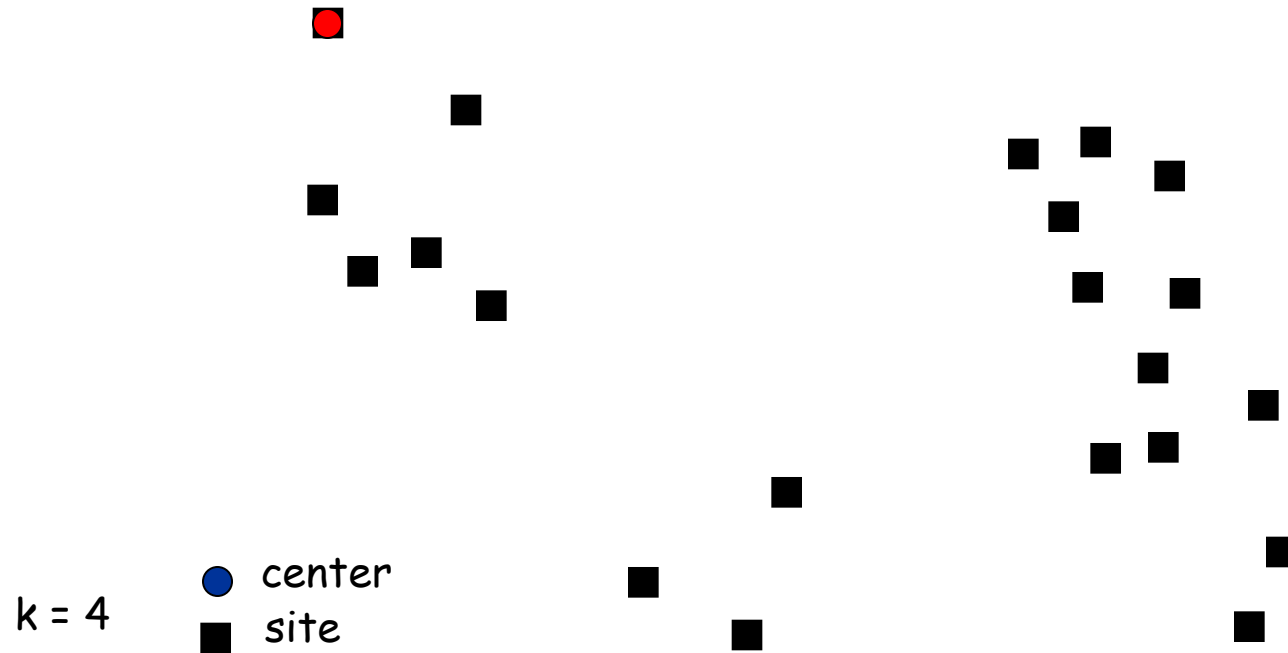
Observation. Upon termination all centers in C are pairwise at least $r(C)$ apart.

Pf. By construction of algorithm.

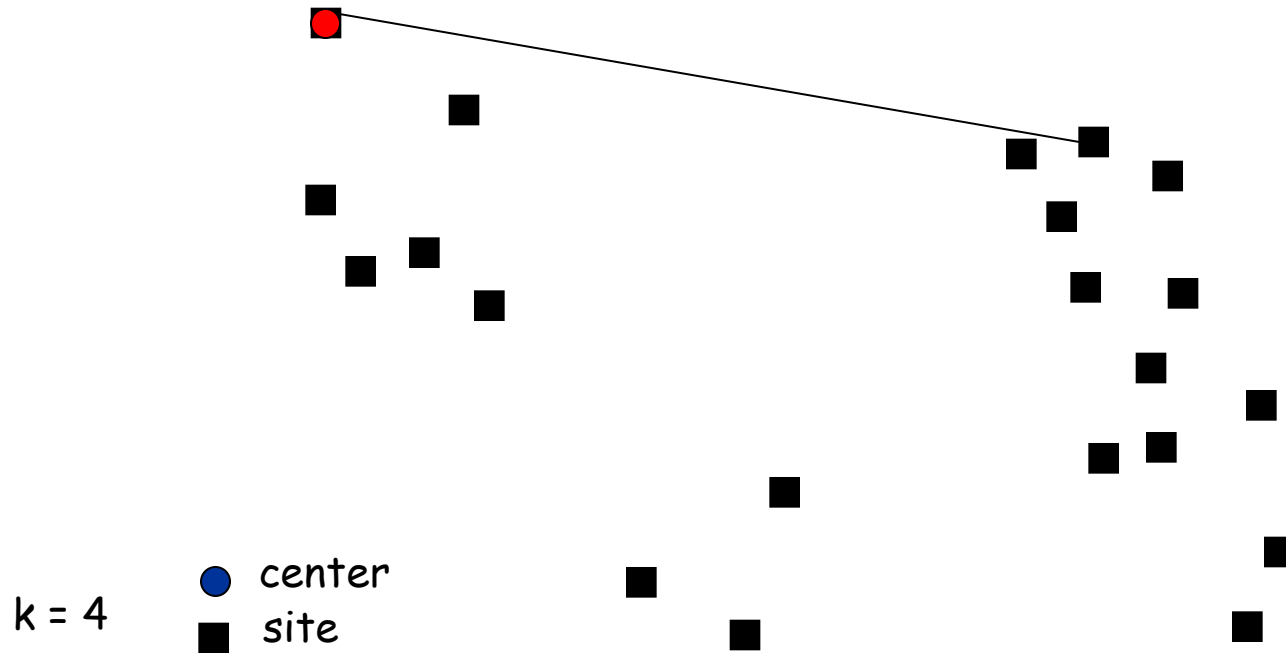
example of execution



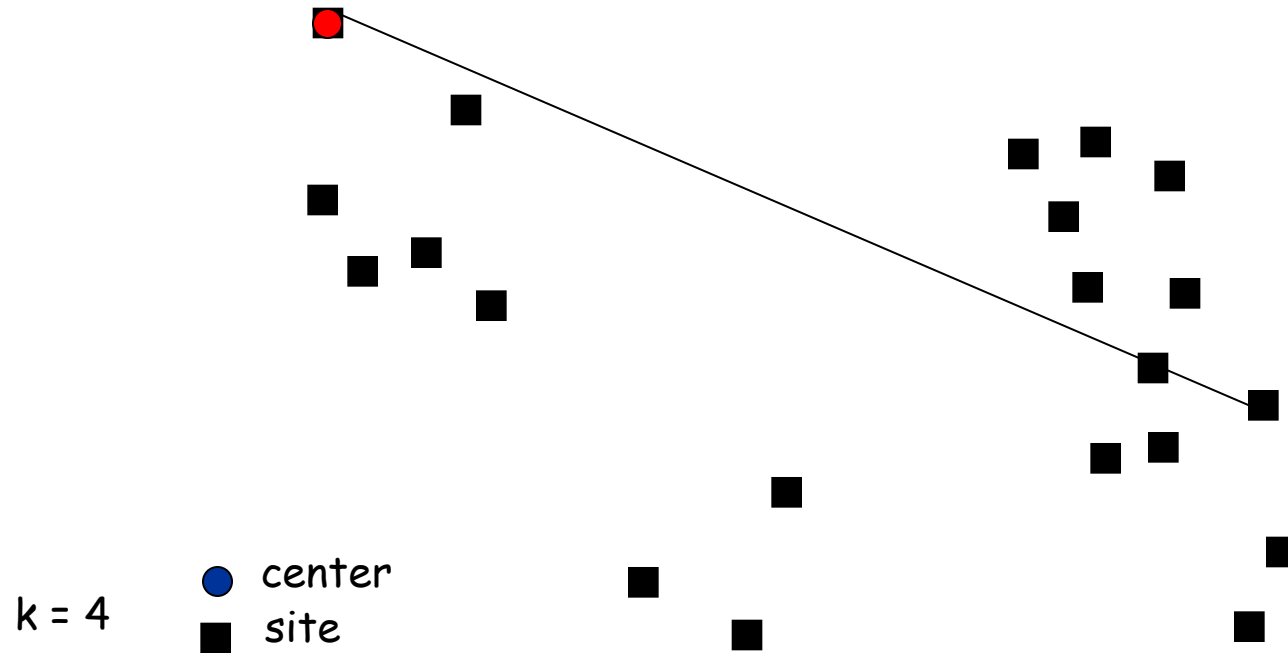
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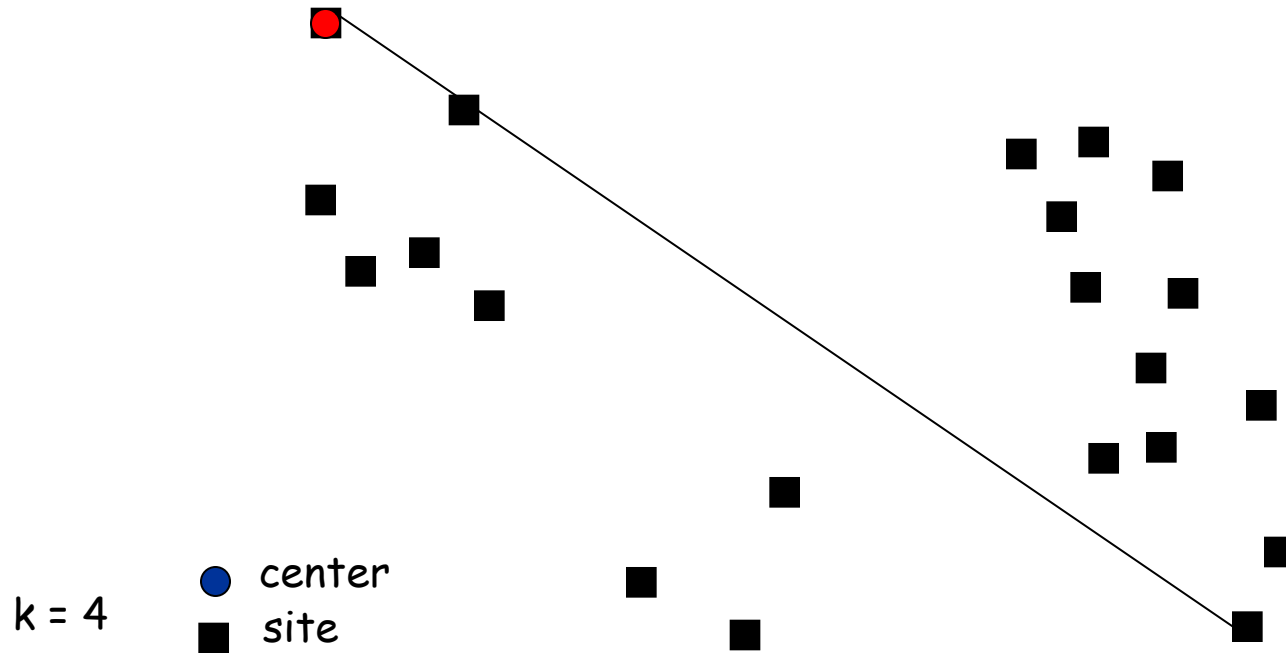
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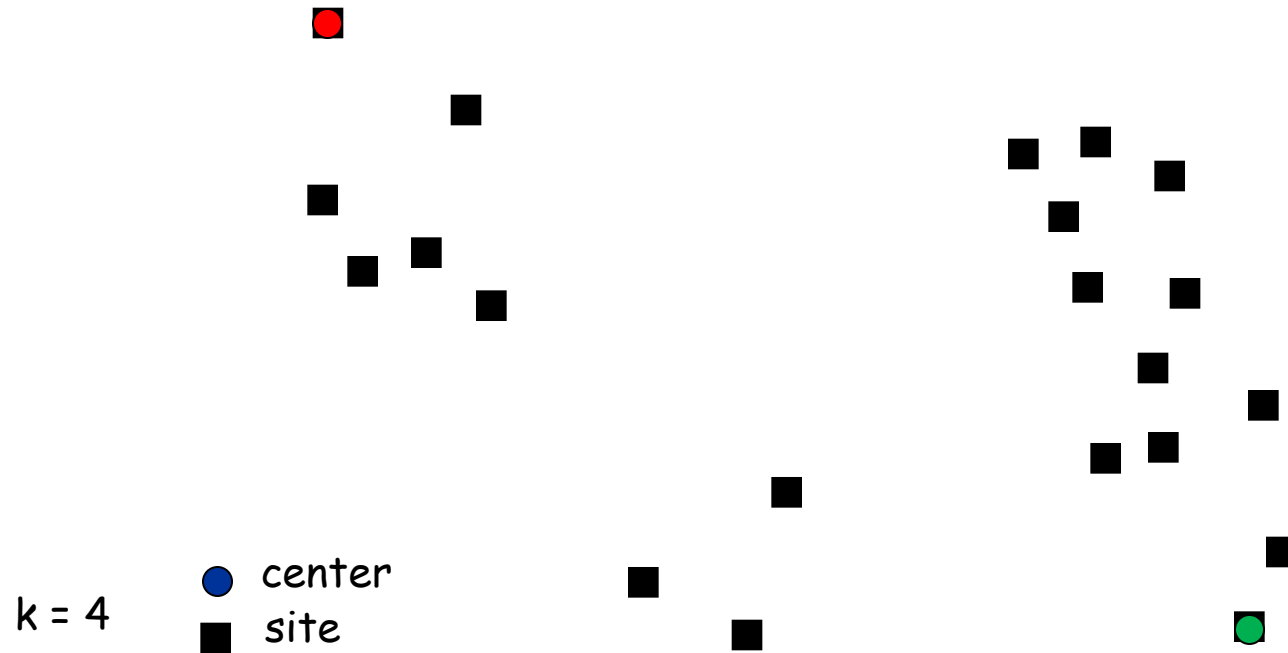
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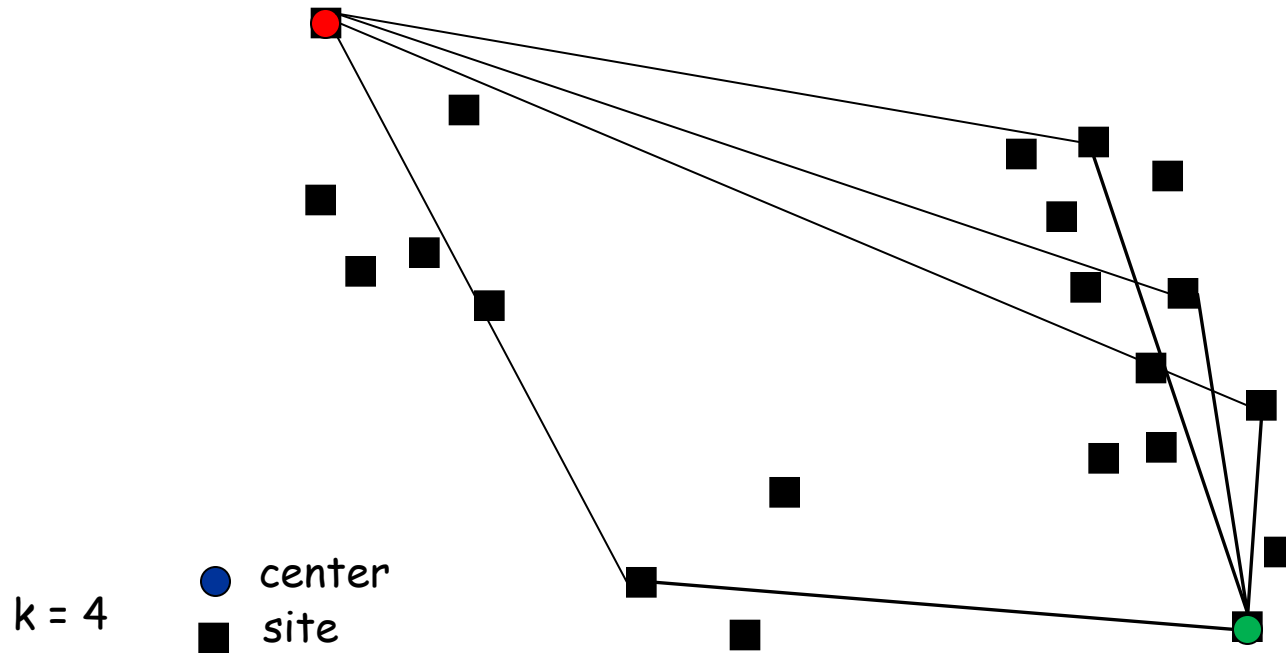
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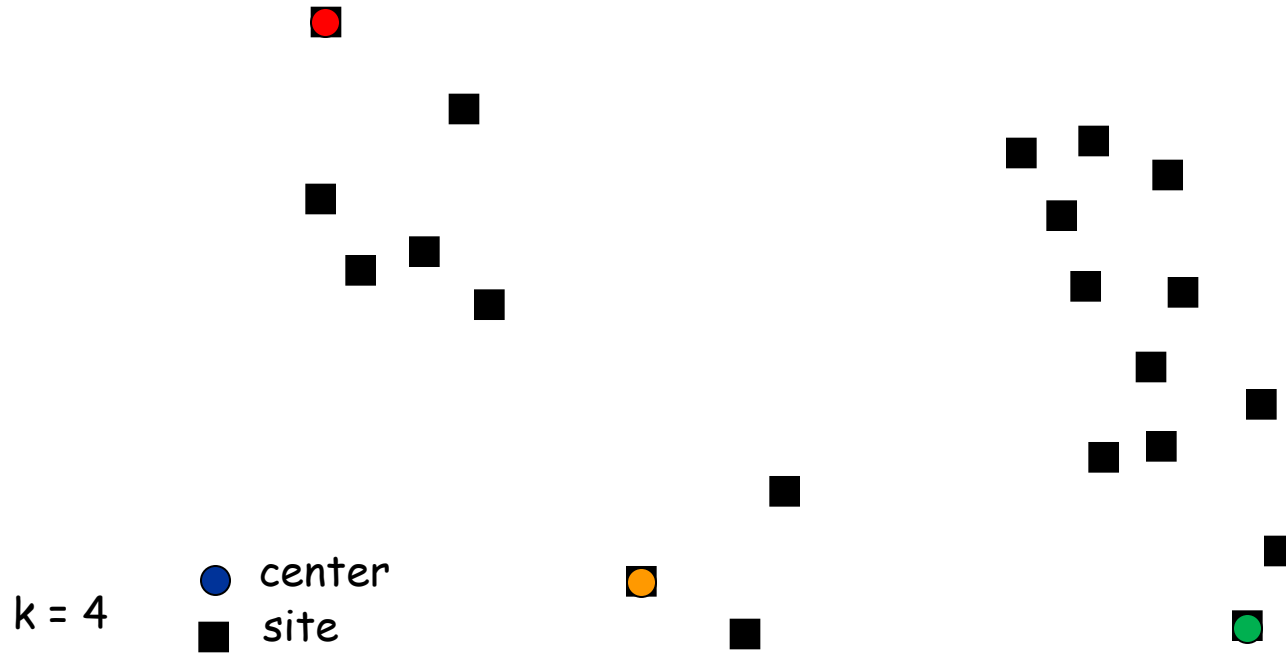
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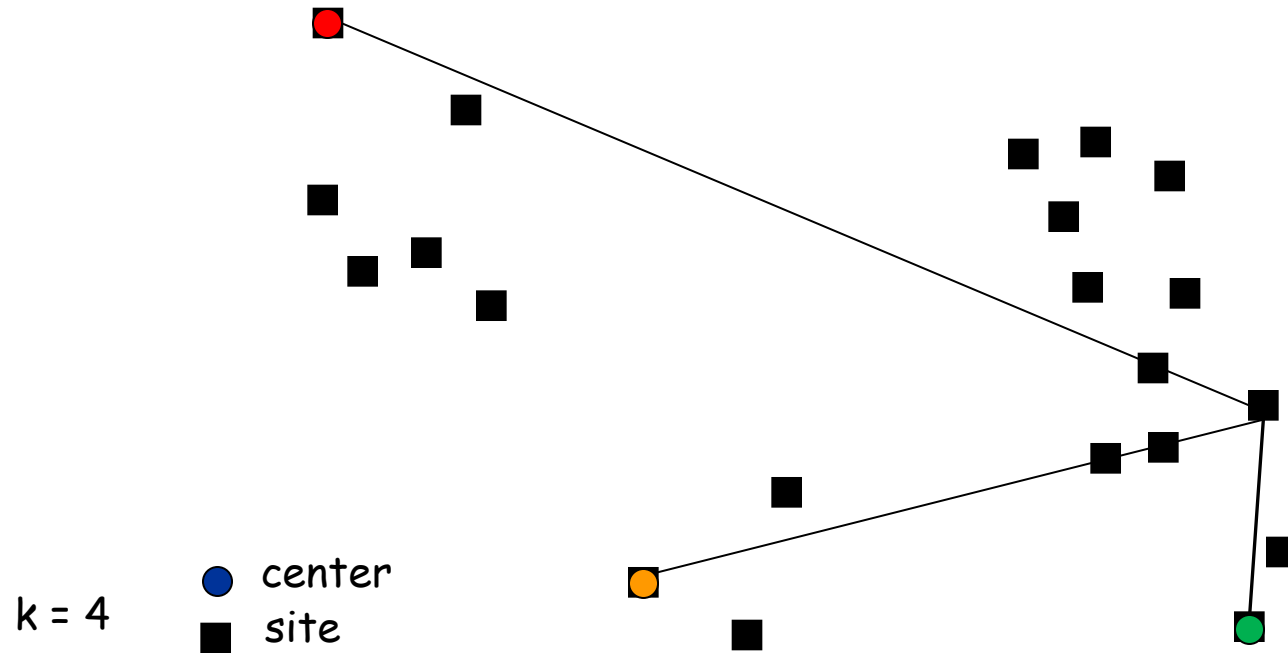
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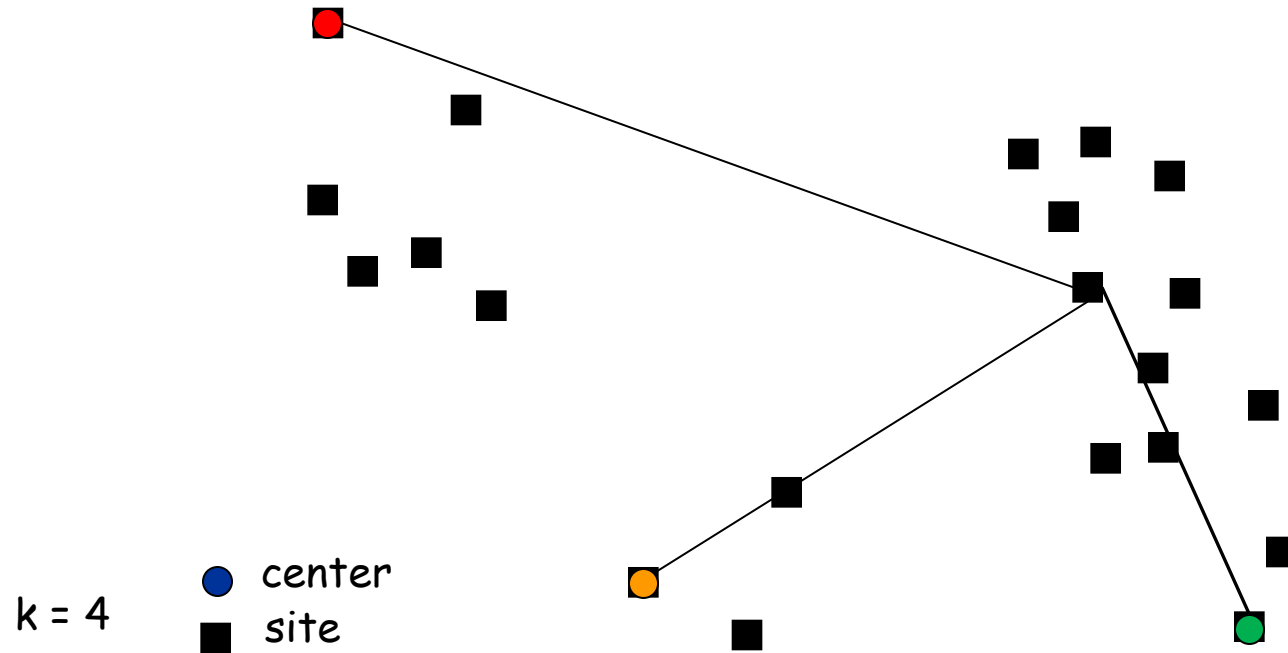
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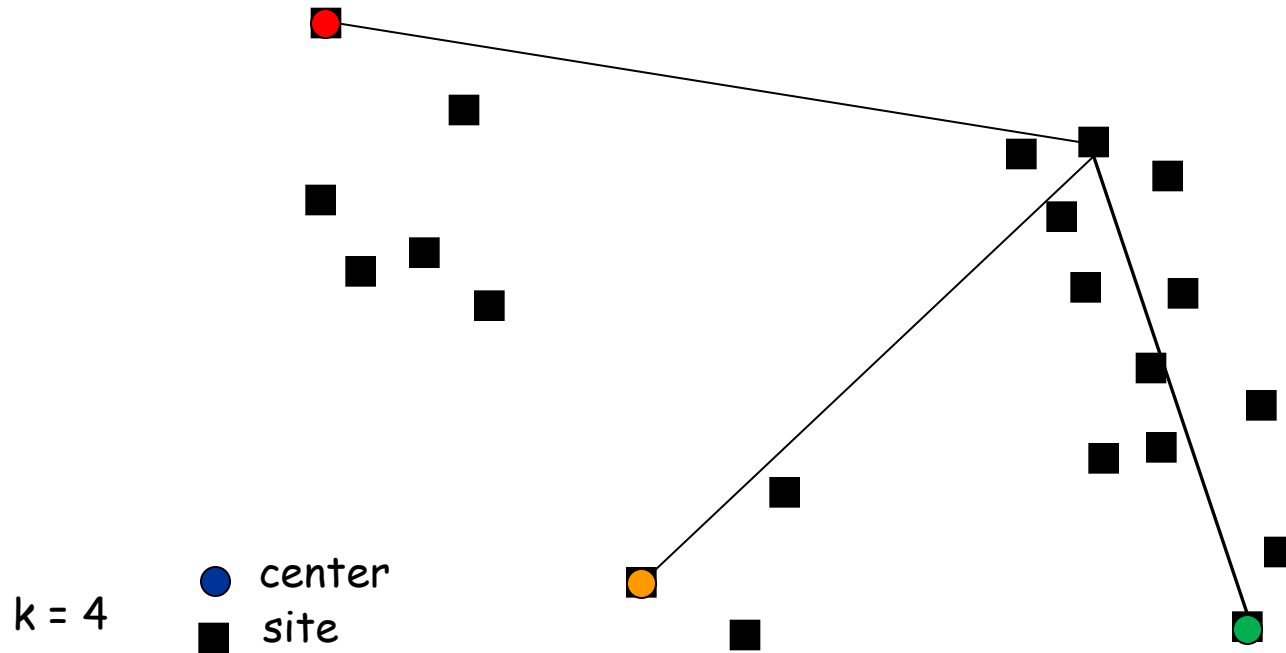
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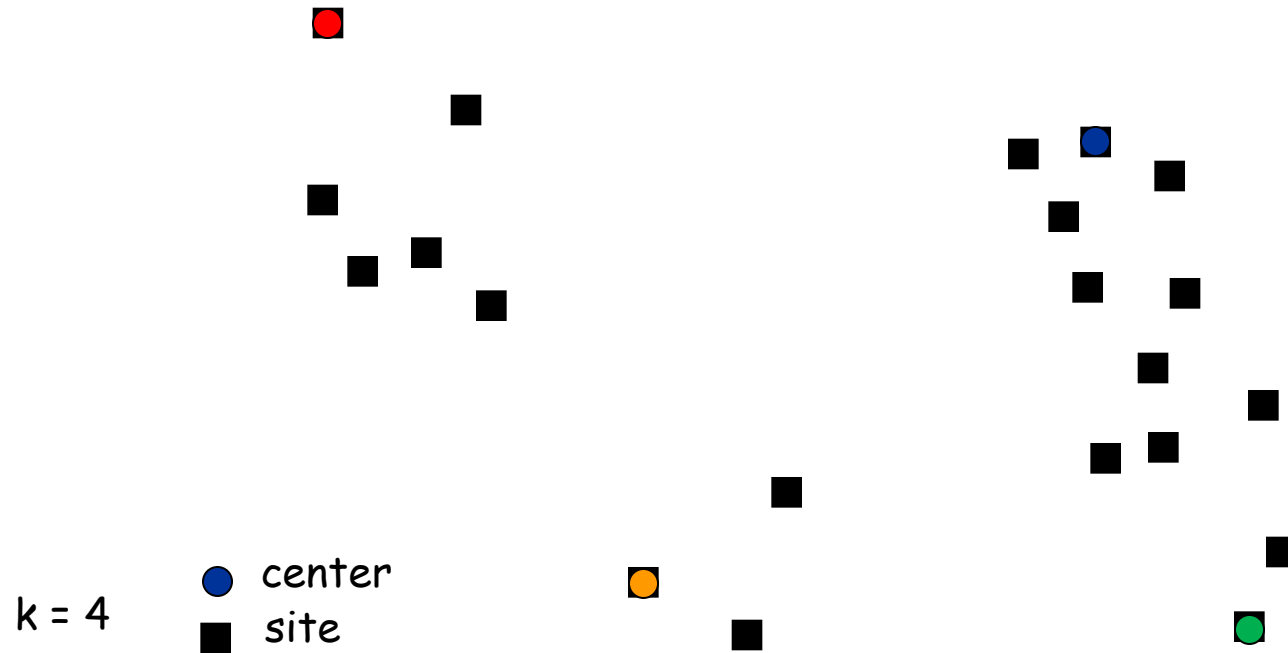
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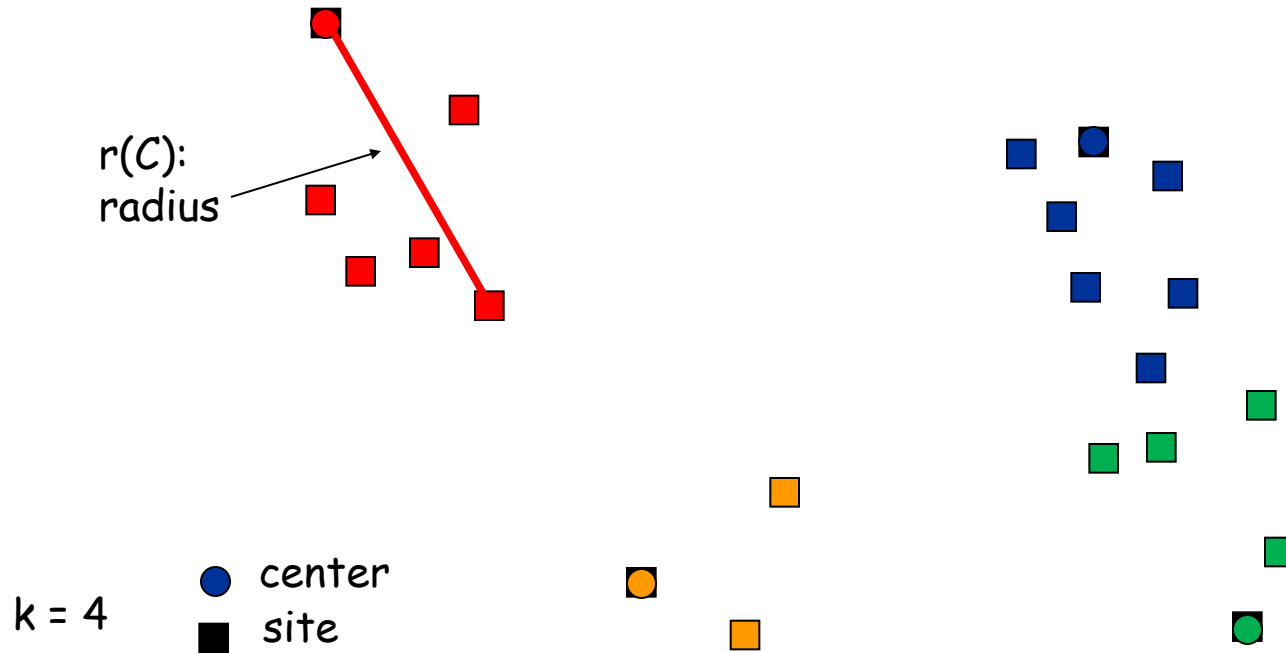
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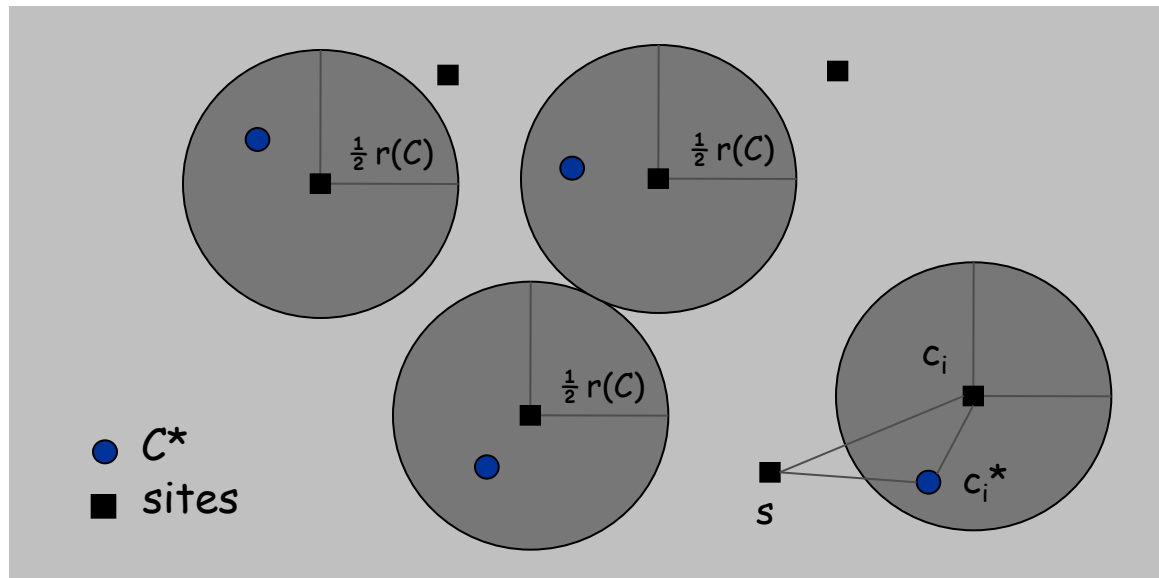
Center Selection: Analysis of Greedy Algorithm

balls are disjoint since all centers in C are pairwise at distance at least $r(C)$

Theorem. Let C^* be an optimal set of centers. Then $r(C) \leq 2r(C^*)$.

Pf. (by contradiction) Assume $r(C^*) < \frac{1}{2} r(C)$.

- For each site c_i in C , consider ball of radius $\frac{1}{2} r(C)$ around it.
- Exactly one c_i^* in each ball;
 - each ball with center $c_i \in C$ must contain a center in C^* (otherwise $\text{dist}(c_i, C^*) \geq \frac{1}{2} r(C) > r(C^*)$);
 - balls are disjoint and $|C| = |C^*|$.



Center Selection: Analysis of Greedy Algorithm

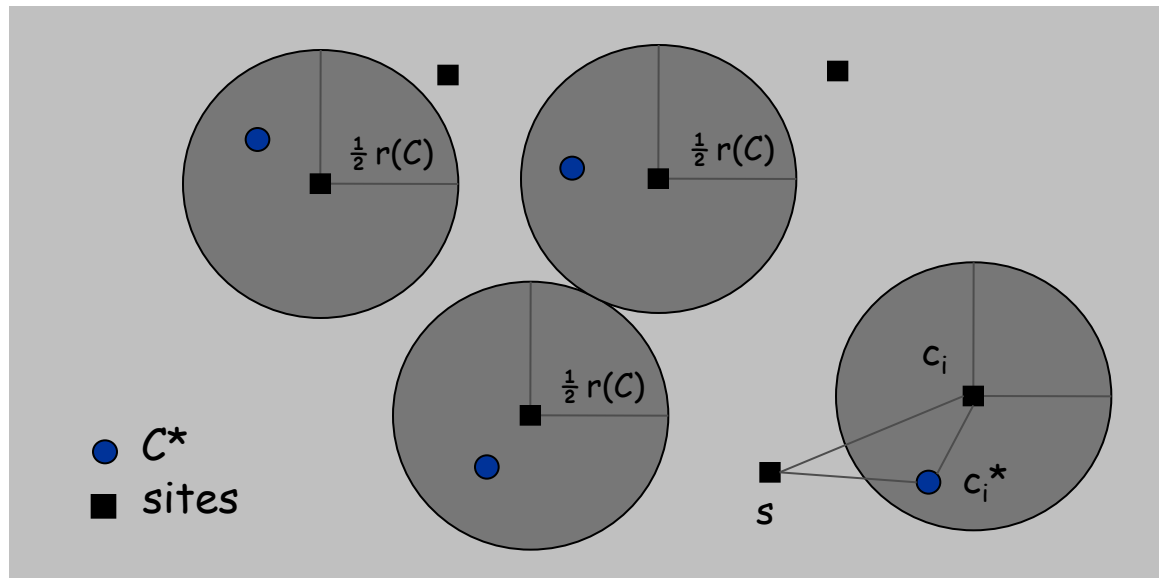
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- For each site c_i in C , consider ball of radius $\frac{1}{2} r(C)$ around it.
- Exactly one c_i^* in each ball; let c_i be the site paired with c_i^* .
- Consider any site s and its closest center c_i^* in C^* .
- $\text{dist}(s, C) \leq \text{dist}(s, c_i) \leq \text{dist}(s, c_i^*) + \text{dist}(c_i^*, c_i) \leq 2r(C^*)$.
- Thus $r(C) \leq 2r(C^*)$. ■

△-inequality

$\leq r(C^*)$ since c_i^* is closest center



Center Selection

Theorem. Let C^* be an optimal set of centers. Then $r(C) \leq 2r(C^*)$.

Theorem. Greedy algorithm is a 2-approximation for center selection problem.

Remark. Greedy algorithm always places centers at sites, but is still within a factor of 2 of best solution that is allowed to place centers anywhere.

↖
e.g., points in the plane

Question. Is there hope of a better approximation?

...very unlikely:

Theorem. Unless $P = NP$, there no ρ -approximation for center-selection problem for any $\rho < 2$.