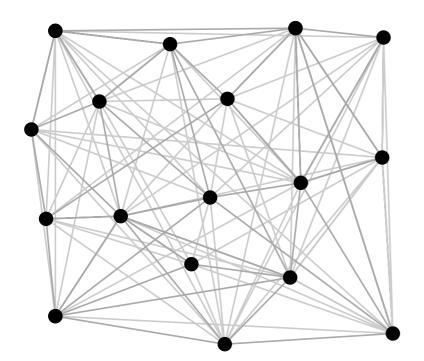
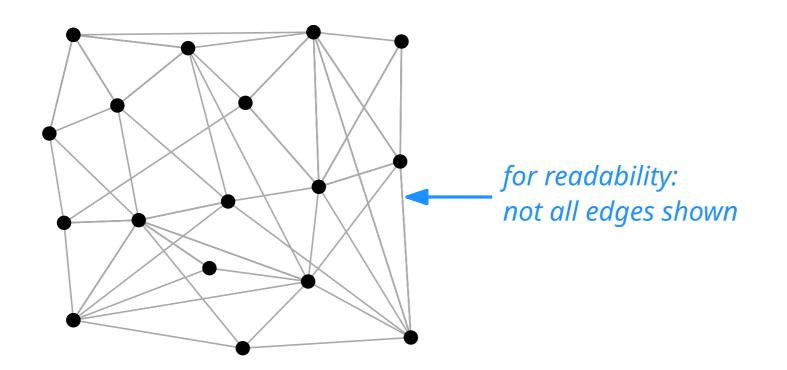
k-CENTER via Parametric Pruning

Given: A complete graph G=(V,E) with edge costs $c\colon E\to \mathbb{Q}_{\geq 0}$ satisfying the triangle inequality

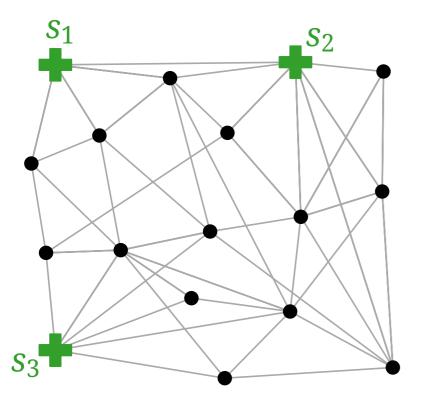


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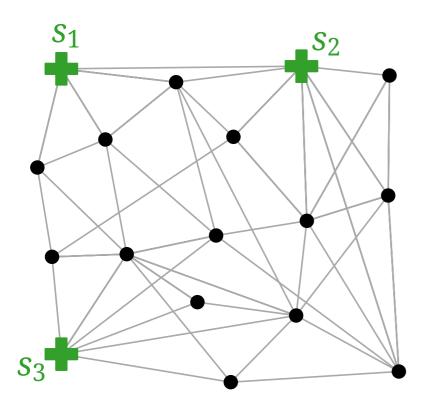


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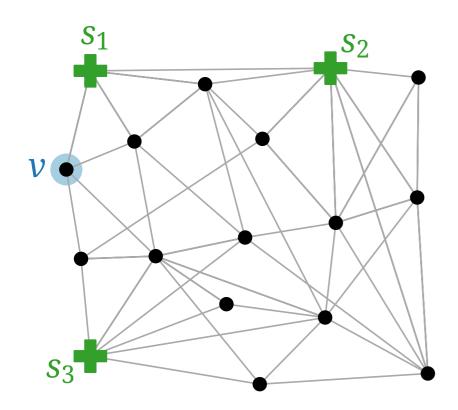
For each vertex set $S \subseteq V$



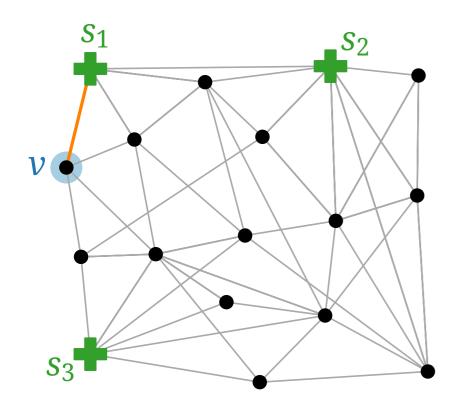
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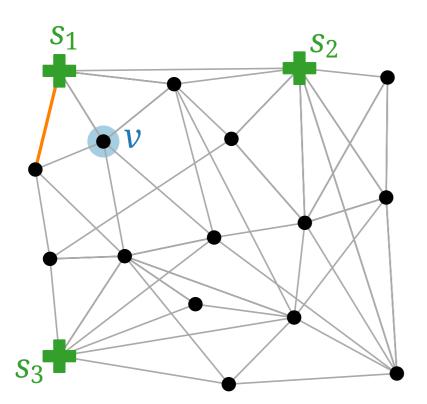
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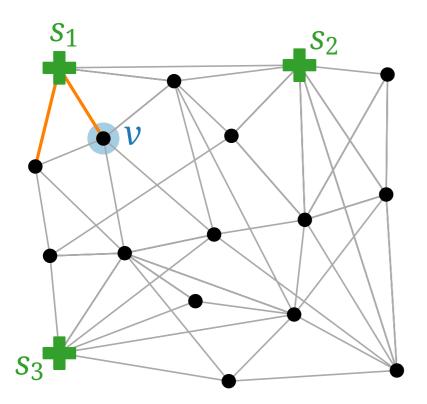
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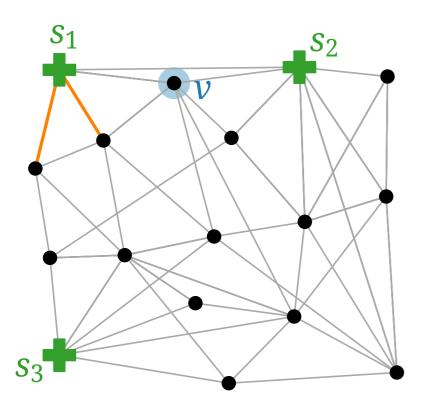
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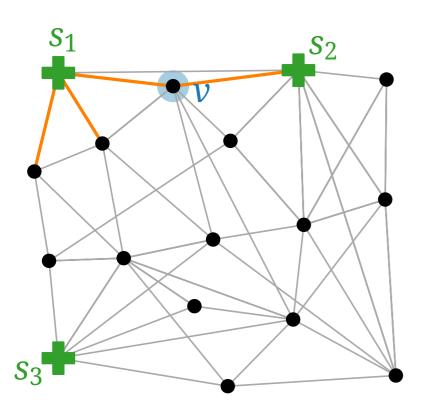
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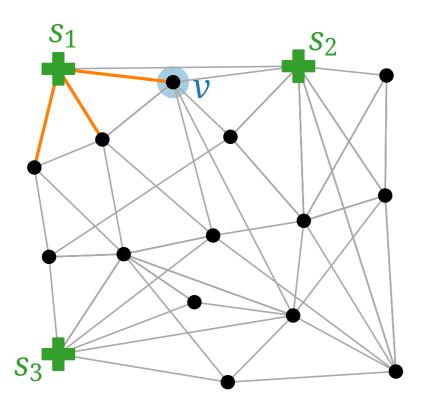
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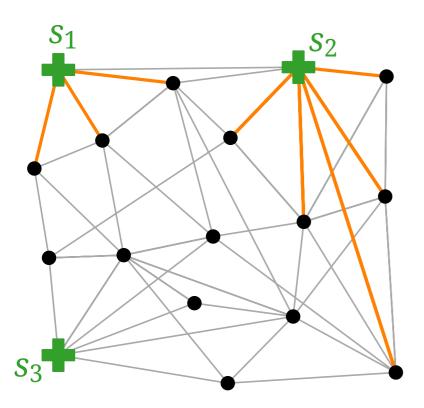
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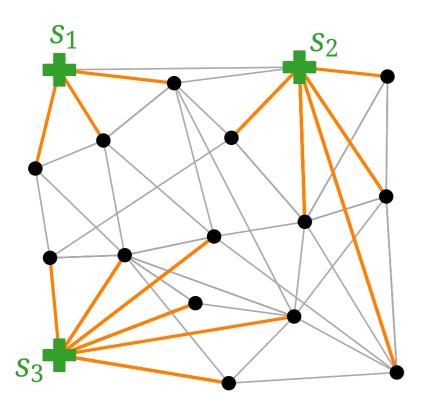
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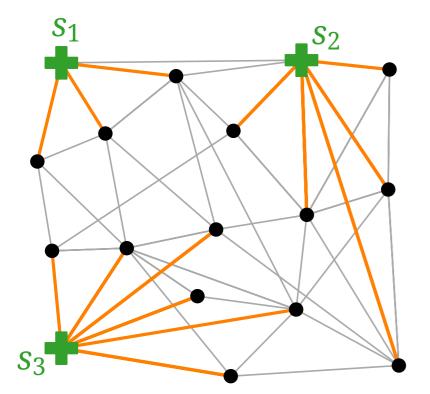


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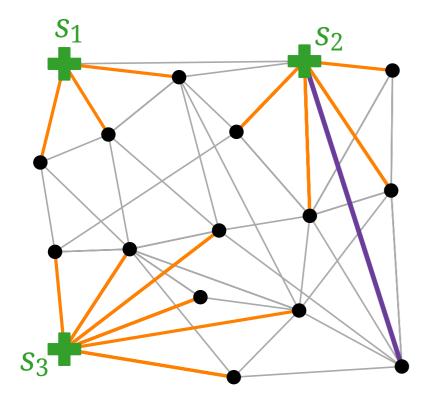
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$$cost(S) := max_{v \in V} c(v, S)$$



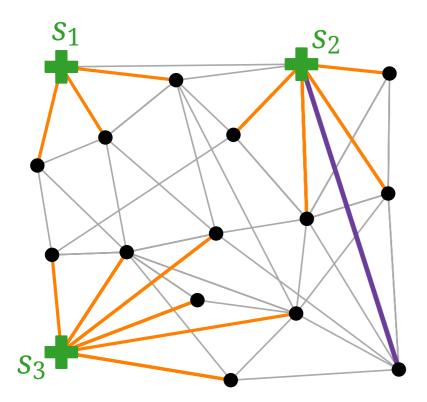
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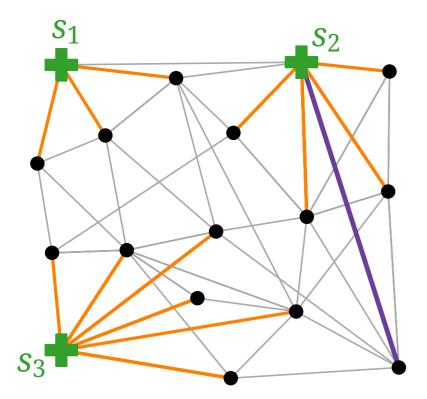
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For each vertex set $S \subseteq V$, c(v, S) is the cost of the cheapest edge from v to a vertex in S.



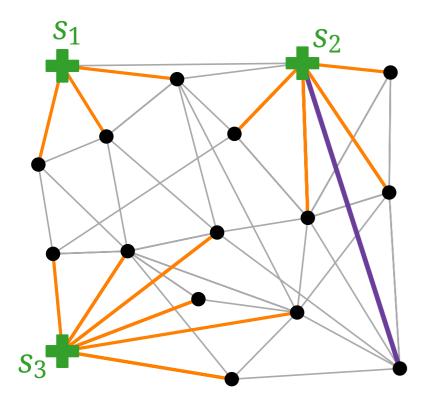
Given: A complete graph G = (V, E) with edge costs $c \colon E \to \mathbb{Q}_{\geq 0}$ satisfying the triangle inequality and a natural number $k \leq |V|$.

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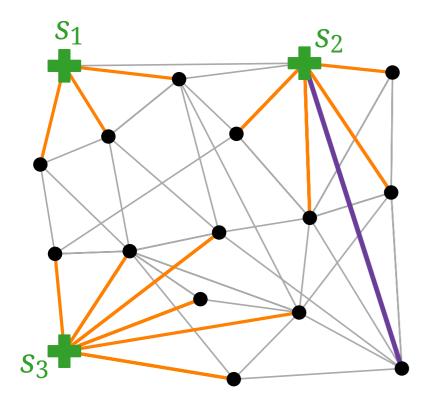
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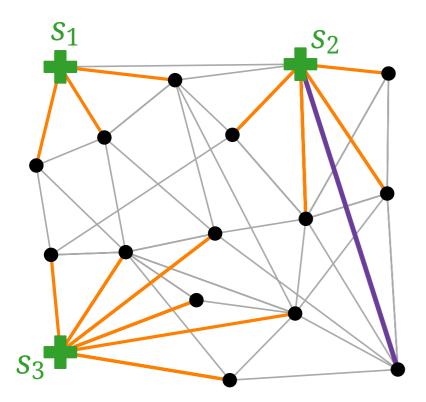
Find: A k-element vertex set S such that $cost(S) := max_{v \in V} c(v, S)$ is minimized.

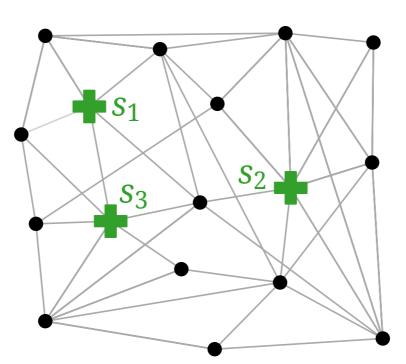


optimal for k = 3? better solution?

Given: A complete graph G = (V, E) with edge costs $c : E \to \mathbb{Q}_{\geq 0}$ satisfying the triangle inequality and a natural number $k \leq |V|$.

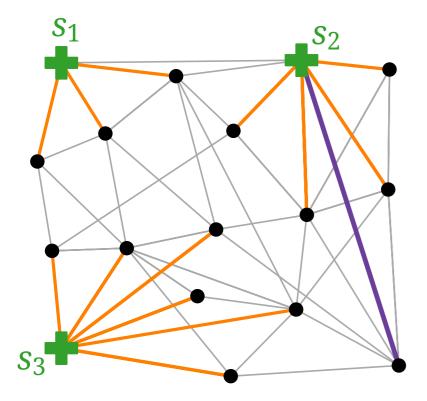
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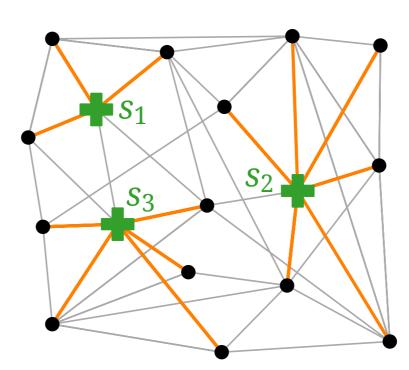




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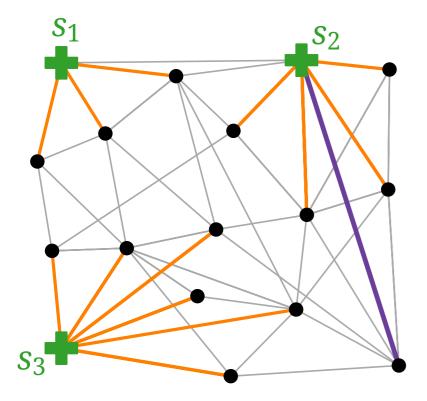
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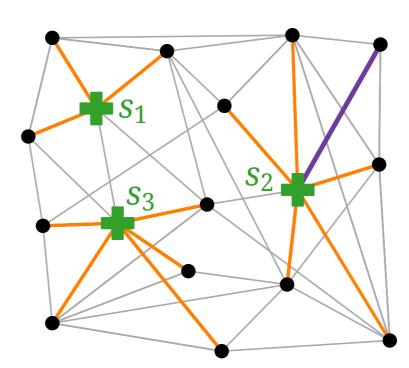




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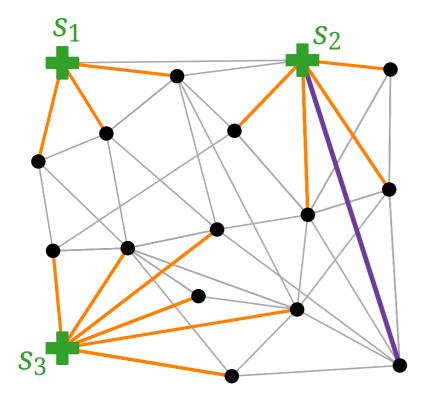


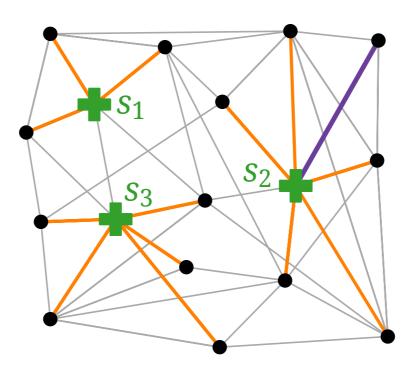
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Find: A k-element vertex set S such that $cost(S) := max_{v \in V} c(v, S)$ is minimized.

today: 2-approximation of metric *k*-center using *parametric pruning.*





Idea 1: Reduce optimization problem to decision problem

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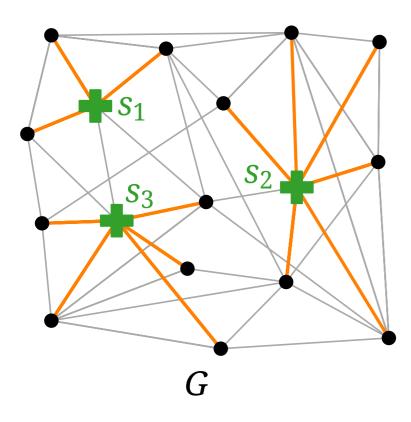
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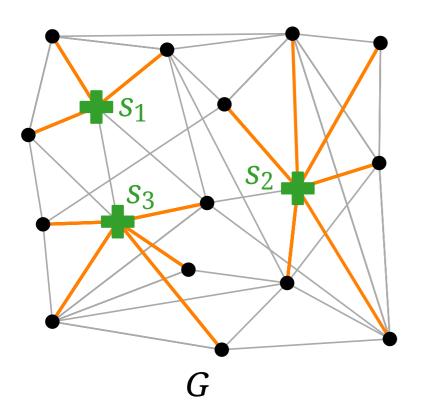
Idea 3 (lower bound): factor- α approximation algorithm consists of two steps

- 1.) use family of instances I(t) to compute lower bound t^* for OPT
- 2.) find solution in instance $I(\alpha t^*)$

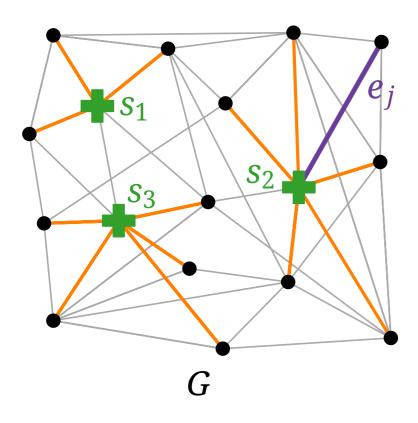
Metric k-center: What is the decision problem? How do the pruned instances I(t) look like?



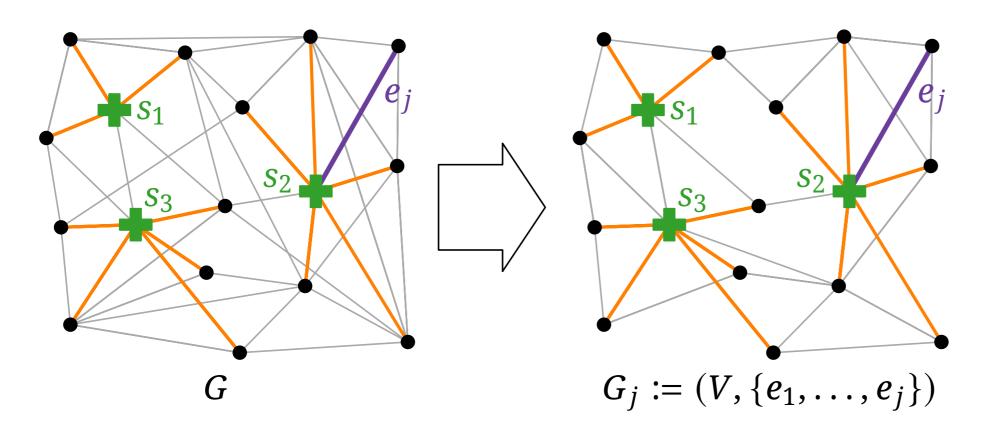
Let $E = \{e_1, \ldots, e_m\}$ with $c(e_1) \leq \ldots \leq c(e_m)$.



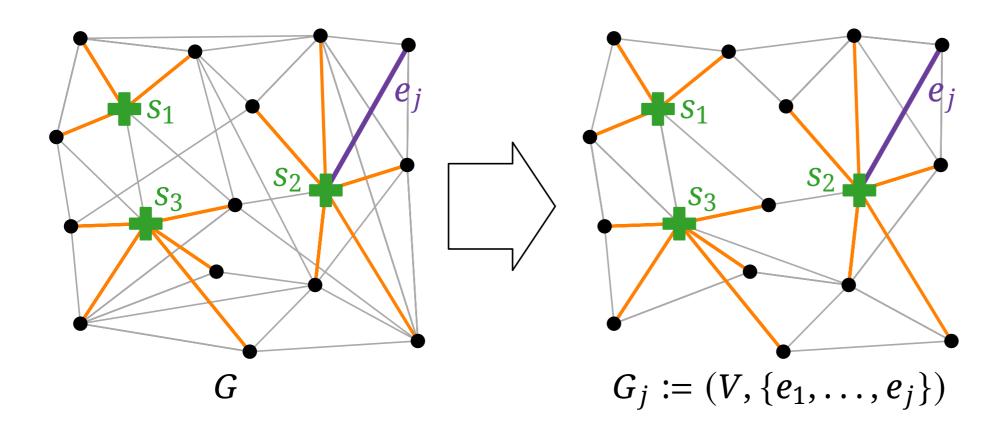
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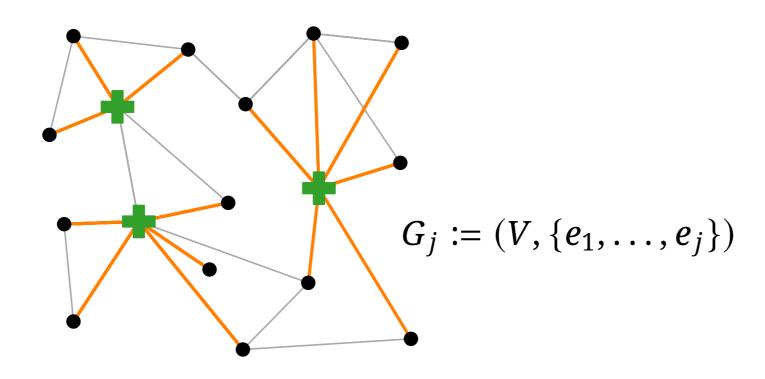


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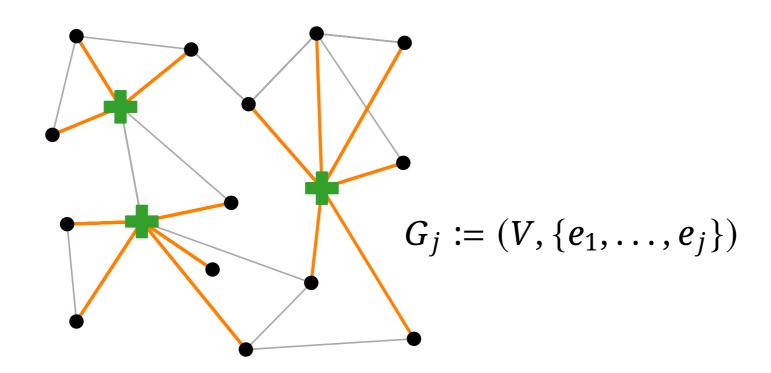


...try each pruned instance G_j .

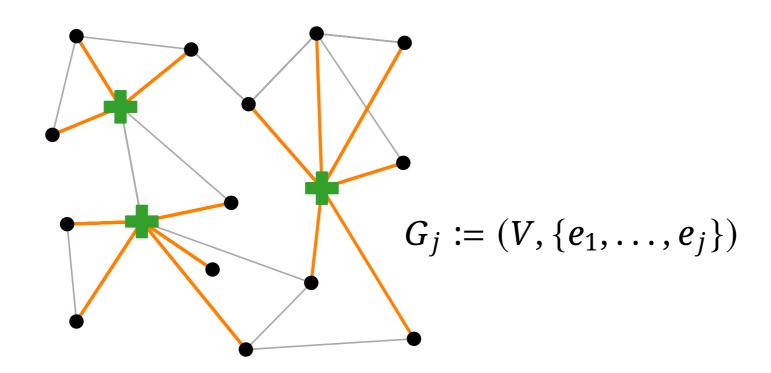
Def.



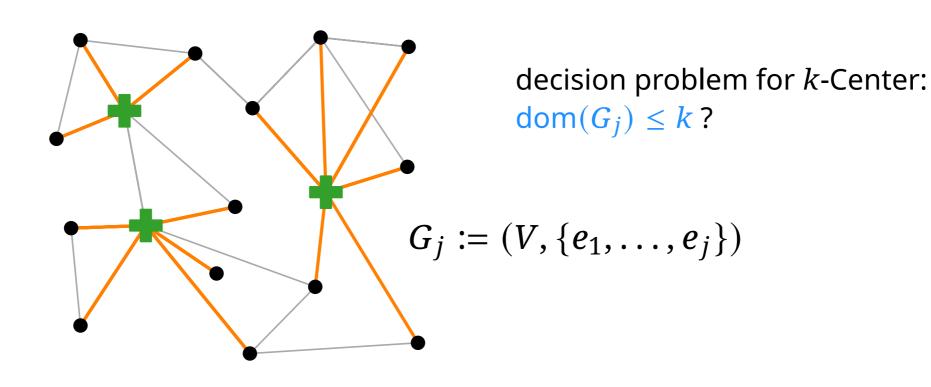
Def. A vertex set D of a graph H is dominating if each vertex is either in D or adjacent to a vertex in D.



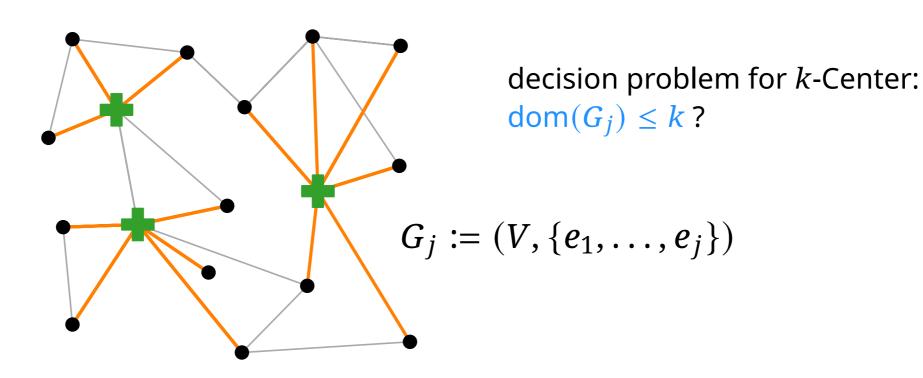
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...but computing dom(H) is NP-hard.

Square of a Graph – Lower bounding k-Center

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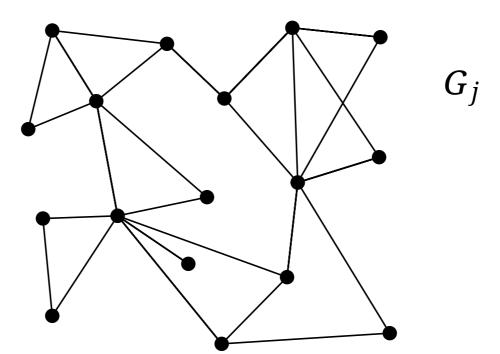
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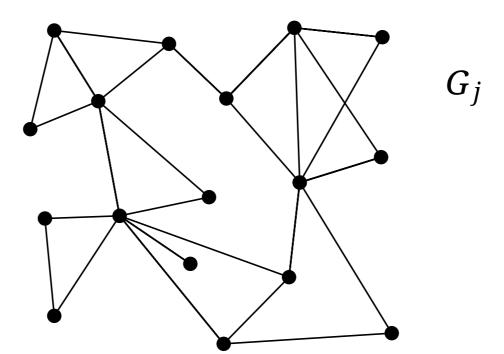
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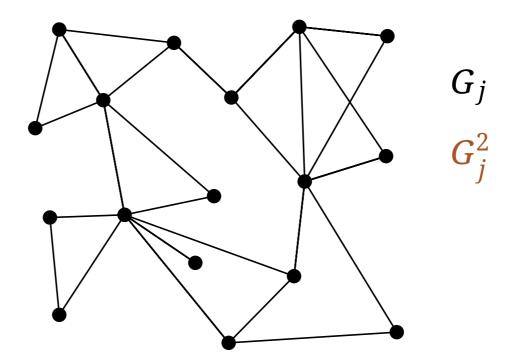
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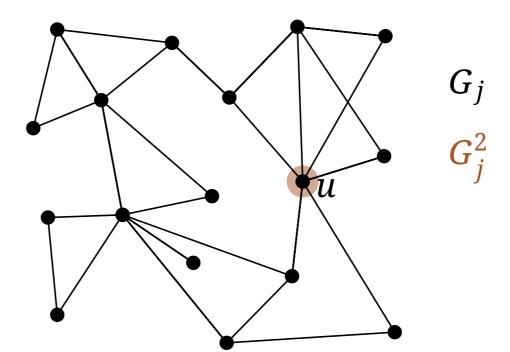
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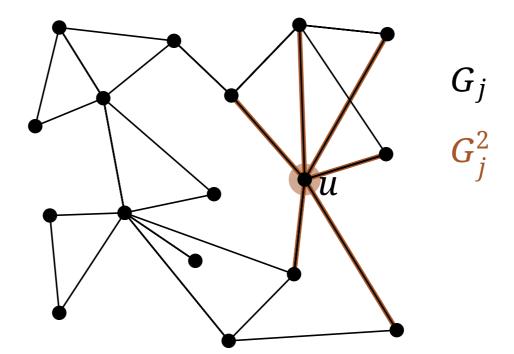
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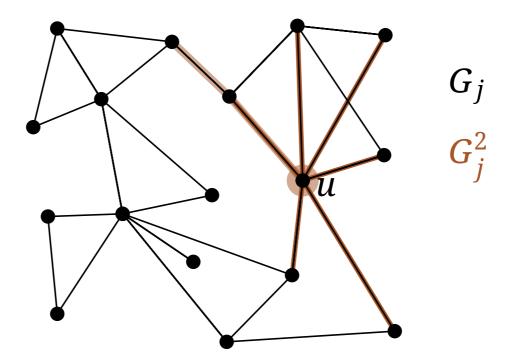
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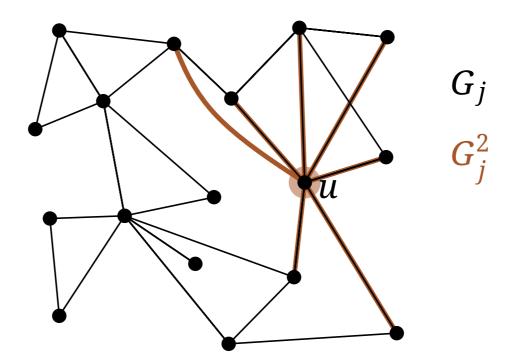
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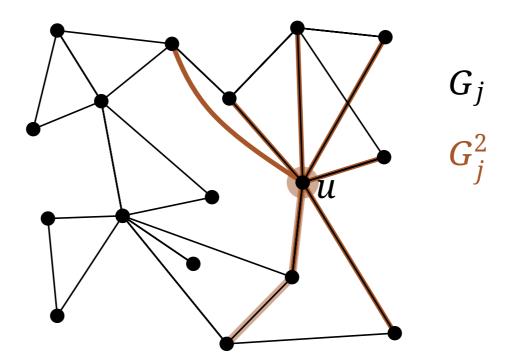
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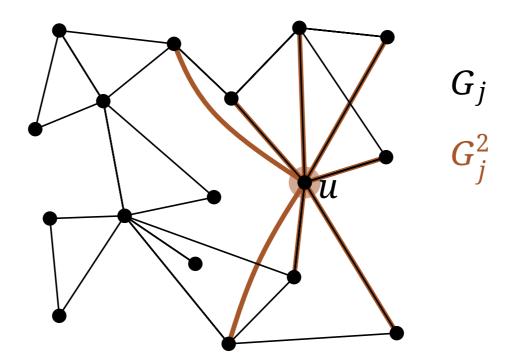
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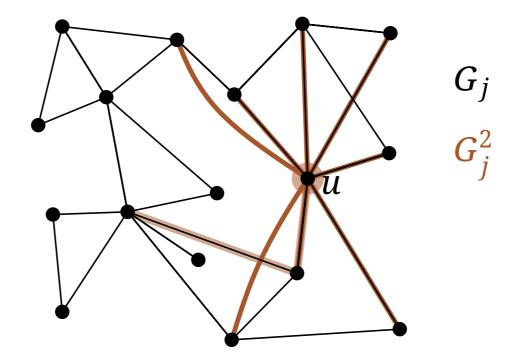
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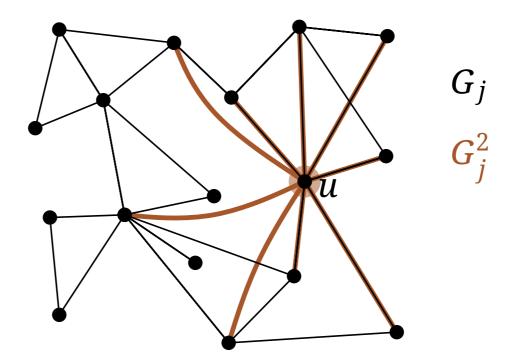
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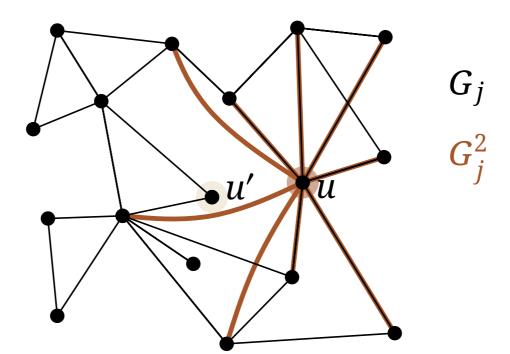
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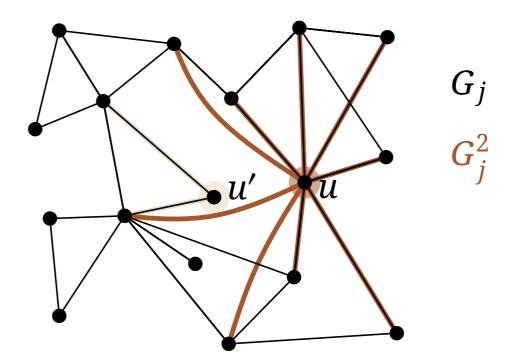
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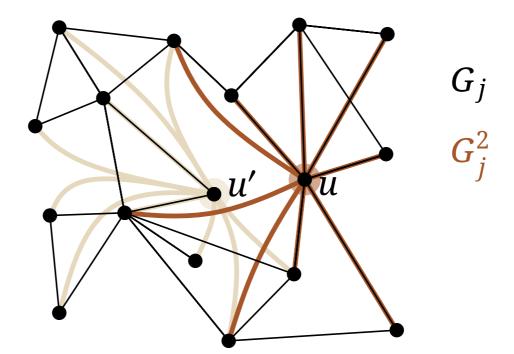
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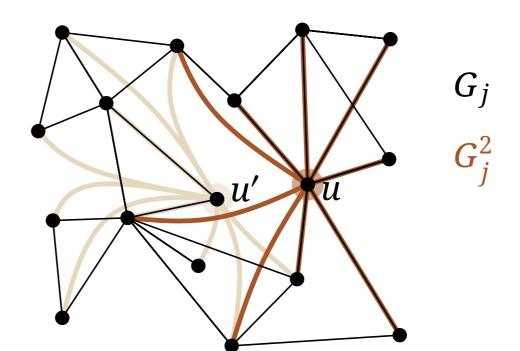


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Obs. If OPT $\geq c(e_j)$ then a dominating set with at most k elements in G_j^2 is a 2-approximation for metric k-CENTER.

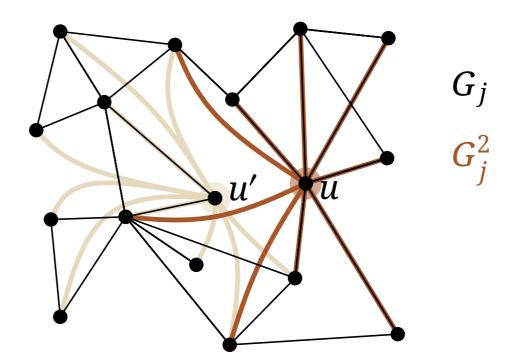


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Why?

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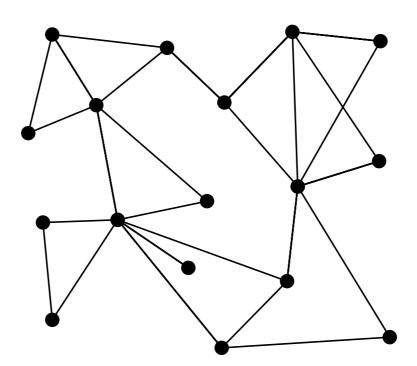
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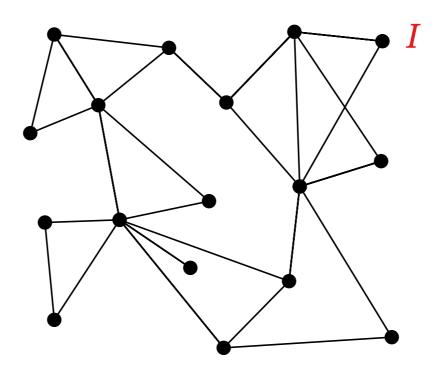
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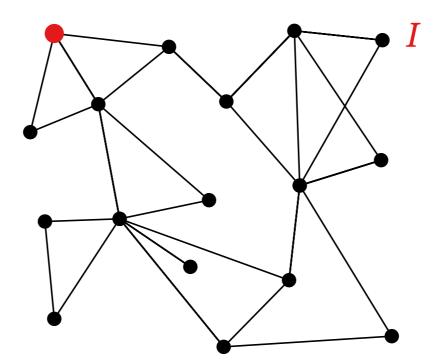
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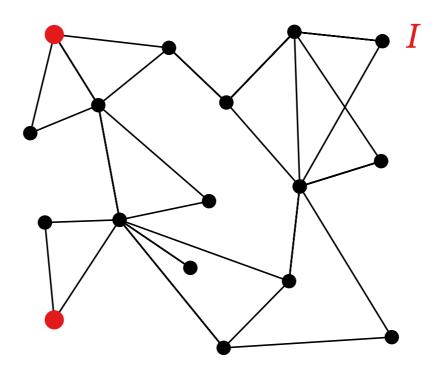
 G_j U' U G_j

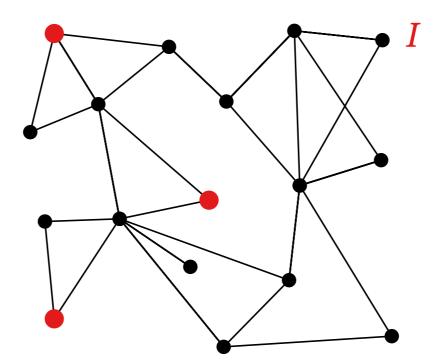
Why? triangle inequality

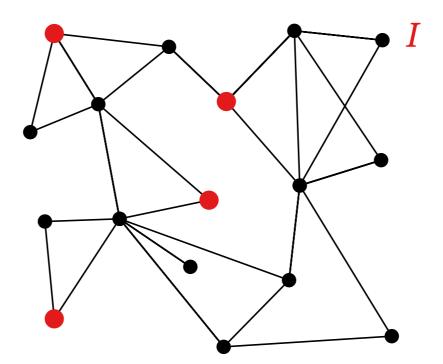


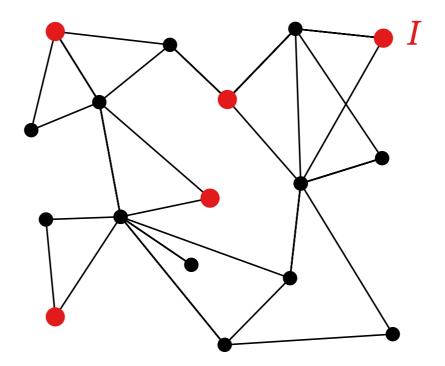


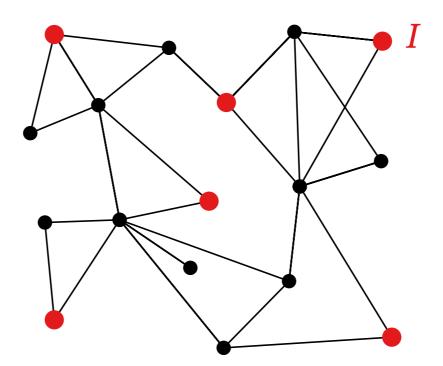


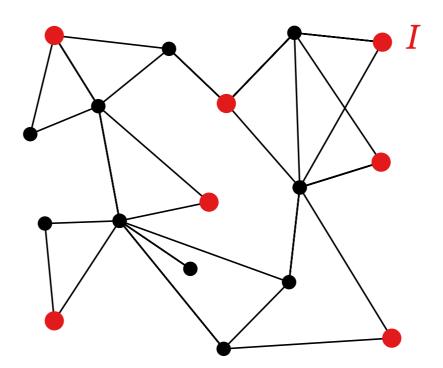




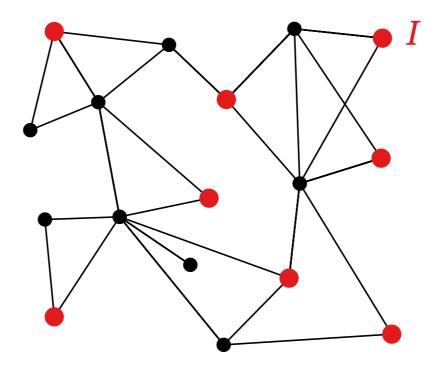




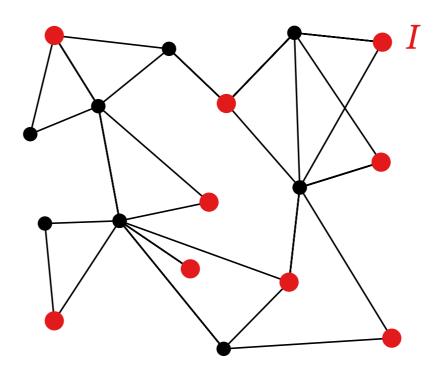




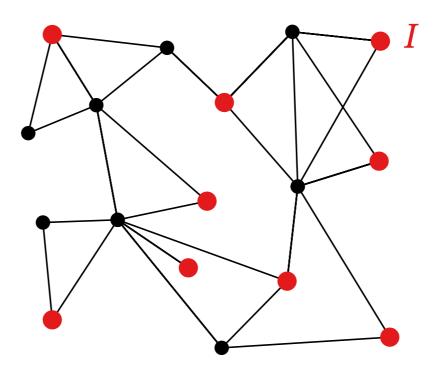
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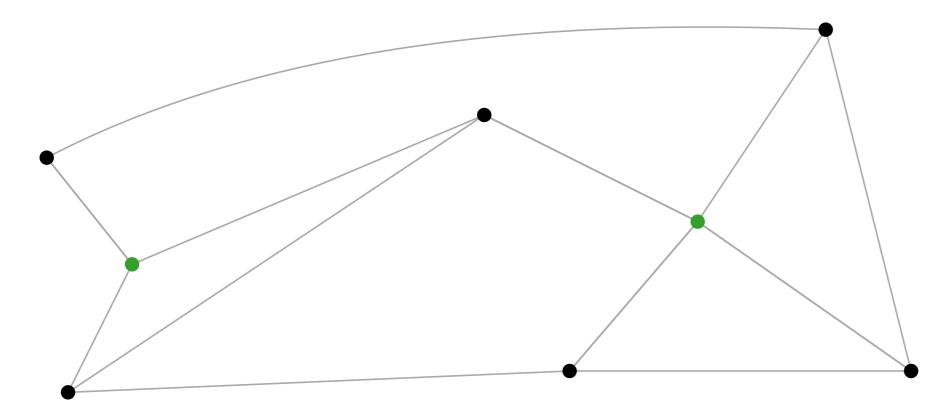
lependent ninating

Obs. Maximal independent sets are dominating sets!

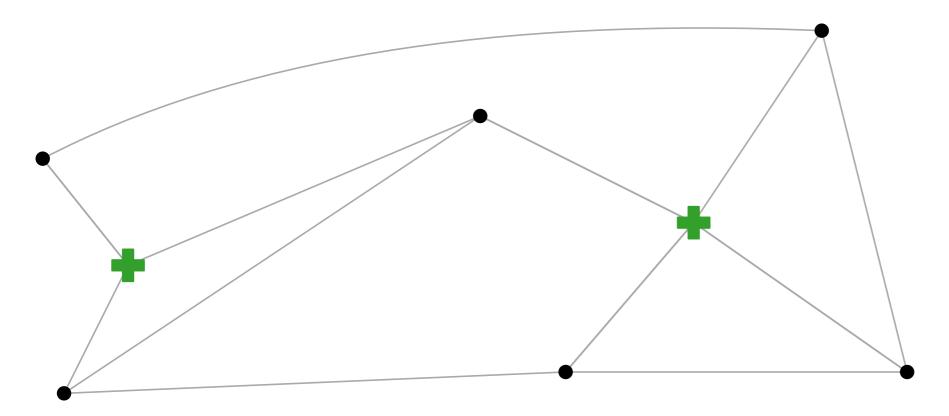
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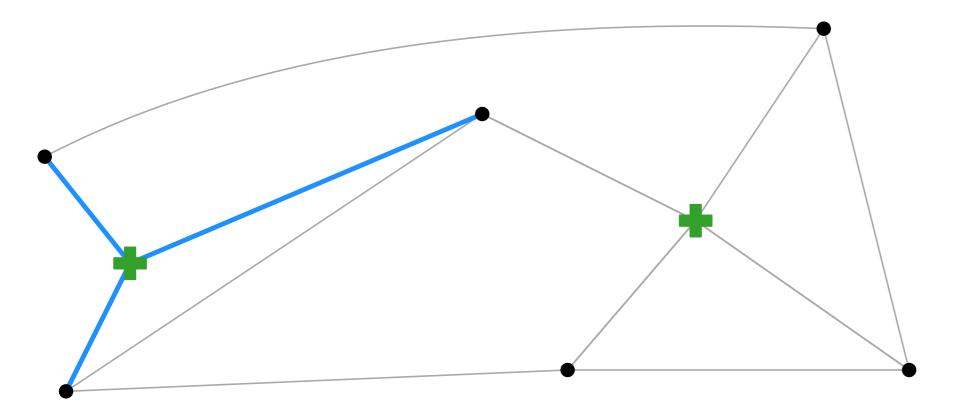
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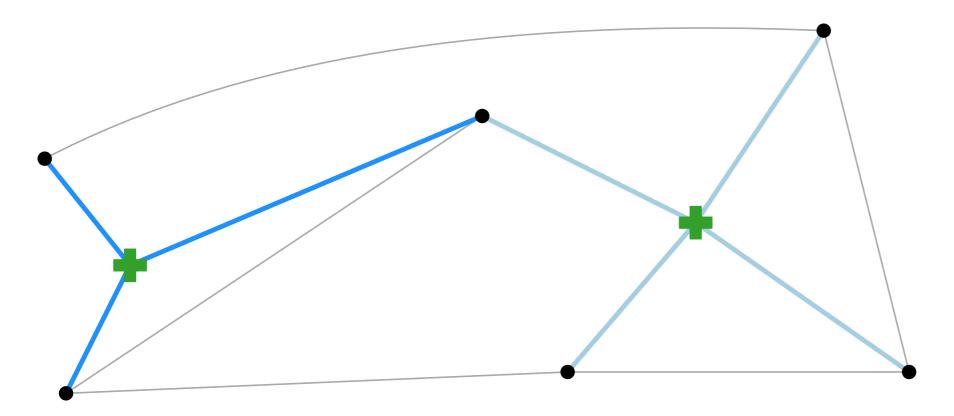
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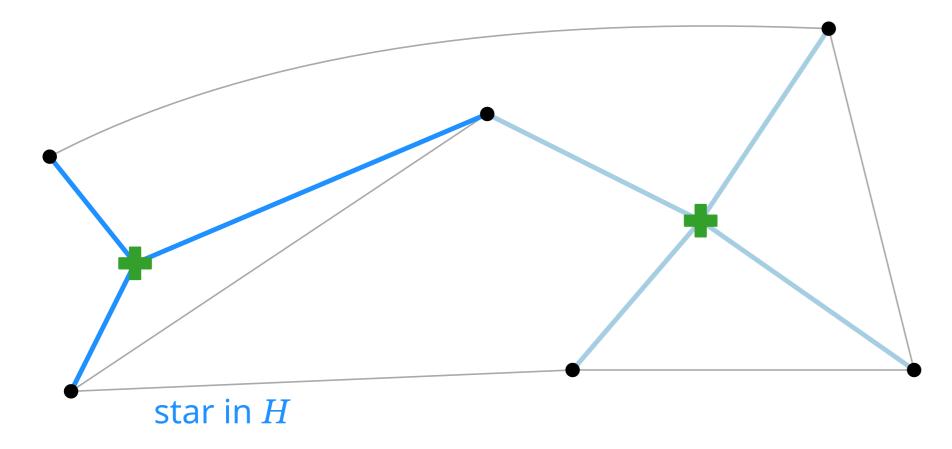
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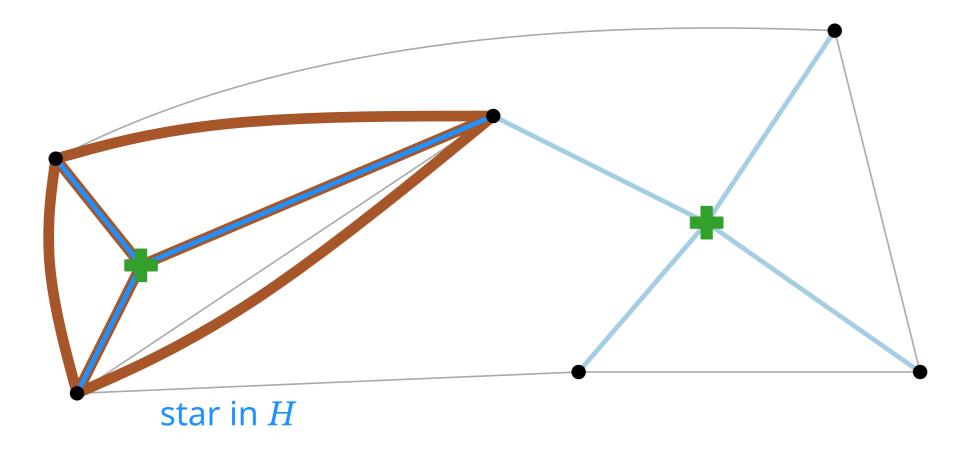
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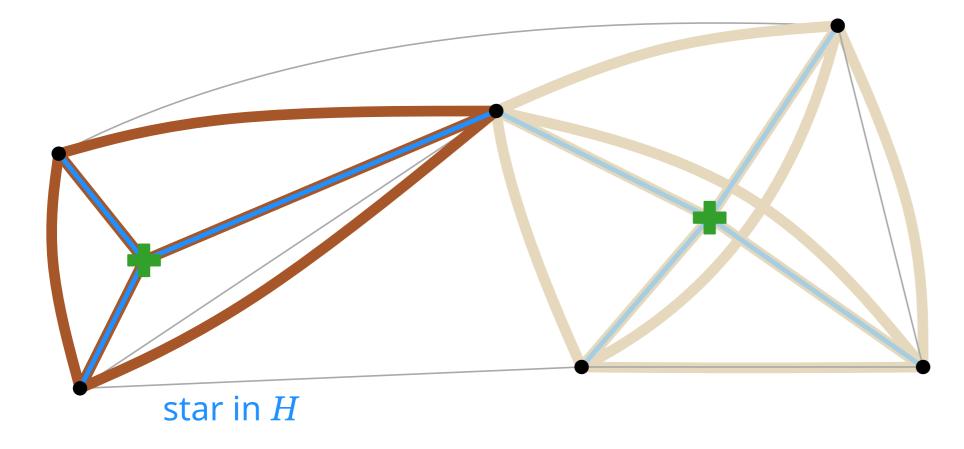
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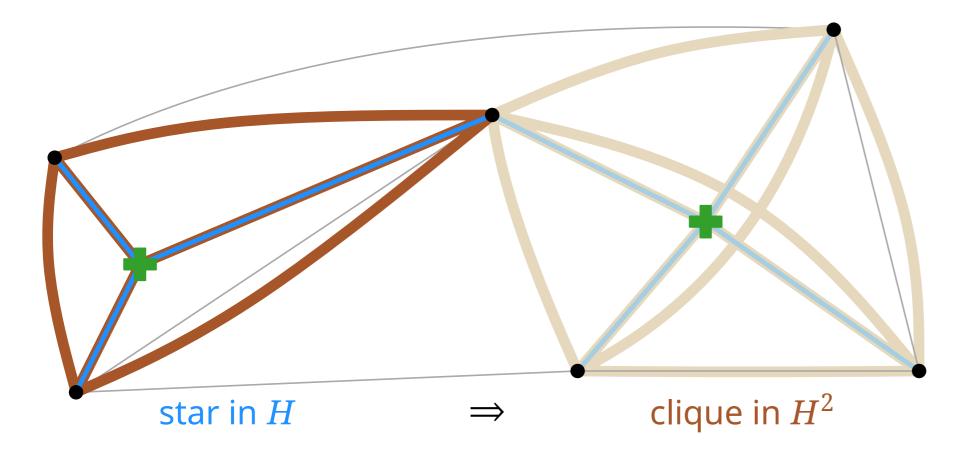
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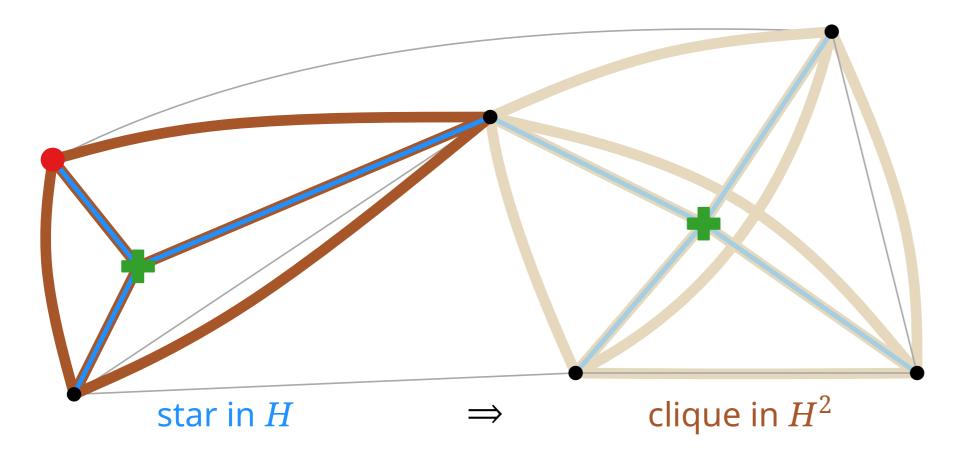
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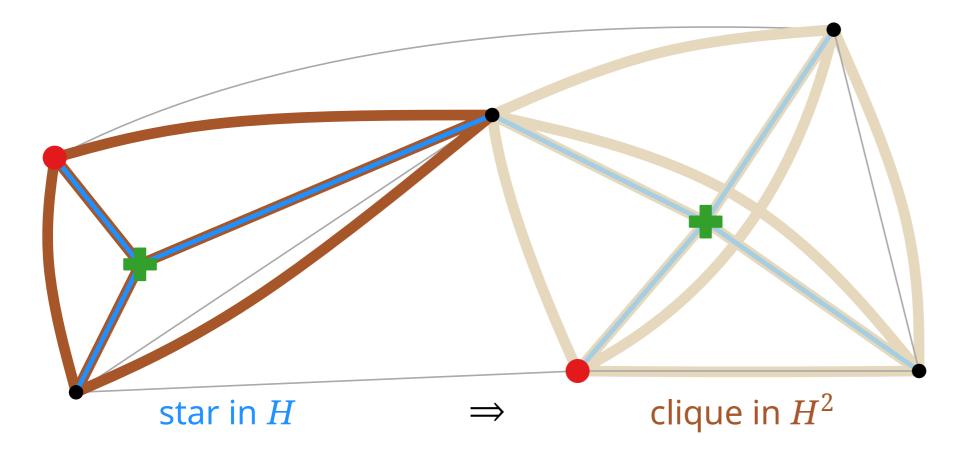
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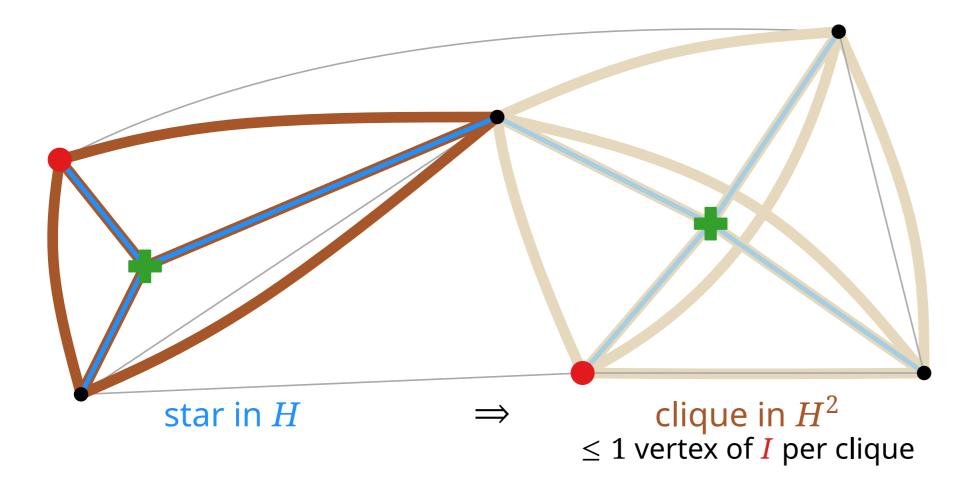
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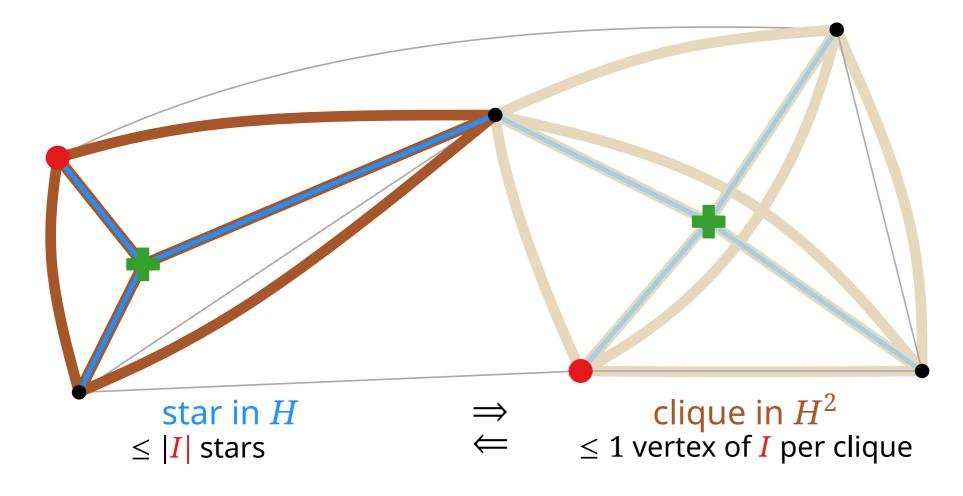
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Sort the edges of G by cost: $c(e_1) \leq \ldots \leq c(e_m)$

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recall: parametric pruning:

- reduce to decision problem, work on pruned instance I(t)
- step 1: Use family of I(t) to compute lower bound t^*
- step 2: find solution in $I(\alpha t^*)$

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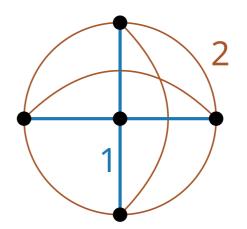
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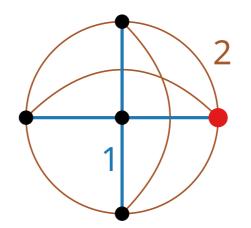
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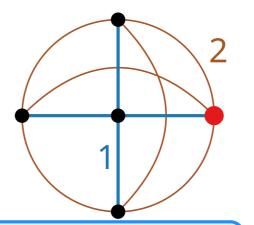
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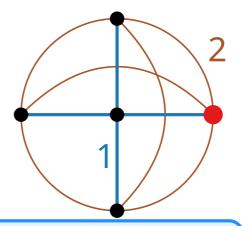
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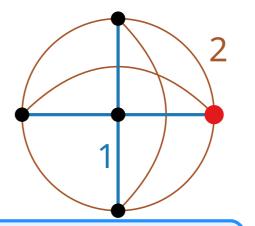


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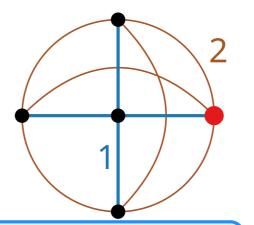
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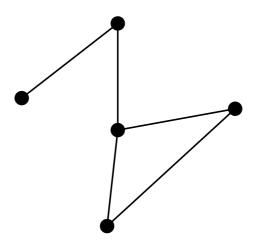


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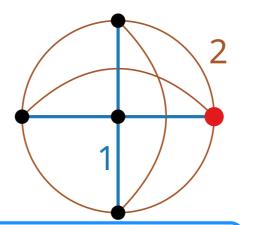
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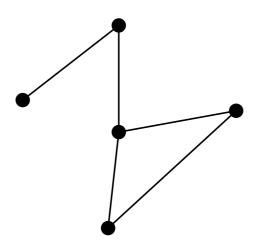


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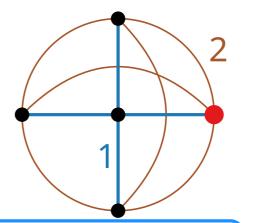
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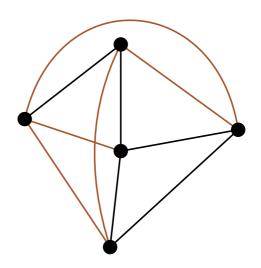


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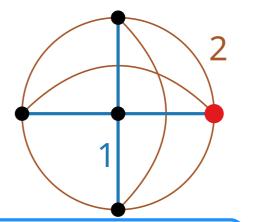
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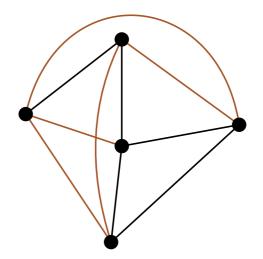
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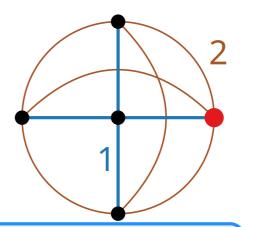
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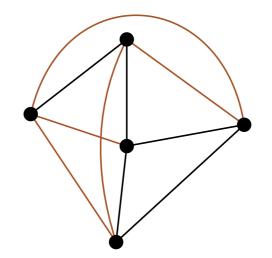
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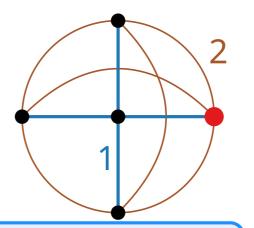
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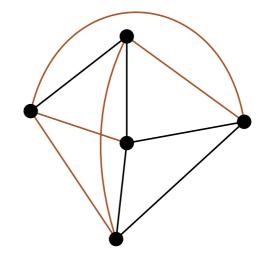
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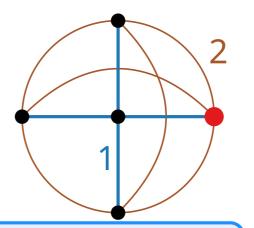
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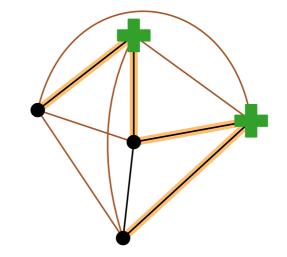
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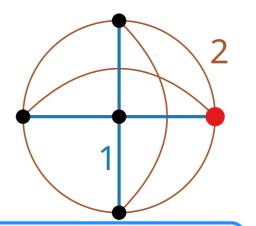
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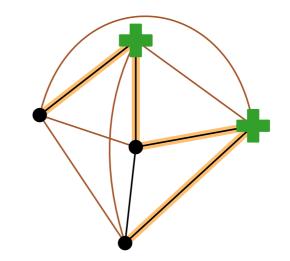
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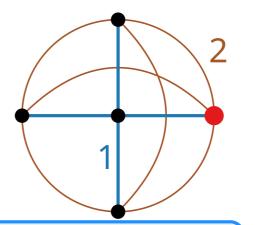
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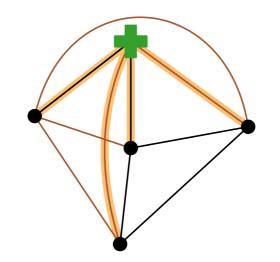
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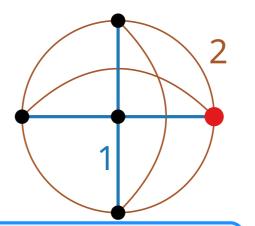
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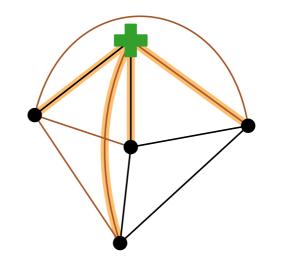
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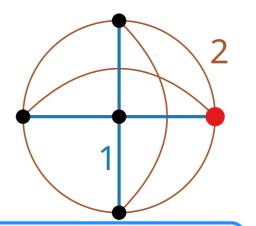
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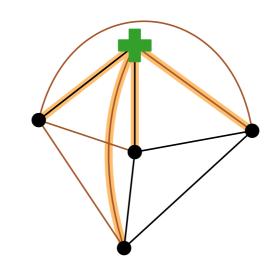
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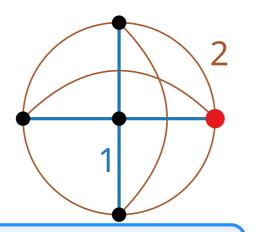
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Why does this prove that,

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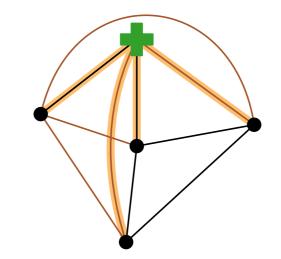
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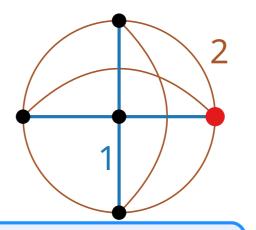
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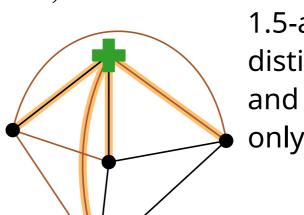
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1.5-approx. allows to distinguish between 1 and 2, and these are the only possible answers.

3-approximation for METRIC-WEIGHTED-CENTER

Metric-k-Center

Given: A complete graph G = (V, E) with metric edge costs $c: E \to \mathbb{Q}_{>0}$ and a natural number $k \leq |V|$.

For $S \subseteq V$, c(v, S) is the cost of the cheapest edge from v to a vertex in S.

Find: A k-element vertex set S such that $cost(S) := max_{v \in V} c(v, S)$ is minimized.

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Given: A complete graph G = (V, E) with metric edge costs $c: E \to \mathbb{Q}_{\geq 0}$ and a natural number $k \leq |V|$.

For $S \subseteq V$, c(v, S) is the cost of the cheapest edge from v to a vertex in S.

Find: A k-element vertex set S such that $cost(S) := max_{v \in V} c(v, S)$ is minimized.

METRIC-K-CENTER WEIGHTED

Given: A complete graph G = (V, E) with metric edge costs

 $c: E \to \mathbb{Q}_{\geq 0}$ and a natural number $k \leq |V|$., vertex weights

 $w: V \to \mathbb{Q}_{\geq 0}$ and a budget $W \in \mathbb{Q}_+$

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```
Algorithm Metric-
                      -CENTER
  Sort the edges of G by cost : c(e_1) \leq \ldots \leq c(e_m)
  for j = 1, \ldots, m do
      Construct G_i^2
      Find a maximal independent set I_j in G_i^2
      if |I_j| \leq k then
        return I_i
```

```
Algorithm Metric-Weighted-CENTER
  Sort the edges of G by cost : c(e_1) \le ... \le c(e_m)
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       Find a maximal independent set I_j in G_i^2
                                              what about the weights?
      if |I_j| \leq k then
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```

```
Algorithm Metric-Weighted-CENTER
   Sort the edges of G by cost : c(e_1) \leq \ldots \leq c(e_m)
   for j = 1, \ldots, m do
       Construct G_{:}^{2}
       Find a maximal independent set I_j in G_i^2
                                                 what about the weights?
       if |I_j| \leq k then
           return I<sub>i</sub>
     s_j(u) := \text{lightest node in } N_{G_i}(u) \cup \{u\}
```

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       Compute S_i := \{ s_i(u) \mid u \in I_i \}
      if |I_j| \le k then w(S_j) \le W return I_j
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      if |I_j| \le k then w(S_j) \le W return I_j u \in I_j
```

```
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      if |I_j| \le k then return I_{S_i}
                          w(S_j) \leq W
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       if |I_j| \le k then return I_{S_i}
                           \sim w(S_j) \leq W
                                                               |s_j(u)|
\leq 3c(e_j) because of
```

triangle inequality

```
Algorithm Metric-Weighted-CENTER
   Sort the edges of G by cost: c(e_1) \leq \ldots \leq c(e_m)
   for j = 1, \ldots, m do
       Construct G_i^2
       Find a maximal independent set I_j in G_i^2
       Compute S_i := \{ s_i(u) \mid u \in I_i \}
      if |I_j| \le k then w(S_j) \le W return S_j
                                                            |s_j(u)| \le 3c(e_j) because of
```

triangle inequality

$$s_j(u) := \text{lightest node in } N_{G_i}(u) \cup \{u\}$$

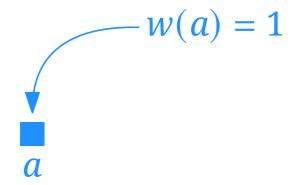
Theorem. The above is a factor-3 approximation algorithm for Metric-Weighted-Center.

Here, we need to have a budget W, and edge costs satisfying the triangle inequality.

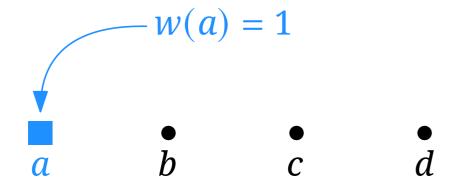
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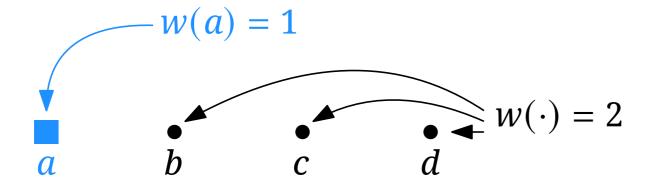
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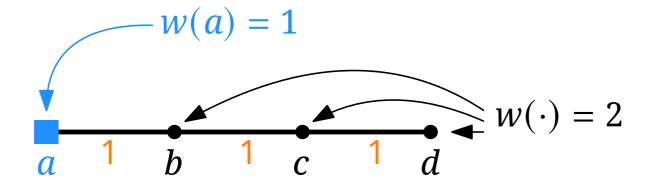
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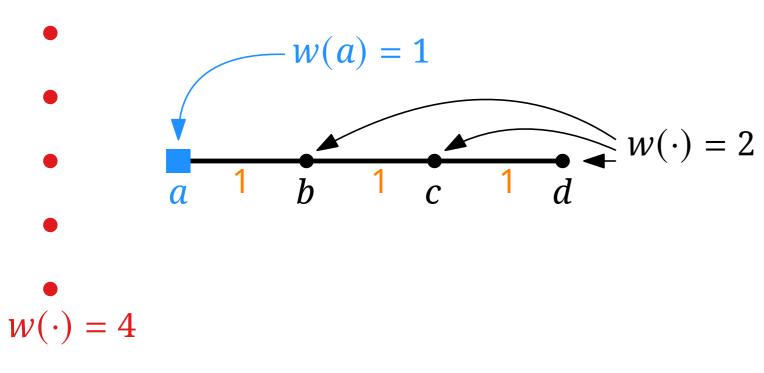
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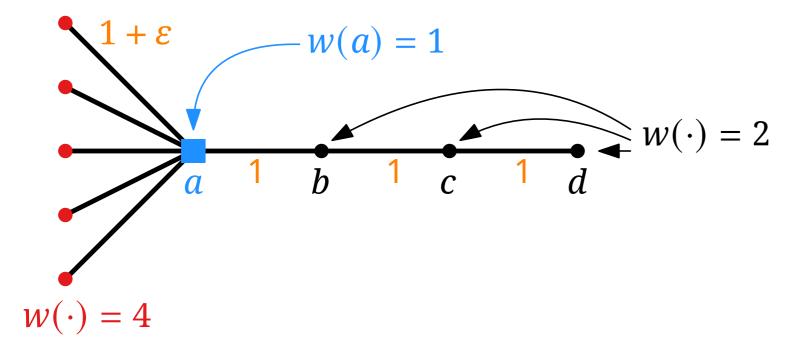
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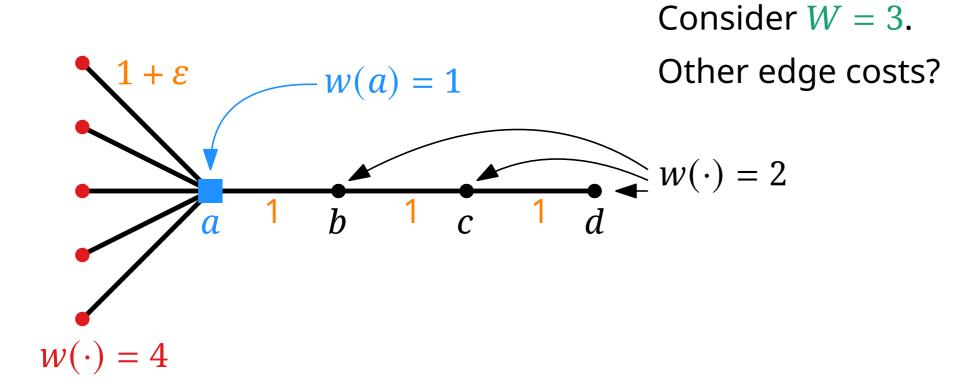


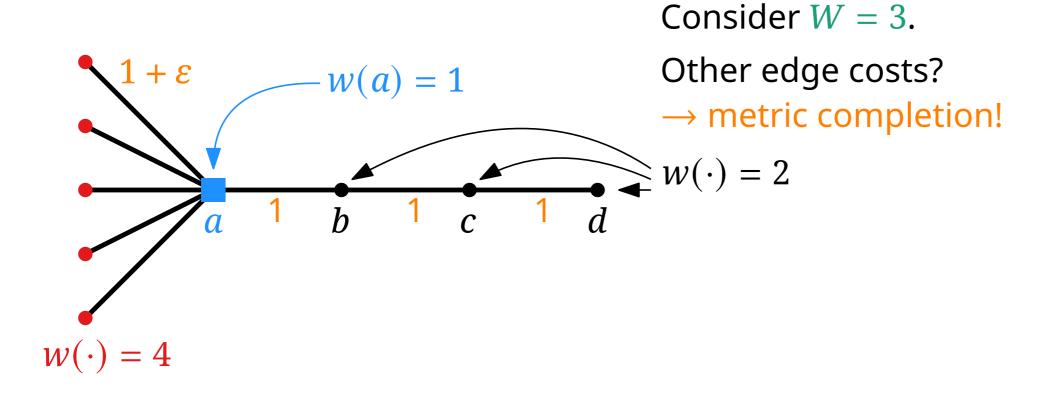
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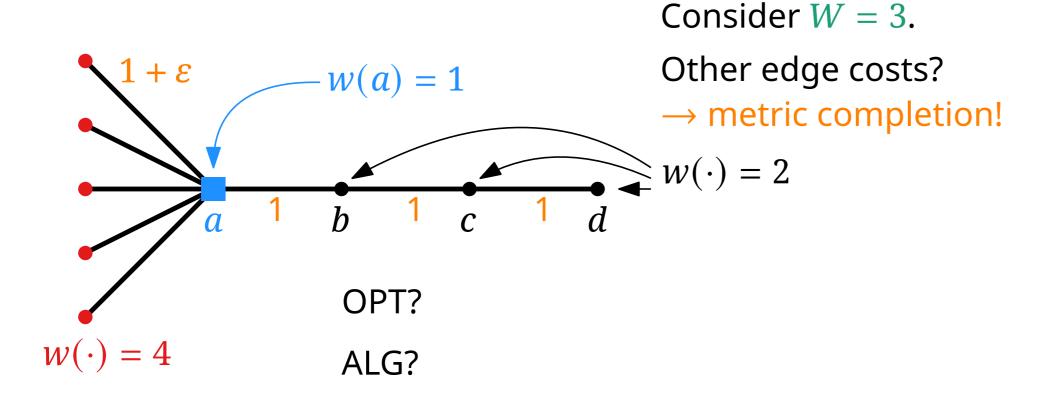


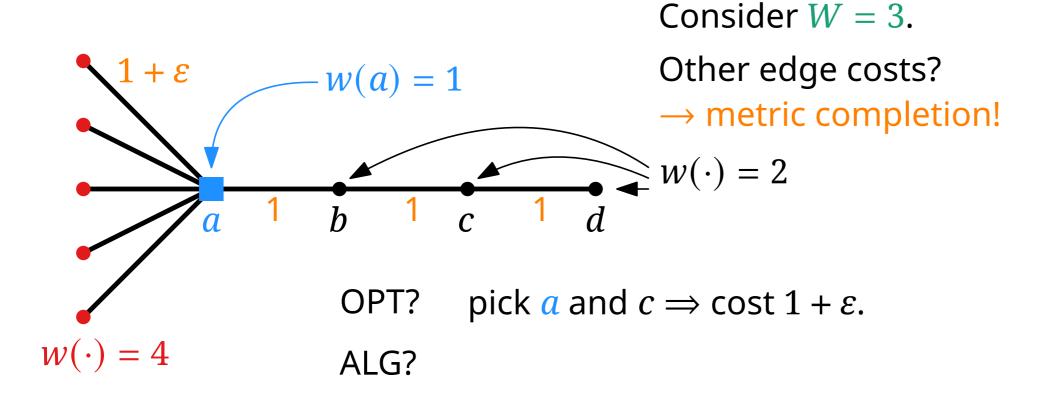
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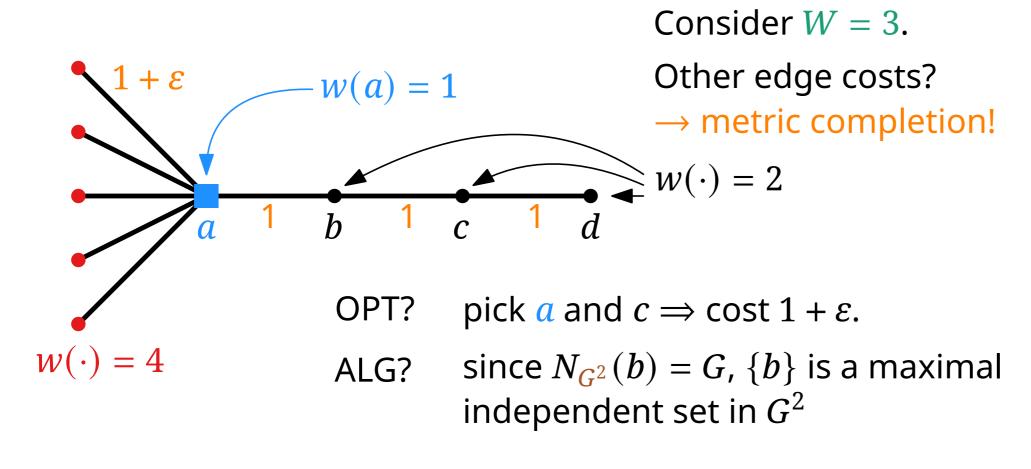


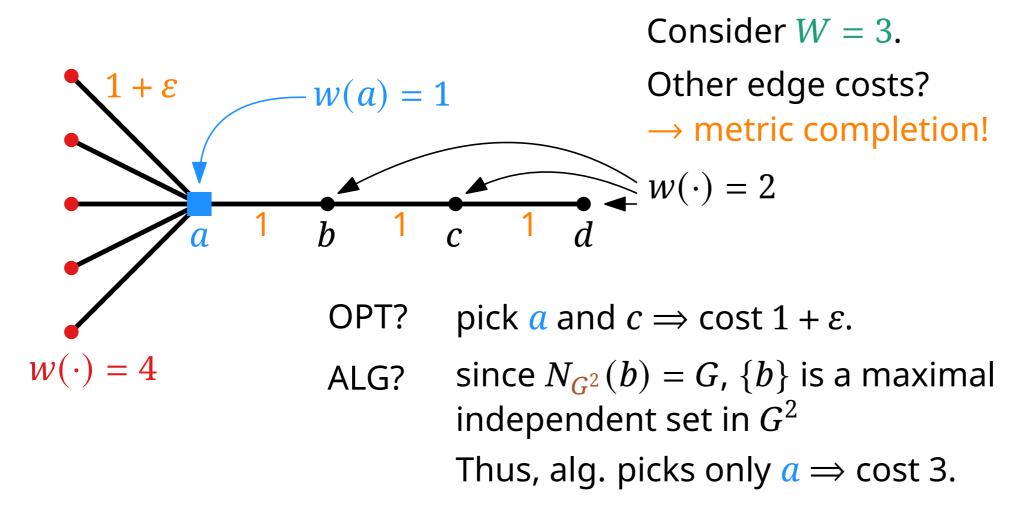




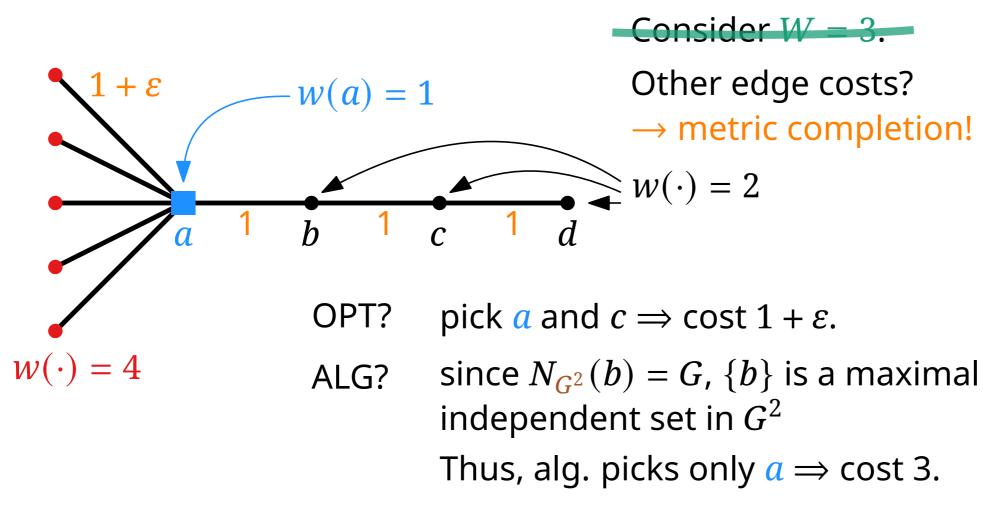






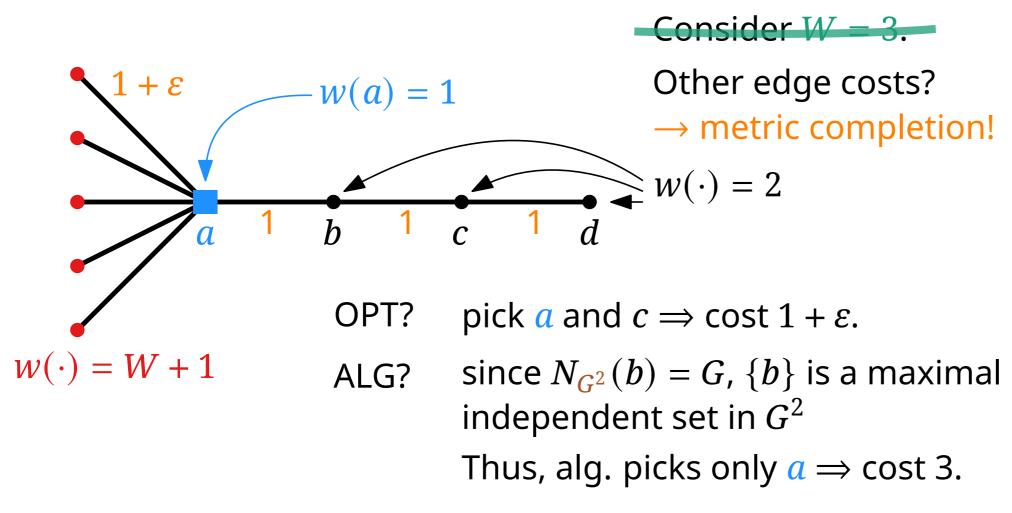


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How can we generalize this to larger W?

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How can we generalize this to larger W?

Summary

```
parametric pruning: decision problem \rightarrow pruned instance I(t) \rightarrow lower bound t^* opt OPT & find solution in I(\alpha t^*)
```

Metric k-Center: 2-approximation using pruned instances G_j , and finding a maximal independent set in G_j^2 (square of G_j)

Metric Weighted-Center: 3-approximation by same algorithm but taking for each v in independent set, lowest weight vertex in neighborhood

Note Alternative 2-approximation algorithm (Gonzalez 1985): Pick an arbitrary vertex as first center, then greedily add the vertex with largest distance (cost) to centers selected, until *k* centers have been selected.