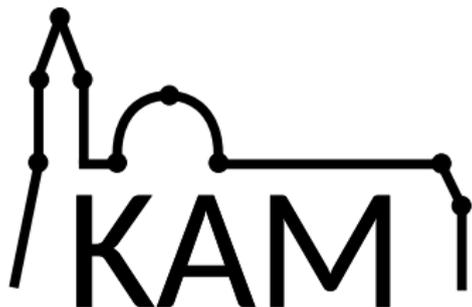


Capacitated k -Center in Low Doubling and Highway Dimension

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OF MATHEMATICS
AND PHYSICS
Charles University

Capacitated k -Center

Input

- ▶ graph $G = (V, E)$ with edge lengths $\ell: E \rightarrow \mathbb{R}^+$,
- ▶ integer k ,
- ▶ capacities $c: V \rightarrow \mathbb{N}$.

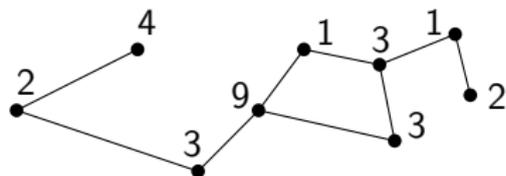
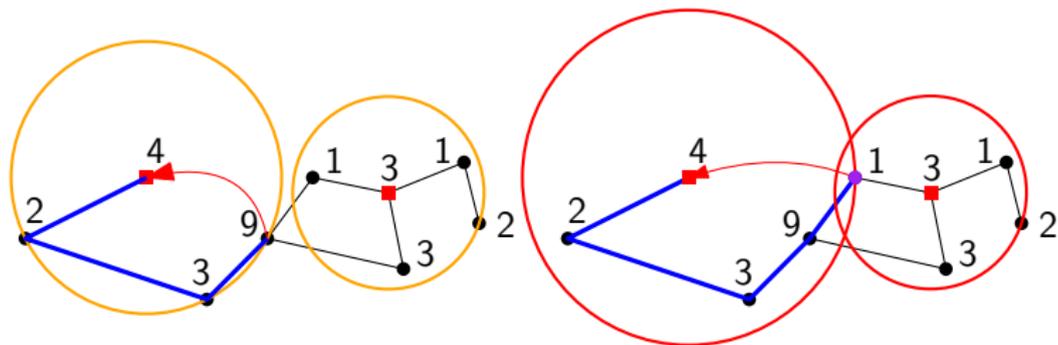


Figure: CKC input with $k = 2$.

Capacitated k -Center: Goal

Find $S \subseteq V$ and an *assignment* $\varphi: (V \setminus S) \rightarrow S$ such that

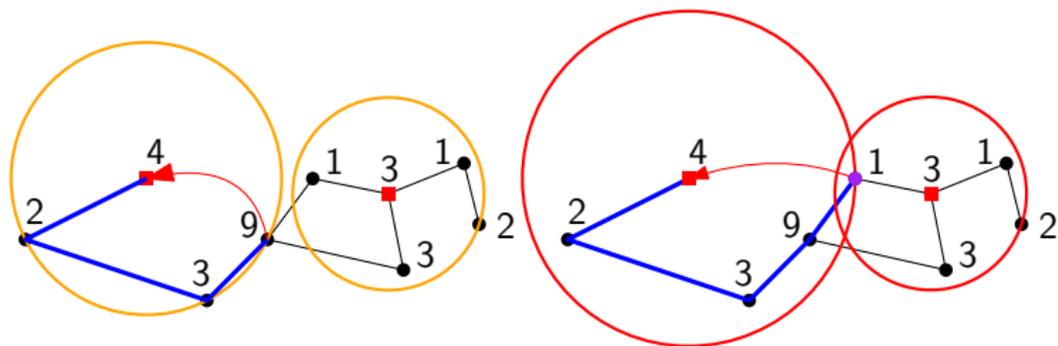
- ▶ $|S| \leq k$,
- ▶ for every $u \in S$, $|\varphi^{-1}(u)| \leq c(u)$, and
- ▶ $\max_{v \in V \setminus S} \text{dist}(v, \varphi(v))$ is minimal.



Capacitated k -Center: Goal

Find $S \subseteq V$ and an *assignment* $\varphi: (V \setminus S) \rightarrow S$ such that

- ▶ $|S| \leq k$,
- ▶ for every $u \in S$, $|\varphi^{-1}(u)| \leq c(u)$, and
- ▶ $\max_{v \in V \setminus S} \text{dist}(v, \varphi(v))$ is minimal.



When $c(u) = |V|$ for every $u \in V \Rightarrow k$ -CENTER.

Capacitated k -Center: Solution Prospects

CAPACITATED k -CENTER is NP-hard.

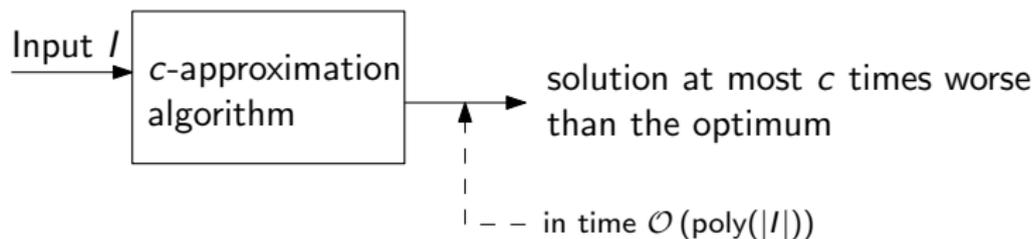
⇒ cannot solve exactly in polynomial time assuming $P \neq NP$.

Capacitated k -Center: Solution Prospects

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⇒ cannot solve exactly in polynomial time assuming $P \neq NP$.

c -approximation algorithm

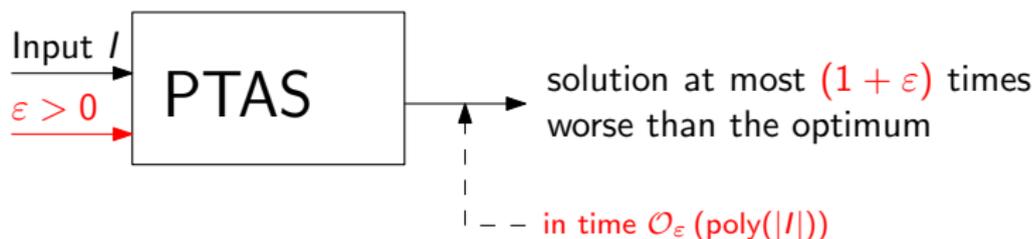


Capacitated k -Center: Solution Prospects

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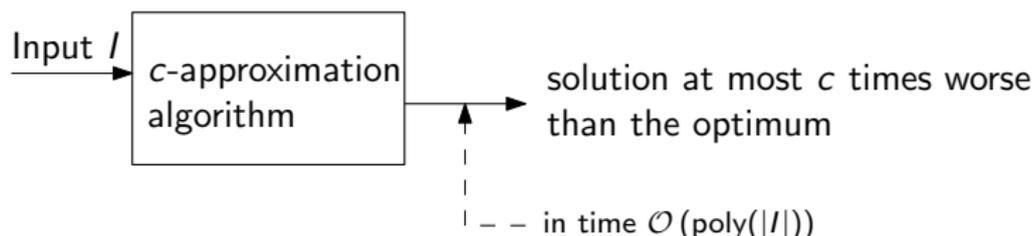
⇒ cannot solve exactly in polynomial time assuming $P \neq NP$.

Polynomial-time approximation scheme



Capacitated k -Center: Solution Prospects

c -approximation algorithm



An, Bhaskara, Chekuri, Gupta, Madan, Svensson. 2015

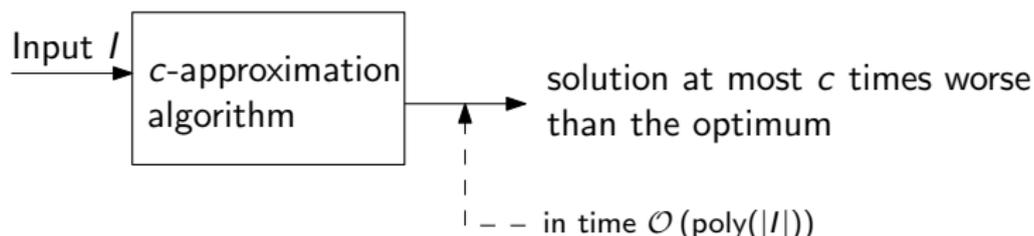
There is a 9-approximation algorithm for CKC .

Cygan, Hajiaghayi, Khuller. 2012

There is no $(3 - \varepsilon)$ -approximation algorithm for CKC unless $P = NP$.

Capacitated k -Center: Solution Prospects

c -approximation algorithm



Cygan, Hajiaghayi, Khuller. 2012

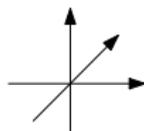
There is no $(3 - \varepsilon)$ -approximation algorithm for CKC unless $P = \text{NP}$.

Question

Are there settings where we can overcome this lower bound?

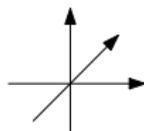
Planar graphs, Euclidean spaces, real world, ...

Special Settings?



	Doubling Dimension (Δ)	
CAPACITATED k -CENTER	<div style="border: 1px solid black; padding: 5px; display: inline-block;">generalizes the dimension of ℓ_q spaces</div>	
k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)} \cdot \text{poly}(n)$ Feldmann, Marx. 2020	
k -MEDIAN, k -MEANS, FACILITY LOCATION	$2^{(1/\varepsilon)^{\mathcal{O}(\Delta^2)}} \cdot \text{poly}(n)$ Cohen-Addad, Feldmann, Saulpic. 2021	
TSP, STEINER TREE	$\exp\{2^{\mathcal{O}(\Delta)} \cdot (4\Delta \log n / \varepsilon)^\Delta\}$ Talwar. 2004	

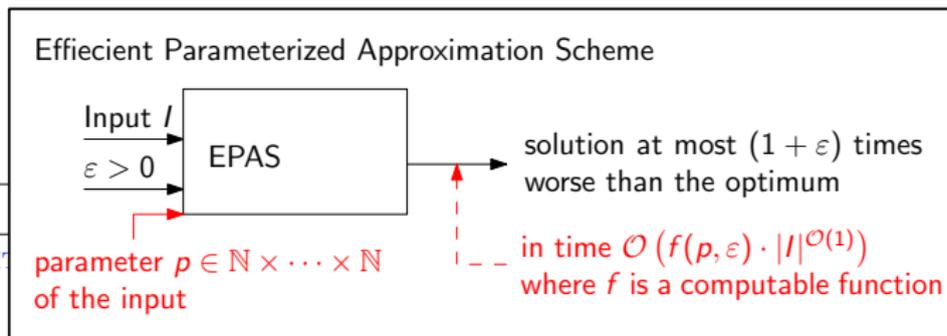
Special Settings?



	Doubling Dimension (Δ)	Highway dimension (h)
CAPACITATED k -CENTER		captures properties of transportation networks
k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)} \cdot \text{poly}(n)$ Feldmann, Marx. 2020	$f(k, h, \varepsilon) \cdot \text{poly}(n)^\dagger$ Becker, Klein, Saulpic. 2018
k -MEDIAN, k -MEANS, FACILITY LOCATION	$2^{(1/\varepsilon)\mathcal{O}(\Delta^2)} \cdot \text{poly}(n)$ Cohen-Addad, Feldmann, Saulpic. 2021	$n^{(2h/\varepsilon)\mathcal{O}(1)}$ Feldmann, Saulpic. 2021
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†: f : computable function

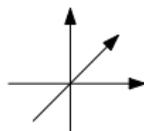
Special Settings?



CAPACITY		
k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)}$ poly(n) Feldmann, Marx. 2020	$f(k, h, \varepsilon)$ poly(n) [†] Becker, Klein, Saulpic. 2018
k -MEDIAN, k -MEANS, FACILITY LOCATION	$2^{(1/\varepsilon)^{\mathcal{O}(\Delta^2)}}$ poly(n) Cohen-Addad, Feldmann, Saulpic. 2021	$n^{(2h/\varepsilon)^{\mathcal{O}(1)}}$ Feldmann, Saulpic. 2021
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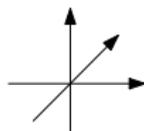
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	Doubling Dimension (Δ)	Highway dimension (h)
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\dagger : f : computable function

\S : unless FPT = W[1]

Doubling Dimension

- ▶ Let $M = (X, \text{dist})$ be a metric space.

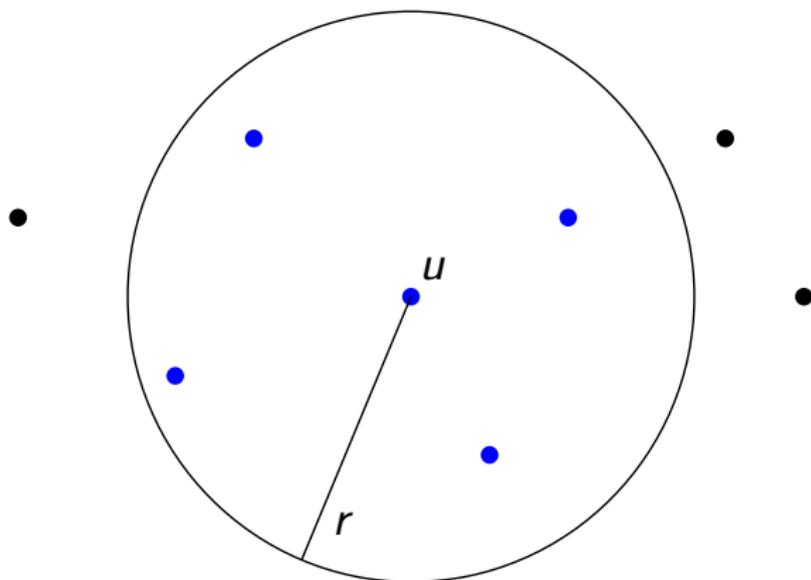
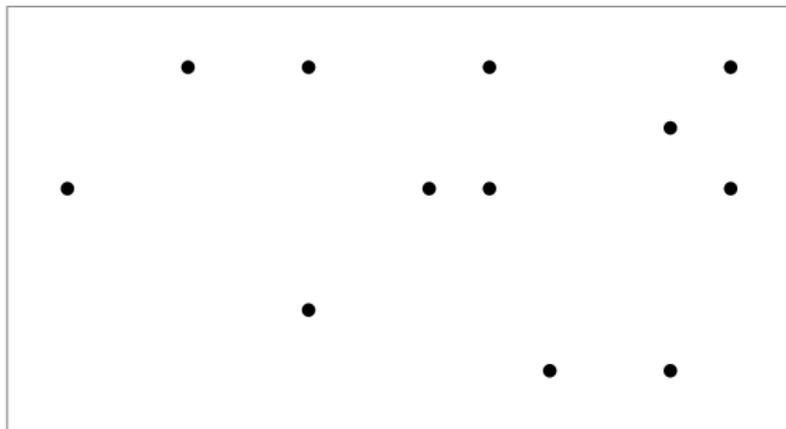


Figure: $B_r(u)$: Ball of radius r .

Doubling Dimension

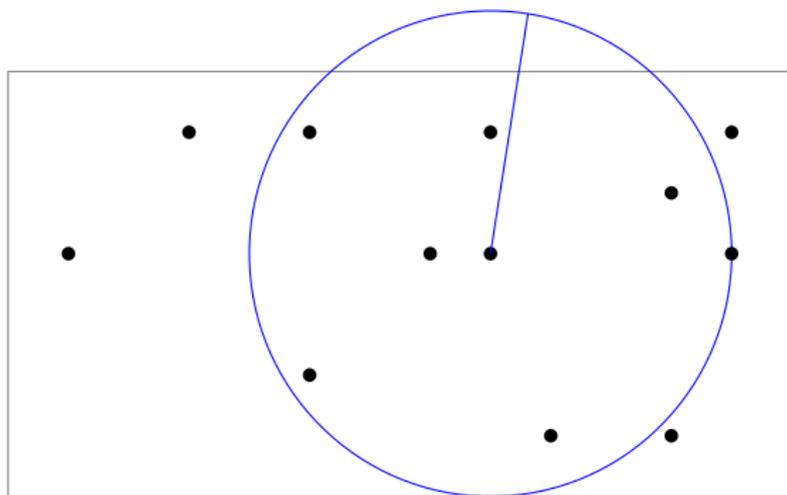
Doubling dimension $\Delta(M)$: smallest $\Delta \in \mathbb{N}$ such that



Doubling Dimension

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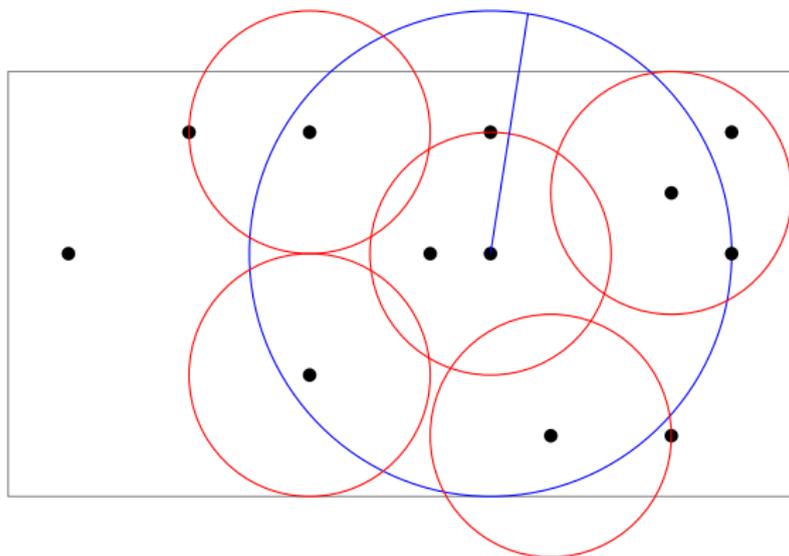
- ▶ the ball $B_r(u)$ for every $u \in X$ and every $r \in \mathbb{R}^+$



Doubling Dimension

Doubling dimension $\Delta(M)$: smallest $\Delta \in \mathbb{N}$ such that

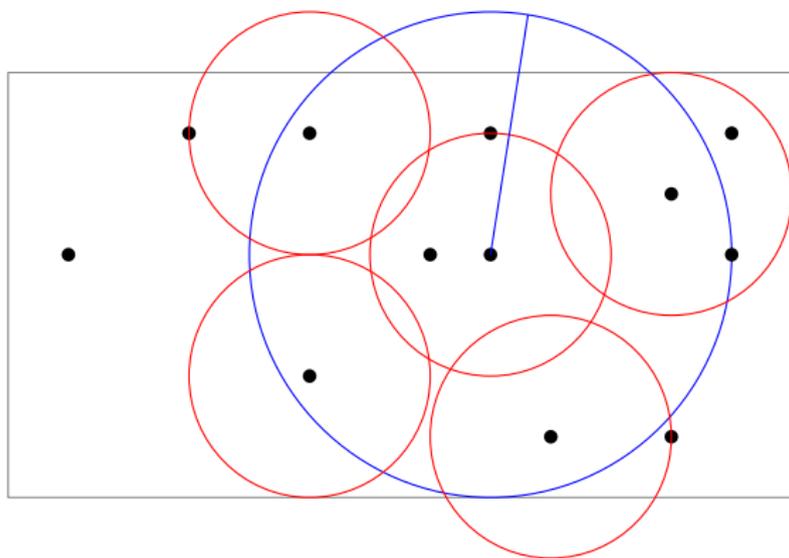
- ▶ the ball $B_r(u)$ for every $u \in X$ and every $r \in \mathbb{R}^+$
- ▶ is contained in $\cup_{v \in V} B_{r/2}(v)$ for some $V \subseteq X$ with $|V| \leq 2^\Delta$.



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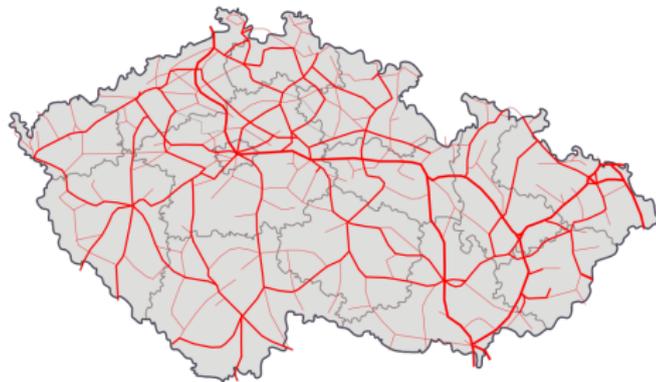
\rightsquigarrow d -dimensional ℓ_q metrics have doubling dimension $\mathcal{O}(d)$.

Highway Dimension: Shortest Path Cover

- ▶ Let G be an edge-weighted graph and fix a *scale* $r \in \mathbb{R}^+$.
- ▶ Let \mathcal{P}_r be the set of paths of G such that
 - ▶ they are a shortest path between their endpoints,
 - ▶ their length is more than r and at most $2r$.



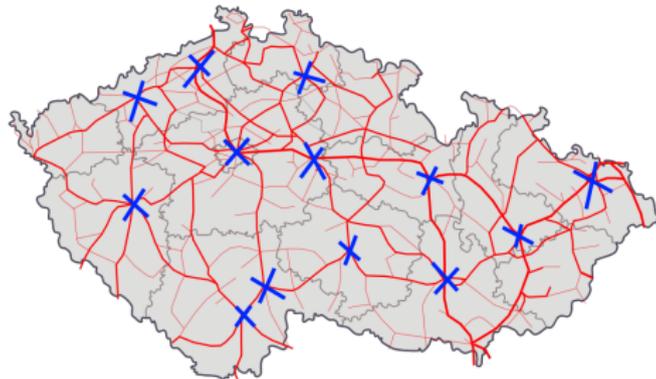
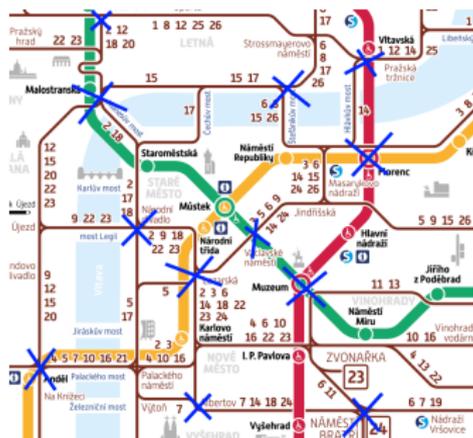
(a) Metro and tram network in Prague city center.



(b) Czech railway network.

Highway Dimension: Shortest Path Cover

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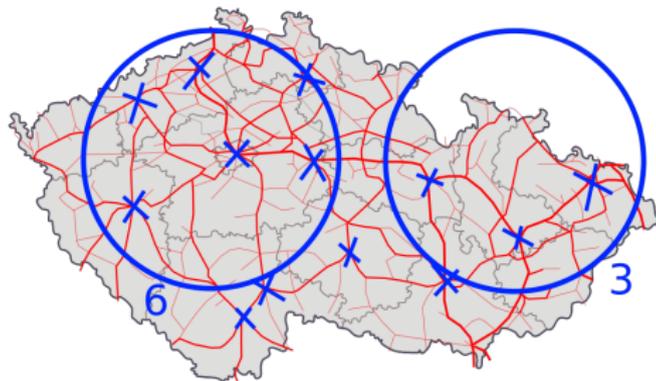
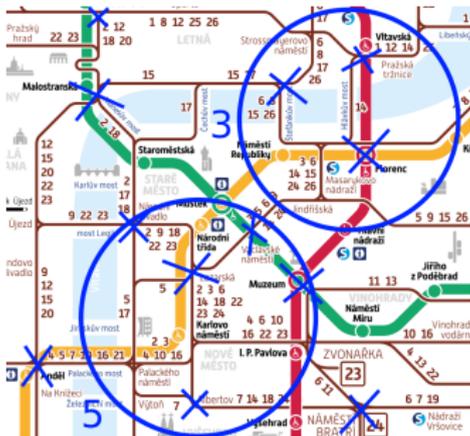
The *shortest path cover* $\text{SPC}_r(G)$ is a hitting set¹ for \mathcal{P}_r .

¹For every $P \in \mathcal{P}_r$ we have $P \cap \text{SPC}_r(G) \neq \emptyset$.

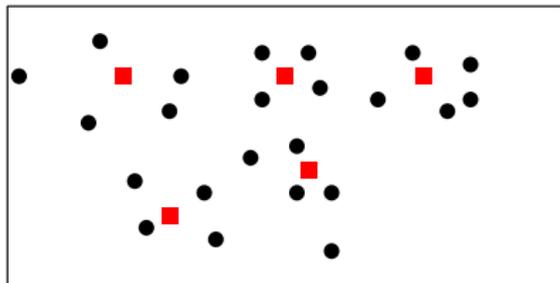
Highway Dimension

highway dimension of an edge-weighted graph G :

- ▶ smallest integer h such that,
- ▶ for any scale $r \in \mathbb{R}^+$,
- ▶ there exists $H := \text{SPC}_r(G)$ so that,
- ▶ $|H \cap B_{2r}(u)| \leq h$ for every $u \in V(G)$.

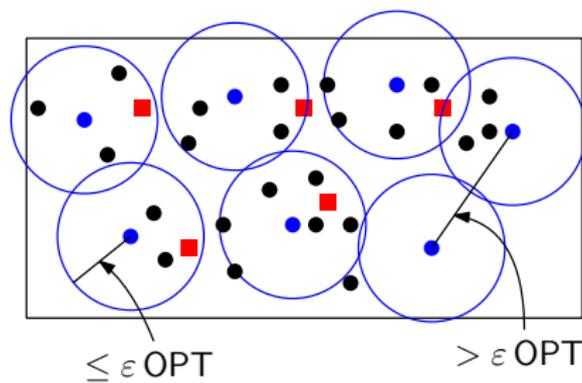


k -CENTER algorithm



■ Optimum solution
of cost OPT.

k-CENTER algorithm

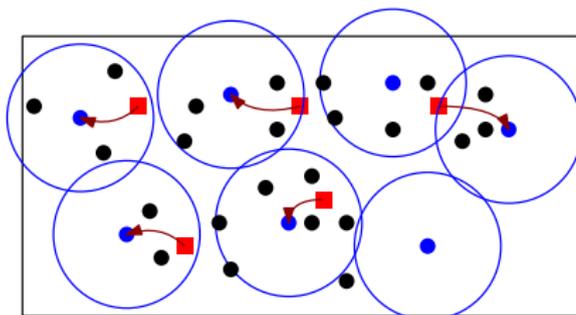


■ Optimum solution of cost OPT.

● Net: $Y \subseteq X$ such that

- $\forall x \in X \exists y \in Y: d(x, y) \leq \epsilon \text{ OPT}$, and
- $\forall y_1 \neq y_2 \in Y: d(y_1, y_2) > \epsilon \text{ OPT}$.

k-CENTER algorithm



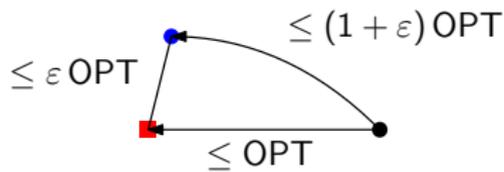
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↪ Replace every optimum center by its nearest net point.

⇒ We get a $(1 + \varepsilon)$ -approximate solution.



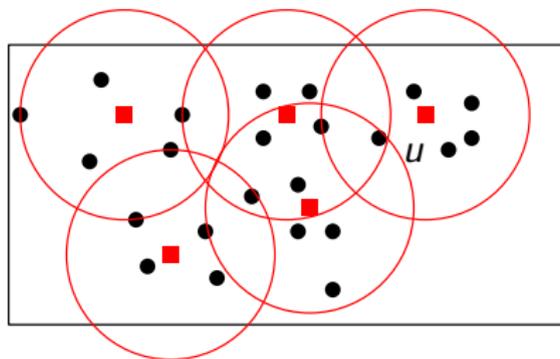
k -CENTER algorithm

```
1: guess OPT
2: for all  $K \subseteq Y$  with  $|K| \leq k$  do
3:   if  $\bigcup_{u \in K} B_{(1+\varepsilon)\text{OPT}}(u) \supseteq V$  then
4:     return  $K$ 
5: return :
```

Bounding net size

Gupta, Krauthgamer, Lee. 2003

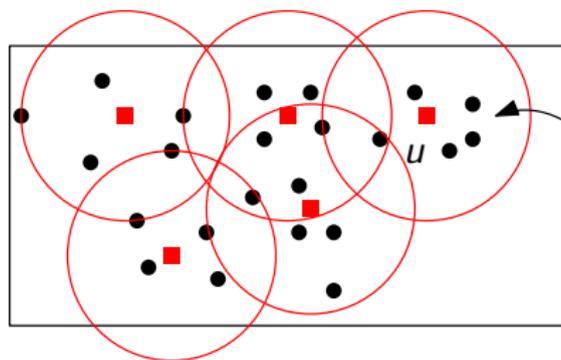
Let M be a metric and $\alpha(M) = \frac{\max_{u,v \in M} \text{dist}(u,v)}{\min_{u \neq v \in M} \text{dist}(u,v)}$. Then for every $M' \subseteq M$ we have $\Delta(M') \leq \Delta(M)$ and $|M| \leq 2^{\mathcal{O}(\Delta \lceil \log_2(\alpha) \rceil)}$.



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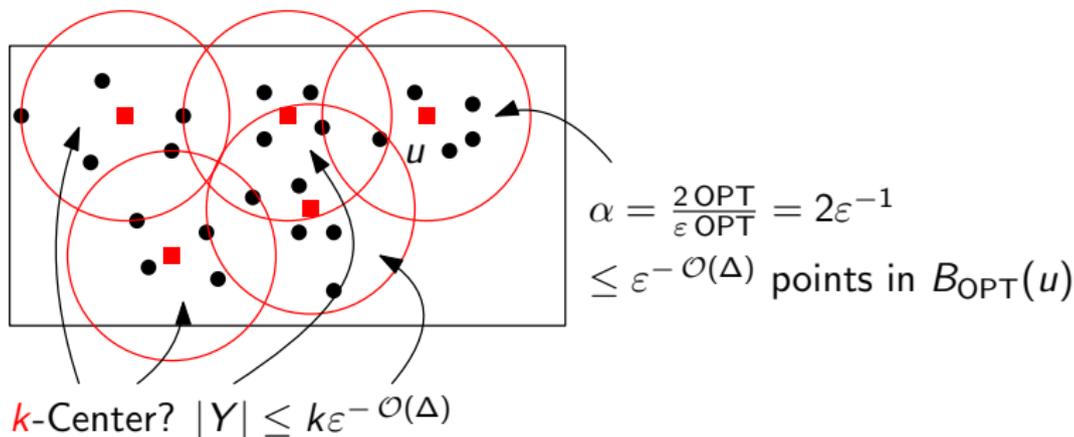
$$\alpha = \frac{2 \text{OPT}}{\varepsilon \text{OPT}} = 2\varepsilon^{-1}$$

$$\leq \varepsilon^{-\mathcal{O}(\Delta)} \text{ points in } B_{\text{OPT}}(u)$$

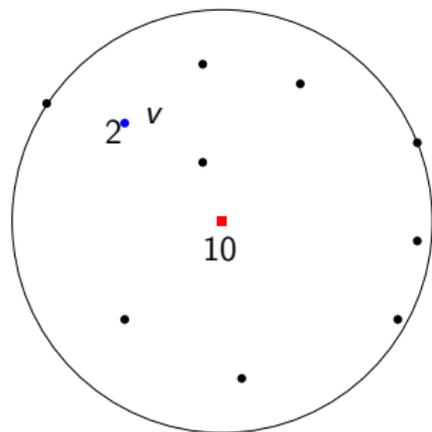
Bounding net size

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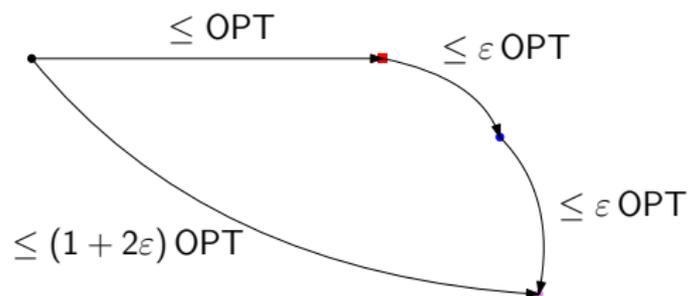
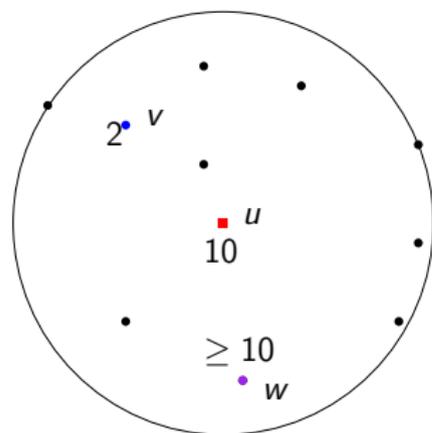


CKC algorithm obstacles



► $c(u) < c(v)$?

CKC algorithm obstacles

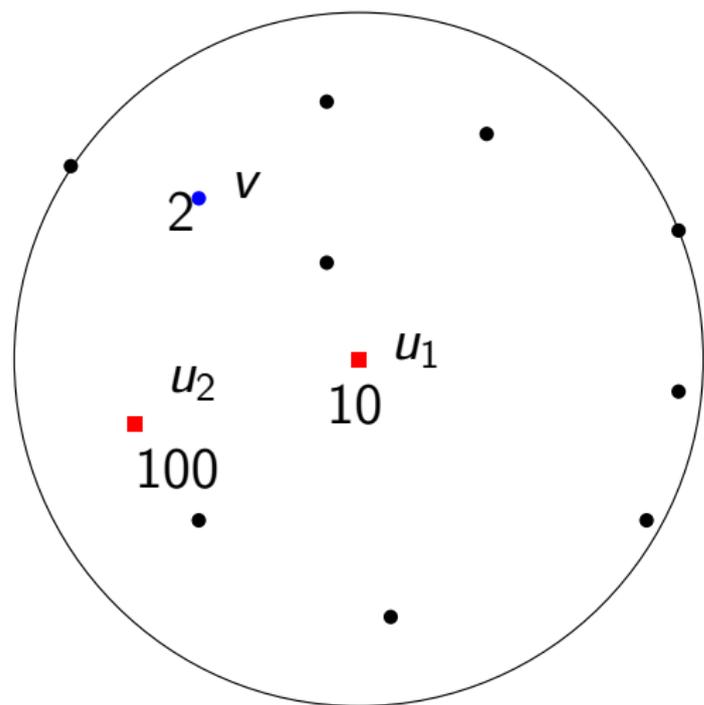


- ▶ $c(u) < c(v)$?
- ▶ w : vertex with highest capacity in $B_{\varepsilon \text{OPT}}(u)$.
- ▶ $c(w) \geq c(v)$ as v , the optimum center, exists.

CKC algorithm

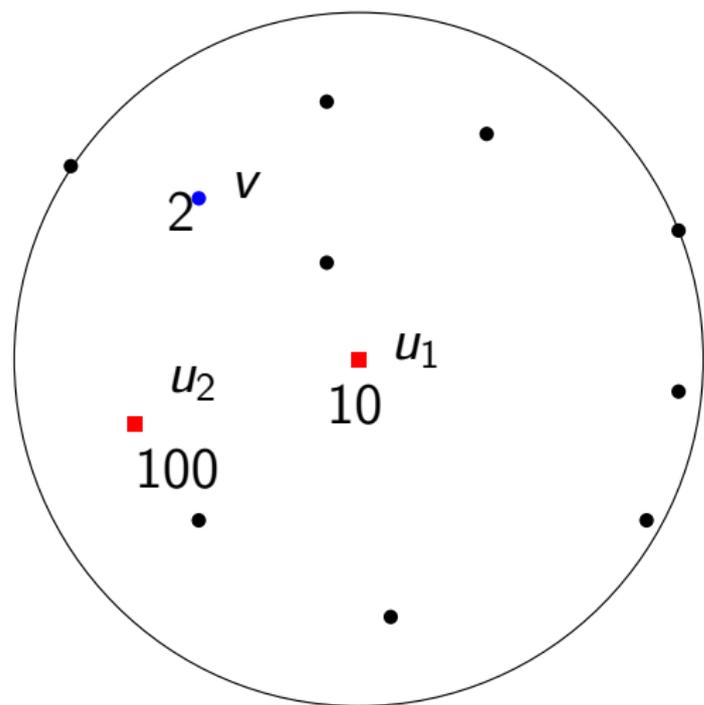
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1: guess OPT
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6:      $S \leftarrow S \cup \{w\}$ 
7:   if ??? then
8:     return  $S$ 
9: return :
```

CKC algorithm obstacles



- ▶ multiple optimum centers in $B_{\epsilon_{\text{OPT}}}(v)$?

CKC algorithm obstacles



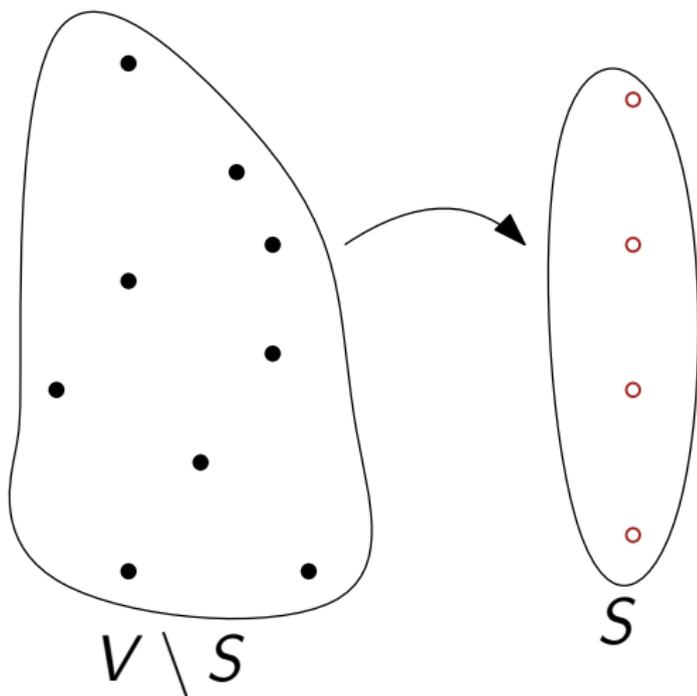
- ▶ multiple optimum centers in $B_{\epsilon_{\text{OPT}}}(v)$?
- ▶ use multisets!

CKC algorithm

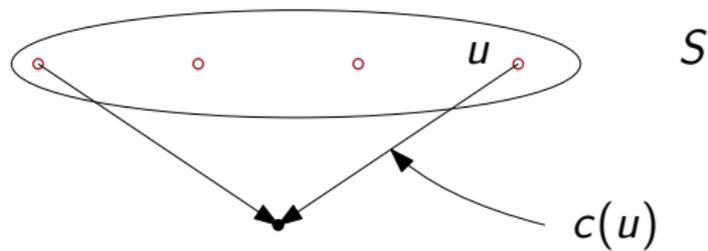
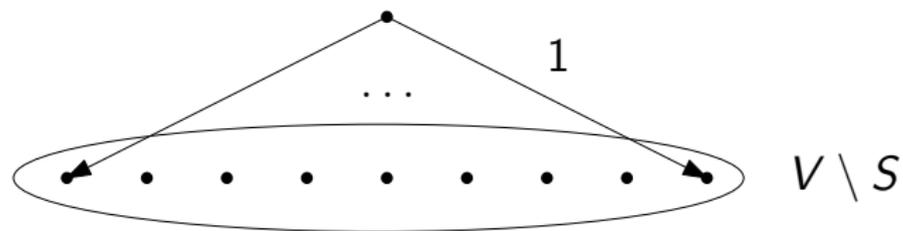
```
1: guess OPT
2: for all  $K \subseteq_{\text{multiset}} Y$  with  $|K| \leq k$  do
3:    $S \leftarrow \emptyset$ 
4:   for all  $v \in K$  do
5:      $w \leftarrow$  vertex with the highest capacity in  $B_{\epsilon_{\text{OPT}}}(v) \setminus S$ 
6:      $S \leftarrow S \cup \{w\}$ 
7:   if ??? then
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```

Solution verification

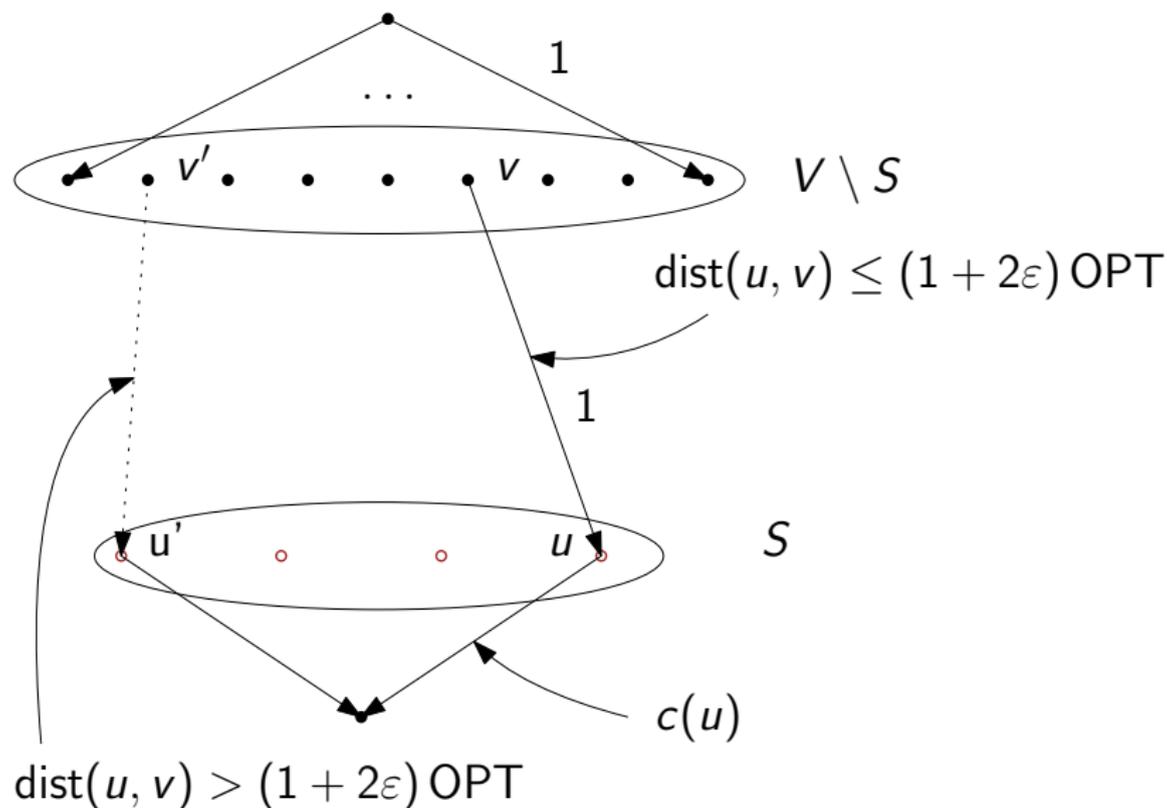
- ▶ Given $S \subseteq V$,
- ▶ \exists assignment of $V \setminus S$ to S
- ▶ respecting capacities of S
- ▶ of cost $(1 + 2\varepsilon) \text{OPT}$?



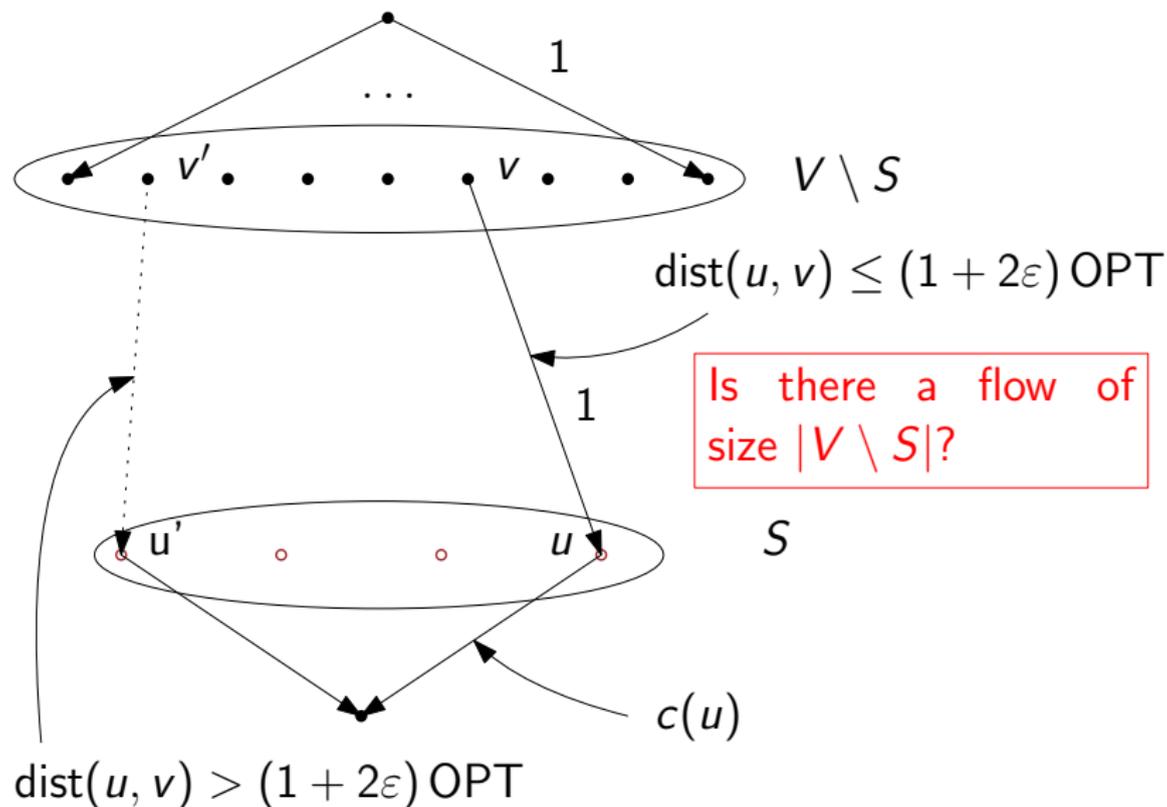
Solution verification: using network flows



Solution verification: using network flows



Solution verification: using network flows



CKC algorithm

- 1: guess OPT
 - 2: **for all** $K \subseteq_{\text{multiset}} Y$ with $|K| \leq k$ **do**
 - 3: $S \leftarrow \emptyset$
 - 4: **for all** $v \in K$ **do**
 - 5: $w \leftarrow$ vertex with the highest capacity in $B_{\epsilon \text{OPT}}(v) \setminus S$
 - 6: $S \leftarrow S \cup \{w\}$
 - 7: **if** \exists flow from $V \setminus S$ to S of size $|V \setminus S|$ **then**
 - 8: **return** assignment corresponding to that flow
 - 9: **return** : (
-

Conclusion

	Doubling Dimension (Δ)	Highway dimension (h)
CAPACITATED k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)} \cdot \text{poly}(n)$ Theorem 2	$\exists c > 1$: no c -approximation in $\mathcal{O}_\varepsilon(f(k, h) \cdot \text{poly}(n))^{\dagger, \S}$ Theorem 1
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\dagger : f : computable function

\S : unless FPT = W[1]

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	Doubling Dimension (Δ)	Highway dimension (h)
CAPACITATED k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)} \cdot \text{poly}(n)$ Theorem 2	$\exists c > 1$: no c -approximation in $\mathcal{O}_\varepsilon(f(k, h) \cdot \text{poly}(n))^{\dagger, \S}$ Theorem 1
k -CENTER	$k^k / \varepsilon^{\mathcal{O}(k\Delta)} \cdot \text{poly}(n)$ Feldmann, Marx. 2020	$f(k, h, \varepsilon) \cdot \text{poly}(n)^{\dagger}$ Becker, Klein, Saulpic. 2018
k -MEDIAN, k -MEANS, FACILITY LOCATION	$2^{(1/\varepsilon)^{\mathcal{O}(\Delta^2)}} \cdot \text{poly}(n)$ Cohen-Addad, Feldmann, Saulpic. 2021	$n^{(2h/\varepsilon)^{\mathcal{O}(1)}}$ Feldmann, Saulpic. 2021
TSP, STEINER TREE	$\exp\{2^{\mathcal{O}(\Delta)} \cdot (4\Delta \log n/\varepsilon)^\Delta\}$ Talwar. 2004	$\exp\{\text{polylog}(n)^{\mathcal{O}(\log^2(h/\varepsilon))}\}$ Feldmann, Fung, Könemann, Post. 2018

\dagger : f : computable function

\S : unless FPT = W[1]

Thank you for your attention!

Questions, comments, ...?