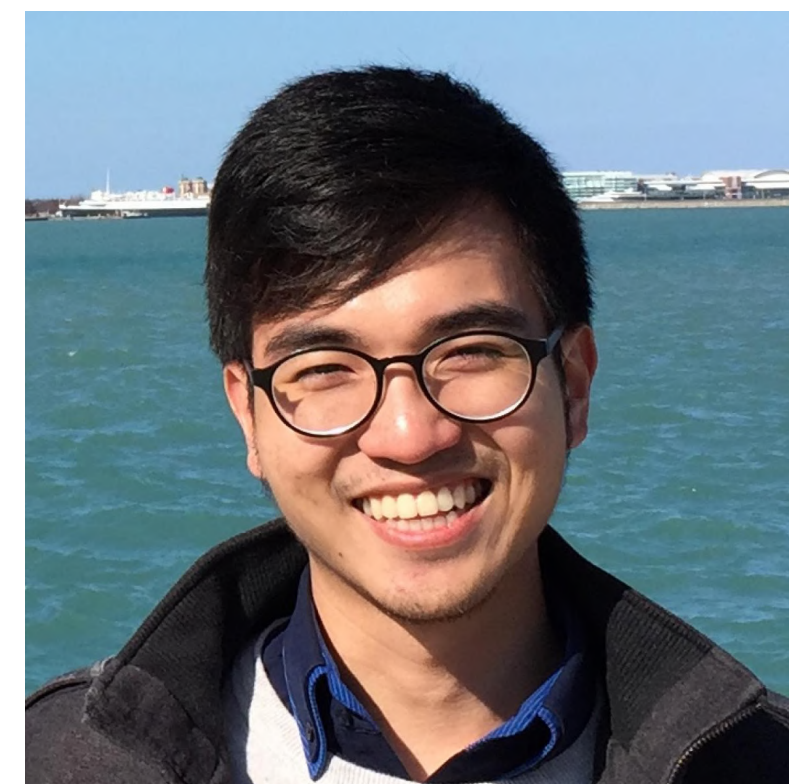


On **Equivalence** of Parameterized Inapproximability of **k-median**, **k-max-coverage**, and **2-CSP**

Karthik C. S.
(Rutgers University)

Joint Work with

Euiwoong Lee
(University of Michigan)



Pasin Manurangsi
(Google Thailand)

2-CSP

k-max-coverage

k-median

2-CSP

k-max-coverage

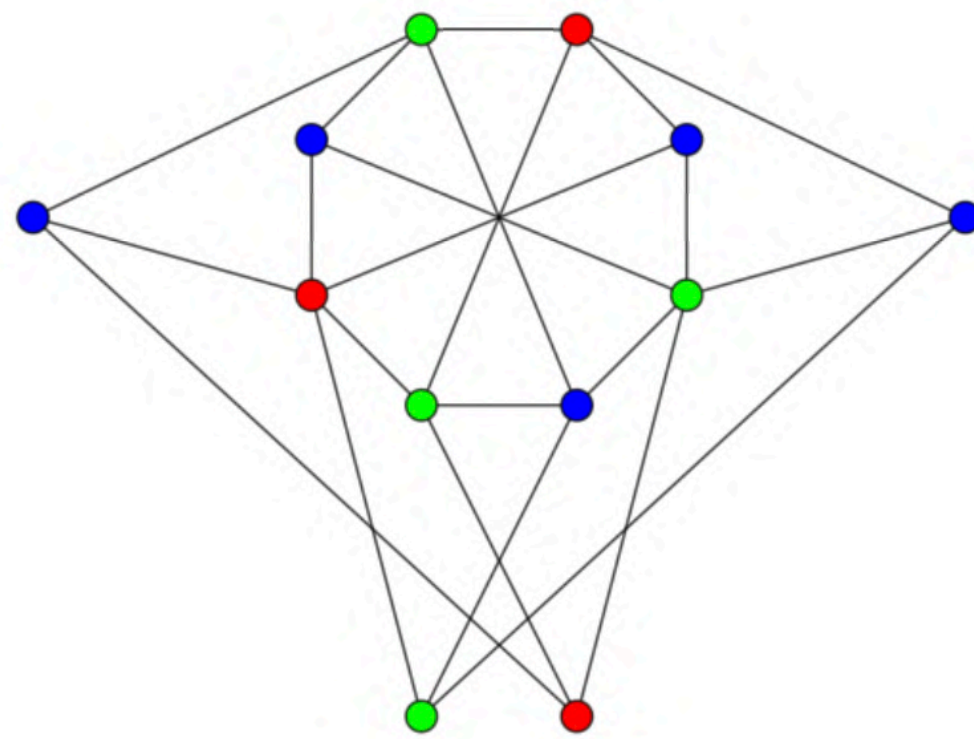
k-median

- **Input:** Graph $G = (V, E)$ and constraints:

$$\mathcal{C} = \{C_e \subseteq \Sigma \times \Sigma\}_{e \in E}$$

- **Output:** Assignment $\sigma : V \rightarrow \Sigma$ maximizing:

$$\Pr_{(u,v) \sim E} [(\sigma(u), \sigma(v)) \in C_{(u,v)}]$$



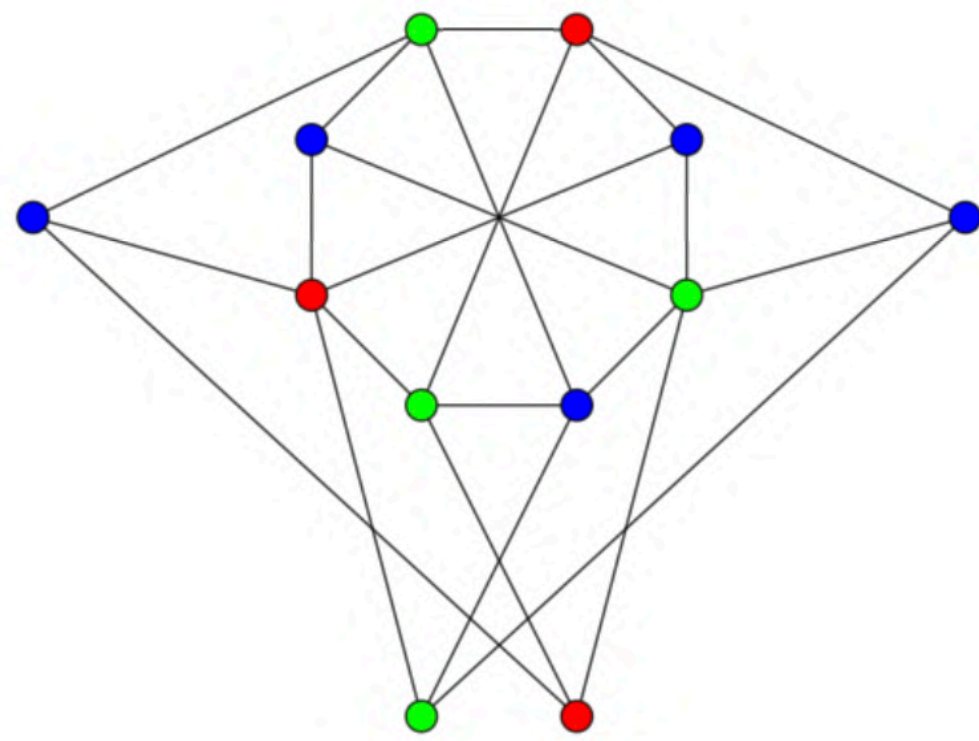
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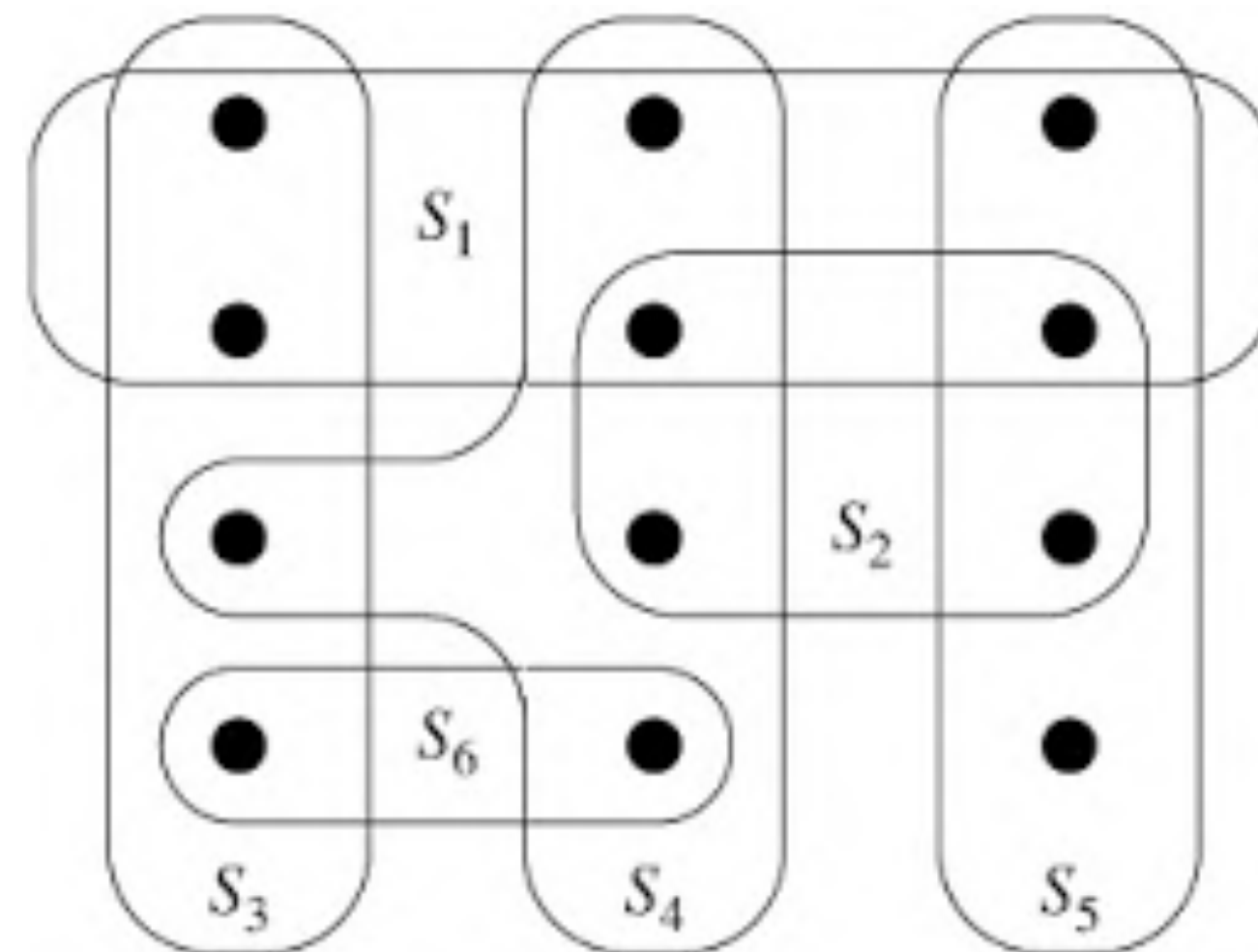
$$\Pr_{(u,v) \sim E} [(\sigma(u), \sigma(v)) \in C_{(u,v)}]$$



k-max-coverage

- **Input:** $S_1, \dots, S_m \subseteq [n]$, and integer k
- **Output:** S_{i_1}, \dots, S_{i_k} maximizing:

$$\frac{\left| \bigcup_{j \in [k]} S_{i_j} \right|}{n}$$



k-median

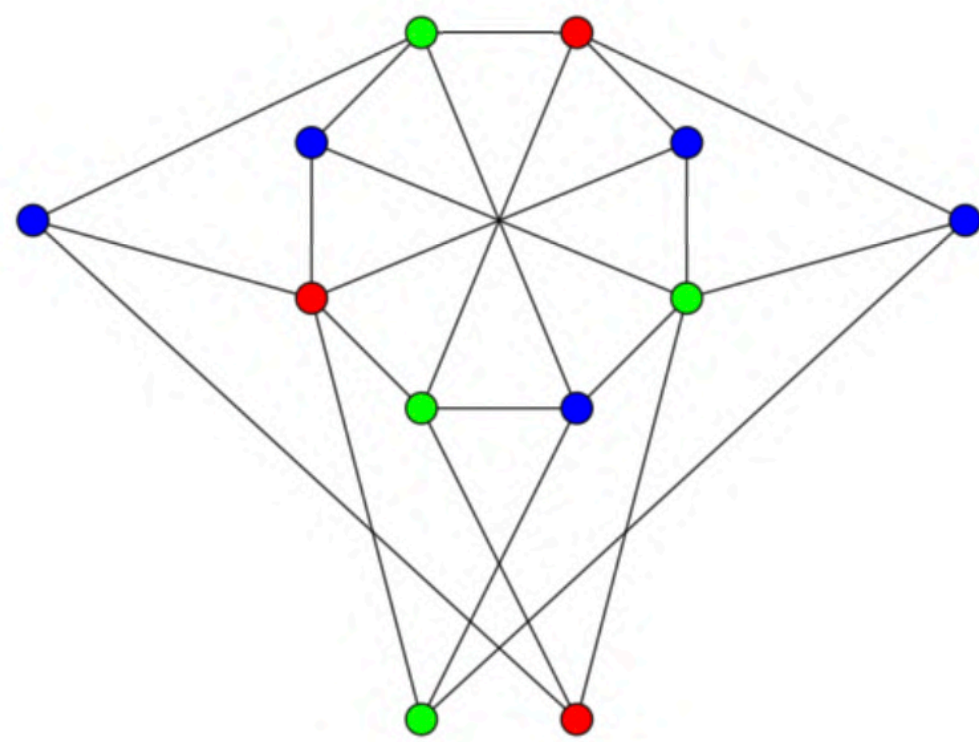
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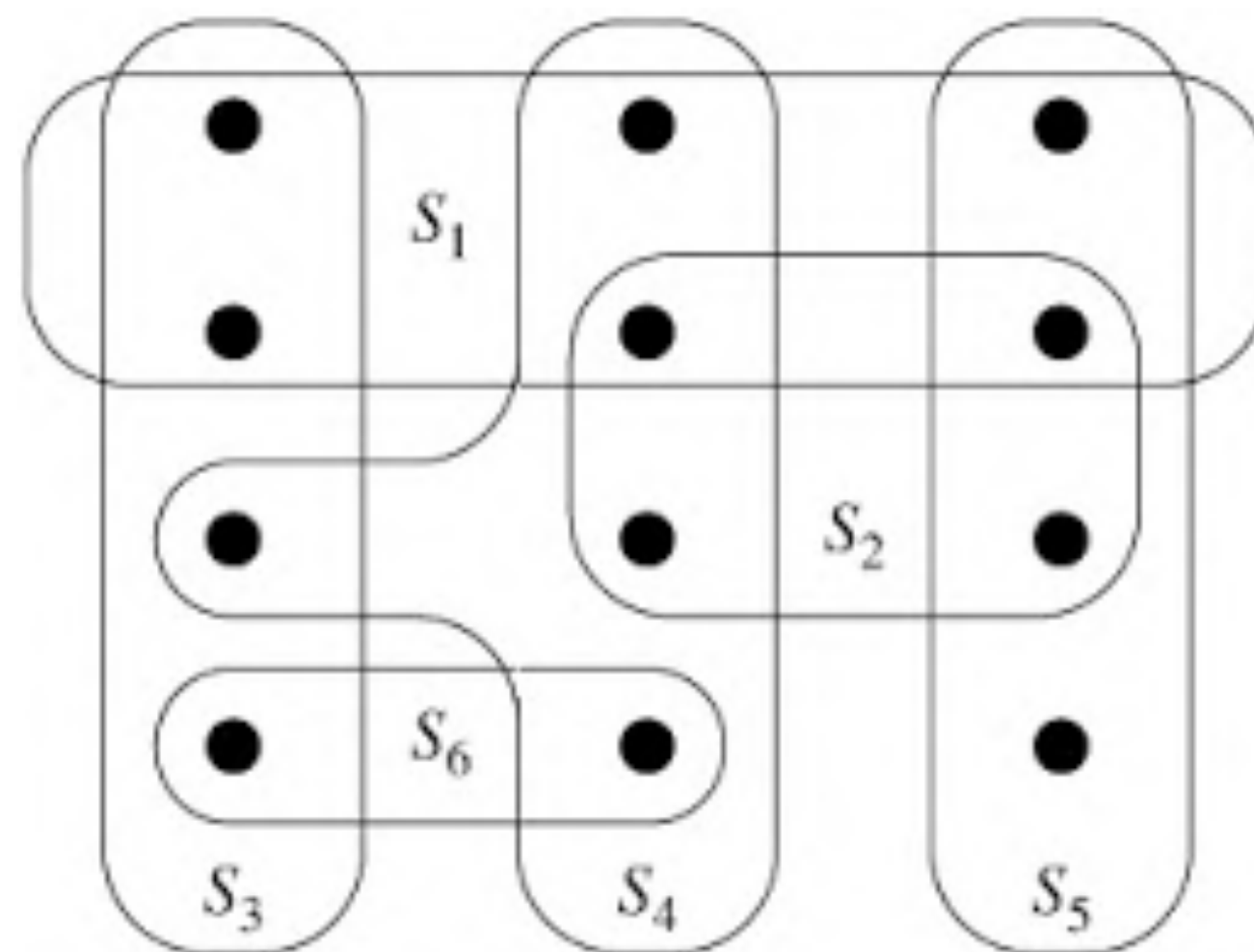
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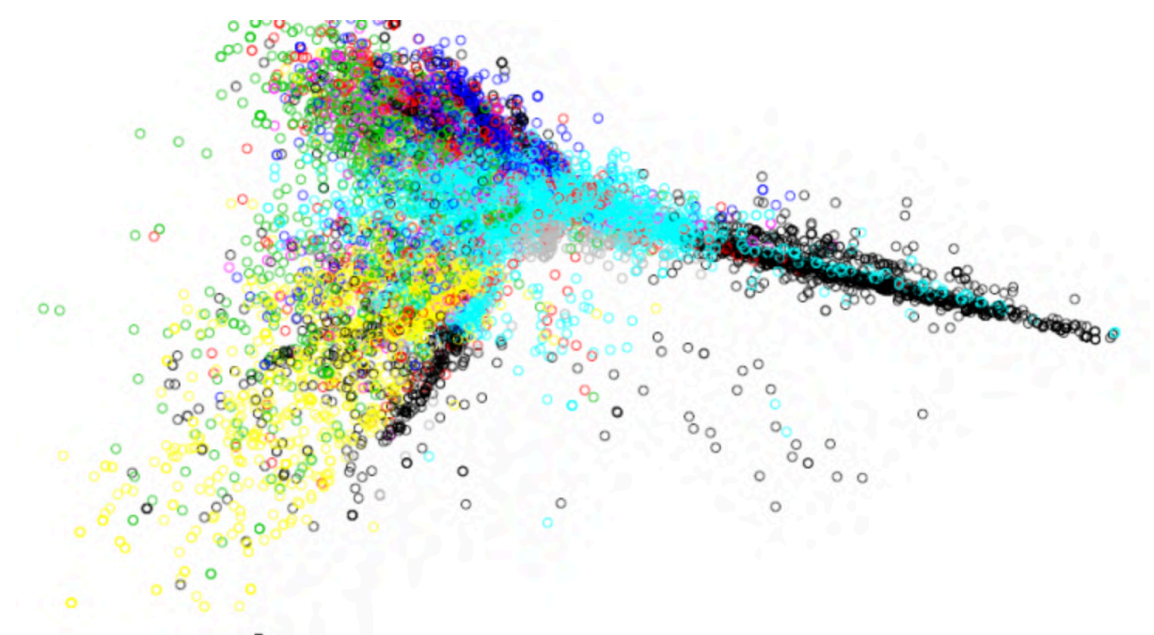
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k-median

- **Input:** Clients C , Facilities F , distance $\Delta : C \times F \rightarrow \mathbb{R}_{\geq 0}$, and integer k
- **Output:** $f_1, \dots, f_k \in F$ minimizing:

$$\sum_{c \in C} \min_{i \in [k]} \Delta(c, f_i)$$



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W[1]-complete

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W[2]-complete

What about FPT Approximation?

FPT Approximation

2-CSP

- **Input:** Graph $G = ([k], E)$ and constraints:

$$\mathcal{C} = \{C_e \subseteq [n] \times [n]\}_{e \in E}$$

- **Completeness:** $\exists \sigma : [k] \rightarrow [n]$

$$\Pr_{(u,v) \sim E} [(\sigma(u), \sigma(v)) \in C_{(u,v)}] \geq c$$

- **Soundness:** $\forall \sigma : [k] \rightarrow [n]$

$$\Pr_{(u,v) \sim E} [(\sigma(u), \sigma(v)) \in C_{(u,v)}] < s$$

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$$\sum_{c \in C} \min_{i \in [k]} \Delta(c, f_i) \leq c$$

- **Soundness:** $\forall f_1, \dots, f_k \in F$

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FPT Approximation

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But we are interested when $c=1$
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Parameterized Inapproximability Hypothesis (PIH) [Lokshtanov-Ramanujan-Saurabh-Zehavi'17]

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But we are interested when $c=1$
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Completeness: $\exists f_1, \dots, f_k \in F$

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FPT Approximation

2-CSP

k-max-coverage

Tight

k-median

Tight



- **Input:** Graph $G = ([k], E)$ and parameterized inapproximability hypothesis (PIH) [Lokshtanov-Ramanujan-Saurabh-Zehavi'17]

- **Input:** $S_1, \dots, S_m \subseteq [n]$, and integer k

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FPT Approximation

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Parameterized Inapproximability Hypothesis

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But we are interested when $c=1$
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Completeness: $\exists f_1, \dots, f_k \in F$

• **Completeness:** $\exists \sigma : [k] \rightarrow [n]$ such that $|S_i \cap \sigma^{-1}(i)| \geq c$
[Lokshtanov-Ramanujan-Saurabh-Zehavi'17]

$$\sum_{c \in C} \min_{i \in [k]} \Delta(c, f_i) \leq c$$

• **Soundness:** $\forall \sigma : [k] \rightarrow [n]$

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ETH \implies PIH
[Guruswami-Lin-Ren-Sun-Wu'24]



FPT Approximation

2-CSP $\xrightarrow{\text{Tight}}$ k-max-coverage $\xrightarrow{\text{Tight}}$ k-median

Parameterized
Inapproximability
Hypothesis
(PIH)
[Lokshtanov-Ramanujan-
Saurabh-Zehavi'17]

Can we prove PIH?

ETH \implies PIH
[Guruswami-Lin-Ren-Sun-Wu'24]

FPT Approximation

2-CSP $\xrightarrow{\text{Tight}}$ k-max-coverage $\xrightarrow{\text{Tight}}$ k-median

Parameterized Inapproximability Hypothesis (PIH)
[Lokshtanov-Ramanujan-Saurabh-Zehavi'17]

Can we prove PIH?

ETH \implies PIH
[Guruswami-Lin-Ren-Sun-Wu'24]

Can we prove parameterized inapproximability of k-max-coverage circumventing PIH?

FPT Approximation



[Guruswami-Lin-Ren-Sun-Wu'24]

Can we prove PIH?

Can we prove parameterized inapproximability of k-max-coverage circumventing PIH?



Once Upon a time ...



Once Upon a time ...





Once Upon a time ...



SIMONS

FOUNDATION



U.S. National
Science
Foundation

Wine Upon a time ...



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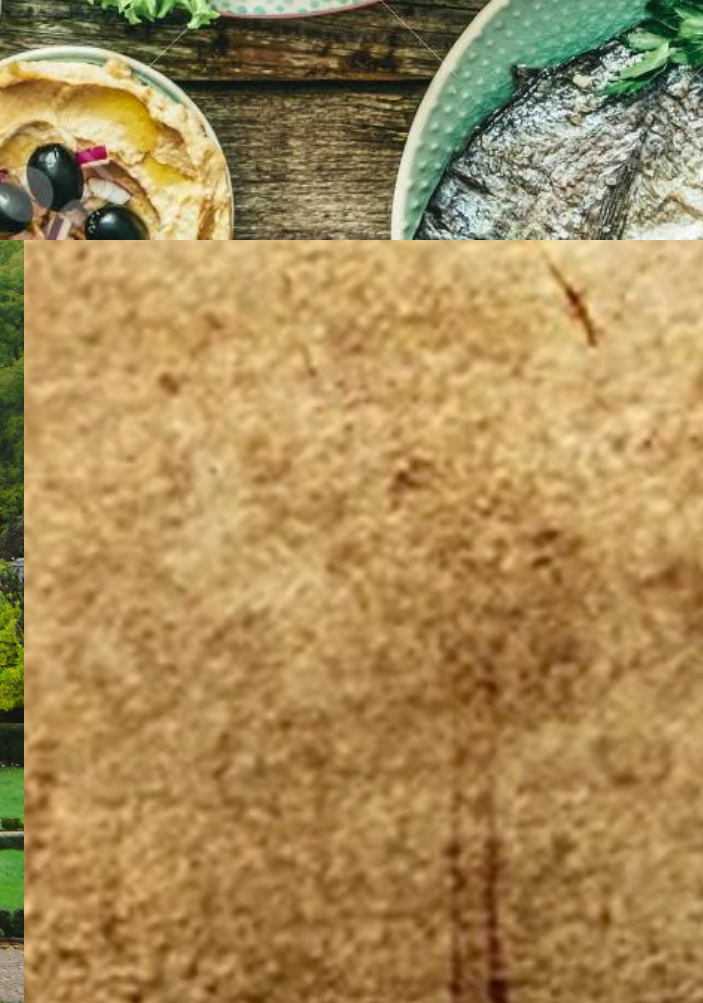
U.S. National
Science
Foundation



Orce ...



... ..



Our Results

Our Results

While *exact* k-max-coverage is $W[2]$ -complete, approximating it to any $1 - \frac{1}{F(k)}$ factor is $W[1]$ -complete.

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If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Our Results

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If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

If approximating k -median to $1 + \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Our Results

While *exact* k -max-coverage is $W[2]$ -complete, approximating it to any $1 - \frac{1}{F(k)}$ factor is $W[1]$ -complete.

FPT Reductions



If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

If approximating k -median to $1 + \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Our Results

While *exact* k-max-coverage is W[2]-complete, approximating it to any $1 - \frac{1}{F(k)}$ factor is W[1]-complete.

FPT Reductions



If approximating k-max-coverage to $1 - \epsilon$ factor is W[1]-complete, then PIH is true.

If approximating k-median to $1 + \epsilon$ factor is W[1]-complete, then PIH is true.

Approximating k-max-coverage to $1 - \epsilon$ factor is W[1]-hard,
 \implies Approximating k-max-coverage to $1 - \frac{1}{e} + o(1)$ factor is W[1]-hard

Gap Amplification

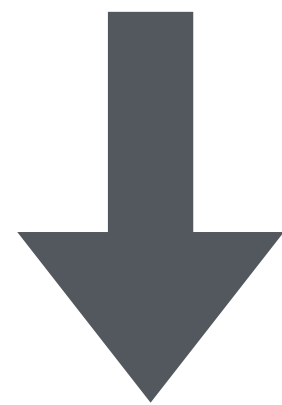
If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Step 1

Step 2

Step 3

$(\tau, (1 - \delta) \cdot \tau)$ k -max-coverage
on universe $[n]$



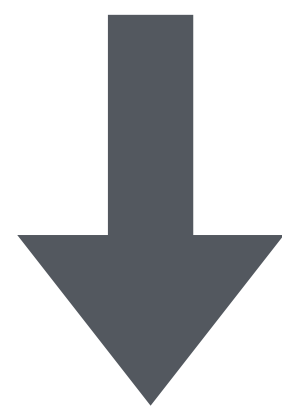
$(\tau', (1 - \epsilon) \cdot \tau')$ k -max-coverage
on universe of size $O_\delta(k \log n)$

τ' and ϵ are functions of δ

If approximating k -max-coverage to $1 - \varepsilon$ factor is $W[1]$ -complete, then PIH is true.

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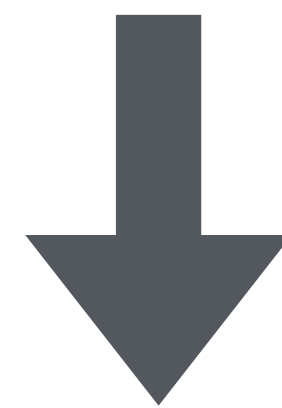


$(\tau', (1 - \varepsilon) \cdot \tau')$ k -max-coverage
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$(c, (1 - \varepsilon) \cdot c)$ Valued 2-CSP

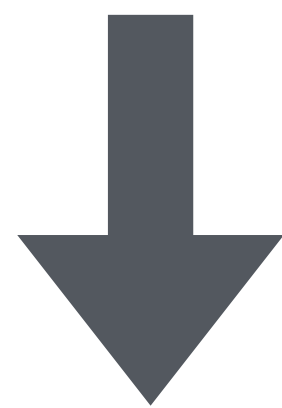
$$c = O_k(\tau')$$

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If approximating k -max-coverage to $1 - \varepsilon$ factor is $W[1]$ -complete, then PIH is true.

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$(\tau, (1 - \delta) \cdot \tau)$ k -max-coverage
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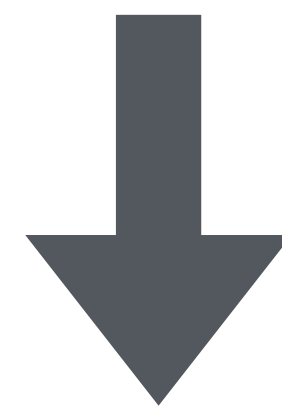


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Step 2

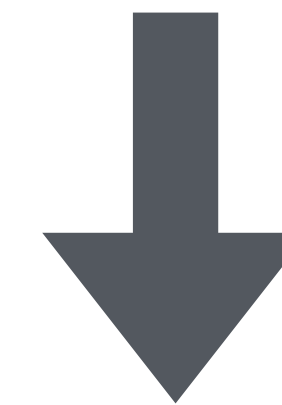
$(\tau', (1 - \varepsilon) \cdot \tau')$ k -max-coverage
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$(c, (1 - \varepsilon) \cdot c)$ Valued 2-CSP
 $c = O_k(\tau')$

Step 3

$(c, (1 - \varepsilon) \cdot c)$ Valued 2-CSP

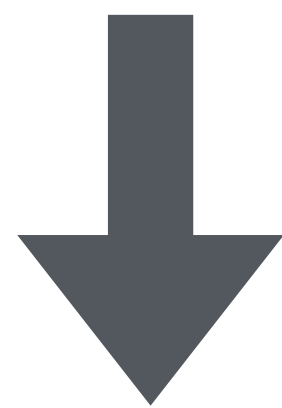


$\left(1, 1 - \frac{\varepsilon}{2}\right)$ 2-CSP

If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Step 1

$(\tau, (1 - \delta) \cdot \tau)$ k -max-coverage
on universe $[n]$



$(\tau', (1 - \epsilon) \cdot \tau')$ k -max-coverage
on universe of size $O_\delta(k \log n)$

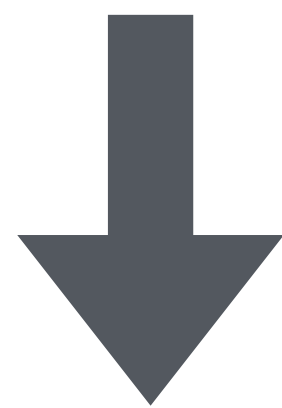
τ' and ϵ are functions of δ

Randomly hash $[n]$ to
universe of size $O_\delta(k \log n)$

If approximating **k-max-coverage** to $1 - \epsilon$ factor is W[1]-complete, then **PIH** is true.

Step 2

$(\tau', (1 - \epsilon) \cdot \tau')$ **k-max-coverage**
on universe of size $O_\delta(k \log n)$



$(c, (1 - \epsilon) \cdot c)$ **Valued 2-CSP**

$$c = O_k(\tau')$$

Partition Universe to $U_1 \dot{\cup} \dots \dot{\cup} U_M$
where $|U_i| = \frac{\log n}{\log k}$

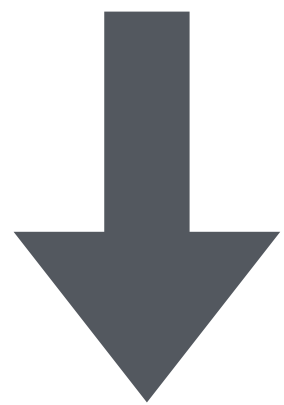
Variable x_i is assigned $f : U_i \rightarrow [k]$
Variable y_j is assigned input set S_j

Constraints between x_i and y_j measure
fraction of U_i mapped to j and covered by S_j

If approximating k -max-coverage to $1 - \epsilon$ factor is W[1]-complete, then PIH is true.

Step 3

$(c, (1 - \epsilon) \cdot c)$ Valued 2-CSP



$\left(1, 1 - \frac{\epsilon}{2}\right)$ 2-CSP

Values of a constraint can take one of $1 + \frac{\log n}{\log k}$ entries

There are at most $\binom{k}{2}$ constraints, so we can enumerate all possibilities

Our Results

While *exact* k -max-coverage is $W[2]$ -complete, approximating it to any $1 - \frac{1}{F(k)}$ factor is $W[1]$ -complete.

If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

If approximating k -median to $1 + \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Our Results

While *exact* k -max-coverage is $W[2]$ -complete, approximating it to any $1 - \frac{1}{F(k)}$ factor is $W[1]$ -complete.

[**K**-Laekhanukit-Manurangsi'19]

If approximating k -max-coverage to $1 - \epsilon$ factor is $W[1]$ -complete, then PIH is true.

Adjusting parameters in prior works

If approximating k -median to $1 + \epsilon$ factor is $W[1]$ -complete, then PIH is true.

[Cohen-Addad-Gupta-Kumar-Lee-Li'19]

Thank you for engaging!